# Package 'splitFeas'

April 11, 2018

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backtrack

Backtracking Line Search

# Description

Given a descent direction backtrack computes a step size that ensures sufficient decrease in an objective.

# Usage

```
backtrack(x, dx, f, df, alpha = 0.01, beta = 0.8)
```

#### **Arguments**

x	Current iterate
dx	Descent direction
f	objective function
df	gradient of objective function
alpha	sufficient decrease parameter
beta	sufficient decrease parameter

ddg

Compute the approximate Hessian of the majorization.

# Description

ddg computes the Hessian of the majorization of the proximity function.

# Usage

```
ddg(x, v, w, hgrad)
```

dg 3

# Arguments

X	non-anchor point
V	weights for first set of constraints
W	weights for second set of constraints
hgrad	Handle for output mapping Jacobian

# **Examples**

```
set.seed(12345)
n <- 10
p <- 2
x <- matrix(rnorm(p),p,1)
v <- 1
w <- 1
A <- matrix(rnorm(n*p),n,p)
hgrad <- function(x) {return(t(A))}
sol <- ddg(x,v,w,hgrad)</pre>
```

dg

Compute the gradient of the majorization.

# Description

dg computes the gradient of the majorization of the proximity function.

# Usage

```
dg(x, xa, v, w, plist1, plist2, h, hgrad)
```

x	non-anchor point
xa	Anchor point
V	weights for first set of constraints
W	weights for second set of constraints
plist1	list of projection functions for first set of constraints; each takes a single point and returns its projection
plist2	list of projection functions for second set of constraints; each takes a single point and returns its projection
h	Function handle for output mapping
hgrad	Handle for output mapping Jacobian

4 nmsfp\_mm

mmqn_step	MM-quasi-Newton step	
mmqri_5ccp	mm quasi newion step	

# Description

mmqn\_step computes a single step.

# Usage

```
mmqn\_step(x, v, w, plist1, plist2, f, df, h, hgrad, woodbury = TRUE)
```

# Arguments

x	Current iterate
V	weights for first set of constraints
w	weights for second set of constraints
plist1	list of projection functions for first set of constraints; each takes a single point and returns its projection
plist2	list of projection functions for second set of constraints; each takes a single point and returns its projection
f	objective function
df	gradient of objective function
h	Function handle for output mapping
hgrad	Handle for output mapping Jacobian
woodbury	Boolean: TRUE to use the Woodbury inversion formula

nmsfp_mm	MM algorithm for nonlinear multiple-sets split feasibility problem

# Description

nmsfp\_mm uses quasi-Newton updates to solve the nonlinear multiple-sets split feasibility problem.

# Usage

```
nmsfp_mm(x0, v, w, plist1, plist2, f, df, h, hgrad, tol = 1e-10, max_iter = 1000)
```

nmsfp\_mmqn 5

# Arguments

x0	Initial iterate
V	weights for first set of constraints
W	weights for second set of constraints
plist1	list of projection functions for first set of constraints; each takes a single point and returns its projection
plist2	list of projection functions for second set of constraints; each takes a single point and returns its projection
f	objective function
df	gradient of objective function
h	Function handle for output mapping
hgrad	Handle for output mapping Jacobian
tol	Stopping tolerance
max_iter	Maximum number of iterations

# See Also

mmqn\_step

nmsfp_mmqn	MM algorithm (accelerated) for nonlinear multiple-sets split feasibil-
	ity problem

# Description

 $nmsfp\_mmqn\ uses\ quasi-Newton\ updates\ to\ solve\ the\ nonlinear\ multiple-sets\ split\ feasibility\ problem.$ 

# Usage

```
nmsfp_mmqn(x0, v, w, plist1, plist2, f, df, h, hgrad, qn = 5, tol = 1e-10, \max_i ter = 1000)
```

x0	Initial iterate
V	weights for first set of constraints
W	weights for second set of constraints
plist1	list of projection functions for first set of constraints; each takes a single point and returns its projection
plist2	list of projection functions for second set of constraints; each takes a single point and returns its projection

6 nmsfp\_sap

f	objective function
df	gradient of objective function
h	Function handle for output mapping
hgrad	Handle for output mapping Jacobian
qn	number of secants
tol	convergence tolerance
max_iter	maximum number of iterations

# See Also

mmqn\_step, qnamm

nmsfp_sap	Self-adaptive	projection-type	method	algorithm	for	nonlinear
multiple-sets split feasibility problem						

# Description

nmsfp\_sap performs the self-adaptive projection-type method of Li et al.

# Usage

```
nmsfp_sap(x0, v, w, plist1, plist2, proj, f, df, h, hgrad, delta = 0.3, mu = 0.7, beta0 = 1, tol = 1e-10, max_iter = 1000)
```

x0	Initial iterate
V	weights for first set of constraints
W	weights for second set of constraints
plist1	list of projection functions for first set of constraints; each takes a single point and returns its projection
plist2	list of projection functions for second set of constraints; each takes a single point and returns its projection
proj	handle to projection operation.
f	objective function
df	gradient of objective function
h	Function handle for output mapping
hgrad	Handle for output mapping Jacobian
delta	step-size parameter
mu	step-size parameter
beta0	initial
tol	Tolerance
max_iter	Maximum number of iterations

nmsfp\_sap\_one\_step 7

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nmsfp	san	one	sten

One step of self-adaptive projection-type method for the NMSFP

# Description

nmsfp\_sap\_one\_step performs the self-adaptive projection-type method of Li et al.

# Usage

```
nmsfp_sap_one_step(x, delta, mu, tau, gamma, df, proj)
```

# **Arguments**

X	current iterate
delta	step-size parameter
mu	step-size parameter
tau	step-size parameter
gamma	step-size parameter
df	handle to gradient of objective function
proj	handle to projection operation.

#### References

Li, Han, and Zhang. (2013) A self-adaptive projection-type method for nonlinear multiple-sets split feasibility problem, Inverse Problems in Science and Engineering, 21(1):155-170.

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Projection onto a ball

# Description

project\_ball computes the Euclidean projection of a point onto a ball.

# Usage

```
project_ball(x, center, r)
```

Х	Point to project
center	Center of the sphere
r	Radius of the sphere

8 project\_halfspace

#### **Examples**

```
set.seed(12345)
p <- 3
center <- rnorm(p)
r <- runif(1)
x <- rnorm(p)
y <- project_ball(x,center,r)</pre>
```

project\_cube

Project onto a cube

#### **Description**

project\_cube computes the Euclidean projection of a point onto a cube.

# Usage

```
project_cube(x, center, r)
```

#### **Arguments**

x Point to projectcenter Center of the squarer Half the length of a side

#### **Examples**

```
set.seed(12345)
p <- 3
center <- matrix(rnorm(p),p,1)
r <- runif(1)
x <- matrix(rnorm(p),p,1)
y <- project_cube(x,center,r)</pre>
```

project\_halfspace

Projection onto a halfspace

# **Description**

project\_halfspace computes the Euclidean projection of a point onto a closed half-space. The function returns the projection onto the set

#### Usage

```
project_halfspace(x, a, b)
```

project\_square 9

# **Arguments**

X	Point to project
a	is the normal vector
b	is the threshold

# **Examples**

```
set.seed(12345)
p <- 3
a <- matrix(rnorm(p),p,1)
a <- a/norm(a,'f')
b <- runif(1)
x <- matrix(rnorm(p),p,1)
y <- project_halfspace(x,a,b)</pre>
```

project\_square

Project onto a square

# **Description**

project\_square computes the Euclidean projection of a point onto a square.

# Usage

```
project_square(x, center, r)
```

# Arguments

x Point to projectcenter Center of the squarer Half the length of a side

#### **Examples**

```
set.seed(12345)
p <- 2
center <- matrix(rnorm(p),p,1)
r <- runif(1)
x <- matrix(rnorm(p),p,1)
y <- project_square(x,center,r)</pre>
```

10 qnamm

proximity	Proximity function

# Description

proximity computes the proximity function.

# Usage

```
proximity(x, v, w, plist1, plist2, h)
```

# Arguments

<ul> <li>w weights for first set of constraints</li> <li>w weights for second set of constraints</li> <li>plist1 list of projection functions for first set of constraints; each takes a single point and returns its projection</li> <li>plist2 list of projection functions for second set of constraints; each takes a single point and returns its projection</li> <li>h Function handle for output mapping</li> </ul>	X	Current iterate
plist1 list of projection functions for first set of constraints; each takes a single point and returns its projection  plist2 list of projection functions for second set of constraints; each takes a single point and returns its projection	V	weights for first set of constraints
and returns its projection  plist2 list of projection functions for second set of constraints; each takes a single point and returns its projection	W	weights for second set of constraints
and returns its projection	plist1	
h Function handle for output mapping	plist2	· ·
	h	Function handle for output mapping

qnamm	Quasi-Newton acceleration of MM algorithm
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# Description

qnamm performs Quasi-Newton acceleration of an MM algorithm.

# Usage

```
qnamm(x, fx_mm, qn, fx_obj, max_iter = 100, tol = 1e-06, ...)
```

Χ	initial iterate
fx_mm	MM algorithm map
qn	number of secants
fx_obj	handle to objective function
max_iter	maximum number of iterations
tol	convergence tolerance
	Additional arguments to pass to fx_mm

softmax 11

#### References

H Zhou, D Alexander, and K Lange. (2011) A quasi-Newton acceleration method for high-dimensional optimization algorithms, Statistics and Computing, 21(2):261-273.

 $\operatorname{softmax}$ 

Compute soft-max

# Description

softmax returns the soft maximum of a collection of reals.

# Usage

```
softmax(x, a = 100)
```

#### **Arguments**

x input

a scaling factor

# Examples

```
set.seed(12345)
n <- 10
x <- rnorm(n)
softmax(x)</pre>
```

split\_feasibility

 $split\_feasibility$ 

# Description

split\_feasibility

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wood\_inv\_solve

Compute the inverse approximate Hessian of the majorization using the Woodbury inversion formula. wood\_inv\_solve computes the inverse of the Hessian term of the majorization of the proximity function using the Woodbury formula. The function mmqn\_step invokes wood\_inv\_solve instead of ddg if the argument woodbury=TRUE. This should be used when  $p \ll n$ .

#### **Description**

Compute the inverse approximate Hessian of the majorization using the Woodbury inversion formula. wood\_inv\_solve computes the inverse of the Hessian term of the majorization of the proximity function using the Woodbury formula. The function mmqn\_step invokes wood\_inv\_solve instead of ddg if the argument woodbury=TRUE. This should be used when p « n.

# Usage

```
wood_inv_solve(x, v, w, hgrad, df)
```

# Arguments

X	non-anchor point	

v weights for first set of constraintsw weights for second set of constraintshgrad Handle for output mapping Jacobian

df Right hand side

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