

# Package ‘comlasso’

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**Type** Package

**Title** Primal path algorithm for compositional data (comlasso)

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**Depends** R (>=3.2.2), Rglpk (>=0.6-2)

**Description** This package provides functions for fitting the entire solution path algorithm for compositional data analysis

**License** GPL(>=2)

**First version** Sep/19/2016

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comlasso	<i>Primal path algorithm for compositional data</i>
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## Description

Fit the entire solution path of lasso under zero sum constraint for regression & classification problem

## Usage

```
comlasso(ltype = c("regression", "classification"), n, p, K, X.raw, y,  
  gam = 1e2, weights = NULL, max.steps = length(K)*min(n,sum(K))+1, lam.min = 0,  
  tol = 1e-08, trace = FALSE)
```

**Arguments**

<code>ltype</code>	choose either one among <code>c("regression", "classification")</code>
<code>n</code>	sample size
<code>p</code>	number of components
<code>K</code>	vector of the number of variables belonging to each component
<code>X.raw</code>	$n$ times $K[1]+K[2]+\dots+K[p]$ matrix
<code>y</code>	response vector with length $n$
<code>gam</code>	Knot value for Huber loss function. For classification, default value of <code>gam</code> is $1e2$ . For classification problem, <code>gam</code> should be $\leq 1$ .
<code>weights</code>	For adaptive lasso penalty,
<code>max.steps</code>	steps: default value is $\min(n, K) + 1$
<code>lam.min</code>	minimum value of $\lambda$ . Default value is 0
<code>tol</code>	relative zero
<code>trace</code>	To check status of solution

**Details**

See example codes

**Value**

return classo object and see `predict.classo`

**Note**

$n < p$  & blockwise zero sum for several components

**Author(s)**

Jong-Jun Jeon, Yondai Kim, Hyowon An, Sungho Won and Hosik Choi (2016) Primal path algorithm for compositional data analysis

**References**

Primal path algorithm for compositional data analysis

**See Also**

`predict.class`, `stability.classo`

**Examples**

```
data(sand)
mdat <- sand[,-1]
x <- as.matrix(mdat[,1:3])/100
x <- log(x)
y <- log(mdat[,4])

n <- 39; p <- 3
# lasso
```

```

fit1 <- comlasso("regression", n=n, p=1, K=p, X.raw=x, y=y, weights=c(1,1,1))
w <- 1/(abs(fit1$beta[[1]])/sum(abs(fit1$beta[[1]])))
fit2 <- comlasso("regression", n=n, p=1, K=p, X.raw=x, y=y, weights=w)

par(mfrow=c(1,2))
matplot(t(fit1$beta.rec[[1]]),type="b",xaxt="n",ylab=expression(hat(beta)[j](lambda)),
  ylim=c(-0.6,0.6), main="lasso")
axis(1,at=c(1,2,3))
abline(h=0,lty=2,lwd=0.7)
matplot(t(fit2$beta.rec[[1]]),type="b",xaxt="n",ylab=expression(hat(beta)[j](lambda)),
  ylim=c(-0.6,0.6), main="alasso")
axis(1,at=c(1,2,3))
abline(h=0,lty=2,lwd=0.7)

```

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predict.comlasso

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*Make predictions or extract coefficients from a fitted comlasso model*


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## Description

While comlasso() produces the entire path of solutions, predict.comlasso allows one to extract a prediction at a particular point along the path.

## Usage

```

## S3 method for class 'comlasso'
predict(object, newx, s, type=c("fit","coefficients"),
  mode=c("step","fraction","norm","lambda"), ...)

```

## Arguments

object	comlasso object
newx	If type="fit", then newx should be the x values at which the fit is required. If type="coefficients", then newx can be omitted.
s	a value, or vector of values, indexing the path. Its values depends on the mode= argument. By default (mode="step"), s should take on values between 0 and p (e.g., a step of 1.3 means .3 of the way between step 1 and 2.)
type	If type="fit", predict returns the fitted values. If type="coefficients", predict returns the coefficients. Abbreviations allowed. mode
mode	mode="step" means the s= argument indexes the comlasso step number, and the coefficients will be returned corresponding to the values corresponding to step s. If mode="fraction", then s should be a number between 0 and 1, and it refers to the ratio of the L1 norm of the coefficient vector, relative to the norm at the full LS solution. mode="norm" means s refers to the L1 norm of the coefficient vector. mode="lambda" uses the lasso regularization parameter for s; for other models it is the maximal correlation (does not make sense for comlasso models).
...	additional parameters

## Details

comlasso is described in detail in Jeon, Kim, An, Won and Choi (2016). It computes the complete lasso solution simultaneously for ALL values of the shrinkage parameter in the same computational cost as a least squares fit without penalization.

**Value**

Either a vector/matrix of fitted values, or a vector/matrix of coefficients.

**Note**

`predict` / `predict.comlasso`

**Author(s)**

Jong-Jun Jeon, Yondai Kim, Hyowon An, Sungho Won and Hosik Choi (2016) Primal path algorithm for compositional data analysis

**References**

Jeon, Kim, An, Won and Choi (2016) "Primal path algorithm for compositional data analysis"

**See Also**

`stability_comlasso`

**Examples**

```
data(sand)
mdat <- sand[,-1]
x <- as.matrix(mdat[,1:3])/100
x <- log(x)
y <- log(mdat[,4])

n <- 39; p <- 3
# lasso
fit <- comlasso("regression", n=n, p=1, K=p, X.raw=x, y=y, weights=c(1,1,1))
predict(fit, x, s=c(fit$lambda,0.5), type="fit",mode="lambda")
predict(fit, x, s=c(0,0.5,1), type="fit",mode="fraction")
predict(fit, x, s=1:2, type="fit",mode="step")
predict(fit, x, s=0.5, type="fit",mode="norm")
```

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