Interactive System for Universal Functional Optimization (UFO) - Version 1997

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Contents

1. Introduction to the UFO system

The universal functional optimization (UFO) system is an interactive modular system for solving both dense medium-size and sparse large-scale optimization problems. The UFO system can be used for the following applications:

1. Formulation and solution of particular optimization problems that are described in chapter 2.
2. Preparation of specialized optimization routines (or subroutines) based on methods described in chapter 3.
3. Designing and testing new optimization methods. The UFO system is a very useful tool for optimization algorithms development.

The special realization of the UFO system, which is described in the subsequent text, makes this system portable and extensible and we continue with its further development.

1.1. Philosophy of the UFO system

The UFO system is an open software system for solving a broad class of optimization problems. An optimization problem solution is processed in four phases. In the first phase the optimization problem is specified and an optimization method is selected. This can be made in three different ways:

1. The full dialogue mode: The problem specification and the method selection are realized by using a conversation between the user and the UFO system.
2. The batch mode: The problem specification and the method selection are realized by using the UFO control language. An input file, written in the UFO control language, has to be prepared and stored.
3. The combined mode: Only a part of the specification is written in the input file. The rest of the specification is obtained as in the dialogue mode. This possibility is usually the best one since the problem functions can be defined beforehand by using a convenient text editor.

The second phase is realized by using the UFO preprocessor. This preprocessor is written in the Fortran 77 language and its output is a Fortran 77 control program. This conception is very advantageous for the following reasons:

1. The Fortran 77 (full ANSI norm) is a sufficiently high and portable programming language. Moreover, this language is very suitable for numerical computations, and a broad class of subroutines is available in this field.
2. A control program, generated by the UFO preprocessor, calls for necessary modules only and its specification is very easy. Moreover, control program global declarations are determined by the problem size, which decreases storage requirements. This way overcomes an impossibility of dynamical declarations in the Fortran 77 language.
3. The UFO system is open. When a new class of optimization problems or optimization methods has to be included, one only needs to change the system templates and prepare new modules. The control program is composed of individual modules by using specifications in the first phase. This fact allows us to create a great number of various optimization methods.

In the third phase, the control program is translated by using a Fortran 77 compiler and a final program is linked by using library modules. In the fourth phase, the final program is executed and results which can be viewed by using extensive output means are obtained.

The above conception is enabled by a special form of source modules. These modules usually consist of two parts, the interface template and the Fortran 77 realization. The interface template is used by the UFO preprocessor only and it serves for the control program generation (the part of control program corresponding to a given module is coded in the template). These templates also contain knowledge bases for an automatic selection of the optimization method. If the UFO system has to be extended then usually only templates, which do not need to be compiled, are changed. Besides interface templates, which are a part of source modules, special templates controlling the UFO reprocessor exist. A batch input file written in the UFO control language is one of these special templates.

The UFO macroprocesor works in two stages. In the first pass, the file P.TMP is created. This file is a control program ancestor containing some macroinstructions and macrovariables which are replaced in the second pass. The control program P.FOR is the result of the second pass.

1.2. Execution of the UFO system

The UFO system contains three basic procedures GENER.BAT, COMPIL.BAT and UFOGO.BAT. The UFO preprocessor is called if the statement

GENER *input_name*

is typed. Then the control program P.FOR, written in the Fortran 77 language, is obtained. Furthermore, the compilation of the control program P.FOR, followed by its loading and executing, is started if the statement

COMPIL *output_name*

is typed. Finally, all the UFO system phases are performed if the statement

UFOGO *input_name*

is typed. Here *input_name* is the first part of the batch file name that is used as an batch input file for the control program generation and *output_name* is the first part of the text file that is used as an text output from the UFO system. The batch file name must always have the form *input_name*.UFO with the extension UFO and the text file name must have the form *output_name*.OUT with the extension OUT. If GENER, UFOGO statements do not contain the batch input file specification then a full dialogue mode is considered (the batch file name is STANDARD.UFO in this case). If COMPIL statement does not contain a text file specification then the standard text file name is P.OUT. The UFOGO statement has the same meaning as the two statements GENER and COMPIL.

First we show how the batch mode proceeds. We suppose that the model function has the form

$$f^F(x) = 100(x_1^2 - x_2)^2 + (x_1 - 1)^2$$

(the Rosenbrock function) and the starting point is $x_1 = -1.2$ and $x_2 = 1.0$. If we prepare the batch input file P.UFO of the form

```
$SET(INPUT)
  X(1)=-1.2D0; X(2)= 1.0D0
$ENDSET
$SET(FMODELF)
  FF=1.0D2*(X(1)**2-X(2))**2+(X(1)-1.0D0)**2
$ENDSET
$NF=2
$NOUT=1
$BATCH
$STANDARD
```

and type the statement UFOGO P.UFO, then the following results appear in the output file P.OUT

```
  0  NIT=  43  NFV= 147  NFG=   0  NDC=   0  NCG=   0  F= .233D-15  G= .164D-07
FF=   .2333078060D-15
X =   .9999999847D+00   .9999999694D+00
TIME= 0:00:00.66
```

Batch files are written in the UFO control language. This language is described in section 1.4. Here we note that a certain experience with the UFO control language can be obtained by using the demo-files PROB01.UFO,..., PROB19.UFO. These demo-files contain 19 test problems described in chapter 6. We can solve them by using the statements UFOGO PROB01,..., UFOGO PROB19.

Besides the batch mode, we can use the full dialogue mode. The full dialogue mode is started if we type the statement UFOGO (without a batch input file specification). An example which demonstrates the full dialogue mode applied to the Rosenbrock function is given in Appendix A.

1.3. The UFO control language

The form of the control program is determined by using statements of the UFO control language. The UFO control language is based on the batch editing language (BEL) [127], which is described in Appendix B, and it contains three types of instructions:

1. Standard Fortran 77 instructions which can be written in the free format.
2. Fortran 77 instructions containing macrovariables. These instructions get a final form after the first pass of the UFO preprocessor.
3. Special macroinstructions. These macroinstructions control the UFO preprocessor execution.

Standard Fortran 77 instructions used in the UFO control language have some extensions and limitations. The main extension is the free format. The instructions do not have a limited length, they can be written everywhere in the input file and if they are written in the same line then the character ';' is used to separate the instructions. The continuation of an instruction is specified by the character '& '. The main limitation concerns the placement of instructions in the control program. Therefore, statement numbers greater than 9999 cannot be used, comments can be introduced by the character $'*'$ only and the only continuation character can be '& '. Also, it is recommended to use identifiers beginning with the character 'W' which are not used in the UFO system

Macrovariables used in the UFO system begin with the character '$' and they are supposed to be of the type character. Their values are always in the form of a string of characters which can be sometimes interpreted as an integer or a real or a logical constant. The chief significance of the macrovariables is their use in substituting their values for their names in the Fortran 77 statements. In this case we place the macrovariable (beginning with '$') in the text, but if it is followed by a letter or digit we have to use brackets. For example if we write

```
      $FLOAT W(100)
```

or

```
      CALL UD$HESF$TYPE$DECOMP$NUMBER
```

or

```
      X(1)=1.0$(P)0
```

and if the values of $FLOAT, $HESF, $TYPE, $DECOMP, $NUMBER and $P are 'REAL*8' (it is standard), 'D', 'L', 'G' '1' and 'D' (it is standard) then we get REAL*8 W(100) or CALL UDDLG1 or X(1)=1.0D0 respectively after the UFO preprocessor application. The values of macrovariables can be defined by assignments as will be shown later.

The macroinstructions are very important for the UFO control language since they make the substitution of texts, change of macrovariables, branching, loops, etc., possible. We briefly describe the most useful of them. A more detailed description is given in Appendix B.

1. Assignment: The assignment of a string of characters for a macrovariable is specified by the macroinstruction $MACRO='value'. For example, we have to set $HESF='D', $TYPE='L', $DECOMP='G', $NUMBER=1 (the integers do not need to be substituted as strings) to obtain the result given above.

2. Insertion of a text: If we write

```
$SET(MACRO)     or     $ADD(MACRO)
   text                      text
$ENDSET                  $ENDADD
```

   then a given text (that can contain a large number of Fortran 77 statements) is inserted into the macrovariable $MACRO. The macroinstruction $SET is used for the definition of a new macrovariable. The macroinstruction $ADD appends a new text into the old macrovariable so that it can be used repeatedly.

3. Logical macrovariables: The macrovariables $INT, $REAL, $LOG and $DEF have logical values. If we write $INT(MACRO) (or $REAL(MACRO) or $LOG(MACRO)), then the resulting value is either .TRUE., if the value of the macrovariable $MACRO is an integer constant (or real constant or logical constant), or .FALSE. in the oposite case. If we write $DEF(MACRO) then the value of $DEF is either .TRUE., if the macrovariable $MACRO was previously defined (by the substitution $MACRO='value' or by using macroinstructions $SET and $ADD), or .FALSE. in the oposite case. This possibility can be used for branching. If we use the macroinstruction $ERASE(MACRO), then the previously defined macrovariable $MACRO becomes undefined (so that $DEF(MACRO)=.FALSE.).

4. List of items macrovariables: Values of macrovariables can be lists of items, i.e. they can have a more complicated form $MACRO='item_1 \item_2\...\item_n' where every item corresponds to one value. The list of items macrovariables use pointers which point out the current items. The current item can be obtained by the macroinstruction $DATA(MACRO) which also moves the pointer to the next item. The macroinstruction $RESTORE(MACRO) returns the pointer to the first item.

5. Branching: This possibility is very similar to the branching in the Fortran 77 language:

```
$IF(condition)
   statements
$ELSEIF(condition)
   statements
$ELSE
   statements
$ENDIF
```

   Conditions can be logical constants .TRUE., .FALSE., or logical macrovariables $INT(MACRO), $REAL(MACRO), $LOG(MACRO)), $DEF(MACRO), or they can have a form of comparisons MACRO=MACRO1, MACRO='value' etc. (besides the relation = we can also use the other relations < or > or <= or >= or <>). Branching is used in the UFO preprocessor stage and it has an influence on the form of the control program.

6. Loops: The basic looping macroinstructions have the form (similarly as in the Fortran 77 or Pascal languages):

```
$DO(MACRO=INDEX1,INDEX2,INDEX3)
   statements
$ENDDO
```

or

```
$REPEAT
   statements
$UNTIL(condition)
```

For example if we set $NF=2, $NC=3 and write

```
$DO(I=1,NF,1)
  $DO(J=1,NC,1)
    CALL UKMCI1($I,$J,$I.0D0+$J.0D0,ICG,JCG,CG)
  $ENDDO
$ENDDO
```

then the UFO preprocessor generates the sequence

```
CALL UKMCI1(1,1,1.0D0+1.0D0,ICG,JCG,CG)
CALL UKMCI1(1,2,1.0D0+2.0D0,ICG,JCG,CG)
CALL UKMCI1(1,3,1.0D0+3.0D0,ICG,JCG,CG)
CALL UKMCI1(2,1,2.0D0+1.0D0,ICG,JCG,CG)
CALL UKMCI1(2,2,2.0D0+2.0D0,ICG,JCG,CG)
CALL UKMCI1(2,3,2.0D0+3.0D0,ICG,JCG,CG)
```

Similarly, if we set $FLOAT='REAL*8' $N=20, $MACRO='X($N)\G($N)\H($N,$N)\.END.', and write

```
$REPEAT
   $I='DATA(MACRO)'
   $FLOAT $I
$UNTIL(I='.END.')
```

then the UFO preprocessor generates the sequence

```
REAL*8 X(20)
REAL*8 G(20)
REAL*8 H(20,20)
```

7. Substitution of a file: Suppose we have a file with a name file_ name.extension. Then we can include it into the control program by using the macroinstructions

```
$INCLUDE('file_name.extension')
```

or

```
$SUBST('file_name.extension')
```

The main difference between these possibilities is that the macroinstruction $INCLUDE includes a text without change (it has to be a regular Fortran 77 text with a fixed format) while the macroinstruction $SUBST substitutes a text executed consecutively by the UFO preprocessor (so it can contain the macrovariables and macroinstructions and it can be written in the free format). Moreover, if this text contains a template, then the macroinstruction $SUBST substitutes only this template. This possibility is widely used for control program generation by using intermediate templates. If the included file has the name file_name.I, then we can use a simpler form without extension. For example, the file UZLINS.I can be substituted by using the macroinstruction $SUBST('UZLINS').

8. Special macroinstructions: Besides macroinstructions of the batch editing language BEL, the UFO control language contains special macroinstructions which control the UFO preprocessor:

$BATCH	- switch to the batch mode.
$DIALOGUE	- switch to the dialogue mode.
$GLOBAL	- global declarations.
$INITIATION	- initiation of the global variables.
$INPUT	- user supplied input.
$OUTPUT	- user supplied output.
$METHOD	- generation of the optimization method.
$MODERASE	- cancelation of the current model.
$METERASE	- cancelation of the current method.
$TSTART	- start of the time measurement.
$TSTOP	- termination of the time measurement and print of the measured time.
$END	- end of the optimization block.
$STANDARD	- standard optimization block: The macroinstruction $STANDARD substitutes the sequence of macroinstructions $GLOBAL, $INITIATION, $MODERASE, $INPUT, $METHOD, $OUTPUT, $TSTOP.

9. Standard macrovariables: The macrovariables $FLOAT or $P have standard values 'REAL*8' or 'D' respectively. This possibility has a meaning for a precision free notation. If we write

   ```
   $FLOAT WA,WB
   WA=2.0$(P)1
   WB=1.0$(P)2
   ```

   then after the UFO preprocessor execution we have

   ```
   REAL*8 WA,WB
   WA=2.0D0
   WB=1.0D2
   ```

   The macrovariables $FLOAT and $P are defined in the installation template and they can be changed when we wish to use single precision computations.

We have described the basic possibilities of the UFO control language that are sufficient for preparing the batch input file. More details are given in subsequent sections and especially in Appendix B. The following example demonstrates the use of the UFO control language for the solution of three collections of optimization problems by two selected methods.

```
$REM ---------------  basic parameters  ---------------

$TOLX='1.0$P-10'; $TOLF='1.0$P-15'; $TOLG='1.0$P-5'; $MIT=800; $MFV=1200
$KOUT=0; $LOUT=1; $MOUT=1

$BATCH
$GLOBAL
$ADD(INTEGER,'\IAG($NA+1)\JAG($MA)')

$REM ---------------  the first method  ---------------

$CLASS='VM'; $TYPE='L'; $DECOMP='M'; $NUMBER=3; $UPDATE='B'

$REM ---------------  the first model   ---------------

$MODEL='AF'; $JACA='S'; $HESF='S'; $NF=100; $NA=500; $MA=2000; $M=9000
$SET(INPUT)
  CALL EIUB14(NF,NA,MA,X,IAG,JAG,FMIN,XMAX,NEXT,IEXT,IERR)
  IF(IERR.NE.0) GO TO 7777
$ENDSET
$SET(FMODELA)
  CALL EAFU14(NF,KA,X,FA,NEXT)
$ENDSET
$SET(GMODELA)
  CALL EAGU14(NF,KA,X,GA,NEXT)
$ENDSET

$REM ---------------  the first solver  ---------------

$INITIATION
$MODERASE
  CALL UYTES1
  DO 7777 NEXT=1,15
  CALL UYTES2
$INPUT
$METHOD
  CALL UYTES3
  7777 CONTINUE

$REM ---------------  the second method ---------------

$METERASE
$CLASS='GN'; $TYPE='L'; $DECOMP='M'; $NUMBER=3; $UPDATE='D'

$REM ---------------  the second model  ---------------

$MODEL='AQ'; $JACA='S'; $HESF='S'; $NF=100; $NA=500; $MA=2000; $M=9000
$SET(INPUT)
  CALL EIUB15(NF,NA,MA,X,IAG,JAG,FMIN,XMAX,NEXT-15,IEXT,IERR)
  IF(IERR.NE.0) GO TO 8888
$ENDSET
$SET(FMODELA)
```

```
  CALL EAFU15(NF,KA,X,FA,NEXT-15)
$ENDSET
$SET(GMODELA)
  CALL EAGU15(NF,KA,X,GA,NEXT-15)
$ENDSET

$REM ---------------  the second solver ---------------

$INITIATION
$MODERASE
  DO 8888 NEXT=16,37
  CALL UYTES2
$INPUT
$METHOD
  CALL UYTES3
  8888 CONTINUE

$REM ---------------  the third model   ---------------

$SET(INPUT)
  CALL EIUB18(NF,NA,MA,X,IAG,JAG,FMIN,XMAX,NEXT-37,IEXT,IERR)
  IF(IERR.NE.0) GO TO 9999
$ENDSET
$SET(FMODELA)
  CALL EAFU18(NF,KA,X,FA,NEXT-37)
$ENDSET
$SET(GMODELA)
  CALL EAGU18(NF,KA,X,GA,NEXT-37)
$ENDSET

$REM ---------------  the third solver  ---------------

$INITIATION
$MODERASE
  DO 9999 NEXT=38,65
  CALL UYTES2
$INPUT
$METHOD
  CALL UYTES3
  9999 CONTINUE

$REM ---------------  the final action  ---------------

  CALL UOTES4
$END
```

1.4. Problem description and method selection by using the UFO control language

If we want to process either the batch mode or the mixed mode we have to prepare a batch input file written in the UFO control language. This input file prescribes the structure of the control program. If some macrovariable is used, it has to have been previously defined. Therefore definitions of macrovariables usually lie in the beginning of the input file. Many macrovariables serves for the defini-

tion of a given optimization problem. The most important among them are macrovariables which define problem functions, specifically the model (or objective) function, approximating functions for nonlinear approximation, constrain functions for nonlinear programming, state functions, initial functions and the terminal function for optimization of dynamical systems. These functions are specified by using special macrovariables whose names are consisted of three parts. The first part can contain letters F, G, D, H or their combinations:

F - function value.
G - gradient with respect to basic variables.
D - gradient with respect to state variables.
H - Hessian matrix with respect to basic variables.
FG - function value and gradient with respect to basic variables.
FD - function value and gradient with respect to state variables.
GD - gradient with respect to basic variables and gradient with respect to state variables.
FGD - function value, gradient with respect to basic variables and gradient with respect to state variables.
FGH - function value, gradient with respect to basic variables and Hessian matrix with respect to basic variables.

The second part always has the form MODEL. The third part can contain letters F, A, C, E, Y and also an additional letter S:

F - the model function or the terminal function.
A - the selected approximating function.
AS - all approximating functions.
C - the selected constraint function.
CS - all constraint functions.
E - the selected state function.
ES - all state functions.
Y - the selected initial function.
YS - all initial functions.

The following combinations are possible:

$FMODELF	$FMODELA	$FMODELC	$FMODELE	$FMODELY
	$FMODELAS	$FMODELCS	$FMODELES	$FMODELYS
$GMODELF	$GMODELA	$GMODELC	$GMODELE	$GMODELY
	$GMODELAS	$GMODELCS	$GMODELES	$GMODELYS
$DMODELF	$DMODELA		$DMODELE	
			$DMODELES	
$HMODELF	$HMODELA	$HMODELC		
	$HMODELAS	$HMODELCS		
$FGMODELF	$FGMODELA	$FGMODELC	$FGMODELE	$FGMODELY
	$FGMODELAS	$FGMODELCS	$FGMODELES	$FGMODELYS
$FDMODELF	$FDMODELA		$FDMODELE	
			$FDMODELES	
$GDMODELF	$GDMODELA		$GDMODELE	
			$GDMODELES	
$FGDMODELF	$FGDMODELA		$FGDMODELE	
			$FGDMODELES	
$FGHMODELF	$FGHMODELA	$FGHMODELC		
	$FGHMODELAS	$FGHMODELCS		

Choice of a suitable way for problem functions definitions is ambiguous and problem dependent. We can only state several remarks:

1. The basic and most general way is the use of different macrovariables for different quantities (values, gradients, Hessian matrices) together with an independent evaluation of individual functions

(the last letter is different from S). This way saves the computer storage and frequently also the computational time.

2. Sometimes, evaluation of gradients require function values. In this case, it can be advantageous to compute values and gradients simultaneously. Similar consideration holds also for Hessian matrices.

3. Even if simultaneous evaluation of all aproximating (constraint, state, initial) functions increase storage requirements, it can be advantageous if complicated computations exist which are common for all such functions and also if a problem has a low dimension or sparse structure. Frequently it is advantageous for the evaluation of state and initial functions when dynamical systems are optimized.

4. If gradients of aproximating (constraint, state, initial) functions are computed simultaneously (the last letter is equal to S), then also function values have to be computed simultaneously. Similarly if Hessian matrices are computed simultaneously, then also function values and gradients have to be computed simultaneously.

A simple example of a batch input file was shown in section 1.2. We repeat it here with some explanations:

```
$SET(INPUT)
  X(1)=-1.2D0; X(2)= 1.0D0
$ENDSET
$SET(FMODELF)
  FF=1.0D2*(X(1)**2-X(2))**2+(X(1)-1.0D0)**2
$ENDSET
$NF=2
$NOUT=1
$BATCH
$STANDARD
```

By using the macrovariable $INPUT, we specify initial values of variables $x_1 = -1.2$ and $x_2 = 1.0$. By using the macrovariable $FMODELF, we specify the model function value (the model function gradient is not specified, it will be computed numerically). The macrovariable $NF defines the number of variables and $NOUT is a print specification. The macroinstruction $BATCH switches the mode to the batch mode. The macroinstruction $STANDARD defines a standard form of the control program. Descriptions of more complicated problems are shown in chapter 5.

In the above example, a direct definition of a model function value is used. We can also use indirect specifications by means of the Fortran 77 subroutines or the files prepared beforehand. Suppose that the model function value is defined by using the subroutine EFFU01 or it is specified in the file FVAL.FOR. Then we can write:

```
$SET(FMODELF)
  CALL EFFU01(NF,X,FF,NEXT)
$ENDSET
```

or

```
$SET(FMODELF)
  $INCLUDE('FVAL.FOR')
$ENDSET
```

or

```
$SET(FMODELF)
  $SUBST('FVAL.FOR')
$ENDSET
```

The last possibility is useful if the model function value specification is written in a free format or it contains the BEL macroinstructions.

If we need to utilize user supplied subroutines, we can include them into the control program using the macrovariable $SUBROUTINES:

```
$SET(SUBROUTINES)
  user supplied subroutines
$ENDSET
```

In this case, some exceptions laid on the text of user supplied subroutines, forced by the UFO preprocessor, have to be satisfied: All comments have to begin by the character '*', the continuation line has to begin by the character '&', the character '$' has to be replaced by '$$' and the character ';' does not have to be present.

The batch input file should also contain the optimization method selection. Fortunately, this selection is not critical since the optimization method can be chosen automatically by using knowledge bases contained in the UFO system templates. Here we will only demonstrate some possibilities. The greatest influence on the optimization method selection have the following macrovariables:

$CLASS	- class of optimization methods (heuristic, conjugate gradient, variable metric, Newton, Gauss-Newton, quasi-Newton, proximal bundle, bundle-Newton).
$TYPE	- type of optimization methods (line search, trust region).
$DECOMP	- type of matrix decomposition (original matrix, Choleski decomposition, inversion).
$NUMBER	- individual methods for direction determination (various direct, various iterative).
$UPDATE	- type of variable metric or quasi-Newton update.

A more detailed description of these choices together with other choices ($MET, $MET1, $MET2, $MET3, $MES, $MES1, $MES2, $MES3, $MOS, $MOS1, $MOS2, $MOS3) is given in section 3.

1.5. The UFO environment

The UFO environment can be used on personal computers (PC) with processors 386/486/586, with the operating system MS DOS version 5.0 or higher and with the Microsoft FORTRAN 77 compiler version 5.0 or higher or the Microsoft Fortran Power Station compiler version 1.0 or higher.

The UFO environment is called by using the statement UFO (program UFO.EXE). It is controlled by using the "pull-down" menu. The main menu is activated by pressing the key <F10> . The UFO environment contains a source program editor whose control is similar to the Word Star editor and, therefore, to the most commonly used source program editors under the MS DOS system (for example Turbo Pascal). All significant statements of the source program editor are available from the UFO environment menu.

Since the UFO environment menu is clearly understood we do not describe it (the description is given in [128]) . We only show the usual way for operating input files. When the batch mode input file is prepared by using the source program editor we press the key <F10> and find the command Run! in the UFO environment menu. This command starts the UFO preprocessor and its action corresponds to the statement UFOGO (with the present input file). An easier possibility is pressing the keys <Alt-1> . Similarly, pressing the keys <Alt-9> has the same effect as the statement GENER. Furthermore, if the control program P.FOR is loaded in the source program editor, pressing the keys <Alt-3> has the same effect as the statement COMPIL and pressing the keys <Alt-5> causes an exit from the UFO environment.

2. Problems solved using the UFO system

The most general problem which can be solved by using the UFO system is a minimization of an objective function $F : R^n \to R$ over a set $X \subset R^n$. The objective function can have several forms determined using the macrovariable $MODEL:

$MODEL='FF' - general optimization. In this case

$$F(x) = \pm f^F(x)$$

where $f^F : R^n \to R$ is a real valued, so-called model function

$MODEL='FL' - linear optimization. In this case

$$F(x) = \pm \ (f^F + \sum_{i=1}^{n} g_i^F x_i)$$

where f^F, g_i^F, $1 \le i \le n$, are real coefficients.

$MODEL='FQ' - quadratic optimization. In this case

$$F(x) = \pm \ (f^F + \sum_{i=1}^{n} (g_i^F + \frac{1}{2} \sum_{j=1}^{n} h_{ij}^F x_j) x_i)$$

where f^F, $g_i^F, 1 \le i \le n,$, $h_{ij}^F, 1 \le i \le n, 1 \le j \le n$, are real coefficients.

$MODEL='AF' - sum of functions minimization. In this case

$$F(x) = \sum_{k=1}^{n_A} f_k^A(x)$$

where $f_k^A : R^n \to R, \ 1 \le k \le n_A$, are real valued, so-called approximating functions.

$MODEL='AQ' - Sum of squares minimization. In this case

$$F(x) = \sum_{k=1}^{n_A} |f_k^A(x)|^2$$

where $f_k^A : R^n \to R, \ 1 \le k \le n_A$, are real valued, so-called approximating functions.

$MODEL='AP' - sum of powers minimization. In this case

$$F(x) = \sum_{k=1}^{n_A} |f_k^A(x)|^r$$

where $f_k^A : R^n \to R, \ 1 \le k \le n_A$, are real valued, so-called approximating functions and $1 < r < \infty$ is a real exponent.

$MODEL='AM' - minimization of maximum (minimax). In this case

$$F(x) = \max_{1 \le k \le n_A} |f_k^A(x)|$$

where $f_k^A : R^n \to R, \ 1 \le k \le n_A$, are real valued, so-called approximating functions.

$MODEL='DF' - minimization of general integral criterion with respect to the state equations. In this case

$$F(x) = \int_{t_A^{min}}^{t_A^{max}} f^A(x, y_A(x,t_A), t_A)\, dt_A + f^F(x, y_A(x, t_A^{max}), t_A^{max})$$

and

$$\frac{dy_A(x,t_A)}{dt_A} = f^E(x, y_A(x,t_A), t_A),\ y^A(x, t_A^{min}) = f^Y(x)$$

where $f^A : R^{n+n_E+1} \to R$ is a real valued, smooth, so-called subintegral function, $f^F : R^{n+n_E+1} \to R$ is a real valued, smooth, so-called terminal function, $f^E : R^{n+n_E+1} \to R^{n_E}$ is a real valued, smooth, so-called state function and $f^Y : R^n \to R^{n_E}$ is a real valued, smooth, so-called initial function.

$MODEL='DQ' - minimization of sum of squares integral criterion with respect to the state equations. In this case

$$F(x) = \frac{1}{2} \int_{t_A^{min}}^{t_A^{max}} \sum_{i=1}^{n_E} w_i^E(t_A)(y_i^A(x,t_A) - y_i^E(t_A))^2\, dt_A + \frac{1}{2} \sum_{i=1}^{n_E} w_i^E (y_i^A(x, t_A^{max}) - y_i^E)^2$$

and

$$\frac{dy_A(x,t_A)}{dt_A} = f^E(x, y_A(x,t_A), t_A),\ y^A(x, t_A^{min}) = f^Y(x)$$

where $f^E : R^{n+n_E+1} \to R^{n_E}$ is a real valued, smooth, so-called state function and $f^Y : R^n \to R^{n_E}$ is a real valued, smooth, so-called initial function.

$MODEL='NO' - solving an initial value problem for a system of ordinary differential equations. In this case

$$\frac{dy_A(t_A)}{dt_A} = f^E(y_A(t_A), t_A),\ y^A(t_A^{min}) = y_A^{min}$$

where $f^E : R^{n_E+1} \to R^{n_E}$ is a real valued, smooth, so-called state function.

The objective function defined by the choice $MODEL='AQ' can be used for the solution of a system of nonlinear equations

$$f_k^A(x) = 0, \qquad 1 \le k \le n_A$$

In this case we suppose $n_A = n$. This case is considered separately, since for $n_A = n$ special methods for systems of nonlinear equations can be used.

The model function $f^F : R^n \to R$ can have several types of Hessian matrices specified by the macrovariable $HESF:

$HESF='D' - dense Hessian matrix.
$HESF='S' - sparse Hessian matrix with a general pattern.
$HESF='NO' - Hessian matrix is not used.

The default option is $HESF='D'. The approximating functions $f_k^A : R^n \to R,\ 1 \le k \le n_A$, can have several types of Jacobian matrices specified by the macrovariable $JACA:

$JACA='D' - dense Jacobian matrix.
$JACA='S' - sparse Jacobian matrix with a general pattern.

$JACA='NO' - Jacobian matrix is not used.

If the approximating functions are used then we can choose several types of the Hessian matrix representation. These types are again specified by the macrovariable $HESF:

$HESF='D' - dense Hessian matrix.
$HESF='S' - sparse Hessian matrix with a general pattern.
$HESF='B' - sparse Hessian matrix with a partitioned pattern
$HESF='NO' - Hessian matrix is not used.

If $JACA='D', then it must be either $HESF='D' or $HESF='NO'. If $JACA='S', then we can specify all types of Hessian matrices ($HESF='D', $HESF='S', $HESF='B', $HESF='NO'). The representation $HESF='B' usually leads to more expensive matrix operations. Therefore, we recommend to prefer the choice $HESF='S' against the choice $HESF='B'.

The subintegral function, terminal function, state function and initial function, appeared in the case of dynamical systems optimization, are considered to be dense. Therefore we cannot use the specifications $HESF='S' or $HESF='B' in this case.

The set $X \subset R^n$ can be whole R^n (unconstrained case) or it can be defined by box constraints

$$\begin{array}{rcl} x_i^L \le x_i & & \text{if } i \in I_1 \\ x_i \le x_i^U & & \text{if } i \in I_2 \\ x_i^L \le x_i \le x_i^U & & \text{if } i \in I_3 \\ x_i^L = x_i & & \text{if } i \in I_5 \end{array}$$

where $I_1 \cup I_2 \cup I_3 \cup I_5 \subset \{i \in N : 1 \le i \le n\}$, by general linear constraints

$$\begin{array}{rcl} c_k^L \le \sum_{i=1}^{n} g_{ki}^C x_i & & \text{if } k \in L_1 \\ \sum_{i=1}^{n} g_{ki}^C x_i \le c_k^U & & \text{if } k \in L_2 \\ c_k^L \le \sum_{i=1}^{n} g_{ki}^C x_i \le c_k^U & & \text{if } k \in L_3 \\ c_k^L = \sum_{i=1}^{n} g_{ki}^C x_i & & \text{if } k \in L_5 \end{array}$$

where g_{ki}^C, $1 \le k \le n_C$, $1 \le i \le n$, are real coefficients and $L_1 \cup L_2 \cup L_3 \cup L_5 \subset \{k \in N : 1 \le k \le n_C\}$, or by general nonlinear constraints

$$\begin{array}{rcl} c_k^L \le f_k^C(x) & & \text{if } k \in N_1 \\ f_k^C(x) \le c_k^U & & \text{if } k \in N_2 \\ c_k^L \le f_k^C(x) \le c_k^U & & \text{if } k \in N_3 \\ c_k^L = f_k^C(x) & & \text{if } k \in N_5 \end{array}$$

where $f_k^C : R^n \to R$, $1 \le k \le n_C$, are real valued, smooth, so-called constraint functions and $N_1 \cup N_2 \cup N_3 \cup N_5 \subset \{k \in N : 1 \le k \le n_C\}$. The constraint functions $f_k^C : R^n \to R$, $1 \le k \le n_C$, can have several types of Jacobian matrices specified by the macrovariable $JACC:

$JACC='D' - dense Jacobian matrix.
$JACC='S' - sparse Jacobian matrix with a general pattern.

If $JACC='D', then must be $HESF='D' or $HESF='NO'. If $JACC='S', then must be $HESF='S' or $HESF='NO'.

There are several limitations in the current version of the UFO system:

1. Minimization of maximum (minimax) and nonsmooth optimization is not implemented in the sparse case.
2. Minimization of dynamical systems is not implemented in the sparse case.
3. Usually the UFO system serves for local optimization. Global optimization can be used only for relatively small ($n \le 20$) dense problems that are unconstrained or that contain box constraints.

These limitations will be consecutively removed in subsequent versions of the UFO system.

In the rest of this report we will use the notation NF, NA, NC instead of n, n_A, n_C and X, FF, GF, HF, FA, GA, FC, GC instead of x, f^F, g^F, h^F, f^A, g^A, f^C, g^C. This new notation corresponds to the notation of the variables and the fields in the UFO system.

2.1. Specification of variables

First we must specify the number of variables using the statement $NF=number_of_variables. If there are no box constraints we set $KBF=0. In the opposite case we set $KBF=1 or $KBF=2. If $KBF=1 or $KBF=2 then

$$\begin{array}{lcll} \mathrm{X(I)} & - & \text{unbounded} & \text{, if IX(I)} = 0 \\ \mathrm{XL(I)} & \le & \mathrm{X(I)} & \text{, if IX(I)} = 1 \\ & & \mathrm{X(I)} \le \mathrm{XU(I)} & \text{, if IX(I)} = 2 \\ \mathrm{XL(I)} & \le & \mathrm{X(I)} \le \mathrm{XU(I)} & \text{, if IX(I)} = 3 \\ \mathrm{X(I)} & - & \text{constant} & \text{, if IX(I)} = 5 \end{array}$$

where 1≤ I≤ NF. The option $KBF=2 must be chosen if IX(I)=3 for at least one index 1≤ I≤ NF. Then two different fields XL(I) and XU(I), 1≤ I≤ NF are declared. In the opposite case we set $KBF=1 and only one common field XL(I)=XU(I), $1 \le$ I≤ NF is declared.

Initial values of variables X(I), 1≤ I≤ NF, types of box constraints IX(I), 1≤ I≤ NF, and lower and upper bounds XL(I) and XU(I), 1≤ I≤ NF, can be specified using macrovariable $INPUT. The default values are IX(I)=0 and XL(I)=XU(I)=0, 1≤ I≤ NF. For example:

$KBF=2; $NF=4
$SET(INPUT)
X(1)=x_1
X(2)=x_2; IX(2)=1; XL(1)=x_1^L
X(3)=x_3; IX(3)=3; XL(3)=x_3^L; XU(3)=x_3^U
X(4)=x_4; IX(4)=5
$ENDSET

The UFO system allows us to use a scaling of variables (for instance if the values of variables differ very much in their magnitude). We set:

$NORMF=1 - scaling parameters XN(I), 1≤ I≤ NF, are determined automatically so that X(I)/XN(I)=1, 1≤I≤NF, for the initial values of variables.

$NORMF=2 - scaling parameters must be specified by the user by means of the macrovariable $INPUT.

The scaling of variables is recommended only in exceptional cases since it increases the computational time and storage requirements. The scaling of variables is suppressed if $NORMF=0 (this value is default). The scaling of variables is not permitted in the case of general constraints (if KBC>0).

2.2. Specification of the model function (dense problems)

If the macrovariable $MODEL is not specified or if $MODEL='FF', then the objective function is defined by the formula

F(X) = + FF(X) if $IEXT = 0 (minimization)

or

F(X) = − FF(X) if $IEXT = 1 (maximization)

Option $IEXT=0 is default.

The model function FF(X) must be defined by the user either directly in the full dialogue mode, or by using corresponding macrovariables in the batch (or mixed) mode. The value of the model function is specified by using the macrovariable $FMODELF:

$SET(FMODELF)
FF = value FF(X)
(for given values of variables X(I), 1≤ I≤ NF)
$ENDSET

The first derivatives of the model function are specified by using the macrovariable $GMODELF:

$SET(GMODELF)
GF(1) = derivative ∂FF(X)/ ∂X(1)
GF(2) = derivative ∂FF(X)/ ∂X(2)
—
GF(NF) = derivative ∂FF(X)/ ∂X(NF)
(for given values of variables X(I), 1≤I≤NF)
$ENDSET

The second derivatives of the model function are specified by using the macrovariable $HMODELF. If $HESF='D', then the Hessian matrix is assumed to be dense and we specify only its upper half:

$SET(HMODELF)
HF(1) = derivative ∂^2FF(X)/ ∂ X(1)^2
HF(2) = derivative ∂^2FF(X)/ ∂X(1)∂X(2)
HF(3) = derivative ∂^2FF(X)/ ∂X(2)^2
HF(4) = derivative ∂^2FF(X)/ ∂X(1) ∂X(3)
HF(5) = derivative ∂^2FF(X)/ ∂X(2) ∂X(3)
HF(6) = derivative ∂^2FF(X)/ ∂X(3)^2
—
HF(NF*(NF+1)/2) = derivative ∂^2FF(X)/ ∂X(NF)^2
(for given values of variables X(I), 1≤I≤NF)
$ENDSET

If the macrovariables $GMODELF or $HMODELF are not defined, we suppose that the first or the second derivatives of the model function are not given analytically. In this case, they are computed numerically by using the UFO system routines whenever it is required. If it is advantageous to compute the first derivatives of the model function FF(X) together with its value, we can include the models

$FMODELF and $GMODELF into the common model $FGMODELF. Similarly we can include the models $FMODELF, $GMODELF and $HMODELF into the common model $FGHMODELF.

To improve the efficiency of the computation, we can specify additional information about the model function FF(X). The first piece of information, useful for an automatic choice of the optimization method, is the computational complexity specified by the macrovariable $KCF:

$KCF=1 - evaluation of the model function FF(X) is very easy (it takes at most $O(NF)$ simple operations).

$KCF=2 - evaluation of the model function FF(X) is of medium complexity (it takes at least $O(NF)$ complicated operations and at most $O(NF^2)$ simple operations).

$KCF=3 - evaluation of the model function FF(X) is extremely difficult (it takes at least $O(NF^2)$ complicated operations or $O(NF^3)$ simple operations).

The option $KCF=2 is default. An additional useful piece of information is the analytical complexity (differentiability and conditioning), which is specified by the macrovariable $KSF:

$KSF=1 - the model function FF(X) is smooth and well-conditioned.

$KSF=2 - the model function FF(X) is smooth but ill-conditioned.

$KSF=3 - the model function FF(X) is nonsmooth.

The option $KSF=1 is default. Other specifications, which can improve the computational efficiency and robustness of optimization methods, are a lower bound of the objective function values and an upper bound of the stepsize. Both these values depend on a definition of the objective function and can be specified by the statements $FMIN=lower_bound (for the objective function) and $XMAX=upper_bound (for the stepsize). We recommend a definition of $FMIN whenever it is possible and a definition of $XMAX whenever the objective function contains exponentials.

If $MODEL='FL', we suppose the model function is linear of the form

$$\mathrm{FF(X)} = \mathrm{FF} + \sum_{\mathrm{I=1}}^{\mathrm{NF}} \mathrm{GF(I)} * \mathrm{X(I)}$$

In this case we need not specify the value and the first derivatives of the model function by the macrovariables $FMODELF and $GMODELF as in the general case. Instead, we must specify the coefficients FF (constant value) and GF(I), 1≤ I≤ NF, (constant gradient) using the macrovariable $INPUT:

```
$ADD(INPUT)
   FF = constant value
   GF(1) = constant derivative ∂FF(X)/∂X(1)
   GF(2) = constant derivative ∂FF(X)/∂X(2)
   ——
   GF(NF) = constant derivative ∂FF(X)/∂X(NF)
$ENDADD
```

If $MODEL='FL', we usually assume that either box constraints or general linear constraints are given. In this case the optimization problem is the linear programming problem.

If $MODEL='FQ', we suppose the model function is quadratic of the form

$$\mathrm{FF(X)} = \mathrm{FF} + \sum_{\mathrm{I=1}}^{\mathrm{NF}} \mathrm{GF(I)} * \mathrm{X(I)} + \frac{1}{2} \sum_{\mathrm{I=1}}^{\mathrm{NF}} \sum_{\mathrm{J=1}}^{\mathrm{NF}} \mathrm{HF(K)} * \mathrm{X(I)} * \mathrm{X(J)}$$

where K=MAX(I,J)*(MAX(I,J)-1)/2+MIN(I,J). In this case we need not specify the value, the first derivatives and the second derivatives of the model function by the macrovariables $FMODELF, $GMODELF and $HMODELF as in the general case. The coefficients FF (constant value) and GF(I), 1≤ I≤ NF, (constant gradient) are specified in the same way as in the linear case. The coefficients HF(K), 1≤ K≤ NF*(NF+1)/2, (the constant Hessian matrix) must be specified using the macrovariable $INPUT. If $HESF='D', then the Hessian matrix is assumed to be dense and we specify only its upper half:

$ADD(INPUT)
HF(1) = constant derivative $\partial^2 FF(X)/\partial X(1)^2$
HF(2) = constant derivative $\partial^2 FF(X)/\partial X(1)\partial\ X(2)$
HF(3) = constant derivative $\partial^2 FF(X)/\partial X(2)^2$
HF(4) = constant derivative $\partial^2 FF(X)/\partial X(1)\partial\ X(3)$
HF(5) = constant derivative $\partial^2 FF(X)/\partial X(2)\partial\ X(3)$
HF(6) = constant derivative $\partial^2 FF(X)/\partial X(3)^2$
—
HF(NF*(NF+1)/2) = constant derivative $\partial^2 FF(X)/\partial X(NF)^2$
$ENDADD

If $MODEL='FQ', we usually assume that either box constraints or general constraints are given. In this case the optimization problem is the quadratic programming problem.

If the model function is linear or quadratic, then the options $KCF and $KSF need not be defined, since they are not used.

2.3. Specification of the model function (sparse problems)

The UFO system contains optimization methods that take into account the sparsity pattern of the Hessian matrix HF. This possibility decreases computational time and storage requirements for large-scale optimization problems. In this case we use the option $HESF='S' which means that the sparsity pattern is specified. All other specifications remain the same as in the case of dense problems. The sparsity pattern of the Hessian matrix is specified by using the macrovariable $INPUT. Two integer vectors IH and JH are used where IH(I), $1 \leq I \leq NF+1$, are pointers and JH(K), $1 \leq K \leq M$, are indices of nonzero elements. Only the upper half of the Hessian matrix is assumed and the nonzero elements are ordered in rows. The number of nonzero elements must be specified using the statement $M=number_of_elements. The number of nonzero elements could be greater then is required (eg. two times) since it is used for the declaration of working fields. For example, if we have the Hessian matrix

$$HF = \begin{pmatrix} h^F_{11}, & h^F_{12}, & h^F_{13}, & 0, & h^F_{15} \\ h^F_{21}, & h^F_{22}, & 0, & h^F_{24}, & 0 \\ h^F_{31}, & 0, & h^F_{33}, & 0, & h^F_{35}, \\ 0, & h^F_{42}, & 0, & h^F_{44}, & 0, \\ h^F_{51}, & 0, & h^F_{53}, & 0, & h^F_{55} \end{pmatrix}$$

then we have to set:

$NF=5
$M=20 (the minimum required value is M=10)
$ADD(INPUT)
IH(1)=1; IH(2)=5; IH(3)=7
IH(4)=9; IH(5)=10; IH(6)=11
JH(1)=1; JH(2)=2; JH(3)=3; JH(4)=5; JH(5)=2
JH(6)=4; JH(7)=3; JH(8)=5; JH(9)=4; JH(10)=5
$ENDADD

All diagonal elements of the sparse Hessian matrix are assumed to be nonzero.

As in the case of the dense problem, second derivatives of the model function can be specified by using the macrovariable $HMODELF. If $HESF='S', then only nonzero elements of the upper half (including the diagonal) of the Hessian matrix are specified. For the above example the specification has the form:

$SET(HMODELF)
HF(1)=h^F_{11}; HF(2)=h^F_{12}; HF(3)=h^F_{13}; HF(4)=h^F_{15}
HF(5)=h^F_{22}; HF(6)=h^F_{24}; HF(7)=h^F_{33}; HF(8)=h^F_{35}
HF(9)=h^F_{44}; HF(10)=h^F_{55}
$ENDSET

If the model function is quadratic (i.e. if $MODEL='FQ') and if $HESF='S', then the coefficients HF(K), 1≤ K≤ M, (constant sparse Hessian matrix) must be specified by using the macrovariable $INPUT. If the matrix given in the above example is the constant sparse Hessian matrix, we use the specification:

$ADD(INPUT)
HF(1)=h^F_{11}; HF(2)=h^F_{12}; HF(3)=h^F_{13}; HF(4)=h^F_{15}
HF(5)=h^F_{22}; HF(6)=h^F_{24}; HF(7)=h^F_{33}; HF(8)=h^F_{35}
HF(9)=h^F_{44}; HF(10)=h^F_{55}
$ENDADD

2.4. Objective functions for discrete approximation

If we set $MODEL='AF', then we suppose that the objective function F(X) has the form:

$$\mathrm{F(X)} = \sum_{\mathrm{KA=1}}^{\mathrm{NA}} \mathrm{FA(KA;X)} \qquad \text{if KBA} = 0$$

or

$$\mathrm{F(X)} = \sum_{\mathrm{KA=1}}^{\mathrm{NA}} \mathrm{AW(KA)} * (\mathrm{FA(KA;X)} - \mathrm{AM(KA)}) \qquad \text{if KBA} = 1$$

where FA(KA;X), 1≤ KA≤ NA, are approximating functions. This form of the objective function is very useful in large-scale optimization when the approximating functions FA(KA;X), 1≤ KA≤ NA, are assumed to have sparse gradients.

If we set $MODEL='AP', then we suppose that the objective function F(X) has the form:

$$\mathrm{F(X)} = \frac{1}{\mathrm{R}} \sum_{\mathrm{KA=1}}^{\mathrm{NA}} |\mathrm{FA(KA;X)}| **\mathrm{R} \qquad \text{if KBA} = 0$$

or

$$\mathrm{F(X)} = \frac{1}{\mathrm{R}} \sum_{\mathrm{KA=1}}^{\mathrm{NA}} |\mathrm{AW(KA)} * (\mathrm{FA(KA;X)} - \mathrm{AM(KA)})| **\mathrm{R} \qquad \text{if KBA} = 1$$

where FA(KA;X), 1≤ KA≤ NA, are approximating functions, and R>1 is a real exponent. The value of the exponent is specified by the choice $REXP=R (default value is $REXP=2). Since the most used value of the exponent is R=2, and since the computations are simplest and the most efficient for such a choice, we can use the specification $MODEL='AQ' in this case (minimization of sum of squares). Moreover, $MODEL='AQ' is formally set whenever we chose $MODEL='AP' and $REXP=2.

If we set $MODEL='AM', then we suppose that the objective function F(X) has the form:

$$\mathrm{F(X)} = \max_{1\le \mathrm{KA}\le \mathrm{NA}} (+\mathrm{FA(KA;X)}) \quad \text{if } \$\mathrm{IEXT} = -1$$

$$\mathrm{F(X)} = \max_{1\le \mathrm{KA}\le \mathrm{NA}} (|\mathrm{FA(KA;X)}|) \quad \text{if } \$\mathrm{IEXT} = 0$$

$$F(X) = \max_{1 \le KA \le NA} (-FA(KA;X)) \quad \text{if \$IEXT} = +1$$

for \$KBA=0, or

$$F(X) = \max_{1 \le KA \le NA} (+AW(KA) * (FA(KA;X) - AM(KA))) \quad \text{if \$IEXT} = -1$$

$$F(X) = \max_{1 \le KA \le NA} (|AW(KA) * (FA(KA;X) - AM(KA))|) \quad \text{if \$IEXT} = 0$$

$$F(X) = \max_{1 \le KA \le NA} (-AW(KA) * (FA(KA;X) - AM(KA))) \quad \text{if \$IEXT} = +1$$

for \$KBA=1, where FA(KA;X), $1 \le KA \le NA$, are approximating functions. The default value is \$IEXT=0 (the minimax or the Chebyshev approximation).

The option \$KBA serves as a decision between a simple objective function and a more complicated one. The simple objective function uses no additional fields, while the more complicated one uses at most two additional fields, AM and AW. The vector AM usually contains frequently used observations which can be included into the functions FA(KA;X), $1 \le KA \le NA$, in the case of the simple objective function. Observations AM(KA), $1 \le KA \le NA$, are specified by using the macrovariable \$INPUT. Their default values are AM(KA)=0, $1 \le KA \le NA$. The vector AW serves for possible scaling specified by the option \$NORMA:

\$NORMA=0 - no scaling is performed. In this case AW(KA)=1, $1 \le KA \le NA$.

\$NORMA=1 - scaling parameters are determined automatically so that AW(KA)=|AM(KA)|, $1 \le KA \le NA$.

\$NORMA=2 - scaling parameters must be specified by the user by means of the macrovariable \$INPUT.

The number of approximating functions NA must be specified, in all the above cases, by using the statement \$NA=number_of_functions.

2.5. Specification of the approximating functions (dense problems)

The approximating functions FA(KA;X), $1 \le KA \le NA$, must be defined by the user either directly in the full dialogue mode, or by using corresponding macrovariables in the batch (or mixed) mode. Values of the approximating functions are specified by using the macrovariables \$FMODELA or \$FMODELAS:

```
$SET(FMODELA)
      FA = value FA(KA;X)
      (for a given index KA and given values of variables X(I), 1≤ I≤ NF)
$ENDSET
```

or

```
$SET(FMODELAS)
      AF(1) = value FA(1;X)
      AF(2) = value FA(2;X)
      ---
      AF(NA) = value FA(NA;X)
$ENDSET
```

The first derivatives of the approximating functions are specified by using the macrovariables \$GMODELA or \$GMODELAS:

$SET(GMODELA)
GA(1) = derivative ∂FA(KA;X)/∂X(1)
GA(2) = derivative ∂FA(KA;X)/∂X(2)
——
GA(NF) = derivative ∂FA(KA;X)/∂X(NF)
(for a given index KA and given values of variables X(I), 1$\leq$ I$\leq$ NF)
$ENDSET

or

$SET(GMODELAS)
AG(1) = derivative ∂FA(1;X)/∂X(1)
AG(2) = derivative ∂FA(1;X)/∂X(2)
——
AG(NF) = derivative ∂FA(1;X)/∂X(NF)
AG(NF+1) = derivative ∂FA(2;X)/∂X(1)
AG(NF+2) = derivative ∂FA(2;X)/∂X(2)
——
AG(NA*NF) = derivative ∂FA(NA;X)/∂X(NF)
$ENDSET

The second derivatives of the approximating functions are specified by using the macrovariables $HMODELA or $HMODELAS. If $JACA='D', then the Hessian matrices are assumed to be dense and we specify only their upper half:

$SET(HMODELA)
HA(1) = derivative ∂^2FA(KA;X)/∂X(1)^2
HA(2) = derivative ∂^2FA(KA;X)/∂X(1)∂X(2)
HA(3) = derivative ∂^2FA(KA;X)/∂X(2)^2
HA(4) = derivative ∂^2FA(KA;X)/∂X(1)∂X(3)
HA(5) = derivative ∂^2FA(KA;X)/∂X(2)∂X(3)
HA(6) = derivative ∂^2FA(KA;X)/∂X(3)^2
——
HA(NF*(NF+1)/2) = derivative ∂^2FA(KA;X)/∂X(NF)^2
(for a given index KA and given values of variables X(I), 1$\leq$ I$\leq$ NF)
$ENDSET

or

$SET(HMODELAS)
AH(1) = derivative ∂^2FA(1;X)/∂X(1)^2
AH(2) = derivative ∂^2FA(1;X)/∂X(1)∂X(2)
AH(3) = derivative ∂^2FA(1;X)/∂X(2)^2
AH(4) = derivative ∂^2FA(1;X)/∂X(1)∂X(3)
AH(5) = derivative ∂^2FA(1;X)/∂X(2)∂X(3)
AH(6) = derivative ∂^2FA(1;X)/∂X(3)^2
——
AH(NF*(NF+1)/2) = derivative ∂^2FA(1;X)/∂X(NF)^2
AH(NF*(NF+1)/2+1) = derivative ∂^2FA(2;X)/∂X(1)^2
AH(NF*(NF+1)/2+2) = derivative ∂^2FA(2;X)/∂X(1)∂X(2)
AH(NF*(NF+1)/2+3) = derivative ∂^2FA(2;X)/∂X(2)^2
——
AH(NA*NF*(NF+1)/2) = derivative ∂^2FA(NA;X)/∂X(NF)^2
$ENDSET

If the macrovariables $GMODELA and $GMODELAS or $HMODELA and $HMODELAS are not defined, we suppose that the first or the second derivatives of the approximating functions are not given analytically. In this case, they are computed numerically by using the UFO system routines, whenever it is required. If it is advantageous to compute first derivatives of the approximating functions FA(KA;X), 1≤ KA≤ NA, together with their values, we can collect the models $FMODELA, $GMODELA into the common model $FGMODELA and the models $FMODELAS, $GMODELAS into the common model $FGMODELAS. Similarly we can collect the models $FMODELA, $GMODELA, $HMODELA into the common model $FGHMODELA and the models $FMODELAS, $GMODELAS, $HMODELAS into the common model $FGHMODELAS.

To improve the efficiency of the computation, we can specify additional information about the approximating functions FA(KA;X), 1≤ KA≤ NA. The first piece of information, useful for an automatic choice of the optimization method, is the computational complexity specified by the macrovariable $KCA:

$KCA=1 - evaluations of the approximating functions FA(KA;X), 1≤ KA≤ NA, are very easy (they take at most $O(NF)$ simple operations).

$KCA=2 - evaluations of the approximating functions FA(KA;X), 1≤ KA≤ NA, are of medium complexity (they take at least O(NF) complicated operations and at most $O(NF^2)$ simple operations).

$KCA=3 - evaluations of the approximating functions FA(KA;X), 1≤ KA≤ NA, are extremely difficult (they take at least $O(NF^2)$ complicated or $O(NF^3)$ simple operations).

The option $KCA=2 is default. An additional useful piece of information is the analytical complexity (conditioning), which is specified by the macrovariable $KSA:

$KSA=1 - the approximating functions FA(KA;X), 1≤ KA≤ NA, are smooth and well-conditioned.

$KSA=2 - the approximating functions FA(KA;X), 1≤ KA≤ NA, are smooth but ill-conditioned.

$KSA=3 - the approximating functions FA(KA;X), 1≤ KA≤ NA, are nonsmooth.

The option $KSA=1 is default.

If some of the approximating functions are linear having the form

$$\mathrm{FA(KA;\,X)} = \sum_{\mathrm{I=1}}^{\mathrm{NF}} \mathrm{AG((KA-1)*NF+I)*X(I)}$$

we can specify them separately. Then the number of linear approximating functions must be specified by using the statement $NAL=number_of_linear_functions (default value is $NAL=0). We always suppose that the first NAL approximating functions are linear. Then the coefficients AG((KA-1)*NF+I), 1≤ KA≤ NAL, 1≤ I≤ NF, are specified using the macrovariable $INPUT and the macrovariables $FMODELA or $FMODELAS, $GMODELA or $GMODELAS, $HMODELA or $HMODELAS are used only for the specification of the nonlinear approximating functions FA(KA;X), NAL<KA≤ NA.

2.6. Specification of the approximating functions (sparse problems)

The UFO system contains optimization methods that take into account the sparsity pattern of the Jacobian matrix AG. This possibility decreases computational time and storage requirements for large-scale optimization problems. In this case, we use the option $JACA='S' which means that the sparsity pattern is specified. All other specifications remain the same as in the case of dense problems. The sparsity pattern of the Jacobian matrix is specified by using the macrovariable $INPUT. Two integer vectors IAG and JAG are used where IAG(KA), 1≤ KA≤ NA+1, are pointers and JAG(K), 1≤ K≤ IAG(NA+1)-1, are indices of nonzero elements. Nonzero elements are ordered by the gradients of the approximating functions. The number of nonzero elements must be specified by using the statement $MA=number_of_elements. For example, if we have the gradients

$$\mathrm{GA}(1;\mathrm{X}) = [g^A_{11}, 0\ \ , 0\ \ , g^A_{14}],$$

$$\mathrm{GA}(2;\mathrm{X}) = [0\ \ , g^A_{22}, 0\ \ , g^A_{24}],$$

$$\mathrm{GA}(3;\mathrm{X}) = [0\ \ , 0\ \ , g^A_{33}, 0\ \ ],$$

$$\mathrm{GA}(4;\mathrm{X}) = [g^A_{41}, g^A_{42}, g^A_{43}, 0\ \ ],$$

$$\mathrm{GA}(5;\mathrm{X}) = [0\ \ , 0\ \ , g^A_{53}, g^A_{54}],$$

and the Jacobian matrix

$$\mathrm{AG}(\mathrm{X}) = \begin{pmatrix} g^A_{11} & ,0 & ,0 & ,g^A_{14} \\ 0 & ,g^A_{22} & ,0 & ,g^A_{24} \\ 0 & ,0 & ,g^A_{33} & ,0 \\ g^A_{41} & ,g^A_{42} & ,g^A_{43} & ,0 \\ 0 & ,0 & ,g^A_{53} & ,g^A_{54} \end{pmatrix}$$

then we have to set:

```
$NA=5
$MA=10
$ADD(INPUT)
      IAG(1)=1; IAG(2)=3; IAG(3)=5
      IAG(4)=6; IAG(5)=9; IAG(6)=11
      JAG(1)=1; JAG(2)=4; JAG(3)=2; JAG(4)=4; JAG(5)=3
      JAG(6)=1; JAG(7)=2; JAG(8)=3; JAG(9)=3; JAG(10)=5
$ENDADD
```

As in the case of the dense problem, the first derivatives of the approximating functions can be specified by using the macrovariables $GMODELA or $GMODELAS. If $JACA='S', then only nonzero elements of the gradients are specified. For the above example the specifications have the form

```
$SET(GMODELA)
      IF (KA.EQ.1) THEN
            GA(1) = ∂FA(1;X)/∂X(1)
            GA(4) = ∂FA(1;X)/∂X(4)
      ELSE IF (KA.EQ.2) THEN
            GA(2) = ∂FA(2;X)/∂X(2)
            GA(4) = ∂FA(2;X)/∂X(4)
      ELSE IF (KA.EQ.3) THEN
            GA(3) = ∂FA(3;X)/∂X(3)
      ELSE IF (KA.EQ.4) THEN
            GA(1) = ∂FA(4;X)/∂X(1)
            GA(2) = ∂FA(4;X)/∂X(2)
            GA(3) = ∂FA(4;X)/∂X(3)
      ELSE
            GA(3) = ∂FA(5;X)/∂X(3)
            GA(4) = ∂FA(5;X)/∂X(4)
      ENDIF
$ENDSET
```

or

```
$SET(GMODELAS)
      AG(1) = ∂FA(1;X)/∂X(1)
      AG(2) = ∂FA(1;X)/∂X(4)
      AG(3) = ∂FA(2;X)/∂X(2)
      AG(4) = ∂FA(2;X)/∂X(4)
      AG(5) = ∂FA(3;X)/∂X(3)
      AG(6) = ∂FA(4;X)/∂X(1)
      AG(7) = ∂FA(4;X)/∂X(2)
      AG(8) = ∂FA(4;X)/∂X(3)
      AG(9) = ∂FA(5;X)/∂X(3)
      AG(10) = ∂FA(5;X)/∂X(4)
$ENDSET
```

As in the case of the dense problem, the second derivatives of the approximating functions can be specified by using the macrovariables $HMODELA or $HMODELAS. If $JACA='S', then only nonzero elements of the Hessian matrices are specified. For the above example the specifications have the form

```
$SET(HMODELA)
      IF (KA.EQ.1) THEN
            HA(1) = ∂²FA(1;X)/∂X(1)²
            HA(2) = ∂²FA(1;X)/∂X(1)∂X(4)
            HA(3) = ∂²FA(1;X)/∂X(4)²
      ELSE IF (KA.EQ.2) THEN
            HA(1) = ∂²FA(2;X)/∂X(2)²
            HA(2) = ∂²FA(2;X)/∂X(2)∂X(4)
            HA(3) = ∂²FA(2;X)/∂X(4)²
      ELSE IF (KA.EQ.3) THEN
            HA(1) = ∂²FA(3;X)/∂X(3)²
      ELSE IF (KA.EQ.4) THEN
            HA(1) = ∂²FA(4;X)/∂X(1)²
            HA(2) = ∂²FA(4;X)/∂X(1)∂X(2)
            HA(3) = ∂²FA(4;X)/∂X(2)²
            HA(4) = ∂²FA(4;X)/∂X(1)∂X(3)
            HA(5) = ∂²FA(4;X)/∂X(2)∂X(3)
            HA(6) = ∂²FA(4;X)/∂X(3)²
      ELSE
            HA(1) = ∂²FA(5;X)/∂X(3)²
            HA(2) = ∂²FA(5;X)/∂X(3)∂X(4)
            HA(3) = ∂²FA(5;X)/∂X(4)²
      ENDIF
$ENDSET
```

or

```
$SET(HMODELAS)
      AH(1) = ∂²FA(1;X)/∂X(1)²
      AH(2) = ∂²FA(1;X)/∂X(1)∂X(4)
      AH(3) = ∂²FA(1;X)/∂X(4)²
      AH(4) = ∂²FA(2;X)/∂X(2)²
      AH(5) = ∂²FA(2;X)/∂X(2)∂X(4)
      AH(6) = ∂²FA(2;X)/∂X(4)²
      AH(7) = ∂²FA(3;X)/∂X(3)²
```

AH(8) = ∂^2FA(4;X)/∂X(1)^2
AH(9) = ∂^2FA(4;X)/∂X(1)∂X(2)
AH(10) = ∂^2FA(4;X)/∂X(2)^2
AH(11) = ∂^2FA(4;X)/∂X(1)∂X(3)
AH(12) = ∂^2FA(4;X)/∂X(2)∂X(3)
AH(13) = ∂^2FA(4;X)/∂X(3)^2
AH(14) = ∂^2FA(5;X)/∂X(3)^2
AH(15) = ∂^2FA(5;X)/∂X(3)∂X(4)
AH(16) = ∂^2FA(5;X)/∂X(4)^2
$ENDSET

Note that dimensions of arrays HA or AH must be specified by the statement $MHA=dimension_of_HA or $MAH=dimension_of_AH.

If some of the approximating functions are linear (i.e. if $NAL>0) and if $JACA='S', then the coefficients AG(K), 1$\leq$ K$\leq$ IAG(NAL+1)$-$1 (constant part of the sparse Jacobian matrix), must be specified by using the macrovariable $INPUT. If the matrix given in the above example is the constant sparse Jacobian matrix, we use the specification:

$ADD(INPUT)
AG(1)=g_{11}^A; AG(2)=g_{14}^A; AG(3)=g_{22}^A; AG(4)=g_{24}^A
AG(5)=g_{33}^A; AG(6)=g_{41}^A; AG(7)=g_{42}^A; AG(8)=g_{43}^A
AG(9)=g_{53}^A; AG(10)=g_{54}^A
$ENDADD

There is another possibility which can be useful when all approximating functions are linear. It is based on the usage of special procedure UKMAI1 that serves for direct input of individual Jacobian matrix elements. The procedure UKMAI1 is formally called by using the statement

CALL UKMAI1(K,I,GAKI,IAG,JAG,AG)

where K is an index of a given approximating function (row of the Jacobian matrix), I is an index of a given variable (column of the Jacobian matrix), and GAKI is a numerical value of the element ∂FA(K;X)/∂X(I). For the example given above we can write:

$ADD(INPUT)
CALL UKMAI1(1,1,g_{11}^A,IAG,JAG,AG)
CALL UKMAI1(1,4,g_{14}^A,IAG,JAG,AG)
CALL UKMAI1(2,2,g_{22}^A,IAG,JAG,AG)
CALL UKMAI1(2,4,g_{24}^A,IAG,JAG,AG)
CALL UKMAI1(3,3,g_{33}^A,IAG,JAG,AG)
CALL UKMAI1(4,1,g_{41}^A,IAG,JAG,AG)
CALL UKMAI1(4,2,g_{42}^A,IAG,JAG,AG)
CALL UKMAI1(4,3,g_{43}^A,IAG,JAG,AG)
CALL UKMAI1(5,3,g_{53}^A,IAG,JAG,AG)
CALL UKMAI1(5,4,g_{54}^A,IAG,JAG,AG)
$ENDADD

The main advantage of the last possibility is the fact that it is not necessary to specify the fields IAG and JAG beforehand.

If we use the option $JACA='S', then we can specify a form of the objective function sparse Hessian matrix. There are four possibilities:

$HESF='D' - dense Hessian matrix.

$HESF='B' - partitioned sparse Hessian matrix. This matrix is a sum of simple Hessian matrices which correspond to the individual approximating functions. Only nonzero blocks are stored.

$HESF='S' - general sparse Hessian matrix (the same as the model function Hessian matrix corresponding to the option $HESF='S').

$HESF='NO' - Hessian matrix is not used.

This specification serves only for an internal realization of optimization methods and has no influence on the user's input. The default option is $HESF='D'.

2.7. Objective functions for optimization of dynamical systems

If we set $MODEL='DF', then we suppose that the objective function F(X) has the form:

$$\mathrm{F(X)} = \int_{\mathrm{TAMIN}}^{\mathrm{TAMAX}} \mathrm{FA(X, YA(TA), TA)}\, d\mathrm{TA} + \mathrm{FF(X, YA(TAMAX), TAMAX)}$$

where FA(X,YA(TA),TA) is a smooth subintegral function and FF(X,YA(TAMAX),TAMAX) is a smooth terminal function. At the same time

$$\frac{d\mathrm{YA(KE;TA)}}{d\mathrm{TA}} = \mathrm{FE(KE; X, YA(TA), TA),\ YA(KE; TAMIN) = FY(KE; X)}$$

where FE(KE;X,YA(TA),TA), 1≤ KE≤ NE, are smooth state functions and FY(KE;X), 1≤ KE≤ NE, are smooth initial functions.

If we set $MODEL='DQ', then we suppose the objective function F(X) has the form:

$$\mathrm{F(X)} = \frac{1}{2}\int_{\mathrm{TAMIN}}^{\mathrm{TAMAX}} \sum_{\mathrm{KE=1}}^{\mathrm{NE}} \mathrm{WE(KE;TA)} * (\mathrm{YA(KE;TA)} - \mathrm{YE(KE;TA)})^2\, d\mathrm{TA}$$
$$+\frac{1}{2}\sum_{\mathrm{KE=1}}^{\mathrm{NE}} \mathrm{EW(KE)} * (\mathrm{YA(KE;TAMAX)} - \mathrm{EY(KE)})^2$$

At the same time

$$\frac{d\mathrm{YA(KE;TA)}}{d\mathrm{TA}} = \mathrm{FE(KE; X, YA(TA), TA),\ YA(KE; TAMIN) = FY(KE; X)}$$

where FE(KE;X,YA(TA),TA), 1≤ KE≤ NE, are smooth state functions and FY(KE;X), 1≤ KE≤ NE, are smooth initial functions.

If we set $MODEL='NO', then we consider the initial value problem

$$\frac{d\mathrm{YA(KE;TA)}}{d\mathrm{TA}} = \mathrm{FE(KE; YA(TA), TA),\ YA(KE; TAMIN) \text{ is given}}$$

where FE(KE;YA(TA),TA), 1≤ KE≤ NE, are smooth state functions.

In all the above cases, the statement $NE=number_of_differential_equations must be used for the specification of number of differential equations NE.

2.8. Specification of the state functions

The state functions FE(KE;X,YA(TA),TA), 1≤ KE≤ NE, must be defined by the user either directly in the full dialogue mode, or by using corresponding macrovariables in the batch (or mixed) mode. Values of the state functions are specified by using the macrovariables $FMODELE or $FMODELES:

```
$SET(FMODELE)
        FE = value FE(KE;X,YA(TA),TA)
        (for a given index KE, given vector of variables X,
        given vector of state variables YA(TA) and given time TA)
$ENDSET
```

or

```
$SET(FMODELES)
        EF(1) = value FE(1;X,YA(TA),TA)
        EF(2) = value FE(2;X,YA(TA),TA)
        ---
        EF(NE) = value FE(NE;X,YA(TA),TA)
$ENDSET
```

The first derivatives of the state functions according to the variables are specified by using the macrovariables $GMODELE or $GMODELES:

```
$SET(GMODELE)
        GE(1) = derivative ∂FE(KE;X,YA(TA),TA)/∂X(1)
        GE(2) = derivative ∂FE(KE;X,YA(TA),TA)/∂X(2)
        ---
        GE(NF) = derivative ∂FE(KE;X,YA(TA),TA)/∂X(NF)
        (for a given index KE, given vector of variables X,
        given vector of state variables YA(TA) and given time TA)
$ENDSET
```

or

```
$SET(GMODELES)
        EG(1) = derivative ∂FE(1;X,YA(TA),TA)/∂X(1)
        EG(2) = derivative ∂FE(1;X,YA(TA),TA)/∂X(2)
        ---
        EG(NF) = derivative ∂FE(1;X,YA(TA),TA)/∂X(NF)
        EG(NF+1) = derivative ∂FE(2;X,YA(TA),TA)/∂X(1)
        EG(NF+2) = derivative ∂FE(2;X,YA(TA),TA)/∂X(2)
        ---
        EG(NE*NF) = derivative ∂FE(NE;X,YA(TA),TA)/∂X(NF)
$ENDSET
```

The first derivatives of the state functions according to the state variables are specified by using the macrovariables $DMODELE or $DMODELES:

```
$SET(DMODELE)
        DE(1) = derivative ∂FE(KE;X,YA(TA),TA)/∂YA(1)
        DE(2) = derivative ∂FE(KE;X,YA(TA),TA)/∂YA(2)
        ---
        DE(NE) = derivative ∂FE(KE;X,YA(TA),TA)/∂YA(NE)
        (for a given index KE, given vector of variables X,
```

given vector of state variables YA(TA) and given time TA)
$ENDSET

or

$SET(DMODELES)
ED(1) = derivative ∂FE(1;X,YA(TA),TA)/∂YA(1)
ED(2) = derivative ∂FE(1;X,YA(TA),TA)/∂YA(2)
——
ED(NE) = derivative ∂FE(1;X,YA(TA),TA)/∂YA(NE)
ED(NE+1) = derivative ∂FE(2;X,YA(TA),TA)/∂YA(1)
ED(NE+2) = derivative ∂FE(2;X,YA(TA),TA)/∂YA(2)
——
ED(NE*NE) = derivative ∂FE(NE;X,YA(TA),TA)/∂YA(NE)
$ENDSET

If it is advantageous to compute first derivatives of the state functions FE(KE;X,YA(TA),TA), 1$\leq$ KE$\leq$ NE, together with their values, we can collect the models $FMODELE, $GMODELE, $DMODELE into the common model $FGDMODELE and the models $FMODELES, $GMODELES, $DMODELES into the common model $FGDMODELES. Partially we can collect the models $FMODELE, $GMODELE or $FMODELE, $DMODELE or $GMODELE, $DMODELE into the common models $FGMODELE or $FDMODELE or $GDMODELE respectively. Similarly we can collect the models $FMODELES, $GMODELES or $FMODELES, $DMODELES or $GMODELES, $DMODELES into the common models $FGMODELES or $FDMODELES or $GDMODELES respectively.

If $MODEL='DQ' we have to define the functions WE(KE;TA) and YE(KE;TA), 1$\leq$ KE$\leq$ NE, for a given index KE and given time TA. These functions can be specified by using the macrovariable $FMODELE together with the state function FE(KE;X,YA(TA),TA):

$SET(FMODELE)
FE = value FE(KE;X,YA(TA),TA)
WE = value WE(KE;TA)
YE = value YE(KE;TA)
(for a given index KE, given vector of variables X,
given vector of state variables YA(TA) and given time TA)
$ENDSET

The default values WE(KE;TA)=1 and YE(KE;TA)=0 cannot be specified, they are supposed automatically.

2.9. Specification of the initial functions

The initial functions FY(KE;X), 1$\leq$ KE$\leq$ NE, must be defined by the user either directly in the full dialogue mode, or by using corresponding macrovariables in the batch (or mixed) mode. Values of the initial functions are specified by using the macrovariables $FMODELY or $FMODELYS:

$SET(FMODELY)
FE = value FY(KE;X)
(for a given index KE and given vector of variables X)
$ENDSET

or

$SET(FMODELYS)
EF(1) = value FY(1;X)
EF(2) = value FY(2;X)

```
        ——
        EF(NE) = value FY(NE;X)
$ENDSET
```

The first derivatives of the initial functions according to the variables are specified by using the macrovariables $GMODELY or $GMODELYS:

```
$SET(GMODELY)
        GE(1) = derivative ∂FY(KE;X)/∂X(1)
        GE(2) = derivative ∂FY(KE;X)/∂X(2)
        ——
        GE(NF) = derivative ∂FY(KE;X)/∂X(NF)
        (for a given index KE and given vector of variables X)
$ENDSET
```

or

```
$SET(GMODELYS)
        EG(1) = derivative ∂FY(1;X)/∂X(1)
        EG(2) = derivative ∂FY(1;X)/∂X(2)
        ——
        EG(NF) = derivative ∂FY(1;X)/∂X(NF)
        EG(NF+1) = derivative ∂FY(2;X)/∂X(1)
        EG(NF+2) = derivative ∂FY(2;X)/∂X(2)
        ——
        EG(NE*NF) = derivative ∂FY(NE;X)/∂X(NF)
$ENDSET
```

If it is advantageous to compute first derivatives of the initial functions FY(KE;X), $1 \leq$ KE$\leq$ NE, together with their values, we can collect the models $FMODELY, $GMODELY into the common model $FGMODELY and the models $FMODELYS, $GMODELYS into the common model $FGMODELYS.

If the initial values YA(KE;TAMIN), $1 \leq$ KE$\leq$ NE, do not depend on the variables X(I), $1 \leq$ I$\leq$ NF, they can be specified by using the macrovariable $INPUT:

```
$ADD(INPUT)
        YA(1) = initial value YA(1,TAMIN)
        YA(2) = initial value YA(2,TAMIN)
        ——
        YA(NE) = initial value YA(NE,TAMIN)
$ENDADD
```

2.10. Specification of the subintegral function

If $MODEL='DF', then the subintegral function FA(X,YA(TA),TA) must be defined by the user either directly in the full dialogue mode, or by using corresponding macrovariables in the batch (or mixed) mode. Value of the subintegral function is specified by using the macrovariable $FMODELA:

```
$SET(FMODELA)
        FA = value FA(X,YA(TA),TA)
        (for a given vector of variables X, given vector of state variables YA(TA)
        and given time TA)
$ENDSET
```

The first derivatives of the subintegral function according to the variables are specified by using the macrovariable $GMODELA:

```
$SET(GMODELA)
      GA(1) = derivative ∂FA(X,YA(TA),TA)/∂X(1)
      GA(2) = derivative ∂FA(X,YA(TA),TA)/∂X(2)
      ——
      GA(NF) = derivative ∂FA(X,YA(TA),TA)/∂X(NF)
      (for a given vector of variables X, given vector of state variables YA(TA)
      and given time TA)
$ENDSET
```

The first derivatives of the subintegral function according to the state variables are specified by using the macrovariable $DMODELA:

```
$SET(DMODELA)
      DA(1) = derivative ∂FA(X,YA(TA),TA)/∂YA(1)
      DA(2) = derivative ∂FA(X,YA(TA),TA)/∂YA(2)
      ——
      DA(NE) = derivative ∂FA(X,YA(TA),TA)/∂YA(NE)
      (for a given vector of variables X, given vector of state variables YA(TA)
      and given time TA)
$ENDSET
```

If it is advantageous to compute first derivatives of the subintegral function FA(X,YA(TA),TA) together with its value, we can collect the models $FMODELA, $GMODELA and $DMODELA into the common model $FGDMODELA. Partially we can collect the models $FMODELA, $GMODELA or $FMODELA, $DMODELA or $GMODELA, $DMODELA into the common models $FGMODELA or $FDMODELA or $GDMODELA respectively.

If $MODEL='DQ' and the objective function contains an integral part, we have to set $MODELA='YES' and define the functions WE(KE;TA) and YE(KE;TA), $1 \leq$ KE$\leq$ NE, by using the macrovariable $FMODELE.

2.11. Specification of the terminal function

If $MODEL='DF', then the terminal function FF(X,YA(TAMAX),TAMAX) must be defined by the user either directly in the full dialogue mode, or by using corresponding macrovariables in the batch (or mixed) mode. Value of the terminal function is specified by using the macrovariable $FMODELF:

```
$SET(FMODELF)
      FF = value FF(X,YA(TAMAX),TAMAX)
      (for a given vector of variables X, given vector of state variables YA(TAMAX)
      and given time TAMAX)
$ENDSET
```

The first derivatives of the terminal function according to the variables are specified by using the macrovariable $GMODELF:

```
$SET(GMODELF)
      GF(1) = derivative ∂FF(X,YA(TAMAX),TAMAX)/∂X(1)
      GF(2) = derivative ∂FF(X,YA(TAMAX),TAMAX)/∂X(2)
      ——
      GF(NF) = derivative ∂FF(X,YA(TAMAX),TAMAX)/∂X(NF)
```

(for a given vector of variables X, given vector of state variables YA(TAMAX)
and given time TAMAX)
$ENDSET

The first derivatives of the terminal function according to the state variables are specified by using the macrovariable $DMODELF:

$SET(DMODELF)
DF(1) = derivative ∂FF(X,YA(TAMAX),TAMAX)/∂YA(1)
DF(2) = derivative ∂FF(X,YA(TAMAX),TAMAX)/∂YA(2)
——
DF(NE) = derivative ∂FF(X,YA(TAMAX),TAMAX)/∂YA(NE)
(for a given vector of variables X, given vector of state variables YA(TAMAX)
and given time TAMAX)
$ENDSET

If it is advantageous to compute first derivatives of the terminal function FF(X,YA(TAMAX),TAMAX) together with its value, we can collect the models $FMODELF, $GMODELF and $DMODELF into the common model $FGDMODELF. Partially we can collect the models $FMODELF, $GMODELF or $FMODELF, $DMODELF or $GMODELF, $DMODELF into the common models $FGMODELF or $FDMODELF or $GDMODELF respectively.

If $MODEL='DQ' and the objective function contains a terminal part, we have to set $MODELF='YES' and define the coefficients EW(KE) and EY(KE), 1≤ KE≤ NE, by using the macrovariable $INPUT:

$ADD(INPUT)
EW(1) = value EW(1); EY(1) = value EY(1)
EW(2) = value EW(2); EY(2) = value EY(2)
——
EW(NE) = value EW(NE); EY(NE) = value EY(NE)
$ENDADD

2.12. Optimization with general constraints.

If there are no general constraints we set $KBC=0. In the oposite case we set $KBC=1 or $KBC=2. If $KBC=1 or $KBC=2 then

FC(KC;X) - unbounded		, if IC(KC) = 0
CL(KC) ≤	FC(KC;X)	, if IC(KC) = 1
	FC(KC;X) ≤ CU(KC)	, if IC(KC) = 2
CL(KC) ≤	FC(KC;X) ≤ CU(KC)	, if IC(KC) = 3
CL(KC) =	FC(KC;X) = CU(KC)	, if IC(KC) = 5

where 1≤ KC≤ NC.The option $KBC=2 must be chosen if IC(KC)=3 for at least one index 1≤ KC≤ NC. Then two different fields XL(K) and XU(KC), 1≤ KC≤ NC are declared. In the opposite case we set $KBC=1 and only one common field XL(KC)=XU(KC), 1≤ KC≤ NC is declared. The number of constraints NC must be specified by using the statement $NC=number_of_functions.

Types of general constraints IC(KC), 1≤ KC≤ NC, and lower and upper bounds XL(KC) and XU(KC), 1≤ KC≤ NC, can be specified by using the macrovariable $INPUT. Default values are IC(KC)=3 and XL(KC)=XU(KC)=0, 1≤ KC≤ NC. For example:

$KBF=2; $NC=3

$ADD(INPUT)
IC(1)=1; CL(1)=c_1^L
IC(2)=1; CL(2)=c_2^L
IC(3)=3; CL(3)=c_3^L; CU(3)=c_3^L
$ENDADD

2.13. Specification of the constraint functions (dense problems)

The constraint functions FC(KC;X), 1≤ KC≤ NC, must be defined by the user either directly in the full dialogue mode, or by using corresponding macrovariables in the batch (or mixed) mode. Values of the constraint functions are specified by using the macrovariables $FMODELC or $FMODELCS:

$SET(FMODELC)
FC = value FC(KC;X)
(for a given index KC and given values of variables X(I), 1≤ I≤ NF)
$ENDSET

or

$SET(FMODELCS)
CF(1) = value FC(1;X)
CF(2) = value FC(2;X)
——
CF(NC) = value FC(NC;X)
$ENDSET

The first derivatives of the constraint functions are specified by using the macrovariables $GMODELC or $GMODELCS:

$SET(GMODELC)
GC(1) = derivative ∂FC(KC;X)/∂X(1)
GC(2) = derivative ∂FC(KC;X)/∂X(2)
——
GC(NF) = derivative ∂FC(KC;X)/∂X(NF)
(for a given index KC and given values of variables X(I), 1≤ I≤ NF)
$ENDSET

or

$SET(GMODELCS)
CG(1) = derivative ∂FC(1;X)/∂X(1)
CG(2) = derivative ∂FC(1;X)/∂X(2)
——
CG(NF) = derivative ∂FC(1;X)/∂X(NF)
CG(NF+1) = derivative ∂FC(2;X)/∂X(1)
CG(NF+2) = derivative ∂FC(2;X)/∂X(2)
——
CG(NC*NF) = derivative ∂FC(NC;X)/∂X(NF)
$ENDSET

The second derivatives of the constraint functions are specified by using the macrovariables $HMODELC or $HMODELCS. If $JACC='D', then the Hessian matrices are assumed to be dense and we specify only their upper half:

$SET(HMODELC)

HC(1) = derivative $\partial^2 FC(KC;X)/\partial X(1)^2$

HC(2) = derivative $\partial^2 FC(KC;X)/\partial X(1)\partial X(2)$

HC(3) = derivative $\partial^2 FC(KC;X)/\partial X(2)^2$

HC(4) = derivative $\partial^2 FC(KC;X)/\partial X(1)\partial X(3)$

HC(5) = derivative $\partial^2 FC(KC;X)/\partial X(2)\partial X(3)$

HC(6) = derivative $\partial^2 FC(KC;X)/\partial X(3)^2$

——

HC(NF*(NF+1)/2) = derivative $\partial^2 FC(KC;X)/\partial X(NF)^2$

(for a given index KC and given values of variables X(I), $1 \leq I \leq NF$)

$ENDSET

or

$SET(HMODELCS)

CH(1) = derivative $\partial^2 FC(1;X)/\partial X(1)^2$

CH(2) = derivative $\partial^2 FC(1;X)/\partial X(1)\partial X(2)$

CH(3) = derivative $\partial^2 FC(1;X)/\partial X(2)^2$

CH(4) = derivative $\partial^2 FC(1;X)/\partial X(1)\partial X(3)$

CH(5) = derivative $\partial^2 FC(1;X)/\partial X(2)\partial X(3)$

CH(6) = derivative $\partial^2 FC(1;X)/\partial X(3)^2$

——

CH(NF*(NF+1)/2) = derivative $\partial^2 FC(1;X)/\partial X(NF)^2$

CH(NF*(NF+1)/2+1) = derivative $\partial^2 FC(2;X)/\partial X(1)^2$

CH(NF*(NF+1)/2+2) = derivative $\partial^2 FC(2;X)/\partial X(1)\partial X(2)$

CH(NF*(NF+1)/2+3) = derivative $\partial^2 FC(2;X)/\partial X(2)^2$

——

CH(NC*NF*(NF+1)/2) = derivative $\partial^2 FC(NC;X)/\partial X(NF)^2$

$ENDSET

If the macrovariables $GMODELC and $GMODELCS or $HMODELC and $HMODELCS are not defined, we suppose that the first or the second derivatives of the constraint functions are not given analytically. In this case, they are computed numericaly, by using the UFO system routines, whenever it is required. If it is advantageous to compute first derivatives of the constraint functions FC(KC;X), $1 \leq KC \leq NC$, together with their values, we can collect the models $FMODELC, $GMODELC into the common model $FGMODELC and the models $FMODELCS, $GMODELCS into the common model $FGMODELCS. Similarly we can collect the models $FMODELC, $GMODELC, $HMODELC into the common model $FGHMODELC and the models $FMODELCS, $GMODELCS, $HMODELCS into the common model $FGHMODELCS.

To improve the efficiency of the computation, we can specify additional information about the constraint functions FC(KC;X), $1 \leq KC \leq NC$. The first piece of information, useful for an automatic choice of the optimization method, is the computational complexity specified by the macrovariable $KCC:

$KCC= 1 - evaluations of the constraint functions FC(KC;X), $1 \leq KC \leq NC$, are very easy (they take at most $O(NF)$ simple operations).

$KCC= 2 - evaluations of the constraint functions FC(KC;X), $1 \leq KC \leq NC$, are of medium complexity (they take at least $O(NF)$ complicated operations and at most $O(NF^2)$ simple operations).

$KCC= 3 - evaluations of the constraint functions FC(KC;X), $1 \leq KC \leq NC$, are extremely difficult (they take at least $O(NF^2)$ complicated or $O(NF^3)$ simple operations).

The option $KCC=2 is default.

If some of the constraint functions are linear having the form

$$\mathrm{FC(KC;X)} = \sum_{\mathrm{I=1}}^{\mathrm{NF}} \mathrm{CG((KC-1)*NF+I)*X(I)}$$

we can specify them separately. Then the number of linear constraint functions must be specified by using the statement \$NCL=number_of_linear_functions (default value is \$NCL=0). We always suppose that the first NCL constraint functions are linear. Then the coefficients CG((KC-1)*NF+I), 1≤ KC≤ NCL, 1≤ I≤ NF, are specified by using the macrovariable \$INPUT and the macrovariables \$FMODELC or \$FMODELCS, \$GMODELC or \$GMODELCS, \$HMODELC or \$HMODELCS are used only for the specification of the nonlinear constraint functions FC(KC;X), NCL<KC≤ NC.

2.14. Specification of the constraint functions (sparse problems)

The UFO system contains optimization methods that take into account the sparsity pattern of the Jacobian matrix CG. This possibility decreases computational time and storage requirements for large-scale optimization problems. In this case, we use option \$JACC='S' which means that the sparsity pattern is specified. All other specifications remain the same as in the case of dense problems. The sparsity pattern of the Jacobian matrix is specified by using the macrovariable \$INPUT. Two integer vectors ICG and JCG are used where ICG(KC), 1≤ KC≤ NC+1, are pointers and JCG(K), 1≤ K≤ ICG(NC+1)-1, are indices of nonzero elements. Nonzero elements are ordered by the gradients of the constraint functions. The number of nonzero elements must be specified by using the statement \$MC=number_of_elements. The number of nonzero elements could be greater then is needed (two times say) since it is used for declaration of working fields. For example if we have the gradients

$$\mathrm{GC(1;X)} = [g_{11}^C, 0\ \ , 0\ \ , g_{14}^C],$$

$$\mathrm{GC(2;X)} = [0\ \ , g_{22}^C, 0\ \ , g_{24}^C],$$

$$\mathrm{GC(3;X)} = [0\ \ , 0\ \ , g_{33}^C, 0\ \ ],$$

$$\mathrm{GC(4;X)} = [g_{41}^C, g_{42}^C, g_{43}^C, 0\ \ ],$$

$$\mathrm{GC(5;X)} = [0\ \ , 0\ \ , g_{53}^C, g_{54}^C],$$

and the Jacobian matrix

$$\mathrm{CG(X)} = \begin{pmatrix} g_{11}^C & ,0 & ,0 & ,g_{14}^C \\ 0 & ,g_{22}^C & ,0 & ,g_{24}^C \\ 0 & ,0 & ,g_{33}^C & ,0 \\ g_{41}^C & ,g_{42}^C & ,g_{43}^C & ,0 \\ 0 & ,0 & ,g_{53}^C & ,g_{54}^C \end{pmatrix}$$

then we have to set:

```
$NC=5
$MC=20 (the minimum required value is MC=10)
$ADD(INPUT)
        ICG(1)=1; ICG(2)=3; ICG(3)=5
        ICG(4)=6; ICG(5)=9; ICG(6)=11
        JCG(1)=1; JCG(2)=4; JCG(3)=2; JCG(4)=4; JCG(5)=3
        JCG(6)=1; JCG(7)=2; JCG(8)=3; JCG(9)=3; JCG(10)=5
$ENDADD
```

As in the case of the dense problem, the first derivatives of the constraint functions can be specified by using the macrovariables $GMODELC or $GMODELCS. If $JACC='S', then only the nonzero elements of the gradients are specified. For the above example the specification has the form:

```
$SET(GMODELC)
      IF (KC.EQ.1) THEN
            GC(1) = ∂FC(1;X)/∂X(1)
            GC(4) = ∂FC(1;X)/∂X(4)
      ELSE IF (KC.EQ.2) THEN
            GC(2) = ∂FC(2;X)/∂X(2)
            GC(4) = ∂FC(2;X)/∂X(4)
      ELSE IF (KC.EQ.3) THEN
            GC(3) = ∂FC(3;X)/∂X(3)
      ELSE IF (KC.EQ.4) THEN
            GC(1) = ∂FC(4;X)/∂X(1)
            GC(2) = ∂FC(4;X)/∂X(2)
            GC(3) = ∂FC(4;X)/∂X(3)
      ELSE
            GC(3) = ∂FC(5;X)/∂X(3)
            GC(4) = ∂FC(5;X)/∂X(4)
      ENDIF
$ENDSET
```

or

```
$SET(GMODELCS)
      CG(1) = ∂FC(1;X)/∂X(1)
      CG(2) = ∂FC(1;X)/∂X(4)
      CG(3) = ∂FC(2;X)/∂X(2)
      CG(4) = ∂FC(2;X)/∂X(4)
      CG(5) = ∂FC(3;X)/∂X(3)
      CG(6) = ∂FC(4;X)/∂X(1)
      CG(7) = ∂FC(4;X)/∂X(2)
      CG(8) = ∂FC(4;X)/∂X(3)
      CG(9) = ∂FC(5;X)/∂X(3)
      CG(10) = ∂FC(5;X)/∂X(4)
$ENDSET
```

As in the case of the dense problem, the second derivatives of the approximating functions can be specified by using the macrovariableS $HMODELC or $HMODELCS. If $JACC='S', then only nonzero elements of the Hessian matrices are specified. For the above example the specifications have the form

```
$SET(HMODELC)
      IF (KC.EQ.1) THEN
            HC(1) = ∂²FC(1;X)/∂X(1)²
            HC(2) = ∂²FC(1;X)/∂X(1)∂X(4)
            HC(3) = ∂²FC(1;X)/∂X(4)²
      ELSE IF (KC.EQ.2) THEN
            HC(1) = ∂²FC(2;X)/∂X(2)²
            HC(2) = ∂²FC(2;X)/∂X(2)∂X(4)
            HC(3) = ∂²FC(2;X)/∂X(4)²
      ELSE IF (KC.EQ.3) THEN
            HC(1) = ∂²FC(3;X)/∂X(3)²
```

```
      ELSE IF (KC.EQ.4) THEN
            HC(1) = ∂²FC(4;X)/∂X(1)²
            HC(2) = ∂²FC(4;X)/∂X(1)∂X(2)
            HC(3) = ∂²FC(4;X)/∂X(2)²
            HC(4) = ∂²FC(4;X)/∂X(1)∂X(3)
            HC(5) = ∂²FC(4;X)/∂X(2)∂X(3)
            HC(6) = ∂²FC(4;X)/∂X(3)²
      ELSE
            HC(1) = ∂²FC(5;X)/∂X(3)²
            HC(2) = ∂²FC(5;X)/∂X(3)∂X(4)
            HC(3) = ∂²FC(5;X)/∂X(4)²
      ENDIF
$ENDSET
```

or

```
$SET(HMODELCS)
      CH(1) = ∂²FC(1;X)/∂X(1)²
      CH(2) = ∂²FC(1;X)/∂X(1)∂X(4)
      CH(3) = ∂²FC(1;X)/∂X(4)²
      CH(4) = ∂²FC(2;X)/∂X(2)²
      CH(5) = ∂²FC(2;X)/∂X(2)∂X(4)
      CH(6) = ∂²FC(2;X)/∂X(4)²
      CH(7) = ∂²FC(3;X)/∂X(3)²
      CH(8) = ∂²FC(4;X)/∂X(1)²
      CH(9) = ∂²FC(4;X)/∂X(1)∂X(2)
      CH(10) = ∂²FC(4;X)/∂X(2)²
      CH(11) = ∂²FC(4;X)/∂X(1)∂X(3)
      CH(12) = ∂²FC(4;X)/∂X(2)∂X(3)
      CH(13) = ∂²FC(4;X)/∂X(3)²
      CH(14) = ∂²FC(5;X)/∂X(3)²
      CH(15) = ∂²FC(5;X)/∂X(3)∂X(4)
      CH(16) = ∂²FC(5;X)/∂X(4)²
$ENDSET
```

Note that dimensions of arrays HC or CH must be specified by the statement $MHC=dimension_of_HC or $MCH=dimension_of_CH.

If some of the constraint functions are linear (i.e. if $NCL>0) and if $JACC='S', then the coefficients CG(K), $1 \leq K \leq$ ICG(NCL+1)-1 (constant part of the sparse Jacobian matrix), must be specified by using the macrovariable $INPUT. If the matrix given in the above example is the constant sparse Jacobian matrix, we use the specification:

```
$ADD(INPUT)
      CG(1)=g11^C; CG(2)=g14^C; CG(3)=g22^C; CG(4)=g24^C
      CG(5)=g33^C; CG(6)=g41^C; CG(7)=g42^C; CG(8)=g43^C
      CG(9)=g53^C; CG(10)=g54^C
$ENDADD
```

There is another possibility which can be useful when all constraint functions are linear. It is based on the usage of a special procedure UKMCI1 that serves for direct input of individual Jacobian matrix elements. The procedure UKMCI1 is formally called by using the statement

```
      CALL UKMCI1(K,I,GCKI,ICG,JCG,CG)
```

where K is an index of a given constraint function (row of the Jacobian matrix), I is an index of a given variable (column of the Jacobian matrix), and GCKI is a numerical value of the element ∂FC(K;X)/∂X(I). For the example given above we can write:

$ADD(INPUT)
CALL UKMCI1(1,1,g^C_{11},ICG,JCG,CG)
CALL UKMCI1(1,4,g^C_{14},ICG,JCG,CG)
CALL UKMCI1(2,2,g^C_{22},ICG,JCG,CG)
CALL UKMCI1(2,4,g^C_{24},ICG,JCG,CG)
CALL UKMCI1(3,3,g^C_{33},ICG,JCG,CG)
CALL UKMCI1(4,1,g^C_{41},ICG,JCG,CG)
CALL UKMCI1(4,2,g^C_{42},ICG,JCG,CG)
CALL UKMCI1(4,3,g^C_{43},ICG,JCG,CG)
CALL UKMCI1(5,3,g^C_{53},ICG,JCG,CG)
CALL UKMCI1(5,4,g^C_{54},ICG,JCG,CG)
$ENDADD

The main advantage of the last possibility is the fact that it is not necessary to specify beforehand the fields ICG and JCG. If the number of constraints are very large then we can use a slightly more complicated procedure UKMCI2, which uses dynamic structures and therefore works more quickly. The procedure UKMCI2 is formally called by using the statement

CALL UKMCI2(K,I,GCKI,ICG,JCG,CG,LCG)

where K is an index of a given constraint function (row of the Jacobian matrix), I is an index of a given variable (column of the Jacobian matrix), GCKI is a numerical value of the element ∂FC(K;X)/∂X(I) and LCG is an auxiliary working field.

2.15. Additional specifications concerning optimization problems

Useful specifications, which can improve the computational efficiency and robustness of the optimization methods, are a lower bound for the objective function value and an upper bound for the stepsize. Both of these values depend on a definition of the objective function and can be specified by the statements $FMIN=lower_bound (for the objective function value) and $XMAX=upper_bound (for the stepsize). We recommend a definition of $FMIN, whenever it is possible, and a definition of $XMAX, whenever the objective function contains exponentials. If the objective function is a sum of powers (or a sum of squares), then automatically $FMIN=0. The default option for the maximum stepsize is $XMAX=1000.

If there are no general constraints and if the number of variables is not greater than 20, then we can use global optimization methods. A decision between local and global optimization is effected by means of macrovariable $EXTREM:

$EXTREM='L' - a local extremum, that usually contains the starting point in its region of attractivity is found.
$EXTREM='G' - all extrema in the given region are found and a global extremum is determined.

The default option is $EXTREM='L'. If $EXTREM='G', we cannot use the common models $FGMODELF and $FGHMODELF for a common specification of the value , the gradient and the Hessian matrix of the model function. Similarly we cannot use the models $FGMODELA or $FGMODELAS and $FGHMODELA or $FGHMODELAS for a common specification of the approximating functions.

The global optimization is performed over a bounded region specified by lower and upper bounds XL(I) and XU(I), 1$\leq$ I$\leq$ NF. If these bounds are not specified (using the macrovariable $INPUT), they are computed from the initial values of variables and from the given maximum stepsize, so that XL(I)=X(I)$-$XMAX and XU(I)=X(I)+XMAX, 1$\leq$ I$\leq$ NF. The maximum stepsize is specified, as in

the case given above, using the statement $XMAX=maximum_stepsize. The default option is again $XMAX=1000.

Additional useful specifications, concerning the solution precision, are bounds used in termination criteria. These bounds can be specified by the macrovariables $TOLX, $TOLF, $TOLB, $TOLG, $TOLC and MIC, MIT, MFV:

$TOLX - lower bound for a relative change of variables.
$TOLF - lower bound for a relative change of function values.
$TOLB - lower bound for the objective function value.
$TOLG - lower bound for the objective function gradient norm.
$TOLC - lower bound for the violated constraint functions.

$MIC - maximum number of penalty function changes.
$MIT - maximum number of iterations.
$MFV - maximum number of function evaluations.

The default values are $TOLX='1.0D-8', $TOLF='1.0D-16', $TOLB='−1.0D60', $TOLG='1.0D-6', $TOLC='1.0D-6' and MIC=5, MIT=500, MFV=1000.

3. Optimization methods in the UFO system

The UFO system has a modular structure. All optimization methods can be set up using the individual simple modules. For example, the sequential quadratic programming variable metric methods for nonlinearly constrained optimization problems are set up by using the modules for the objective function evaluation, penalty function definition, direction determination, quadratic programming solution, stepsize selection, and variable metric update. Optimization methods contained in the UFO system can be roughly divided into two groups. The first group contains methods for unconstrained and linearly constrained optimization problems, while the second group contains methods for general nonlinear programming problems. Methods for general nonlinear programming problems, i.e. for problems with nonlinear constraints, are classified by their realization form which is determined by using the macrovariable $FORM:

$FORM='SQ' - sequential (or recursive) quadratic programming methods for general dense problems.
$FORM='SM' - sequential (or recursive) minimax optimization methods for general dense problems.
$FORM='SE' - inexact sequential (or recursive) quadratic programming methods for sparse equality constrained problems.

Sections 3.1 - 3.14 concern methods for unconstrained and linearly constrained problems. These methods do not use the macrovariable $FORM for a classification. Methods for general nonlinear programming problems are described in Sections 3.15 - 3.17. Basic parts of optimization methods are described in Sections 3.18 - 3.22. Section 3.23 is devoted to global optimization methods.

Methods for unconstrained and linearly constrained problems contained in the UFO system can be partitioned into several classes which are specified by using the macrovariable $CLASS:

$CLASS='HM' - heuristic methods for small-size problems. This class contains the pattern search method and the simplex method.
$CLASS='CD' - conjugate direction methods that use no matrices. This class contains conjugate direction methods, variable metric methods with limited storage and difference versions of the truncated Newton method.
$CLASS='VM' - variable metric methods that use an approximation of the Hessian matrix which is updated in each iteration.
$CLASS='MN' - modified Newton methods that use the Hessian matrix computed either analytically or numerically.
$CLASS='GN' - modified Gauss-Newton methods for nonlinear least squares problems that use the normal equation matrix as an approximation of the Hessian matrix. These methods are also realized by using the Jacobian matrix representation.
$CLASS='QN' - quasi-Newton methods for nonlinear least squares problems and nonlinear equations.
$CLASS='QL' - quasi-Newton methods with limited storage for sparse nonlinear least squares problems and sparse nonlinear equations.
$CLASS='BD' - biconjugate direction methods for nonlinear equations.
$CLASS='BR' - modified Brent method for nonlinear equations.
$CLASS='LP' - special methods for linear programming problems.
$CLASS='QP' - special methods for quadratic programming problems.
$CLASS='BM' - proximal bundle methods for nonsmooth optimization.
$CLASS='BN' - bundle-Newton methods for nonsmooth optimization.

The individual methods from the above classes can be chosen by using additional specifications. The most important ones concerning direction determination and stepsize selection, are type of the method, kind of the matrix decomposition and number of the method. The type of the method is specified by the macrovariable $TYPE:

$TYPE='L' - line search methods.
$TYPE='G' - general trust region methods .
$TYPE='T' - special trust region methods for nonlinear least squares problems.
$TYPE='M' - modified Marquardt methods for nonlinear least squares problems.
$TYPE='P' - pattern search method of Hooke and Jeeves.
$TYPE='S' - simplex method of Nelder and Mead.

The kind of the matrix decomposition is specified by the macrovariable $DECOMP:

$DECOMP='M' - the symmetric matrix is used as an input for the direction determination.
$DECOMP='G' - the LDL^T decomposition without permutations is used as an input for the direction determination. This decomposition is usually obtained by the Gill-Murray algorithm [46].
$DECOMP='S' - the LDL^T decomposition with permutations is used as an input for the direction determination. This decomposition is usually obtained by the Schnabel-Eskow algorithm [116].
$DECOMP='B' - the block LDL^T decomposition with permutations is used as an input for the direction determination. This decomposition is usually obtained by the Bunch-Parlett algorithm [14].
$DECOMP='I' - the inverse of a symmetric matrix is used as an input for the direction determination.
$DECOMP='R' - the $R^T R$ decomposition without permutation is used as an input for the direction determination. This decomposition is usually obtained by the recursive QR factorization [65].
$DECOMP='C' - the $R^T R$ decomposition with permutations is used as an input for the direction determination. This decomposition is usually obtained by an application of the rank revealing algoritm [17].
$DECOMP='A' - the rectangular matrix is used as an input for the direction determination.
$DECOMP='Q' - the QR decomposition of a rectangular matrix without permutations is used as an input for the direction determination. This decomposition is usually obtained by using the Householder reflection with the explicitly stored orthogonal matrix Q.
$DECOMP='E' - the general square matrix is used as an input for the direction determination in the case NA=NF (system of nonlinear equations).
$DECOMP='K' - the indefinite Karush-Kuhn-Tucker matrix is used as an input for the direction determination in the equality constrained case.

The macrovariable $DECOMP is also used for the selection of conjugate direction methods. In this case it does not concern the kind of matrix decomposition.

The serial number of the method is specified by the macrovariable $NUMBER. It determines an individual realization of the direction determination.

Additional information about specifications $TYPE, $DECOMP, $NUMBER is given in Section 3.19.

All options used for the method selection have default values which follows from the knowledge bases coded in the individual templates. Therefore, they need not be specified by the user. The possibilities we describe can be of service to users that are familiar with optimization methods.

Almost all optimization methods have different realizations for three different representations of the objective function. If $HESF='D', then dense variants for either unconstrained problems or box constrained problems or linearly constrained problems (with dense linear constraints specified by $JACC='D') can be used. If $HESF='S', then sparse variants for either unconstrained problems or box constrained problems or linearly constrained problems (with sparse linear constraints specified by $JACC='S') can be used. If $JACA='S' and $HESF='B', then partitioned variants for either unconstrained problems or box constrained problems can be used. Partitioned variants of optimization methods are usually less efficient due to more expensive matrix operations. Therefore, we recommend to prefer sparse variants against the partitioned ones.

3.1. Heuristic methods

Heuristic (or comparative) methods are specified by the statement $CLASS='HM'. These methods can be used only for small-size problems (with at most 10 variables). The main advantage of the heuristic methods is that they do not require continuity of the objectiver function.

The individual heuristic methods are specified by the macrovariable $TYPE:

$TYPE='P' - pattern search method of Hooke and Jeeves [59].
$TYPE='S' - simplex method of Nelder and Mead [100].

The default value is $TYPE='P'.

3.2. Conjugate direction methods

Conjugate direction methods are specified by the statement $CLASS='CD'. These methods are very efficient for large problems with computationally simple objective functions ($KCF=1 or $KCA=1). The main advantage of conjugate direction methods is that matrices are not used (implicitly $HESF='NO'). This fact highly decreases storage requirements.

The individual conjugate direction methods are specified by the macrovariable $DECOMP:

$DECOMP='C' - conjugate gradient methods. These methods are the simplest ones from all conjugate direction methods and they require the fewest storage requirements. However, they usually consume a greater number of function evaluations then other conjugate direction methods.

$DECOMP='V' - variable metric methods with limited storage. These methods allow us to prescribe storage requirements using the number of VM steps (the number of necessary used vectors is approximately two times greater than the number of VM steps). The number of VM steps is specified by the macrovariable $MF. Variable metric methods with limited storage usually consume fewer function evaluations then conjugate gradient methods.

$DECOMP='M' - inexact difference version of the modified Newton method [27]. This method is implemented either as the line search method or as the trust region method. It can be very efficient but, since it consumes a greater number of gradient evalutions, it can be slower then other conjugate direction methods, particularly if the objective function is more complicated ($KCF>1 or $KCA>1).

There are two families of conjugate gradient methods implemented in the UFO system:

$NUMBER=1 - basic conjugate gradient methods described in [74]. The individual methods are specified by using the macrovariables $MET, $MET1 and $MET2.

$NUMBER=2 - generalized conjugate gradient methods introduced in [61]. The individual methods are specified by using the macrovariable $MET1.

If $MET=0, then the steepest descent method is used. If $MET=1, then the Fletcher-Reeves method [38] is used. If $MET=2, then the Polak-Ribiere method [106] is used. If $MET=3, then the Hestenes-Stiefel method [57] is used. The macrovariable $MET1 specifies the restart procedure as it is described in [74]. If $MET1=1, then a restarted CG method with positive parameter is used. If $MET1=2, then a bounded CG method with positive parameter is used. If $MET1=3, then a bounded CG method with positive lower bound is used. If $MET1=4, then a CG method with the Powell restart is used. If $MET1=5, then a CG method with the test on conjugacy is used. If $MET1=6, then a CG method with the test on orthogonality is used. The macrovariable $MET2 specifies the scaling parameter as it is described in [74] ($MET2=1 for suppressed scaling and $MET2=2 for scaling in each iteraton).

Similarly, the UFO system contains two variable metric methods with limited storage:

$NUMBER=1 - The BFGS method with limited storage described in [101]. The default number of VM steps is $MF=5.

$NUMBER=2 - The extended BFGS method with limited storage described in [62]. The default number of VM steps is $MF=3.

Both these methods are realized by using various scaling techniques [69], specified by the macrovariable $MET1. If $MET1=1, then scaling is suppressed. If $MET1=2, then scalar scaling is used. If $MET1=3, then diagonal scaling is used. If $MET1=4, then scalar and diagonal scalings are used simultaneously.

The possible specifications (type-decomp-number) for the conjugate direction methods in the unconstrained case are:

L-C-1, L-V-1,
L-C-2, L-V-2,
L-M-3,
G-M-3.

The default choice is L-C-1. In both the box constrained and the linearly constrained cases we cannot use specifications with $DECOMP='M'. Conjugate direction methods can be used also for sparse linear constraints when $JACC='S'.

3.3. Variable metric methods

Variable metric methods are specified by the statement $CLASS='VM'. These methods are most commonly used for either unconstrained or linearly constrained optimizations. Variable metric methods use a symmetric (usually positive definite) matrix which is updated in every iteration in such a way that it approximates the Hessian matrix of the objective function as precisely as possible. In the UFO system, the variable metric methods are realized in three different forms (for $HESF='D', $HESF='S' and $HESF='B') depending on the Hessian matrix specification.

There are two families of variable metric methods for dense problems ($HESF='D') which are distinguished using the macrovariable $UPDATE:

$UPDATE='B' - the Broyden family [11]. Variable metric methods from this family are the most commonly used ones since they are very robust and efficient.

$UPDATE='D' - the Davidon family [24]. Variable metric methods from this family are similar to the previous ones. The only difference is that projections into the new subspace are computed. This guarantees the quadratic termination property even in the case of an imperfect line search.

The default value is $UPDATE='B'.

Individual variable metric methods are specified by using the macrovariables $MET, $MET1, and $MET2. The macrovariable $MET determines the variable metric update. If $MET=1, then the BFGS method [11], [33], [49], [118] is used. If $MET=2, then the DFP method [23], [37] is used. If $MET=3, then the Hoshino method [60] is used. If $MET=4, then the safeguarded rank-one method [73] is used. If $MET=5, then the optimally conditioned method [24] is used. If $MET=6, then the rank-one based method [73] from the preconvex part of the Broyden family is used. If $MET=7, then the variationally derived method [76] from the preconvex part of the Broyden family is used. If $MET=8, then the heuristic method [79] is used. If $MET=9, then the method [139] derived from the matrix decomposition is used. If $MET=10, then the method [140] which minimizes the angle between the direction vector and the negative gradient is used. If $MET=11, then the method [79] which minimizes the norm of the direction vector is used. If $MET=12, then the least prior deviation method [95] is used. The default value is $MET=1. If we specify $DECOMP='M', then we can use only the values $MET=1,2,3,4.

The macrovariable $MET1 determines the Oren (scaling) parameter [103]. If $MET1=1, then no scaling is used. If $MET1=2, then initial scaling [119] is used. If $MET1=3, then controlled scaling [76] is used. If $MET1=4, then simple controlled scaling [84] is used. If $MET1=5, then scaling in each

iteration is used. The default value is $MET1=3. The scaling parameter is determined by using heuristic rules given in [79].

The macrovariable $MET2 determines a value of the Biggs (nonquadratic model) parameter [4]. If $MET2=1, then the unit value is used. If $MET2=2, then the Spedicato value [120] is used. If $MET2=3, then the modified Spedicato value [79] is used. If $MET2=4, then the value determined from the homogeneous model [79] is used. If $MET2=5, then the value determined from the cubic model [5] is used. The default value is $MET2=2.

The macrovariable $MET3 determines the Powell correction [110]. If $MET3=1 then the Powell correction is suppressed. If $MET3=2 then the Powell correction is applied.

The possible specifications (type-decomposition-number) for dense variable metric methods in the unconstrained case are:

L-G-1,	L-S-1,	L-B-1,	L-I-1,	L-M-1,
				L-M-3,
G-G-1,	G-S-1,	G-B-1,		G-M-1,
G-G-2,	G-S-2,	G-B-2,		G-M-2,
				G-M-3,
				G-M-4,
				G-M-5,
				G-M-7.

The default choice is L-I-1. In both the box constrained and the linearly constrained cases we cannot use specifications with $DECOMP='B'.

If the Hessian matrix is sparse with a general pattern ($HESF='S'), then the sparse variable metric methods, that preserve this pattern, are used. The individual variable metric updates (or families) are specified by using the macrovariable $UPDATE:

$UPDATE='M' - the simple Marwill projection [91]. This update can be used only if $DECOMP='M'.

$UPDATE='G' - the fractioned Marwill projection [134]. This update can be used only if $DECOMP='M' and $NUMBER=3.

$UPDATE='T' - the fractioned Toint projection (the best method given in [134]). This update can be used only if $DECOMP='M' and $NUMBER=3.

$UPDATE='B' - the partitioned variable metric updates from the Broyden family [53]. These updates can be used only if $MODEL='AF' or $MODEL='AQ' or $MODEL='AP'.

The default value is $UPDATE='M'.

Fractioned updates with the specifications $UPDATE='G' or $UPDATE='T' can be used only in the unconstrained case. If $UPDATE='B', then the particular update is specified by using the macrovariable $MET. If $MET=1, then the BFGS method is used. If $MET=2, then the DFP method is used. If $MET=3, then the Hoshino method is used. If $MET=4, then the safeguarded rank-one method is used. The default value is $MET=1.

If $DECOMP='G', then less efficient sparse product form updates from the Broyden family are used. In this case, the values $MET=1,2,3 can be used.

The possible specifications (type-decomposition-number) for sparse variable metric methods in the unconstrained case are:

L-G-1,	L-M-1,
	L-M-3,
G-G-1,	G-M-1,
	G-M-2,
	G-M-3,
	G-M-4,
	G-M-5,
	G-M-7.

The default choice is L-M-3. In both the box constrained and the linearly constrained cases we can use only specifications with $DECOMP='M' and $NUMBER=3. Similarly, if the fractioned updates ($UPDATE='T' and $UPDATE='G') are required, then only specifications with $DECOMP='M' and $NUMBER=3 can be used.

If the Hessian matrix is sparse with a partitioned pattern ($HESF='B'), then only the partitioned variable metric updates, specified by the choice $UPDATE='B', can be used. These updates are the same as in the case in which the Hessian matrix is sparse with a general pattern, but the partitioned realization is usually less efficient than the general one due to more expensive matrix operations.

The possible specifications (type-decomposition-number) for partitioned variable metric methods in the unconstrained case are:

L-M-3,
G-M-3.

The default choice is L-M-3.

3.4. Modified Newton methods

Modified Newton methods are specified by the statement $CLASS='MN'. These methods use the Hessian matrix of the objective function which is computed either analytically or numerically. The UFO system performs a numerical computation of the Hessian matrix automatically whenever the macrovariable $HMODELF (or $FGHMODELF) is not defined. Modified Newton methods are realized in three different forms (for $HESF='D', $HESF='S' and $HESF='B') depending on the Hessian matrix specification. Even if the modified Newton methods can be realized as the line search methods ($TYPE='L'), it is more advantageous to realize them as the trust region methods ($TYPE='G').

If the Hessian matrix is dense ($HESF='D'), then all second derivatives have to be given analytically or they are computed numericaly by using differences of gradients. The possible specifications (type-decomposition-number) for dense modified Newton methods in the unconstrained case are:

L-G-1,	L-S-1,	L-B-1,	L-M-1,
L-G-2,	L-S-2,	L-B-2,	L-M-2,
			L-M-3,
G-G-1,	G-S-1,	G-B-1,	G-M-1,
G-G-2,	G-S-2,	G-B-2,	G-M-2,
			G-M-3,
			G-M-4,
			G-M-5,
			G-M-7.

The default choice is G-M-7. In both the box constrained and the linearly constrained cases we cannot use specifications with $DECOMP='S' and $DECOMP='B'. The choice L-G-1 differs from the choice L-G-2. The last one corresponds to the combination of both the Newton and the conjugate gradient methods.

If the Hessian matrix is sparse with a general pattern ($HESF='S'), we have two possibilities. If $MODEL='FF', then only the structurally nonzero second order derivatives have to be given analytically by using the prescribed pattern. Numerical computation of the second derivatives is based on the fact that a substantially lower number of differences has to be used in comparison with the dense case. The determination of suitable differences is a combinatorial problem equivalent to some graph coloring problem [18], [19]. If $MODEL='AF' or $MODEL='AQ' or $MODEL='AP', then only the nonzero second derivatives of the approximating functions have to be given analytically by using the prescribed pattern. Numerical computation of the second derivatives is based on the fact that the approximating functions depend on a minor number of variables so that the number of differences is substantially lower in comparison to the dense case.

If $MODEL='AQ' (sum of squares), then the combination [82] of both the modified Newton and the modified Gauss-Newton methods can be used. This choice is possible by using the macrovariable $MET. If $MET=1, then the modified Newton method is used. If $MET=2, then the combined method is used. The default value is $MET=2.

The possible specifications (type-decomposition-number) for sparse modified Newton methods in the unconstrained case are:

L-G-1, L-M-1,
L-M-3,
G-G-1, G-M-1,
G-M-2,
G-M-3,
G-M-4,
G-M-5,
G-M-7.

The default choice is G-M-3. In the box constrained case we can only use specifications with $DECOMP='M' and $NUMBER=3. The choice L-M-1 differs from the choice L-M-2. The last one corresponds to the incomplete Gill-Murray decomposition.

If the Hessian matrix is sparse with a partitioned pattern ($HESF='B'), then a computation of the second order derivatives is the same as in the case when the Hessian matrix is sparse with a general pattern, but the partitioned realization is usually less efficient than the general one due to more expensive matrix operations.

If $MODEL='AQ' (sum of squares), then the combination of both the modified Newton and the modified Gauss-Newton methods can be used. This choice is possible by using the macrovariable $MET like the dense case. The possible specifications (type-decomposition-number) for partitioned modified Newton methods in the unconstrained case are:

L-M-3,
G-M-3.

The default choice is G-M-3.

3.5. Modified Gauss-Newton methods for nonlinear least squares and nonlinear equations

Modified Gauss-Newton methods are specified by the statement $CLASS='GN'. These methods are special optimization methods for either nonlinear least squares ($MODEL='AQ') or nonlinear least powers ($MODEL='AP') problems. Modified Gauss-Newton methods are based on the fact that the first term in the Hessian matrix expression, the so-called normal equation matrix, depending on the first derivatives of the approximating functions only, is a good approximation of the whole Hessian matrix. The second term in the Hessian matrix expression can be approximated by using the variable metric updates.

Modified Gauss-Newton methods are realized in four different forms (for $HESF='D', $HESF='S', $HESF='B', $HESF='NO') depending on the Hessian matrix specification. Even if the modified Gauss-Newton methods can be realized as the line search methods ($TYPE='L'), it is more advantageous to realize them as the trust region methods ($TYPE='G').

If the Hessian matrix is specified to be dense ($HESF='D'), then the normal equation matrix is also dense. In this case, we can use hybrid methods with dense updates:

$UPDATE='NO' - no update is used. The method utilizes the normal equation matrix (the first part of the Hessian matrix expression).

$UPDATE='S' - the Dennis structured approach [28] is used. The second part of the Hessian matrix is approximated by using modified variable metric updates. This part is added to the normal equation matrix if the conditions for leaving the modified Gauss-Newton method are satisfied.

$UPDATE='F' - the Fletcher hybrid approach [3], [39] is used. The Hessian matrix is approximated either by the normal equation matrix or by the matrix obtained by using the variable metric updates. The decision between the two cases is based on the rate of function value decrease and on the normal equation matrix conditioning.

$UPDATE='B' - a variable metric update from the Broyden class is applied either to the normal equation matrix or to the previous approximation of the Hessian matrix if conditions for leaving the modified Gauss-Newton method are satisfied [82].

The default value is $UPDATE='NO'.

Individual variable metric updates from the above families are specified by using the macrovariable $MET. If $MET=1, then the BFGS method is used. If $MET=2, then the DFP method is used. If $MET=3, then the Hoshino method is used. If $MET=4, then the original (unsafeguarded) rank-one method is used. The value $MET=4 is allowed only if $UPDATE='S' and it is the default in this case. The value $MET=1 is the default in the other cases.

Variable metric updates ($UPDATE=F or $UPDATE='B') can be realized either as simple updates (normal equation matrix is updated) or as cumulative updates (previous approximation of the Hessian matrix is updated), as it is described in [82]. Decision between these possibilities is mediated by the macrovariable $MOT1. If $MOT1=0, then the cumulative update is used. If $MOT1=1, then the simple update is used.

In the dense case, the modified Gauss-Newton methods can be realized with additional special matrix decompositions that cannot be used in other cases. If $DECOMP='R', then the recursive QR decomposition [106] is used with an additional correction of the upper triangular matrix R. If $DECOMP='C' then, moreover, the upper triangular matrix R is changed by using the rank revealing algorithm [17] that can improve its conditioning. The possible specifications (type-decomposition-number) for dense modified Gauss-Newton methods in the unconstrained case are:

L-G-1 ,	L-S-1,	L-B-1,	L-R-1,	L-C-1,	L-M-1,
					L-M-3,
G-G-1,	G-S-1,	G-B-1,	G-R-1,	G-C-1,	G-M-1,
G-G-2,	G-S-2,	G-B-2,	G-R-2,	G-C-2,	G-M-2,
					G-M-3,
					G-M-4,
					G-M-5,
					G-M-7,
T-G-1,	T-S-1,		T-R-1,	T-C-1,	T-M-1,
T-G-2,					
	T-S-7,			T-7-5,	T-M-7,
					M-M-1.

The default choice is G-M-7. In both the box constrained and the linearly constrained cases we cannot use specifications $DECOMP='S', $DECOMP='R', $DECOMP='C'. If $DECOMP='S' or $DECOMP='C', then variable metric updates cannot be used ($UPDATE='NO'). The specification $UPDATE='S' can be used only if $DECOMP='M'.

If the Hessian matrix is specified to be sparse with a general pattern ($HESF='S'), then the normal equation matrix has the same structure. In this case, we can use hybrid methods with sparse updates:

$UPDATE='NO' - no update is used. The method utilizes the normal equation matrix (the first part of the Hessian matrix expression).

$UPDATE='S' - the Dennis structured approach [28] is used. The second part of the Hessian matrix is approximated by using modified variable metric updates. This part is added to the normal equation matrix if conditions for leaving the modified Gauss-Newton method are satisfied.

$UPDATE='D' - the Brown-Dennis structured approach [13] is used. The Hessian matrices of approximating functions are approximated by using variable metric updates. These matrices serve for approximating the second part of the Hessian matrix which is added to the normal equation matrix if conditions for leaving the modified Gauss-Newton method are satisfied.

$UPDATE='B' - a variable metric update from the Broyden class is applied either to the normal equation matrix or to the previous approximation of the Hessian matrix if conditions for leaving the modified Gauss-Newton method are satisfied [82].

$UPDATE='M' - a sparse update based on the Marwill projection is applied either to the normal equation matrix or to the previous approximation of the Hessian matrix if conditions for leaving the modified Gauss-Newton method are satisfied [82].

The default value is $UPDATE='NO'.

Individual variable metric updates from the above families are specified by using the macrovariable $MET like the dense case. The value $MET=4 is allowed only if either $UPDATE='S' or $UPDATE='D' and it is the default in this case. The value $MET=1 is the default in the other cases excepting the case $UPDATE='M' in which the macrovariable $MET is not utilized.

Variable metric updates ($UPDATE=M or $UPDATE='B') can be realized either as simple updates (normal equation matrix is updated) or as cumulative updates (previous approximation of the Hessian matrix is updated). Decision between these possibilities is mediated by the macrovariable $MOT1 similarly as in the dense case.

If $UPDATE='D', then we can use several switches for utilizing variable metric updates specified by the macrovariable $MOT2. If $MOT2=0, then the Fletcher and Xu switch [39] is used. If $MOT2=1, then a modification of the Fletcher and Xu switch is used. If $MOT2=2, then the Denis and Welsch switch [31] is used. If $MOT2=3, then the Ramsin and Wedin switch [112] is used. The default value is $MOT2=0.

The possible specifications (type-decomposition-number) for sparse Gauss-Newton methods in the unconstrained case are:

L-G-1,	L-M-1,
	L-M-3,
G-G-1,	G-M-1,
G-G-2,	G-M-2,
	G-M-3,
	G-M-4,
	G-M-5,
	G-M-7,
T-G-1,	T-M-1,
	T-M-7,
	M-M-1.

The default choice is G-M-3. In the box constrained case we can use only specifications with $DECOMP='M' and $NUMBER=3.

If the Hessian matrix is specified to be sparse with a partitioned pattern ($HESF='B') then the normal equation matrix has the same structure. If that is the case, then we can use hybrid methods with partitioned updates $UPDATE='NO', $UPDATE='S', $UPDATE='D', $UPDATE='F', $UPDATE='B', whose details were already explained above. Note that the partitioned realization is usually less efficient than the general one due to more expensive matrix operations.

The possible specifications (type-decomposition-number) for partitioned Gauss-Newton methods are:

L-M-3,
G-M-3.

The default choice is G-M-3.

If the Hessian matrix is not specified ($HESF='NO'), then the normal equation matrix is not used. Instead of that the Jacobian matrix, defining a linear least squares problem, is utilized in each iteration. Such, so-called, normal equation free, Gauss-Newton methods are realized in two different forms (for $JACA='D' and $JACA='S') depending on the Jacobian matrix specification.

If the Jacobian matrix is specified to be dense ($JACA='D'), then we cannot use hybrid methods with variable metric updates (only the specification $UPDATE='NO is permitted). Moreover, dense, normal equation free, Gauss-Newton methods can be used only in the unconstrained case.

The possible specifications (type-decomposition-number) for dense, normal equation free, Gauss-Newton methods are:

L-Q-1,	L-A-1,	L-E-1,
	L-A-3,	L-E-3,
	L-A-4,	L-E-4,
		L-E-5,
G-Q-1,	G-A-1,	G-E-1,
G-Q-2,		G-E-2,
	G-A-3,	G-E-3,
	G-A-4,	G-E-4,
		G-E-5,
	G-A-7.	

The default choice is G-A-3 for least squares problems and G-E-3 for systems of nonlinear equations. The specification $DECOMP='E' can be used only if NA=NF (system of nonlinear equations).

If the Jacobian matrix is specified to be sparse ($JACA='S'), then we can use hybrid methods with simple variable metric updates:

$UPDATE='NO' - no update is used. The method utilizes original Jacobian matrix.

$UPDATE='V' - the simple factorized BFGS update [82] is used. The second order information is approximated by the unsymmetric rank-one update of the Jacobian matrix.

$UPDATE='R' - the simple factorized rank-one update [82] is used. The second order information is approximated by the addition of a dense row to the Jacobian matrix.

If $UPDATE='V' or $UPDATE='R', then we can use several switches for utilizing variable metric updates, specified by the macrovariable $MOT2, like the case with the specification $HESF='S' described above. The default value is $MOT2=0.

The main advantage of sparse, normal equation free, Gauss-Newton methods consists in the fact that the normal equation matrix is dense if the sparse Jacobian matrix has at least one dense row. If this is the case, then the classical Gauss-Newton methods cannot be used. On the other hand, the normal equation matrix has often a lower number of nonzero elements then the Jacobian one. As a result, the classical Gauss-Newton methods are more efficient in this case.

The possible specifications (type-decomposition-number) for sparse, normal equation free, Gauss-Newton methods are:

L-A-1,	L-E-1,
L-A-3,	L-E-3,
L-A-4,	L-E-4,
	L-E-5,
G-A-1,	G-E-1,
	G-E-2,
G-A-3,	G-E-3,
G-A-4,	G-E-4,
	G-E-5,
G-A-7.	

The default choice is G-A-3 for least squares problems and G-E-3 for systems of nonlinear equations. The specification $DECOMP='E' can be used only if NA=NF (system of nonlinear equations). In the box constrained case we can use only specifications with either $NUMBER=3 or $NUMBER=4. The choice L-E-1 differs from the choice L-E-2. The last one corresponds to the incomplete LU decomposition.

3.6. Quasi-Newton methods for nonlinear least squares and nonlinear equations

Quasi-Newton methods are specified by the statement $CLASS='QN'. These methods are special optimization methods for nonlinear least squares ($MODEL='AQ') problems including systems of nonlinear equations in the case when the first derivatives are not specified analytically (the macrovariable $GMODELA is not defined). Quasi-Newtod methods use a rectangular matrix which is updated in every iteration in such a way that it aproximates the Jacobian matrix as precisely as possible. In the UFO system, the quasi-Newton methods are realized in two different forms (for $JACA='D' and $JACA='S') depending on the Jacobian matrix specification.

There are two possibilities for dense problems ($JACA='D') which are distinguished by using the macrovariable $UPDATE:

$UPDATE='NO' - no update is used. Every approximation of the Jacobian matrix is computed numerically by using differences.
$UPDATE='B' - the Broyden family [12] of rank-one updates is used in almost all iterations. Only after restart the Jacobian matrix is approximated numerically by using differences.

When $UPDATE='B', then the individual quasi-Newton methods are specified by using the macrovariable $MET. If $MET=1, then the first (good) Broyden update [12] is used. If $MET=2, then the second Broyden update [12] is used. If $MET=3, then the second Greenstadt update [122] is used. If $MET=4, then the first Greenstadt update [122] is used. If $MET=5, then the first Todd OC update [63] is used. If $MET=6, then the first Todd OCX update [63] is used. If $MET=7, then the second Todd OC update [63] is used. If $MET=8, then the second Todd OCX update [63] is used. The default value is $MET=1. Dense quasi-Newton methods can be used only in the unconstrained case.

The possible specifications (type-decomposition-number) for dense quasi-Newton methods are:

L-Q-1,	L-A-1,	L-E-1,
	L-A-3,	L-E-3,
	L-A-4,	L-E-4,
		L-E-5,
G-Q-1,	G-A-1,	G-E-1,
G-Q-2,		G-E-2,
	G-A-3,	G-E-3,
	G-A-4,	G-E-4,
		G-E-5,
	G-A-7.	

The default choice is G-Q-3. The specification $DECOMP='E' can be used only if NA=NF (system of nonlinear equations).

If the Jacobian matrix is sparse with a general pattern ($JACA='S'), then there are two possibilities for computing an approximation of the Jacobian matrix by the differences. These possibilities are distinguished by using the macrovariable $NUMDER:

$NUMDER=1 - derivatives of individual approximating functions are computed.
$NUMDER=2 - the Coleman-More [20] graph coloring algorithm is used.

Moreover, various sparse quasi-Newton updates that preserve pattern of the Jacobian matrix can be used.

If $NUMDER=1, then there are three choices of the quasi-Newton updates which are specified by the macrovariable $UPDATE:

$UPDATE='NO' - no update is used. Every approximation of the Jacobian matrix is computed numerically by using differences.

$UPDATE='B' - sparse quasi-Newton updates are used in almost all iterations. Only after restart, the Jacobian matrix is approximated numerically by using differences.

$UPDATE='S' - modified Newton methods such as row scaling update are used in almost all iterations. Only after restart the Jacobian matrix is approximated numerically by using differences.

If $NUMDER=2, then there are four choices of the quasi-Newton updates which are specified by the macrovariable $UPDATE:

$UPDATE='NO' - no update is used. Every approximation of the Jacobian matrix is computed numerically by using differences.

$UPDATE='B' - sparse quasi-Newton updates [117] are used in almost all iterations. Only after restart the Jacobian matrix is approximated numerically by using differences.

$UPDATE='S' - modified Newton methods such as row scaling update are used in almost all iterations. Only after restart the Jacobian matrix is approximated numerically by using differences.

$UPDATE='C' - cyclic column determination methods are used in almost all iterations. Only after restart the Jacobian matrix is approximated numerically by using differences.

When $UPDATE='B', then the individual quasi-Newton methods are specified by using the macrovariable $MET. If $MET=1, then the Schubert update [117] is used. If $MET=2, then the Bogle-Perkins update [10] is used. If $MET=3, then the column update [92] is used. When $UPDATE='S' and $MET=0 then the modified Newton method is used. When $UPDATE='S' and $MET=1 then the row scaling update [92] is used. When $UPDATE='C' and $MET=0 then the cyclic column determination method [67] is used. When $UPDATE='S' and $MET=1 then the cyclic column determination method [67] followed by the Schubert update [117] is used.

The possible specifications (type-decomposition-number) for sparse quasi-Newton methods are:

L-A-1,	L-E-1,
L-A-3,	L-E-3,
L-A-4,	L-E-4,
	L-E-5,
G-A-1,	G-E-1,
	G-E-2,
G-A-3,	G-E-3,
G-A-4,	G-E-4,
	G-E-5,
G-A-7.	

The default choice is G-A-3 for least squares problems and G-E-3 for systems of nonlinear equations. The specification $DECOMP='E' can be used only if NA=NF (system of nonlinear equations). In the box constrained case we can use only specifications with either $NUMBER=3 or $NUMBER=4. The choice L-E-1 differs from the choice L-E-2. The later one corresponds to the incomplete LU decomposition.

3.7. Quasi-Newton methods with limited storage for nonlinear equations

Quasi-Newton methods with limited storage are specified by the statement $CLASS='QL'. The number of QN steps is specified by the macrovariable $MF. These methods are special methods for solving sparse systems of nonlinear equations ($MODEL='AQ') in a case in which the first derivatives are not specified analytically (the macrovariable $GMODELA is not defined). Therefore, only the case NA=NF is permitted. Quasi-Newtod methods with limited storage use an initial approximation of the sparse Jacobian matrix together with several small-size matrices which are updated in every iteration in such a

way that they aproximate the Jacobian matrix as precisely as possible. There are two possibilities which are distinguished by using the macrovariable $UPDATE:

$UPDATE='NO' - no update is used. Every approximation of the Jacobian matrix is computed numerically by using differences.

$UPDATE='B' - the Broyden good update of rank-one with limited storage [16] is used in almost all iterations. Only after restart the Jacobian matrix is approximated numerically by using differences.

The possible specifications (type-decomposition-number) for quasi-Newton methods with limited storage are:

L-A-3,	L-E-3,
L-A-4,	L-E-4,
	L-E-5,
G-A-3,	G-E-3,
G-A-4,	G-E-4,
	G-E-5,

The default choice is G-E-3.

Besides the quasi-Newtod methods with limited storage, this class contains inverse column scaling methods which are chosen by using the specification $DECOMP='I'. There are two possibilities which are distinguished by using the macrovariable $UPDATE:

$UPDATE='NO' - no update is used. Every approximation of the Jacobian matrix is computed numerically by using differences.

$UPDATE='B' - the inverse column scaling update [93] is used in almost all iterations. Only after restart the Jacobian matrix is approximated numerically by using differences.

The possible specifications (type-decomposition-number) for inverse column scaling methods are:

L-I-1,
L-I-3.

If $NUMBER=1, then a complete LU decomposition is used. If $NUMBER=3, then a combination of direct and iterative methods is used. The default choice is L-I-3.

3.8. Biconjugate direction methods for nonlinear equations

Biconjugate direction methods are specified by the statement $CLASS='BD'. These methods are special methods for solving systems of nonlinear equations ($MODEL='AQ') in the case when the first derivatives are not specified analytically (the macrovariable $GMODELA is not defined). Therefore only the case NA=NF is permitted. Biconjugate direction methods are very efficient for large problems with computationally simple functions in nonlinear equations ($KCA=1). The main advantage of biconjugate direction methods is that matrices are not used. This fact highly decreases storage requirements.

The individual biconjugate direction methods are specified by the macrovariable $DECOMP:

$DECOMP='E' - inexact difference version of the Newton method for systems of nonlinear equations [80]. This method is implemented either as the line search method or as the trust region method and it is based on smoothed CGS algorithm.

Iterative methods for solving linearized equations can be modified by using tridiagonal decomposition. This possibility is determined by the macrovariable $MOS2. If $MOS2=0, then tridiagonal decomposition is not used. If $MOS2=1, then tridiagonal decomposition is used before the iterative process. If

$MOS2=2, then tridiagonal decomposition is used as a preconditioner. If $MOS2=3, then both previous cases are assumed. The default value is $MOS2=0.

The possible specifications (type-decomposition-number) for the biconjugate direction methods are:

L-E-3,
L-E-4,
L-E-5,
G-E-3,
G-E-4,
G-E-5.

The default choice is G-E-3.

3.9. Modified Brent method for nonlinear equations

The Brent method is specified by the statement $CLASS='BR'. This method is a special method for solving dense systems of nonlinear equations ($MODEL='AQ') in the case when the first derivatives are not specified analytically (the macrovariable $GMODELA is not defined). Therefore, only the case NA=NF is permitted. The Brent method does not need any additional specifications (macrovariables $TYPE, $DECOMP, $NUMBER are not used).

3.10. Simplex type methods for linear programming problems

Simplex type methods for linear programming problems are specified by the statement $CLASS='LP'. These methods are realized in two different forms (for $JACC='D' and $JACC='S') depending on the constraint Jacobian matrix specification.

If the constraint Jacobian matrix is dense ($JACC='D'), then we can use two different linear programming methods based on the active set strategy:

$NUMBER=1 - primal reduced gradient (null-space) method (like the method proposed in [45]), which is a special implementation of the steepest descent reduced gradient method.

$NUMBER=2 - primal projected gradient (range-space) method which is a special implementation of the steepest descent projected gradient method.

The possible specifications (type-number) for dense linear programming methods are L-1 and L-2. The default choice is L-1.

If the constraint Jacobian matrix is sparse ($JACC='S'), then we can use one linear programming method based on the simplex algorithm:

$NUMBER=1 - primal reduced gradient (null-space) method which is described in [133].

The possible specification (type-number) for sparse linear programming methods is L-1.

3.11. Interior point methods for sparse linear programming problems

Interior point methods for sparse linear programming problems are specified by using the statement $CLASS='LI'. These methods, based on an infeasible primal-dual predictor-corrector strategy, can be used only in the sparse case when $JACC='S'. Individual methods are chosen by using the macrovariable $MLP:

$MLP=1 - the first algorithm of Miao [94].

$MLP=1 - the second algorithm of Miao [94].

$MLP=3 - the Mizuno algorithm [96].

All these methods can be realized in three form depending on a way of solving linear generalized Karush-Kuhn-Tucker system:

$NUMBER=1 - direct solution based on the Gill-Murray decomposition applied to the Schur complement.

$NUMBER=2 - direct solution based on the Bunch-Parlett decomposition applied to the original Karush-Kuhn-Tucker system.

$NUMBER=3 - iterative solution based on the conjugate gradient method applied to the Schur complement.

The possible specifications (type-number) for interior point methods are L-1, L-2 and L-3. The default choice is L-1.

3.12. Simplex type methods for quadratic programming problems

Simplex type methods for quadratic programming problems are specified by using the statement $CLASS='QP'. These methods are realized in two different forms (for $JACC='D' and $JACC='S') depending on the constraint Jacobian matrix specification.

If the constraint Jacobian matrix is dense ($JACC='D'), then we can use three different quadratic programming methods based on the active set strategy:

$NUMBER=1 - primal reduced gradient (null-space) method (like the method proposed in [47]) which is a special implementation of the Newton reduced gradient method.

$NUMBER=2 - primal projected gradient (range-space) method (like the method proposed in [35]) which is a special implementation of the Newton projected gradient method.

$NUMBER=3 - dual projected gradient (range-space) method (like the method proposed in [50]).

The possible specifications (type-number) for dense quadratic programming methods are L-1, L-2, and L-3. The default choice is L-1.

If the constraint Jacobian matrix is sparse ($JACC='S'), then we can use one quadratic programming method based on the simplex algorithm:

$NUMBER=1 - primal reduced gradient (null-space) method which is described in [133].

The possible specification (type-number) for sparse linear programming methods is L-1.

3.13. Proximal bundle methods for nonsmooth optimization

Proximal bundle methods for nonsmooth optimization problems are specified by the statement $CLASS ='BM'. These methods use a solution of the special quadratic programming subproblem derived from the cutting plane approach. This subproblem is in fact the same as in recursive quadratic programming methods for minimax problems. Proximal bundle methods are realized only for unconstrained or linearly constrained dense problems ($JACA='D'). The special quadratic programming subproblem can be solved by using the following methods:

$NUMBER=1 - dual projected gradient (range-space) method proposed in [70].

$NUMBER=2 - primal projected gradient (range-space) method which is a special implementation of the Newton projected gradient method.

Proximal bundle methods are realized only as line search methods in two modifications, which are specified by the macrovariable $MEX. If $MEX=0, then a convex version is assumed. If $MEX=1, then a nonconvex version is assumed and we can define a measure of nonconvexity using the macrovariable $ETA5. The default value is $ETA5=0.25. The possible specifications (type-number) for bundle methods are L-1 and L-2. The default choice is L-1. There are implemented various methods for computing of the weight parameter which are chosen by using the macrovariables $MOS and $MES2. If $MOS=1

and $MES2=1, then weights are updated using curvature of the one-dimensional quadratic function. If $MOS=1 and $MES2=2, then weights are updated using minimum position estimate (suitable for polyhedral and nearly polyhedral functions). If $MOS=2, then weights are updated using the quasi-Newton condition. Proximal bundle methods are used whenever $KSF=3 or $KSA=3. They can be also used for minimax problems as it is shown in Section 3.14.

3.14. Bundle-Newton methods for nonsmooth optimization

Bundle-Newton methods for nonsmooth optimization problems are specified by the statement $CLASS ='BN'. These methods use a solution of the special quadratic programming subproblem derived from the cutting plane approach which contains second order information. This subproblem is in fact the same as in recursive quadratic programming methods for minimax problems. Bundle-Newton methods are realized only for unconstrained or linearly constrained dense problems ($JACA='D'). The special quadratic programming subproblem can be solved by using the following methods:

$NUMBER=1	- dual projected gradient (range-space) method proposed in [70].
$NUMBER=2	- primal projected gradient (range-space) method which is a special implementation of the Newton projected gradient method.

A nonconvex version is assumed and we can define a measure of nonconvexity using the macrovariable $ETA5. The default value is $ETA5=0.25. The possible specifications (type-number) for bundle methods are L-1 and L-2. The default choice is L-1. Bundle-Newton methods can be used when $KSF=3 or $KSA=3. They can be also used for minimax problems as it is shown in Section 3.14.

3.15. Variable metric bundle methods for nonsmooth optimization

Variable metric bundle methods for nonsmooth optimization problems are specified by the statement $CLASS ='BV'. These methods are based on a special realization of the BFGS variable metric method. This realization uses special null steps and restarts. Stepsize selection is based on a polyhedral approximation obtained using bundles of points and subgradients. Variable metric bundle methods are realized only for unconstrained dense problems ($JACA='D'). Variable metric bundle methods can be used when $KSF=3 or $KSA=3. They can be also used for minimax problems as it is shown in Section 3.16.

3.16. Methods for minimax problems.

Minimax problems are specified by the choice $MODEL='AM'. These problems can be solved using four classes of methods:

$CLASS='BM'	- proximal bundle methods.
$CLASS='BN'	- bundle-Newton methods.
$CLASS='LP'	- recursive linear programming methods.
$CLASS='VM'	- recursive quadratic programming variable metric methods. An approximation of Lagrangian function Hessian matrix is updated in each iteration using the variable metric updates belonging to the Broyden family.
$CLASS='MN'	- recursive quadratic programming modified Newton methods. The Lagrangian function Hessian matrix is computed in each iteration either analytically or numerically.

the default value is $CLASS='VM'. Variable metric methods are the same as in Section 3.3 with the choice $DECOMP='G' and $UPDATE='B' (values $MET=1 - $MET=12 can be used). Similarly, modified Newton methods are the same as in Section 3.4 with the choice $DECOMP='G' (the Gill-Murray decomposition is used).

Even if minimax problems can be solved by using bundle methods described in Sections 3.13 - 3.15, it is more efficient to use recursive linear programming or recursive quadratic programming methods that utilize a special structure of minimax problems.

Recursive linear programming methods are realized as trust region methods with box constrained subproblems. The special linear programming subproblem, which is derived from the minimax problem, is solved by a primal projected gradient (range-space) method which is a special implementation of the steepest descent method.

Recursive quadratic programming methods are realized in three different forms:

$TYPE='L' - line search methods.
$TYPE='G' - general trust region methods .
$TYPE='C' - general trust region methods with second order corrections [40].

If $TYPE='L', then The special line search method ($MES=5) described in [71] can be used successfully.

The special quadratic programming subproblem, which is derived from the minimax problem, can be solved by using two different methods:

$NUMBER=1 - dual projected gradient (range-space) method proposed in [70].
$NUMBER=2 - primal projected gradient (range-space) method which is a special implementation of the Newton projected gradient method.

All of the above methods are realized only for dense unconstrained or linearly constrained problems. The possible specification (type-number) for recursive linear programming methods is G-1. The possible specifications (type-number) for recursive quadratic programming methods are:

L-1,
L-2,
G-1,
G-2,
C-1,
C-2.

The default choice is L-1.

3.17. Recursive quadratic programming methods for nonlinear programming problems.

Recursive quadratic programming methods for nonlinear programming problems are specified by the statement $FORM='SQ'. These methods belong to two following classes:

$CLASS='VM' - recursive quadratic programming variable metric methods. An approximation of Lagrangian function Hessian matrix is updated in each iteration using variable metric updates.
$CLASS='MN' - recursive quadratic programming modified Newton methods. The Lagrangian function Hessian matrix is computed in each iteration either analytically or numerically.

the default value is $CLASS='VM'. Variable metric methods are the same as in Section 3.3 with the choice $DECOMP='G' and $UPDATE='B' (values $MET=1 - $MET=12 can be used). Similarly, modified Newton methods are the same as in Section 3.4 with the choice $DECOMP='G' (the Gill-Murray decomposition is used).

Recursive quadratic programming methods for nonlinear programming problems are realized as line search methods ($TYPE='L') with the l_1-exact penalty function. They are like the methods proposed in [110]. The special line search method ($MES=5) for l_1-exact penalty function can be used successfully. The quadratic programming subproblem can be solved by using two different methods:

$NUMBER=1 - dual projected gradient (range-space) method (like the method proposed in [50]).
$NUMBER=2 - primal projected gradient (range-space) method (like the method proposed in [35]) which is a special implementation of the Newton projected gradient method.

Recursive quadratic programming methods are realized only for dense nonlinear programming problems. The possible specifications (type-number) for these methods are L-1 and L-2. The default choice is L-1.

3.18. Recursive minimax optimization methods for nonlinear programming problems.

Recursive minimax optimization methods for nonlinear programming problems are specified by the statement $FORM='SM'. These methods belong to two following classes:

$CLASS='VM' - recursive minimax optimization variable metric methods. An approximation of Lagrangian function Hessian matrix is updated in each iteration using variable metric updates.

$CLASS='MN' - recursive minimax optimization modified Newton methods. The Lagrangian function Hessian matrix is computed in each iteration either analytically or numerically.

the default value is $CLASS='VM'. Variable metric methods are the same as in Section 3.3 with the choice $DECOMP='G' and $UPDATE='B' (values $MET=1 - $MET=12 can be used). Similarly, modified Newton methods are the same as in Section 3.4 with the choice $DECOMP='G' (the Gill-Murray decomposition is used).

Recursive minimax optimization methods for nonlinear programming problems are based on the transformation of a nonlinear programming problem to a sequence of minimax problems with l_∞-exact penalty function (see [71]). These methods are realized as line search methods ($TYPE='L'). The special line search method ($MES=5) for l_∞-exact penalty function can be used successfully. The special quadratic programming subproblem, derived from the minimax formulation, can be solved by using two different methods:

$NUMBER=1 - dual projected gradient (range-space) method proposed in [70].

$NUMBER=2 - primal projected gradient (range-space) method which is a special implementation of the Newton projected gradient method.

Recursive quadratic programming methods are realized only for dense nonlinear programming problems. The possible specifications (type-number) for these methods are L-1 and L-2. The default choice is L-1.

3.19. Inexact recursive quadratic programming methods for large sparse equality constrained nonlinear programming problems.

Inexact recursive quadratic programming methods for equality constrained nonlinear programming problems are specified by the statement $FORM='SE'. These methods, which are designed for large sparse problems, belong to the following class:

$CLASS='MN' - inexact recursive quadratic programming modified Newton methods. The Lagrangian function Hessian matrix is computed in each iteration either analytically or numerically.

Inexact recursive quadratic programming methods for equality constrained nonlinear programming problems are based either on an inexact solution of the Karush-Kuhn-Tucker system [89] or on a decomposition of Lagrangian function Hessian matrix followed by an inexact solution of a range space system for the Lagrange multipliers [85]. The first approach, specified by the choice $DECOMP='K', is realized in three variants:

$NUMBER=1 - exact sparse Bunch-Parlett decomposition [32] of the indefinite Karush-Kuhn-Tucker system.

$NUMBER=3 - inexact smoothed conjugate gradient method for the indefinite Karush-Kuhn-Tucker system with a precision control based on various penalty functions.

$NUMBER=4 - inexact MINRES method for the indefinite Karush-Kuhn-Tucker system with a precision control based on various penalty functions.

A particular realization of both inexact smoothed conjugate gradient method and inexact MINRES method depends on specifications given by the macrovariables $MOS1, $MOS2, $MOS3. The macrovariable $MOS1 specifies a precision control and a choice of penalty parameter. If $MOS1=0, then a precision control is suppressed. If $MOS1=1, then a presision control, together with a basic choice of the penalty parameter, is used. If $MOS1=2, then a presision control, together with an extended choice of the penalty parameter, based on condition of positive definitness, is used. The macrovariable $MOS2 specifies a preconditioning technique (see section 3.21). The macrovariable $MOS3 specifies residual smoothing of the conjugate gradient method. If $MOS3=0, then a residual smoothing is suppressed. If $MOS3=1, then a one-dimensional residual smoothing is used.

The second approach, specified by the choice $DECOMP='G', is realized in two variants:

$NUMBER=3 - sparse Gill-Murray decomposition of the Lagrangian function Hessian matrix followed by the inexact smoothed conjugate gradient method for positive definite range space system with a precision control based on various penalty functions.

$NUMBER=4 - sparse Bunch-Parlett decomposition of the Lagrangian function Hessian matrix followed by inexact MINRES method for an indefinite range space system with a precision control based on various penalty functions.

Individual penalty functions are determined by using the macrovariable $MEP. If $MEP=1, then the l_1 exact penalty function is used. If $MEP=2, then the augmented Lagrangian function is used. If $MEP=3, then the combined l_1 and augmented Lagrangian function is used.

The UFO system allows us to choose a second order correction for overcoming the Maratos effect, various Lagrange multipliers updates and various forms of augmented Lagrangian function. This is affected by the macrovariables $MEP1, $MEP2, $MEP3. The macrovariable $MEP1 specifies a second order correction. If $MEP1=1, then the second order correction is suppressed. If $MEP1=2, then the second order correction is determined as being a least squares solution of the shifted constraint system. The macrovariable $MEP2 specifies estimates of Lagrange multipliers at the begining of each iteration. If $MEP2=1, then the initial estimate is taken from the previous iteration. If $MEP2=2, then the initial estimate is determined as being a least squares solution of the first part of the Karush-Kuhn-Tucker system. The macrovariable $MEP3 specifies penalty term of the augmented Lagrangian function. If $MEP3=1, then the basic penalty term is used. If $MEP3=2, then the extended Boggs-Tolle [9] penalty term is used.

The possible specifications (type-decomposition-number) for inexact recursive quadratic programming methods for equality constrained nonlinear programming problems are

L-K-1,
L-K-3, L-G-3,
L-K-4, L-G-4.

The default choice is L-K-3.

3.20. Methods for initial value problems for ordinary differential equations

Methods for initial value problems for ordinary differential equations are specified by using the macrovariable $SOLVER. The UFO system contains five types of integration methods:

$SOLVER='DP5' - the Dormand and Prince method of the fifth order with a stepsize control for nonstiff problems.

$SOLVER='DP8' - the Dormand and Prince method of the eighth order with a stepsize control for nonstiff problems.

$SOLVER='EX1' - the extrapolation method with a stepsize control, based on the midpoint rule, for nonstiff problems.
$SOLVER='RD5' - the Radau method of the fifth order with a stepsize control for stiff problems.
$SOLVER='RS4' - the Rosenbrock method of the fourth order with a stepsize control for stiff problems.

The default value is $SOLVER='DP8'. These methods, described in [55], use a stepsize control based on a local truncation error.

A solution to an initial value problem for ordinary differential equations can be stored for subsequent processing. An extent of stored data is determined by using the macrovariable $MED. If $MED=0, then no data are stored. If $MED=1, then data in all solution steps are stored. If $MED=2, then data in equidistant mesh points are stored. The number of mesh points is specified by using the statement $NA=number_of_mesh_points in the last case.

3.21. Methods for direction determination

Optimization methods, contained in the UFO system, are usually implemented in such a way that they use the same modules for direction determination. These modules, realized with different kinds of matrix decomposition, are distinguished by using the macrovariables $TYPE and $NUMBER. The meaning of the specification $TYPE was explained above. Now we will explain the specification $NUMBER.

If $TYPE='L', then line search methods are supposed. In this case, relatively simple procedures for direction determination are used. There are five possibilities:

$NUMBER=1 - direct methods for solving linear systems based on various matrix decompositions. These decompositions are interesting, especially in the sparse case. The Gill-Murray decomposition [46] of the Hessian matrix is applied if $DECOMP='M' and $MOS2=0 or if $DECOMP='G'. The Schnabel-Eskow decomposition [116] of the Hessian matrix is used if $DECOMP='M' and $MOS2=1 or if $DECOMP='S'. The Choleski decomposition of the Hessian matrix is utilized if $DECOMP='R' or $DECOMP='C'. The Bunch-Parlett decomposition [14] of the Hessian matrix is applied if $DECOMP='B'. The inverse matrix is used if $DECOMP='I'. The orthogonal QR decomposition [135] of the Jacobian matrix is utilized if $DECOMP='A' or $DECOMP='Q'. The complete LU decomposition [25] of the Jacobian matrix is applied if $DECOMP='E'. The Bunch-Parlett decomposition [32] of the sparse Karush-Kuhn-Tucker matrix is used if $DECOMP='K'. Moreover, symbolic decomposition is always determined before the iterative process in the sparse case, so that only numerical computations with known factors are carried out in the subsequent iterations.

$NUMBER=2 - an alternative possibility to the previous case. The direct solution is combined with a conjugate gradient direction if the Hessian matrix is indefinite. This possibility can be advantageously used in connection with the modified Newton method.

$NUMBER=3 - inexact iterative methods. The conjugate gradient method [27] for solving linear systems with the Hessian matrix is applied if $DECOMP='M'. The CGLS method [105] for solving linear least squares problems with the Jacobian matrix is used if $DECOMP='A'. The smoothed CGS method [131] for solving linear systems with the Jacobian matrix is utilized if $DECOMP='E'. The smoothed conjugate gradient method [89] for a linear system with the Karush-Kuhn-Tucker matrix is applied if $DECOMP='K'. The precision is specified by the macrovariable $MOS. If $MOS=1, then the simple strategy is used. If $MOS=2, then the geometric decreasing strategy is used. If $MOS=3, then the harmonic decreasing strategy is used. If $DECOMP='M' and $HESF='S', then the conjugate gradient method can be preconditioned by using the incomplete Gill-Murray (IGM) decomposition. This possibility is specified by the macrovariable $MOS2. If

$MOS2=0 then preconditioning is suppressed. If $MOS2=1, then IGM decomposition is used. Similarly, if $DECOMP='E' and $JACA='S', then the smoothed CGS method can be preconditioned by using either the incomplete LU (ILU) decomposition or the SSOR iteration. This possibility is specified by the macrovariable $MOS2. If $MOS2=0, then preconditioning is suppressed. If $MOS2=1, then ILU decomposition is used. If $MOS2=2, then SSOR iteration is used. Finally, if $DECOMP='K' then the smoothed conjugate gradient method can be preconditioned by using various preconditioners. This possibility is specified by the macrovariable $MOS2. If $MOS2=0 then preconditioning is suppressed. If ABS($MOS2)=1, then the block diagonal positive definite preconditioner [136] is used. If ABS($MOS2)=2, then the indefinite preconditioner [89] based on a diagonal approximation of the Schur complement is used. If ABS($MOS2)=3, then the indefinite preconditioner [89] based on a diagonal perturbation of the Schur complement is used. If ABS($MOS2)=4, then the indefinite preconditioner [89] based on a diagonal approximation of the Hessian matrix is used. In the later cases, a complete Gill-Murray decomposition is used if $MOS2 is negative and an incomplete Gill-Murray decomposition is used if $MOS2 is positive.

$NUMBER=4 - inexact iterative methods. The LSQR method [105] for solving linear least squares problems with the Jacobian matrix is applied if $DECOMP='A'. The GMRES method [115] for solving linear systems with the Jacobian matrix is used if $DECOMP='E'. The MINRES method for solving linear systems with the Karush-Kuhn-Tucker matrix is utilized if $DECOMP='K'. The precision is specified by the macrovariable $MOS as in the previous case.

$NUMBER=5 - inexact iterative methods. The smoothed BICGSTAB method [137] for solving linear systems with the sparse Jacobian matrix is used if $DECOMP='E'. The QMR method [41] for solving linear systems with the Karush-Kuhn-Tucker matrix is used if $DECOMP='K'. The precision is specified by the macrovariable $MOS as in the previous case.

If the line search method is used then a descent property of the determined direction is tested. If

$$-s^T g \geq \varepsilon_0 \parallel s \parallel\parallel g \parallel$$

where $s^T g$ is the directional derivative, s is the direction, and g is the objective function gradient, then the direction is accepted. In the opposite case the optimization method is restarted. The value ε_0 is specified using the macrovariable $EPS0.

If $TYPE='G', then trust region methods are supposed. The initial trust region radius can be specified by the statement $XDEL=trust_region_radius, but the default automatically derived value is recommended. Trust region methods can be internally scaled. This way is very advantageous for nonlinear regression problems containing exponentials. The trust region scaling is specified by the macrovariable $MOS1. If $MOS1=1, then no scaling is performed. If $MOS1=2, then the scaling coefficients are derived from the normal equation matrix diagonal elements [78]. There are six possibilities:

$NUMBER=1 - so-called single dog-leg methods based on various matrix decompositions. These decompositions are interesting especially in the sparse case. The Gill-Murray decomposition [46] of the Hessian matrix is applied if $DECOMP='M' and $MOS2=0 or if $DECOMP='G'. The Schnabel-Eskow decomposition [116] is used if $DECOMP='M' and $MOS2=1 or if $DECOMP='S'. The Choleski decomposition of the Hessian matrix is utilized if $DECOMP='R' or $DECOMP='C'. The Bunch-Parlett decomposition [14] of the Hessian matrix is applied if $DECOMP='B'. The orthogonal QR decomposition [135] of the Jacobian matrix is utilized if $DECOMP='A' or $DECOMP='Q'. The complete LU decomposition [25] of the Jacobian matrix is applied

if $DECOMP='E'. Moreover, symbolic decomposition is always determined before the iterative process in the sparse case, so that only numerical computations with known factors are carried out in the subsequent iterations The individual dog-leg methods are specified by the macrovariable $MOS. If $MOS=1, then the single dog-leg method [107] is used. If $MOS=2, then the double dog-leg method [29] is used. If $MOS=3, then the triple dog-leg method is used. If $MOS=4, then the optimum dog-leg method [15] is used.

$NUMBER=2 - an alternative possibility to the previous case. The so-called multiple dog-leg methods (combinations of single dog-leg methods and conjugate gradient methods) [81] are supposed. The number of dog-leg steps is specified by the statement $MOS=number_of_steps.

$NUMBER=3 - iterative trust region methods. The conjugate gradient trust region method [124] with the Hessian matrix is applied if $DECOMP='M'. The CGLS trust region method [77] with the Jacobian matrix is used if $DECOMP='A'. The smoothed CGS trust region method [87] with the Jacobian matrix is utilized if $DECOMP='E'. The precision is specified by the macrovariable $MOS. If $MOS=1, then the simple strategy is used. If $MOS=2, then the geometric decreasing strategy is used. If $MOS=3, then the harmonic decreasing strategy is used. If $DECOMP='M' and $HESF='S', then the conjugate gradient method can be preconditioned by using the incomplete Gill-Murray (IGM) decomposition. This possibility is specified by the macrovariable $MOS2. If $MOS2=0 then preconditioning is suppressed. If $MOS2=1, then IGM decomposition is used. Similarly, if $DECOMP='E' and $JACA='S', then the smoothed CGS method can be preconditioned by using either the incomplete LU (ILU) decomposition or the SSOR iteration. This possibility is specified by the macrovariable $MOS2. If $MOS2=0, then preconditioning is suppressed. If $MOS2=1, then ILU decomposition is used. If $MOS2=2, then SSOR iteration is used.

$NUMBER=4 - iterative trust region methods. The combined Lanczos and CG trust region method [81] with the Hessian matrix is applied if $DECOMP='M'. The LSQR trust region method [77] with the Jacobian matrix is used if $DECOMP='A'. The GMRES trust region method [87] with the Jacobian matrix is utilized if $DECOMP='E'. The precision is specified by the macrovariable $MOS as in the previous case. Iterative methods can be again preconditioned. This possibility is specified by the macrovariable $MOS2 as in the previous case.

$NUMBER=5 - iterative trust region methods. The combined CG and Lanczos trust region method [81] with the Hessian matrix is applied if $DECOMP='M'. The smoothed BICGSTAB trust region method [87] with the Jacobian matrix is utilized if $DECOMP='E'. The precision is specified by the macrovariable $MOS as in the previous case. Iterative methods can be again preconditioned. This possibility is specified by the macrovariable $MOS2 as in the previous case.

$NUMBER=7 - an optimum locally constrained trust region method [99]. The Gill-Murray decomposition [46] of the Hessian matrix is applied if $DECOMP='M' and $MOS2=0. The Schnabel-Eskow decomposition [116] of the Hessian matrix is used if $DECOMP='M' and $MOS2=1 or if $DECOMP='S'. The special augmented Jacobian matrix is used if $DECOMP='A'.

If $TYPE='T', then only the specifications $NUMBER=1, $NUMBER=2 and $NUMBER=7 can be used. These specifications have the same meaning as in the case $TYPE='G', but the implementation is simpler. If $NUMBER=7, then the simplified optimum locally constrained trust region method [78] is used.

If $TYPE='M' then Levenberg-Marquardt type methods are supposed:

$NUMBER=1 - a modified Marquardt method proposed by Fletcher [34].

3.22. Methods for stepsize selection

Stepsize selection is a very important part of optimization methods. The UFO system contains two types of stepsize selection procedures: line search methods and trust region methods. Line search methods are realized in two modifications specified by the macrovariable $SEARCH:

$SEARCH='B' - basic line search methods based on various interpolation and extrapolation formulas.
$SEARCH='M' - mixed line search methods which control the maximum stepsize like the trust region methods.

The choice of individual line search procedures is influenced by the order of directional derivatives being used. This order can be specified by the macrovariable $KDS. The value of the macrovariable $KDS is usually derived internally from the order of analytically supplied partial derivatives. If this order is zero, then always $KDS=0. In the opposite case, the value of the macrovariable $KDS can be specified by the user. If $KDS=0, then only the function values are used during the line search. If $KDS=1, then the function values and the first directional derivatives are used. If $KDS=2 then, in addition, the Hessian matrices or their approximations are computed during the line search (this case is very useful for a line search implementation of modified Gauss-Newton methods).

The particular interpolation and extrapolation rule is specified by the macrovariable $MES. If $KDS=0 then we have the following possibilities:

$MES=1 - The uniformly increasing extrapolation or bisection interpolation is used.
$MES=2 - Two point quadratic extrapolation or interpolation is used.
$MES=3 - Three point quadratic extrapolation or interpolation is used.
$MES=4 - Three point cubic extrapolation or interpolation is used.
$MES=5 - Special extrapolation or interpolation is used based on the special form of the objective function.

If $KDS=1 or $KDS=2, then the following possibilities, based on the first directional derivatives, can be used:

$MES=1 - the uniformly increasing extrapolation or bisection interpolation is used.
$MES=2 - quadratic extrapolation or interpolation (with one directional derivative) is used.
$MES=3 - quadratic extrapolation or interpolation (with two directional derivatives) is used.
$MES=4 - cubic extrapolation or interpolation [23] is used.
$MES=5 - conic extrapolation or interpolation [6] is used.

More detailed specifications concerning line search selection can be chosen using macrovariables $MES1, $MES2, $MES3:

$MES1=1 - constant extrapolation is used.
$MES1=2 - extrapolation specified by the macrovariable $MES is used.
$MES1=3 - extrapolation is suppressed.
$MES2=1 - standard line search termination criterion is used.
$MES2=2 - special termination criterion for nonconvex functions is used.
$MES2=3 - line search is terminated after at least two function evaluations.
$MES3=1 - safeguard against rounding errors is suppressed.
$MES3=2 - first level of safeguard is used.
$MES3=3 - second level of safeguard is used.

Another useful specification for the line search selection is a termination criterion, which is determined by using the macrovariable $KTERS:

$KTERS<0 - nonmonotone line search procedure proposed in [54] is used. The absolute value of the macrovariable $KTERS, which cannot be greater then 10, gives the number of nonmonotone steps.

$KTERS=1 - perfect stepsize. The relative precision of the stepsize parameter is given by the value $EPS3.

$KTERS=2 - the Goldstein stepsize [51]. The termination precision is given by the value $EPS1.

$KTERS=3 - the Curry-Altman stepsize [22] (Wolfe conditions). The termination precision is given by the values $EPS1 and $EPS2.

$KTRES=4 - the extended Curry-Altman stepsize [36] (strict Wolfe conditions). The termination precision is given by the values $EPS1 and $EPS2.

$KTERS=5 - the Armijo stepsize [2]. The termination is given by the value $EPS1.

$KTERS=6 - the first stepsize. The stepsize selection is terminated after the first function evaluation.

The last useful specification for the line search methods is an initial stepsize choice which is determined by the macrovariable $INITS. The initial stepsize is usually computed by the rule

$$\alpha \ = \min(c_1, -c_2(\Delta F / s^T g))$$

where $s^T g$ is the initial directional derivative and $\Delta F = F - F_{min}$ or $\Delta F = F_{old} - F$ if the value of the macrovariable $INITS is positive or negative, respectively. The absolute value of the macrovariable $INITS determines the coefficients c_1 and c_2 If |INITS|=1, then $c_1 = 1$ and $c_2 = 0$. If |INITS|=2, then $c_1 = 1$ and $c_2 = 4$. If |INITS|=3, then $c_1 = 1$ and $c_2 = 2$. If |INITS|=4, then $c_1 = 0$ and $c_2 = 2$.

Trust region methods are also realized in two modifications specified by the macrovariable $SEARCH:

$SEARCH='B' - basic trust region methods with stepsize control based on the comparison of both the actual and the predicted function decreases.

$SEARCH='M' - mixed trust region methods which use interpolation formulas for stepsize reduction like the line search methods [102].

Trust region methods are also influenced using the macrovariable $KTERS. If $KTERS<0, then nonmonotone trust region procedure proposed in [26] is used. The absolute value of the macrovariable $KTERS, which cannot be greater then 10, gives the number of nonmonotone steps.

3.23. Methods for numerical differentiation

The UFO system computes derivatives of the model function (of the approximating functions, of the constraint funcions) numerically whenever they are not given analytically. This is made possible by the macroprocessor that generates a corresponding part of the control program. The main problem of a numerical differentiation is a difference determination which has to be chosen in such a way that the total influence of both the cancellation and the roundoff error is as small as possible. There are three possibilities in the UFO system which are distinguished using the macrovatiable $MCG:

$MCG=0 - a simple difference determination described in [30] is used.

$MCG=1 - an optimum difference determination proposed in [48] is used.

$MCG=2 - an optimum difference determination proposed in [126] is used.

The default option is $MCG=2. The above possibilities are used for a computation of the model function first order derivatives. The other (second order derivatives or derivatives of the approximating functions and constraint functions) are always computed with the simple difference determination.

3.24. Methods for objective function evaluation in the case of dynamical systems optimization

If either $MODEL='DF' or $MODEL='DQ', then the objective function is computed from the solution of an initial value problem for ordinary differential equations. The initial value problem is solved and the integral criterion is evaluated by using integration methods specified by the macrovariable $SOLVER as it is described above. If the partial derivatives of all the used functions are given analytically, then the gradient of the objective function is computed by integration methods. There are two possibilities specified by the macrovariable $SYSTEM:

$SYSTEM='F' - forward integration using an augmented systen of ordinary differential equations.
$SYSTEM='B' - backward integration using the adjoint system of ordinary differential equations.

The default value is $SYSTEM='F'. In the case of modified Gauss-Newton methods ($CLASS='GN'), an approximation of the Hessian matrix is also computed by using forward integration of an augmented system.

3.25. Global optimization methods

Global optimization methods are used if $EXTREM='G' is specified. Global optimization methods use local optimization ones for finding local minima. Therefore the particular local optimization method has to be chosen by using the macrovariables $CLASS and $TYPE and others. Individual global optimization methods are specified by using the macrovariables $GCLASS and $GTYPE. The UFO system contains four classes of global optimization methods:

$GCLASS=1 - random search methods. These methods are simple and robust, but less efficient.
$GCLASS=2 - continuation methods. These methods use some penalty functions which are adjusted after reaching an arbitrary local minimum so that another local minimum is found.
$GCLASS=3 - clustering methods. These methods are based on randomly generated sample points which are processed using clustering algorithms to determine attractivity regions (clusters) of the individual minima. The attractivity regions (clusters) obtained are not searched repeatedly.
$GCLASS=4 - multi-level methods. Modern stochastic methods that involve a combination of sampling and local search techniques. These methods combine strong theoretical properties with an attractive computational behaviour. These methods are simpler, but more efficient than clustering methods.

If $GCLASS=1, then we can choose four types of global optimization methods:

$GTYPE=1 - singlestart methods. Random points, uniformly distributed in the given region, are generated and a local minimization method is started from the point with the lowest function value.
$GTYPE=2 - multistart methods. Random points, uniformly distributed in a given region, are generated and a local minimization is started from every point. Obtained local minima are compared and selected.
$GTYPE=3 - modified multistart methods. Random points, distributed in a given region uniformly, are generated and a local minimization is started whenever a point is found which has a lower function value than that reached up to date.
$GTYPE=3 - Bayesian reduced multistart methods [7]. Random samples of points are repeatedly generated. Every random sample is reduced and a local minimization is started from all points belonging to the reduced sample. Obtained local minima are compared and selected. This process is repeated while the Bayessian termination criterion is not satisfied.

If $GCLASS=2, then we can choose three types of global optimization methods:

$GTYPE=1 - tunneling function methods [66]. These methods consist of two phases: a local minimization phase and a tunneling phase. The starting point for the second phase is the local minimum. At the end of the tunneling phase a new point is found which has a function value equal or lower then the starting point.

$GTYPE=2 - combined tunneling function and random search methods. In this case a random search is used in the tunneling phase, if minimization of a tunneling function has failed to find a new starting point.

$GTYPE=3 - filled function methods [42], [43]. The idea of filled function methods is based on a filled function. This function has a maximum in the point of a known minimum of the objective function. On the other hand, this function does not have minimizers or saddle points in any basin of a higher minimizer of the objective function, but it does have a minimizer or saddle point in a basin of a lower minimizer of the objective function.

If $GCLASS=3, then we can choose two types of global optimization methods:

$GTYPE=1 - density clustering method [8]. Density clustering refers to a class of clustering techniques by using nonparametric probability density estimates to form clusters. All unclustered points from a reduced sample, which are within the threshold distance from the seed point, are added to the cluster.

$GTYPE=2 - single linkage clustering method [8]. In this case, the next two clusters to be merged are those for which the distance between the nearest points is the smallest. When this distance becomes larger than the threshold distance, the procedure is stopped. Starting with each point in a separate cluster, the points at distances less than the threshold distance are linked. A cluster is recognised as a set of points linked together.

If $GCLASS=4, then we can choose three types of global optimization methods:

$GTYPE=1 - multi level single linkage method [114]. In this case, the function values of the sample points are used in a very simple manner to obtain a very powerful method. The local search procedure is applied to every sample point, except if there is another sample point within the critical distance which has a smaller function value. Clusters can be constructed by associating a point to a local minimum, if there exists a chain of points linking it to that minimum. This is done so that the distance between each

- successive pair is, at most, equal to the critical distance and the function value is decreasing along the chain. A point in this way could be assigned to more than one minimum.

$GTYPE=2 - multi level mode analysis method [114]. This method is a generalization of the mode analysis method. Region is partitioned into cells. After sample reduction, it is determined which cells contain enough points to be "full". For each full cell the function value of the cell is defined to be equal to the smallest function value of any of the sample points in the cell. Finally, for every full cell, local minimization is applied except if a cell has a neighbouring cell which is full and has a smaller function value.

$GTYPE=3 - modified multi level single linkage method. This is a multi level single linkage method with some modifications that are described in [114].

The number of points randomly generated in the given region can be specified by using the macrovariable $MNRND. The default value is usually 100+20*NF. Since it depends on the number of variables and for NF>20 it is too large, we recommend the use of global optimization methods up to 20 variables only. If we use clustering or multi level single linkage methods ($GCLASS=3 or $GCLASS=4), then we can specify additional parameters.

$MNLMIN - maximum considered number of local minima. The default value is 50+20*NF.

$GAMA - reduction of random sample (typically 0.1D0 - 0.2D0). Greater value of GAMA usually leads to a greater number of local minima, but it requires a greater amount of work.

$SIGMA - parameter of cluster or single linkage termination (typically 1 - 8).

4. Output specifications in the UFO system

The UFO system has many output possibilities including the graphical pictures. These output possibilities can be divided into five basic groups.

4.1. Basic screen output

The basic screen output can be used only if $GRAPH='NO' and $DISPLAY='NO'. In this case, individual rows corresponding to iterations and final results are printed on the screen consequently. A print level of the screen output is determined by using the macrovariables $MOUT and $NOUT. The macrovariable $MOUT can have the following values:

$MOUT= 0 -	Screen output is suppressed.
$MOUT=± 1 -	Standard output. The final results appear on the screen.
$MOUT=± 2 -	Extended output. Additional information from every iteration appear on the screen.
$MOUT=± 3 -	Extended output. Additional final results of linear or quadratic programming subproblems appear on the screen.
$MOUT=± 4 -	Extended output. Additional information from every iteration of linear or quadratic programming subproblems appear on the screen.

If $MOUT>0, then a standard line of final results is printed, while if $MOUT<0 then a modified line of final results, containing termination criterion, is printed.

The macrovariable $NOUT can have the following values:

$NOUT= 0 -	Short final results (scalar variables) appear on the screen.
$NOUT= 1 -	Extended final results (vectors) appear on the screen.

4.2. Extended screen output

If we want to use an extended screen output, we have to set $DISPLAY='YES' (the default value is $DISPLAY='NO'). This type of screen output consists of text pages which correspond to individual iterations and final results. Final results are divided into several groups which can be displayed successively. We can change the displayed group by typing particular characters from the keyboard.

Change of the displayed group of final results:

F - (function) :	Value of the objective function and statistics.
V - (variables) :	Values of variables if NF>0 (with their bounds if KBF>0).
A - (approximation) :	Values of approximating functions if NA>0 (with their prescribed values if KBA>0). Values of selected components of a solution of the set of ordinary differential equations at the prescribed mesh points if NE>0.
C - (constraints) :	Values of constraint functions if NC>0 (with their bounds if KBC>0).
D - (data) :	Data which specify the problem solved (sizes of problem and additional specifications).
O - (options) :	Options which specify the method used.

Exit:

Q - (quit) :	Exit from the extended screen output.

After typing each character we must use ENTER.

Besides these possibilities we can stop every iteration for scanning the iterative process. It is specified if we set $SCAN='YES' (the default value is $SCAN='NO') If $SCAN='NO', then the output of iterations is suppressed. Scanning of the iterative process can be terminated by using the character '!' from the keyboard.

4.3. Graphical screen output

The graphical output can be used only on PC computers under the MS DOS system. This possibility is not allowed on the UNIX workstations. If we want to use a graphical output, we have to set $GRAPH='YES' (the default value is $GRAPH='NO'). In this case, both iterations and final results appear in the graphical mode. Graphical form of final results can be specified in detail using macrovariables $PATH ('NO', 'YES', 'EXTENDED'), $MAP ('NO', 'YES', 'EXTENDED'), $HIL ('NO', 'YES') and $ISO ('NO', 'YES'). Final results are divided into several groups which can be displayed successively. We can change the displayed group by typing particular characters from the keyboard.

Change of the displayed group of final results:

F - (function) :	Value of the objective function and statistics.
V - (variables) :	Values of variables if NF>0 (with their bounds if KBF>0).
A - (approximation) :	Values of approximating functions if NA>0 (with their prescribed values if KBA>0). Values of selected components of a solution of the set of ordinary differential equations at the prescribed mesh points if NE>0.
C - (constraints) :	Values of constraint functions if NC>0 (with their bounds if KBC>0).
D - (data) :	Data which specify the problem solved (sizes of problem and additional specifications).
O - (options) :	Options which specify the method used.
T - (path) :	Values of the objective function and selected variables (we can change these variables during the graphical output, if we have specified $PATH='EXTENDED') in the last NPA iterations (only if $PATH='YES' or $PATH='EXTENDED').

Exit:

Q - (quit) :	Exit from the graphical output.
X - (quit) :	Exit from the control system.

After typing each character we must use ENTER.

Besides these possibilities we can stop every iteration for scanning the iterative process. It is specified if we set $SCAN='YES' (the default value is $SCAN='NO'). In every iteration we can choose all possibilities F, V, A, C, D, O as above. If we have chosen either V (variables) or A (approximation) or C (constraints), then intermediate results can be displayed graphically by typing G (graph) from the keyboard. In all these cases we can execute a single iteration typing ENTER merely; in the highest graphics level we can execute all iterations until the k-th by entering the number k or all remaining iterations by typing the character '!' from the keyboard.

Besides text representations in the graphical mode, which are essentially like the ones in the extended screen output (with the choice $DISPLAY='YES'), we can chose several types of graphical data representation.

a) Graphical picture:

If we have chosen either V (variables) or A (approximation) or C (constraints), then results can be displayed graphically by typing G (graph) from the keyboard. A graphical picture appears on the screen in this case. It contains either values of variables with indices I, $1 \leq \mathrm{I} \leq \mathrm{NF}$, or values of the approximating functions with indices KA, $1 \leq \mathrm{KA} \leq \mathrm{NA}$, or values of the constraint functions with indices KC, $1 \leq \mathrm{KC} \leq \mathrm{NC}$. If we have chosen A (approximation) in the case of NE>0, then the graphical picture contains a component (with the index VAR) of a solution of the set of ordinary differential equations at the mesh points AT(KA), $1 \leq \mathrm{KA} \leq \mathrm{NA}$. We have to define the index VAR from the keyboard in this case. The graphical picture can be changed by typing particular characters from the keyboard.

Change of representation:

V - (values) :	Values are drawn.
O - (ordinates) :	Values and ordinates from zero axis are drawn.

C - (curves) : Values are connected by a curve.
M - (mixed) : Curve and ordinates are drawn.

Change of graph (if either KBF>0 or KBA>0 or KBC>0):

F - (functions) : Either values of variables X(I), $1 \leq I \leq NF$, or values of approximating functions AF(KA), $1 \leq KA \leq NA$, or values of constraint functions CF(KC), $1 \leq KC \leq NC$, are demonstrated.
A - (approximation) : Either values of variables X(I) together with their bounds XL(I) and XU(I), $1 \leq I \leq NF$, or values of approximating functions AF(KA) together with their prescribed values AM(KA) , $1 \leq KA \leq NA$, or values of constraint functions CF(KC) together with their bounds CL(KC) and CU(KC), $1 \leq KC \leq NC$, are demonstrated.
D - (differences) : Either differences between variables and their bounds or differences between approximating functions and their prescribed values or differences between constraint functions and their bounds are demonstrated.

Continuation (if either $NF > 200$ or $NA > 200$ or $NC > 200$):

P - (previous) : Previous set of at most 200 values is drawn.
N - (next) : Next set of at most 200 values is drawn.

New graph or return:

W - (new) : This possibility can be used only if NE>0. Then a new component (with a new index VAR) of a solution of the set of ordinary differential equations is drawn. We have to define a new index VAR from the keyboard in this case.
Q - (quit) : Return to the displayed group of final results.

After typing each character we must use ENTER.

If we have chosen F (function) as a group of final results, we can use additional graphical representations.

b) Two dimensional orbit:

If NE>1, we can draw an orbit of two components of a solution of the set of ordinary differential equations by typing G (graph) from the keyboard. We have to define an index VAR for every selected component of a solution (according to the text appeared on the screen). Two dimensional orbit can be changed by typing particular characters from the keyboard.

Change of the orbit:

V - (values) : Values are drawn.
C - (curves) : Values are connected by a curve.

New orbit or return:

W - (new) : New components of a solution of the set of ordinary differential equations are drawn. We have to define new two indices from the keyboard in this case.
Q - (quit) : Return to the displayed group of final results.

After typing each character we must use ENTER.

c) Three dimensional orbit:

If NE>2, we can draw an orbit of three components of a solution of the set of ordinary differential equations by typing P (picture) from the keyboard. We have to define an index VAR for every selected component of a solution (according to the text appeared on the screen). Three dimensional orbit can be changed by typing particular characters from the keyboard.

Change of the orbit:

V - (values) : Values are drawn.

C - (curves) : Values are connected by a curve.
O - (rotate) : Rotation of values or curves about a vertical axis by a subsequently entered angle Dfi.
T - (tilt) : Tilting rotated values or curves by a subsequently entered angle Dtheta.
A - (axes) : Drawing a picture with rotated and tilted axes.
S - (scale) : Scaling of rotated and tilted values or curves to make full use of the screen.

New orbit or return:

W - (new) : New components of a solution of the set of ordinary differential equations are drawn. We have to define new three indices from the keyboard in this case.
Q - (quit) : Return to the displayed group of final results.

After typing each character we must use ENTER.

d) Colored map of the objective function:

If we have specified either $MAP='YES' or $MAP='EXTENDED' (default value is $MAP='NO'), we can draw a colored map of the objective function by typing M (map) from the keyboard. This picture can be changed by typing particular characters from the keyboard.

Change of the map:

L - (linear) : Linear scale of the colored map.
G - (logarithmic) : Logarithmic scale of the colored map.
R - (refinement) : Refinement of the colored map.
B - (back) : Back refinement of the colored map.
N - (inverse) : Colored map of the objective function negation.

Another type of picture, new map or return:

H - (hills) : Drawing an objective function surface with respect to visibility.
I - (isolines) : Drawing objective function contours.
W - (new) : Selection of new variables and drawing a new colored map.
Q - (quit) : Return to the displayed group of final results.

After typing each character we must use ENTER.

If we set $MAP='YES', then one picture for two variables is drawn. If we set $MAP='EXTENDED', then three pictures for all combinations of two from three variables are drawn. In both cases we have to define, from the keyboard, an index VAR and bounds XL(VAR), XU(VAR) for every used variable (according to the text appeared on the screen). Note that the choice $MAP='EXTENDED' excludes the choices $HIL='YES' and $ISO='YES', so that the other pictures cannot be used.

e) Objective function surface:

If we have specified $HIL='YES' (default value is $HIL='NO'), we can draw an objective function surface with respect to visibility by typing H (hills) from the keyboard. This picture can be changed by typing particular characters from the keyboard.

Change of the surface:

L - (linear) : Linear scale of the surface.
G - (logarithmic) : Logarithmic scale of the surface.
R - (refinement) : Refinement of the surface.
B - (back) : Back refinement of the surface.
O - (rotate) : Rotation of the surface about a vertical axis by a subsequently entered angle Dfi.
T - (tilt) : Tilting the rotated surface by a subsequently entered angle Dtheta.
F - (face) : Facing the rotated surface (drawing the rotated surface without tilting).
N - (inverse) : Surface of the objective function negation.

Another type of picture, new surface or return:

M - (map) :	Drawing a colored map of the objective function.
I - (isolines) :	Drawing objective function contours.
W - (new) :	Selection of new variables and drawing new surface.
Q - (quit) :	Return to the displayed group of final results.

After typing each character we must use ENTER.

Before drawing the objective function surface we have to define, from the keyboard, an index VAR and bounds XL(VAR), XU(VAR) for every used variable (according to the text appeared on the screen).

f) Objective function contours:

If we have specified $ISO='YES' (default value is $ISO='NO'), we can draw an objective function contours by typing I (isolines) from the keyboard. This picture can be changed by typing particular characters from the keyboard.

Change of contours:

L - (linear) :	Linear scale of contours.
G - (logarithmic) :	Logarithmic scale of contours.
R - (refinement) :	Refinement of contours.
B - (back) :	Back refinement of contours.
O - (color) :	Coloring of contours and used levels.
N - (inverse) :	Inverse coloring of contours and used levels.

Another type of picture, new contours or return:

M - (map) :	Drawing a colored map of the objective function.
H - (hills) :	Drawing an objective function surface with respect to visibility.
W - (new) :	Selection of new variables and drawing a new surface.
Q - (quit) :	Return to the displayed group of final results.

After typing each character we must use ENTER.

g) Graphical path of the objective function and selected variables:

If we have chosen T (path), then we can display the values of the objective function as a function graph by typing G (graph) or draw the objective function contours with the path in the last NPA iterations. The graph can be changed in the same way as in a).

Change of contours:

L - (linear) :	Linear scale of contours.
G - (logarithmic) :	Logarithmic scale of contours.
R - (refinement) :	Refinement of contours.
B - (back) :	Back refinement of contours.
Z - (zoom) :	Zoom of the path for the number of last iterations entered.

Another type of picture, new contours or return:

W - (new) :	Selection of new variables and drawing a new contours (only if we have specified $PATH='EXTENDED').
Q - (quit) :	Return to the displayed group of final results.

After typing each character we must use ENTER.

Before drawing the objective function contours we have to define, from the keyboard, an index VAR and bounds XL(VAR), XU(VAR) for every used variable (according to the text appeared on the screen).

4.4. Text file output

The UFO system contains a great number of text file output procedures which are controlled by using the macrovariables $KOUT, $KOUT1, $KOUT2, $KOUT3, and $LOUT. These text file output procedures are useful especially for debugging new optimization methods. The UFO system works with the output file P.OUT. The Fortran number of this output file defines the common variable IWR. The macrovariables $KOUT, $KOUT1, $KOUT2, $KOUT3 determines what is printed and the macrovariable $LOUT has an influence to the extent of the print.

The macrovariable $KOUT can have the following values:

$KOUT= 0 -	Text file output is suppressed (the file P.OUT is empty) .
$KOUT= ± 1 -	Standard output. The heading and the final results are printed together with selected information on each accepted iteration.
$KOUT= ± 2 -	Extended output. Additional information, obtained from stepsize selection, is printed.
$KOUT= ± 3 -	Extended output. Additional information, obtained from direction determination and variable metric update, is printed.
$KOUT= ± 4 -	Extended output. Additional information, obtained from linear constraint addition and deletion, is printed.
$KOUT= ± 5 -	Extended output. Additional information, obtained from numerical differentiation, is printed.

If $KOUT>0, then a standard heading is printed, while if $KOUT<0 then an extended heading, containing problem specifications and optimization options, is printed.

A selection of iterations, accepted for print, is controlled by the contents of the macrovariables $KOUT1, $KOUT2, $KOUT3. If KOUT1≤ KOUT2 then only the iterations whose numbers are between KOUT1 and KOUT2 are assumed, but KOUT3−1 ones are always omitted (KOUT1 is a lower bound, KOUT2 is an upper bound and KOUT3 is a step). Similarly, if KOUT1>KOUT2, then only the iterations whose numbers are less than KOUT2 or greater then KOUT1 are assumed, but KOUT3−1 ones are always omitted. If $KOUT3=0, then no iterations are assumed.

While the macrovariable $KOUT specifies which information is printed, the macrovariable $LOUT specifies how much information is printed:

$LOUT= 0 -	Basic output. The basic information (1 row if $KOUT=1) is printed in each accepted iteration.
$LOUT=± 1 -	Extended output. Additional scalars, together with vector of variables, are printed.
$LOUT=± 2 -	Extended output. Additional vectors (usually gradients) are printed.
$LOUT=± 3 -	Extended output. Aditional matrices (usually Hessian matrices) are printed.
$LOUT=± 4 -	The most extended output. All useful data are printed.

If $LOUT>0, then basic part of the information is printed. If $LOUT<0, then a more extensive part of the information is printed.

The macrovariable $LOUT has an additional significance. If $KOUT=0 and $LOUT>0, then a copy of the basic screen output is provided. If $KOUT=0 and $LOUT<0, then paper saving print is assumed. In the last case, only several rows are printed for every solution. This type of output is useful for simultaneous tests of optimization methods.

To show a typical basic output which corresponds to the choices $KOUT=1, $KOUT3=0 and $LOUT=0 we propose the following results from unconstrained optimization:

```
UNCONSTRAINED MINIMIZATION USING UFO SYSTEM
-------------------------------------------
OPTIMIZATION SUBROUTINE : U1FDU1
DIRECTION DETERMINATION : UDDLI1
STEP SIZE DETERMINATION : US0L01
FUNCTION  DETERMINATION : UF1F01
GRADIENT  DETERMINATION : UF0GS2
H MATRIX  DETERMINATION :
VARIABLE METRIC  UPDATE : UUDBI1

PROBLEM
-------
NF = 2               KDF= 0 KSF= 1 KCF= 2 KBF= 0 ISNF= 1 NORMF= 0
NA = 0 NAL= 0 MAL= 0 KDA=-1 KSA= 0 KCA= 0 KBA= 0 ISNA= 0 NORMA= 0
NC = 0 NCL= 0 MCL= 0 KDC=-1 KSC= 0 KCC= 0 KBC= 0 ISNC= 0 NORMC= 0

FINAL RESULTS
-------------
FF= -.3072281498D+03
X = -.6228926480D+01  .4363683132D+01

TERMINATION: ITERM=4 GRAD TOL F=-.307D+03 G= .480D-06 D= .148D-07

STATISTICS
----------
NIT = 14                    NDEC = 0
NFV = 58  NAV = 0  NCV = 0  NRES = 6
NFG =  0  NAG = 0  NCG = 0  NREM = 0
NFH =  0  NAH = 0  NCH = 0  NADD = 0
```

Here the optimization subroutines used are listed on the top followed by problem specifications. After brief results, the termination causes are written. The termination cause ITERM=4 (GRAD TOL) corresponds to the attainment of the required gradient norm, F is the objective function value, G is the maximum absolute value of gradient elements and D is the maximum relative change of variables. The statistics contain the number of iterations NIT, the number of decompositions NDEC, the number of restarts NRES, the number of constraint deletions or additions NREM or NADD respectively, and a set of data concerns numbers (N) of model function (F) or approximating functions (A) or constraint functions (C) values (V) or gradients (G) or Hessian matrices (H) evaluations respectively.

4.5. User supplied output

The UFO system allows to the utilization of both the user supplied output subroutines and the post-processing subroutines. These subroutines can be included into the control program by using the macrovariable $OUTPUT:

$SET(OUTPUT)
    Calling the user supplied output subroutines.
    Calling the post-processing subroutines.
$ENDSET

Parameters of the user supplied output subroutines and post-processing subroutines must satisfy the

UFO conventions. For example, the vector of variables, the model function value, the model function gradient must be denoted X, FF, GF respectively (see chapter 2).

4.6. Storing final results

If we set $OUTPUTDATA='YES', then final values of the variables X(I), $1 \leq I \leq$ NF, are stored in the file P.DAT. Similarly, if we set $INPUTDATA='YES', then values of the variables X(I), $1 \leq I \leq$ NF, from the file P.DAT are used as input data for the new optimization process.

4.7. Tracing in the UFO control program

Tracing in the control program is a useful tool for debugging optimization algorithms on main-frames. If this is the case, then we will specify $TRACE='YES'. Besides simple tracing, we can prescribe scalar integer or real variables whose values will be printed together with labels. This possibility can be specified by using the macrovariables $IDEB and $RDEB:

$IDEB = 'list of integer variables separated by commas'
$RDEB = 'list of real variables separated by commas'

If the macrovariables $IDEB or $RDEB are not specified, then no integer or real variables are printed.

Tracing is executed only in the accepted iterations whose numbers are determined by using the macrovariables $KOUT1, $KOUT2, $KOUT3 (see Section 4.5).

4.8. Error messages

If we use the specification $MOUT>0 (basic screen output), then nonstandard terminations are indicated. The message consists of three parts: the name of a critical subroutine, the number of a message, and an explanation text. This possibility serves especially for a debugging and no details are given here.

5. Special tools of the UFO system

The UFO system contains special tools that facilitate the user's activity. There are tools for checking the correctness of optimization problems and for testing optimization methods.

5.1. Checking external subroutines

The values, gradients, Hessian matrices of the model function or the approximating functions or the constraint functions are specified by using the macrovariables $FMODELF, $GMODELF, $HMODELF or $FMODELA, $GMODELA, $HMODELA or $FMODELC, $GMODELC, $HMODELC, respectively. Sometimes checking the correctness of these models is needed. If this is the case, then both the analytical and the numerical differentiation can be compared. Checking optimization problems can be specified by using the macrovariable $TEST. If $TEST='NO', then no checking is performed. If $TEST='YES', then both the analytical and the numerical differentiation is executed before optimization is started and the derivatives obtained are printed. Only the derivatives that are analytically specified (the first, the second) are checked. Finally, if $TEST='ONLY', then only checking is performed and optimization is not started. An output of checking an optimization problem has the following form:

```
STANDARD TEST OF EXTERNAL SUBROUTINES
-------------------------------------
PROBLEM NO    1

PROBLEM
-------
NF = 2                    KDF =  2  KSF = 1  KCF = 2  NORMF = 0
NA = 0  NAL = 0  MAL = 0  KDA = -1  KSA = 0  KCA = 0  NORMA = 0
NC = 3  NCL = 0  MCL = 0  KDC =  1  KSC = 0  KCC = 2  NORMC = 0

PARAMETERS
----------
X    =   -.2000000000D+01    .1000000000D+01

DERIVATIVES
-----------
FF A =    .9090000000D+03
GF N =   -.2405999822D+04   -.6000004263D+03
GF A =   -.2406000000D+04   -.6000000000D+03
HF N =    .4402000148D+04    .8000000070D+03    .2000000002D+03
HF A =    .4402000000D+04    .8000000000D+03    .2000000000D+03

FC A =   -.1000000000D+01
GC N =    .1000000000D+01    .2000000032D+01
GC A =    .1000000000D+01    .2000000000D+01

FC A =    .5000000000D+01
GC N =   -.4000000070D+01    .9999999930D+00
GC A =   -.4000000000D+01    .1000000000D+01

FC A =    .5000000000D+01
GC N =   -.4000000070D+01    .2000000042D+01
GC A =   -.4000000000D+01    .2000000000D+01
```

Here the letter 'N' indicates a numerical differentiation and the letter 'A' indicates an analytical differentiation.

5.2. Testing optimization methods

The UFO system contains a great number of subroutines (collections of test problems) that serve for testing optimization methods. All of these subroutines begin with the letter 'E' (external). Input subroutines have the second letter 'I' and the third letter 'U' or 'L' or 'N' for an unconstrained or linearly constrained or nonlinearly constrained problems, respectively. The model specification subroutines have the second letter 'F' or 'A' or 'C' or 'E' or 'Y' for a model function or approximating functions or constraint functions or state functions or initial functions, respectively, and the third letter 'F' or 'G' or 'H' for values or gradients or Hessian matrices, respectively. The fourth letter is always 'U' or 'D' or 'S' or 'B' for universal or dense or sparse or partitioned problems, respectively. The last two digits specify individual test problems collections. When we want to carry out a test of the selected method, we use the specicications $COLLECTION='YES' and $NEXT=number_of_test_problems in the input batch file.

Tests corresponding to individual test problems collections are realized by using the following test input files:

TEST01*.UFO - Tests for unconstrained optimization (15 dense problems from [21], [73]). External subroutines EIUD01, EFFU01, EFGU01, EFHD01 are used.

TEST02*.UFO - Tests for sum of squares minimization (30 dense problems from [98]). External subroutines EIUD02, EAFU02, EAGU02, EAHD02 are used.

TEST03*.UFO - Tests for linearly constrained optimization (16 dense problems from [58]). External subroutines EILD03, EFFU03, EFGU03 are used.

TEST04*.UFO - Tests for medium-size linear programming (6 dense problems). External subroutine EILD04 is used.

TEST05*.UFO - Tests for medium-size quadratic programming (5 dense problems). External subroutine EILD05 is used.

TEST06*.UFO - Tests for minimax (7 dense problems from [71]). External subroutines EIUD06, EAFU06, EAGU06, EAHD06 are used.

TEST07*.UFO - Tests for inequality constrained nonlinear programming (34 dense problems from [58]). External subroutines EIND07, EFFU07, EFGU07, ECFU07, ECGU07 are used.

TEST08*.UFO - Tests for equality constrained nonlinearp rogramming (31 dense problems from [58]). External subroutines EIND08, EFFU08, EFGU08, ECFU08, ECGU08 are used.

TEST09*.UFO - Tests for unconstrained global optimization (13 problems from [141]). External subroutines EIUD09, EFFU09, EFGU09 are used.

TEST10*.UFO - Tests for unconstrained optimization (15 sparse problems from [73], [134]). External subroutines EIUS10, EFFU10, EFGU10, EFHS10 are used.

TEST11*.UFO - Tests for large-scale linear programming (18 sparse problems). External subroutine EILS11 is used.

TEST12*.UFO - Tests for large-scale quadratic programming (11 sparse problems). External subroutine EILS12 is used.

TEST13*.UFO - Tests for linearly constrained optimization (8 sparse problems). External subroutines EILS13, EFFU13, EFGU13 are used.

TEST14*.UFO - Tests for sum of functions minimization (15 sparse problems from [73], [134]). External subroutines EIUB14, EAFU14, EAGU14 are used.

TEST15*.UFO - Tests for sum of squares minimization (24 sparse problems from [77]). External subroutines EIUB15, EAFU15, EAGU15 are used.

TEST16*.UFO - Extended tests for unconstrained optimization (80 dense problems from [21], [73], [98]). External subroutines EIUD16, EFFU16, EFGU16, EFHD16 are used.

TEST17*.UFO - Tests for nonlinear equations solutions (30 dense problems). External subroutines EIUD17, EAFU17, EUGU17 are used.

TEST18*.UFO - Tests for nonlinear equations (32 sparse problems from [80]). External subroutines EIUS18, EAFU18, EAGU18 are used.

TEST19*.UFO - Tests for nonsmooth unconstrained optimization (24 dense problems from [90], [64]). External subroutines EIUD19, EFFU19, EFGU19, EFHD19 are used.

TEST20*.UFO - Tests for equality constrained sparse nonlinear programming (18 sparse problems from [86]). External subroutines EIUB20, EIUS20, EIND20, EINS20, EFFU20, EFGU20, EAFU20, EAGU20, ECFU20, ECGU20 are used.

TEST21*.UFO - Tests for optimization of dynamical systems (4 dense problems). External subroutines EIUD21, EEFU21, EEGU21, EYFU21, EYGU21 are used.

TEST22*.UFO - Tests for linearly constrained minimax optimization (6 dense problems from [71]). External subroutines EIUD22, EAFU22, EAGU22, EAHD22 are used.

TEST24*.UFO - Tests for sum of squares minimization (115 dense problems from [21], [73], [77], [80], [98]). External subroutines EIUD24, EAFU24, EAGU24 are used.

TEST25*.UFO - Tests for sum of functions minimization (65 sparse problems from [73], [77], [80], [134]). External subroutines EIUB14, EAFU14, EAGU14, EIUB15, EAFU15, EAGU15, EIUB18, EAFU18, EAGU18 are used.

TEST32*.UFO - Tests for sum of squares minimization (6 dense problems from [75]). External subroutines EIUD32, EAFU32, EAGU32 are used.

TEST33*.UFO - Tests for sum of squares minimization (6 dense problems from [75]). External subroutines EIUD33, EAFU33, EAGU33 are used.

In these input files, all necessary macrovariables are defined and the external subroutines are called. The external subroutines with the last two digits 01, ..., 23 are briefly described in the text files E01.TXT, ..., E24.TXT.

To demonstrate the use of the test input file we perform a test of sum of squares minimization by using hybrid method realized as a trust region method. The test input file TEST02.UFO has the form:

```
$SET(INPUT)
  CALL EIUD02(NF,NA,NAL,X,FMIN,XMAX,NEXT,IEXT,IERR)
  IF(IERR.NE. 0) GO TO $$ENDTEST
$ENDSET
$SET(FMODELA)
  CALL EAFU02(NF,KA,X,FA,NEXT)
$ENDSET
$SET(GMODELA)
  CALL EAGU02(NF,KA,X,GA,NEXT)
$ENDSET
$NF=12
$NA=400
$KOUT=0
$LOUT=1
$MOUT=1
$MIT=500
$MFV=1000
$MODEL='AQ'
$CLASS='GN'
$TYPE='G'
$DECOMP='M'
$NUMBER=7
$UPDATE='F'
$TOLX='1.0$P-16'
$TOLF='1.0$P-16'
```

```
$TOLB='1.0$P-16'
$TOLG='1.0$P-6'
$COLLECTION='YES'
$NEXT=30
$BATCH

$STANDARD
```

The result (screen output) obtained has the following form (each row corresponds to one test problem and the last row is a summary):

```
   1  NIT=  12  NFV=  15  NFG=  13  NDC=  28  NCG=   0  F= .256D-29  G= .440D-13
   2  NIT=  21  NFV=  28  NFG=  22  NDC=  51  NCG=   0  F= .245D+02  G= .678D-06
   3  NIT=  33  NFV=  34  NFG=  34  NDC=  64  NCG=   0  F= .204D-22  G= .581D-06
   4  NIT=  13  NFV=  16  NFG=  14  NDC=  27  NCG=   0  F= .380D-19  G= .276D-03
   5  NIT=   6  NFV=   7  NFG=   7  NDC=   8  NCG=   0  F= .142D-15  G= .807D-07
   6  NIT=  11  NFV=  17  NFG=  12  NDC=  28  NCG=   0  F= .622D+02  G= .790D-07
   7  NIT=   7  NFV=   8  NFG=   8  NDC=  11  NCG=   0  F= .203D-26  G= .622D-12
   8  NIT=   5  NFV=   6  NFG=   6  NDC=   8  NCG=   0  F= .411D-02  G= .261D-08
   9  NIT=   1  NFV=   2  NFG=   2  NDC=   1  NCG=   0  F= .564D-08  G= .177D-07
  10  NIT= 127  NFV= 133  NFG= 128  NDC= 275  NCG=   0  F= .440D+02  G= .608D-04
  11  NIT=  69  NFV=  77  NFG=  70  NDC= 177  NCG=   0  F= .592D-20  G= .213D-06
  12  NIT=  12  NFV=  14  NFG=  13  NDC=  29  NCG=   0  F= .114D-20  G= .601D-10
  13  NIT=  10  NFV=  11  NFG=  11  NDC=  14  NCG=   0  F= .169D-09  G= .247D-06
  14  NIT=  41  NFV=  50  NFG=  42  NDC=  82  NCG=   0  F= .142D-23  G= .280D-10
  15  NIT=  11  NFV=  14  NFG=  12  NDC=  29  NCG=   0  F= .154D-03  G= .511D-06
  16  NIT=  24  NFV=  55  NFG=  24  NDC= 103  NCG=   0  F= .429D+05  G= .153D-05
 USOG01: ( 6) MAXIMUM NUMBER OF REDUCTIONS
  17  NIT=  22  NFV=  24  NFG=  23  NDC=  61  NCG=   0  F= .273D-04  G= .476D-07
  18  NIT=  31  NFV=  41  NFG=  32  NDC= 144  NCG=   0  F= .283D-02  G= .404D-06
  19  NIT=  13  NFV=  15  NFG=  14  NDC=  29  NCG=   0  F= .219D-01  G= .373D-07
  20  NIT=   7  NFV=   8  NFG=   8  NDC=  28  NCG=   0  F= .325D-09  G= .245D-10
  21  NIT=  12  NFV=  15  NFG=  13  NDC=  28  NCG=   0  F= .149D-28  G= .446D-13
  22  NIT=  10  NFV=  11  NFG=  11  NDC=  14  NCG=   0  F= .506D-09  G= .247D-06
  23  NIT=  20  NFV=  25  NFG=  21  NDC=  41  NCG=   0  F= .439D-04  G= .197D-06
  24  NIT=  24  NFV=  34  NFG=  25  NDC= 107  NCG=   0  F= .308D-03  G= .354D-06
  25  NIT=  10  NFV=  11  NFG=  11  NDC=  10  NCG=   0  F= .125D-25  G= .190D-11
  26  NIT=   9  NFV=  13  NFG=  10  NDC=  21  NCG=   0  F= .138D-06  G= .491D-07
  27  NIT=   6  NFV=   7  NFG=   7  NDC=   9  NCG=   0  F= .946D-18  G= .140D-08
  28  NIT=   7  NFV=   8  NFG=   8  NDC=  18  NCG=   0  F= .211D-10  G= .379D-06
  29  NIT=   2  NFV=   3  NFG=   3  NDC=   2  NCG=   0  F= .799D-13  G= .203D-06
  30  NIT=   5  NFV=   6  NFG=   6  NDC=   6  NCG=   0  F= .138D-26  G= .145D-12
 TOTAL   NIT=     581   NFV=     708   NFG=     610   NDC=    1453 *    29
         NCG=       0   NRS=       1   NAD=       0   NRM=       0
```

6. Application of the UFO system (examples)

Before the solution of a given problem, the input file containing the problem description and other specifications for macroprocessor must usually be prepared. This input file can contain only the macroinstruction $STANDARD (input file STANDARD.UFO). Then a full dialogue is processed. However, a more advantageous possibility is to prepare an input file containing a problem description while a method selection is left to the dialogue. Moreover, since a method selection can be made automatically by using knowledge bases coded in UFO templates, the batch mode is recommended.

When writing input file instructions, we have to observe some conventions. Since a control program contains a great number of common variables, we recommend using variables beginning with the letter 'W' for a problem description to avoid their double use. Real variables of this type should be declared at the beginning of the control program by the statement $FLOAT (for example $FLOAT W,W1,W2). Simple integers I,J,K,L need not be declared. We recommend using statement numbers less than 10000 for a problem description to avoid their double use.

The basic implementation of the UFO system is in a double precision arithmetic. Therefore, usually $FLOAT='REAL*8' and $P='D'. We recommend writing real constants always in the form of $P or D specification (for example 1.0$P 2, 4.0$P-1 or 1.0D 2, 4.0D-1) since the conversions from a single precision, that depend on a compiler, can be incorrect. Instead of the constants 0.0D0, 1.0D0, 2.0D0, 3.0D0, 4.0D0, 5.0D0, 1.0D1, we can use the common variables ZERO, HALF, ONE, TWO, THREE, FOUR, FIVE, TEN which contain corresponding values.

In the following text, we demonstrate the application of the UFO system to 19 typical problems. Every example consists of the problem description, the problem specification (input file), comments to the problem specification and the problem solution (basic screen output). All input files contain necessary data and can be used in the batch mode. These input files are included to the UFO system as the demofiles PROB01.UFO,... ,PROB19.UFO.

6.1. Optimization with simple bounds

a) Problem description:

Suppose we have to find a maximum of the objective function

$$F(x) = \frac{1}{n!}\left(\prod_{i=1}^{n} x_i\right) - 2$$

with simple bounds $0 \le x_i \le i$ for $1 \le i \le n$, where $n = 5$. The starting point is $x_i = 2$ for $1 \le i \le n$. The solution point is $x_i = i$ for $1 \le i \le n$ and the corresponding maximum value of the objective function is $F = -1.0$.

b) Problem specification (input file):

```
$FLOAT W
$SET(INPUT)
  DO 1 I=1,NF
  X(I)=2.D0 ; XL(I)=0.D0 ; XU(I)=DBLE(I) ; IX(I)=3
1 CONTINUE
$ENDSET
$SET(FGMODELF)
  W=1.D0
  DO 2 I=1,NF
  W=W*X(I)/DBLE(I)
2 CONTINUE
  FF=W-2.D0
  DO 3 I=1,NF
```

```
  GF(I)=W/X(I)
3 CONTINUE
$ENDSET
$IEXT=1
$NF=5
$KBF=2
$MOUT=2
$NOUT=1
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify initial values and simple bounds for variables. By using the macrovariable $FGMODELF we specify analytically the value and the gradient of the model function. Because we look for a maximum, we set $IEXT=1.

d) Problem solution (basic screen output):

```
         NIT=    0   NFV=    1   NFG=    1   F=  .187D+01   G=  .667D-01
         NIT=    1   NFV=    4   NFG=    4   F=  .155D+01   G=  .150D+00
         NIT=    2   NFV=    7   NFG=    7   F=  .120D+01   G=  .200D+00
         NIT=    3   NFV=    9   NFG=    9   F=  .100D+01   G=  .000D+00
  0  NIT=   3  NFV=   9  NFG=   9  NDC=   0  NCG=   0  F= .100D+01  G= .000D+00
 FF=  -.1000000000D+01
 X =   .1000000000D+01   .2000000000D+01   .3000000000D+01   .4000000000D+01
       .5000000000D+01
```

6.2. Minimization of the sum of squares

a) problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \frac{1}{2}\sum_{i=1}^{m}\left(x_4 e^{-x_1 t_i} + x_5 e^{-x_2 t_i} + x_6 e^{-x_3 t_i} - y_i\right)^2$$

where $m = 20$, $t_i = i/10$ and $y_i = e^{-t_i} - 5e^{-10t_i} + 3e^{-4t_i}$ for $1 \le i \le m$. The starting point is $x_1 = 1, x_2 = 2, x_3 = 1, x_4 = 1, x_5 = 1, x_6 = 1$. The solution point is $x_1 = 1, x_2 = 10, x_3 = 4, x_4 = 1, x_5 = 5, x_6 = 3$ and the corresponding minimum value of the objective function is $F = 0.0$

b) Problem specification (input file):

```
$FLOAT W,WA,WB,WC
$SET(INPUT)
  X(1)=1.D0 ; X(2)=2.D0 ; X(3)=1.D0
  X(4)=1.D0 ; X(5)=1.D0 ; X(6)=1.D0
  DO 1 KA=1,NA
    W=0.1D0*DBLE(KA)
    AM(KA)=EXP(-W)-5.D0*EXP(-1.0D1*W)+3.D0*EXP(-4.D0*W)
  1 CONTINUE
  XMAX=1.D1
  FMIN=0.D0
```

```
$ENDSET
$SET(FMODELA)
  W=0.1D0*FLOAT(KA)
  WA=EXP(-W*X(1))
  WB=EXP(-W*X(2))
  WC=EXP(-W*X(3))
  FA=X(4)*WA-X(5)*WB+X(6)*WC
$ENDSET
$NF=6
$NA=20
$NAL=0
$KBA=1
$MOUT=2
$NOUT=1
$MODEL='AQ'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable \$INPUT we specify the initial values of variables and the vector AM containing values $y_i, 1 \leq i \leq m$. Since the approximating functions contain exponentials, we define the maximum stepsize \$XMAX=10. By using the macrovariable \$FGMODELA we specify analytically the values of the approximating function. The gradients of the approximating functions are computed numerically. For the sum of squares minimization we set \$MODEL='AQ'. The specification \$KBA=1 indicates that the vector AM is used.

d) Problem solution (basic screen output):

```
NIT=    0   NFV=    7   NFG=    0   F=  .465D+00   G=  .675D+00
NIT=    1   NFV=   14   NFG=    0   F=  .264D+00   G=  .323D+00
NIT=    2   NFV=   21   NFG=    0   F=  .142D+00   G=  .426D-01
NIT=    3   NFV=   28   NFG=    0   F=  .116D+00   G=  .615D-01
NIT=    4   NFV=   35   NFG=    0   F=  .730D-01   G=  .638D-01
NIT=    5   NFV=   43   NFG=    0   F=  .518D-01   G=  .342D-01
NIT=    6   NFV=   50   NFG=    0   F=  .405D-01   G=  .123D+00
NIT=    7   NFV=   57   NFG=    0   F=  .269D-01   G=  .163D-01
NIT=    8   NFV=   64   NFG=    0   F=  .183D-01   G=  .704D-01
NIT=    9   NFV=   71   NFG=    0   F=  .123D-01   G=  .492D-01
NIT=   10   NFV=   79   NFG=    0   F=  .105D-01   G=  .789D-02
NIT=   11   NFV=   86   NFG=    0   F=  .771D-02   G=  .175D-01
NIT=   12   NFV=   93   NFG=    0   F=  .368D-02   G=  .462D-01
NIT=   13   NFV=  100   NFG=    0   F=  .219D-02   G=  .860D-01
NIT=   14   NFV=  107   NFG=    0   F=  .175D-02   G=  .781D-01
NIT=   15   NFV=  114   NFG=    0   F=  .494D-03   G=  .215D-02
NIT=   16   NFV=  122   NFG=    0   F=  .412D-03   G=  .323D-02
NIT=   17   NFV=  129   NFG=    0   F=  .301D-03   G=  .154D-01
NIT=   18   NFV=  136   NFG=    0   F=  .186D-03   G=  .117D-01
NIT=   19   NFV=  143   NFG=    0   F=  .114D-03   G=  .115D-01
NIT=   20   NFV=  150   NFG=    0   F=  .892D-04   G=  .155D-01
NIT=   21   NFV=  158   NFG=    0   F=  .267D-04   G=  .465D-02
NIT=   22   NFV=  166   NFG=    0   F=  .174D-04   G=  .110D-02
```

```
         NIT=   23   NFV=  173   NFG=    0   F=  .122D-04   G=  .413D-02
         NIT=   24   NFV=  180   NFG=    0   F=  .795D-05   G=  .482D-02
         NIT=   25   NFV=  187   NFG=    0   F=  .591D-05   G=  .532D-02
         NIT=   26   NFV=  194   NFG=    0   F=  .538D-05   G=  .556D-02
         NIT=   27   NFV=  201   NFG=    0   F=  .282D-10   G=  .145D-04
         NIT=   28   NFV=  208   NFG=    0   F=  .114D-21   G=  .303D-10
  0  NIT=  28  NFV= 208  NFG=   0  NDC= 105  NCG=   0  F= .114D-21  G= .303D-10
 F =    .1141152049D-21
 X =    .4000000000D+01   .1000000000D+02   .1000000000D+01   .3000000000D+01
        .5000000000D+01   .1000000000D+01
```

6.3. Minimax approximation

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \max_{1 \le i \le m} |\frac{x_1 + t_i x_2}{1 + t_i x_3 + t_i^2 x_4 + t_i^3 x_5} - y_i|$$

where $m = 21$, $t_i = (i-1)/10 - 1$ and $y_i = e^{-t_i}$ for $1 \le i \le m$. Starting point is $x_1 = 0.5$, $x_2 = 0$, $x_3 = 0$, $x_4 = 0$, $x_5 = 0$. The solution point is $x_1 = 0.9998, x_2 = 0.2536$, $x_3 = -0.7466$, $x_4 = 0.2452$, $x_5 = -0.3749$ and the corresponding minimum value of the objective function is $F = 0.000122371$.

b) Problem specification (input file):

```
$FLOAT W
$SET(INPUT)
  X(1)=0.5D0 ; X(2)=0.0D0 ; X(3)=0.0D0
  X(4)=0.0D0 ; X(5)=0.0D0
$ENDSET
$SET(FMODELA)
  W=0.1D0*DBLE(KA-1)-1.0D0
  FA=(X(1)+W*X(2))/(1.0D0+W*(X(3)+W*(X(4)+W*X(5))))-EXP(W)
$ENDSET
$MODEL='AM'
$NF=5
$NA=21
$NAL=0
$MOUT=2
$NOUT=1
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables. By using the macrovariable $FMODELA we specify analytically the values of the approximating functions. The gradients of the approximating functions are computed numerically. For minimax approximation we set $MODEL='AM'.

d) Problem solution (basic screen output):

```
         NIT=    0   NFV=    6   NFG=    0   F=  .222D+01   G=  .100D+61
         NIT=    1   NFV=   13   NFG=    0   F=  .425D+00   G=  .783D+00
         NIT=    2   NFV=   19   NFG=    0   F=  .827D-01   G=  .223D+00
         NIT=    3   NFV=   25   NFG=    0   F=  .126D-01   G=  .114D+00
         NIT=    4   NFV=   31   NFG=    0   F=  .669D-02   G=  .290D-01
         NIT=    5   NFV=   37   NFG=    0   F=  .563D-02   G=  .253D-01
         NIT=    6   NFV=   43   NFG=    0   F=  .128D-02   G=  .515D-01
         NIT=    7   NFV=   49   NFG=    0   F=  .631D-03   G=  .712D-02
         NIT=    8   NFV=   55   NFG=    0   F=  .134D-03   G=  .176D-02
         NIT=    9   NFV=   61   NFG=    0   F=  .122D-03   G=  .792D-05
         NIT=   10   NFV=   67   NFG=    0   F=  .122D-03   G=  .189D-08
  0  NIT=  10  NFV=  67  NFG=   0  NDC=   0  NCG=   0  F= .122D-03  G= .189D-08
 F =    .1223712513D-03
 X =    .9998776287D+00   .2535884404D+00  -.7466075717D+00   .2452015019D+00
       -.3749029101D-01
```

6.4. Nonsmooth optimization

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = -x_1 + 2 * (x_1^2 + x_2^2 - 1) + \frac{7}{4}|x_1^2 + x_2^2 - 1|$$

Starting point is $x_1 = -1,\ x_2 = -1.$. The solution point is $x_1 = 1,\ x_2 = 0$, and the corresponding minimum value of the objective function is $F = -1.0$.

b) Problem specification (input file):

```
$FLOAT W
$SET(INPUT)
  X(1)=-1.D0
  X(2)=-1.D0
$ENDSET
$SET(FGMODELF)
  W=X(1)**2+X(2)**2-1.D0
  FF=-X(1)+2.D0*W+1.75D0*ABS(W)
  W=SIGN(3.5$P 0,W)+4.D0
  GF(1)=W*X(1)-1.D0
  GF(2)=W*X(2)
$ENDSET
$NF=2
$KSF=3
$MOUT=2
$NOUT=1
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable \$INPUT we specify the initial values of variables. By using the macrovariable \$FGMODELF we specify analytically the value and the gradient of the objective function. For nonsmooth optimization we set \$KSF=3.

d) Problem solution (basic screen output):

```
         NIT=    0   NFV=    1   NFG=    1   F=  .475D+01   G=  .100D+61
         NIT=    1   NFV=    3   NFG=    3   F= -.379D+00   G=  .850D+01
         NIT=    2   NFV=    4   NFG=    4   F= -.606D+00   G=  .933D+00
         NIT=    3   NFV=    5   NFG=    5   F=  .925D+01   G=  .802D+00
         NIT=    4   NFV=    6   NFG=    6   F= -.728D+00   G=  .802D+00
         NIT=    5   NFV=    7   NFG=    7   F=  .871D+00   G=  .348D+00
         NIT=    6   NFV=    8   NFG=    8   F= -.828D+00   G=  .722D+00
         NIT=    7   NFV=    9   NFG=    9   F= -.844D+00   G=  .162D+00
         NIT=    8   NFV=   10   NFG=   10   F= -.999D+00   G=  .984D-01
         NIT=    9   NFV=   11   NFG=   11   F= -.998D+00   G=  .114D+00
         NIT=   10   NFV=   12   NFG=   12   F= -.999D+00   G=  .501D+00
         NIT=   11   NFV=   13   NFG=   13   F= -.100D+01   G=  .530D-01
         NIT=   12   NFV=   14   NFG=   14   F= -.100D+01   G=  .454D-05
         NIT=   13   NFV=   15   NFG=   15   F= -.100D+01   G=  .986D-07
  0  NIT=  13  NFV=  15  NFG=  15  NDC=   0  NCG=   0  F=-.100D+01  G= .986D-07
 FF=  -.1000000000D+01
 X =   .1000000000D+01   .0000000000D+00
```

6.5. Optimization with linear constraints

a) problem specification:

Suppose we have to find a minimum of the objective function

$$F(x) = (x_1 - x_2)^2 + (x_3 - 1)^2 + (x_4 - 1)^4 + (x_5 - 1)^6$$

over the set given by the linear constraints

$$x_1 + x_2 + x_3 + 4x_4 = 7$$

$$x_3 + 5x_5 = 6$$

The starting point is $x_1 = 10$, $x_2 = 7$, $x_3 = 2$, $x_4 = 3$, $x_5 = 0.8$. The solution point is $x_1 = 1$, $x_2 = 1$, $x_3 = 1$, $x_4 = 1$, $x_5 = 1$ and the corresponding minimum value of the objective function is $F = 0.0$

b) Problem specification (input file):

```
$SET(INPUT)
  X(1)= 1.D1 ; X(2)= 7.D0 ; X(3)= 2.D0
  X(4)=-3.D0 ; X(5)=0.8D0
  IC(1)=5 ; CL(1)=7.D0
  CG(1)=1.D0 ; CG(2)=1.D0 ; CG(3)=1.D0
  CG(4)=4.D0 ; CG(5)=0.D0
  IC(2)=5 ; CL(2)=6.D0
  CG(6)=0.D0 ; CG(7)=0.D0 ; CG(8)=1.D0
  CG(9)=0.D0 ; CG(10)=5.D0
  FMIN =0.D0
$ENDSET
$SET(FMODELF)
  FF=(X(1)-X(2))**2+(X(3)-1.D0)**2+ &
```

```
     (X(4)-1.D0)**4+(X(5)-1.D0)**6
$ENDSET
$SET(GMODELF)
  GF(1)= 2.D0*(X(1)-X(2))
  GF(2)=-2.D0*(X(1)-X(2))
  GF(3)= 2.D0*(X(3)-1.D0)
  GF(4)= 4.D0*(X(4)-1.D0)**3
  GF(5)= 6.D0*(X(5)-1.D0)**5
$ENDSET
$NF=5
$NC=2
$NCL=2
$KBC=1
$MOUT=2
$NOUT=1
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables and types and values of the general linear constraints. Since there are only the equality constraints, we can specify only the left sides (CL(1) and CL(2)) and we can set $KBC=1. The specification $FMIN=0 is used, since the objective function value cannot be less then zero. By using the macrovariable $FMODELF we specify analytically the value of the model function. By using the macrovariable $GMODELF we specify analytically the gradient of the model function.

d) Problem solution (basic screen output):

```
NIT=    0   NFV=    1   NFG=    1   F=  .266D+03   G=  .853D+02
NIT=    1   NFV=    2   NFG=    2   F=  .234D+02   G=  .911D+01
NIT=    2   NFV=    3   NFG=    3   F=  .908D+01   G=  .573D+01
NIT=    3   NFV=    4   NFG=    4   F=  .118D+01   G=  .213D+01
NIT=    4   NFV=    5   NFG=    5   F=  .328D+00   G=  .713D+00
NIT=    5   NFV=    6   NFG=    6   F=  .197D+00   G=  .356D+00
NIT=    6   NFV=    7   NFG=    7   F=  .697D-01   G=  .176D+00
NIT=    7   NFV=    8   NFG=    8   F=  .179D-01   G=  .115D+00
NIT=    8   NFV=    9   NFG=    9   F=  .564D-02   G=  .887D-01
NIT=    9   NFV=   10   NFG=   10   F=  .219D-02   G=  .438D-01
NIT=   10   NFV=   11   NFG=   11   F=  .838D-03   G=  .639D-02
NIT=   11   NFV=   12   NFG=   12   F=  .193D-03   G=  .101D-01
NIT=   12   NFV=   13   NFG=   13   F=  .550D-04   G=  .137D-02
NIT=   13   NFV=   14   NFG=   14   F=  .146D-04   G=  .117D-02
NIT=   14   NFV=   15   NFG=   15   F=  .487D-05   G=  .220D-02
NIT=   15   NFV=   16   NFG=   16   F=  .120D-05   G=  .118D-02
NIT=   16   NFV=   17   NFG=   17   F=  .791D-06   G=  .108D-02
NIT=   17   NFV=   18   NFG=   18   F=  .367D-06   G=  .403D-03
NIT=   18   NFV=   19   NFG=   19   F=  .899D-07   G=  .133D-03
NIT=   19   NFV=   20   NFG=   20   F=  .273D-07   G=  .111D-04
NIT=   20   NFV=   21   NFG=   21   F=  .842D-08   G=  .900D-04
NIT=   21   NFV=   22   NFG=   22   F=  .513D-08   G=  .150D-03
NIT=   22   NFV=   23   NFG=   23   F=  .172D-08   G=  .246D-04
```

```
        NIT=   23   NFV=   24   NFG=   24   F=  .104D-08   G=  .772D-05
        NIT=   24   NFV=   25   NFG=   25   F=  .278D-09   G=  .207D-04
        NIT=   25   NFV=   26   NFG=   26   F=  .828D-10   G=  .122D-04
        NIT=   26   NFV=   27   NFG=   27   F=  .318D-10   G=  .291D-05
        NIT=   27   NFV=   28   NFG=   28   F=  .604D-11   G=  .801D-06
  0  NIT=  27  NFV=  28  NFG=  28  NDC=   0  NCG=   0  F= .604D-11  G= .801D-06
 FF=   .6044149141D-11
 X =   .1003116995D+01   .1003116712D+01   .9999997424D+00   .9984416378D+00
       .1000000052D+01
```

6.6. Minimax approximation with linear constraints

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \max(f_1(x),\ f_2(x),\ f_3(x))$$

with

$$f_1(x) = -\exp(x_1 - x_2)$$
$$f_2(x) = \sinh(x_1 - 1) - 1$$
$$f_3(x) = -\log(x_2) - 1$$

over the set given by the box constraint $x_2 \geq 1/100$ and the linear constraint

$$\frac{5}{100}\, x_1 - x_2 + \frac{1}{2} \geq 0.$$

Starting point is $x_1 = -1,\ x_2 = 1/100$. The solution point is $x_1 = 1.5264,\ x_2 = 0.5763$ and the corresponding minimum value of the objective function is $F = -0.448910$.

b) Problem specification (input file):

```
$SET(INPUT)
  X(1)=-1.D 0 ;                IX(1)=0
  X(2)= 1.D-2 ; XL(2)= 1.D-2 ; IX(2)=1
                CL(1)=-5.D-1 ; IC(1)=1
  CG(1)=5.D-2 ; CG(2)=-1.D 0
$ENDSET
$SET(FMODELA)
  IF (KA.EQ.1) FA=-EXP(X(1)-X(2))
  IF (KA.EQ.2) FA= SINH(X(1)-1.D0)-1.D0
  IF (KA.EQ.3) FA=-LOG(X(2))-1.D0
$ENDSET
$MODEL='AM'
$IEXT=-1
$NF=2
$NA=3
$NC=1
$NCL=1
$KBF=1
$KBC=1
```

```
$MOUT=2
$NOUT=1
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables and types and values of both the box constraints and the general linear constraints. Since there are only one-sided constraints, we specify only the left sides (XL(2) and CL(1)) and we can set $KBF=1 and $KBC=1. By using the macrovariable $FMODELA we specify analytically the values of the approximating functions. The gradients of the approximating functions are computed numerically. For minimax approximation we set $MODEL='AM' and $IEXT=-1.

d) Problem solution (basic screen output):

```
         NIT=    0   NFV=    3   NFG=    0   F=  .361D+01   G=  .100D+61
         NIT=    1   NFV=    6   NFG=    0   F=  .198D+01   G=  .363D+00
         NIT=    2   NFV=    9   NFG=    0   F=  .682D+00   G=  .487D+00
         NIT=    3   NFV=   12   NFG=    0   F= -.277D+00   G=  .595D+00
         NIT=    4   NFV=   15   NFG=    0   F= -.396D+00   G=  .429D+00
         NIT=    5   NFV=   18   NFG=    0   F= -.426D+00   G=  .973D-01
         NIT=    6   NFV=   21   NFG=    0   F= -.449D+00   G=  .294D-02
         NIT=    7   NFV=   24   NFG=    0   F= -.449D+00   G=  .129D-04
  0  NIT=   8  NFV=  24  NFG=   0  NDC=   0  NCG=   0  F=-.449D+00  G= .263D-09
 F =  -.4489107840D+00
 X =   .1526434617D+01   .5763217308D+00
```

6.7. Optimization with nonlinear constraints (nonlinear programming)

a) Problem description:

Suppose we have to find a maximum of the objective function

$$F(x) = x_1 x_3$$

over the set given by the simple bounds $x_1 \geq 0,\ x_3 \geq 0,\ x_5 \geq 0,\ x_7 \geq 0$ and by the nonlinear constraints

$$(x_4 - x_6)^2 + (x_5 - x_7)^2 \geq 4$$

$$\frac{x_3 x_4 - x_2 x_5}{\sqrt{x_2^2 + x_3^2}} \geq 1$$

$$\frac{x_3 x_6 - x_2 x_7}{\sqrt{x_2^2 + x_3^2}} \geq 1$$

$$\frac{x_1 x_3 + (x_2 - x_1) x_5 - x_3 x_4}{\sqrt{(x_2 - x_1)^2 + x_3^2}} \geq 1$$

$$\frac{x_1 x_3 + (x_2 - x_1) x_7 - x_3 x_6}{\sqrt{(x_2 - x_1) + x_3^2}} \geq 1$$

The starting point is $x_1 = 3.0,\ x_2 = 0.0,\ x_3 = 2.0,\ x_4 = -1.5,\ x_5 = 1.5,\ x_6 = 5.0,\ x_7 = 0.0$. The solution point is $x_1 = 4.828,\ x_2 = 0.000, x_3 = 4.828,\ x_4 = 1.000,\ x_5 = 2.414,\ x_6 = 2.414,\ x_7 = 1.000$ and the corresponding minimum value of the objective function is $F = 23.3137$.

b) Problem specification (input file):

```
$FLOAT W
$SET(INPUT)
  X(1)= 3.0D0 ; XL(1)= 0.0D0 ; IX(1)= 1
  X(2)= 0.0D0
  X(3)= 2.0D0 ; XL(3)= 0.0D0 ; IX(3)= 1
  X(4)=-1.5D0
  X(5)= 1.5D0 ; XL(5)= 1.0D0 ; IX(5)= 1
  X(6)= 5.0D0
  X(7)= 0.0D0 ; XL(7)= 1.0D0 ; IX(7)= 1
  CL(1)=4.0D0 ; IC(1)= 1
  CL(2)=1.0D0 ; IC(2)= 1
  CL(3)=1.0D0 ; IC(3)= 1
  CL(4)=1.0D0 ; IC(4)= 1
  CL(5)=1.0D0 ; IC(5)= 1
$ENDSET
$SET(FMODELF)
  FF=X(1)*X(3)
$ENDSET
$SET(FMODELC)
  IF (KC.LE.0) THEN
  ELSE IF (KC.EQ.1) THEN
    FC=(X(4)-X(6))**2+(X(5)-X(7))**2
  ELSE IF (KC.EQ.2) THEN
    W=SQRT(X(2)**2+X(3)**2)
    FC=(X(3)*X(4)-X(2)*X(5))/W
  ELSE IF (KC.EQ.3) THEN
    W=SQRT(X(2)**2+X(3)**2)
    FC=(X(3)*X(6)-X(2)*X(7))/W
  ELSE IF (KC.EQ.4) THEN
    W=SQRT((X(2)-X(1))**2+X(3)**2)
    FC=(X(1)*X(3)+(X(2)-X(1))*X(5)-X(3)*X(4))/W
  ELSE IF (KC.EQ.5) THEN
    W=SQRT((X(2)-X(1))**2+X(3)**2)
    FC=(X(1)*X(3)+(X(2)-X(1))*X(7)-X(3)*X(6))/W
  ENDIF
$ENDSET
$NF=7
$NC=5
$NCL=0
$KBF=1
$KBC=1
$MOUT=2
$NOUT=1
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify initial values and simple bounds for variables and types and values of the general constraints. Since there are only one-sided simple bounds and one-sided general constraints, we set $KBF=1 and $KBC=1. By using the macrovariable $FMODELF we specify analytically the value of the model function. The gradient of the model function is computed numerically.

d) Problem solution (basic screen output):

```
     NIC=  0 NIT=   0 NFV=    8 NFG=    0 F=  .600D+01 C=  .294D+01 G=  .000D+00
     NIC=  0 NIT=   1 NFV=   19 NFG=    0 F=  .332D+02 C=  .961D+00 G=  .267D+01
     NIC=  0 NIT=   2 NFV=   31 NFG=    0 F=  .290D+02 C=  .807D-01 G=  .107D+01
     NIC=  0 NIT=   3 NFV=   43 NFG=    0 F=  .247D+02 C=  .168D-01 G=  .104D+01
     NIC=  0 NIT=   4 NFV=   55 NFG=    0 F=  .237D+02 C=  .356D-01 G=  .704D+00
     NIC=  0 NIT=   5 NFV=   68 NFG=    0 F=  .235D+02 C=  .472D-01 G=  .949D+00
     NIC=  0 NIT=   6 NFV=   81 NFG=    0 F=  .233D+02 C=  .104D-02 G=  .240D+00
     NIC=  0 NIT=   7 NFV=   94 NFG=    0 F=  .233D+02 C=  .238D-04 G=  .795D-01
     NIC=  0 NIT=   8 NFV=  107 NFG=    0 F=  .233D+02 C=  .394D-05 G=  .288D-01
     NIC=  0 NIT=   9 NFV=  120 NFG=    0 F=  .233D+02 C=  .129D-04 G=  .194D-01
     NIC=  0 NIT=  10 NFV=  133 NFG=    0 F=  .233D+02 C=  .672D-07 G=  .885D-03
     NIC=  0 NIT=  11 NFV=  145 NFG=    0 F=  .233D+02 C=  .187D-08 G=  .228D-03
     NIC=  0 NIT=  12 NFV=  158 NFG=    0 F=  .233D+02 C=  .581D-12 G=  .203D-05
   0 NIC=  0 NIT=  13 NFV=  158 NFG=    0 F=  .233D+02 C=  .581D-12 G=  .416D-06
 FF=   .2331370850D+02
 X =   .4828427080D+01   .3557932046D-06   .4828427170D+01   .1000000178D+01
       .2414213628D+01   .2414213675D+01   .1000000000D+01
```

6.8. Global optimization

a) Problem description:

Suppose we have to find a global minimum of the objective function

$$F(x) = (x_1 - 3)^2(x_1 + 5)^2 + (x_2 - 2)^2(x_2 + 3)^2 - x_1^2 x_2^2$$

over the set given by the inequalities $-12 \leq x_1 \leq 10$ and $-12 \leq x_2 \leq 10$. The starting point is $x_1 = 0$, $x_2 = 0$. The solution point is $x_1 = -7.3300$, $x_2 = -6.4475$ and the global minimum value of the objective function is $F = -806.077$.

b) Problem specification (input file):

```
$SET(INPUT)
  XL(1)=-12.D0 ; XU(1)=10.D0
  XL(2)=-12.D0 ; XU(2)=10.D0
$ENDSET
$SET(FMODELF)
  FF=((X(1)-3.D0)*(X(1)+5.D0))**2+ &
     ((X(2)-2.D0)*(X(2)+3.D0))**2-(X(1)*X(2))**2
$ENDSET
$NF=2
$MOUT=1
$EXTREM='G'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify bounds defining the investigated region. By using the macrovariable $FMODELF we specify analytically the value of the model function. The gradient of the model function is computed numerically. Since we require to find a global minimum we set $EXTREM='G'.

d) Problem solution (basic screen output):

```
   0  NIT=    55  NFV=   657  NEX=   4  F= -.806D+03

   1.EXTREM : F =  -.8060772623D+03
              X =  -.7329989894D+01   -.6447506450D+01

   2.EXTREM : F =  -.3072281498D+03
              X =  -.6228926481D+01    .4363683136D+01

   3.EXTREM : F =  -.1504539067D+03
              X =   .3836710563D+01   -.4317610586D+01

   4.EXTREM : F =  -.5795091449D+02
              X =   .3368245241D+01    .2827173198D+01
```

6.9. Large scale optimization (sparse Hessian matrix)

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \sum_{i=1}^{n} \left((3 - 2x_i)x_i - x_{i-1} - x_{i+1} + 1\right)^2, \quad x_{n+1} = x_0 = 0$$

where $n = 100$. The starting point is $x_i = -1$ for $1 \leq i \leq n$. The minimum value of the objective function is $F = 0.0$

b) Problem specification (input file):

```
$FLOAT A
$SET(INPUT)
  DO 1 I=1,NF
    X(I)=-1.0D0
    J=2*(I-1)+1
    IH(I)=J
    JH(J)=I
    JH(J+1)=I+1
1 CONTINUE
  IH(NF+1)=2*NF
$ENDSET
$SET(FMODELF)
  FF=0.0D0
  DO 2 J=1,NF
    A=(3.0D0-2.0D0*X(J))*X(J)+1.0D0
    IF (J.GT.1) A=A-X(J-1)
```

```
    IF (J.LT.NF) A=A-X(J+1)
    FF=FF+A*A
2 CONTINUE
$ENDSET
$SET(GMODELF)
  GF(1)=0.0D0
  DO 3 J=1,NF
    A=(3.0D0-2.0D0*X(J))*X(J)+1.0D0
    IF (J.GT.1) A=A-X(J-1)
    IF (J.LT.NF) A=A-X(J+1)
    A=A+A
    GF(J)=GF(J)+A*(3.0D0-4.0D0*X(J))
    IF (J.GT.1) GF(J-1)=GF(J-1)-A
    IF (J.LT.NF) GF(J+1)=-A
3 CONTINUE
$ENDSET
$NF=100
$M=500
$MOUT=2
$HESF='S'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables and the sparsity pattern of the Hessian matrix. The sparse Hessian matrix, indicated by the statement $HESF='S', is tridiagonal so that the number of its upper half nonzero elements is 2*NF−1=199. We set $M=500, since a greater space is needed for sparse matrix processing. By using the macrovariable $FMODELF we specify analytically the value of the model function. By using the macrovariable $GMODELF we specify analytically the gradient of the model function.

d) problem solution (basic screen output):

```
         NIT=    0   NFV=    1   NFG=    4   F=  .410D+03   G=  .380D+02
         NIT=    1   NFV=    2   NFG=    8   F=  .513D+02   G=  .123D+02
         NIT=    2   NFV=    3   NFG=   12   F=  .616D+01   G=  .694D+01
         NIT=    3   NFV=    4   NFG=   16   F=  .241D+00   G=  .141D+01
         NIT=    4   NFV=    5   NFG=   20   F=  .760D-03   G=  .122D+00
         NIT=    5   NFV=    6   NFG=   24   F=  .244D-05   G=  .636D-02
         NIT=    6   NFV=    7   NFG=   28   F=  .390D-07   G=  .822D-03
         NIT=    7   NFV=    8   NFG=   32   F=  .106D-09   G=  .481D-04
         NIT=    8   NFV=    9   NFG=   36   F=  .539D-12   G=  .401D-05
         NIT=    9   NFV=   10   NFG=   40   F=  .387D-14   G=  .271D-06
  0  NIT=   9  NFV=  10  NFG=  40  NDC=   0  NCG=   2  F= .387D-14  G= .271D-06
```

6.10. Large-scale optimization (sparse Jacobian matrix)

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \sum_{i=1}^{n} f_i^A(x)$$

where n=100 and

$$\begin{aligned} f_i^A(x) &= \left((3-2x_i)x_i - x_{i+1} + 1\right)^2 && , \ i = 1 \\ f_i^A(x) &= \left((3-2x_i)x_i - x_{i-1} - x_{i+1} + 1\right)^2 && , 2 \leq i \leq n-1 \\ f_i^A(x) &= \left((3-2x_i)x_i - x_{i-1} + 1\right)^2 && , \ i = n \end{aligned}$$

The starting point is $x_i = -1$ for $1 \leq i \leq n$. The minimum value of the objective function is $F = 0.0$ (This problem is equivalent to the previous problem).

b) Problem specification (input file):

```
$FLOAT A
$SET(INPUT)
  DO 1 I=1,NF
    X(I)=-1.0D0
1 CONTINUE
  L=1
  DO 2 I=1,NF
    IAG(I)=L
    IF (I.GT.1) THEN
    JAG(L)=I-1
    L=L+1
    ENDIF
    JAG(L)=I
    L=L+1
    IF (I.LT.NF) THEN
    JAG(L)=I+1
    L=L+1
    ENDIF
2 CONTINUE
  IAG(NF+1)=L
$ENDSET
$SET(FMODELA)
  A=(3.0D0-2.0D0*X(KA))*X(KA)+1.0D0
  IF (KA.GT.1) A=A-X(KA-1)
  IF (KA.LT.NF) A=A-X(KA+1)
  FA=A*A
$ENDSET
$SET(GMODELA)
  A=(3.0D0-2.0D0*X(KA))*X(KA)+1.0D0
  IF (KA.GT.1) A=A-X(KA-1)
  IF (KA.LT.NF) A=A-X(KA+1)
  A=A+A
  GA(KA)=A*(3.0D0-4.0D0*X(KA))
  IF (KA.GT.1) GA(KA-1)=-A
  IF (KA.LT.NF) GA(KA+1)=-A
```

```
$ENDSET
$NF=100
$NA=100
$MA=300
$M=600
$MOUT=2
$MODEL='AF'
$JACA='S'
$HESF='B'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables and the sparsity pattern of the Jacobian matrix. The sparse Jacobian matrix, indicated by the statement $JACA='S', is tridiagonal and the number of its nonzero elements is 3*NF−2=298. Therefore, we set $MA=300. Since we use the partitioned Hessian matrix, indicated by the statement $HESF='B', we must specify the number of its nonzero elements (it is 6*NF−2). Therefore, we set $M=600. By using the macrovariable $FMODELA we specify analytically the values of the approximating functions. By using the macrovariable $GMODELA we specify analytically the gradients of the approximating functions. For the sum of values minimization we set $MODEL='AF'.

d) problem solution (basic screen output):

```
        NIT=    0   NFV=    1   NFG=    1   F=  .410D+03   G=  .380D+02
        NIT=    1   NFV=    2   NFG=    2   F=  .562D+02   G=  .887D+01
        NIT=    2   NFV=    3   NFG=    3   F=  .332D+02   G=  .563D+01
        NIT=    3   NFV=    5   NFG=    5   F=  .803D+01   G=  .672D+01
        NIT=    4   NFV=    7   NFG=    7   F=  .250D+01   G=  .753D+01
        NIT=    5   NFV=    9   NFG=    9   F=  .113D+01   G=  .617D+01
        NIT=    6   NFV=   10   NFG=   10   F=  .272D-01   G=  .730D+00
        NIT=    7   NFV=   11   NFG=   11   F=  .167D-03   G=  .647D-01
        NIT=    8   NFV=   12   NFG=   12   F=  .103D-05   G=  .365D-02
        NIT=    9   NFV=   13   NFG=   13   F=  .178D-07   G=  .834D-03
        NIT=   10   NFV=   14   NFG=   14   F=  .226D-09   G=  .981D-04
        NIT=   11   NFV=   15   NFG=   15   F=  .289D-11   G=  .995D-05
        NIT=   12   NFV=   16   NFG=   16   F=  .128D-12   G=  .216D-05
        NIT=   13   NFV=   17   NFG=   17   F=  .219D-15   G=  .782D-07
0  NIT=  13  NFV=  17  NFG=  17  NDC=   0  NCG=   4  F= .219D-15  G= .782D-07
```

6.11. Large-scale sum of squares optimization (sparse Jacobian matrix)

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \sum_{i=1}^{n} \left(f_i^A(x)\right)^2$$

where $n = 100$ and

$$f_i^A(x) \quad = \quad (3 - 2x_i)x_i - x_{i+1} + 1 \qquad\qquad , \; i = 1$$

$$
\begin{aligned}
f_i^A(x) &= (3-2x_i)x_i - x_{i-1} - x_{i+1} + 1 \quad , 2 \le i \le n-1 \\
f_i^A(x) &= (3-2x_i)x_i - x_{i-1} + 1 \quad , \; i = n
\end{aligned}
$$

The starting point is $x_i = -1$ for $1 \le i \le n$. The minimum value of the objective function is $F = 0.0$ (This problem is equivalent to the previous problem).

b) Problem specification (input file):

```
$SET(INPUT)
  DO 1 I=1,NF
    X(I)=-1.0D0
1 CONTINUE
  L=1
  DO 2 I=1,NA
    IAG(I)=L
    IF (I.GT.1) THEN
    JAG(L)=I-1
    L=L+1
    ENDIF
    JAG(L)=I
    L=L+1
    IF (I.LT.NA) THEN
    JAG(L)=I+1
    L=L+1
    ENDIF
2 CONTINUE
  IAG(NA+1)=L
$ENDSET
$SET(FMODELA)
  I=KA
  FA=(3.0D0-2.0D0*X(I))*X(I)+1.0D0
  IF (I.GT.1) FA=FA-X(I-1)
  IF (I.LT.NA) FA=FA-X(I+1)
$ENDSET
$SET(GMODELA)
  I=KA
  GA(I)=3.0D0-4.0D0*X(I)
  IF (I.GT.1) GA(I-1)=-1.0D0
  IF (I.LT.NA) GA(I+1)=-1.0D0
$ENDSET
$NF=100
$NA=100
$MA=300
$M=600
$MOUT=2
$MODEL='AQ'
$JACA='S'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables and the sparsity pattern of the Jacobian matrix. The sparse Jacobian matrix, indicated by the statement $JACA='S', is tridiagonal and the number of its nonzero elements is 3*NF−2=298. Therefore, we set $MA=300. Since we do not use the sparse Hessian matrix, we do not specify the number of its nonzero elements. By using the macrovariable $FMODELA we specify analytically the values of the approximating functions. By using the macrovariable $GMODELA we specify analytically the gradients of the approximating functions. For the sum of squares minimization we set $MODEL='AQ'.

d) problem solution (basic screen output):

```
        NIT=    0   NFV=    1   NFG=    1   F=  .205D+03   G=  .190D+02
        NIT=    1   NFV=    2   NFG=    2   F=  .636D+01   G=  .230D+01
        NIT=    2   NFV=    3   NFG=    3   F=  .228D-01   G=  .118D+00
        NIT=    3   NFV=    4   NFG=    4   F=  .487D-06   G=  .532D-03
        NIT=    4   NFV=    5   NFG=    5   F=  .235D-15   G=  .128D-07
  0  NIT=   4  NFV=   5  NFG=   5  NDC=   4  NCG=   0  F= .235D-15  G= .128D-07
```

6.12. Large-scale nonlinear equations

a) Problem description:

Suppose we have to solve a system of the nonlinear equations

$$\begin{aligned} f_i^A(x) &= (3-2x_i)x_i - x_{i+1} + 1 = 0 && , i = 1 \\ f_i^A(x) &= (3-2x_i)x_i - x_{i-1} - x_{i+1} + 1 = 0 && , 2 \le i \le n-1 \\ f_i^A(x) &= (3-2x_i)x_i - x_{i-1} + 1 = 0 && , i = n \end{aligned}$$

where n=100. The starting point is $x_i = -1$ for $1 \le i \le n$. The minimum value of the objective function is $F = 0.0$ (This problem is equivalent to the previous problem).

b) Problem specification (input file):

```
$SET(INPUT)
  DO 1 I=1,NF
    X(I)=-1.0D0
1 CONTINUE
$ENDSET
$SET(FMODELA)
  I=KA
  FA=(3.0D0-2.0D0*X(I))*X(I)+1.0D0
  IF (I.GT.1) FA=FA-X(I-1)
  IF (I.LT.NA) FA=FA-X(I+1)
$ENDSET
$NF=100
$NA=100
$MOUT=2
$MODEL='AQ'
$JACA='NO'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables. By using the macrovariable $FMODELA we specify analytically the values of functions in the nonlinear equations. For solving nonlinear equations we set $MODEL='AQ'.

d) problem solution (basic screen output):

```
         NIT=    0   NFV=    1   F=  .205D+03
         NIT=    1   NFV=    7   F=  .526D+01
         NIT=    2   NFV=   14   F=  .163D-01
         NIT=    3   NFV=   21   F=  .250D-06
         NIT=    4   NFV=   28   F=  .620D-16
   0  NIT=     4  NFV=    28  NDC=    4  NCG=    1  F= .620D-16
```

6.13. Large-scale linear programming

a) Problem description:

Suppose we have to find a maximum of the linear function

$$F(x) = \sum_{i=1}^{n} (-1)^i x_i$$

with simple bounds $-20 \leq x_i \leq 20,\ 1 \leq x_i \leq n$, and linear constraints

$$-x_i + x_{i+1} - x_{i+2} = i,\ \ 1 \leq i \leq n_C$$

where $n = 20$ and $n_C = 18$. The starting point is not given. The maximum value of the linear objective function is $F = 7.0$

b) Problem specification (input file):

```
$SET(INPUT)
  DO 1 I=1,NF
    IX(I)=3
    XL(I)=-2.0D1
    XU(I)= 2.0D1
    GF(I)=FLOAT((-1)**I)
1 CONTINUE
  DO 2 KC=1,NC
    IC(KC)=5
    CL(KC)=FLOAT(KC)
    CALL UKMCI1(KC,KC,-1.0D0,ICG,JCG,CG)
    CALL UKMCI1(KC,KC+1,1.0D0,ICG,JCG,CG)
    CALL UKMCI1(KC,KC+2,-1.0D0,ICG,JCG,CG)
2 CONTINUE
$ENDSET
$IEXT=1
$NF=20
$NC=18
$NCL=18
$MC=200
$KBF=2
```

```
$KBC=1
$MOUT=2
$NOUT=1
$MODEL='FL'
$JACC='S'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable \$INPUT we specify bounds for variables and the sparsity pattern with numerical values of the constraint Jacobian matrix. We use the procedure UKMCI1. The sparse Jacobian matrix, indicated by the statement \$JACC='S', is tridiagonal and the number of its nonzero elements is 3*(NF−2)=54. We set \$MC=200 as a sufficiently large dimension for auxiliary fields. The option \$MODEL='FL' indicates the linear programming problem.

d) Problem solution (basic screen output):

```
 NUMITR=    1 IJNEW=   20 IJOLD=   15 KINP=   0 IU=   48 F= .980D+04
 NUMITR=    2 IJNEW=   19 IJOLD=   20 KINP=   0 IU=   49 F= .208D+04
 NUMITR=    3 IJNEW=    0 IJOLD=   20 KINP=   0 IU=   49 F= .000D+00
  0 NUMITR=    3 NEL=    3 NREF=   1 KINP=   0 IU=   49 F= .000D+00 ITERL= 1
 NUMITR=    1 IJNEW=   15 IJOLD=   19 KINP=   0 IU=   49 F=-.900D+01
 NUMITR=    2 IJNEW=   20 IJOLD=   18 KINP=   0 IU=   48 F=-.700D+01
 NUMITR=    3 IJNEW=    0 IJOLD=   18 KINP=   0 IU=   48 F=-.700D+01
  0 NUMITR=    3 NEL=    3 NREF=   1 KINP=   0 IU=   48 F=-.700D+01 ITERL= 2
  0  NIT=   0  NFV=   0  NFG=   0  NDC=   0  NCG=   0  F= .700D+01  G= .000D+00
FF=   .7000000000D+01
X =  -.2000000000D+01   .0000000000D+00   .1000000000D+01  -.1000000000D+01
     -.5000000000D+01  -.8000000000D+01  -.8000000000D+01  -.6000000000D+01
     -.5000000000D+01  -.7000000000D+01  -.1100000000D+02  -.1400000000D+02
     -.1400000000D+02  -.1200000000D+02  -.1100000000D+02  -.1300000000D+02
     -.1700000000D+02  -.2000000000D+02  -.2000000000D+02  -.1800000000D+02
```

6.14. Large-scale quadratic programming

a) Problem description:

Suppose we have to find a minimum of the quadratic function

$$F(x) = \sum_{i=1}^{k-2} (x_{k+i+1} - x_{k+i})^2$$

with simple bounds $\alpha_i \le x_i \le \alpha_{i+1},\ 0.4(\alpha_{i+2} - \alpha_i) \le x_{k+i} \le 0.6(\alpha_{i+2} - \alpha_i),\ 1 \le i \le k-1,\ \alpha_k \le x_k \le \alpha_{k+1}$, and linear constraints

$$x_{k+i} - x_{i+1} + x_i = 0,\ 1 \le i \le k-1$$

where $\alpha_i = 1 + (101/100)^i,\ 1 \le i \le k+1$, and where $n = 2k - 1 = 41,\ n_C = k - 1 = 20$. The starting point is not given. The minimum value of the quadratic objective function is $F = 2.29133$.

b) Problem specification (input file):

```
$FLOAT WA,WB,WC
$SET(INPUT)
  WA=1.00D0; WB=2.01D0
  DO 1 I=1,NC
    J=I+NC+1
    WC=1.0D0+(1.01D0)**(I+1)
    IX(I)=3; XL(I)=WA; XU(I)=WB
    IX(J)=3; XL(J)=0.4D0*(WC-WA); XU(J)=0.6D0*(WC-WA)
    GF(I)=0.0D0
    GF(J)=0.0D0
    WA=WB; WB=WC
    IC(I)=5; CL(I)=0.0D0
    CALL UKMCI1(I,J,1.0D0,ICG,JCG,CG)
    CALL UKMCI1(I,I,1.0D0,ICG,JCG,CG)
    CALL UKMCI1(I,I+1,-1.0D0,ICG,JCG,CG)
    IH(I)=1
1 CONTINUE
    IX(NC+1)=3; XL(NC+1)=WA; XU(NC+1)=WB
    GF(NC+1)=1
    IH(NC+1)=1; IH(NC+2)=1
    K=NC+2
    DO 2 I=K,NF-1
    IH(I+1)=IH(I)+2
2 CONTINUE
  IH(NF+1)=IH(NF)+1
  J=1
  DO 3 I=K,NF
    JH(J)=I; JH(J+1)=I+1
    HF(J)=2.0D0; HF(J+1)=-2.0D0
    IF (I.EQ.K.OR.I.EQ.NF)  HF(J)=1.0D0
    J=J+2
3 CONTINUE
$ENDSET
$NF=41
$NC=20
$NCL=20
$MC=500
$M=100
$MCOLS=100
$MROWS=50
$KBF=2
$KBC=1
$MOUT=2
$NOUT=1
$MODEL='FQ'
$JACC='S'
$HESF='S'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify bounds for variables, the sparsity pattern with numerical values of the model Hessian matrix, and the sparsity pattern with numerical values of the constraint Jacobian matrix. We use the procedure UKMCI1. The sparse Hessian matrix, indicated by the statement $HESF='S', is very simple and the number of its upper half nonzero elements is 2*(N−NC)−3=39. We set M$=100 as a sufficiently large dimension for working fields. The sparse Jacobian matrix, indicated by the statement $JACC='S', is tridiagonal and the number of its nonzero elements is 3*NC=60. We set $MC=200 as a sufficiently large dimension for working fields. The option $MODEL='FQ' indicates the linear programming problem.

d) Problem solution (basic screen output):

```
NUMITR=    1 IJNEW=   33 IJOLD=   33 KINP=   0 IU=   39 F= .496D+00
NUMITR=    2 IJNEW=   36 IJOLD=   36 KINP=   0 IU=   39 F= .393D+00
NUMITR=    3 IJNEW=   35 IJOLD=   35 KINP=   0 IU=   39 F= .261D+00
NUMITR=    4 IJNEW=   39 IJOLD=   39 KINP=   0 IU=   39 F= .164D+00
NUMITR=    5 IJNEW=   31 IJOLD=   31 KINP=   0 IU=   39 F= .905D-01
NUMITR=    6 IJNEW=   26 IJOLD=   26 KINP=   0 IU=   39 F= .416D-01
NUMITR=    7 IJNEW=   37 IJOLD=   10 KINP=   0 IU=   39 F= .123D-01
NUMITR=    8 IJNEW=   28 IJOLD=   28 KINP=   0 IU=   39 F= .435D-02
NUMITR=    9 IJNEW=    0 IJOLD=   28 KINP=   0 IU=   39 F= .000D+00
 0 NUMITR=    9 NEL=    6 NREF=   1 KINP=   0 IU=   39 F= .000D+00 ITERL= 1
NAQ=  1 NIQ=    0 NSBSP=  14 NCGR=  0 IU=   39 F=  .2291583D+01 G= .686D-01
NAQ=  1 NIQ=    1 NSBSP=  13 NCGR=  3 IU=   39 F=  .2291563D+01 G= .683D-01
NAQ=  1 NIQ=    2 NSBSP=  12 NCGR=  6 IU=   39 F=  .2291553D+01 G= .668D-01
NAQ=  1 NIQ=    3 NSBSP=  11 NCGR= 10 IU=   39 F=  .2291537D+01 G= .668D-01
NAQ=  1 NIQ=    4 NSBSP=  10 NCGR= 14 IU=   39 F=  .2291512D+01 G= .455D-01
NAQ=  1 NIQ=    5 NSBSP=   9 NCGR= 19 IU=   38 F=  .2291501D+01 G= .291D-01
NAQ=  2 NIQ=    0 NSBSP=  14 NCGR= 19 IU=   37 F=  .2291496D+01 G= .436D-01
NAQ=  2 NIQ=    1 NSBSP=  13 NCGR= 24 IU=   37 F=  .2291495D+01 G= .435D-01
NAQ=  2 NIQ=    2 NSBSP=  12 NCGR= 30 IU=   37 F=  .2291495D+01 G= .432D-01
NAQ=  2 NIQ=    3 NSBSP=  11 NCGR= 36 IU=   37 F=  .2291495D+01 G= .429D-01
NAQ=  2 NIQ=    4 NSBSP=  10 NCGR= 41 IU=   36 F=  .2291482D+01 G= .434D-01
NAQ=  2 NIQ=    5 NSBSP=   9 NCGR= 47 IU=   38 F=  .2291360D+01 G= .125D-01
NAQ=  3 NIQ=    0 NSBSP=  10 NCGR= 47 IU=   38 F=  .2291359D+01 G= .131D-01
NAQ=  3 NIQ=    1 NSBSP=   9 NCGR= 50 IU=   38 F=  .2291354D+01 G= .108D-01
NAQ=  3 NIQ=    2 NSBSP=   8 NCGR= 53 IU=   38 F=  .2291348D+01 G= .621D-02
NAQ=  4 NIQ=    0 NSBSP=   9 NCGR= 53 IU=   38 F=  .2291348D+01 G= .621D-02
NAQ=  4 NIQ=    1 NSBSP=   8 NCGR= 55 IU=   40 F=  .2291345D+01 G= .373D-02
NAQ=  5 NIQ=    0 NSBSP=  10 NCGR= 55 IU=   40 F=  .2291345D+01 G= .437D-02
NAQ=  6 NIQ=    0 NSBSP=  11 NCGR= 55 IU=   40 F=  .2291345D+01 G= .411D-02
NAQ=  6 NIQ=    1 NSBSP=  11 NCGR= 56 IU=   40 F=  .2291344D+01 G= .396D-02
NAQ=  6 NIQ=    2 NSBSP=  10 NCGR= 57 IU=   40 F=  .2291343D+01 G= .318D-02
NAQ=  6 NIQ=    3 NSBSP=   9 NCGR= 58 IU=   40 F=  .2291342D+01 G= .253D-02
NAQ=  6 NIQ=    4 NSBSP=   8 NCGR= 59 IU=   41 F=  .2291342D+01 G= .246D-02
NAQ=  6 NIQ=    5 NSBSP=   7 NCGR= 61 IU=   41 F=  .2291341D+01 G= .233D-02
NAQ=  6 NIQ=    6 NSBSP=   6 NCGR= 67 IU=   41 F=  .2291341D+01 G= .197D-02
NAQ=  6 NIQ=    7 NSBSP=   5 NCGR= 72 IU=   40 F=  .2291339D+01 G= .638D-03
NAQ=  7 NIQ=    0 NSBSP=   7 NCGR= 72 IU=   40 F=  .2291339D+01 G= .194D-02
NAQ=  7 NIQ=    1 NSBSP=   6 NCGR= 77 IU=   39 F=  .2291335D+01 G= .190D-02
NAQ=  7 NIQ=    2 NSBSP=   5 NCGR= 82 IU=   39 F=  .2291335D+01 G= .186D-02
NAQ=  7 NIQ=    3 NSBSP=   4 NCGR= 86 IU=   36 F=  .2291334D+01 G= .149D-02
NAQ=  8 NIQ=    0 NSBSP=   6 NCGR= 86 IU=   36 F=  .2291334D+01 G= .178D-02
```

```
 NAQ=  8 NIQ=    1 NSBSP=   5 NCGR= 90 IU=   36 F=  .2291331D+01 G= .187D-02
 NAQ=  8 NIQ=    2 NSBSP=   4 NCGR= 93 IU=   36 F=  .2291331D+01 G= .944D-03
 NAQ=  8 NIQ=    3 NSBSP=   4 NCGR= 94 IU=   36 F=  .2291330D+01 G= .000D+00
  0 NAQ=  8 NIQ=    3 NSBSP=   4 NCGR= 94 IU=   36 F=  .2291330D+01 G= .000D+00
  0  NIT=   0  NFV=   0  NFG=   0  NDC=   0  NCG=  94  F= .229D+01  G= .000D+00
FF=   .2291330455D+01
X =   .1601960000D+01   .2010000000D+01   .2022180600D+01   .2034483006D+01
      .2046647093D+01   .2056885037D+01   .2065335158D+01   .2073869780D+01
      .2082856706D+01   .2093685273D+01   .2106259114D+01   .2119580857D+01
      .2133035817D+01   .2146625327D+01   .2160350732D+01   .2172205073D+01
      .2181910478D+01   .2191338011D+01   .2200859818D+01   .2210476844D+01
      .2220190040D+01   .4080400000D+00   .1218060000D-01   .1230240600D-01
      .1216408687D-01   .1023794428D-01   .8450120803D-02   .8534622011D-02
      .8986925666D-02   .1082856706D-01   .1257384126D-01   .1332174283D-01
      .1345496026D-01   .1358950986D-01   .1372540496D-01   .1185434141D-01
      .9705405081D-02   .9427532305D-02   .9521807628D-02   .9617025705D-02
      .9713195962D-02
```

6.15. Large-scale optimization with linear constraints

a) Problem description:

The problem we have solved is in fact the Hock and Schittkowski problem number 119 (see [44]) which has 16 variables and 8 linear constraints. The minimum value of the objective function is $F = 244.899$.

b) Problem specification (input field):

```
$FLOAT WI,WJ
$SET(INPUT)
  DO 1 I=1,NF
    X(I)=10.0D0; XL(I)=0.0D0; XU(I)=5.0D0; IX(I)=3
1 CONTINUE
  IH( 1)= 1; IH( 2)= 6; IH( 3)=10; IH( 4)=15; IH( 5)=19
  IH( 6)=24; IH( 7)=27; IH( 8)=30; IH( 9)=33; IH(10)=36
  IH(11)=38; IH(12)=40; IH(13)=42; IH(14)=44; IH(15)=45
  IH(16)=46; IH(17)=47;
  JH( 1)= 1; JH( 2)= 4; JH( 3)= 7; JH( 4)= 8; JH( 5)=16
  JH( 6)= 2; JH( 7)= 3; JH( 8)= 7; JH( 9)=10;
  JH(10)= 3; JH(11)= 7; JH(12)= 9; JH(13)=10; JH(14)=14
  JH(15)= 4; JH(16)= 7; JH(17)=11; JH(18)=15;
  JH(19)= 5; JH(20)= 6; JH(21)=10; JH(22)=12; JH(23)=16
  JH(24)= 6; JH(25)= 8; JH(26)=15;
  JH(27)= 7; JH(28)=11; JH(29)=13;
  JH(30)= 8; JH(31)=10; JH(32)=15;
  JH(33)= 9; JH(34)=12; JH(35)=16;
  JH(36)=10; JH(37)=14;
  JH(38)=11; JH(39)=13;
  JH(40)=12; JH(41)=14;
  JH(42)=13; JH(43)=14;
  JH(44)=14;
  JH(45)=15;
  JH(46)=16;
```

```
      DO 2 I=1,NC
        IC(I)=5
    2 CONTINUE
      CL(1)= 2.5D0
      CL(2)= 1.1D0
      CL(3)=-3.1D0
      CL(4)=-3.5D0
      CL(5)= 1.3D0
      CL(6)= 2.1D0
      CL(7)= 2.3D0
      CL(8)=-1.5D0
      CALL UKMCI1(1, 1, 0.22D0,ICG,JCG,CG)
      CALL UKMCI1(1, 2, 0.20D0,ICG,JCG,CG)
      CALL UKMCI1(1, 3, 0.19D0,ICG,JCG,CG)
      CALL UKMCI1(1, 4, 0.25D0,ICG,JCG,CG)
      CALL UKMCI1(1, 5, 0.15D0,ICG,JCG,CG)
      CALL UKMCI1(1, 6, 0.11D0,ICG,JCG,CG)
      CALL UKMCI1(1, 7, 0.12D0,ICG,JCG,CG)
      CALL UKMCI1(1, 8, 0.13D0,ICG,JCG,CG)
      CALL UKMCI1(1, 9, 1.00D0,ICG,JCG,CG)
      CALL UKMCI1(2, 1,-1.46D0,ICG,JCG,CG)
      CALL UKMCI1(2, 3,-1.30D0,ICG,JCG,CG)
      CALL UKMCI1(2, 4, 1.82D0,ICG,JCG,CG)
      CALL UKMCI1(2, 5,-1.15D0,ICG,JCG,CG)
      CALL UKMCI1(2, 7, 0.80D0,ICG,JCG,CG)
      CALL UKMCI1(2,10, 1.00D0,ICG,JCG,CG)
      CALL UKMCI1(3, 1, 1.29D0,ICG,JCG,CG)
      CALL UKMCI1(3, 2,-0.89D0,ICG,JCG,CG)
      CALL UKMCI1(3, 5,-1.16D0,ICG,JCG,CG)
      CALL UKMCI1(3, 6,-0.96D0,ICG,JCG,CG)
      CALL UKMCI1(3, 8,-0.49D0,ICG,JCG,CG)
      CALL UKMCI1(3,11, 1.00D0,ICG,JCG,CG)
      CALL UKMCI1(4, 1,-1.10D0,ICG,JCG,CG)
      CALL UKMCI1(4, 2,-1.06D0,ICG,JCG,CG)
      CALL UKMCI1(4, 3, 0.95D0,ICG,JCG,CG)
      CALL UKMCI1(4, 4,-0.54D0,ICG,JCG,CG)
      CALL UKMCI1(4, 6,-1.78D0,ICG,JCG,CG)
      CALL UKMCI1(4, 7,-0.41D0,ICG,JCG,CG)
      CALL UKMCI1(4,12, 1.00D0,ICG,JCG,CG)
      CALL UKMCI1(5, 4,-1.43D0,ICG,JCG,CG)
      CALL UKMCI1(5, 5, 1.51D0,ICG,JCG,CG)
      CALL UKMCI1(5, 6, 0.59D0,ICG,JCG,CG)
      CALL UKMCI1(5, 7,-0.33D0,ICG,JCG,CG)
      CALL UKMCI1(5, 8,-0.43D0,ICG,JCG,CG)
      CALL UKMCI1(5,13, 1.00D0,ICG,JCG,CG)
      CALL UKMCI1(6, 2,-1.72D0,ICG,JCG,CG)
      CALL UKMCI1(6, 3,-0.33D0,ICG,JCG,CG)
      CALL UKMCI1(6, 5, 1.62D0,ICG,JCG,CG)
      CALL UKMCI1(6, 6, 1.24D0,ICG,JCG,CG)
      CALL UKMCI1(6, 7, 0.21D0,ICG,JCG,CG)
      CALL UKMCI1(6, 8,-0.26D0,ICG,JCG,CG)
      CALL UKMCI1(6,14, 1.00D0,ICG,JCG,CG)
```

```
  CALL UKMCI1(7, 1, 1.12D0,ICG,JCG,CG)
  CALL UKMCI1(7, 4, 0.31D0,ICG,JCG,CG)
  CALL UKMCI1(7, 7, 1.12D0,ICG,JCG,CG)
  CALL UKMCI1(7, 9,-0.36D0,ICG,JCG,CG)
  CALL UKMCI1(7,15, 1.00D0,ICG,JCG,CG)
  CALL UKMCI1(8, 2, 0.45D0,ICG,JCG,CG)
  CALL UKMCI1(8, 3, 0.26D0,ICG,JCG,CG)
  CALL UKMCI1(8, 4,-1.10D0,ICG,JCG,CG)
  CALL UKMCI1(8, 5, 0.58D0,ICG,JCG,CG)
  CALL UKMCI1(8, 7,-1.03D0,ICG,JCG,CG)
  CALL UKMCI1(8, 8, 0.10D0,ICG,JCG,CG)
  CALL UKMCI1(8,16, 1.00D0,ICG,JCG,CG)
$ENDSET
$SET(FGMODELF)
  FF=0.0D0
  DO 3 I=1,NF
    GF(I)=0.0D0
3 CONTINUE
  DO 5 I=1,NF
    WI=X(I)*(X(I)+1.0D0)+1.0D0
   K1=IH(I)
   K2=IH(I+1)-1
    DO 4 K=K1,K2
      J=JH(K)
      WJ=X(J)*(X(J)+1.0D0)+1.0D0
      FF=FF+WI*WJ
      GF(I)=GF(I)+(2.0D0*X(I)+1.0D0)*WJ
      GF(J)=GF(J)+WI*(2.0D0*X(J)+1.0D0)
  4 CONTINUE
5 CONTINUE
$ENDSET
$NF=16
$M=100
$NC=8
$NCL=8
$MC=200
$KBF=2
$KBC=1
$MOUT=2
$NOUT=1
$JACC='S'
$HESF='S'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify bounds for variables, the sparsity pattern with numerical values of the model Hessian matrix, and the sparsity pattern with numerical values of the constraint Jacobian matrix. We use the procedure UKMCI1. The sparse Hessian matrix is indicated by the statement $HESF='S'. The sparse Jacobian matrix is indicated by the statement $JACC='S'. The option $MODEL='FF' indicates a general objective function. By using the macrovariable $FGMODELF we

specify analytically the value and the gradient of the model function.

d) Problem solution (basic screen output):

```
         NIT=    0   NFV=    2   NFG=    2   F=  .421D+04   G=  .100D+01
         NIT=    1   NFV=    3   NFG=    3   F=  .325D+03   G=  .000D+00
         NIT=    2   NFV=    4   NFG=    4   F=  .252D+03   G=  .000D+00
         NIT=    3   NFV=    5   NFG=    5   F=  .246D+03   G=  .152D+02
         NIT=    4   NFV=    6   NFG=    6   F=  .245D+03   G=  .174D+01
         NIT=    5   NFV=    7   NFG=    7   F=  .245D+03   G=  .182D+01
         NIT=    6   NFV=    8   NFG=    8   F=  .245D+03   G=  .502D-01
         NIT=    7   NFV=    9   NFG=    9   F=  .245D+03   G=  .170D-01
         NIT=    8   NFV=   10   NFG=   10   F=  .245D+03   G=  .952D-04
         NIT=    9   NFV=   11   NFG=   11   F=  .245D+03   G=  .000D+00
  0  NIT=   9  NFV=  11  NFG=  11  NDC=   0  NCG=   0  F= .245D+03  G= .000D+00
 FF=   .2448996975D+03
 X =   .3984822768D-01   .7919832074D+00   .2028707289D+00   .8443603623D+00
       .1269907517D+01   .9347380182D+00   .1681960181D+01   .1553019234D+00
       .1567869438D+01  -.5551115123D-16   .0000000000D+00   .0000000000D+00
       .6602062241D+00   .0000000000D+00   .6742558679D+00   .0000000000D+00
```

6.16. Large-scale optimization with nonlinear equality constraints

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \sum_{i=1}^{n} \left(f_i^A(x)\right)^2$$

where $n = 100$ and

$$\begin{aligned} f_i^A(x) &= (3-2x_i)x_i - x_{i+1} + 1 && , i = 1 \\ f_i^A(x) &= (3-2x_i)x_i - x_{i-1} - x_{i+1} + 1 && , 2 \le i \le n-1 \\ f_i^A(x) &= (3-2x_i)x_i - x_{i-1} + 1 && , i = n \end{aligned}$$

over the set given by the nonlinear equality constraints

$$8x_i(x_i^2 - x_{i-1}) - 2(1-x_i) + 4(x_i - x_{i+1}^2) + x_{i-1}^2 - x_{i-2} + x_{i+1} - x_{i+2}^2 = 0,\ 3 \le i \le n-2$$

The starting point is $x_i = -1,\ 1 \le i \le n$. The minimum value of the objective function is $F = 5.29056$.

b) Problem specification (input file):

```
$FLOAT WA,WB
$SET(INPUT)
  DO 1 I=1,NF
  X(I)=-1.D0
1 CONTINUE
```

```
  M=0
  IH(1)=1
  DO 2 I=1,NF
  M=M+1
  JH(M)=I
  IF (I.LE.NF-1) THEN
  M=M+1
  JH(M)=I+1
  ENDIF
  IF (I.LE.NF-2) THEN
  M=M+1
  JH(M)=I+2
  ENDIF
  IH(I+1)=M+1
2 CONTINUE
  MC=0
  ICG(1)=1
  DO 3 I=3,NF-2
  MC=MC+1
  JCG(MC)=I-2
  MC=MC+1
  JCG(MC)=I-1
  MC=MC+1
  JCG(MC)=I
  MC=MC+1
  JCG(MC)=I+1
  MC=MC+1
  JCG(MC)=I+2
  ICG(I-1)=MC+1
3 CONTINUE
  DO 4 KC=1,NC
  IC(KC)=5
  CL(KC)=0.D0
4 CONTINUE
$ENDSET
$SET(FMODELF)
  FF=0.D0
  DO 5 J=1,NF
  WA=(3.D0-2.D0*X(J))*X(J)+1.D0
  IF (J.GT. 1) WA=WA-X(J-1)
  IF (J.LT.NF) WA=WA-X(J+1)
  FF=FF+WA**2
5 CONTINUE
$ENDSET
$SET(GMODELF)
  DO 6 J=1,NF
  GF(J)=0.D0
6 CONTINUE
  DO 7 J=1,NF
  WA=(3.D0-2.D0*X(J))*X(J)+1.D0
  IF (J.GT. 1) WA=WA-X(J-1)
  IF (J.LT.NF) WA=WA-X(J+1)
```

```
  WB=2.D0*WA
  GF(J)=GF(J)+WB*(3.D0-4.D0*X(J))
  IF (J.GT. 1) GF(J-1)=GF(J-1)-WB
  IF (J.LT.NF) GF(J+1)=GF(J+1)-WB
7 CONTINUE
$ENDSET
$SET(FMODELC)
  K=KC+2
  FC=8.D0*X(K)*(X(K)**2-X(K-1))-2.D0*(1.D0-X(K))+
 & 4.D0*(X(K)-X(K+1)**2)+X(K-1)**2-X(K-2)+X(K+1)-
 & X(K+2)**2
$ENDSET
$SET(GMODELC)
  K=KC+2
  GC(K-2)=-1.D0
  GC(K-1)=-8.D0*X(K)+2.D0*X(K-1)
  GC(K)=24.D0*X(K)**2-8.D0*X(K-1)+6.D0
  GC(K+1)=-8.D0*X(K+1)+1.D0
  GC(K+2)=-2.D0*X(K+2)
$ENDSET
$NF=100
$M=1500
$NC=96
$NCL=0
$MC=500
$KBC=1
$MOUT=2
$JACC='S'
$HESF='S'
$FORM='SE'
$FMIN=0
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify the initial values of variables, the sparsity pattern of the objective Hessian matrix, the sparsity pattern of the constraint Jacobian matrix, and the constraint specifications. The sparse Hessian matrix, indicated by the statement $HESF='S', is tridiagonal so that the number of its upper half nonzero elements is 2*NF−1=199. We set $M=1500, since a greater space is needed for sparse matrix processing. The sparse Jacobian matrix is indicated by the statement $JACC='S'. Since there are only the equality constraints, we can specify only the left sides CL(KC), $1 \leq \mathrm{KC} \leq \mathrm{NC}$, and we can set $KBC=1. The specification $FMIN=0 is used, since the objective function value cannot be less then zero. By using the macrovariable $FMODELF we specify analytically the value of the model function. By using the macrovariable $GMODELF we specify analytically the gradient of the model function. By using the macrovariable $FMODELC we specify analytically the values of the constraint functions. By using the macrovariable $GMODELC we specify analytically the gradients of the constraint functions. The choice $FORM='SE' correspond to inexact recursive quadratic programming methods for equality constrained problems.

d) problem solution (basic screen output):

```
     NIC=  0 NIT=   0 NFV=    1 NFG=   10 F=  .410D+03 C=  .280D+02 G=  .380D+02
     NIC=  0 NIT=   1 NFV=    2 NFG=   20 F=  .347D+03 C=  .881D+01 G=  .126D+02
     NIC=  0 NIT=   2 NFV=    3 NFG=   30 F=  .246D+03 C=  .286D+01 G=  .489D+01
     NIC=  0 NIT=   3 NFV=    4 NFG=   40 F=  .217D+03 C=  .127D+01 G=  .643D+01
     NIC=  0 NIT=   4 NFV=    5 NFG=   50 F=  .748D+02 C=  .844D+00 G=  .899D+01
     NIC=  0 NIT=   5 NFV=    7 NFG=   60 F= -.215D+00 C=  .592D+00 G=  .106D+02
     NIC=  0 NIT=   6 NFV=    8 NFG=   70 F=  .556D+01 C=  .575D-01 G=  .154D+01
     NIC=  0 NIT=   7 NFV=    9 NFG=   80 F=  .535D+01 C=  .459D-02 G=  .575D+00
     NIC=  0 NIT=   8 NFV=   10 NFG=   90 F=  .530D+01 C=  .392D-02 G=  .234D+00
     NIC=  0 NIT=   9 NFV=   11 NFG=  100 F=  .529D+01 C=  .111D-02 G=  .617D-01
     NIC=  0 NIT=  10 NFV=   12 NFG=  110 F=  .529D+01 C=  .247D-03 G=  .135D-01
     NIC=  0 NIT=  11 NFV=   13 NFG=  120 F=  .529D+01 C=  .364D-04 G=  .168D-02
     NIC=  0 NIT=  12 NFV=   14 NFG=  130 F=  .529D+01 C=  .863D-06 G=  .426D-04
     NIC=  0 NIT=  13 NFV=   15 NFG=  140 F=  .529D+01 C=  .630D-09 G=  .310D-07
   0 NIC=  0 NIT=  14 NFV=   15 NFG=  140 F=  .529D+01 C=  .630D-09 G=  .310D-07
```

6.17. Optimization of dynamical systems - general integral criterion

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \frac{1}{2}\int_0^T (y_1^2(t) + y_2^2(t))dt + \frac{1}{2}(y_1^2(T) + y_2^2(T))$$

where $T = 1.5$ and where

$$\frac{dy_1(t)}{dt} = y_2(t), \qquad y_1(0) = x_1$$

$$\frac{dy_2(t)}{dt} = (1 - y_1^2(t))y_2(t) - y_1(t), \quad y_2(0) = 1$$

b) Problem specification (input field):

```
$SET(INPUT)
  X(1)=ZERO
  TA=ZERO
  TAMAX=1.5D 0
$ENDSET
$SET(FMODELF)
  FF=HALF*(YA(1)**2+YA(2)**2)
$ENDSET
$SET(DMODELF)
  DF(1)=YA(1)
  DF(2)=YA(2)
$ENDSET
$SET(FMODELA)
  FA=HALF*(YA(1)**2+YA(2)**2)
$ENDSET
$SET(DMODELA)
  DA(1)=YA(1)
  DA(2)=YA(2)
$ENDSET
```

```
$SET(FMODELE)
  GO TO (1,2) KE
1 FE=YA(2)
  GO TO 3
2 FE=(ONE-YA(1)**2)*YA(2)-YA(1)
3 CONTINUE
$ENDSET
$SET(DMODELE)
  GO TO (4,5) KE
4 DE(1)=ZERO
  DE(2)=ONE
  GO TO 6
5 DE(1)=-ONE-TWO*YA(1)*YA(2)
  DE(2)=ONE-YA(1)**2
6 CONTINUE
$ENDSET
$SET(FMODELY)
  GO TO (7,8) KE
7 FE=X(1)
  GO TO 9
8 FE=ONE
9 CONTINUE
$ENDSET
$SET(GMODELY)
   GO TO (10,11) KE
10 GE(1)=ONE
   GO TO 12
11 GE(1)=ZERO
12 CONTINUE
$ENDSET
$NF=1
$NE=2
$MODEL='DF'
$MOUT=2
$NOUT=1
$TOLR='1.0$P-9'
$TOLA='1.0$P-9'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify as the initial value of the variable x_1 as the initial and terminal times 0 and T, respectively. By using the macrovariables $FMODELA and $DMODELA we specify subintegral function and by using the macrovariables $FMODELF and $DMODELF we specify terminal function. Right hand sides of the differential equations are specified by using the macrovariables $FMODELE and $DMODELE, while initial values and their derivatives are given by using the macrovariables $FMODELY and $GMODELY. The option $MODEL='DF' indicates general integral criterion.

d) Problem solution (basic screen output):

```
        NIT=    0   NFV=    1   NFG=    0   F=  .276D+01   G=  .242D+01
        NIT=    1   NFV=    3   NFG=    0   F=  .197D+01   G=  .513D+00
        NIT=    2   NFV=    4   NFG=    0   F=  .194D+01   G=  .468D-02
        NIT=    3   NFV=    5   NFG=    0   F=  .194D+01   G=  .122D-03
        NIT=    4   NFV=    6   NFG=    0   F=  .194D+01   G=  .205D-07
  0  NIT=   4  NFV=   6  NFG=   0  NDC=   0  NCG=   0  F= .194D+01  G= .205D-07
 FF=   .7671653645D+00
 X =   .6169838477D+00
```

6.18. Optimization of dynamical systems - special integral criterion

a) Problem description:

Suppose we have to find a minimum of the objective function

$$F(x) = \frac{1}{2}\int_0^T (y_1(t) - 1/(1+t))^2 dt$$

where $T = 1$ and where

$$\frac{dy_1(t)}{dt} = -x_1 y_1(t),\ y_1(0) = x_2$$

b) Problem specification (input field):

```
$SET(INPUT)
  X(1)=2.0D 0
  X(2)=0.0D 0
  TA=ZERO
  TAMAX=ONE
$ENDSET
$SET(FMODELE)
  FE=-X(1)*YA(1)**2
  YE=ONE/(ONE+TA)
  WE=ONE
$ENDSET
$SET(GDMODELE)
  GE(1)=-YA(1)**2
  GE(2)=ZERO
  DE(1)=-TWO*X(1)*YA(1)
$ENDSET
$SET(FMODELY)
  FE=X(2)
$ENDSET
$SET(GMODELY)
  GE(1)=ZERO
  GE(2)=ONE
$ENDSET
$MODELA='YES'
$NF=2
$NE=1
$MODEL='DQ'
$CLASS='GN'
```

```
$UPDATE='F'
$MOUT=2
$NOUT=1
$TOLR='1.0$P-9'
$TOLA='1.0$P-9'
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable $INPUT we specify as the initial values of the variables x_1 and x_2 as the initial and terminal times 0 and T, respectively. The right hand side of the differential equation is specified by using the macrovariables $FMODELE and $GDMODELE, while initial values and their derivatives are given by using the macrovariables $FMODELY and $GMODELY. The option $MODEL='DQ' together with $MODELA='YES' indicates special integral criterion.

d) Problem solution (basic screen output):

```
         NIT=    0   NFV=    1   NFG=    1   F=  .250D+00   G=  .693D+00
         NIT=    1   NFV=    3   NFG=    2   F=  .338D-01   G=  .114D+00
         NIT=    2   NFV=    5   NFG=    3   F=  .160D-02   G=  .613D-02
         NIT=    3   NFV=    7   NFG=    4   F=  .120D-04   G=  .225D-02
         NIT=    4   NFV=    9   NFG=    5   F=  .191D-08   G=  .300D-04
         NIT=    5   NFV=   11   NFG=    6   F=  .279D-15   G=  .200D-08
  0  NIT=   5  NFV=  11  NFG=   6  NDC=   7  NCG=   0  F= .279D-15  G= .200D-08
 F =   .2793083032D-15
 X =   .9999999725D+00   .9999999990D+00
```

6.19. Initial value problem for ordinary differential equations

a) Problem description:

Suppose we have to find a solution of the Van der Pol equation

$$\frac{dy_1(t)}{dt} = y_2(t), \qquad y_1(0) = 2$$

$$\frac{dy_1(t)}{dt} = (1 - y_1^2(t))y_2(t) - y_1(t), \quad y_2(0) = 0$$

in the interval $0 \leq t \leq T$ where $T = 20$.

b) Problem specification (input field):

```
$SET(INPUT)
  YA(1)=2.0D0
  YA(2)=0.0D0
  TA=0.0D0
  TAMAX=1.0D1
$ENDSET
$SET(FMODELE)
  IF (KE.EQ.1) THEN
    FE=YA(2)
  ELSE
```

```
    FE=(1.0D0-YA(1)**2)*YA(2)-YA(1)
  ENDIF
$ENDSET
$NA=21
$NE=2
$MODEL='NO'
$MED=2
$NOUT=1
$BATCH
$STANDARD
```

c) Comments on the problem specification:

By using the macrovariable \$INPUT we specify as the initial values of the variables y_1 and y_2 as the initial and terminal times 0 and T, respectively. Right hand sides of the differential equations are specified by using the macrovariable \$FMODELE. The option \$MODEL='NO' indicates integration of a system of ordinary differential equations.

d) Problem solution (basic screen output):

```
 0  NSTP=  47  NACC=  35  NREJ=  12  NEV =1228
   1  AT =   .0000000000D+00
      AY =   .2000000000D+01   .0000000000D+00
   2  AT =   .5000000000D+00
      AY =   .1837719210D+01  -.5345234547D+00
   3  AT =   .1000000000D+01
      AY =   .1508144241D+01  -.7802180795D+00
   4  AT =   .1500000000D+01
      AY =   .1040932817D+01  -.1124320556D+01
   5  AT =   .2000000000D+01
      AY =   .3233165976D+00  -.1832974600D+01
   6  AT =   .2500000000D+01
      AY =  -.8409663144D+00  -.2677481056D+01
   7  AT =   .3000000000D+01
      AY =  -.1866073894D+01  -.1021060289D+01
   8  AT =   .3500000000D+01
      AY =  -.1981111880D+01   .2780991398D+00
   9  AT =   .4000000000D+01
      AY =  -.1741768298D+01   .6246661652D+00
  10  AT =   .4500000000D+01
      AY =  -.1369680461D+01   .8749049693D+00
  11  AT =   .5000000000D+01
      AY =  -.8370774151D+00   .1307088938D+01
  12  AT =   .5500000000D+01
      AY =   .1492104708D-01   .2187559874D+01
  13  AT =   .6000000000D+01
      AY =   .1279043316D+01   .2437814560D+01
  14  AT =   .6500000000D+01
      AY =   .1981657547D+01   .3834607828D+00
  15  AT =   .7000000000D+01
      AY =   .1920152630D+01  -.4358385236D+00
  16  AT =   .7500000000D+01
```

```
     AY =    .1630052536D+01  -.7027644971D+00
17   AT =    .8000000000D+01
     AY =    .1213232754D+01  -.9878137411D+00
18   AT =    .8500000000D+01
     AY =    .5979891894D+00  -.1543326967D+01
19   AT =    .9000000000D+01
     AY =   -.4129152480D+00  -.2526902993D+01
20   AT =    .9500000000D+01
     AY =   -.1638546749D+01  -.1781241644D+01
21   AT =    .1000000000D+02
     AY =   -.2008340784D+01   .3290648928D-01
```

7. Model examples for demonstration of graphical output

Here we introduce several problem specifications (input files) which demonstrates application of graphical output. These input files are included to the UFO system as the demo-files PROC01.UFO,... ,PROC08.UFO. Corresponding grafical pictures are included in the appendix. The recommended data for graphical pictures are introduced in lines which begin by the directive $REM.

7.1. Nonlinear regression

```
$SET(INPUT)
  LDIM=5
  X(1)=7.0D20
  X(2)=1.0D4
  X(3)=2.2D0
  X(4)=1.01D0
  X(5)=7.0D17
  X(6)=7.0D3
  X(7)=1.6D0
  X(8)=1.01D0
  X(9)=1.0D16
  X(10)=4.0D3
  X(11)=1.5D0
  X(12)=1.01D0
  X(13)=2.0D15
  X(14)=4.0D3
  X(15)=1.3D0
  X(16)=1.01D0
  X(17)=1.0D16
  X(18)=5.0D2
  X(19)=1.2D0
  X(20)=1.01D0
  BETA=5.95D0
  CALL BIUD01(NF,LDIM,NA,X,XL,XU,IX,AT,AM)
$ENDSET
$SET(FMODELA)
  CALL BAFU01(NF,LDIM,KA,NA,X,AT,FA,BETA)
$ENDSET
$SET(GMODELA)
  CALL BAGU01(NF,LDIM,KA,NA,X,AT,GA,BETA)
$ENDSET
$NF=30
$NA=500
$KOUT=0
$KOUT1=0
$KOUT2=100
$KOUT3=1
$LOUT=0
$MOUT=2
$MIT=100
$MODEL='AQ'
$CLASS='GN'
$TYPE='G'
```

```
$DECOMP='M'
$NUMBER=7
$UPDATE='F'
$TOLX='1.0$P-16'
$TOLF='1.0$P-16'
$TOLB='1.0$P-16'
$TOLG='1.0$P-6'
$KBA=1
$KBF=2
$GRAPH='YES'
$SCAN='YES'
$BATCH
$ADD(REAL,'\BETA\AT($NA)')
$ADD(SUBROUTINES)
      SUBROUTINE BIUD01(N,L,NA,X,XL,XU,IX,AT,AM)
      INTEGER N,L,NA,IX(N),I,K
      REAL*8 X(N),XL(N),XU(N),AT(NA),AM(NA)
      N=4*L
      K=0
      DO 1 I=1,L
      X(K+1)=LOG(X(K+1))
      XL(K+1)=LOG(1.0D+0)
      XU(K+1)=LOG(1.0D+40)
      IX(K+1)=3
      X(K+2)=LOG(X(K+2))
      XL(K+2)=LOG(1.0D+0)
      XU(K+2)=LOG(1.0D+10)
      IX(K+2)=3
      XL(K+3)=1.0D-2
      XU(K+3)=1.0D+2
      IX(K+3)=3
      XL(K+4)=1.00001D0
      XU(K+4)=1.00000D1
      IX(K+4)=3
      K=K+4
    1 CONTINUE
      OPEN (11,FILE='PROC01.DAT',STATUS='OLD')
      NA=0
    2 NA=NA+1
      READ (11,'(2D14.6)',ERR=3) AT(NA),AM(NA)
      GO TO 2
    3 NA=NA-1
      RETURN
      END
      SUBROUTINE BAFU01(N,L,KA,NA,X,AT,FA,BETA)
      INTEGER N,L,KA,NA
      REAL*8 X(N),AT(NA),FA,Q(8),QD(8)
      REAL*8 ARG,POM,BK,B6INT,BETA
      INTEGER J,K
      COMMON /BCOM/ Q,QD
      DATA BK /8.617385D-5/
      FA=0.0D 0
```

```
      K=0
      DO 1 J=1,L
      ARG=X(K+3)/(BK*AT(KA))
      IF (KA.EQ.1) THEN
      Q(J)=B6INT(AT(KA),ARG)
      FA=FA+EXP(X(K+1)+X(K+2)-ARG)
      ELSE
      POM=X(K+4)-1.0D0
      FA=FA+EXP(X(K+1)+X(K+2)-ARG)*
     & (1.0D0+(POM/BETA)*EXP(X(K+1))*(B6INT(AT(KA),ARG)-
     & Q(J)))**(-X(K+4)/POM)
      ENDIF
      K=K+4
    1 CONTINUE
      RETURN
      END
      SUBROUTINE BAGU01(N,L,KA,NA,X,AT,GA,BETA)
      INTEGER N,L,KA,NA
      REAL*8 X(N),AT(NA),GA(N)
      REAL*8 FAC,ARG,POM,POW,BK,B6INT,B6INTD,A,B,C,D,E,F,G
      REAL*8 Q(8),QD(8),QQ,QQD,BETA
      INTEGER J,K
      COMMON /BCOM/ Q,QD
      DATA BK /8.617385D-5/
      K=0
      DO 1 J=1,L
      FAC=1.0D0/(BK*AT(KA))
      ARG=FAC*X(K+3)
      IF (KA.EQ.1) THEN
      Q(J)=B6INT(AT(KA),ARG)
      QD(J)=FAC*B6INTD(AT(KA),ARG)
      QQ=0.0D0
      QQD=0.0D0
      ELSE
      QQ=B6INT(AT(KA),ARG)-Q(J)
      QQD=FAC*B6INTD(AT(KA),ARG)-QD(J)
      ENDIF
      POM=X(K+4)-1.0D0
      POW=-X(K+4)/POM
      A=EXP(X(K+1)+X(K+2)-ARG)
      B=EXP(X(K+1))
      G=B*QQ
      C=(1.0D0+(POM/BETA)*G)
      D=C**POW
      E=POW*D/C
      F=POM*POM
      GA(K+1)=A*(D+E*(POM/BETA)*G)
      GA(K+2)=A*D
      GA(K+3)=A*(-FAC*D+E*(POM/BETA)*B*QQD)
      GA(K+4)=A*D*(LOG(C)/F+POW*G/(C*BETA))
      K=K+4
    1 CONTINUE
```

```
      RETURN
      END
      FUNCTION B6INT(T,X)
      REAL*8 T,X,B6INT
      REAL*8 A1,A2,A3,A4,A5,A6,B1,B2,B3,B4,B5,B6
      DATA A1,A2,A3,A4,A5,A6 /41.0D+0, 590.0D+0, 3648.0D+0,
     &                      9432.0D+0, 8028.0D+0, 720.0D+0/
      DATA B1,B2,B3,B4,B5,B6 /42.0D+0, 630.0D+0, 4200.0D+0,
     &                    12600.0D+0, 15120.0D+0, 5040.0D+0/
      B6INT=(1.0D0-(A6+X*(A5+X*(A4+X*(A3+X*(A2+X*(A1+X))))))/
     &     (B6+X*(B5+X*(B4+X*(B3+X*(B2+X*(B1+X)))))))*EXP(-X)*T
      RETURN
      END
      FUNCTION B6INTD(T,X)
      REAL*8 T,X,B6INTD
      REAL*8 A1,A2,A3,A4,A5,A6,B1,B2,B3,B4,B5,B6
      REAL*8 C1,C2,C3,C4,C5,D1,D2,D3,D4,D5,DIS,DEN,DISD,DEND
      DATA A1,A2,A3,A4,A5,A6 /41.0D+0, 590.0D+0, 3648.0D+0,
     &                      9432.0D+0, 8028.0D+0, 720.0D+0/
      DATA B1,B2,B3,B4,B5,B6 /42.0D+0, 630.0D+0, 4200.0D+0,
     &                    12600.0D+0, 15120.0D+0, 5040.0D+0/
      DATA C1,C2,C3,C4,C5 /205.0D+0, 2360.0D+0, 10944.0D+0,
     &                                18863.0D+0, 8028.0D+0/
      DATA D1,D2,D3,D4,D5 /210.0D+0, 2520.0D+0, 12600.0D+0,
     &                              25200.0D+0, 15120.0D+0/
      DIS=A6+X*(A5+X*(A4+X*(A3+X*(A2+X*(A1+X)))))
      DEN=B6+X*(B5+X*(B4+X*(B3+X*(B2+X*(B1+X)))))
      DISD=C5+X*(C4+X*(C3+X*(C2+X*(C1+6.0D0*X))))
      DEND=D5+X*(D4+X*(D3+X*(D2+X*(D1+6.0D0*X))))
      B6INTD=((DIS-DISD+DEND*DIS/DEN)/DEN-1.0D0)*EXP(-X)*T
      RETURN
      END
$ENDADD
$STANDARD
```

7.2. Nonlinear minimax optimization

```
$FLOAT W
$SET(INPUT)
  X(1)=0.5D0 ; X(2)=0.0D0 ; X(3)=0.0D0
  X(4)=0.0D0 ; X(5)=0.0D0
$ENDSET
$SET(FMODELA)
  W=0.1D0*DBLE(KA-1)-1.0D0
  FA=(X(1)+W*X(2))/(1.0D0+W*(X(3)+W*(X(4)+W*X(5))))-EXP(W)
$ENDSET
$MODEL='AM'
$NF=5
$NA=21
$NAL=0
$GRAPH='YES'
```

```
$MAP='YES'
$HIL='YES'
$ISO='YES'
$PATH='EXTENDED'
$BATCH
$STANDARD

$REM VAR=1, XL=-5, XU=5
$REM VAR=3, XL=-5, XU=5
```

7.3. Transformer network design

```
$SET(INPUT)
  NEXT=4
  CALL EIUD06(NF,NA,NAL,X,FMIN,XMAX,NEXT,IEXT,IERR)
$ENDSET
$SET(FMODELA)
  CALL EAFU06(NF,KA,X,FA,NEXT)
$ENDSET
$SET(GMODELA)
  CALL EAGU06(NF,KA,X,GA,NEXT)
$ENDSET
$NF=6
$NA=11
$NAL=0
$MOUT=1
$MODEL='AM'
$GRAPH='YES'
$MAP='YES'
$HIL='YES'
$ISO='YES'
$PATH='EXTENDED'
$BATCH
$STANDARD

$REM VAR=1, XL=-5, XU=5
$REM VAR=3, XL=-5, XU=5
```

7.4. Global optimization

```
$SET(INPUT)
  NEXT=4
  CALL EIUD09(NF,XL,XU,NEXT,IERR)
$ENDSET
$SET(FMODELF)
  CALL EFFU09(NF,X,FF,NEXT)
$ENDSET
$NF=4
$MOUT=1
$GCLASS=1
```

```
$GRAPH='YES'
$MAP='YES'
$HIL='YES'
$ISO='YES'
$EXTREM='G'
$BATCH
$STANDARD

$REM VAR=1, XL=-3.8, XU=3.8
$REM VAR=2, XL=-3.8, XU=3.8
```

7.5. Nonsmooth optimization

```
$SET(INPUT)
  NEXT=17
  CALL EIUD19(NF,X,FMIN,XMAX,NEXT,IEXT,IERR)
  MA=NF+3
$ENDSET
$SET(FMODELF)
  CALL EFFU19(NF,X,FF,NEXT)
$ENDSET
$SET(GMODELF)
  CALL EFGU19(NF,X,GF,NEXT)
$ENDSET
$KSF=3
$NF=30
$MOUT=-1
$MODEL='FF'
$GRAPH='YES'
$MAP='YES'
$HIL='YES'
$ISO='YES'
$PATH='YES'
$BATCH
$STANDARD

$REM VAR=1, XL=-5, XU=5
$REM VAR=4, XL=-5, XU=5
```

7.6. Nonlinear equations

```
$SET(INPUT)
  DO 1 I=1,NF
    X(I)=-1.0D0
1 CONTINUE
$ENDSET
$SET(FMODELA)
  I=KA
  FA=(3.0D0-2.0D0*X(I))*X(I)+1.0D0
  IF (I.GT.1) FA=FA-X(I-1)
```

```
  IF (I.LT.NA) FA=FA-X(I+1)
$ENDSET
$NF=100
$NA=100
$MOUT=1
$MODEL='AQ'
$JACA='NO'
$GRAPH='YES'
$BATCH
$STANDARD
```

7.7. Ordinary differential equations

```
$FLOAT W1,W2,W3,W4
$SET(INPUT)
  TA=0.0D0
  YA(1)=0.994D0
  YA(2)=0.0D0
  YA(3)=0.0D0
  YA(4)=-2.0015851063790825224053786224D0
  TAMAX=17.065216560157962558891720624 9D0
$ENDSET
$SET(FMODELE)
  W1=0.012277471D0
  W2=1.D0-W1
  W3=(YA(1)+W1)**2+YA(2)**2
  W3=W3*SQRT(W3)
  W4=(YA(1)-W2)**2+YA(2)**2
  W4=W4*SQRT(W4)
  GO TO (1,2,3,4) KE
1 FE=YA(3)
  GO TO 5
2 FE=YA(4)
  GO TO 5
3 FE=YA(1)+2*YA(4)-W2*(YA(1)+W1)/W3-W1*(YA(1)-W2)/W4
  GO TO 5
4 FE=YA(2)-2*YA(3)-W2*YA(2)/W3-W1*YA(2)/W4
5 CONTINUE
$ENDSET
$NE=4
$NA=2000
$MODEL='NO'
$SOLVER='DP5'
$MOUT=-1
$TOLR='1.0$P-9'
$TOLA='1.0$P-9'
$MED=1
$GRAPH='YES'
$BATCH
$STANDARD
```

7.8. The Lorenz attractor

```
$FLOAT W1,W2,W3
$SET(INPUT)
  W1=10.0D0
  W2=28.0D0
  W3=8.0D0/3.0D0
  TA=0.0D0
  YA(1)=-8.0D0
  YA(2)= 8.0D0
  YA(3)=W2-1.0D0
  TAMAX=50.0D0
$ENDSET
$SET(FMODELE)
  GO TO (1,2,3) KE
1 FE=-W1*YA(1)+W1*YA(2)
  GO TO 4
2 FE=-YA(1)*YA(3)+W2*YA(1)-YA(2)
  GO TO 4
3 FE=YA(1)*YA(2)-W3*YA(3)
4 CONTINUE
$ENDSET
$NE=3
$NA=2000
$MODEL='NO'
$SOLVER='DP8'
$MOUT=-1
$TOLR='1.0$P-9'
$TOLA='1.0$P-9'
$MED=1
$GRAPH='YES'
$BATCH
$STANDARD
```

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Index of macrovariables

Appendix A. Demonstration of the full dialogue mode

Suppose that the model function has the form

$$f^F(x) = 100(x_1^2 - x_2)^2 + (x_1 - 1)^2$$

(the Rosenbrock function) and the starting point is $x_1 = -1.2$ and $x_2 = 1.0$. If we type the statement UFOGO (without batch input file specification), then the following questions (which we supplement together with answers) appear on the screen.

```
UFO PREPROCESSOR    V.3.1.

                    ? INPUT ( ) ?
   USER SUPPLIED INPUT:

        HERE THE STARTING POINT, BOUNDS FOR VARIABLES,
        TYPES OF CONSTRAINTS, THE STRUCTURE OF SPARSE PROBLEM,
        AND OTHER INPUT DATA HAVE TO BE SPECIFIED.

X(1) = -1.2D0; X(2) = 1.0D0
                    ? GRAPH (NO) ?
   SPECIFICATION OF GRAPHICAL OUTPUT
      NO    - GRAPHICAL OUTPUT SUPPRESSED
      YES   - GRAPHICAL OUTPUT REQUIRED
                    ? DISPLAY (NO) ?
   SPECIFICATION OF EXTENDED SCREEN OUTPUT
      NO    - EXTENDED SCREEN OUTPUT SUPPRESSED
      YES   - EXTENDED SCREEN OUTPUT REQUIRED
                    ? MODEL (FF) ?
   TYPE OF OBJECTIVE FUNCTION
      FF - GENERAL FUNCTION
      FL - LINEAR FUNCTION
      FQ - QUADRATIC FUNCTION
      AF - SUM OF FUNCTIONS
      AQ - SUM OF SQUARES
      AP - SUM OF POWERS
      AM - MINIMAX
      DF - DIFFERENTIAL SYSTEM WITH GENERAL INTEGRAL CRITERION
      DQ - DIFFERENTIAL SYSTEM WITH INTEGRAL OF SQUARES
      NO - MODEL IS NOT SPECIFIED
                    ? NF (0) ?
   NUMBER OF VARIABLES

2
```

? IEXT (0) ?

TYPE OF EXTREMUM
0 - MINIMUM
1 - MAXIMUM

? FMODELF (*) ?

MODEL OF OBJECTIVE FUNCTION

FF = <FORTRAN_EXPRESSION>

FF = 1.0D2*(X(1)**2 - X(2))**2 + (X(1) - 1.0D0)**2

? GMODELF (*) ?

MODEL OF GRADIENT OF OBJECTIVE FUNCTION

GF(1) = <FORTRAN_EXPRESSION>
GF(2) = <FORTRAN_EXPRESSION>
.
.
GF(NF) = <FORTRAN_EXPRESSION>

? HMODELF (*) ?

MODEL OF HESSIAN MATRIX

HF(1) = <FORTRAN_EXPRESSION>
HF(2) = <FORTRAN_EXPRESSION>
.
.
HF(M) = <FORTRAN_EXPRESSION>

? KCF (2) ?

COMPLEXITY OF THE OBJECTIVE FUNCTION
1 - EASY COMPUTED FUNCTION
2 - REASONABLE BUT NOT EASY COMPUTED FUNCTION
3 - EXTREMELY COMPLICATED FUNCTION

? KSF (1) ?

SMOOTHNESS OF THE OBJECTIVE FUNCTION:
1 - SMOOTH AND WELL-CONDITIONED FUNCTION
2 - SMOOTH BUT ILL-CONDITIONED FUNCTION
3 - NONSMOOTH FUNCTION

? HESF (D) ?

TYPE OF HESSIAN MATRIX:
D - DENSE
S - SPARSE WITH KNOWN (GENERAL) STRUCTURE
NO - HESSIAN MATRIX IS NOT USED

? KBF (0) ?

TYPE OF SIMPLE BOUNDS:
0 - NO SIMPLE BOUNDS
1 - ONE SIDED SIMPLE BOUNDS
2 - TWO SIDED SIMPLE BOUNDS

? KBC (0) ?

TYPE OF GENERAL CONSTRAINTS:
0 - NO GENERAL CONSTRAINTS
1 - ONE SIDED GENERAL CONSTRAINTS
2 - TWO SIDED GENERAL CONSTRAINTS

? EXTREM (L) ?

TYPE OF OPTIMIZATION
L - LOCAL OPTIMIZATION
G - GLOBAL OPTIMIZATION

? NORMF (0) ?

SCALING SPECIFICATION FOR VARIABLES:
0 - NO SCALING IS PERFORMED
1 - SCALING FACTORS ARE DETERMINED AUTOMATICALLY
2 - SCALING FACTORS ARE SUPPLIED BY USER

? INPUTDATA (NO) ?

READ INPUT VALUES OF X (YES OR NO)

? TEST (NO) ?

STANDARD TEST OF EXTERNAL SUBROUTINES:
NO - NO TEST
YES - PERFORM TEST BEFORE SOLUTION
AFTER - PERFORM TEST AFTER SOLUTION
ONLY - PERFORM TEST WITHOUT SOLUTION

? KOUT (0) ?

LEVEL OF TEXT FILE OUTPUT:
ABS(KOUT)=0 - NO PRINT OR PAPER SAVING PRINT
ABS(KOUT)=1 - STANDARD PRINT OF ITERATIONS
ABS(KOUT)=2 - ADDITIONAL PRINT OF STEPSIZE SELECTION
ABS(KOUT)=3 - ADDITIONAL PRINT OF DIRECTION DETERMINATION
AND VARIABLE METRIC UPDATE
ABS(KOUT)=4 - ADDITINAL PRINT OF CONSTRAINT HANDLING
ABS(KOUT)=5 - ADDITIONAL PRINT OF NUMERICAL DIFFERENTIATION
KOUT<0 - ADDITIONAL PRINT OF DATA AND OPTIONS IN THE HEADING

```
                    ? LOUT (1) ?
  LEVEL OF TEXT FILE OUTPUT:
     0 - NO PRINT
     1 - COPY OF THE BASIC SCREEN OUTPUT
    -1 - PAPER SAVING PRINT

                    ? MOUT (1) ?
  LEVEL OF BASIC SCREEN OUTPUT:
     ABS(MOUT)=0 - NO OUTPUT
     ABS(MOUT)=1 - FINAL OUTPUT
     ABS(MOUT)=2 - ADDITIONAL OUTPUT IN EACH ITERATION
     ABS(MOUT)=3 - ADDITIONAL FINAL OUTPUT OF LINEAR OR
                   QUADRATIC PROGRAMMING
     ABS(MOUT)=4 - ADDITIONAL OUTPUT IN EACH ITERATION
                   OF LINEAR OR QUADRATIC PROGRAMMING
  MOUT<0 - FINAL OUTPUT WITH TERMINATION CRITERION

                    ? NOUT (0) ?
  LEVEL OF BASIC SCREEN OUTPUT:
     0 - BASIC FINAL OUTPUT
     1 - EXTENDED FINAL OUTPUT

1
                    ? MSELECT (1) ?
  SELECTION OF OPTIMIZATION METHOD
     1 - AUTOMATICAL SELECTION OF METHOD
     2 - MANUAL SELECTION OF METHOD
     3 - MANUAL SELECTION OF METHOD AND IMPORTANT PARAMETERS
     4 - MANUAL SELECTION OF METHOD AND ALL PARAMETERS

                    ? OUTPUT ( ) ?
  USER SUPPLIED OUTPUT:

        HERE THE RESULTS OBTAINED IN THE OPTIMIZATION PROCESS
        CAN BE USED FOR ADDITIONAL COMPUTATIONS AND FOR A
        SPECIFIC OUTPUT.

                  ? OUTPUTDATA (NO) ?
  WRITE OUTPUT VALUES OF X (YES OR NO)

UFO PREPROCESSOR    STOP
```

Each question is represented by one frame that contains the contents of the question (name of the macrovariable which has to be defined), the default value (in the brackets) and an explanation of the requirement. If no default value is wanted then the corresponding value or text has to be typed. The dialogue can be ended by pressing the key <! > .

The result of the UFO preprocessor action is the following control program (reported in a slightly shortened form) consisting of global declarations, input specifications, problem definition, method real-

ization and control variables adjustement:

```
*
*     ------------------
*     GLOBAL DECLARATIONS
*     ------------------
*
      INTEGER ITIME
      INTEGER IMD
      INTEGER IX(1)
      REAL*8 UXVDOT
      REAL*8 GF(2)
      REAL*8 X(2)
      REAL*8 HD(2)
      REAL*8 HF(2*(2+1)/2)
      REAL*8 S(2)
      REAL*8 ALF
      REAL*8 BET
      REAL*8 XO(2)
      REAL*8 GO(2)
      INTEGER IMB
*
*     commons placed here were omitted
*     since they require a large space
*
*
*     ------------------
*     END OF DECLARATIONS
*     ------------------
*
      CALL UYCLEA
      CALL UYINTP
*
*     ----------
*     METHOD (1)
*     ----------
*
      CALL UYINT1
      X(1)=-1.2D0
      X(2)=1.0D0
      M=NF*(NF+1)/2
      CALL UYTIM1(ITIME)
      CALL UYCLST
      NDECF=0
      CALL UO0FU1
*
*     ---------------------
*     VARIABLE METRIC METHOD
*     TEMPLATE : U1FDU1
*     ---------------------
*
      ASSIGN 11030 TO IMD
```

```
      CALL UYPRO1('UXFU',1)
      CALL UYPRO2
11010 CONTINUE
*
*     -----------------
*     MODEL DESCRIPTION
*     -----------------
*
11500 CALL UF1F01(NF,GF,GF,FF,F)
      GOTO (11540,11510,11520) ISB+1
11510 CONTINUE
      ASSIGN 11610 TO IMB
11600 CONTINUE
      NFV=NFV+1
      FF=1.2D0*(X(1)**2-X(2))**2+(X(1)-1.0D0)**2
      GOTO IMB
11610 CONTINUE
      GOTO 11500
11520 CONTINUE
      CALL UF0GS2(NF,X,IX,X,GF,FF,HD,R,SNORM,1.0D-15,1.0D-15,2,1)
      GOTO (11500,11530) ISB+1
11530 CONTINUE
      ASSIGN 11810 TO IMB
      GOTO 11600
11810 CONTINUE
      GO TO 11520
11540 CONTINUE
*
*     ------------------------
*     END OF MODEL DESCRIPTION
*     ------------------------
*
      GO TO IMD
11030 CONTINUE
      CALL UYTRUG(X,GF,GF)
      CALL UO2FU3(X,GF,HF,X,X)
      CALL UYFUT1
      IF(ITERM.NE.0) GOTO 11090
11040 CONTINUE
      ASSIGN 11040 TO IMD
      CALL UUDSD1(N,HF,1)
      GOTO (11050,11010) ISB+1
11050 CONTINUE
      IF(ITERM.NE.0) GOTO 11090
      CALL UYCPSD(IX,HF,HD)
      CALL UYTRUH(X,HF)
*
*     ------------------------
*     DIRECTION DETERMINATION
*     TEMPLATE : UDGLG1
*     ------------------------
*
```

```
      CALL UOD1D1
      IF (IDECF.LT.0) THEN
      IDECF=9
      INF=0
      ENDIF
      IF (IDECF.EQ.0) THEN
      TDXX(1:4)='INV '
*
*     INVERSION
*
      ALF=ETA2
      CALL UXDPGF(N,HF,INF,ALF,BET)
      CALL UXDPGI(N,HF)
      NDECF=NDECF+1
      IDECF=9
      ELSE IF (IDECF.EQ.9) THEN
      ELSE
      ITERD=-1
      TDXX='BAD DEC9'
      CALL UOERR1('UDDLI1',1)
      GO TO 12530
      ENDIF
      GNORM=SQRT(UXVDOT(N,GF,GF))
*
*     NEWTON LIKE STEP
*
      CALL UXDSMM(N,HF,GF,S)
      CALL UXVNEG(N,S,S)
      INITD=MAX(ABS(INITD),1)
      ITERD=1
      IF(INF.EQ.0) THEN
      TDXX(5:8)=' POS'
      ELSEIF(INF.LT.0) THEN
      TDXX(5:8)=' ZER'
      ELSE
      TDXX(5:8)=' NEG'
      ENDIF
      SNORM=SQRT(UXVDOT(N,S,S))
      NRED=INF
      CALL UOD1D5(ALF,SIG,INF)
12530 CALL UOD1D2(N,GF,S)
*
*     ------------------------------
*     END OF DIRECTION DETERMINATION
*     ------------------------------
*
      CALL UD1TL1(GF,S)
      IF(ITERM.NE.0) GOTO 11090
      IF(IREST.NE.0) GOTO 11040
      CALL UYTRUS(X,X,XO,GF,GO,S,S)
11070 CONTINUE
      ASSIGN 11070 TO IMD
```

```
      CALL US0L01(EPS1,R0,RP,R,F0,FP,F,P0,PP,FMIN,FMAX,PAR1,PAR2,RMAX,RM
     &    IN,SNORM,MODE,KTERS,MES,MES1,MES2,INITS,MRED)
      GOTO (11074,11072) ISB+1
11072 CONTINUE
      CALL UXVDIR(NF,R,S,X0,X)
      GOTO 11010
11074 CONTINUE
      IF (ITERS.LE.0) THEN
      CALL UYZER0(X,X0)
      IF(IDIR.EQ.0) THEN
      CALL UYRES1(TSXX)
      CALL UYSET1
      GO TO 11040
      ELSE IF (MOT3.EQ.0) THEN
      CALL UYSET1
      GO TO 11040
      ELSE
      ITERD=0
      ENDIF
      ENDIF
      IF(KD.GT.LD) THEN
      ASSIGN 11080 TO IMD
      GO TO 11010
      ENDIF
11080 CONTINUE
      TXFU=TUXX
      CALL UYUPSD(X,IX,X0,GF,G0,HD)
      CALL UYTRUD(X,X,X0,GF,G0)
      CALL UUDBI1(N,HF,GF,S,X0,G0,R,F,F0,P,P0,PAR1,PAR2,1.0D 60,8)
      IF(IDIR.EQ.0) THEN
      IF(ITERH.NE.0) CALL UYRES1('UPDATE  ')
      GOTO 11030
      ELSE
      GOTO 11040
      ENDIF
11090 CONTINUE
      IF(ITERM.LT.0) TXFU=TDXX
      CALL UYEPI1(1)
      CALL U01FU2(X,X,X,X)
*
*     -----------------
*     END OF METHOD (1)
*     -----------------
*
      CALL UYTIM2(ITIME)
      END
*
*     ------------------------
*     INITIATION OF METHOD (1)
*     ------------------------
*
      SUBROUTINE UYINT1
```

```
*
*     commons placed here were omitted
*     since they require a large space
*
      REAL*8 XDELS,RPF1S,RPF2S,RPF3S,RGF1S,RGF2S,RGF3S
      COMMON/UMCLST/ XDELS,RPF1S,RPF2S,RPF3S,RGF1S,RGF2S,RGF3S
      ETA0=1.0D-15
      ETA9=1.0D 60
      ITR=6
      IRD=5
      IWR=2
*
*     many other assignments follows which were
*     omitted since they require a large space
*
      END
*
*     ---------------------
*     INITIATION OF PROBLEM
*     ---------------------
*
      SUBROUTINE UYINTP
*
*     commons placed here were omitted
*     since they require a large space
*
      NF=2
      IEXT=0
      KCF=2
      KSF=1
      KBF=0
      KBC=0
      NORMF=0
      KDF=0
      KDA=-1
      KDC=-1
      KDE=-1
      KDY=-1
      END
*     ----------------------------------------
*     BROYDEN CLASS OF VARIABLE METRIC UPDATES
*     TEMPLATE : UUDBI1
*     ----------------------------------------
      SUBROUTINE UUDBI1(N,H,G,S,XO,GO,R,F,FO,P,PO,PAR1,PAR2,ETA9,MET)
*
*     commons placed here were omitted
*     since they require a large space
*
*
      REAL*8 H(N*(N+1)/2),G(N),S(N),XO(N),GO(N),R,F,FO,P,PO,PAR1,PAR2,ET
     &    A9
      REAL*8 AA,CC
```

```
      COMMON /UMFUN1/ AA,CC
      REAL*8 UXVDOT,UNFUN1
      REAL*8 UXDPGP
      REAL*8 DEN,DIS,POM,POM3,POM4,A,B,C,GAM,RHO,PAR
      INTEGER IUPDT
      LOGICAL L1,L2,L3
      EXTERNAL UNFUN1
      IF (MET.LE.0) GO TO 22
      CALL UOU1D1(N,XO,GO)
      IF (IDECF.NE.9) THEN
      ITERH=-1
      TUXX='BAD DEC9'
      CALL UOERR1('UUDBI2',1)
      GO TO 22
      ENDIF
      L1=ABS(3).GE.3.OR.ABS(3).EQ.2.AND.NIT.EQ.KIT
      L3=.NOT.L1
*
*     DETERMINATION OF THE PARAMETERS A, B, C
*
      B=UXVDOT(N,XO,GO)
      IF (B.LE.ZERO) THEN
      ITERH=2
      TUXX='B - NEG.'
      GO TO 22
      ENDIF
      CALL UXDSMM(N,H,GO,S)
      A=UXVDOT(N,GO,S)
      IF (A.LE.ZERO) THEN
      ITERH=1
      TUXX='A - NEG.'
      GO TO 22
      ENDIF
      IF(MET.GE.4.OR.L1) THEN
      IF (ITERD.NE.1) THEN
      MET=1
      C=ZERO
      ELSE
      C=-R*PO
      IF (C.LE.ZERO) THEN
      ITERH=3
      TUXX='C - NEG.'
      GO TO 22
      ENDIF
      ENDIF
      ELSE
      C=ZERO
      ENDIF
*
*     DETERMINATION OF THE PARAMETER RHO (NONQUADRATIC PROPERTIES)
*
      RHO=HALF*B/(FO-F+P)
```

```
      IF(RHO.LE.1.0D-2) RHO=ONE
      IF(RHO*1.0D-2.GE.ONE) RHO=ONE
      AA=A/B
      CC=C/B
      IUPDT=0
      IF (L1) THEN
*
*     DETERMINATION OF THE PARAMETER GAM (SELF SCALING)
*
      IF (C.LE.ZERO) THEN
      PAR=A/B
      POM3=0.8D 0
      POM4=8.0D 0
      ELSE
      PAR=SQRT(A/C)
      POM3=0.7D 0
      POM4=6.0D 0
      ENDIF
      GAM=RHO/PAR
      IF (NIT.NE.KIT) THEN
      L2=PAR2.LE.ZERO
      L3=L2.AND.ABS(PAR1).LE.0.2D 0
      L3=L3.OR.(.NOT.L2.AND.GAM.GT.ONE)
      L3=L3.OR.(L2.AND.PAR1.LT.ZERO.AND.GAM.GT.ONE)
      L3=L3.OR.(L2.AND.PAR1.GT.ZERO.AND.GAM.LT.ONE)
      L3=L3.OR.GAM.LT.POM3
      L3=L3.OR.GAM.GT.POM4
      ENDIF
      ENDIF
      IF (L3) THEN
      GAM=ONE
      PAR=RHO/GAM
      ENDIF
*
*     NEW UPDATE
*
      POM=ONE/(AA*CC)
      DEN=MAX(POM+1.0D-15,(C/A)**0.2D 0)
      POM=(DEN-POM)/MAX(1.0D-15,ONE-POM)
      TUXX='NEW     '
   20 CONTINUE
*
*     GENERAL UPDATE
*
      DEN=PAR+POM*AA
      DIS=POM/DEN
      CALL UXDSMU(N,H,(PAR*DIS-ONE)/A,S)
      CALL UXVDIR(N,-DIS,S,XO,S)
      CALL UXDSMU(N,H,DEN/B,S)
   21 CONTINUE
      ITERH=0
      IF (GAM.EQ.ONE) GO TO 22
```

```
*
*     SCALING
*
      CALL UXDSMS(N,H,GAM)
   22 CONTINUE
      CALL UOU1D2(N,H,S,RHO,GAM,PAR,A,B,C,POM,ETA9)
      RETURN
      END
```

The results (screen output) obtained using this control program have the following form:

```
   0  NIT=  16  NFV=  60  NFG=   0  NDC=   0  NCG=   0  F= .169D-11  G= .605D-05
 FF=   .1685701450D-11
 X =   .9999997761D+00   .1000000720D+01
 TIME= 0:00:00.11
```

Appendix B. The BEL interpreter

The BEL (Batch Editor Language) interpreter, developed as a part of the UFO project, is especially determined for the generation of computer programs, batch editing of texts, preparation of print files, filtering of text files etc. The BEL interpreter allows us to generate a prescribed output file from the input file (template) which is a mixture of text lines and special instructions.

The UFO system is organized in such a way that a control program may not be written in the FORTRAN language immediately. Instead, the procedure written in the UFO control language is supplied. By using the installation template, the compiler of the UFO control language (UFOCLP - UFO Control Language Preprocessor) generates the table of symbols which is together with the user supplied procedure offered to the BEL interpreter. The BEL interpreter then generates the resulting control program which is written in the FORTRAN language.

B.1. General description

Although the BEL interpreter can be used in various general applications, it was developed especially for the generation of FORTRAN programs. It is:

1. Interpreter, since instructions contained in the input text are interpreted and immediately realized.
2. Batch editor, since it serves for editing batch files.
3. Macroprocessor, since it makes it possible to define or modify special macrovariables which can be substituted to the processed text.

The macrovariable can be an integer constant, a logical constant, a string of characters, a set of text lines, a set of BEL instructions, even a text file.

The BEL interpreter requires an input text file and a table of symbols. The input text file (template) consists of standard text lines together with the BEL instructions. The table of symbols contains names and values of used macrovariables.

The BEL instructions, contained in the input text file, can be of two types:

1. Directives, i.e. control instructions and instructions for manipulation with a table of symbols. These instructions begin with the special character CHDIR. In the subsequent text, we will suppose that CHDIR='$' ('$' is the default value).
2. Substitutions, i.e.instructions for substitution of macrovariables into the text. These instructions begin with the special character CHSUB. In the subsequent text, we will suppose that CHSUB='$' ('$' is the default value).

The BEL interpreter works in the following way:

1. The line of the input file is read.
2. The line is recognized and if the character CHSUB is found, then a pertinent substitution is realized.
3. If the first character (different from blank) is CHDIR, then the line is a directive line. The recognized directive is realized.

This process is repeated until the directive $END or the end of the file is found. Note that we suppose that CHSUB and CHDIR have the same values. This is allowed, since the correct meaning is recognized from the context.

At the end of this subsection, we stress some specific features and advantages of the BEL interpreter.

1. The substitution is recursive. The depth of recursion only depends on the declared work space size.

2. The substitution is allowed in both the text lines and the directives.

3. Names and values of macrovariables can have an arbitrary length which again only depends on the declared work space size.

4. The set of directives is relatively small with a consistent syntax. It contains all important instructions ($IF-$ELSEIF-$ELSE-$ENDIF, $DO-$ENDDO, $REPEAT-$UNTIL etc.)

5. The control parameters (CHDIR, CHSUB etc.) can be changed during the work of the BEL interpreter. This makes it possible to generate a program written in the BEL language which can be immedialely processed.

6. The BEL interpreter is a fully portable device. It can be implemented in an arbitrary system containing FORTRAN 77 compiler.

B.2. List of instructions

Substitutions:

$INTEGER	- substitute by the absolute label computed from the relative label.
$NAME, $(NAME)	- substitute by the value of the macrovariable NAME.
$DATA(NAME)	- substitute by a new item from the list of items which is a value of the macrovariable NAME.
$DEF(NAME)	- substitute by '.TRUE.' if the macrovariable NAME is defined in the table of symbols. Otherwise substitute by '.FALSE.'
$INT(NAME)	- substitute by '.TRUE.' if the value of the macrovariable NAME is an integer constant. Otherwise substitute by '.FALSE.'
$LOG(NAME)	- substitute by '.TRUE.' if the value of the macrovariable NAME is a logical constant. Otherwise substitute by '.FALSE.'
$REAL(NAME)	- substitute by '.TRUE.' if the value of the macrovariable NAME is a real constant. Otherwise substitute by '.FALSE.'
$$	- substitute '$' (replace '$$' by '$'). This makes possible to insert the character CHSUB into the text.

Directives:

$ADD	- add a value to a macrovariable.
$ADD, $ENDADD	- add text lines to a macrovariable.
$CLEAR	- clear value of a macrovariable which is a list of items type.
$DO, $ENDDO	- cycle.
$EXIT	- termination of the BEL interpreter work.
$ERASE	- erase a macrovariable from the table of symbols.
$IF, $ELSEIF, $ELSE, $ENDIF	- conditioned instruction.
$HELP, $CHECK	- set a default value to a macrovariable which has not been previously defined.
$INCLUDE	- insert a macrovariable or a text file into the output file.
$OPTION	- change some optional parameter of the BEL interpreter.
$REM	- remark.
$REPEAT, $UNTIL	- cycle.
$RESTORE	- adjust the list of items pointer to the first item.
$REWIND	- rewind the file on a given unit.
$SET	- set a value to a macrovariable.
$SET, $ENDSET	- set text lines to a macrovariable.
$SUBST	- substitute a text file into the input file.

B.3. Special characters

The following special characters are important for the BEL interpreter work:

$ - CHSUB (Substitution Character) - this is the first character in every substitution. If '$' should be inserted into the text, then we have to use '$$'.

$ - CHDIR (Directive Character) - if the first character on the line is CHDIR, then the line is a directive line (CHSUB and CHDIR are distinguished by the context).

& - CHCON (Continuation Character) - if the last character on the line is CHCON, then it is assumed that the logical line continues on the next physical line.

; - CHEOL (End Of Line Character) - this character specifies the end of the logical line if it does not coincide with the end of the physical line. This makes it possible to write several logical lines by using the same physical line.

\ - CHDS (Data Separator Character) - this character separates individual items in the list of items type macrovariable.

The use of special characters can be demonstrated by the following simple example. Assume that the input text has the form

```
$A='Paul\Peter\Jane\Mary'
This is a list of my brothers and sis&
ters:
$DO(I=1,4); $DATA(A); $ENDDO
```

Then the output from the BEL interpreter has the form

```
This is a list of my brothers and sisters:
Paul
Peter
Jane
Mary
```

The special characters can be changed by the directive $OPTION. But no special character has to be an alphabet or a digit. Moreover, different special characters have to differ (with the exception of CHSUB and CHDIR).

B.4. Description of instructions

This subsection contains a detailed description of the syntax and action of individual BEL instructions. The following definitions will be used:

<digit> ::= 0 | 1 | 2 | 3 | ..... | 9

<alphabet> ::= A | B | C | D | ..... | Z

<character> ::= an arbitrary character with exception of apostrophe

<integer constant> ::= (+ | -) <digit> {<digit>}

<logical constant> ::= .TRUE. | .FALSE.

<macroname> ::= <alphabet> {<alphabet> | <digit>}

<string of characters> ::= '{<character> | ''}

<text> ::= <string of characters> '{; <string of characters>}

<list of items> ::= <string of characters> '{\ <string of characters>}

Substitutions:

$INTEGER

Syntax:
The type of INTEGER is an integer constant. Although it can have an arbitrary value, an application to the control program generation requires that it is positive and less then LABEL2 (see the directive $OPTION).

Action:
The integer constant INTEGER is a relative label in a given template. The absolute label, substituted into the control program, is computed by the formula LABEL=LABEL1+K*LABEL2, where LABEL1 and LABEL2 are options of the BEL interpreter (see the directive $OPTION) and K is a serial number of application of the directive $SUBST.

Example:

$10

generates

10010

if the main template is used or

10110

after the first application of the directive $SUBST.

$NAME, $(NAME)

Syntax:
The type of NAME is a macroname. This substitution has two forms either $NAME or $(NAME). The latter form is required if the substitution appears inside a continuous string of characters to separate the NAME from the adjacent text.

Action:
The string '$NAME' is replaced by the value of the macrovariable NAME.

Example:

$A='UFO'
$A SYSTEM

generates

UFO SYSTEM

$DATA(NAME)

Syntax:
The type of NAME is a macroname.

Action:
The string '$DATA(NAME)' is replaced by the next item of the list of items which is a value of the macrovariable NAME. If the next item does not exist, then the list of items pointer is returned to the first item. Additional information is contained in the description of the directive $RESTORE.

Example:

$LIST='ITEM1\ITEM2\ITEM3'
$DATA(LIST)
$DATA(LIST)

$DATA(LIST)
$DATA(LIST)

generates

ITEM1
ITEM2
ITEM3
ITEM1

$DEF(NAME)

Syntax:
The type of NAME is a macroname.

Action:
If the macrovariable NAME is defined in the table of symbols, then the string '$DEF(NAME)' is replaced by the logical constant .TRUE., otherwise it is replaced by the logical constant .FALSE..

Example:

$A=10
$DEF(A)

generates

.TRUE.

$INT(NAME)

Syntax:
The type of NAME is a macroname.

Action:
If the value of the macrovariable NAME is an integer constant, then the string '$INT(NAME)' is replaced by the logical constant .TRUE., otherwise it is replaced by the logical constant .FALSE..

Example:

$A=-25
$INT(A)

generates

.TRUE.

$LOG(NAME)

Syntax:
The type of NAME is a macroname.

Action:
If the value of the macrovariable NAME is a logical constant, then the string '$INT(NAME)' is replaced by the logical constant .TRUE., otherwise it is replaced by the logical constant .FALSE..

Example:

$A=.FALSE.
$LOG(A)

generates

.TRUE.

$REAL(NAME)

Syntax:
The type of NAME is a macroname.

Action:
If the value of the macrovariable NAME is a real constant (i.e. string of character which satisfies syntactic rules for FORTRAN real constants), then the string '$REAL(NAME)' is replaced by the logical constant .TRUE., otherwise it is replaced by the logical constant .FALSE..

Example:

```
$A='-0.09D-12'
$REAL(A)
```

generates

.TRUE.

$$

Action:
The string '$$' is replaced by the character '$'. This substitution alows us to insert the character '$' into the generated text or into the macrovariable.

Example:

```
$I='NAME'
$$DEF($I)
```

generates

$DEF(NAME)

Directives:

$ADD(NAME1,NAME2 or VALUE)

Syntax:
The type of NAME1 and NAME2 is a macroname.
The type of VALUE is an integer constant or a logical constant or a string of characters.

Action:
The value of the macrovariable NAME2 or the VALUE is added to the value of the macrovariable NAME1 (the resulting value of the macrovariable NAME1 is $NAME1$NAME2 in the first case).

Example:

```
$NAME='TOM'
$ADD(NAME,' JONES')
Name: $NAME
```

generates

Name: TOM JONES

$ADD(NAME)
TEXT
$ENDADD

Syntax:
The type of NAME is a macroname.
The type of TEXT is text.

Action:
The TEXT is added to the value of the macrovariable NAME.

Example:

```
$SET(A)
   Day: 31
$ENDSET
$ADD(A)
   Month: December
   Year: 1997
$ENDADD
```

generates

```
   Day: 31
   Month: December
   Year: 1997
```

Remark: Only substitutions are realized in the text TEXT (not directives).

$CLEAR(NAME)

Syntax:
The type of NAME is a macroname.

Action:
This directive clears a list of items type value of the macrovariable NAME, i.e. it deletes all duplications of items. Small and capital letters of items are not distinguished.

Example:

```
$DECL='N\IX(N)\N\M\ I\J\N\M'
$CLEAR(DECL)
$END='$DATA(DECL)'
$REPEAT
   $I='$DATA(DECL)'
   INTEGER $I
$UNTIL(I=END)
```

generates

```
   INTEGER IX(N)
   INTEGER M
   INTEGER I
   INTEGER J
   INTEGER N
```

$DO(NAME=INDEX1,INDEX2,INDEX3)
TEXT
$ENDDO

Syntax:
The type of NAME is a macroname.
The type of INDEX1, INDEX2, INDEX3 is a macroname or an integer constant.
The type of TEXT is text.

Action:
This directive has a similar meaning as the statement DO in the FORTRAN language:
NAME is the cycle counter.

INDEX1 is the initial value of the cycle counter.
INDEX2 is the final value of the cycle counter.
INDEX3 is the change of the cycle counter after a cycle step.
If INDEX3 is not present, then the default value INDEX1=1 is assumed.
The cycle counter NAME does not have to be changed in the cycle step.
The value INDEX3 does not have to be equal to 0.
The body of the cycle is terminated by $ENDDO.
If INDEX1>INDEX2 and INDEX3>0 or INDEX1<INDEX2 and INDEX3<0, then the cycle is not realized.
Cycles can be nested. The maximum depth of nested cycles is 20.

Example:

```
$A='X\Y\Z'
$DO(I=1,5,2)
   A($I,1)=C($I)+$DATA(A)
$ENDDO
```

generates

```
   A(1,1)=C(1)+X
   A(3,1)=C(3)+Y
   A(5,1)=C(5)+Z
```

$ERASE(NAME)

Syntax:
The type of NAME is a macroname.

Action:
The macrovariable NAME is erased from the table of symbols.

Example:

```
$A=1
$DEF(A)
$ERASE(A)
$DEF(A)
```

generates

```
   .TRUE.
   .FALSE.
```

$EXIT

Action:
The directive $EXIT has the same meaning as the end of the file achievement. If nested files are processed (see the description of the directive $SUBST), then the directive $EXIT realizes to return to the hihger level file (if the higher level file does not exist, then $EXIT has the same meaning as $STOP).

$HELP
TEXT
$CHECK(NAME,DEFAULT,TYPE,LEVEL,TRANSFER)

Syntax:
The type of TEXT is text.
The type of NAME is a macroname.
The type of DEFAULT is either a macroname or an integer constant or a logical constant or a string of characters.

The type of TYPE is either list of items or one of the strings INT (integer), LOG (logical), REAL (real).
The type of LEVEL is an integer constant.
The type of TRANSFER is a logical constant.

Action:
The text TEXT appears on the screen if the dialogue mode is used. The value of the macrovariable $NAME is checked to have the type TYPE. If the macrovariable $NAME is not defined or if it has a wrong value then the value DEFAULT is used. The value of LEVEL gives the lowest level of the dialogue (1,2,3 or 4) from which the text TEXT appears on the screen. The value of TRANSFER specifies transfer of the variable $NAME into the control program (YES if transfer is accepted or NO if transfer is suppressed).

Example:

```
$HELP
   TYPE OF HESSIAN MATRIX:
        D - DENSE
        B - SPARSE WITH KNOWN (PARTITIONED) STRUCTURE
        S - SPARSE WITH KNOWN (GENERAL) STRUCTURE
        NO - HESSIAN MATRIX IN NOT USED
$CHECK(HESF,'NO','D\B\S\NO',1,NO)
```

$IF(CONDITION) LINE

Syntax:
The CONDITION can be of the following types:
The type of CONDITION is a macroname and a value of CONDITION is a logical constant.
The type of CONDITION is a logical constant (.TRUE. or .FALSE.).
The type of CONDITION is a string of the form PART1<operator>PART2.
The type of PART1 and PART2 can be a macroname or an integer constant or a logical constant or a string (values of PART1 and PART2 have to be of the same type) and <operator> can have the following forms:

= equal to
<> not equal to
< less than (for integer values only)
<= less than or equal to (for integer values only)
> greater than (for integer values only)
>= greater than or equal to (for integer values only)

LINE is either text line or directive.

Action:
If the condition CONDITION is satisfied then LINE is generated into the output file (if it is a text line) or carried out (if it is a directive). If values of PART1 and PART2 are strings, then small and capital letters are not distinguished and blanks are ignored.

Example:

```
$A='J O H N'
$IF(A='John') Yes
$IF(A<>'Mary') No
```

generates

```
   Yes
   No
```

$IF(CONDITION1)
  TEXT1
$ELSEIF(CONDITION2)
  TEXT3
  .
  .
  .
$ELSE
  TEXT
$ENDIF

Syntax:
CONDITION1 and CONDITION2 have the same syntax and meaning as CONDITION in the previous case. The number of repeated $ELSEIF is not limited, $ELSEIF or $ELSE can be omitted.

Action:
This directive has a similar meaning as the conditioned statement IF-ELSEIF-ELSE-ENDIF in the FORTRAN language. The conditioned statements can be nested. The maximum depth of nested conditioned statements is 20.

Example:

```
$A=10
$L=.FALSE.
$IF(A=10)
   A = A + 1
   B = B + 1
   $IF(L)
   C = C + 1
   $ENDIF
$ELSE
   WRITE(*,*) I
$ENDIF
```

generates

```
   A = A + 1
   B = B + 1
```

$INCLUDE(NAME)

Syntax:
The type of NAME is a macroname.

Action:
The directive $INCLUDE(NAME) is a special case of substitution. This directive makes us possible to insert (into generated text) one or more lines, which were previously assigned to the macrovariable NAME. In contrast with the standard substitution $NAME, the inserted lines are not processed by the BEL interpreter, so that directives are not carried out.

Example:

```
$SET(LINES)
   $ADD(A)
   X = Y + Z
   CALL SUB(X)
   $ENDADD
$ENDSET
```

$INCLUDE(LINES)

generates

```
$ADD(A)
X = Y + Z
CALL SUB(X)
$ENDADD
```

$INCLUDE('FILE')

Syntax:
The type of FILE is a string.

Action:
The directive $INCLUDE('FILE') is a special case of substitution. This directive makes us possible to insert (into generated text) the text, which is stored in the file with the name FILE. The inserted text is not processed by the BEL interpreter, so that directives are not carried out.

Example:

$INCLUDE('C:\UFO\UMCOMN.I')

includes FORTRAN common blocks into the generated text (these common blocks are stored in the file C:\UFO\UMCOMN.I.

$OPTION(OPTIONNAME=NAME or VALUE)

Syntax:
OPTIONNAME is a selected name from the table of optional parameters (see below).
The type of NAME is a macroname. The value of NAME has to be an integer constant or a logical constant or a string of character and has to correspond to the type of OPTIONNAME.
The type of VALUE has to be an integer constant or a logical constant or a string of character and has to correspond to the type of OPTIONNAME.

Action:
This directive makes us possible to change selected optional parameter of the BEL interpreter. Optional parameters are contained in the following table.

Name	Type	Default	Description
CHDIR	char.	$	see B.3
CHEOL	char.	;	see B.3
CHCON	char.	&	see B.3
CHDS	char.	\	see B.3
ILNLEN	int.	80	physical length of the input line
OLNLEN	int.	80	physical length of the input line
IUNIT	int.	-	No. of the input file unit
OUNIT	int.	-	No. of the output file unit
INUNIT	int.	-	No. of the $INCLUDE files unit
IIUNIT	int.	-	No. of the interactive mode input unit
OIUNIT	int.	-	No. of the interactive mode output unit
DIALOG	int.	1	level of dialogue (0 or 1 or 2)
MODERW	int.	1	READ/WRITE mode (1 or 2 or 3)
LABEL1	int.	10000	initial label
LABEL2	int.	100	difference between two consecutive labels
LSUBS	int.	.TRUE.	substitutions carried out
LOUT	int.	.TRUE.	output file created
LSMLET	int.	.TRUE.	small letters used in instructions
LFORTO	int.	.TRUE.	output in standard FORTRAN format
LFRFMT	int.	.TRUE.	input in free FORTRAN format (used only if LFORTO=.TRUE.)

$REM

Action:
The rest of the line (following after $REM) is ignored by the BEL interpreter. The directive $REM is used for remarks.

$REPEAT
TEXT
$UNTIL(CONDITION)

Syntax:
The type of TEXT is text.
CONDITION has the same syntax and meaning as that in the directive $IF(...).

Action:
This directive has a similar meaning as the statement REPEAT-UNTIL in the PASCAL language:
The cycle is terminated whenever the condition CONDITION is satisfied (at least one realization is carried out).
Cycles can be nested. The maximum depth of nested cycles is 20.

Example:

```
$N=20
$REAL='X($N)\G($N)\H($N,$N)\.END.'
$REPEAT
   $I=$DATA(REAL)
   REAL $I
$UNTIL(I='.END.')
```

generates

```
   REAL X(20)
   REAL G(20)
```

```
   REAL H(20,20)
```

$RESTORE(NAME)

Syntax:
The type of NAME is a macroname.

Action:
The directive $RESTORE(NAME) can only be used if the value of the macrovariable NAME is a list of items. Such a macrovariable uses a pointer which points out the next called item. The directive $RESTORE adjust this pointer to point out the first item of the list (if the end of this list is found, then the pointer is adjusted to point out the first item without the application of the directive $RESTORE).

Example:

```
$A='X\Y\Z'
   $DATA(A)
   $DATA(A)
$RESTORE(A)
   $DATA(A)
```

generates

```
   X
   Y
   X
```

$REWIND(UNIT)

Syntax:
The type of UNIT is an integer constant.

Action:
The file opened on the unit with the number UNIT is rewound so that it can again be read from the first record (numbering of I/O units is used in the FORTRAN language).

$NAME1 = NAME2 or VALUE)

$SET(NAME1 = NAME2 or VALUE)

Syntax:
The type of NAME1 and NAME2 is a macroname.
The type of VALUE is an integer constant or a logical constant or a string of characters.
This directive has two forms. The latter form is used if the a macroname is identical with some directive (e.g. $SET(REM='REMARK')).

Action:
The new macrovariable with the name NAME1 and the value equal to the value of the macrovariable NAME2 or constant VALUE is inserted into the table of symbols. If the macrovariable NAME1 has already been defined in the table of symbols, then it is changed.

$SET(NAME)
TEXT
$ENDSET

Syntax:
The type of NAME is a macroname.
The type of TEXT is text.

Action:

The macrovariable NAME is inserted into the table of symbols with the value TEXT. If the macrovariable NAME has already been defined in the table of symbols, then it is changed.

Example:

```
$SET(INIT)
   CALL EIUD01(NF,X,FMIN,XMAX,NEXT,IEXT,IERR)
   IF (IERR.NE.0) GO TO $$ENDTEST
$ENDSET
$INIT
```

generates

```
   CALL EIUD01(NF,X,FMIN,XMAX,NEXT,IEXT,IERR)
   IF (IERR.NE.0) GO TO $ENDTEST
```

Remark: Only substitutions are realized in the text TEXT (not directives).

$SUBST('FILE')

Syntax:
The type of FILE is a string.

Action:
This directive performs the following actions:
The new reference label is computed (using the parameters LABEL1 and LABEL2 of the BEL interpreter).
The file with the name FILE is opened.
This file is processed by the BEL interpreter.
The file with the name FILE is closed.
The old reference label is restored.

This directive is similar to the directive $INCLUDE('FILE'). But the inserted text is now processed by the BEL interpreter. All substitutions and directives are carried out. The directive $SUBST('FILE') serves for the division of large texts into segments and makes the generation of texts possible by using conditioned branching. This is advantageously used for generation of the control program in the UFO system where templates corresponding individual subroutines are such segments.

Example:

```
$INCLUDE('C:\UFO\UMCOMN.I')
```

includes FORTRAN common blocks into the generated text (these common blocks are stored in the file C:\UFO\UMCOMN.I.

Appendix C. Graphical screen output