# Package 'superpc'

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Type Package
Title Supervised Principal Components
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$ \begin{array}{c} \textbf{Description} & \text{Does prediction in the case of a censored survival outcome, or a regression outcome, using the ``supervised principal component" approach. 'Superpc' is especially useful for high-dimensional data when the number of features p dominates the number of samples n (p >> n paradigm), as generated, for instance, by high-throughput technologies. \\ \end{array} $
<b>Depends</b> R ( $>= 3.5.0$ )
Imports survival, stats, graphics, grDevices
NeedsCompilation no
<pre>URL http://www-stat.stanford.edu/~tibs/superpc,</pre>
https://github.com/jedazard/superpc
Repository CRAN
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R topics documented:
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superpc.cv

Cross-validation for supervised principal components

# Description

This function uses a form of cross-validation to estimate the optimal feature threshold in supervised principal components

# Usage

#### **Arguments**

fit	Object returned by superpc.train
data	Data object of form described in superpc.train documentation
n.threshold	Number of thresholds to consider. Default 20.
n.fold	Number of cross-validation folds. default is around 10 (program pick a convenient value based on the sample size
folds	List of indices of cross-validation folds (optional)
n.components	Number of cross-validation components to use: 1,2 or 3.
min.features	Minimum number of features to include in determining range for threshold. Default 5.

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max.features Maximum number of features to include in determining range for threshold.

Default is total number of features in the dataset

compute.fullcv Should full cross-validation be done?

compute.preval Should full pre-validation be done?

xl.mode Used by Excel interface only xl.time Used by Excel interface only

xl.prevfit Used by Excel interface only

#### **Details**

This function uses a form of cross-validation to estimate the optimal feature threshold in supervised principal components. To avoid prolems with fitting Cox models to samll validation datastes, it uses the "pre-validation" approach of Tibshirani and Efron (2002)

#### Value

threshold Vector of thresholds considered

nonzero Number of features exceeding each value of the threshold

scor.preval Likelihood ratio scores from pre-validation

scor Full CV scores

folds Indices of CV folds used

featurescores.folds

Feature scores for each fold

v.preval The pre-validated predictors

type problem type call calling sequence

#### Author(s)

• "Eric Bair, Ph.D."

• "Jean-Eudes Dazard, Ph.D."

• "Rob Tibshirani, Ph.D."

Maintainer: "Jean-Eudes Dazard, Ph.D."

#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

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#### **Examples**

superpc.decorrelate

Decorrelate features with respect to competing predictors

#### Description

Fits a linear model to the features as a function of some competing predictors. Replaces the features by the residual from this fit. These "decorrelated" features are then used in the superpc model building process, to explicitly look for predictors that are independent of the competing predictors. Useful for example, when the competing predictors are clinical predictors like stage, grade etc.

#### Usage

# Arguments

x matrix of features. Different features in different rows, one observation per column competing.predictors

List of one or more competing predictors. Discrete predictors should be factors

#### Value

Returns lm (linear model) fit of rows of x on compeiting predictors.

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#### Author(s)

- "Eric Bair, Ph.D."
- "Jean-Eudes Dazard, Ph.D."
- "Rob Tibshirani, Ph.D."

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#### References

• E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.

• E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

#### **Examples**

```
set.seed(332)
#generate some data
x \leftarrow matrix(rnorm(50*30), ncol=30)
y <- 10 + svd(x[1:50,])$v[,1] + .1*rnorm(30)
censoring.status <- sample(c(rep(1,20), rep(0,10)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
competing.predictors <- list(pred1=rnorm(30),</pre>
                               pred2=as.factor(sample(c(1,2),
                                                 replace=TRUE,
                                                 size=30)))
#decorrelate x. Remember to decorrelate test data in the same way, before making predictions.
foo <- superpc.decorrelate(x, competing.predictors)</pre>
xnew <- t(foo$res)</pre>
#now use xnew in superpc
data <- list(x=xnew,</pre>
              y=y,
              censoring.status=censoring.status,
              featurenames=featurenames)
a <- superpc.train(data, type="survival")</pre>
#etc.
```

superpc.fit.to.outcome

Fit predictive model using outcome of supervised principal components

superpc.fit.to.outcome

#### **Description**

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Fit predictive model using outcome of supervised principal components, via either coxph (for surival data) or lm (for regression data)

# Usage

#### **Arguments**

fit Object returned by superpc.train.

data.test Data object for prediction. Same form as data object documented in superpc.train.

score Supervised principal component score, from superpc.predict.

competing.predictors

Optional - a list of competing predictors to be included in the model.

print Should a summary of the fit be printed? Default TRUE.

iter.max Max number of iterations used in predictive model fit. Default 5. Currently only

relevant for Cox PH model.

#### Value

Returns summary of coxph or lm fit.

# Author(s)

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- "Rob Tibshirani, Ph.D."

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#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

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#### **Examples**

```
set.seed(332)
#generate some data
x <- matrix(rnorm(50*30), ncol=30)</pre>
y \leftarrow 10 + svd(x[1:50,])v[,1] + .1*rnorm(30)
ytest <- 10 + \text{svd}(x[1:50,])$v[,1] + .1*rnorm(30)
censoring.status <- sample(c(rep(1,20), rep(0,10)))
censoring.status.test <- sample(c(rep(1,20), rep(0,10)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
data <- list(x=x,</pre>
              censoring.status=censoring.status,
              featurenames=featurenames)
data.test <- list(x=x,</pre>
                   censoring.status=censoring.status.test,
                   featurenames=featurenames)
a <- superpc.train(data, type="survival")</pre>
fit <- superpc.predict(a,</pre>
                         data,
                         data.test,
                         threshold=1.0,
                         n.components=1,
                         prediction.type="continuous")
superpc.fit.to.outcome(a,
                         data,
                         fit$v.pred)
```

superpc.listfeatures Return a list of the important predictors

# Description

Return a list of the important predictor

#### Usage

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#### **Arguments**

data Data object

train.obj Object returned by superpc.train

fit.red Object returned by superpc.predict.red, applied to training set

fitred.cv (Optional) object returned by superpc.predict.red.cv

num.features Number of features to list. Default is all features.

component.number

Number of principal component (1,2, or 3) used to determine feature importance scores

#### Value

Returns matrix of features and their importance scores, in order of decreasing absolute value of importance score. The importance score is the correlation of the reduced predictor and the full supervised PC predictor. It also lists the raw score- for survival data, this is the Cox score for that feature; for regression, it is the standardized regression coefficient. If fitted.cv is supplied, the function also reports the average rank of the gene in the cross-validation folds, and the proportion of times that the gene is chosen (at the given threshold) in the cross-validation folds.

#### Author(s)

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#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

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superpc.lrtest.curv

Compute values of likelihood ratio test from supervised principal components fit

#### **Description**

Compute values of likelihood ratio test from supervised principal components fit

# Usage

# **Arguments**

object Object returned by superpc.train.

data List of training data, of form described in superpc.train documentation.

newdata List of test data; same form as training data.

n. components Number of principal components to compute. Should be 1,2 or 3.

threshold Set of thresholds for scores; default is n.threshold values equally spaced over

the range of the feature scores.

n. threshold Number of thresholds to use; default 20. Should be 1,2 or 3.

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#### Value

1rtest Values of likelihood ratio test statistic

comp2 Description of 'comp2'

threshold Thresholds used

num. features Number of features exceeding threshold

type Type of outcome variable

call calling sequence

## Author(s)

• "Eric Bair, Ph.D."

• "Jean-Eudes Dazard, Ph.D."

• "Rob Tibshirani, Ph.D."

Maintainer: "Jean-Eudes Dazard, Ph.D."

#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

```
set.seed(332)
#generate some data
x \leftarrow matrix(rnorm(50*30), ncol=30)
y <- 10 + svd(x[1:50,])$v[,1] + .1*rnorm(30)
ytest <- 10 + \text{svd}(x[1:50,]) v[,1] + .1*rnorm(30)
censoring.status <- sample(c(rep(1,20), rep(0,10)))
censoring.status.test <- sample(c(rep(1,20), rep(0,10)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
data <- list(x=x,</pre>
              censoring.status=censoring.status,
              featurenames=featurenames)
data.test <- list(x=x,</pre>
                   y=ytest,
                   censoring.status=censoring.status.test,
                   featurenames=featurenames)
a <- superpc.train(data, type="survival")</pre>
aa <- superpc.lrtest.curv(a, data, data.test)</pre>
#superpc.plot.lrtest(aa)
```

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superpc.news

Display the superpc Package News

# **Description**

Function to display the log file NEWS of updates of the **superpc** package.

# Usage

```
superpc.news(...)
```

# **Arguments**

.. Further arguments passed to or from other methods.

#### Value

None.

# Note

End-user function.

# Author(s)

- "Eric Bair, Ph.D."
- "Jean-Eudes Dazard, Ph.D."
- "Rob Tibshirani, Ph.D."

Maintainer: "Jean-Eudes Dazard, Ph.D."

## References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

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# **Description**

Plot likelhiood ratio test statistics from output of superpc.predict

#### Usage

#### **Arguments**

```
object.lrtestcurv
Output from superpc.lrtest.curv
call.win.metafile
For use by PAM Excel interface
```

#### Author(s)

- "Eric Bair, Ph.D."
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Maintainer: "Jean-Eudes Dazard, Ph.D."

#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

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superpc.plotcv

Plot output from superpc.cv

#### **Description**

Plots pre-validation results from plotcy, to aid in choosing best threshold

#### Usage

#### **Arguments**

object Object returned by superpc.cv.

cv.type Type of cross-validation used - "full" (Default; this is "standard" cross-validation; recommended) and "preval"- pre-validation.

smooth Should plot be smoothed? Only relevant to "preval". Default FALSE.

smooth.df Degrees of freedom for smooth.spline, default 10. If NULL, then degrees of freedom is estimated by cross-validation.

call.win.metafile

Ignore: for use by PAM Excel program.

Additional plotting args to be passed to matplot.

# Author(s)

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#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

#### **Examples**

superpc.plotred.lrtest

Plot likelihood ratio test statistics from supervised principal components predictor

#### **Description**

Plot likelihood ratio test statistics from supervised principal components predictor

#### Usage

#### **Arguments**

```
object.lrtestred
```

Output from either superpc.predict.red or superpc.predict.redcv

call.win.metafile

Used only by PAM Excel interface call to function

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#### Author(s)

- "Eric Bair, Ph.D."
- "Jean-Eudes Dazard, Ph.D."
- "Rob Tibshirani, Ph.D."

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#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

```
## Not run:
set.seed(332)
#generate some data
x \leftarrow matrix(rnorm(50*30), ncol=30)
y <- 10 + svd(x[1:50,])$v[,1] + .1*rnorm(30)
ytest <- 10 + \text{svd}(x[1:50,])$v[,1] + .1*rnorm(30)
censoring.status <- sample(c(rep(1,20), rep(0,10)))
censoring.status.test \leftarrow sample(c(rep(1,20), rep(0,10)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
data <- list(x=x,</pre>
              censoring.status=censoring.status,
              featurenames=featurenames)
data.test <- list(x=x,</pre>
                    censoring.status=censoring.status.test,
                    featurenames=featurenames)
a <- superpc.train(data, type="survival")</pre>
aa <- superpc.cv(a, data)</pre>
fit.red <- superpc.predict.red(a,</pre>
                                  data,
                                  data.test,
                                  .6)
fit.redcv <- superpc.predict.red.cv(fit.red,</pre>
                                        data,
                                        .6)
superpc.plotred.lrtest(fit.redcv)
## End(Not run)
```

superpc.predict

superpc.predict	Form principal components predictor from a trained superpc object

#### Description

Computes supervised principal components, using scores from "object"

#### Usage

#### **Arguments**

object Obect returned by superpc.train

data List of training data, of form described in superpc.train documentation,

newdata List of test data; same form as training data

threshold Threshold for scores: features with abs(score) > threshold are retained. n.components Number of principal components to compute. Should be 1,2 or 3.

prediction.type

"continuous" for raw principal component(s); "discrete" for principal component categorized in equal bins; "nonzero" for indices of features that pass the

threshold

n. class Number of classes into which predictor is binned (for prediction.type="discrete"

#### Value

v.pred Supervised principal componients predictor
u U matrix from svd of feature matrix x
d singual values from svd of feature matrix x
which.features Indices of features exceeding threshold

n.components Number of supervised principal components requested

call calling sequence

#### Author(s)

• "Eric Bair, Ph.D."

• "Jean-Eudes Dazard, Ph.D."

• "Rob Tibshirani, Ph.D."

Maintainer: "Jean-Eudes Dazard, Ph.D."

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#### References

• E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.

• E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

#### **Examples**

```
set.seed(332)
#generate some data
x <- matrix(rnorm(50*30), ncol=30)</pre>
y \leftarrow 10 + svd(x[1:50,])v[,1] + .1*rnorm(30)
ytest <- 10 + \text{svd}(x[1:50,])$v[,1] + .1*rnorm(30)
censoring.status <- sample(c(rep(1,20), rep(0,10)))
censoring.status.test <- sample(c(rep(1,20), rep(0,10)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
data <- list(x=x,</pre>
              censoring.status=censoring.status,
              featurenames=featurenames)
data.test <- list(x=x,</pre>
                   censoring.status=censoring.status.test,
                   featurenames=featurenames)
a <- superpc.train(data, type="survival")</pre>
fit <- superpc.predict(a,</pre>
                         data,
                         data.test,
                         threshold=1.0,
                         n.components=1)
plot(fit$v.pred, ytest)
```

superpc.predict.red Feature selection for supervised principal components

#### **Description**

Forms reduced models to approximate the supervised principal component predictor.

#### Usage

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```
n.components=3,
n.shrinkage=20,
shrinkages=NULL,
compute.lrtest=TRUE,
sign.wt="both",
prediction.type=c("continuous", "discrete"),
n.class=2)
```

# **Arguments** fit

Training data object, of form described in superpc.train dcoumentation

data.test
Test data object; same form as train

threshold
Feature score threshold; usually estimated from superpc.cv

Number of principal components to examine; should equal 1,2, etc up to the number of components used in training

n.shrinkage
Number of shrinkage values to consider. Default 20.

shrinkages
Shrinkage values to consider. Default NULL.

compute.1rtest Should the likelihood ratio test be computed? Default TRUE sign.wt Signs of feature weights allowed: "both", "pos", or "neg"

prediction.type

Type of prediction: "continuous" (Default) or "discrete". In the latter, superprc

score is divided into n.class groups

n.class Number of groups for discrete predictor. Default 2.

Object returned by superpc.train

#### **Details**

Soft-thresholding by each of the "shrinkages" values is applied to the PC loadings. This reduce the number of features used in the model. The reduced predictor is then used in place of the supervised PC predictor.

#### Value

shrinkages	Shrinkage values used
lrtest.reduced	Likelihood ratio tests for reduced models
num.features	Number of features used in each reduced model
feature.list	List of features used in each reduced model
coef	Least squares coefficients for each reduced model
import	Importance scores for features
wt	Weight for each feature, in constructing the reduced predictor
v.test	Outcome predictor from reduced models. Array of n.shrinkage by (number of test observations)

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v.test.1df	Outcome combined predictor from reduced models. Array of n.shrinkage by (number of test observations)
n.components	Number of principal components used
type	Type of outcome
call	calling sequence

#### Author(s)

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• "Rob Tibshirani, Ph.D."

Maintainer: "Jean-Eudes Dazard, Ph.D."

#### References

• E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.

• E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

```
set.seed(332)
#generate some data
x \leftarrow matrix(rnorm(50*30), ncol=30)
y <- 10 + svd(x[1:50,])$v[,1] + .1*rnorm(30)
ytest <- 10 + \text{svd}(x[1:50,])$v[,1] + .1*rnorm(30)
censoring.status <- sample(c(rep(1,20), rep(0,10)))
censoring.status.test <- sample(c(rep(1,20), rep(0,10)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
data <- list(x=x,</pre>
              censoring.status=censoring.status,
              featurenames=featurenames)
data.test <- list(x=x,</pre>
                   y=ytest,
                   censoring.status=censoring.status.test,
                   featurenames=featurenames)
a <- superpc.train(data, type="survival")</pre>
fit.red <- superpc.predict.red(a,</pre>
                                 data.test,
                                 threshold=.6)
superpc.plotred.lrtest(fit.red)
```

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```
superpc.predict.red.cv
```

Cross-validation of feature selection for supervised principal components

# Description

Applies superpc.predict.red to cross-validation folds generates in superpc.cv. Uses the output to evaluate reduced models, and compare them to the full supervised principal components predictor.

# Usage

#### **Arguments**

fitred Output of superpc.predict.red
--------------------------------------

fitcv Output of superpc.cv data Training data object

threshold Feature score threshold; usually estimated from superpc.cv sign.wt Signs of feature weights allowed: "both", "pos", or "neg"

#### Value

1rtest.reduced Likelihood ratio tests for reduced models

components Number of supervised principal components used

v.preval.red Outcome predictor from reduced models. Array of num.reduced.models by

(number of test observations)

type Type of outcome call calling sequence

#### Author(s)

- "Eric Bair, Ph.D."
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- "Rob Tibshirani, Ph.D."

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#### References

• E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.

• E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

#### **Examples**

```
## Not run:
set.seed(332)
#generate some data
x <- matrix(rnorm(50*20), ncol=20)</pre>
y \leftarrow 10 + svd(x[1:10,])$v[,1] + .1*rnorm(20)
ytest <- 10 + \text{svd}(x[1:10,])$v[,1] + .1*rnorm(20)
censoring.status <- sample(c(rep(1,15), rep(0,5)))
censoring.status.test <- sample(c(rep(1,15), rep(0,5)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
data <- list(x=x,</pre>
              y=y,
              censoring.status=censoring.status,
              featurenames=featurenames)
data.test <- list(x=x,</pre>
                    y=ytest,
                    censoring.status=censoring.status.test,
                    featurenames=featurenames)
a <- superpc.train(data, type="survival")</pre>
aa <- superpc.cv(a, data)</pre>
fit.red <- superpc.predict.red(a,</pre>
                                  data,
                                  data.test,
                                  threshold=.6)
fit.redcv <- superpc.predict.red.cv(fit.red,</pre>
                                        aa,
                                        data,
                                        threshold=.6)
## End(Not run)
```

superpc.predictionplot

Plot outcome predictions from superpc

# **Description**

Plots outcome predictions from superpc

#### Usage

# Arguments

train.obj	Object returned by superpc.train	
data	List of training data, of form described in superpc.train documentation	
data.test	List of test data; same form as training data	
threshold	Threshold for scores: features with abs(score) > threshold are retained.	
n.components	Number of principal components to compute. Should be 1,2 or 3.	
	Number of classes for survival stratification. Only applicable for survival data. Default $2$ .	
shrinkage	Shrinkage to be applied to feature loadings. Default is NULL, meaning no shrinkage	
call.win.metafile		
	Used only by Excel interface call to function	

Used only by Excel interface call to function

#### Author(s)

- "Eric Bair, Ph.D."
- "Jean-Eudes Dazard, Ph.D."
- "Rob Tibshirani, Ph.D."

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#### References

- E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.
- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

```
set.seed(332)
#generate some data
x <- matrix(rnorm(50*30), ncol=30)
y <- 10 + svd(x[1:50,])$v[,1] + .1*rnorm(30)
ytest <- 10 + svd(x[1:50,])$v[,1] + .1*rnorm(30)</pre>
```

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superpc.rainbowplot

Make rainbow plot of superpc and compeiting predictors

# **Description**

Makes a heatmap display of outcome predictions from superpc, along with expected survival time, and values of competing predictors.

#### Usage

#### **Arguments**

data List of (test) data, of form described in superpc.train documentation pred Superpc score from superpc.predict or superpc.predict.red sample.labels Vector of sample labels of test data competing.predictors

List of competing predictors to be plotted call.win.metafile

Used only by Excel interface call to function

#### Details

Any censored survival times are estimated by E(T|T>C), where C is the observed censoring time and the Kaplan-Meier estimate from the training set is used to estimate the expectation.

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#### Author(s)

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- E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

```
set.seed(332)
#generate some data
x \leftarrow matrix(rnorm(50*30), ncol=30)
y \leftarrow 10 + svd(x[1:50,]) v[,1] + .1*rnorm(30)
ytest <- 10 + \text{svd}(x[1:50,])$v[,1] + .1*rnorm(30)
censoring.status <- sample(c(rep(1,20), rep(0,10)))
censoring.status.test <- sample(c(rep(1,20), rep(0,10)))
featurenames <- paste("feature", as.character(1:50), sep="")</pre>
competing.predictors.test <- list(pred1=rnorm(30),</pre>
                                     pred2=as.factor(sample(c(1,2),
                                                      replace=TRUE,
                                                       size=30)))
data <- list(x=x,</pre>
              censoring.status=censoring.status,
              featurenames=featurenames)
data.test <- list(x=x,</pre>
                   y=ytest,
                   censoring.status=censoring.status.test,
                   featurenames=featurenames)
sample.labels <- paste("te", as.character(1:20), sep="")</pre>
a <- superpc.train(data, type="survival")</pre>
pred <- superpc.predict(a,</pre>
                          data,
                          data.test,
                          threshold=.25,
                          n.components=1)$v.pred
superpc.rainbowplot(data,
                      pred,
                      sample.labels,
                      competing.predictors=competing.predictors.test)
```

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superpc.train	Prediction by supervised principal components

#### **Description**

Does prediction of a quantitative regression or survival outcome, by the supervised principal components method.

# Usage

#### **Arguments**

data	Data object with components x- p by n matrix of features, one observation per column; y- n-vector of outcome measurements; censoring.status- n-vector of censoring censoring.status (1= died or event occurred, 0=survived, or event was censored), needed for a censored survival outcome
type	Problem type: "survival" for censored survival outcome, or "regression" for simple quantitative outcome
s0.perc	Factor for denominator of score statistic, between 0 and 1: the percentile of standard deviation values added to the denominator. Default is 0.5 (the median)

# **Details**

Compute wald scores for each feature (gene), for later use in superpc.predict and superpc.cv

#### Value

feature.scores

Score for each feature (gene)

type problem type

s0.perc Factor for denominator of score statistic

call calling sequence

# Author(s)

- "Eric Bair, Ph.D."
- "Jean-Eudes Dazard, Ph.D."
- "Rob Tibshirani, Ph.D."

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#### References

• E. Bair and R. Tibshirani (2004). "Semi-supervised methods to predict patient survival from gene expression data." PLoS Biol, 2(4):e108.

• E. Bair, T. Hastie, D. Paul, and R. Tibshirani (2006). "Prediction by supervised principal components." J. Am. Stat. Assoc., 101(473):119-137.

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