# Package: quadprog (via r-universe)

September 20, 2024

Type Package		
Title Functions to Solve Quadratic Programming Problems		
Version 1.5-8		
<pre>Date 2019-11-20 Author S original by Berwin A. Turlach <berwin.turlach@gmail.com> R</berwin.turlach@gmail.com></pre>		
		<b>Date/Publication</b> 2019-11-20 08:20:02 UTC
		Contents
		solve.QPsolve.QP.compact
		Index
		solve.QP Solve a Quadratic Programming Problem
Decemention		

This routine implements the dual method of Goldfarb and Idnani (1982, 1983) for solving quadratic programming problems of the form  $\min(-d^Tb+1/2b^TDb)$  with the constraints  $A^Tb>=b_0$ .

2 solve.QP

### Usage

```
solve.QP(Dmat, dvec, Amat, bvec, meq=0, factorized=FALSE)
```

# Arguments

Dmat matrix appearing in the quadratic function to be minimized.

dvec vector appearing in the quadratic function to be minimized.

Amat matrix defining the constraints under which we want to minimize the quadratic

function.

bvec vector holding the values of  $b_0$  (defaults to zero).

meq the first meq constraints are treated as equality constraints, all further as inequal-

ity constraints (defaults to 0).

factorized logical flag: if TRUE, then we are passing  $R^{-1}$  (where  $D = R^T R$ ) instead of the

matrix D in the argument Dmat.

### Value

a list with the following components:

solution vector containing the solution of the quadratic programming problem.

value scalar, the value of the quadratic function at the solution

unconstrained.solution

vector containing the unconstrained minimizer of the quadratic function.

iterations vector of length 2, the first component contains the number of iterations the

algorithm needed, the second indicates how often constraints became inactive

after becoming active first.

Lagrangian vector with the Lagragian at the solution.

iact vector with the indices of the active constraints at the solution.

# References

D. Goldfarb and A. Idnani (1982). Dual and Primal-Dual Methods for Solving Strictly Convex Quadratic Programs. In J. P. Hennart (ed.), Numerical Analysis, Springer-Verlag, Berlin, pages 226–239.

D. Goldfarb and A. Idnani (1983). A numerically stable dual method for solving strictly convex quadratic programs. *Mathematical Programming*, **27**, 1–33.

### See Also

solve.QP.compact

solve.QP.compact 3

## **Examples**

```
##
## Assume we want to minimize: -(0.5.0) %*% b + 1/2 b^T b
## under the constraints: A^T b >= b0
## with b0 = (-8,2,0)^T
           (-4 2 0)
## and
        A = (-3 \ 1 \ -2)
##
            (0 0 1)
##
## we can use solve.QP as follows:
Dmat
          \leftarrow matrix(0,3,3)
diag(Dmat) <- 1</pre>
dvec
          <-c(0,5,0)
          <- matrix(c(-4,-3,0,2,1,0,0,-2,1),3,3)
Amat
          <-c(-8,2,0)
bvec
solve.QP(Dmat,dvec,Amat,bvec=bvec)
```

solve.QP.compact

Solve a Quadratic Programming Problem

# Description

This routine implements the dual method of Goldfarb and Idnani (1982, 1983) for solving quadratic programming problems of the form  $\min(-d^Tb + 1/2b^TDb)$  with the constraints  $A^Tb >= b_0$ .

## Usage

```
solve.QP.compact(Dmat, dvec, Amat, Aind, bvec, meq=0, factorized=FALSE)
```

# **Arguments**

Dmat	matrix appearing in the quadratic function to be minimized.
dvec	vector appearing in the quadratic function to be minimized.
Amat	matrix containing the non-zero elements of the matrix $A$ that defines the constraints. If $m_i$ denotes the number of non-zero elements in the $i$ -th column of $A$ then the first $m_i$ entries of the $i$ -th column of Amat hold these non-zero elements. (If $maxmi$ denotes the maximum of all $m_i$ , then each column of Amat may have arbitrary elements from row $m_i+1$ to row $maxmi$ in the $i$ -th column.)
Aind	matrix of integers. The first element of each column gives the number of non-zero elements in the corresponding column of the matrix $A$ . The following entries in each column contain the indexes of the rows in which these non-zero elements are.
bvec	vector holding the values of $b_0$ (defaults to zero).
meq	the first meq constraints are treated as equality constraints, all further as inequality constraints (defaults to $0$ ).

factorized logical flag: if TRUE, then we are passing  $R^{-1}$  (where  $D=R^TR$ ) instead of the matrix D in the argument Dmat.

4 solve.QP.compact

### Value

a list with the following components:

solution vector containing the solution of the quadratic programming problem.

value scalar, the value of the quadratic function at the solution

unconstrained.solution

vector containing the unconstrained minimizer of the quadratic function.

iterations vector of length 2, the first component contains the number of iterations the

algorithm needed, the second indicates how often constraints became inactive

after becoming active first.

Lagrangian vector with the Lagragian at the solution.

iact vector with the indices of the active constraints at the solution.

## References

D. Goldfarb and A. Idnani (1982). Dual and Primal-Dual Methods for Solving Strictly Convex Quadratic Programs. In J. P. Hennart (ed.), Numerical Analysis, Springer-Verlag, Berlin, pages 226–239.

D. Goldfarb and A. Idnani (1983). A numerically stable dual method for solving strictly convex quadratic programs. *Mathematical Programming*, **27**, 1–33.

#### See Also

```
solve.QP
```

### **Examples**

```
##
## Assume we want to minimize: -(0 5 0) %*% b + 1/2 b^T b
                              A^T b >= b0
## under the constraints:
## with b0 = (-8,2,0)^T
## and
            (-4 \ 2 \ 0)
##
        A = (-3 \ 1 \ -2)
            (0 0 1)
##
## we can use solve.QP.compact as follows:
##
Dmat
           \leftarrow matrix(0,3,3)
diag(Dmat) <- 1</pre>
dvec
           <-c(0,5,0)
Aind
           <- rbind(c(2,2,2),c(1,1,2),c(2,2,3))
           \leftarrow rbind(c(-4,2,-2),c(-3,1,1))
Amat
           <-c(-8,2,0)
bvec
solve.QP.compact(Dmat,dvec,Amat,Aind,bvec=bvec)
```

# **Index**

```
* optimize

solve.QP, 1

solve.QP.compact, 3

solve.QP, 1, 4

solve.QP.compact, 2, 3
```