

Package ‘predbayescor’

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Title Classification rule based on Bayesian naive Bayes models with feature selection bias corrected

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Depends R (>= 2.5.1)

Description This software is used to predict the binary response based on high dimensional features, for example gene expression data. The data are modelled with Bayesian naive Bayes models. When a large number of features are available, one may like to select only a subset of features to use, typically those features strongly correlated with the response in training cases. Such a feature selection procedure is however invalid since the relationship between the response and the features will appear stronger. This package provides a way to avoid this bias and yields well-calibrated prediction for the test cases.

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URL [\{\}url{http://www.r-project.org}](http://www.r-project.org)[\protect\T1\textbraceright, \{\}url{http://math.usask.ca/~longhai}](http://math.usask.ca/~longhai)[\protect\T1\textbraceright](http://math.usask.ca/~longhai)

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evaluate_by_loss	<i>calculating the total loss of prediction results</i>
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Description

Calculates the average loss of predictions based on threshold with threshold. Note that this threshold has 1-1 mapping with the ratio of the loss of assigning 0 to 1 to the loss of assigning 1 to 0: $\text{threshold} = 1 - 1 / (1 + \text{ratio})$.

Usage

```
evaluate_by_loss ( y.true, pred.prob, threshold=0.5)
```

Arguments

y.true	a vector containing the true response.
pred.prob	a vector containing the predictive probabilities.
threshold	When predictive probability is greater than threshold, the response is predicted as 1.

Value

loss	the average loss, with attrib "sd" storing the estimate of the standard error of this loss.
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gendata.bayes	<i>Generate binary data with Bayesian naive Bayes Models</i>
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Description

"gendata.bayes" generates data (both training and test data) with Bayesian naive Bayes model. The prior distribution of "theta" is uniform(0,1). The value of "alpha" is given by argument alpha, which controls the the overall relationship between the response and the predictor variables.

Usage

```
gendata.bayes(n0, n1, m0, m1, p, alpha)
```

Arguments

n0	the number of class 0 in training data
n1	the number of class 1 in training data
m0	the number of class 0 in test data
m1	the number of class 1 in test data
p	the number of features
alpha	a parameter controlling the dependency between the features and the response

Value

train	the training data, with the row standing for the cases and the first column being the response
test	the test data, of the same format as "train"

See Also

[predict_bayes](#)

predict_bayes	<i>Classification rule based on Bayesian naive Bayes models with feature selection bias corrected</i>
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Description

predict_bayes predicts the binary response based on high dimensional binary features modeled by Bayesian naive Bayes models. It also accepts real values but they will be converted into binary by thresholding at the medians estimated from the data. A smaller number of features can be selected based on the correlations with the response. The bias due to the selection procedure can be corrected. cv.bayes is the short-cut function for cross-validation with predict_bayes.

Usage

```
predict_bayes(
  test, train, is.binary.features=FALSE, k,
  subset.sel=1:nrow(train),
  theta0=0, no.theta=20,
  alpha.shape=0.5, alpha.rate=5, no.alpha=5,
  correct=TRUE, no.theta.adj=20)

cv.bayes(
  data, is.binary.features=FALSE, no.folds=10, k,
  theta0=0, no.theta=20,
  alpha.shape=0.5, alpha.rate=5, no.alpha=5,
  correct=TRUE, no.theta.adj=20)
```

Arguments

test	a test data, a matrix, i.e. the data for which we want to predict the responses. The row stands for the cases. The first column is the binary response, which could be NA if they are missing.
train	a training data, of the same format as test
data	a data used in cross-validation, of the same format as test
no.folds	the number of blocks the data is divided into in cross-validation

is.binary.features	the indicator whether the features are binary
k	the number of features retained
subset.sel	the indice of training cases used to select features
theta0	the prior of "theta" is uniform over (theta0,1-theta0)
no.theta	the parameter in Simpson's rule used to evaluate the integration w.r.t. "theta". The integrant is evaluated at $2*(no.theta)+1$ points.
alpha.shape	the shape parameter of the inverse Gamma, which is the prior distribution of "alpha"
alpha.rate	the rate parameter of the inverse Gamma, as above
no.alpha	the number of "alpha"'s used in mid-point rule, which is used to approximate the integral with respect to "alpha".
correct	the indicator whether the correction method shall be applied
no.theta.adj	a parameter of Simpson's rule, which is used to evaluate the integration with respect to "theta" in calculating the adjustment factor

Value

prediction	a matrix showing the detailed prediction result: the 1st column being the true responses, the 2nd being the predicted responses, the 3rd being the predictive probabilities of class 1 and the 4th being the indicator whether wrong prediction is made.
amlp	the average minus log probabilities
error.rate	the ratio of wrong prediction
mse	the average square error of the predictive probabilities
summary.pred	tabular display of the predictive probabilities and the actual fraction of class 1.
alpha.prior.adj.post	a matrix showing the detailed information about the "alpha"'s, the 1st column being the values of "alpha"'s, the 2nd being the adjustment factor, i.e. probability that feature is discarded by the cutoff used in the feature selection, the 3rd being the log of the 2nd column times the numbers of discarded features, the 4th being the posterior probabilities
features.selected	The features selected using correlation criterion

References

<http://math.usask.ca/~longhai/doc/naivebayes/naivebayes.abstract.html>

See Also

[gendata.bayes](#)

Examples

```
#generate a dataset
d <- gendata.bayes(100,100,500,500,1000,400)

#do prediction with correction applied
pred.d.cor <- predict_bayes(d$test,d$train,TRUE,10,,0,20,0.5,5,20,TRUE,40)

#do prediction without correction applied
pred.d.uncor <- predict_bayes(d$test,d$train,TRUE,10,,0,20,0.5,5,20,FALSE,40)

#do 5-fold cross-validation on the training data with correction applied
cv.dtr.cor <- cv.bayes(d$train,TRUE,5,10,0,20,0.5,5,20,TRUE,40)
```

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