

# NAG Fortran Library Routine Document

## E04NKF/E04NKA

**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

E04NKF/E04NKA solves sparse linear programming or quadratic programming problems.

E04NKA is a version of E04NKF that has additional parameters in order to make it safe for use in multithreaded applications (see Section 5 below). The initialisation routine E04WBF **must** have been called prior to calling E04NKA.

## 2 Specifications

### 2.1 Specification for E04NKF

```

SUBROUTINE E04NKF(N, M, NNZ, IOBJ, NCOLH, QPHX, A, HA, KA, BL, BU,
1          START, NAMES, NNAME, CRNAME, NS, XS, ISTATE, MINIZ,
2          MINZ, NINF, SINP, OBJ, CLAMDA, IZ, LENIZ, Z, LENZ,
3          IFAIL)
    INTEGER      N, M, NNZ, IOBJ, NCOLH, HA(NNZ), KA(N+1), NNAME, NS,
1          ISTATE(N+M), MINIZ, MINZ, NINF, IZ(LENIZ), LENIZ,
2          LENZ, IFAIL
    real         A(NNZ), BL(N+M), BU(N+M), XS(N+M), SINP, OBJ,
1          CLAMDA(N+M), Z(LENZ)
    CHARACTER*1  START
    CHARACTER*8  NAMES(5), CRNAME(NNAME)
    EXTERNAL     QPHX

```

### 2.2 Specification for E04NKA

```

SUBROUTINE E04NKA(N, M, NNZ, IOBJ, NCOLH, QPHX, A, HA, KA, BL, BU,
1          START, NAMES, NNAME, CRNAME, NS, XS, ISTATE, MINIZ,
2          MINZ, NINF, SINP, OBJ, CLAMDA, IZ, LENIZ, Z, LENZ,
3          IUSER, RUSER, LWSAV, IWSAV, RWSAV, IFAIL)
    INTEGER      N, M, NNZ, IOBJ, NCOLH, HA(NNZ), KA(N+1), NNAME, NS,
1          ISTATE(N+M), MINIZ, MINZ, NINF, IZ(LENIZ), LENIZ,
2          LENZ, IUSER(*), IWSAV(380), IFAIL
    real         A(NNZ), BL(N+M), BU(N+M), XS(N+M), SINP, OBJ,
1          CLAMDA(N+M), Z(LENZ), RUSER(*), RWSAV(285)
    LOGICAL      LWSAV(20)
    CHARACTER*1  START
    CHARACTER*8  NAMES(5), CRNAME(NNAME)
    EXTERNAL     QPHX

```

Before calling E04NKA, or either of the option setting routines E04NLA or E04NMA, routine E04WBF **must** be called. The specification for E04WBF is:

```

SUBROUTINE E04WBF(RNAME, CWSAV, LCWSAV, LWSAV, LLWSAV, IWSAV, LIWSAV,
1          RWSAV, LRWSAV, IFAIL)
    INTEGER      LCWSAV, LLWSAV, IWSAV(LIWSAV), LIWSAV, LRWSAV, IFAIL
    real         RWSAV(LRWSAV)
    LOGICAL      LWSAV(LLWSAV)
    CHARACTER*6  RNAME
    CHARACTER*80  CWSAV(LCWSAV)

```

E04WBF should be called with `RNAME = 'E04NKA'`. `LCWSAV`, `LLWSAV`, `LIWSAV` and `LRWSAV`, the declared lengths of `CWSAV`, `LWSAV`, `IWSAV` and `RWSAV` respectively, must satisfy:

$$\text{LCWSAV} \geq 1$$

$$\text{LLWSAV} \geq 20$$

$$\text{LIWSAV} \geq 380$$

$$\text{LRWSAV} \geq 285$$

The contents of the arrays `CWSAV`, `LWSAV`, `IWSAV` and `RWSAV` **must not** be altered between calling routines `E04WBF` and `E04NKA`, `E04NLA` or `E04NMA`.

### 3 Description

E04NKF/E04NKA is designed to solve a class of quadratic programming problems that are assumed to be stated in the following general form:

$$\underset{x \in R^n}{\text{minimize}} \ f(x) \quad \text{subject to} \quad l \leq \begin{Bmatrix} x \\ Ax \end{Bmatrix} \leq u, \quad (1)$$

where  $x$  is a set of variables,  $A$  is an  $m$  by  $n$  matrix and the objective function  $f(x)$  may be specified in a variety of ways depending upon the particular problem to be solved. The optional parameter **Maximize** (see Section 11.2) may be used to specify an alternative problem in which  $f(x)$  is maximized. The possible forms for  $f(x)$  are listed in Table 1 below, in which the prefixes FP, LP and QP stand for ‘feasible point’, ‘linear programming’ and ‘quadratic programming’ respectively,  $c$  is an  $n$  element vector and  $H$  is the  $n$  by  $n$  second-derivative matrix  $\nabla^2 f(x)$  (the *Hessian matrix*).

Problem type	Objective function $f(x)$	Hessian matrix $H$
FP	Not applicable	Not applicable
LP	$c^T x$	Not applicable
QP	$c^T x + \frac{1}{2} x^T H x$	Symmetric positive semi-definite

**Table 1**

For LP and QP problems, the unique global minimum value of  $f(x)$  is found. For FP problems,  $f(x)$  is omitted and the routine attempts to find a feasible point for the set of constraints. For QP problems, you must also provide a subroutine that computes  $Hx$  for any given vector  $x$ . ( $H$  need not be stored explicitly.) If  $H$  is the zero matrix, the routine will still solve the resulting LP problem; however, this can be accomplished more efficiently by setting `NCOLH = 0` (see Section 5).

The defining feature of a *convex* QP problem is that the matrix  $H$  must be *positive semi-definite*, i.e., it must satisfy  $x^T H x \geq 0$  for all  $x$ . Otherwise,  $f(x)$  is said to be *nonconvex* and it may be more appropriate to call `E04UGF`/`E04UGA` instead.

E04NKF/E04NKA is intended to solve large-scale linear and quadratic programming problems in which the constraint matrix  $A$  is *sparse* (i.e., when the number of zero elements is sufficiently large that it is worthwhile using algorithms which avoid computations and storage involving zero elements). The routine also takes advantage of sparsity in  $c$ . (Sparsity in  $H$  can be exploited in the subroutine that computes  $Hx$ .) For problems in which  $A$  can be treated as a *dense* matrix, it is usually more efficient to use `E04MFF`/`E04MFA`, `E04NCF`/`E04NCA` or `E04NFF`/`E04NFA`.

The upper and lower bounds on the  $m$  elements of  $Ax$  are said to define the *general constraints* of the problem. Internally, E04NKF/E04NKA converts the general constraints to equalities by introducing a set of *slack variables*  $s$ , where  $s = (s_1, s_2, \dots, s_m)^T$ . For example, the linear constraint  $5 \leq 2x_1 + 3x_2 \leq +\infty$  is replaced by  $2x_1 + 3x_2 - s_1 = 0$ , together with the bounded slack  $5 \leq s_1 \leq +\infty$ . The problem defined by (1) can therefore be re-written in the following equivalent form:

$$\underset{x \in R^n, s \in R^m}{\text{minimize}} \ f(x) \quad \text{subject to} \quad Ax - s = 0, \quad l \leq \begin{Bmatrix} x \\ s \end{Bmatrix} \leq u.$$

Since the slack variables  $s$  are subject to the same upper and lower bounds as the elements of  $Ax$ , the bounds on  $Ax$  and  $x$  can simply be thought of as bounds on the combined vector  $(x, s)$ . (In order to

indicate their special role in QP problems, the original variables  $x$  are sometimes known as ‘column variables’, and the slack variables  $s$  are known as ‘row variables’.)

Each LP or QP problem is solved using an *active-set* method. This is an iterative procedure with two phases: a *feasibility phase*, in which the sum of infeasibilities is minimized to find a feasible point; and an *optimality phase*, in which  $f(x)$  is minimized by constructing a sequence of iterations that lies within the feasible region.

A constraint is said to be *active* or *binding* at  $x$  if the associated element of either  $x$  or  $Ax$  is equal to one of its upper or lower bounds. Since an active constraint in  $Ax$  has its associated slack variable at a bound, the status of both simple and general upper and lower bounds can be conveniently described in terms of the status of the variables  $(x, s)$ . A variable is said to be *nonbasic* if it is temporarily fixed at its upper or lower bound. It follows that regarding a general constraint as being *active* is equivalent to thinking of its associated slack as being *nonbasic*.

At each iteration of an active-set method, the constraints  $Ax - s = 0$  are (conceptually) partitioned into the form

$$Bx_B + Sx_S + Nx_N = 0,$$

where  $x_N$  consists of the nonbasic elements of  $(x, s)$  and the *basis matrix*  $B$  is square and non-singular. The elements of  $x_B$  and  $x_S$  are called the *basic* and *superbasic* variables respectively; with  $x_N$  they are a permutation of the elements of  $x$  and  $s$ . At a QP solution, the basic and superbasic variables will lie somewhere between their upper or lower bounds, while the nonbasic variables will be equal to one of their bounds. At each iteration,  $x_S$  is regarded as a set of independent variables that are free to move in any desired direction, namely one that will improve the value of the objective function (or sum of infeasibilities). The basic variables are then adjusted in order to ensure that  $(x, s)$  continues to satisfy  $Ax - s = 0$ . The number of superbasic variables ( $n_S$  say) therefore indicates the number of degrees of freedom remaining after the constraints have been satisfied. In broad terms,  $n_S$  is a measure of *how nonlinear* the problem is. In particular,  $n_S$  will always be zero for FP and LP problems.

If it appears that no improvement can be made with the current definition of  $B$ ,  $S$  and  $N$ , a nonbasic variable is selected to be added to  $S$ , and the process is repeated with the value of  $n_S$  increased by one. At all stages, if a basic or superbasic variable encounters one of its bounds, the variable is made nonbasic and the value of  $n_S$  is decreased by one.

Associated with each of the  $m$  equality constraints  $Ax - s = 0$  is a *dual variable*  $\pi_i$ . Similarly, each variable in  $(x, s)$  has an associated *reduced gradient*  $d_j$  (also known as a *reduced cost*). The reduced gradients for the variables  $x$  are the quantities  $g - A^T \pi$ , where  $g$  is the gradient of the QP objective function; and the reduced gradients for the slack variables  $s$  are the dual variables  $\pi$ . The QP subproblem is optimal if  $d_j \geq 0$  for all nonbasic variables at their lower bounds,  $d_j \leq 0$  for all nonbasic variables at their upper bounds and  $d_j = 0$  for all superbasic variables. In practice, an *approximate* QP solution is found by slightly relaxing these conditions on  $d_j$  (see the description of the optional parameter **Optimality Tolerance** in Section 11.2).

The process of computing and comparing reduced gradients is known as *pricing* (a term first introduced in the context of the simplex method for linear programming). To ‘price’ a nonbasic variable  $x_j$  means that the reduced gradient  $d_j$  associated with the relevant active upper or lower bound on  $x_j$  is computed via the formula  $d_j = g_j - a_j^T \pi$ , where  $a_j$  is the  $j$ th column of  $(A \ -I)$ . (The variable selected by such a process and the corresponding value of  $d_j$  (i.e., its reduced gradient) are the quantities +S and dJ in the monitoring file output; see Section 12.) If  $A$  has significantly more columns than rows (i.e.,  $n \gg m$ ), pricing can be computationally expensive. In this case, a strategy known as *partial pricing* can be used to compute and compare only a subset of the  $d_j$ ’s.

E04NKF/E04NKA is based on SQOPT, which is part of the SNOPT package described in Gill *et al.* (1997), which in turn utilizes routines from the MINOS package (see Murtagh and Saunders (1995)). It uses stable numerical methods throughout and includes a reliable basis package (for maintaining sparse  $LU$  factors of the basis matrix  $B$ ), a practical anti-degeneracy procedure, efficient handling of linear constraints and bounds on the variables (by an active-set strategy), as well as automatic scaling of the constraints. Further details can be found in Section 10.

## 4 References

- Fourer R (1982) Solving staircase linear programs by the simplex method *Math. Programming* **23** 274–313
- Gill P E and Murray W (1978) Numerically stable methods for quadratic programming *Math. Programming* **14** 349–372
- Gill P E, Murray W and Saunders M A (1997) SNOPT: An SQP Algorithm for Large-scale Constrained Optimization *Numerical Analysis Report 97-2* Department of Mathematics, University of California, San Diego
- Gill P E, Murray W, Saunders M A and Wright M H (1987) Maintaining *LU* factors of a general sparse matrix *Linear Algebra and its Applics.* **88/89** 239–270
- 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
- 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
- Hall J A J and McKinnon K I M (1996) The Simplest Examples where the Simplex Method Cycles and Conditions where EXPAND Fails to Prevent Cycling *Report MS 96-100* Department of Mathematics and Statistics, University of Edinburgh
- Murtagh B A and Saunders M A (1995) MINOS 5.4 User's Guide *Report SOL 83-20R* Department of Operations Research, Stanford University

## 5 Parameters

- 1: N – INTEGER *Input*  
*On entry:*  $n$ , the number of variables (excluding slacks). This is the number of columns in the linear constraint matrix  $A$ .  
*Constraint:*  $N \geq 1$ .
- 2: M – INTEGER *Input*  
*On entry:*  $m$ , the number of general linear constraints (or slacks). This is the number of rows in  $A$ , including the free row (if any; see IOBJ below).  
*Constraint:*  $M \geq 1$ .
- 3: NNZ – INTEGER *Input*  
*On entry:* the number of non-zero elements in  $A$ .  
*Constraint:*  $1 \leq \text{NNZ} \leq N \times M$ .
- 4: IOBJ – INTEGER *Input*  
*On entry:* if  $\text{IOBJ} > 0$ , row IOBJ of  $A$  is a free row containing the non-zero elements of the vector  $c$  appearing in the linear objective term  $c^T x$ . If  $\text{IOBJ} = 0$ , there is no free row, i.e., the problem is either an FP problem (in which case IOBJ must be set to zero), or a QP problem with  $c = 0$ .  
*Constraint:*  $0 \leq \text{IOBJ} \leq M$ .
- 5: NCOLH – INTEGER *Input*  
*On entry:*  $n_H$ , the number of leading non-zero columns of the Hessian matrix  $H$ . For FP and LP problems, NCOLH must be set to zero.  
*Constraint:*  $0 \leq \text{NCOLH} \leq N$ .

- 6: QPHX – SUBROUTINE, supplied by the NAG Fortran Library or the user. *External Procedure*

For QP problems, you must supply a version of QPHX to compute the matrix product  $Hx$ . If  $H$  has zero rows and columns, it is most efficient to order the variables  $x = (y \ z)^T$  so that

$$Hx = \begin{pmatrix} H_1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} y \\ z \end{pmatrix} = \begin{pmatrix} H_1 y \\ 0 \end{pmatrix},$$

where the nonlinear variables  $y$  appear first as shown. For FP and LP problems, QPHX will never be called by E04NKF/E04NKA and hence QPHX may be the dummy routine E04NKH/E04NKH (NKUE04/NKUE54 in some implementations: see the Users' Note for your implementation for details).

The specification of QPHX for E04NKF is:

```
SUBROUTINE QPHX(NSTATE, NCOLH, X, HX)
  INTEGER          NSTATE, NCOLH
  real            X(NCOLH), HX(NCOLH)
```

The specification of QPHX for E04NKA is:

```
SUBROUTINE QPHX(NSTATE, NCOLH, X, HX, IUSER, RUSER)
  INTEGER          NSTATE, NCOLH, IUSER(*)
  real            X(NCOLH), HX(NCOLH), RUSER(*)
```

- 1: NSTATE – INTEGER *Input*

*On entry:* if NSTATE = 1, then E04NKF/E04NKA is calling QPHX for the first time. This parameter setting allows you to save computation time if certain data must be read or calculated only once. If NSTATE ≥ 2, then E04NKF/E04NKA is calling QPHX for the last time. This parameter setting allows you to perform some additional computation on the final solution. In general, the last call to QPHX is made with NSTATE = 2 + IFAIL (see Section 6). Otherwise, NSTATE = 0.

- 2: NCOLH – INTEGER *Input*

*On entry:* this is the same parameter NCOLH as supplied to E04NKF/E04NKA (see above).

- 3: X(NCOLH) – **real** array *Input*

*On entry:* the first NCOLH elements of the vector  $x$ .

- 4: HX(NCOLH) – **real** array *Output*

*On exit:* the product  $Hx$ .

**Note:** the following are additional parameters for specific use of QPHX with E04NKA. Users of E04NKF therefore need not read the remainder of this description.

- 5: IUSER(\*) – INTEGER array *User Workspace*

- 6: RUSER(\*) – **real** array *User Workspace*

QPHX is called from E04NKA with the parameters IUSER and RUSER as supplied to E04NKA. You are free to use the arrays IUSER and RUSER to supply information to QPHX.

QPHX must be declared as EXTERNAL in the (sub)program from which E04NKF/E04NKA is called. Parameters denoted as *Input* must **not** be changed by this procedure.

- 7: A(NNZ) – *real* array Input/Output
- On entry:* the non-zero elements of  $A$ , ordered by increasing column index. Note that elements with the same row and column indices are not allowed.
- On exit:* used as internal workspace prior to being restored and hence is unchanged.
- 8: HA(NNZ) – INTEGER array Input
- On entry:* HA( $i$ ) must contain the row index of the non-zero element stored in A( $i$ ), for  $i = 1, 2, \dots, \text{NNZ}$ . Note that the row indices for a column may be supplied in any order.
- Constraint:*  $1 \leq \text{HA}(i) \leq M$ , for  $i = 1, 2, \dots, \text{NNZ}$ .
- 9: KA(N+1) – INTEGER array Input
- On entry:* KA( $j$ ) must contain the index in A of the start of the  $j$ th column, for  $j = 1, 2, \dots, N$ . To specify the  $j$ th column as empty, set KA( $j$ ) = KA( $j + 1$ ). Note that the first and last elements of KA must be such that KA(1) = 1 and KA(N + 1) = NNZ + 1.
- Constraints:*
- $$\begin{aligned} \text{KA}(1) &= 1, \\ \text{KA}(j) &\geq 1, \text{ for } j = 2, 3, \dots, N, \\ \text{KA}(N + 1) &= \text{NNZ} + 1, \\ 0 &\leq \text{KA}(j + 1) - \text{KA}(j) \leq M, \text{ for } j = 1, 2, \dots, N. \end{aligned}$$
- 10: BL(N+M) – *real* array Input/Output
- On entry:*  $l$ , the lower bounds for all the variables and general constraints, in the following order. The first N elements of BL must contain the bounds on the variables  $x$ , and the next M elements the bounds for the general linear constraints  $Ax$  (or slacks  $s$ ) and the free row (if any). To specify a non-existent lower bound (i.e.,  $l_j = -\infty$ ), set BL( $j$ )  $\leq -\text{bigbnd}$ , where  $\text{bigbnd}$  is the value of the optional parameter **Infinite Bound Size** (default value =  $10^{20}$ ; see Section 11.2). To specify the  $j$ th constraint as an *equality*, set BL( $j$ ) = BU( $j$ ) =  $\beta$ , say, where  $|\beta| < \text{bigbnd}$ . Note that the lower bound corresponding to the free row must be set to  $-\infty$  and stored in BL(N + IOBJ).
- On exit:* used as internal workspace prior to being restored and hence is unchanged.
- Constraints:*
- $$\text{BL}(N + \text{IOBJ}) \leq -\text{bigbnd} \text{ when } \text{IOBJ} > 0.$$
- (See also the description for BU below.)
- 11: BU(N+M) – *real* array Input/Output
- On entry:*  $u$ , the upper bounds for all the variables and general constraints, in the following order. The first N elements of BL must contain the bounds on the variables  $x$ , and the next M elements the bounds for the general linear constraints  $Ax$  (or slacks  $s$ ) and the free row (if any). To specify a non-existent upper bound (i.e.,  $u_j = +\infty$ ), set BU( $j$ )  $\geq \text{bigbnd}$ . Note that the upper bound corresponding to the free row must be set to  $+\infty$  and stored in BU(N + IOBJ).
- On exit:* used as internal workspace prior to being restored and hence is unchanged.
- Constraints:*
- $$\begin{aligned} \text{BU}(N + \text{IOBJ}) &\geq \text{bigbnd} \text{ when } \text{IOBJ} > 0, \\ \text{BL}(j) &\leq \text{BU}(j), \text{ for } j = 1, 2, \dots, N + M, \\ |\beta| &< \text{bigbnd} \text{ when } \text{BL}(j) = \text{BU}(j) = \beta. \end{aligned}$$

- 12: START – CHARACTER\*1 *Input*  
*On entry:* indicates how a starting basis is to be obtained.  
 If START = 'C', then an internal Crash procedure will be used to choose an initial basis matrix  $B$ .  
 If START = 'W', then a basis is already defined in ISTATE (probably from a previous call).  
*Constraint:* START = 'C' or 'W'.
- 13: NAMES(5) – CHARACTER\*8 array *Input*  
*On entry:* a set of names associated with the so-called MPSX form of the problem, as follows:  
     NAMES(1) must contain the name for the problem (or be blank);  
     NAMES(2) must contain the name for the free row (or be blank);  
     NAMES(3) must contain the name for the constraint right-hand side (or be blank);  
     NAMES(4) must contain the name for the ranges (or be blank);  
     NAMES(5) must contain the name for the bounds (or be blank).  
 (These names are used in the monitoring file output; see Section 12.)
- 14: NNAME – INTEGER *Input*  
*On entry:* the number of column (i.e., variable) and row names supplied in CRNAME.  
 If NNAME = 1, there are no names. Default names will be used in the printed output.  
 If NNAME = N + M, all names must be supplied.  
*Constraint:* NNAME = 1 or N + M.
- 15: CRNAME(NNAME) – CHARACTER\*8 array *Input*  
*On entry:* the optional column and row names, respectively.  
 If NNAME = 1, CRNAME is not referenced and the printed output will use default names for the columns and rows.  
 If NNAME = N + M, the first N elements must contain the names for the columns and the next M elements must contain the names for the rows. Note that the name for the free row (if any) must be stored in CRNAME(N + IOBJ).
- 16: NS – INTEGER *Input/Output*  
*On entry:*  $n_S$ , the number of superbasics. For QP problems, NS need not be specified if START = 'C', but must retain its value from a previous call when START = 'W'. For FP and LP problems, NS need not be initialised.  
*On exit:* the final number of superbasics. This will be zero for FP and LP problems.
- 17: XS(N+M) – *real* array *Input/Output*  
*On entry:* the initial values of the variables and slacks  $(x, s)$ . (See the description for ISTATE below.)  
*On exit:* the final values of the variables and slacks  $(x, s)$ .
- 18: ISTATE(N+M) – INTEGER array *Input/Output*  
*On entry:* if START = 'C', the first N elements of ISTATE and XS must specify the initial states and values, respectively, of the variables  $x$ . (The slacks  $s$  need not be initialised.) An internal Crash procedure is then used to select an initial basis matrix  $B$ . The initial basis matrix will be triangular (neglecting certain small elements in each column). It is chosen from various rows and columns of  $(A \ -I)$ . Possible values for ISTATE( $j$ ) are as follows:

**ISTATE(*j*) State of XS(*j*) during Crash procedure**

0 or 1	Eligible for the basis
2	Ignored
3	Eligible for the basis (given preference over 0 or 1)
4 or 5	Ignored

If nothing special is known about the problem, or there is no wish to provide special information, you may set  $ISTATE(j) = 0$  and  $XS(j) = 0.0$ , for  $j = 1, 2, \dots, N$ . All variables will then be eligible for the initial basis. Less trivially, to say that the  $j$ th variable will probably be equal to one of its bounds, set  $ISTATE(j) = 4$  and  $XS(j) = BL(j)$  or  $ISTATE(j) = 5$  and  $XS(j) = BU(j)$  as appropriate.

Following the Crash procedure, variables for which  $ISTATE(j) = 2$  are made superbasic. Other variables not selected for the basis are then made nonbasic at the value  $XS(j)$  if  $BL(j) \leq XS(j) \leq BU(j)$ , or at the value  $BL(j)$  or  $BU(j)$  closest to  $XS(j)$ .

If  $START = 'W'$ ,  $ISTATE$  and  $XS$  must specify the initial states and values, respectively, of the variables and slacks ( $x, s$ ). If E04NKF/E04NKA has been called previously with the same values of  $N$  and  $M$ ,  $ISTATE$  already contains satisfactory information.

*Constraints:*

if  $START = 'C'$ ,  $0 \leq ISTATE(j) \leq 5$ , for  $j = 1, 2, \dots, N$ ;  
 if  $START = 'W'$ ,  $0 \leq ISTATE(j) \leq 3$ , for  $j = 1, 2, \dots, N + M$ .

*On exit:* the final states of the variables and slacks ( $x, s$ ). The significance of each possible value of  $ISTATE(j)$  is as follows:

**ISTATE(*j*) State of variable *j* Normal value of XS(*j*)**

0	Nonbasic	$BL(j)$
1	Nonbasic	$BU(j)$
2	Superbasic	Between $BL(j)$ and $BU(j)$
3	Basic	Between $BL(j)$ and $BU(j)$

If  $NINF = 0$ , basic and superbasic variables may be outside their bounds by as much as the value of the optional parameter **Feasibility Tolerance** (default value =  $\max(10^{-6}, \sqrt{\epsilon})$ , where  $\epsilon$  is the *machine precision*; see Section 11.2). Note that unless the optional parameter **Scale Option** = 0 (default value = 2; see Section 11.2) is specified, the **Feasibility Tolerance** applies to the variables of the scaled problem. In this case, the variables of the original problem may be as much as 0.1 outside their bounds, but this is unlikely unless the problem is very badly scaled.

Very occasionally some nonbasic variables may be outside their bounds by as much as the **Feasibility Tolerance**, and there may be some nonbasic variables for which  $XS(j)$  lies strictly between its bounds.

If  $NINF > 0$ , some basic and superbasic variables may be outside their bounds by an arbitrary amount (bounded by  $SINF$  if **Scale Option** = 0).

19: MINIZ – INTEGER

*Output*

*On exit:* the minimum value of LENIZ required to start solving the problem. If  $IFAIL = 12$ , E04NKF/E04NKA may be called again with LENIZ suitably larger than MINIZ. (The bigger the better, since it is not certain how much workspace the basis factors need.)

20: MINZ – INTEGER

*Output*

*On exit:* the minimum value of LENZ required to start solving the problem. If  $IFAIL = 13$ , E04NKF/E04NKA may be called again with LENZ suitably larger than MINZ. (The bigger the better, since it is not certain how much workspace the basis factors need.)

21: NINF – INTEGER

*Output*

*On exit:* the number of infeasibilities. This will be zero if  $IFAIL = 0$  or 1.



- 22: SINF – *real* Output  
*On exit:* the sum of infeasibilities. This will be zero if NINF = 0. (Note that E04NKF/E04NKA does *not* attempt to compute the minimum value of SINF if IFAIL = 3.)
- 23: OBJ – *real* Output  
*On exit:* the value of the objective function. If NINF = 0, OBJ includes the quadratic objective term  $\frac{1}{2}x^T Hx$  (if any). If NINF > 0, OBJ is just the linear objective term  $c^T x$  (if any). For FP problems, OBJ is set to zero.
- 24: CLAMDA(N+M) – *real* array Output  
*On exit:* a set of Lagrange multipliers for the bounds on the variables and the general constraints. More precisely, the first N elements contain the multipliers (*reduced costs*) for the bounds on the variables, and the next M elements contain the multipliers (*shadow prices*) for the general linear constraints.
- 25: IZ(LENIZ) – INTEGER array Workspace  
26: LENIZ – INTEGER Input  
*On entry:* the dimension of the array IZ as declared in the (sub)program from which E04NKF/E04NKA is called.  
*Constraint:* LENIZ ≥ 1.
- 27: Z(LENZ) – *real* array Workspace  
28: LENZ – INTEGER Input  
*On entry:* the dimension of the array Z as declared in the (sub)program from which E04NKF/E04NKA is called.  
*Constraint:* LENZ ≥ 1.  
The amounts of workspace provided (i.e., LENIZ and LENZ) and required (i.e., MINIZ and MINZ) are (by default for E04NKF) output on the current advisory message unit NADV (as defined by X04ABF). Since the minimum values of LENIZ and LENZ required to start solving the problem are returned in MINIZ and MINZ, respectively, you may prefer to obtain appropriate values from the output of a preliminary run with LENIZ and LENZ set to 1. (E04NKF/E04NKA will then terminate with IFAIL = 12.)
- 29: IFAIL – INTEGER Input/Output  
**Note:** for E04NKA, IFAIL does not occur in this position in the parameter list. See the additional parameters described below.  
*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 (see Section 6).  
For environments where it might be inappropriate to halt program execution when an error is detected, the value -1 or 1 is recommended. If the output of error messages is undesirable, then the value 1 is recommended. Otherwise, because for this routine the values of the output parameters may be useful even if IFAIL ≠ 0 on exit, the recommended value is -1. **When the value -1 or 1 is used it is essential to test the value of IFAIL on exit.**  
E04NKF/E04NKA returns with IFAIL = 0 if the reduced gradient (Norm rg; see Section 8.1) is negligible, the Lagrange multipliers (Lagr Mult; see Section 8.1) are optimal and  $x$  satisfies the constraints to the accuracy requested by the value of the optional parameter **Feasibility Tolerance** (default value =  $\max(10^{-6}, \sqrt{\epsilon})$ , where  $\epsilon$  is the *machine precision*; see Section 11.2).

**Note:** the following are additional parameters for specific use with E04NKA. Users of E04NKF therefore need not read the remainder of this section.

29: IUSER(\*) – INTEGER array User Workspace

**Note:** the dimension of the array IUSER must be at least 1.

IUSER is not used by E04NKA, but is passed directly to the external procedure QPHX and may be used to pass information to that routine.

30: RUSER(\*) – *real* array User Workspace

**Note:** the dimension of the array RUSER must be at least 1.

RUSER is not used by E04NKA, but is passed directly to the external procedure QPHX and may be used to pass information to that routine.

31: LWSAV(20) – LOGICAL array Workspace

32: IWSAV(380) – INTEGER array Workspace

33: RWSAV(285) – *real* array Workspace

The arrays LWSAV, IWSAV and RWSAV **must not** be altered between calls to any of the routines E04WBF, E04NKA, E04NLA or E04NMA.

34: IFAIL – INTEGER Input/Output

**Note:** see the parameter description for IFAIL above.

## 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 detected by the routine:

IFAIL = 1

Weak solution found. The final  $x$  is not unique, although  $x$  gives the global minimum value of the objective function.

IFAIL = 2

The problem is unbounded (or badly scaled). The objective function is not bounded below in the feasible region.

IFAIL = 3

The problem is infeasible. The general constraints cannot all be satisfied simultaneously to within the value of the optional parameter **Feasibility Tolerance** (default value =  $\max(10^{-6}, \sqrt{\epsilon})$ , where  $\epsilon$  is the *machine precision*; see Section 11.2).

IFAIL = 4

Too many iterations. The value of the optional parameter **Iteration Limit** (default value =  $\max(50, 5(n + m))$ ; see Section 11.2) is too small.

IFAIL = 5

The reduced Hessian matrix  $Z^T H Z$  (see Section 10.2) exceeds its assigned dimension. The value of the optional parameter **Superbasics Limit** (default value =  $\min(n_H + 1, n)$ ; see Section 11.2) is too small.

IFAIL = 6

The Hessian matrix  $H$  appears to be indefinite. This sometimes occurs because the values of the optional parameters **LU Factor Tolerance** (default value = 100.0) and **LU Update Tolerance** (default value = 10.0; see Section 11.2) are too large. Check also that subroutine QPHX has been coded correctly and that all relevant elements of  $Hx$  have been assigned their correct values.

IFAIL = 7

An input parameter is invalid.

IFAIL = 8

Numerical error in trying to satisfy the general constraints. The basis is very ill-conditioned.

IFAIL = 9

Not enough integer workspace for the basis factors. Increase LENIZ and rerun E04NKF/E04NKA.

IFAIL = 10

Not enough real workspace for the basis factors. Increase LENZ and rerun E04NKF/E04NKA.

IFAIL = 11

The basis is singular after 15 attempts to factorize it (adding slacks where necessary). Either the problem is badly scaled or the value of the optional parameter **LU Factor Tolerance** (default value = 100.0; see Section 11.2) is too large.

IFAIL = 12

Not enough integer workspace to start solving the problem. Increase LENIZ to at least MINIZ and rerun E04NKF/E04NKA.

IFAIL = 13

Not enough real workspace to start solving the problem. Increase LENZ to at least MINZ and rerun E04NKF/E04NKA.

## 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 a description of the printed output.

### 8.1 Description of the Printed Output

This section describes the intermediate printout and final printout produced by E04NKF/E04NKA. 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 for E04NKF, by default no output is produced by E04NKA).

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 *on completion* of the given iteration.

Itn                      is the iteration count.

Step                    is the step taken along the computed search direction.

Ninf	is the number of violated constraints (infeasibilities). This will be zero during the optimality phase.
Sinf/Objective	<p>is the value of the current objective function. If <math>x</math> is not feasible, Sinf gives the sum of the magnitudes of constraint violations. If <math>x</math> is feasible, Objective 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 Ninf is zero) will give the value of the true objective at the first feasible point.</p> <p>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.</p>
Norm rg	is $\ d_S\ $ , the Euclidean norm of the reduced gradient (see Section 10.3). During the optimality phase, this norm will be approximately zero after a unit step. For FP and LP problems, Norm rg is not printed.
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.	
Variable	gives the name of the variable. If NNAME = 1, a default name is assigned to the $j$ th variable for $j = 1, 2, \dots, n$ . If NNAME = N + M, the name supplied in CRNAME( $j$ ) is assigned to the $j$ th variable.
State	<p>gives the state of the variable (LL if nonbasic on its lower bound, UL if nonbasic on its upper bound, EQ if nonbasic and fixed, FR if nonbasic and strictly between its bounds, BS if basic and SBS if superbasic).</p> <p>A key is sometimes printed before State to give some additional information about the state of a variable. Note that unless the optional parameter <b>Scale Option</b> = 0 (default value = 2; see Section 11.2) is specified, the tests for assigning a key are applied to the variables of the scaled problem.</p> <p>A <i>Alternative optimum possible</i>. The variable is nonbasic, but its reduced gradient is essentially zero. This means that if the variable were allowed to start moving away from its bound, there would be no change in the value of the objective function. The values of the other free variables <i>might</i> 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 <i>might</i> also change.</p> <p>D <i>Degenerate</i>. The variable is basic or superbasic, but it is equal to (or very close to) one of its bounds.</p> <p>I <i>Infeasible</i>. The variable is basic or superbasic and is currently violating one of its bounds by more than the value of the optional parameter <b>Feasibility Tolerance</b> (default value = <math>\max(10^{-6}, \sqrt{\epsilon})</math>, where <math>\epsilon</math> is the <i>machine precision</i>; see Section 11.2).</p> <p>N <i>Not precisely optimal</i>. The variable is nonbasic or superbasic. If the value of the reduced gradient for the variable exceeds the value of the optional parameter <b>Optimality Tolerance</b> (default value = <math>\max(10^{-6}, \sqrt{\epsilon})</math>; see Section 11.2), the solution would not be declared optimal because the reduced gradient for the variable would not be considered negligible.</p>
Value	is the value of the variable at the final iteration.
Lower Bound	is the lower bound specified for the variable. None indicates that $BL(j) \leq -bigbnd$ .
Upper Bound	is the upper bound specified for the variable. None indicates that $BU(j) \geq bigbnd$ .

Lagr Mult	is the Lagrange multiplier for the associated bound. This will be zero if State is FR. If $x$ is optimal, the multiplier should be non-negative if State is LL, non-positive if State is UL and zero if State is BS or SBS.
Residual	is the difference between the variable Value 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 linear constraints is the same as that given above for variables, with ‘variable’ replaced by ‘constraint’,  $n$  replaced by  $m$ , CRNAME( $j$ ) replaced by CRNAME( $n + j$ ), BL( $j$ ) and BU( $j$ ) are replaced by BL( $n + j$ ) and BU( $n + j$ ) respectively, and with the following change in the heading:

Constrnt                      gives the name 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 Residual 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  $f(x) = c^T x + \frac{1}{2}x^T Hx$ , where

$$c = (-200.0, -2000.0, -2000.0, -2000.0, -2000.0, 400.0, 400.0)^T$$

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

subject to the bounds

$$\begin{aligned} 0 &\leq x_1 \leq 200 \\ 0 &\leq x_2 \leq 2500 \\ 400 &\leq x_3 \leq 800 \\ 100 &\leq x_4 \leq 700 \\ 0 &\leq x_5 \leq 1500 \\ 0 &\leq x_6 \\ 0 &\leq x_7 \end{aligned}$$

and to the linear constraints

$$\begin{aligned} x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 &= 2000 \\ 0.15x_1 + 0.04x_2 + 0.02x_3 + 0.04x_4 + 0.02x_5 + 0.01x_6 + 0.03x_7 &\leq 60 \\ 0.03x_1 + 0.05x_2 + 0.08x_3 + 0.02x_4 + 0.06x_5 + 0.01x_6 &\leq 100 \\ 0.02x_1 + 0.04x_2 + 0.01x_3 + 0.02x_4 + 0.02x_5 &\leq 40 \\ 0.02x_1 + 0.03x_2 + 0.01x_5 &\leq 30 \\ 1500 &\leq 0.70x_1 + 0.75x_2 + 0.80x_3 + 0.75x_4 + 0.80x_5 + 0.97x_6 \\ 250 &\leq 0.02x_1 + 0.06x_2 + 0.08x_3 + 0.12x_4 + 0.02x_5 + 0.01x_6 + 0.97x_7 \leq 300 \end{aligned}$$

The initial point, which is infeasible, is

$$x_0 = (0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0)^T.$$

The optimal solution (to five figures) is

$$x^* = (0.0, 349.40, 648.85, 172.85, 407.52, 271.36, 150.02)^T.$$

One bound constraint and four linear constraints are active at the solution. Note that the Hessian matrix  $H$  is positive semi-definite.

## 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.

**Note:** the following program illustrates the use of E04NKF. An equivalent program illustrating the use of E04NKA is available with the supplied Library and is also available from the NAG web site.

```
*      E04NKF Example Program Text.
*      Mark 19 Revised. NAG Copyright 1999.
*      .. Parameters ..
      INTEGER          NIN, NOUT
      PARAMETER        (NIN=5,NOUT=6)
      INTEGER          IDUMMY
      PARAMETER        (IDUMMY=-11111)
      INTEGER          NMAX, MMAX, NNZMAX, LENIZ, LENZ
      PARAMETER        (NMAX=100,MMAX=100,NNZMAX=100,LENIZ=10000,
+                      LENZ=10000)
*      .. Local Scalars ..
      real             OBJ, SINF
      INTEGER          I, ICOL, IFAIL, IOBJ, J, JCOL, M, MINIZ, MINZ, N,
+                      NCOLH, NINF, NNAME, NNZ, NS
      CHARACTER        START
*      .. Local Arrays ..
      real             A(NNZMAX), BL(NMAX+MMAX), BU(NMAX+MMAX),
+                      CLAMDA(NMAX+MMAX), XS(NMAX+MMAX), Z(LENZ)
      INTEGER          HA(NNZMAX), ISTATE(NMAX+MMAX), IZ(LENIZ),
+                      KA(NMAX+1)
      CHARACTER*8      CRNAME(NMAX+MMAX), NAMES(5)
*      .. External Subroutines ..
      EXTERNAL          E04NKF, QPHX
*      .. Executable Statements ..
      WRITE (NOUT,*) 'E04NKF Example Program Results'
      Skip heading in data file.
      READ (NIN,*)
      READ (NIN,*) N, M
      IF (N.LE.NMAX .AND. M.LE.MMAX) THEN
*
*          Read NNZ, IOBJ, NCOLH, START and NNAME from data file.
*
      READ (NIN,*) NNZ, IOBJ, NCOLH, START, NNAME
*
*          Read NAMES and CRNAME from data file.
*
      READ (NIN,*) (NAMES(I),I=1,5)
      READ (NIN,*) (CRNAME(I),I=1,NNAME)
*
*          Read the matrix A from data file. Set up KA.
*
      JCOL = 1
      KA(JCOL) = 1
      DO 40 I = 1, NNZ
*
*          Element ( HA( I ), ICOL ) is stored in A( I ).
*
      READ (NIN,*) A(I), HA(I), ICOL
*
      IF (ICOL.LT.JCOL) THEN
*
*          Elements not ordered by increasing column index.
*
      WRITE (NOUT,99999) 'Element in column', ICOL,
+          ' found after element in column', JCOL, '. Problem',
+          ' abandoned.'
      STOP
      ELSE IF (ICOL.EQ.JCOL+1) THEN
```

```

*
*           Index in A of the start of the ICOL-th column equals I.
*
*           KA(ICOL) = I
*           JCOL = ICOL
*           ELSE IF (ICOL.GT.JCOL+1) THEN
*
*           Index in A of the start of the ICOL-th column equals I,
*           but columns JCOL+1,JCOL+2,...,ICOL-1 are empty. Set the
*           corresponding elements of KA to I.
*
*           DO 20 J = JCOL + 1, ICOL - 1
*               KA(J) = I
20          CONTINUE
*           KA(ICOL) = I
*           JCOL = ICOL
40          END IF
*           CONTINUE
*
*           KA(N+1) = NNZ + 1
*
*           IF (N.GT.ICOL) THEN
*
*           Columns N,N-1,...,ICOL+1 are empty. Set the corresponding
*           elements of KA accordingly.
*
*           DO 60 I = N, ICOL + 1, -1
*               IF (KA(I).EQ.IDUMMY) KA(I) = KA(I+1)
60          CONTINUE
*           END IF
*
*           Read BL, BU, ISTATE and XS from data file.
*
*           READ (NIN,*) (BL(I),I=1,N+M)
*           READ (NIN,*) (BU(I),I=1,N+M)
*           IF (START.EQ.'C') THEN
*               READ (NIN,*) (ISTATE(I),I=1,N)
*           ELSE IF (START.EQ.'W') THEN
*               READ (NIN,*) (ISTATE(I),I=1,N+M)
*           END IF
*           READ (NIN,*) (XS(I),I=1,N)
*
*           Solve the QP problem.
*
*           IFAIL = -1
*
*           CALL E04NKF(N,M,NNZ,IOBJ,NCOLH,QPHX,A,HA,KA,BL,BU,START,NAMES,
+              NNAME,CRNAME,NS,XS,ISTATE,MINIZ,MINZ,NINF,SINF,OBJ,
+              CLAMDA,IZ,LENIZ,Z,LENZ,IFAIL)
*
*           END IF
*           STOP
*
99999 FORMAT (/1X,A,I5,A,I5,A,A)
END
*
SUBROUTINE QPHX(NSTATE,NCOLH,X,HX)
*
* Routine to compute H*x. (In this version of QPHX, the Hessian
* matrix H is not referenced explicitly.)
*
* .. Parameters ..
real TWO
PARAMETER (TWO=2.0e+0)
* .. Scalar Arguments ..
INTEGER NCOLH, NSTATE
* .. Array Arguments ..
real HX(NCOLH), X(NCOLH)
* .. Executable Statements ..
*
HX(1) = TWO*X(1)

```

```

      HX(2) = TWO*X(2)
      HX(3) = TWO*(X(3)+X(4))
      HX(4) = HX(3)
      HX(5) = TWO*X(5)
      HX(6) = TWO*(X(6)+X(7))
      HX(7) = HX(6)
*
      RETURN
      END

```

## 9.2 Program Data

E04NKF Example Program Data

```

7  8      :Values of N and M
48 8 7 'C' 15 :Values of NNZ, IOBJ, NCOLH, START and NNAME
'...' '...' '...' '...' '...' :End of NAMES
'...' '...' '...' '...' '...'
'...' '...' '...' '...' '...'
'...' '...' '...' '...' '...' :End of CRNAME
0.02 7 1
0.02 5 1
0.03 3 1
1.00 1 1
0.70 6 1
0.02 4 1
0.15 2 1
-200.00 8 1
0.06 7 2
0.75 6 2
0.03 5 2
0.04 4 2
0.05 3 2
0.04 2 2
1.00 1 2
-2000.00 8 2
0.02 2 3
1.00 1 3
0.01 4 3
0.08 3 3
0.08 7 3
0.80 6 3
-2000.00 8 3
1.00 1 4
0.12 7 4
0.02 3 4
0.02 4 4
0.75 6 4
0.04 2 4
-2000.00 8 4
0.01 5 5
0.80 6 5
0.02 7 5
1.00 1 5
0.02 2 5
0.06 3 5
0.02 4 5
-2000.00 8 5
1.00 1 6
0.01 2 6
0.01 3 6
0.97 6 6
0.01 7 6
400.00 8 6
0.97 7 7
0.03 2 7
1.00 1 7
400.00 8 7
0.0 0.0 4.0E+02 1.0E+02 0.0 0.0 0.0 2.0E+03
-1.0E+25 -1.0E+25 -1.0E+25 -1.0E+25 1.5E+03 2.5E+02 -1.0E+25 :End of BL
2.0E+02 2.5E+03 8.0E+02 7.0E+02 1.5E+03 1.0E+25 1.0E+25 2.0E+03

```



```

6.0E+01    1.0E+02    4.0E+01    3.0E+01    1.0E+25    3.0E+02    1.0E+25 :End of BU
0      0      0      0      0      0      0      0 :End of ISTATE
0.0    0.0    0.0    0.0    0.0    0.0    0.0    0.0 :End of XS

```

### 9.3 Program Results

E04NKF Example Program Results

\*\*\* E04NKF

\*\*\* Start of NAG Library implementation details \*\*\*

Implementation title: Generalised Base Version

Precision: FORTRAN double precision

Product Code: FLBAS20D

Mark: 20A

\*\*\* End of NAG Library implementation details \*\*\*

Parameters

-----

Frequencies.

Check frequency.....	60	Expand frequency.....	10000
Factorization frequency.	100		

LP Parameters.

Scale tolerance.....	9.00E-01	Feasibility tolerance...	1.00E-06
Iteration limit.....	75	Scale option.....	2
Optimality tolerance....	1.00E-06	Partial price.....	10
Crash tolerance.....	1.00E-01	Pivot tolerance.....	2.05E-11
Crash option.....	2		

QP objective.

Objective variables.....	7	Hessian columns.....	7
Superbasics limit.....	7		

Miscellaneous.

Variables.....	7	Linear constraints.....	8
LU factor tolerance.....	1.00E+02	LU update tolerance.....	1.00E+01
LU singularity tolerance	2.05E-11	Monitoring file.....	-1
EPS (machine precision).	1.11E-16	Print level.....	10
Infinite bound size.....	1.00E+20	Infinite step size.....	1.00E+20
COLD start.....		MINIMIZE.....	

Workspace provided is IZ( 10000), Z( 10000).  
 To start solving the problem we need IZ( 428), Z( 358).

Itn	Step	Ninf	Sinf/Objective	Norm rg
Itn	0	-- Infeasible.		
0	0.0E+00	1	1.152891E+03	0.0E+00
1	4.3E+02	0	0.000000E+00	0.0E+00
Itn	1	-- Feasible point found (for 1 equality constraints).		
1	0.0E+00	0	0.000000E+00	0.0E+00
1	0.0E+00	0	1.460000E+06	0.0E+00
Itn	1	-- Feasible QP solution.		
2	8.7E-02	0	9.409959E+05	0.0E+00
3	5.3E-01	0	-1.056552E+06	0.0E+00
4	1.0E+00	0	-1.462190E+06	4.1E-12
5	1.0E+00	0	-1.698092E+06	1.8E-12
6	4.6E-02	0	-1.764906E+06	7.0E+02
7	1.0E+00	0	-1.811946E+06	9.1E-13
8	1.7E-02	0	-1.847325E+06	1.7E+02
9	1.0E+00	0	-1.847785E+06	5.2E-12

Variable	State	Value	Lower Bound	Upper Bound	Lagr Mult	Residual
...X1...	LL	0.00000	.	200.00	2361.	.
...X2...	BS	349.399	.	2500.0	-1.0616E-12	349.4

...X3...	SBS	648.853	400.00	800.00	-4.3952E-12	151.1
...X4...	SBS	172.847	100.00	700.00	-2.2739E-12	72.85
...X5...	BS	407.521	.	1500.0	-2.0673E-12	407.5
...X6...	BS	271.356	.	None	7.4549E-13	271.4
...X7...	BS	150.023	.	None	4.7101E-13	150.0

Constrnt	State	Value	Lower Bound	Upper Bound	Lagr Mult	Residual
..ROW1..	EQ	2000.00	2000.0	2000.0	-1.2901E+04	.
..ROW2..	BS	49.2316	None	60.000	-1.3490E-11	-10.77
..ROW3..	UL	100.000	None	100.00	-2325.	.
..ROW4..	BS	32.0719	None	40.000	.	-7.928
..ROW5..	BS	14.5572	None	30.000	.	-15.44
..ROW6..	LL	1500.00	1500.0	None	1.4455E+04	.
..ROW7..	LL	250.000	250.00	300.00	1.4581E+04	.
..COST..	BS	-2.988690E+06	None	None	-1.000	-2.9887E+06

Exit E04NKF - Optimal QP solution found.

Final QP objective value = -1847785.

Exit from QP problem after 9 iterations.

**Note:** the remainder of this document is intended for more advanced users. Section 10 contains a detailed algorithm description that may be needed in order to understand Section 11 and Section 12. Section 11 describes the optional parameters that may be set by calls to E04NLF/E04NLA and/or E04NMF/E04NMA. Section 12 describes the quantities that can be requested to monitor the course of the computation.

## 10 Algorithmic Details

This section contains a description of the method used by E04NKF/E04NKA.

### 10.1 Overview

E04NKF/E04NKA is based on an inertia-controlling method that maintains a Cholesky factorization of the reduced Hessian (see below). The method is similar to that of Gill and Murray (1978), and is described in detail by Gill *et al.* (1991). 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 the routine or appear in the printed output.

The method used has two distinct 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 (the printed quantity *Sinf*; see Section 12) to the quadratic objective function (the printed quantity *Objective*; see Section 12).

In general, an iterative process is required to solve a quadratic program. Given an iterate  $(x, s)$  in both the original variables  $x$  and the slack variables  $s$ , a new iterate  $(\bar{x}, \bar{s})$  is defined by

$$\begin{pmatrix} \bar{x} \\ \bar{s} \end{pmatrix} = \begin{pmatrix} x \\ s \end{pmatrix} + \alpha p, \quad (2)$$

where the *step length*  $\alpha$  is a non-negative scalar (the printed quantity *Step*; see Section 12), and  $p$  is called the *search direction*. (For simplicity, we shall consider a typical iteration and avoid reference to the index of the iteration.) Once an iterate is feasible (i.e., satisfies the constraints), all subsequent iterates remain feasible.

## 10.2 Definition of the Working Set and Search Direction

At each iterate  $(x, s)$ , a *working set* of constraints is defined to be a linearly independent subset of the constraints that are satisfied ‘exactly’ (to within the value of the optional parameter **Feasibility Tolerance**; see Section 11.2). The working set is the current prediction of the constraints that hold with equality at a solution of the LP or QP problem. Let  $m_W$  denote the number of constraints in the working set (including bounds), and let  $W$  denote the associated  $m_W$  by  $(n + m)$  *working set matrix* consisting of the  $m_W$  gradients of the working set constraints.

The search direction is defined so that constraints in the working set remain *unaltered* for any value of the step length. It follows that  $p$  must satisfy the identity

$$Wp = 0. \quad (3)$$

This characterization allows  $p$  to be computed using any  $n$  by  $n_Z$  full-rank matrix  $Z$  that spans the null space of  $W$ . (Thus,  $n_Z = n - m_W$  and  $WZ = 0$ .) The null space matrix  $Z$  is defined from a sparse *LU* factorization of part of  $W$  (see (6) and (7) below). The direction  $p$  will satisfy (3) if

$$p = Zp_Z, \quad (4)$$

where  $p_Z$  is any  $n_Z$ -vector.

The working set contains the constraints  $Ax - s = 0$  and a subset of the upper and lower bounds on the variables  $(x, s)$ . Since the gradient of a bound constraint  $x_j \geq l_j$  or  $x_j \leq u_j$  is a vector of all zeros except for  $\pm 1$  in position  $j$ , it follows that the working set matrix contains the rows of  $(A \ -I)$  and the unit rows associated with the upper and lower bounds in the working set.

The working set matrix  $W$  can be represented in terms of a certain column partition of the matrix  $(A \ -I)$  by (conceptually) partitioning the constraints  $Ax - s = 0$  so that

$$Bx_B + Sx_S + Nx_N = 0, \quad (5)$$

where  $B$  is a square non-singular basis and  $x_B$ ,  $x_S$  and  $x_N$  are the basic, superbasic and nonbasic variables respectively. The nonbasic variables are equal to their upper or lower bounds at  $(x, s)$ , and the superbasic variables are independent variables that are chosen to improve the value of the current objective function. The number of superbasic variables is  $n_S$  (the printed quantity *Ns*; see Section 12). Given values of  $x_N$  and  $x_S$ , the basic variables  $x_B$  are adjusted so that  $(x, s)$  satisfies (5).

If  $P$  is a permutation matrix such that  $(A \ -I)P = (B \ S \ N)$ , then  $W$  satisfies

$$WP = \begin{pmatrix} B & S & N \\ 0 & 0 & I_N \end{pmatrix}, \quad (6)$$

where  $I_N$  is the identity matrix with the same number of columns as  $N$ .

The null space matrix  $Z$  is defined from a sparse *LU* factorization of part of  $W$ . In particular,  $Z$  is maintained in ‘reduced gradient’ form, using the LUSOL package (see Gill *et al.* (1991)) to maintain sparse *LU* factors of the basis matrix  $B$  that alters as the working set  $W$  changes. Given the permutation  $P$ , the null space basis is given by

$$Z = P \begin{pmatrix} -B^{-1}S \\ I \\ 0 \end{pmatrix}. \quad (7)$$

This matrix is used only as an operator, i.e., it is never computed explicitly. Products of the form  $Zv$  and  $Z^Tg$  are obtained by solving with  $B$  or  $B^T$ . This choice of  $Z$  implies that  $n_Z$ , the number of ‘degrees of freedom’ at  $(x, s)$ , is the same as  $n_S$ , the number of superbasic variables.

Let  $g_Z$  and  $H_Z$  denote the *reduced gradient* and *reduced Hessian* of the objective function:

$$g_Z = Z^Tg \quad \text{and} \quad H_Z = Z^THZ, \quad (8)$$

where  $g$  is the objective gradient at  $(x, s)$ . Roughly speaking,  $g_Z$  and  $H_Z$  describe the first and second derivatives of an  $n_S$ -dimensional *unconstrained* problem for the calculation of  $p_Z$ . (The condition estimator of  $H_Z$  is the quantity *Cond Hz* in the monitoring file output; see Section 12.)

At each iteration, an upper triangular factor  $R$  is available such that  $H_Z = R^T R$ . Normally,  $R$  is computed from  $R^T R = Z^T H Z$  at the start of the optimality phase and then updated as the QP working set changes. For efficiency, the dimension of  $R$  should not be excessive (say,  $n_S \leq 1000$ ). This is guaranteed if the number of nonlinear variables is ‘moderate’.

If the QP problem contains linear variables,  $H$  is positive semi-definite and  $R$  may be singular with at least one zero diagonal element. However, an inertia-controlling strategy is used to ensure that only the last diagonal element of  $R$  can be zero. (See Gill *et al.* (1991) for a discussion of a similar strategy for indefinite quadratic programming.)

If the initial  $R$  is singular, enough variables are fixed at their current value to give a non-singular  $R$ . This is equivalent to including temporary bound constraints in the working set. Thereafter,  $R$  can become singular only when a constraint is deleted from the working set (in which case no further constraints are deleted until  $R$  becomes non-singular).

### 10.3 Main Iteration

If the reduced gradient is zero,  $(x, s)$  is a constrained stationary point on the working set. During the feasibility phase, the reduced gradient will usually be zero only at a vertex (although it may be zero elsewhere in the presence of constraint dependencies). During the optimality phase, a zero reduced gradient implies that  $x$  minimizes the quadratic objective function when the constraints in the working set are treated as equalities. At a constrained stationary point, Lagrange multipliers  $\lambda$  are defined from the equations

$$W^T \lambda = g(x). \quad (9)$$

A Lagrange multiplier  $\lambda_j$  corresponding to an inequality constraint in the working set is said to be *optimal* if  $\lambda_j \leq \sigma$  when the associated constraint is at its *upper bound*, or if  $\lambda_j \geq -\sigma$  when the associated constraint is at its *lower bound*, where  $\sigma$  depends on the value of the optional parameter **Optimality Tolerance** (see Section 11.2). If a multiplier is non-optimal, the objective function (either the true objective or the sum of infeasibilities) can be reduced by continuing the minimization with the corresponding constraint excluded from the working set. (This step is sometimes referred to as ‘deleting’ a constraint from the working set.) If optimal multipliers occur during the feasibility phase but the sum of infeasibilities is non-zero, there is no feasible point and the routine terminates immediately with IFAIL = 3 (see Section 6).

The special form (6) of the working set allows the multiplier vector  $\lambda$ , the solution of (9), to be written in terms of the vector

$$d = \begin{pmatrix} g \\ 0 \end{pmatrix} - (A \quad -I)^T \pi = \begin{pmatrix} g - A^T \pi \\ \pi \end{pmatrix}, \quad (10)$$

where  $\pi$  satisfies the equations  $B^T \pi = g_B$ , and  $g_B$  denotes the basic elements of  $g$ . The elements of  $\pi$  are the Lagrange multipliers  $\lambda_j$  associated with the equality constraints  $Ax - s = 0$ . The vector  $d_N$  of nonbasic elements of  $d$  consists of the Lagrange multipliers  $\lambda_j$  associated with the upper and lower bound constraints in the working set. The vector  $d_S$  of superbasic elements of  $d$  is the reduced gradient  $g_Z$  in (8). The vector  $d_B$  of basic elements of  $d$  is zero, by construction. (The Euclidean norm of  $d_S$  and the final values of  $d_S$ ,  $g$  and  $\pi$  are the quantities Norm rg, Reduced Gradnt, Obj Gradient and Dual Activity in the monitoring file output; see Section 12.)

If the reduced gradient is not zero, Lagrange multipliers need not be computed and the search direction is given by  $p = Zp_Z$  (see (7) and (11)). The step length is chosen to maintain feasibility with respect to the satisfied constraints.

There are two possible choices for  $p_Z$ , depending on whether or not  $H_Z$  is singular. If  $H_Z$  is non-singular,  $R$  is non-singular and  $p_Z$  in (4) is computed from the equations

$$R^T R p_Z = -g_Z, \quad (11)$$

where  $g_Z$  is the reduced gradient at  $x$ . In this case,  $(x, s) + p$  is the minimizer of the objective function subject to the working set constraints being treated as equalities. If  $(x, s) + p$  is feasible,  $\alpha$  is defined to be unity. In this case, the reduced gradient at  $(\bar{x}, \bar{s})$  will be zero, and Lagrange multipliers are computed at

the next iteration. Otherwise,  $\alpha$  is set to  $\alpha_M$ , the step to the ‘nearest’ constraint along  $p$ . This constraint is then added to the working set at the next iteration.

If  $H_Z$  is singular, then  $R$  must also be singular, and an inertia-controlling strategy is used to ensure that only the last diagonal element of  $R$  is zero. (See Gill *et al.* (1991) for a discussion of a similar strategy for indefinite quadratic programming.) In this case,  $p_Z$  satisfies

$$p_Z^T H_Z p_Z = 0 \quad \text{and} \quad g_Z^T p_Z \leq 0, \quad (12)$$

which allows the objective function to be reduced by any step of the form  $(x, s) + \alpha p$ , where  $\alpha > 0$ . The vector  $p = Z p_Z$  is a direction of unbounded descent for the QP problem in the sense that the QP objective is linear and decreases without bound along  $p$ . If no finite step of the form  $(x, s) + \alpha p$  (where  $\alpha > 0$ ) reaches a constraint not in the working set, the QP problem is unbounded and the routine terminates immediately with IFAIL = 2 (see Section 6). Otherwise,  $\alpha$  is defined as the maximum feasible step along  $p$  and a constraint active at  $(x, s) + \alpha p$  is added to the working set for the next iteration.

## 10.4 Miscellaneous

If the basis matrix is not chosen carefully, the condition of the null space matrix  $Z$  in (7) could be arbitrarily high. To guard against this, the routine implements a ‘basis repair’ feature in which the LUSOL package (see Gill *et al.* (1991)) is used to compute the rectangular factorization

$$\begin{pmatrix} B & S \end{pmatrix}^T = LU, \quad (13)$$

returning just the permutation  $P$  that makes  $PLP^T$  unit lower triangular. The pivot tolerance is set to require  $|PLP^T|_{ij} \leq 2$ , and the permutation is used to define  $P$  in (6). It can be shown that  $\|Z\|$  is likely to be little more than unity. Hence,  $Z$  should be well-conditioned *regardless of the condition of  $W$* . This feature is applied at the beginning of the optimality phase if a potential  $B - S$  ordering is known.

The EXPAND procedure (see Gill *et al.* (1989)) is used to reduce the possibility of cycling at a point where the active constraints are nearly linearly dependent. Although there is no absolute guarantee that cycling will not occur, the probability of cycling is extremely small (see Hall and McKinnon (1996)). The main feature of EXPAND is that the feasibility tolerance is increased at the start of every iteration. This allows a positive step to be taken at every iteration, perhaps at the expense of violating the bounds on  $(x, s)$  by a small amount.

Suppose that the value of the optional parameter **Feasibility Tolerance** (see Section 11.2) is  $\delta$ . Over a period of  $K$  iterations (where  $K$  is the value of the optional parameter **Expand Frequency**; see Section 11.2), the feasibility tolerance actually used by the routine (i.e., the *working* feasibility tolerance) increases from  $0.5\delta$  to  $\delta$  (in steps of  $0.5\delta/K$ ).

At certain stages the following ‘resetting procedure’ is used to remove small constraint infeasibilities. First, all nonbasic variables are moved exactly onto their bounds. A count is kept of the number of non-trivial adjustments made. If the count is non-zero, the basic variables are recomputed. Finally, the working feasibility tolerance is reinitialised to  $0.5\delta$ .

If a problem requires more than  $K$  iterations, the resetting procedure is invoked and a new cycle of iterations is started. (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 the routine 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.

The EXPAND procedure not only allows a positive step to be taken at every iteration, but also provides a potential *choice* of constraints to be added to the working set. All constraints at a distance  $\alpha$  (where  $\alpha \leq \alpha_M$ ) 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. This strategy helps keep the basis matrix  $B$  well-conditioned.

## 11 Optional Parameters

Several optional parameters in E04NKF/E04NKA define choices in the problem specification or the algorithm logic. In order to reduce the number of formal parameters of E04NKF/E04NKA 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 E04NLF/E04NLA and E04NMF/E04NMA prior to a call to E04NKF/E04NKA.

E04NLF/E04NLA 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 E04NLF (IOPTNS, INFORM)
```

can then be used to read the file on unit IOPTNS. INFORM will be zero on successful exit. E04NLF/E04NLA should be consulted for a full description of this method of supplying optional parameters.

E04NMF/E04NMA can be called to supply options directly, one call being necessary for each optional parameter. For example,

```
CALL E04NMF ('Print Level = 5')
```

E04NMF/E04NMA 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 E04NKF/E04NKA (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	60
Crash Option	2
Crash Tolerance	0.1
Defaults	
Expand Frequency	10000
Factorization Frequency	100
Feasibility Tolerance	$\max(10^{-6}, \sqrt{\epsilon})$
Infinite Bound Size	$10^{20}$
Infinite Step Size	$\max(\text{bigbnd}, 10^{20})$
Iteration Limit	$\max(50, 5(n + m))$
List/Nolist	<b>List</b> ( <b>Nolist</b> for E04NKA)
LU Factor Tolerance	100.0
LU Singularity Tolerance	$\epsilon^{0.67}$
LU Update Tolerance	10.0
Maximize/Minimize	<b>Minimize</b>
Monitoring File	-1
Optimality Tolerance	$\max(10^{-6}, \sqrt{\epsilon})$
Partial Price	10

<b>Pivot Tolerance</b>	$\epsilon^{0.67}$
<b>Print Level</b>	10 (0 for E04NKA)
<b>Scale Option</b>	2
<b>Scale Tolerance</b>	0.9
<b>Superbasics Limit</b>	$\min(n_H + 1, n)$
<b>Rank Tolerance</b>	$100\epsilon$

## 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. The qualifier may be omitted. The letters  $i$  and  $r$  denote INTEGER and *real* values required with certain options. The default value of an option is used whenever the condition  $|i| \geq 100000000$  is satisfied. The number  $\epsilon$  is a generic notation for *machine precision* (see X02AJF).

**Check Frequency**  $i$  Default = 60

Every  $i$ th iteration after the most recent basis factorization, a numerical test is made to see if the current solution  $(x, s)$  satisfies the linear constraints  $Ax - s = 0$ . If the largest element of the residual vector  $r = Ax - s$  is judged to be too large, the current basis is refactorized and the basic variables recomputed to satisfy the constraints more accurately. If  $i < 0$ , the default value is used. If  $i = 0$ , the value  $i = 99999999$  is used and effectively no checks are made.

**Crash Option**  $i$  Default = 2

Note that this option does not apply when START = 'W' (see Section 5).

If START = 'C', an internal Crash procedure is used to select an initial basis from various rows and columns of the constraint matrix  $(A \ -I)$ . The value of  $i$  determines which rows and columns are initially eligible for the basis, and how many times the Crash procedure is called. If  $i = 0$ , the all-slack basis  $B = -I$  is chosen. If  $i = 1$ , the Crash procedure is called once (looking for a triangular basis in all rows and columns of the linear constraint matrix  $A$ ). If  $i = 2$ , the Crash procedure is called twice (looking at any *equality* constraints first followed by any *inequality* constraints). If  $i < 0$  or  $i > 2$ , the default value is used.

If  $i = 1$  or 2, certain slacks on inequality rows are selected for the basis first. (If  $i = 2$ , numerical values are used to exclude slacks that are close to a bound.) The Crash procedure then makes several passes through the columns of  $A$ , searching for a basis matrix that is essentially triangular. A column is assigned to 'pivot' on a particular row if the column contains a suitably large element in a row that has not yet been assigned. (The pivot elements ultimately form the diagonals of the triangular basis.) For remaining unassigned rows, slack variables are inserted to complete the basis.

**Crash Tolerance**  $r$  Default = 0.1

This value allows the Crash procedure to ignore certain 'small' non-zero elements in the constraint matrix  $A$  while searching for a triangular basis. For each column of  $A$ , if  $a_{\max}$  is the largest element in the column, other non-zeros in that column are ignored if they are less than (or equal to)  $a_{\max} \times r$ .

When  $r > 0$ , the basis obtained by the Crash procedure may not be strictly triangular, but it is likely to be non-singular and almost triangular. The intention is to obtain a starting basis with more column variables and fewer (arbitrary) slacks. A feasible solution may be reached earlier for some problems. If  $r < 0$  or  $r \geq 1$ , 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 = 10000

This option is part of an anti-cycling procedure (see Section 10.4) designed to allow progress even on highly degenerate problems.

For LP 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 E04NKF/E04NKA (i.e., the *working* feasibility tolerance) increases from  $0.5\delta$  to  $\delta$  (in steps of  $0.5\delta/i$ ).

For QP problems, the same procedure is used for iterations in which there is only one superbasic variable. (Cycling can only occur when the current solution is at a vertex of the feasible region.) Thus, zero steps are allowed if there is more than one superbasic variable, but otherwise positive steps are enforced.

Increasing the value of  $i$  helps reduce the number of slightly infeasible nonbasic basic variables (most of which are eliminated during the resetting procedure). However, it also diminishes the freedom to choose a large pivot element (see **Pivot Tolerance** below).

If  $i < 0$ , the default value is used. If  $i = 0$ , the value  $i = 99999999$  is used and effectively no anti-cycling procedure is invoked.

**Factorization Frequency** $i$ 

Default = 100

If  $i > 0$ , at most  $i$  basis changes will occur between factorizations of the basis matrix. For LP problems, the basis factors are usually updated at every iteration. For QP problems, fewer basis updates will occur as the solution is approached. The number of iterations between basis factorizations will therefore increase. During these iterations a test is made regularly according to the value of **Check Frequency** (see above) to ensure that the linear constraints  $Ax - s = 0$  are satisfied. If necessary, the basis will be refactored before the limit of  $i$  updates is reached. If  $i \leq 0$ , the default value is used.

**Feasibility Tolerance** $r$ Default =  $\max(10^{-6}, \sqrt{\epsilon})$ 

If  $r \geq \epsilon$ ,  $r$  defines the maximum acceptable *absolute* violation in each constraint at a ‘feasible’ point (including slack variables). For example, if the variables and the coefficients in the linear constraints are of order unity, and the latter are correct to about 5 decimal digits, it would be appropriate to specify  $r$  as  $10^{-5}$ . If  $r < \epsilon$ , the default value is used.

E04NKF/E04NKA attempts to find a feasible solution before optimizing the objective function. If the sum of infeasibilities cannot be reduced to zero, the problem is assumed to be *infeasible*. Let  $S_{\text{inf}}$  be the corresponding sum of infeasibilities. If  $S_{\text{inf}}$  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 the routine does *not* attempt to find the minimum value of  $S_{\text{inf}}$ .

If the constraints and variables have been scaled (see **Scale Option** below), then feasibility is defined in terms of the scaled problem (since it is more likely to be meaningful).

**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.



<b><u>Iteration Limit</u></b>	$i$	Default = $\max(50, 5(n + m))$
<b><u>Iters</u></b>		
<b><u>Itns</u></b>		

The value of  $i$  specifies the maximum number of iterations allowed before termination. Setting  $i = 0$  and **Print Level**  $> 0$  means that the workspace needed to start solving the problem will be computed and printed, but no iterations will be performed. If  $i < 0$ , the default value is used.

<b><u>List</u></b>	Default for E04NKF = <b>List</b>
<b><u>Nolist</u></b>	Default for E04NKA = <b>Nolist</b>

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.

<b><u>LU Factor Tolerance</u></b>	$r_1$	Default = 100.0
<b><u>LU Update Tolerance</u></b>	$r_2$	Default = 10.0

The values of  $r_1$  and  $r_2$  affect the stability and sparsity of the basis factorization  $B = LU$ , during refactorization and updates respectively. The lower triangular matrix  $L$  is a product of matrices of the form

$$\begin{pmatrix} 1 & \\ \mu & 1 \end{pmatrix}$$

where the multipliers  $\mu$  will satisfy  $|\mu| \leq r_i$ . The default values of  $r_1$  and  $r_2$  usually strike a good compromise between stability and sparsity. For large and relatively dense problems, setting  $r_1$  and  $r_2$  to 25 (say) may give a marked improvement in sparsity without impairing stability to a serious degree. Note that for band matrices it may be necessary to set  $r_1$  in the range  $1 \leq r_1 < 2$  in order to achieve stability. If  $r_1 < 1$  or  $r_2 < 1$ , the default value is used.

<b><u>LU Singularity Tolerance</u></b>	$r$	Default = $\epsilon^{0.67}$
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If  $r > 0$ ,  $r$  defines the singularity tolerance used to guard against ill-conditioned basis matrices. Whenever the basis is refactorized, the diagonal elements of  $U$  are tested as follows. If  $|u_{jj}| \leq r$  or  $|u_{jj}| < r \times \max_i |u_{ij}|$ , the  $j$ th column of the basis is replaced by the corresponding slack variable. If  $r \leq 0$ , the default value is used.

<b><u>Maximize</u></b>	Default = <b>Minimize</b>
<b><u>Minimize</u></b>	

This option specifies the required direction of the optimization. It applies to both linear and nonlinear terms (if any) in the objective function. Note that if two problems are the same except that one minimizes  $f(x)$  and the other maximizes  $-f(x)$ , their solutions will be the same but the signs of the dual variables  $\pi_i$  and the reduced gradients  $d_j$  (see Section 10.3) will be reversed.

<b><u>Monitoring File</u></b>	$i$	Default = $-1$
-------------------------------	-----	----------------

If  $i \geq 0$  and **Print Level**  $> 0$  (see below), monitoring information produced by E04NKF/E04NKA is sent to a file with logical unit number  $i$ . If  $i < 0$  and/or **Print Level**  $= 0$ , the default value is used and hence no monitoring information is produced.

<b><u>Optimality Tolerance</u></b>	$r$	Default = $\max(10^{-6}, \sqrt{\epsilon})$
------------------------------------	-----	--

If  $r \geq \epsilon$ ,  $r$  is used to judge the size of the reduced gradients  $d_j = g_j - \pi^T a_j$ . By definition, the reduced gradients for basic variables are always zero. Optimality is declared if the reduced gradients for any nonbasic variables at their lower or upper bounds satisfy  $-r \times \max(1, \|\pi\|) \leq d_j \leq r \times \max(1, \|\pi\|)$ , and if  $|d_j| \leq r \times \max(1, \|\pi\|)$  for any superbasic variables. If  $r < \epsilon$ , the default value is used.

**Partial Price** $i$ 

Default = 10

Note that this option does not apply to QP problems.

This option is recommended for large FP or LP problems that have significantly more variables than constraints (i.e.,  $n \gg m$ ). It reduces the work required for each pricing operation (i.e., when a nonbasic variable is selected to enter the basis). If  $i = 1$ , all columns of the constraint matrix  $(A \ -I)$  are searched. If  $i > 1$ ,  $A$  and  $-I$  are partitioned to give  $i$  roughly equal segments  $A_j, K_j$ , for  $j = 1, 2, \dots, p$  (modulo  $p$ ). If the previous pricing search was successful on  $A_{j-1}, K_{j-1}$ , the next search begins on the segments  $A_j, K_j$ . If a reduced gradient is found that is larger than some dynamic tolerance, the variable with the largest such reduced gradient (of appropriate sign) is selected to enter the basis. If nothing is found, the search continues on the next segments  $A_{j+1}, K_{j+1}$ , and so on. If  $i \leq 0$ , the default value is used.

**Pivot Tolerance** $r$ Default =  $\epsilon^{0.67}$ 

If  $r > 0$ ,  $r$  is used to prevent columns entering the basis if they would cause the basis to become almost singular. If  $r \leq 0$ , the default value is used.

**Print Level** $i$ 

Default for E04NKF = 10

Default for E04NKA = 0

The value of  $i$  controls the amount of printout produced by E04NKF/E04NKA, as indicated below. A detailed description of the printed output is given in Section 8.1 (summary output at each iteration and the final solution) and Section 12 (monitoring information at each iteration). Note that the summary output will not exceed 80 characters per line and that the monitoring information will not exceed 120 characters per line. 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 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**

- 0 No output.
- 1 The final solution only.
- 5 One long line of output for each iteration (no printout of the final solution).
- $\geq 10$  The final solution and one long line of output for each iteration.
- $\geq 20$  The final solution, one long line of output for each iteration, matrix statistics (initial status of rows and columns, number of elements, density, biggest and smallest elements, etc.), details of the scale factors resulting from the scaling procedure (if **Scale Option** = 1 or 2; see below), basis factorization statistics and details of the initial basis resulting from the Crash procedure (if **START** = 'C'; see Section 5).

If **Print Level**  $> 0$  and the unit number defined by **Monitoring File** is the same as that defined by X04ABF, then the summary output is suppressed.

**Scale Option** $i$ 

Default = 2

This option enables you to scale the variables and constraints using an iterative procedure due to Fourer (1982), which attempts to compute row scales  $r_i$  and column scales  $c_j$  such that the scaled matrix coefficients  $\bar{a}_{ij} = a_{ij} \times (c_j/r_i)$  are as close as possible to unity. This may improve the overall efficiency on some problems. (The lower and upper bounds on the variables and slacks for the scaled problem are redefined as  $\bar{l}_j = l_j/c_j$  and  $\bar{u}_j = u_j/c_j$  respectively, where  $c_j \equiv r_{j-n}$  if  $j > n$ .)

If  $i = 0$ , no scaling is performed. If  $i = 1$ , all rows and columns of the constraint matrix  $A$  are scaled. If  $i = 2$ , an additional scaling is performed that may be helpful when the solution  $x$  is large; it takes into account columns of  $(A \ -I)$  that are fixed or have positive lower bounds or negative upper bounds. If  $i < 0$  or  $i > 2$ , the default value is used.

**Scale Tolerance** $r$ 

Default = 0.9

Note that this option does not apply when **Scale Option** = 0 (see above).

If  $0 < r < 1$ ,  $r$  is used to control the number of scaling passes to be made through the constraint matrix  $A$ . At least 3 (and at most 10) passes will be made. More precisely, let  $a_p$  denote the largest column ratio (i.e.,  $\frac{\text{'biggest' element}}{\text{'smallest' element}}$  in some sense) after the  $p$ th scaling pass through  $A$ . The scaling procedure is terminated if  $a_p \geq a_{p-1} \times r$  for some  $p \geq 3$ . Thus, increasing the value of  $r$  from 0.9 to 0.99 (say) will probably increase the number of passes through  $A$ . If  $r \leq 0$  or  $r \geq 1$ , the default value is used.

**Superbasics Limit** $i$ Default =  $\min(n_H + 1, n)$ 

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

The value of  $i$  specifies 'how nonlinear' you expect the QP problem to be. If  $i \leq 0$ , the default value is used.

## 12 Description of Monitoring Information

This section describes the intermediate printout and final printout which constitutes the monitoring information produced by E04NKF/E04NKA. (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.

When **Print Level** = 5 or  $\geq 10$  and **Monitoring File**  $\geq 0$ , the following line of intermediate printout ( $< 120$  characters) is produced at every iteration on the unit number specified by **Monitoring File**. Unless stated otherwise, the values of the quantities printed are those in effect *on completion* of the given iteration.

Itn	is the iteration count.
pp	is the partial price indicator. The variable selected by the last pricing operation came from the pp-th partition of $A$ and $-I$ . Note that pp is reset to zero whenever the basis is refactorized.
dj	is the value of the reduced gradient (or reduced cost) for the variable selected by the pricing operation at the start of the current iteration.
+S	is the variable selected by the pricing operation to be added to the superbasic set.
-S	is the variable chosen to leave the superbasic set.
-B	is the variable removed from the basis (if any) to become nonbasic.
-B	is the variable chosen to leave the set of basics (if any) in a special basic $\leftrightarrow$ superbasic swap. The entry under -S has become basic if this entry is non-zero, and nonbasic otherwise. The swap is done to ensure that there are no superbasic slacks.
Step	is the value of the step length $\alpha$ taken along the current search direction $p$ . The variables $x$ have just been changed to $x + \alpha p$ . If a variable is made superbasic during the current iteration (i.e., +S is positive), Step will be the step to the nearest bound. During the optimality phase, the step can be greater than unity only if the reduced Hessian is not positive-definite.
Pivot	is the $r$ th element of a vector $y$ satisfying $By = a_q$ whenever $a_q$ (the $q$ th column of the constraint matrix $(A \ -I)$ ) replaces the $r$ th column of the basis matrix $B$ . Wherever possible, Step is chosen so as to avoid extremely small values of Pivot (since they may cause the basis to be nearly singular). In extreme cases, it may be necessary to increase the value of the optional parameter <b>Pivot Tolerance</b>

	(default value = $\epsilon^{0.67}$ ) to exclude very small elements of $y$ from consideration during the computation of Step.
Ninf	is the number of violated constraints (infeasibilities). This will be zero during the optimality phase.
Sinf/Objective	is the value of the current objective function. If $x$ is not feasible, Sinf gives the sum of the magnitudes of constraint violations. If $x$ is feasible, Objective 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 Ninf 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.
L	is the number of non-zeros in the basis factor $L$ . Immediately after a basis factorization $B = LU$ , this entry contains lenL. Further non-zeros are added to L when various columns of $B$ are later replaced. (Thus, L increases monotonically.)
U	is the number of non-zeros in the basis factor $U$ . Immediately after a basis factorization $B = LU$ , this entry contains lenU. As columns of $B$ are replaced, the matrix $U$ is maintained explicitly (in sparse form). The value of U may fluctuate up or down; in general, it will tend to increase.
Ncp	is the number of compressions required to recover workspace in the data structure for $U$ . This includes the number of compressions needed during the previous basis factorization. Normally, Ncp should increase very slowly. If it does not, increase LENIZ and LENZ by at least $L + U$ and rerun E04NKF/E04NKA (possibly using START = 'W'; see Section 5).
Norm rg	is $\ d_S\ $ , the Euclidean norm of the reduced gradient (see Section 10.3). During the optimality phase, this norm will be approximately zero after a unit step. For FP and LP problems, Norm rg is not printed.
Ns	is the current number of superbasic variables. For FP and LP problems, Ns is not printed.
Cond Hz	is a lower bound on the condition number of the reduced Hessian (see Section 10.2). The larger this number, the more difficult the problem. For FP and LP problems, Cond Hz is not printed.

When **Print Level**  $\geq 20$  and **Monitoring File**  $\geq 0$ , the following lines of intermediate printout (< 120 characters) are produced on the unit number specified by **Monitoring File** whenever the matrix  $B$  or  $B_S = (B \ S)^T$  is factorized. Gaussian elimination is used to compute an  $LU$  factorization of  $B$  or  $B_S$ , where  $PLP^T$  is a lower triangular matrix and  $PUQ$  is an upper triangular matrix for some permutation matrices  $P$  and  $Q$ . The factorization is stabilized in the manner described under the optional parameter **LU Factor Tolerance** (default value = 100.0; see Section 11.2).

Factorize	is the factorization count.
Demand	is a code giving the reason for the present factorization as follows:
<b>Code</b>	<b>Meaning</b>
0	First $LU$ factorization.
1	The number of updates reached the value of the optional parameter <b>Factorization Frequency</b> (default value = 50 or 100).
2	The number of non-zeros in the updated factors has increased significantly.
7	Not enough storage to update factors.
10	Row residuals too large (see the description for the optional parameter <b>Check Frequency</b> ).
11	Ill-conditioning has caused inconsistent results.
Iteration	is the iteration count.

Nonlinear	is the number of nonlinear variables in the current basis $B$ (not printed if $B_S$ is factorized).
Linear	is the number of linear variables in $B$ (not printed if $B_S$ is factorized).
Slacks	is the number of slack variables in $B$ (not printed if $B_S$ is factorized).
Elms	is the number of non-zeros in $B$ (not printed if $B_S$ is factorized)
Density	is the percentage non-zero density of $B$ (not printed if $B_S$ is factorized). More precisely, $\text{Density} = 100 \times \text{Elms} / (\text{Nonlinear} + \text{Linear} + \text{Slacks})^2$ .
Compressns	is the number of times the data structure holding the partially factorized matrix needed to be compressed, in order to recover unused workspace. Ideally, it should be zero. If it is more than 3 or 4, increase LENIZ and LENZ and rerun E04NKF/E04NKA (possibly using <code>START = 'W'</code> ; see Section 5).
Merit	is the average Markowitz merit count for the elements chosen to be the diagonals of $PUQ$ . Each merit count is defined to be $(c - 1)(r - 1)$ , where $c$ and $r$ are the number of non-zeros in the column and row containing the element at the time it is selected to be the next diagonal. Merit is the average of $m$ such quantities. It gives an indication of how much work was required to preserve sparsity during the factorization.
lenL	is the number of non-zeros in $L$ .
lenU	is the number of non-zeros in $U$ .
Increase	is the percentage increase in the number of non-zeros in $L$ and $U$ relative to the number of non-zeros in $B$ . More precisely, $\text{Increase} = 100 \times (\text{lenL} + \text{lenU} - \text{Elms}) / \text{Elms}$ .
m	is the number of rows in the problem. Note that $m = U_t + L_t + \text{bp}$ .
Ut	is the number of triangular rows of $B$ at the top of $U$ .
d1	is the number of columns remaining when the density of the basis matrix being factorized reached 0.3.
Lmax	is the maximum subdiagonal element in the columns of $L$ . This will not exceed the value of the optional parameter <b>LU Factor Tolerance</b> (default value = 5.0 or 100.0).
Bmax	is the maximum non-zero element in $B$ (not printed if $B_S$ is factorized).
BSmax	is the maximum non-zero element in $B_S$ (not printed if $B$ is factorized).
Umax	is the maximum non-zero element in $U$ , excluding elements of $B$ that remain in $U$ unchanged. (For example, if a slack variable is in the basis, the corresponding row of $B$ will become a row of $U$ without modification. Elements in such rows will not contribute to $U_{\max}$ . If the basis is strictly triangular then <i>none</i> of the elements of $B$ will contribute and $U_{\max}$ will be zero.)  Ideally, $U_{\max}$ should not be significantly larger than $B_{\max}$ . If it is several orders of magnitude larger, it may be advisable to reset the <b>LU Factor Tolerance</b> to some value nearer unity.  $U_{\max}$ is not printed if $B_S$ is factorized.
Umin	is the magnitude of the smallest diagonal element of $PUQ$ (not printed if $B_S$ is factorized).

Growth	is the value of the ratio $U_{\max}/B_{\max}$ , which should not be too large.  Providing $L_{\max}$ is not large (say $< 10.0$ ), the ratio $\max(B_{\max}, U_{\max})/U_{\min}$ is an estimate of the condition number of $B$ . If this number is extremely large, the basis is nearly singular and some numerical difficulties might occur. (However, an effort is made to avoid near-singularity by using slacks to replace columns of $B$ that would have made $U_{\min}$ extremely small and the modified basis is refactorized.)  Growth is not printed if $B_S$ is factorized.
Lt	is the number of triangular columns of $B$ at the left of $L$ .
bp	is the size of the ‘bump’ or block to be factorized nontrivially after the triangular rows and columns of $B$ have been removed.
d2	is the number of columns remaining when the density of the basis matrix being factorized has reached 0.6.

When **Print Level**  $\geq 20$  and **Monitoring File**  $\geq 0$ , the following lines of intermediate printout ( $< 80$  characters) are produced on the unit number specified by **Monitoring File** whenever **START** = ‘C’ (see Section 5). They refer to the number of columns selected by the Crash procedure during each of several passes through  $A$ , whilst searching for a triangular basis matrix.

Slacks	is the number of slacks selected initially.
Free cols	is the number of free columns in the basis, including those whose bounds are rather far apart.
Preferred	is the number of ‘preferred’ columns in the basis (i.e., $ISTATE(j) = 3$ for some $j \leq n$ ). It will be a subset of the columns for which $ISTATE(j) = 3$ was specified.
Unit	is the number of unit columns in the basis.
Double	is the number of double columns in the basis.
Triangle	is the number of triangular columns in the basis.
Pad	is the number of slacks used to pad the basis (to make it a non-singular triangle).

When **Print Level**  $\geq 20$  and **Monitoring File**  $\geq 0$ , the following lines of intermediate printout ( $< 80$  characters) are produced on the unit number specified by **Monitoring File**. They refer to the elements of the NAMES array (see Section 5).

Name	gives the name for the problem (blank if none).
Status	gives the exit status for the problem (i.e., Optimal soln, Weak soln, Unbounded, Infeasible, Excess itns, Error condn or Feasble soln) followed by details of the direction of the optimization (i.e., (Min) or (Max)).
Objective	gives the name of the free row for the problem (blank if none).
RHS	gives the name of the constraint right-hand side for the problem (blank if none).
Ranges	gives the name of the ranges for the problem (blank if none).
Bounds	gives the name of the bounds for the problem (blank if none).

When **Print Level** = 1 or  $\geq 10$  and **Monitoring File**  $\geq 0$ , the following lines of final printout ( $< 120$  characters) are produced on the unit number specified by **Monitoring File**.

Let  $a_j$  denote the  $j$ th column of  $A$ , for  $j = 1, 2, \dots, n$ . The following describes the printout for each column (or variable). A full stop (.) is printed for any numerical value that is zero.

Number	is the column number $j$ . (This is used internally to refer to $x_j$ in the intermediate output.)
Column	gives the name of $x_j$ .
State	gives the state of the variable (LL if nonbasic on its lower bound, UL if nonbasic on its upper bound, EQ if nonbasic and fixed, FR if nonbasic and strictly between its bounds, BS if basic and SBS if superbasic).

A key is sometimes printed before `State` to give some additional information about the state of a variable. Note that unless the optional parameter **Scale Option** = 0 (default value = 1 or 2) is specified, the tests for assigning a key are applied to the variables of the scaled problem.

- A *Alternative optimum possible.* The variable is nonbasic, but its reduced gradient is essentially zero. This means that if the variable were allowed to start moving away from its bound, there would be no change in the value of 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 basic or superbasic, but it is equal to (or very close to) one of its bounds.
- I *Infeasible.* The variable is basic or superbasic and is currently violating one of its bounds by more than the value of the optional parameter **Feasibility Tolerance** (default value =  $\max(10^{-6}, \sqrt{\epsilon})$ , where  $\epsilon$  is the *machine precision*; see Section 11.2).
- N *Not precisely optimal.* The variable is nonbasic or superbasic. If the value of the reduced gradient for the variable exceeds the value of the optional parameter **Optimality Tolerance** (default value =  $\max(10^{-6}, \sqrt{\epsilon})$ ; see Section 11.2), the solution would not be declared optimal because the reduced gradient for the variable would not be considered negligible.

Activity	is the value of $x_j$ at the final iterate.
Obj Gradient	is the value of $g_j$ at the final iterate. For FP problems, $g_j$ is set to zero.
Lower Bound	is the lower bound specified for the variable. None indicates that $BL(j) \leq -bigbnd$ .
Upper Bound	is the upper bound specified for the variable. None indicates that $BU(j) \geq bigbnd$ .
Reduced Gradnt	is the value of $d_j$ at the final iterate (see Section 10.3). For FP problems, $d_j$ is set to zero.
$m + j$	is the value of $m + j$ .

Let  $v_i$  denote the  $i$ th row of  $A$ , for  $i = 1, 2, \dots, m$ . The following describes the printout for each row (or constraint). A full stop (.) is printed for any numerical value that is zero.

Number	is the value of $n + i$ . (This is used internally to refer to $s_i$ in the intermediate output.)
Row	gives the name of $v_i$ .
State	gives the state of $v_i$ (LL if active on its lower bound, UL if active on its upper bound, EQ if active and fixed, BS if inactive when $s_i$ is basic and SBS if inactive when $s_i$ is superbasic).

A key is sometimes printed before `State` to give some additional information about the state of a variable. Note that unless the optional parameter **Scale Option** = 0 (default value = 1 or 2) is specified, the tests for assigning a key are applied to the variables of the scaled problem.

- A *Alternative optimum possible.* The variable is nonbasic, but its reduced gradient is essentially zero. This means that if the variable were allowed to start moving away from its bound, there would be no change in the value of 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 basic or superbasic, but it is equal to (or very close to) one of its bounds.
- I *Infeasible.* The variable is basic or superbasic and is currently violating one of its bounds by more than the value of the optional parameter **Feasibility Tolerance** (default value =  $\max(10^{-6}, \sqrt{\epsilon})$ , where  $\epsilon$  is the *machine precision*; see Section 11.2).
- N *Not precisely optimal.* The variable is nonbasic or superbasic. If the value of the reduced gradient for the variable exceeds the value of the optional parameter **Optimality Tolerance** (default value =  $\max(10^{-6}, \sqrt{\epsilon})$ ; see Section 11.2), the solution would not be declared optimal because the reduced gradient for the variable would not be considered negligible.

Activity	is the value of $v_i$ at the final iterate.
Slack Activity	is the value by which the row differs from its nearest bound. (For the free row (if any), it is set to Activity.)
Lower Bound	is the lower bound specified for the variable. None indicates that $BL(j) \leq -bigbnd$ .
Upper Bound	is the upper bound specified for the variable. None indicates that $BU(j) \geq bigbnd$ .
Dual Activity	is the value of the dual variable $\pi_i$ (the Lagrange multiplier for $v_i$ ; see Section 10.3). For FP problems, $\pi_i$ is set to zero.
i	gives the index $i$ of the $i$ th row.

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

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