

Package ‘LowRankQP’

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Title Low Rank Quadratic Programming

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Maintainer ORPHANED

Description This package contains routines and documentation for solving quadratic programming problems where the hessian is represented as the product of two matrices.

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X-CRAN-Comment

Previous maintainer address bounced in Oct 2012, maintainer is unresponsive so orphaned 2012-12-06.

NeedsCompilation yes

R topics documented:

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|-----------|--|
| LowRankQP | <i>Solve Low Rank Quadratic Programming Problems</i> |
|-----------|--|

Description

This routine implements a primal-dual interior point method solving quadratic programming problems of the form

$$\begin{array}{ll}
\min & d^T \alpha + 1/2 \alpha^T H \alpha \\
\text{such that} & A \alpha = b \\
& 0 \leq \alpha \leq u
\end{array}$$

with dual

$$\begin{array}{ll}
\min & 1/2 \alpha^T H \alpha + \beta^T b + \xi^T u \\
\text{such that} & H \alpha + c + A^T \beta - \xi = 0 \\
& \xi, \beta \geq 0
\end{array}$$

where $H = V$ if V is square and $H = VV^T$ otherwise.

Usage

```
LowRankQP(Vmat, dvec, Amat, bvec, uvec, method="PFCF", verbose=FALSE, niter=200)
```

Arguments

| | |
|---------|---|
| Vmat | 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 (defaults to zero). |
| uvec | vector holding the values of u . |
| method | Method used for inverting $H+D$ where D is full rank diagonal. If V is square: <ul style="list-style-type: none"> • 'LU': Use LU factorization. (More stable) • 'CHOL': Use Cholesky factorization. (Faster) If V is not square: <ul style="list-style-type: none"> • 'SMW': Use Sherman-Morrison-Woodbury (Faster) • 'PFCF': Use Product Form Cholesky Factorization (More stable) |
| verbose | Display iterations of LowRankQP. |
| niter | Number of iteration to perform. |

Value

a list with the following components:

| | |
|-------|--|
| alpha | vector containing the solution of the quadratic programming problem. |
| beta | vector containing the solution of the dual of quadratic programming problem. |
| xi | vector containing the solution of the dual quadratic programming problem. |
| zeta | vector containing the solution of the dual quadratic programming problem. |

References

- Ormerod, J.T., Wand, M.P. and Koch, I. (2005). Penalised spline support vector classifiers: computational issues, in A.R. Francis, K.M. Matakawie, A. Oshlack, G.K. Smyth (eds). Proceedings of the 20th International Workshop on Statistical Modelling, Sydney, Australia, pp. 33-47.
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- Ferris, M. C. and Munson, T. S. (2003). Interior point methods for massive support vector machines. SIAM Journal on Optimization, 13, 783-804.
- Fine, S. and Scheinberg, K. (2001). Efficient SVM training using low-rank kernel representations. Journal of Machine Learning Research, 2, 243-264.
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Examples

```
library(LowRankQP)

# Assume we want to minimize: (0 -5 0 0 0 0) %*% alpha + 1/2 alpha[1:3]^T alpha[1:3]
# under the constraints:      A^T alpha = b
# with b = (-8,  2,  0)^T
# and      (-4  2  0)
#      A = (-3  1 -2)
#           ( 0  0  1)
#           (-1  0  0)
#           ( 0 -1  0)
#           ( 0  0 -1)
# alpha >= 0
#
# (Same example as used in quadprog)
#
# we can use LowRankQP as follows:

Vmat      <- matrix(0,6,6)
diag(Vmat) <- c(1, 1,1,0,0,0)
dvec      <- c(0,-5,0,0,0,0)
Amat      <- matrix(c(-4,-3,0,-1,0,0,2,1,0,0,-1,0,0,-2,1,0,0,-1),6,3)
bvec      <- c(-8,2,0)
uvec      <- c(100,100,100,100,100,100)
LowRankQP(Vmat,dvec,t(Amat),bvec,uvec,method="CHOL")

# Now solve the same problem except use low-rank V

Vmat      <- matrix(c(1,0,0,0,0,0,0,1,0,0,0,0,0,1,0,0,0),6,3)
dvec      <- c(0,-5,0,0,0,0)
Amat      <- matrix(c(-4,-3,0,-1,0,0,2,1,0,0,-1,0,0,-2,1,0,0,-1),6,3)
bvec      <- c(-8,2,0)
uvec      <- c(100,100,100,100,100,100)
LowRankQP(Vmat,dvec,t(Amat),bvec,uvec,method="SMW")
```

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