

## H02CBF – NAG Fortran Library Routine Document

**Note.** Before using this routine, please read the Users' Note for your implementation to check the interpretation of bold italicised terms and other implementation-dependent details.

**Note.** This routine uses optional parameters to define choices in the problem specification and in the details of the algorithm. If you wish to use default settings for all of the optional parameters, you need only read Section 1 to Section 9 of this document. Refer to the additional Section 10, Section 11 and Section 12 for a detailed description of the algorithm, the specification of the optional parameters and a description of the monitoring information produced by the routine.

### 1 Purpose

H02CBF solves general quadratic programming problems with integer constraints on the variables. It is not intended for large sparse problems.

### 2 Specification

```

SUBROUTINE H02CBF(N, NCLIN, A, LDA, BL, BU, CVEC, H, LDH, QPHESS,
1          INTVAR, LINTVR, MDEPTH, ISTATE, XS, OBJ, AX,
2          CLAMDA, STRTGY, IWRK, LIWRK, WRK, LWRK, MONIT,
3          IFAIL)
  INTEGER   N, NCLIN, LDA, LDH, INTVAR(LINTVR), LINTVR,
1          MDEPTH, ISTATE(N+NCLIN), STRTGY, IWRK(LIWRK),
2          LIWRK, LWRK, IFAIL
  real      A(LDA,*), BL(N+NCLIN), BU(N+NCLIN), CVEC(*),
1          H(LDH,*), XS(N+NCLIN), OBJ, AX(*),
2          CLAMDA(N+NCLIN), WRK(LWRK)
  EXTERNAL MONIT, QPHESS

```

### 3 Description

H02CBF uses a 'Branch and Bound' algorithm in conjunction with E04NFF to try and determine integer solutions to a general quadratic programming problem. Only when the problem is linear and the matrix  $H$  is positive definite can the technique be guaranteed to work; but often useful results can be obtained for a wider class of problems.

Branch and bound consists firstly of obtaining a solution without any of the variables  $x = (x_1, x_2, \dots, x_n)^T$  constrained to be integer. Suppose  $x_1$  ought to be integer, but at the optimal value just computed  $x_1 = 2.4$ . A constraint  $x_1 \leq 2$  is added to the system and the second problem solved. A constraint  $x_1 \geq 3$  gives rise to a third sub-problem. In a similar manner a whole series of sub-problems may be generated, corresponding to integer constraints on the variables. The sub-problems are all solved using E04NFF.

In practice the routine tries to compute an integer solution as quickly as possible using a depth-first approach, since this helps determine a realistic cut-off value. If we have a cut-off value, say the value of the function at this first integer solution, and any sub-problem,  $W$  say, has a solution value greater than this cut-off value, then subsequent sub-problems of  $W$  must have solutions greater than the value of the solution at  $W$  and therefore need not be computed. Thus a knowledge of a good cut-off value can result in fewer sub-problems being solved and thus speed up the operation of the routine. (See the description of MONIT in Section 5 for details of how users can supply their own cut-off value.)

### 4 References

- [1] Gill P E, Hammarling S, Murray W, Saunders M A and Wright M H (1986) User's guide for LSSOL (Version 1.0) *Report SOL 86-1* Department of Operations Research, Stanford University
- [2] Gill P E and Murray W (1978) Numerically stable methods for quadratic programming *Math. Programming* **14** 349–372

- [3] Gill P E, Murray W, Saunders M A and Wright M H (1984) Procedures for optimization problems with a mixture of bounds and general linear constraints *ACM Trans. Math. Software* **10** 282–298
- [4] Gill P E, Murray W, Saunders M A and Wright M H (1989) A practical anti-cycling procedure for linearly constrained optimization *Math. Programming* **45** 437–474
- [5] Gill P E, Murray W, Saunders M A and Wright M H (1991) Inertia-controlling methods for general quadratic programming *SIAM Rev.* **33** 1–36
- [6] Pardalos P M and Schnitger G (1988) Checking local optimality in constrained quadratic programming is NP-hard *Operations Research Letters* **7** 33–35
- [7] Gill P E, Murray W and Wright M H (1981) *Practical Optimization* Academic Press

## 5 Parameters

1: N — INTEGER *Input*

*On entry:*  $n$ , the number of variables.

*Constraint:*  $N > 0$ .

2: NCLIN — INTEGER *Input*

*On entry:*  $m_L$ , the number of general linear constraints.

*Constraint:*  $NCLIN \geq 0$ .

3: A(LDA,\*) — **real** array *Input*

**Note:** the second dimension of the array A must be at least N when  $NCLIN > 0$ , and at least 1 when  $NCLIN = 0$ .

*On entry:* the  $i$ th row of A must contain the coefficients of the  $i$ th general linear constraint, for  $i = 1, 2, \dots, m_L$ .

If  $NCLIN = 0$  then the array A is not referenced.

4: LDA — INTEGER *Input*

*On entry:* the first dimension of the array A as declared in the (sub)program from which H02CBF is called.

*Constraint:*  $LDA \geq \max(1, NCLIN)$ .

5: BL(N+NCLIN) — **real** array *Input*

6: BU(N+NCLIN) — **real** array *Input*

*On entry:* BL must contain the lower bounds and BU the upper bounds, for all the constraints in the following order. The first  $n$  elements of each array must contain the bounds on the variables, and the next  $m_L$  elements the bounds for the general linear constraints (if any). To specify a non-existent lower bound (i.e.,  $l_j = -\infty$ ), set  $BL(j) \leq -bigbnd$ , and to specify a non-existent upper bound (i.e.,  $u_j = +\infty$ ), set  $BU(j) \geq bigbnd$ ; the default value of *bigbnd* is  $10^{20}$ , but this may be changed by the optional parameter **Infinite Bound Size** (see Section 11.2). To specify the  $j$ th constraint as an *equality*, set  $BL(j) = BU(j) = \beta$ , say, where  $|\beta| < bigbnd$ .

*Constraints:*

$$BL(j) \leq BU(j), \text{ for } j = 1, 2, \dots, N+NCLIN,$$

$$|\beta| < bigbnd \text{ when } BL(j) = BU(j) = \beta.$$

**7:** CVEC(\*) — *real* array *Input*

**Note:** the dimension of the array CVEC must be at least  $N$  when the problem is of type LP, QP2 (the default) or QP4, and at least 1 otherwise.

*On entry:* the coefficients of the explicit linear term of the objective function when the problem is of type LP, QP2 (the default) and QP4.

If the problem is of type FP, QP1, or QP3, CVEC is not referenced.

**8:** H(LDH,\*) — *real* array *Input*

**Note:** the second dimension of the array H must be at least  $N$  if it is to be used to store  $H$  explicitly, and at least 1 otherwise.

*On entry:* H may be used to store the quadratic term  $H$  of the QP objective function if desired. In some cases, the user need not use H to store  $H$  explicitly (see the specification of subroutine QPHESS below). The elements of H are referenced only by subroutine QPHESS. The number of rows of  $H$  is denoted by  $m$ , whose default value is  $n$ . (The optional parameter **Hessian Rows** may be used to specify a value of  $m < n$ ; see Section 11.2).

If the default version of QPHESS is used and the problem is of type QP1 or QP2 (the default), the first  $m$  rows and columns of H must contain the leading  $m$  by  $m$  rows and columns of the symmetric Hessian matrix  $H$ . Only the diagonal and upper triangular elements of the leading  $m$  rows and columns of H are referenced. The remaining elements need not be assigned.

If the default version of QPHESS is used and the problem is of type QP3 or QP4, the first  $m$  rows of H must contain an  $m$  by  $n$  upper trapezoidal factor of the symmetric Hessian matrix  $H^T H$ . The factor need not be of full rank, i.e., some of the diagonal elements may be zero. However, as a general rule, the larger the dimension of the leading non-singular sub-matrix of H, the fewer iterations will be required. Elements outside the upper trapezoidal part of the first  $m$  rows of H need not be assigned.

If a non-default version of QPHESS is supplied, then in some cases it may be desirable to use a one-dimensional array to transmit data to QPHESS. (This is illustrated in the example program in Section 9 of the document for H02CCF.) H is then declared as a vector with dimension (LDH), where  $LDH \geq N \times (N+1)/2$ .

In other situations, it may be desirable to compute  $Hx$  or  $H^T Hx$  without accessing H – for example, if  $H$  or  $H^T H$  is sparse or has special structure. The parameters H and LDH may then refer to any convenient array.

If the problem is of type FP or LP, H is not referenced.

**9:** LDH — INTEGER *Input*

*On entry:* the first dimension of the array H as declared in the (sub)program from which H02CBF is called.

*Constraints:*

if the problem is of type QP1, QP2 (the default), QP3 or QP4,  $LDH \geq N$  or at least the value of the optional parameter **Hessian Rows** (default value =  $n$ ; see Section 11.2).

if the problem is of type FP or LP,  $LDH \geq 1$ .

**10:** QPHESS — SUBROUTINE, supplied by the NAG Fortran Library or the user. *External Procedure*

In general, the user need not provide a version of QPHESS, because a ‘default’ subroutine with name E04NFU is included in the Library (NFUE04 in some implementations: see the Users’ Note for your implementation for details). However, the algorithm of H02CBF requires only the product of  $H$  or  $H^T H$  and a vector  $x$ ; and in some cases the user may obtain increased efficiency by providing a version of QPHESS that avoids the need to define the elements of the matrices  $H$  or  $H^T H$  explicitly.

QPHESS is not referenced if the problem is of type FP or LP, in which case QPHESS may be the routine E04NFU (NFUE04 in some implementations).

Its specification is:

SUBROUTINE QPHESS(N, JTHCOL, H, LDH, X, HX)		
INTEGER	N, JTHCOL, LDH	
<i>real</i>	H(LDH,*), X(N), HX(N)	
<b>1:</b>	N — INTEGER	<i>Input</i>
	<i>On entry:</i> this is the same parameter N as supplied to H02CBF (see above).	
<b>2:</b>	JTHCOL — INTEGER	<i>Input</i>
	<i>On entry:</i> JTHCOL specifies whether or not the vector $x$ is a column of the identity matrix. If $JTHCOL = j > 0$ , then the vector $x$ is the $j$ th column of the identity matrix, and hence $Hx$ or $H^T Hx$ is the $j$ th column of $H$ or $H^T H$ , respectively, which may in some cases require very little computation and QPHESS may be coded to take advantage of this. However special code is not necessary because $x$ is always stored explicitly in the array X. If $JTHCOL = 0$ , $x$ has no special form.	
<b>3:</b>	H(LDH,*) — <i>real</i> array	<i>Input</i>
	<i>On entry:</i> this is the same parameter H as supplied to H02CBF (see above).	
<b>4:</b>	LDH — INTEGER	<i>Input</i>
	<i>On entry:</i> this is the same parameter LDH as supplied to H02CBF (see above).	
<b>5:</b>	X(N) — <i>real</i> array	<i>Input</i>
	<i>On entry:</i> the vector $x$ .	
<b>6:</b>	HX(N) — <i>real</i> array	<i>Output</i>
	<i>On exit:</i> the product $Hx$ if the problem is of type QP1 or QP2 (the default), or the product $H^T Hx$ if the problem is of type QP3 or QP4.	

QPHESS must be declared as EXTERNAL in the (sub)program from which H02CBF is called. Parameters denoted as *Input* must **not** be changed by this procedure.

- 11:** INTVAR(LINTVR) — INTEGER array *Input*  
*On entry:* INTVAR( $i$ ) must contain the index of the solution vector  $x$  which is required to be integer. For example, if  $x_1$  and  $x_3$  are constrained to take integer values then INTVAR(1) might be set to 1 and INTVAR(2) to 3. The order in which the indices are specified is important, since this determines the order in which the sub-problems are generated. As a rule-of-thumb, the important variables should always be specified first. Thus, in the above example, if  $x_3$  relates to a more important quantity than  $x_1$ , then it might be advantageous to set INTVAR(1) = 3 and INTVAR(2) = 1. If  $k$  is the smallest integer such that INTVAR( $k$ ) is less than or equal to zero then H02CBF assumes that  $k - 1$  variables are constrained to be integer; components INTVAR( $k + 1$ ), ..., INTVAR(LINTVR) are *not* referenced.
- 12:** LINTVR — INTEGER *Input*  
*On entry:* the dimension of the array INTVAR as declared in the calling (sub)program. Often LINTVR is the number of variables that are constrained to be integer.  
*Constraint:* LINTVR > 0.
- 13:** MDEPTH — INTEGER *Input*  
*On entry:* the maximum depth (i.e., number of extra constraints) that H02CBF may insert before admitting failure.  
*Suggested value:* MDEPTH =  $3 \times N/2$ .  
*Constraint:* MDEPTH  $\geq 1$ .

- 14:** ISTATE(N+NCLIN) — INTEGER array *Input/Output*  
*On entry:* ISTATE need not be set if the (default) **Cold Start** option is used.

If the **Warm Start** option has been chosen (see Section 11.2), ISTATE specifies the desired status of the constraints at the start of the feasibility phase. More precisely, the first  $n$  elements of ISTATE refer to the upper and lower bounds on the variables, and the next  $m_L$  elements refer to the general linear constraints (if any). Possible values for ISTATE( $j$ ) are as follows:

ISTATE( $j$ )	Meaning
0	The corresponding constraint should <i>not</i> be in the initial working set.
1	The constraint should be in the initial working set at its lower bound.
2	The constraint should be in the initial working set at its upper bound.
3	The constraint should be in the initial working set as an equality. This value must not be specified unless $BL(j) = BU(j)$ .

The values  $-2$ ,  $-1$  and  $4$  are also acceptable but will be reset to zero by the routine. If H02CBF has been called previously with the same values of N and NCLIN, ISTATE already contains satisfactory information. (See also the description of the optional parameter **Warm Start** in Section 11.2). The routine also adjusts (if necessary) the values supplied in XS to be consistent with ISTATE.

*Constraint:*  $-2 \leq ISTATE(j) \leq 4$ , for  $j = 1, 2, \dots, N+NCLIN$ .

*On exit:* the status of the constraints in the working set at the point returned in XS. The significance of each possible value of ISTATE( $j$ ) is as follows:

ISTATE( $j$ )	Meaning
$-2$	The constraint violates its lower bound by more than the feasibility tolerance.
$-1$	The constraint violates its upper bound by more than the feasibility tolerance.
0	The constraint is satisfied to within the feasibility tolerance, but is not in the working set.
1	This inequality constraint is included in the working set at its lower bound.
2	This inequality constraint is included in the working set at its upper bound.
3	This constraint is included in the working set as an equality. This value of ISTATE can occur only when $BL(j) = BU(j)$ .
4	This corresponds to optimality being declared with $XS(j)$ being temporarily fixed at its current value. This value of ISTATE can occur only when $IFAIL = 1$ on exit.

- 15:** XS(N+NCLIN) — *real* array *Input/Output*  
*On entry:* an initial estimate of the solution.

*On exit:* the point at which H02CBF terminated. If  $IFAIL = 0, 1$  or  $3$ , XS contains an estimate of the solution.

- 16:** OBJ — *real* *Output*  
*On exit:* the value of the objective function at  $x$  if  $x$  is feasible, or the sum of infeasibilities at  $x$  otherwise. If the problem is of type FP and  $x$  is feasible, OBJ is set to zero.

- 17:** AX(\*) — *real* array *Output*  
**Note:** the dimension of the array AX must be at least  $\max(1, NCLIN)$ .  
*On exit:* the final values of the linear constraints  $Ax$ .

If  $NCLIN = 0$  then AX is not referenced.

- 18:** CLAMDA(N+NCLIN) — *real* array *Output*  
*On exit:* the values of the Lagrange multipliers for each constraint with respect to the current working set. The first  $n$  elements contain the multipliers for the bound constraints on the variables, and the next  $m_L$  elements contain the multipliers for the general linear constraints (if any). If  $ISTATE(j) = 0$  (i.e., constraint  $j$  is not in the working set), CLAMDA( $j$ ) is zero. If  $x$  is optimal, CLAMDA( $j$ ) should be non-negative if  $ISTATE(j) = 1$ , non-positive if  $ISTATE(j) = 2$  and zero if  $ISTATE(j) = 4$ .

**19: STRGY — INTEGER***Input*

*On entry:* STRGY determines a branching strategy to be used throughout the computation, as follows:

STRGY	Meaning
0	Always left branch first i.e., impose an upper bound constraint on the variable first.
1	Always right branch first i.e., impose a lower bound constraint on the variable first.
2	Branch towards the nearest integer i.e., if $x_k = 2.4$ then impose an upper bound constraint $x_k \leq 2$ , whereas if $x_k = 2.6$ then impose the lower bound constraint $x_k \geq 3.0$ .
3	A random choice is made between a left-hand and a right-hand branch.

*Constraint:* STRGY = 0, 1, 2 or 3.

**20: IWRK(LIWRK) — INTEGER array***Workspace***21: LIWRK — INTEGER***Input*

*On entry:* the dimension of the array IWRK as declared in the (sub)program from which H02CBF is called.

*Constraint:* LIWRK  $\geq 2 \times N + 3 + 2 \times MDEPTH$ .

**22: WRK(LWRK) — *real* array***Workspace***23: LWRK — INTEGER***Input*

*On entry:* the dimension of the array WRK as declared in the (sub)program from which H02CBF is called.

*Constraints:*

For problems QP2 (the default) and QP4,

$$LWRK \geq 2 \times N^2 + 8 \times N + 5 \times NCLIN + 4 \times MDEPTH \text{ if } NCLIN > 0,$$

$$LWRK \geq N^2 + 9 \times N + 4 \times MDEPTH \text{ if } NCLIN = 0.$$

For problems QP1 and QP3,

$$LWRK \geq 2 \times N^2 + 8 \times N + 5 \times NCLIN + 4 \times MDEPTH \text{ if } NCLIN > 0,$$

$$LWRK \geq N^2 + 8 \times N + 4 \times MDEPTH \text{ if } NCLIN = 0.$$

If the problem is of type LP,

$$LWRK \geq 9 \times N + 1 + 4 \times MDEPTH \text{ if } NCLIN = 0,$$

$$LWRK \geq 2 \times N^2 + 9 \times N + 5 \times NCLIN + 4 \times MDEPTH \text{ if } NCLIN \geq N,$$

$$LWRK \geq 2 \times (NCLIN+1)^2 + 9 \times N + 5 \times NCLIN + 4 \times MDEPTH \text{ otherwise.}$$

If the problem is of type FP,

$$LWRK \geq 8 \times N + 1 + 4 \times MDEPTH, \text{ if } NCLIN = 0,$$

$$LWRK \geq 2 \times N^2 + 8 \times N + 5 \times NCLIN + 4 \times MDEPTH \text{ if } NCLIN \geq N,$$

$$LWRK \geq 2 \times (NCLIN+1)^2 + 8 \times N + 5 \times NCLIN + 4 \times MDEPTH \text{ otherwise.}$$

**24: MONIT — SUBROUTINE, supplied by the NAG Fortran Library or the user. *External Procedure***

This routine may be used to print out intermediate output and to affect the course of the computation. Specifically, it allows the user to specify a realistic value for the cut-off value (see Section 3) and to terminate the algorithm. If the user does not require any intermediate output, has no estimate of the cut-off value and requires an exhaustive tree search then MONIT may be the dummy routine H02CBU (CBUH02 in some implementations).

Its specification is:

```

SUBROUTINE MONIT(INTFND, NODES, DEPTH, OBJ, X(N), BSTVAL,
1          BSTSOL(N), BL(N), BU(N), N, HALT, COUNT)
INTEGER    INTFND, NODES, DEPTH, N, COUNT
real     OBJ, X(N), BSTVAL, BSTSOL(N), BL(N), BU(N)
LOGICAL    HALT

```

- 1:** INTFND — INTEGER *Input*  
*On entry:* specifies the number of integer solutions obtained so far.
- 2:** NODES — INTEGER *Input*  
*On entry:* specifies the number of nodes (sub-problems) solved so far.
- 3:** DEPTH — INTEGER *Input*  
*On entry:* specifies the depth in the tree of sub-problems the algorithm has now reached.
- 4:** OBJ — **real** *Input*  
*On entry:* specifies the value of the objective function of the end of the latest sub-problem.
- 5:** X(N) — **real** array *Input*  
*On entry:* specifies the values of the independent variables at the end of the latest sub-problem.
- 6:** BSTVAL — **real** *Input/Output*  
*On entry:* normally specifies the value of the best integer solution found so far.  
*On exit:* may be set a cut-off value by the sophisticated user as follows. Before an integer solution has been found BSTVAL will be set by H02CBF to the largest machine representable number (see X02ALF). If the user knows that the solution being sought is a much smaller number, then BSTVAL may be set to this number as a cut-off value (see Section 3). Beware of setting BSTVAL too small, since then no integer solutions will be discovered. Also make sure that BSTVAL is set using a statement of the form
- IF (INTFND.EQ.0) BSTVAL = *cut-off value*
- on entry* to MONIT. This statement will not prevent the normal operation of the algorithm when subsequent integer solutions are found. It would be a grievous mistake to unconditionally set BSTVAL and if you have any doubts whatsoever about the correct use of this parameter then you are strongly recommended to leave it unchanged.
- 7:** BSTSOL(N) — **real** array *Input*  
*On entry:* specifies the solution vector which gives rise to the best integer solution value so far discovered.
- 8:** BL(N) — **real** array *Input*  
*On entry:* BL(*i*) specifies the current lower bounds on the variable  $x_i$ .
- 9:** BU(N) — **real** array *Input*  
*On entry:* BU(*i*) specifies the current upper bounds on the variable  $x_i$ .
- 10:** N — INTEGER *Input*  
*On entry:* specifies the number of variables.
- 11:** HALT — LOGICAL *Input/Output*  
*On entry:* HALT will have the value .FALSE..  
*On exit:* by setting HALT to .TRUE., the user may terminate the algorithm prematurely. This facility may be useful if the user is content with *any* integer solution, or with any integer solution that fits certain criteria. Under these circumstances setting HALT = .TRUE. can save considerable unnecessary computation.
- 12:** COUNT — INTEGER *Input*  
*On entry:* specifies the number of integer solutions found so far.

MONIT must be declared as EXTERNAL in the (sub)program from which H02CBF is called. Parameters denoted as *Input* must **not** be changed by this procedure.

**25: IFAIL — INTEGER**

*Input/Output*

*On entry:* IFAIL must be set to 0,  $-1$  or 1. Users who are unfamiliar with this parameter should refer to Chapter P01 for details.

*On exit:* IFAIL = 0 unless the routine detects an error or gives a warning (see Section 6).

**For this routine**, because the values of output parameters may be useful even if IFAIL  $\neq$  0 on exit, users are recommended to set IFAIL to  $-1$  before entry. **It is then essential to test the value of IFAIL on exit.**

## 6 Error Indicators and Warnings

If on entry IFAIL = 0 or  $-1$ , explanatory error messages are output on the current error message unit (as defined by X04AAF).

Errors or warnings specified by the routine:

IFAIL =  $-1$

Algorithm terminated at user request (HALT = .TRUE.).

IFAIL = 1

Input parameter error immediately detected.

IFAIL = 2

No integer solution found. (Check that BSTVAL has not been set too small.)

IFAIL = 3

MDEPTH is too small. Increase the value of MDEPTH and re-enter H02CBF.

IFAIL = 4

The basic problem (without integer constraints) is unbounded.

IFAIL = 5

The basic problem is infeasible.

IFAIL = 6

The basic problem requires too many iterations.

IFAIL = 7

The basic problem has a reduced Hessian which exceeds its assigned dimension.

IFAIL = 8

The basic problem has an invalid parameter setting.

IFAIL = 9

The basic problem, as defined, is not standard.

IFAIL = 10

LIWRK is too small.

IFAIL = 11

LWRK is too small.

IFAIL = 12

An internal error has occurred within the routine. Please contact NAG with details of the call to H02CBF.

## 7 Accuracy

The routine implements a numerically stable active set strategy and returns solutions that are as accurate as the condition of the problem warrants on the machine.

## 8 Further Comments

This section contains some comments on scaling and a description of the printed output.

### 8.1 Scaling

Sensible scaling of the problem is likely to reduce the number of iterations required and make the problem less sensitive to perturbations in the data, thus improving the condition of the problem. In the absence of better information it is usually sensible to make the Euclidean lengths of each constraint of comparable magnitude. See the E04 Chapter Introduction and Gill *et al.*[7] for further information and advice.

### 8.2 Description of the Printed Output

This section describes the (default) intermediate printout and final printout produced by H02CBF. The intermediate printout is a subset of the monitoring information produced by the routine at every iteration (see Section 12). The level of printed output can be controlled by the user (see the description of the optional parameter **Print Level** in Section 11.2). Note that the intermediate printout and final printout are produced only if **Print Level**  $\geq 10$  (the default).

The following line of summary output (< 80 characters) is produced at every iteration. In all cases, the values of the quantities printed are those in effect *oncompletion* of the given iteration.

<b>Itn</b>	is the iteration count.
<b>Step</b>	is the step taken along the computed search direction. If a constraint is added during the current iteration, <b>Step</b> will be the step to the nearest constraint. When the problem is of type LP, the step can be greater than one during the optimality phase.
<b>Ninf</b>	is the number of violated constraints (infeasibilities). This will be zero during the optimality phase.
<b>Sinf/Objective</b>	is the value of the current objective function. If $x$ is not feasible, <b>Sinf</b> gives a weighted sum of the magnitudes of constraint violations. If $x$ is feasible, <b>Objective</b> is the value of the objective function. The output line for the final iteration of the feasibility phase (i.e., the first iteration for which <b>Ninf</b> is zero) will give the value of the true objective at the first feasible point. During the optimality phase, the value of the objective function will be non-increasing. During the feasibility phase, the number of constraint infeasibilities will not increase until either a feasible point is found, or the optimality of the multipliers implies that no feasible point exists. Once optimal multipliers are obtained, the number of infeasibilities can increase, but the sum of infeasibilities will either remain constant or be reduced until the minimum sum of infeasibilities is found.
<b>Norm Gz</b>	is $\ Z_R^T g_{FR}\ $ , the Euclidean norm of the reduced gradient with respect to $Z_R$ (see Section 10.2 and Section 10.4). During the optimality phase, this norm will be approximately zero after a unit step.

The final printout includes a listing of the status of every variable and constraint.

The following describes the printout for each variable. A full stop (.) is printed for any numerical value that is zero.

<b>Varbl</b>	gives the name (V) and index $j$ , for $j = 1, 2, \dots, n$ of the variable.
<b>State</b>	gives the state of the variable (FR if neither bound is in the working set, EQ if a fixed variable, LL if on its lower bound, UL if on its upper bound, TF if temporarily fixed at its current value). If <b>Value</b> lies outside the upper or lower bounds by more than the <b>Feasibility Tolerance</b> (default value = $\sqrt{\epsilon}$ , where $\epsilon$ is the <i>machine precision</i> ; see Section 11.2), <b>State</b> will be ++ or -- respectively.

A key is sometimes printed before **State** to give some additional information about the state of a variable.

- A *Alternative optimum possible.* The variable is active at one of its bounds, but its Lagrange multiplier is essentially zero. This means that if the variable were allowed to start moving away from its bound, there would be no change to the objective function. The values of the other free variables *might* change, giving a genuine alternative solution. However, if there are any degenerate variables (labelled D), the actual change might prove to be zero, since one of them could encounter a bound immediately. In either case the values of the Lagrange multipliers might also change.
- D *Degenerate.* The variable is free, but it is equal to (or very close to) one of its bounds.
- I *Infeasible.* The variable is currently violating one of its bounds by more than the **Feasibility Tolerance**.

Value	is the value of the variable at the final iterate.
Lower Bound	is the lower bound specified for the variable. <b>None</b> indicates that $BL(j) \leq -bigbnd$ .
Upper Bound	is the upper bound specified for the variable. <b>None</b> indicates that $BU(j) \geq bigbnd$ .
Lagr Mult	is the Lagrange multiplier for the associated bound. This will be zero if <b>State</b> is <b>FR</b> unless $BL(j) \leq -bigbnd$ and $BU(j) \geq bigbnd$ , in which case the entry will be blank. If $x$ is optimal, the multiplier should be non-negative if <b>State</b> is <b>LL</b> , and non-positive if <b>State</b> is <b>UL</b> .
Slack	is the difference between the variable <b>Value</b> and the nearer of its (finite) bounds $BL(j)$ and $BU(j)$ . A blank entry indicates that the associated variable is not bounded (i.e., $BL(j) \leq -bigbnd$ and $BU(j) \geq bigbnd$ ).

The meaning of the printout for general constraints is the same as that given above for variables, with ‘variable’ replaced by ‘constraint’,  $BL(j)$  and  $BU(j)$  are replaced by  $BL(n+j)$  and  $BU(n+j)$  respectively, and with the following change in the heading:

L Con                      gives the name (L) and index  $j$ , for  $j = 1, 2, \dots, m_L$  of the linear constraint.

Note that movement off a constraint (as opposed to a variable moving away from its bound) can be interpreted as allowing the entry in the **Slack** column to become positive.

Numerical values are output with a fixed number of digits; they are not guaranteed to be accurate to this precision.

## 9 Example

To minimize the quadratic function  $f(x) = c^T x + \frac{1}{2} x^T H x$ , where

$$c = (-0.02, -0.2, -0.2, -0.2, -0.2, 0.04, 0.04)^T$$

$$H = \begin{pmatrix} 2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 2 & 0 & 0 & 0 \\ 0 & 0 & 2 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & -2 & -2 \\ 0 & 0 & 0 & 0 & 0 & -2 & -2 \end{pmatrix}$$

subject to the bounds

$$\begin{aligned} -0.01 &\leq x_1 \leq 0.01 \\ -0.1 &\leq x_2 \leq 0.15 \\ -0.01 &\leq x_3 \leq 0.03 \\ -0.04 &\leq x_4 \leq 0.02 \\ -0.1 &\leq x_5 \leq 0.05 \\ -0.01 &\leq x_6 \\ -0.01 &\leq x_7 \end{aligned}$$

to the general constraints

$$\begin{array}{rcl}
 x_1 + & x_2 + & x_3 + & x_4 + & x_5 + & x_6 + & x_7 = & -0.13 \\
 0.15x_1 + 0.04x_2 + 0.02x_3 + 0.04x_4 + 0.02x_5 + 0.01x_6 + 0.03x_7 & \leq & -0.0049 \\
 0.03x_1 + 0.05x_2 + 0.08x_3 + 0.02x_4 + 0.06x_5 + 0.01x_6 & \leq & -0.0064 \\
 0.02x_1 + 0.04x_2 + 0.01x_3 + 0.02x_4 + 0.02x_5 & \leq & -0.0037 \\
 0.02x_1 + 0.03x_2 & & + 0.01x_5 & \leq & -0.0012 \\
 -0.0992 \leq 0.70x_1 + 0.75x_2 + 0.80x_3 + 0.75x_4 + 0.80x_5 + 0.97x_6 & & & & & & & \\
 -0.003 \leq 0.02x_1 + 0.06x_2 + 0.08x_3 + 0.12x_4 + 0.02x_5 + 0.01x_6 + 0.97x_7 & \leq & 0.002
 \end{array}$$

and the variable  $x_4$  is constrained to be integer.

The initial point, which is infeasible, is

$$x_0 = (-0.01, -0.03, 0.0, -0.01, -0.1, 0.02, 0.01)^T.$$

The optimal solution (to five figures) is

$$x^* = (-0.01, -0.073328, -0.00025809, 0.0, -0.063354, 0.014109, 0.0028312)^T.$$

The document for H02CCF includes an example program to solve the same problem using some of the optional parameters described in Section 11.

## 9.1 Program Text

**Note.** The listing of the example program presented below uses bold italicised terms to denote precision-dependent details. Please read the Users' Note for your implementation to check the interpretation of these terms. As explained in the Essential Introduction to this manual, the results produced may not be identical for all implementations.

```

*      H02CBF Example Program Text.
*      Mark 19 Release. NAG Copyright 1999.
*      .. Parameters ..
      INTEGER          NIN, NOUT, LINTVR
      PARAMETER        (NIN=5,NOUT=6,LINTVR=1)
      INTEGER          NMAX, NCMAX
      PARAMETER        (NMAX=10,NCMAX=10)
      INTEGER          LDA, LDH
      PARAMETER        (LDA=NCMAX,LDH=NMAX)
      INTEGER          LIWORK, LWORK, MDEPTH
      PARAMETER        (LIWORK=1000,LWORK=10000,MDEPTH=30)
*      .. Local Scalars ..
      real            OBJ
      INTEGER          I, IFAIL, J, N, NCLIN, STRGY
*      .. Local Arrays ..
      real            A(LDA,NMAX), AX(NCMAX), BL(NMAX+NCMAX),
+                   BU(NMAX+NCMAX), CLAMDA(NMAX+NCMAX), CVEC(NMAX),
+                   H(LDH,NMAX), WORK(LWORK), X(NMAX)
      INTEGER          INTVAR(LINTVR), ISTATE(NMAX+NCMAX), IWORK(LIWORK)
*      .. External Subroutines ..
      EXTERNAL         E04NFU, H02CBF, H02CBU, H02CDF
*      .. Executable Statements ..
      WRITE (NOUT,*) 'H02CBF Example Program Results'
*      Skip heading in data file
      READ (NIN,*)
      READ (NIN,*) N, NCLIN
      IF (N.LE.NMAX .AND. NCLIN.LE.NCMAX) THEN
*
*          Read CVEC, A, BL, BU, X and H from data file
*
      READ (NIN,*) (CVEC(I),I=1,N)
      READ (NIN,*) ((A(I,J),J=1,N),I=1,NCLIN)
      READ (NIN,*) (BL(I),I=1,N+NCLIN)

```

```

READ (NIN,*) (BU(I),I=1,N+NCLIN)
READ (NIN,*) (X(I),I=1,N)
READ (NIN,*) ((H(I,J),J=1,N),I=1,N)
*
STRTRY = 2
INTVAR(1) = 4
*
CALL HO2CDF('NoList')
CALL HO2CDF('Print Level = 0')
*
Solve the problem
*
IFAIL = 0
*
CALL HO2CBF(N,NCLIN,A,LDA,BL,BU,CVEC,H,LDH,E04NFU,INTVAR,
+          LINTVR,MDEPTH,ISTATE,X,OBJ,AX,CLAMDA,STRTRY,IWORK,
+          LIWORK,WORK,LWORK,HO2CBU,IFAIL)
*
Print out the best integer solution found
*
WRITE (NOUT,99999) OBJ, (I,X(I),I=1,N)
*
END IF
STOP
*
99999 FORMAT (' Optimal Integer Value is = ',E20.8,/' Components are ',
+          /(' x(',I3,') = ',F15.8))
END

```

## 9.2 Program Data

### H02CBF Example Program Data

```

7 7                                     :Values of N and NCLIN
-0.02 -0.20 -0.20 -0.20 -0.20  0.04  0.04 :End of CVEC
 1.00  1.00  1.00  1.00  1.00  1.00  1.00
 0.15  0.04  0.02  0.04  0.02  0.01  0.03
 0.03  0.05  0.08  0.02  0.06  0.01  0.00
 0.02  0.04  0.01  0.02  0.02  0.00  0.00
 0.02  0.03  0.00  0.00  0.01  0.00  0.00
 0.70  0.75  0.80  0.75  0.80  0.97  0.00
 0.02  0.06  0.08  0.12  0.02  0.01  0.97 :End of matrix A
-0.01 -0.10 -0.01 -0.04 -0.10 -0.01 -0.01
-0.13 -1.0E+25 -1.0E+25 -1.0E+25 -1.0E+25 -9.92E-02 -3.0E-03 :End of BL
 0.01  0.15  0.03  0.02  0.05  1.0E+25  1.0E+25
-0.13 -4.9E-03 -6.4E-03 -3.7E-03 -1.2E-03  1.0E+25  2.0E-03 :End of BU
-0.01 -0.03  0.00 -0.01 -0.10  0.02  0.01 :End of X
 2.00  0.00  0.00  0.00  0.00  0.00  0.00
 0.00  2.00  0.00  0.00  0.00  0.00  0.00
 0.00  0.00  2.00  2.00  0.00  0.00  0.00
 0.00  0.00  0.00  0.00  2.00  0.00  0.00
 0.00  0.00  0.00  0.00  0.00 -2.00 -2.00
 0.00  0.00  0.00  0.00  0.00 -2.00 -2.00 :End of matrix H

```

### 9.3 Program Results

```

H02CBF Example Program Results
Optimal Integer Value is =      0.37469662E-01
Components are
x( 1) =      -0.01000000
x( 2) =      -0.07332830
x( 3) =      -0.00025809
x( 4) =       0.00000000
x( 5) =      -0.06335433
x( 6) =       0.01410944
x( 7) =       0.00283128

```

The remainder of this document is intended for more advanced users. Section 10 contains a detailed description of the algorithm which may be needed in order to understand Section 11 and Section 12. Section 11 describes the optional parameters which may be set by calls to *H02CCF* and/or *H02CDF*. Section 12 describes the quantities which can be requested to monitor the course of the computation.

## 10 Algorithmic Details

H02CBF implements a basic Branch and bound algorithm (see Section 3) using E04NFF as its basic sub-problem solver. See below for details of its algorithm.

### 10.1 Overview

H02CBF is based on an inertia-controlling method that maintains a Cholesky factorization of the reduced Hessian (see below). The method is based on that of Gill and Murray [2], and is described in detail by Gill *et al.* [5]. Here we briefly summarize the main features of the method. Where possible, explicit reference is made to the names of variables that are parameters of H02CBF or appear in the printed output. H02CBF has two phases: finding an initial feasible point by minimizing the sum of infeasibilities (the *feasibility phase*), and minimizing the quadratic objective function within the feasible region (the *optimality phase*). The computations in both phases are performed by the same subroutines. The two-phase nature of the algorithm is reflected by changing the function being minimized from the sum of infeasibilities to the quadratic objective function. The feasibility phase does *not* perform the standard simplex method (i.e., it does not necessarily find a vertex), except in the LP case when  $m_L \leq n$ . Once any iterate is feasible, all subsequent iterates remain feasible.

H02CBF has been designed to be efficient when used to solve a *sequence* of related problems – for example, within a sequential quadratic programming method for nonlinearly constrained optimization (e.g., E04UCF). In particular, the user may specify an initial working set (the indices of the constraints believed to be satisfied exactly at the solution); see the discussion of the optional parameter **Warm Start** in Section 11.2.

In general, an iterative process is required to solve a quadratic program. (For simplicity, we shall always consider a typical iteration and avoid reference to the index of the iteration.) Each new iterate  $\bar{x}$  is defined by

$$\bar{x} = x + \alpha p \tag{1}$$

where the *step length*  $\alpha$  is a non-negative scalar, and  $p$  is called the *search direction*.

At each point  $x$ , a working set of constraints is defined to be a linearly independent subset of the constraints that are satisfied ‘exactly’ (to within the tolerance defined by the optional parameter **Feasibility Tolerance**; see Section 11.2). The working set is the current prediction of the constraints that hold with equality at the solution of a linearly constrained QP problem. The search direction is constructed so that the constraints in the working set remain *unaltered* for any value of the step length. For a bound constraint in the working set, this property is achieved by setting the corresponding element of the search direction to zero. Thus, the associated variable is *fixed*, and specification of the working set induces a partition of  $x$  into *fixed* and *free* variables. During a given iteration, the fixed variables are

effectively removed from the problem; since the relevant elements of the search direction are zero, the columns of  $A$  corresponding to fixed variables may be ignored.

Let  $m_W$  denote the number of general constraints in the working set and let  $n_{FX}$  denote the number of variables fixed at one of their bounds ( $m_W$  and  $n_{FX}$  are the quantities **Lin** and **Bnd** in the monitoring file output from H02CBF; see Section 12). Similarly, let  $n_{FR}$  ( $n_{FR} = n - n_{FX}$ ) denote the number of free variables. At every iteration, *the variables are re-ordered so that the last  $n_{FX}$  variables are fixed*, with all other relevant vectors and matrices ordered accordingly.

## 10.2 Definition of the Search Direction

Let  $A_{FR}$  denote the  $m_W$  by  $n_{FR}$  sub-matrix of general constraints in the working set corresponding to the free variables, and let  $p_{FR}$  denote the search direction with respect to the free variables only. The general constraints in the working set will be unaltered by any move along  $p$  if

$$A_{FR}p_{FR} = 0. \quad (2)$$

In order to compute  $p_{FR}$ , the  $TQ$  factorization of  $A_{FR}$  is used:

$$A_{FR}Q_{FR} = (0 \ T), \quad (3)$$

where  $T$  is a non-singular  $m_W$  by  $m_W$  upper triangular matrix (i.e.,  $t_{ij} = 0$  if  $i > j$ ), and the non-singular  $n_{FR}$  by  $n_{FR}$  matrix  $Q_{FR}$  is the product of orthogonal transformations (see Gill *et al.* [3]). If the columns of  $Q_{FR}$  are partitioned so that

$$Q_{FR} = (Z \ Y),$$

where  $Y$  is  $n_{FR}$  by  $m_W$ , then the  $n_Z$  ( $n_Z = n_{FR} - m_W$ ) columns of  $Z$  form a basis for the null space of  $A_{FR}$ . Let  $n_R$  be an integer such that  $0 \leq n_R \leq n_Z$ , and let  $Z_R$  denote a matrix whose  $n_R$  columns are a subset of the columns of  $Z$ . (The integer  $n_R$  is the quantity **Zr** in the monitoring output from H02CBF. In many cases,  $Z_R$  will include *all* the columns of  $Z$ .) The direction  $p_{FR}$  will satisfy (2) if

$$p_{FR} = Z_R p_R, \quad (4)$$

where  $p_R$  is any  $n_R$ -vector.

Let  $Q$  denote the  $n$  by  $n$  matrix

$$Q = \begin{pmatrix} Q_{FR} & \\ & I_{FX} \end{pmatrix},$$

where  $I_{FX}$  is the identity matrix of order  $n_{FX}$ . Let  $H_Q$  and  $g_Q$  denote the  $n$  by  $n$  *transformed Hessian* and *transformed gradient*

$$H_Q = Q^T H Q \quad \text{and} \quad g_Q = Q^T (c + Hx)$$

and let the matrix of first  $n_R$  rows and columns of  $H_Q$  be denoted by  $H_R$  and the vector of the first  $n_R$  elements of  $g_Q$  be denoted by  $g_R$ . The quantities  $H_R$  and  $g_R$  are known as the *reduced Hessian* and *reduced gradient* of  $f(x)$ , respectively. Roughly speaking,  $g_R$  and  $H_R$  describe the first and second derivatives of an *unconstrained* problem for the calculation of  $p_R$ .

At each iteration, a triangular factorization of  $H_R$  is available. If  $H_R$  is positive-definite,  $H_R = R^T R$ , where  $R$  is the upper triangular Cholesky factor of  $H_R$ . If  $H_R$  is not positive-definite,  $H_R = R^T D R$ , where  $D = \text{diag}(1, 1, \dots, 1, \mu)$ , with  $\mu \leq 0$ .

The computation is arranged so that the reduced-gradient vector is a multiple of  $e_R$ , a vector of all zeros except in the last (i.e.,  $n_R$ th) position. This allows the vector  $p_R$  in (4) to be computed from a single back-substitution

$$R p_R = \gamma e_R \quad (5)$$

where  $\gamma$  is a scalar that depends on whether or not the reduced Hessian is positive-definite at  $x$ . In the positive-definite case,  $x + p$  is the minimizer of the objective function subject to the constraints (bounds and general) in the working set treated as equalities. If  $H_R$  is not positive-definite,  $p_R$  satisfies the conditions

$$p_R^T H_R p_R < 0 \quad \text{and} \quad g_R^T p_R \leq 0,$$

which allow the objective function to be reduced by any positive step of the form  $x + \alpha p$ .

### 10.3 The Main Iteration

If the reduced gradient is zero,  $x$  is a constrained stationary point in the subspace defined by  $Z$ . During the feasibility phase, the reduced gradient will usually be zero only at a vertex (although it may be zero at non-vertices in the presence of constraint dependencies). During the optimality phase, a zero reduced gradient implies that  $x$  minimizes the quadratic objective when the constraints in the working set are treated as equalities. At a constrained stationary point, Lagrange multipliers  $\lambda_C$  and  $\lambda_B$  for the general and bound constraints are defined from the equations

$$A_{\text{FR}}^T \lambda_C = g_{\text{FR}} \quad \text{and} \quad \lambda_B = g_{\text{FX}} - A_{\text{FX}}^T \lambda_C. \quad (6)$$

Given a positive constant  $\delta$  of the order of the *machine precision*, a Lagrange multiplier  $\lambda_j$  corresponding to an inequality constraint in the working set is said to be *optimal* if  $\lambda_j \leq \delta$  when the associated constraint is at its *upper bound*, or if  $\lambda_j \geq -\delta$  when the associated constraint is at its *lower bound*. If a multiplier is non-optimal, the objective function (either the true objective or the sum of infeasibilities) can be reduced by deleting the corresponding constraint (with index `Jdel`; see Section 12) from the working set.

If optimal multipliers occur during the feasibility phase and the sum of infeasibilities is non-zero, there is no feasible point, and the user can force H02CBF to continue until the minimum value of the sum of infeasibilities has been found; see the discussion of the optional parameter **Minimum Sum of Infeasibilities** in Section 11.2. At such a point, the Lagrange multiplier  $\lambda_j$  corresponding to an inequality constraint in the working set will be such that  $-(1 + \delta) \leq \lambda_j \leq \delta$  when the associated constraint is at its *upper bound*, and  $-\delta \leq \lambda_j \leq (1 + \delta)$  when the associated constraint is at its *lower bound*. Lagrange multipliers for equality constraints will satisfy  $|\lambda_j| \leq 1 + \delta$ .

If the reduced gradient is not zero, Lagrange multipliers need not be computed and the non-zero elements of the search direction  $p$  are given by  $Z_{\text{RPR}}$  (see (4) and (5)). The choice of step length is influenced by the need to maintain feasibility with respect to the satisfied constraints. If  $H_R$  is positive-definite and  $x + p$  is feasible,  $\alpha$  will be taken as unity. In this case, the reduced gradient at  $\bar{x}$  will be zero, and Lagrange multipliers are computed. Otherwise,  $\alpha$  is set to  $\alpha_M$ , the step to the ‘nearest’ constraint (with index `Jadd`; see Section 12), which is added to the working set at the next iteration.

Each change in the working set leads to a simple change to  $A_{\text{FR}}$ : if the status of a general constraint changes, a *row* of  $A_{\text{FR}}$  is altered; if a bound constraint enters or leaves the working set, a *column* of  $A_{\text{FR}}$  changes. Explicit representations are recurred of the matrices  $T$ ,  $Q_{\text{FR}}$  and  $R$ ; and of vectors  $Q^T g$ , and  $Q^T c$ . The triangular factor  $R$  associated with the reduced Hessian is only updated during the optimality phase.

One of the most important features of H02CBF is its control of the conditioning of the working set, whose nearness to linear dependence is estimated by the ratio of the largest to smallest diagonal elements of the  $TQ$  factor  $T$  (the printed value `Cond T`; see Section 12). In constructing the initial working set, constraints are excluded that would result in a large value of `Cond T`.

H02CBF includes a rigorous procedure that prevents the possibility of cycling at a point where the active constraints are nearly linearly dependent (see Gill *et al.* [4]). The main feature of the anti-cycling procedure is that the feasibility tolerance is increased slightly at the start of every iteration. This not only allows a positive step to be taken at every iteration, but also provides, whenever possible, a *choice* of constraints to be added to the working set. Let  $\alpha_M$  denote the maximum step at which  $x + \alpha_M p$  does not violate any constraint by more than its feasibility tolerance. All constraints at a distance  $\alpha$  ( $\alpha \leq \alpha_M$ ) along  $p$  from the current point are then viewed as acceptable candidates for inclusion in the working set. The constraint whose normal makes the largest angle with the search direction is added to the working set.

### 10.4 Choosing the Initial Working Set

At the start of the optimality phase, a positive-definite  $H_R$  can be defined if enough constraints are included in the initial working set. (The matrix with no rows and columns is positive-definite by definition, corresponding to the case when  $A_{\text{FR}}$  contains  $n_{\text{FR}}$  constraints.) The idea is to include as many general constraints as necessary to ensure that the reduced Hessian is positive-definite.

Let  $H_Z$  denote the matrix of the first  $n_Z$  rows and columns of the matrix  $H_Q = Q^T H Q$  at the beginning of the optimality phase. A partial Cholesky factorization is used to find an upper triangular matrix

$R$  that is the factor of the largest positive-definite leading sub-matrix of  $H_Z$ . The use of interchanges during the factorization of  $H_Z$  tends to maximize the dimension of  $R$ . (The condition of  $R$  may be controlled using the optional parameter **Rank Tolerance**; see Section 11.2.) Let  $Z_R$  denote the columns of  $Z$  corresponding to  $R$ , and let  $Z$  be partitioned as  $Z = (Z_R \ Z_A)$ . A working set for which  $Z_R$  defines the null space can be obtained by including *the rows of  $Z_A^T$*  as ‘artificial constraints’. Minimization of the objective function then proceeds within the subspace defined by  $Z_R$ , as described in Section 10.2.

The artificially augmented working set is given by

$$\bar{A}_{\text{FR}} = \begin{pmatrix} Z_A^T \\ A_{\text{FR}} \end{pmatrix}, \quad (7)$$

so that  $p_{\text{FR}}$  will satisfy  $A_{\text{FR}}p_{\text{FR}} = 0$  and  $Z_A^T p_{\text{FR}} = 0$ . By definition of the  $TQ$  factorization,  $\bar{A}_{\text{FR}}$  automatically satisfies the following:

$$\bar{A}_{\text{FR}}Q_{\text{FR}} = \begin{pmatrix} Z_A^T \\ A_{\text{FR}} \end{pmatrix}Q_{\text{FR}} = \begin{pmatrix} Z_A^T \\ A_{\text{FR}} \end{pmatrix}(Z_R \ Z_A \ Y) = (0 \ \bar{T}),$$

where

$$\bar{T} = \begin{pmatrix} I & 0 \\ 0 & T \end{pmatrix},$$

and hence the  $TQ$  factorization of (7) is available trivially from  $T$  and  $Q_{\text{FR}}$  without additional expense.

The matrix  $Z_A$  is not kept fixed, since its role is purely to define an appropriate null space; the  $TQ$  factorization can therefore be updated in the normal fashion as the iterations proceed. No work is required to ‘delete’ the artificial constraints associated with  $Z_A$  when  $Z_R^T g_{\text{FR}} = 0$ , since this simply involves repartitioning  $Q_{\text{FR}}$ . The ‘artificial’ multiplier vector associated with the rows of  $Z_A^T$  is equal to  $Z_A^T g_{\text{FR}}$ , and the multipliers corresponding to the rows of the ‘true’ working set are the multipliers that would be obtained if the artificial constraints were not present. If an artificial constraint is ‘deleted’ from the working set, an **A** appears alongside the entry in the **Jdel** column of the monitoring file output (see Section 12).

The number of columns in  $Z_A$  and  $Z_R$ , the Euclidean norm of  $Z_R^T g_{\text{FR}}$ , and the condition estimator of  $R$  appear in the monitoring file output as **Art**, **Zr**, **Norm Gz** and **Cond Rz** respectively (see Section 12).

Under some circumstances, a different type of artificial constraint is used when solving a linear program. Although the algorithm of H02CBF does not usually perform simplex steps (in the traditional sense), there is one exception: a linear program with fewer general constraints than variables (i.e.,  $m_L \leq n$ ). (Use of the simplex method in this situation leads to savings in storage.) At the starting point, the ‘natural’ working set (the set of constraints exactly or nearly satisfied at the starting point) is augmented with a suitable number of ‘temporary’ bounds, each of which has the effect of temporarily fixing a variable at its current value. In subsequent iterations, a temporary bound is treated as a standard constraint until it is deleted from the working set, in which case it is never added again. If a temporary bound is ‘deleted’ from the working set, an **F** (for ‘Fixed’) appears alongside the entry in the **Jdel** column of the monitoring file output (see Section 12).

## 11 Optional Parameters

Several optional parameters in H02CBF define choices in the problem specification or the algorithm logic. In order to reduce the number of formal parameters of H02CBF these optional parameters have associated *default values* that are appropriate for most problems. Therefore, the user need only specify those optional parameters whose values are to be different from their default values.

The remainder of this section can be skipped by users who wish to use the default values for *all* optional parameters. A complete list of optional parameters and their default values is given in Section 11.1.

Optional parameters may be specified by calling one, or both, of the routines H02CCF and H02CDF prior to a call to H02CBF.

H02CCF reads options from an external options file, with **Begin** and **End** as the first and last lines respectively and each intermediate line defining a single optional parameter. For example,

```

Begin
  Print Level = 5
End

```

The call

```
CALL H02CCF (IOPTNS, INFORM)
```

can then be used to read the file on unit IOPTNS. INFORM will be zero on successful exit. H02CCF should be consulted for a full description of this method of supplying optional parameters.

H02CDF can be called to supply options directly, one call being necessary for each optional parameter. For example,

```
CALL H02CDF ('Print Level = 5')
```

H02CDF should be consulted for a full description of this method of supplying optional parameters.

All optional parameters not specified by the user are set to their default values. Optional parameters specified by the user are unaltered by H02CBF (unless they define invalid values) and so remain in effect for subsequent calls unless altered by the user.

## 11.1 Optional Parameter Checklist and Default Values

For easy reference, the following list shows all the valid keywords and their default values. The symbol  $\epsilon$  represents the *machine precision* (see X02AJF).

Optional Parameters	Default Values
Check frequency	50
Cold/Warm start	<b>Cold Start</b>
Crash tolerance	0.01
Defaults	
Expand frequency	5
Feasibility phase iteration limit	$\max(50, 5(n + m_L))$
Feasibility tolerance	$\sqrt{\epsilon}$
Hessian rows	$n$
Infinite bound size	$10^{20}$
Infinite step size	$\max(\text{bigbnd}, 10^{20})$
Iteration limit	$\max(50, 5(n + m_L))$
List/Nolist	<b>List</b>
Maximum degrees of freedom	$n$
Minimum sum of infeasibilities	<b>No</b>
Monitoring file	-1
Optimality phase iteration limit	$\max(50, 5(n + m_L))$
Optimality tolerance	$\epsilon^{0.8}$
Print level	10
Problem type	QP2
Rank tolerance	$100\epsilon$

## 11.2 Description of the Optional Parameters

The following list (in alphabetical order) gives the valid options. For each option, we give the keyword, any essential optional qualifiers, the default value, and the definition. The minimum abbreviation of each keyword is underlined. If no characters of an optional qualifier are underlined, the qualifier may be omitted. The letter *a* denotes a phrase (character string) that qualifies an option. The letters *i* and *r* denote INTEGER and *real* values required with certain options. The number  $\epsilon$  is a generic notation for *machine precision* (see X02AJF).

**Check Frequency**  $i$  Default = 50

Every  $i$ th iteration, a numerical test is made to see if the current solution  $x$  satisfies the constraints in the working set. If the largest residual of the constraints in the working set is judged to be too large, the current working set is refactorized and the variables are recomputed to satisfy the constraints more accurately. If  $i \leq 0$ , the default value is used.

**Cold Start** Default = **Cold Start**

**Warm Start**

This option specifies how the initial working set is chosen. With a **Cold Start**, H02CBF chooses the initial working set based on the values of the variables and constraints at the initial point. Broadly speaking, the initial working set will include equality constraints and bounds or inequality constraints that violate or ‘nearly’ satisfy their bounds (to within **Crash Tolerance**; see below).

With a **Warm Start**, the user must provide a valid definition of every element of the array ISTATE (see Section 5 for the definition of this array). H02CBF will override the user’s specification of ISTATE if necessary, so that a poor choice of the working set will not cause a fatal error. For instance, any elements of ISTATE which are set to  $-2$ ,  $-1$  or  $4$  will be reset to zero, as will any elements which are set to  $3$  when the corresponding elements of BL and BU are not equal. A warm start will be advantageous if a good estimate of the initial working set is available – for example, when H02CBF is called repeatedly to solve related problems.

**Crash Tolerance**  $r$  Default = 0.01

This value is used in conjunction with the optional parameter **Cold Start** (the default value) when H02CBF selects an initial working set. If  $0 \leq r \leq 1$ , the initial working set will include (if possible) bounds or general inequality constraints that lie within  $r$  of their bounds. In particular, a constraint of the form  $a_j^T x \geq l$  will be included in the initial working set if  $|a_j^T x - l| \leq r(1 + |l|)$ . If  $r < 0$  or  $r > 1$ , the default value is used.

**Defaults**

This special keyword may be used to reset all optional parameters to their default values.

**Expand Frequency**  $i$  Default = 5

This option is part of an anti-cycling procedure designed to guarantee progress even on highly degenerate problems.

The strategy is to force a positive step at every iteration, at the expense of violating the constraints by a small amount. Suppose that the value of the optional parameter **Feasibility Tolerance** is  $\delta$ . Over a period of  $i$  iterations, the feasibility tolerance actually used by H02CBF (i.e., the *working* feasibility tolerance) increases from  $0.5\delta$  to  $\delta$  (in steps of  $0.5\delta/i$ ).

At certain stages the following ‘resetting procedure’ is used to remove constraint infeasibilities. First, all variables whose upper or lower bounds are in the working set are moved exactly onto their bounds. A count is kept of the number of non-trivial adjustments made. If the count is positive, iterative refinement is used to give variables that satisfy the working set to (essentially) *machine precision*. Finally, the working feasibility tolerance is reinitialized to  $0.5\delta$ .

If a problem requires more than  $i$  iterations, the resetting procedure is invoked and a new cycle of  $i$  iterations is started with  $i$  incremented by 10. (The decision to resume the feasibility phase or optimality phase is based on comparing any constraint infeasibilities with  $\delta$ .)

The resetting procedure is also invoked when H02CBF reaches an apparently optimal, infeasible or unbounded solution, unless this situation has already occurred twice. If any non-trivial adjustments are made, iterations are continued.

If  $i \leq 0$ , the default value is used. If  $i \geq 9999999$ , no anti-cycling procedure is invoked.

**Feasibility Phase Iteration Limit**  $i_1$  Default =  $\max(50, 5(n + m_L))$

**Optimality Phase Iteration Limit**  $i_2$  Default =  $\max(50, 5(n + m_L))$

The scalars  $i_1$  and  $i_2$  specify the maximum number of iterations allowed in the feasibility and optimality phases. **Optimality Phase Iteration Limit** is equivalent to **Iteration Limit**. Setting  $i_1 = 0$  and

**Print Level**  $> 0$  means that the workspace needed will be computed and printed, but no iterations will be performed. If  $i_1 < 0$  or  $i_2 < 0$ , the default value is used.

**Feasibility Tolerance**  $r$  Default =  $\sqrt{\epsilon}$

If  $r \geq \epsilon$ ,  $r$  defines the maximum acceptable *absolute* violation in each constraint at a ‘feasible’ point. For example, if the variables and the coefficients in the general constraints are of order unity, and the latter are correct to about 6 decimal digits, it would be appropriate to specify  $r$  as  $10^{-6}$ . If  $0 \leq r < \epsilon$ , the default value is used.

H02CBF attempts to find a feasible solution before optimizing the objective function. If the sum of infeasibilities cannot be reduced to zero, the optional parameter **Minimum Sum of Infeasibilities** (see below) can be used to find the minimum value of the sum. Let **Sinf** be the corresponding sum of infeasibilities. If **Sinf** is quite small, it may be appropriate to raise  $r$  by a factor of 10 or 100. Otherwise, some error in the data should be suspected.

Note that a ‘feasible solution’ is a solution that satisfies the current constraints to within the tolerance  $r$ .

**Hessian Rows**  $i$  Default =  $n$

Note that this option does not apply to problems of type FP or LP.

This specifies  $m$ , the number of rows of the Hessian matrix  $H$ . The default value of  $m$  is  $n$ , the number of variables of the problem.

If the problem is of type QP,  $m$  will usually be  $n$ , the number of variables. However, a value of  $m$  less than  $n$  is appropriate for QP3 or QP4 if  $H$  is an upper trapezoidal matrix with  $m$  rows. Similarly,  $m$  may be used to define the dimension of a leading block of non-zeros in the Hessian matrices of QP1 or QP2, in which case the last  $n - m$  rows and columns of  $H$  are assumed to be zero. In the QP case,  $m$  should not be greater than  $n$ ; if it is, the last  $m - n$  rows of  $H$  are ignored.

If  $i < 0$  or  $i > n$ , the default value is used.

**Infinite Bound Size**  $r$  Default =  $10^{20}$

If  $r > 0$ ,  $r$  defines the ‘infinite’ bound *bigbnd* in the definition of the problem constraints. Any upper bound greater than or equal to *bigbnd* will be regarded as plus infinity (and similarly any lower bound less than or equal to  $-bigbnd$  will be regarded as minus infinity). If  $r \leq 0$ , the default value is used.

**Infinite Step Size**  $r$  Default =  $\max(bigbnd, 10^{20})$

If  $r > 0$ ,  $r$  specifies the magnitude of the change in variables that will be considered a step to an unbounded solution. (Note that an unbounded solution can occur only when the Hessian is not positive-definite.) If the change in  $x$  during an iteration would exceed the value of  $r$ , the objective function is considered to be unbounded below in the feasible region. If  $r \leq 0$ , the default value is used.

**Iteration Limit**  $i$  Default =  $\max(50, 5(n + m_L))$

**Iters**

**Itns**

See **Feasibility Phase Iteration Limit** above.

**List** Default = **List**

**Nolist**

Normally each optional parameter specification is printed as it is supplied. **Nolist** may be used to suppress the printing and **List** may be used to restore printing.

**Maximum Degrees of Freedom**  $i$  Default =  $n$

Note that this option does not apply to problems of type FP or LP.

This places a limit on the storage allocated for the triangular factor  $R$  of the reduced Hessian  $H_R$ . Ideally,  $i$  should be set slightly larger than the value of  $n_R$  expected at the solution. It need not be larger than  $m_N + 1$ , where  $m_N$  is the number of variables that appear nonlinearly in the quadratic objective function. For many problems it can be much smaller than  $m_N$ .

For quadratic problems, a minimizer may lie on any number of constraints, so that  $n_R$  may vary between 1 and  $n$ . The default value of  $i$  is therefore the number of variables  $n$ . If **Hessian Rows**  $m$  is specified, the default value of  $i$  is the same number,  $m$ .

**Minimum Sum of Infeasibilities** No Default = No  
**Minimum Sum of Infeasibilities** Yes

If no feasible point exists for the constraints, this option is used to control whether or not H02CBF will calculate a point that minimizes the constraint violations. If **Minimum Sum of Infeasibilities** = No, H02CBF will terminate as soon as it is evident that no feasible point exists for the constraints. The final point will generally not be the point at which the sum of infeasibilities is minimized. If **Minimum Sum of Infeasibilities** = Yes, H02CBF will continue until the sum of infeasibilities is minimized.

**Monitoring File**  $i$  Default = -1

If  $i \geq 0$  and **Print Level**  $\geq 5$  (see below), monitoring information produced by H02CBF at every iteration is sent to a file with logical unit number  $i$ . If  $i < 0$  and/or **Print Level**  $< 5$ , no monitoring information is produced.

**Optimality Phase Iteration Limit**  $i$  Default =  $\max(50, 5(n + m_L))$

See **Feasibility Phase Iteration Limit** above.

**Optimality Tolerance**  $r$  Default =  $\epsilon^{0.8}$

If  $r \geq \epsilon$ ,  $r$  defines the tolerance used to determine if the bounds and general constraints have the right 'sign' for the solution to be judged to be optimal.

If  $0 \leq r < \epsilon$ , the default value is used.

**Print Level**  $i$  Default = 10

The value of  $i$  controls the amount of printout produced by H02CBF, as indicated below. A detailed description of the printed output is given in Section 8.2 (summary output at each iteration and the final solution) and Section 12 (monitoring information at each iteration). If  $i < 0$ , the default value is used.

The following printout is sent to the current advisory message unit (as defined by X04ABF):

$i$	Output
0	No output.
1	The final solution only.
5	One line of summary output (< 80 characters; see Section 8.2) for each iteration (no printout of the final solution).
$\geq 10$	The final solution and one line of summary output for each iteration.

The following printout is sent to the logical unit number defined by the optional parameter **Monitoring File** (see above):

$i$	Output
$< 5$	No output.
$\geq 5$	One long line of output (> 80 characters; see Section 12) for each iteration (no printout of the final solution).
$\geq 20$	At each iteration, the Lagrange multipliers, the variables $x$ , the constraint values $Ax$ and the constraint status.
$\geq 30$	At each iteration, the diagonal elements of the upper triangular matrix $T$ associated with the $TQ$ factorization (3) (see Section 10.2) of the working set, and the diagonal elements of the upper triangular matrix $R$ .

If **Print Level**  $\geq 5$  and the unit number defined by **Monitoring File** is the same as that defined by X04ABF, then the summary output is suppressed.

**Problem Type**  $a$  Default = QP2

This option specifies the type of objective function to be minimized during the optimality phase. The following are the five optional keywords and the dimensions of the arrays that must be specified in order to define the objective function:

LP	H not referenced, CVEC(N) required;
QP1	H(LDH,*) symmetric, CVEC not referenced;

QP2	H(LDH,*) symmetric, CVEC(N) required;
QP3	H(LDH,*) upper trapezoidal, CVEC not referenced;
QP4	H(LDH,*) upper trapezoidal, CVEC(N) required.

For problems of type FP, the objective function is omitted and neither H nor CVEC are referenced.

The following keywords are also acceptable. The minimum abbreviation of each keyword is underlined.

<i>a</i>	<b>Option</b>
<u>Q</u> uadratic	QP2
<u>L</u> inear	LP
<u>F</u> easible	FP

In addition, the keyword QP is equivalent to the default option QP2.

If  $H = 0$ , i.e., the objective function is purely linear, the efficiency of H02CBF may be increased by specifying *a* as LP.

**Rank Tolerance** *r* Default = 100ε

Note that this option does not apply to problems of type FP or LP.

This parameter enables the user to control the condition number of the triangular factor  $R$  (see Section 10). If  $\rho_i$  denotes the function  $\rho_i = \max\{|R_{11}|, |R_{22}|, \dots, |R_{ii}|\}$ , the dimension of  $R$  is defined to be smallest index  $i$  such that  $|R_{i+1,i+1}| \leq \sqrt{r}|\rho_{i+1}|$ . If  $r \leq 0$ , the default value is used.

### **Warm Start**

See **Cold Start** above.

## 12 Description of Monitoring Information

This section describes the long line of output (> 80 characters) which forms part of the monitoring information produced by H02CBF. (See also the description of the optional parameters **Monitoring File** and **Print Level** in Section 11.2). The level of printed output can be controlled by the user.

To aid interpretation of the printed results, the following convention is used for numbering the constraints: indices 1 through  $n$  refer to the bounds on the variables, and indices  $n + 1$  through  $n + m_L$  refer to the general constraints. When the status of a constraint changes, the index of the constraint is printed, along with the designation L (lower bound), U (upper bound), E (equality), F (temporarily fixed variable) or A (artificial constraint).

When **Print Level**  $\geq 5$  and **Monitoring File**  $\geq 0$ , the following line of output is produced at every iteration on the unit number specified by **Monitoring File**. In all cases, the values of the quantities printed are those in effect *on completion* of the given iteration.

<b>Itn</b>	is the iteration count.
<b>Jdel</b>	is the index of the constraint deleted from the working set. If <b>Jdel</b> is zero, no constraint was deleted.
<b>Jadd</b>	is the index of the constraint added to the working set. If <b>Jadd</b> is zero, no constraint was added.
<b>Step</b>	is the step taken along the computed search direction. If a constraint is added during the current iteration (i.e., <b>Jadd</b> is positive), <b>Step</b> will be the step to the nearest constraint. When the problem is of type LP, the step can be greater than one during the optimality phase.
<b>Ninf</b>	is the number of violated constraints (infeasibilities). This will be zero during the optimality phase.

<b>Sinf/Objective</b>	is the value of the current objective function. If $x$ is not feasible, <b>Sinf</b> gives a weighted sum of the magnitudes of constraint violations. If $x$ is feasible, <b>Objective</b> is the value of the objective function. The output line for the final iteration of the feasibility phase (i.e., the first iteration for which <b>Ninf</b> is zero) will give the value of the true objective at the first feasible point. During the optimality phase, the value of the objective function will be non-increasing. During the feasibility phase, the number of constraint infeasibilities will not increase until either a feasible point is found, or the optimality of the multipliers implies that no feasible point exists. Once optimal multipliers are obtained, the number of infeasibilities can increase, but the sum of infeasibilities will either remain constant or be reduced until the minimum sum of infeasibilities is found.
<b>Bnd</b>	is the number of simple bound constraints in the current working set.
<b>Lin</b>	is the number of general linear constraints in the current working set.
<b>Art</b>	is the number of artificial constraints in the working set, i.e., the number of columns of $Z_A$ (see Section 10.4).
<b>Zr</b>	is the number of columns of $Z_R$ (see Section 10.2). <b>Zr</b> is the dimension of the subspace in which the objective function is currently being minimized. The value of <b>Zr</b> is the number of variables minus the number of constraints in the working set; i.e., $\mathbf{Zr} = n - (\mathbf{Bnd} + \mathbf{Lin} + \mathbf{Art})$ . The value of $n_Z$ , the number of columns of $Z$ (see Section 10.2) can be calculated as $n_Z = n - (\mathbf{Bnd} + \mathbf{Lin})$ . A zero value of $n_Z$ implies that $x$ lies at a vertex of the feasible region.
<b>Norm Gz</b>	is $\ Z_R^T g_{\text{FR}}\ $ , the Euclidean norm of the reduced gradient with respect to $Z_R$ . During the optimality phase, this norm will be approximately zero after a unit step.
<b>NOpt</b>	is the number of non-optimal Lagrange multipliers at the current point. <b>NOpt</b> is not printed if the current $x$ is infeasible or no multipliers have been calculated. At a minimizer, <b>NOpt</b> will be zero.
<b>Min Lm</b>	is the value of the Lagrange multiplier associated with the deleted constraint. If <b>Min Lm</b> is negative, a lower bound constraint has been deleted, if <b>Min Lm</b> is positive, an upper bound constraint has been deleted. If no multipliers are calculated during a given iteration, <b>Min Lm</b> will be zero.
<b>Cond T</b>	is a lower bound on the condition number of the working set.
<b>Cond Rz</b>	is a lower bound on the condition number of the triangular factor $R$ (the Cholesky factor of the current reduced Hessian; see Section 10.2). If the problem is specified to be of type LP, <b>Cond Rz</b> is not printed.
<b>Rzz</b>	is the last diagonal element $\mu$ of the matrix $D$ associated with the $R^T D R$ factorization of the reduced Hessian $H_R$ (see Section 10.2). <b>Rzz</b> is only printed if $H_R$ is not positive-definite (in which case $\mu \neq 1$ ). If the printed value of <b>Rzz</b> is small in absolute value, then $H_R$ is approximately singular. A negative value of <b>Rzz</b> implies that the objective function has negative curvature on the current working set.

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