

# Package ‘bnclassify’

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**Title** Learning Bayesian Network Classifiers from Data

**Description** Algorithms for learning Bayesian network classifiers from data.

**Version** 0.2.0

**URL** <http://github.com/bmihaljevic/bnclassify>

**BugReports** <http://github.com/bmihaljevic/bnclassify/issues>

**Depends** R (>= 3.2.0)

## Imports

assertthat (>= 0.1),entropy(>= 1.2.0),crossval(>= 1.0.2),graph(>= 1.42.0),matrixStats(>= 0.14.0),pryr(>= 0.1.1),RBGL  
(8)

**Suggests** gRain(>= 1.2-3),gRbase(>= 1.7-

0.1),mlr(>= 2.2),testthat(>= 0.8.1),knitr(>= 1.10.5),microbenchmark(>= 1.4-  
2),ParamHelpers(>= 1.5),Rgraphviz(>= 2.8.1),rmarkdown(>= 0.7)

**License** GPL (>= 2)

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**VignetteBuilder** knitr

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as_mlr	<i>Bridge to mlr</i>
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**Description**

Only as\_mlr should be called by the user.

**Usage**

```
makeRLearner.bnc()
```

```
as_mlr(x, dag, id = "1")
```

```
predictLearner.bnc(.learner, .model, .newdata, ...)
```

**Arguments**

x	A <a href="#">bnc_bn</a> object.
dag	A logical. Whether to repeat structure learning on each fold or just parameter learning.
id	A character.
.learner, .task, .subset, .weights, .model, .newdata	Internal.

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bnc	<i>Learn a structure and parameters.</i>
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**Description**

A convenience function to learn the structure and parameters in a single call.

**Usage**

```
bnc(dag_learner, class, dataset, smooth, dag_args = NULL)
```

**Arguments**

dag_learner	A character. Name of the structure learning function.
class	A character. Name of the class variable.
dataset	The data frame from which to learn network structure and parameters.
smooth	A nonnegative numeric. The smoothing value for Bayesian parameter estimation.
dag_args	A list. Optional additional arguments to dag_learner.

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`bnclassify`*Algorithms for learning Bayesian network classifiers from data.*

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**Description**

Algorithms for learning Bayesian network classifiers from data.

**References**

Bielza C and Larrañaga P (2014), Discrete Bayesian network classifiers: A survey. *ACM Computing Surveys*, **47**(1), Article 5.

Friedman N, Geiger D and Goldszmidt M (1997). Bayesian network classifiers. *Machine Learning*, **29**, pp. 131–163.

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`car`*Car Evaluation Data Set.*

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**Description**

Car Evaluation Data Set.

**Format**

A data.frame with 7 columns and 1728 rows.

**Source**

<http://sourceforge.net/projects/weka/files/datasets/UCIandStatLib/uci-20070111.tar.gz>

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`chowliu`*Chow-Liu ODE.*

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**Description**

Chow-Liu ODE.

**Usage**

```
chowliu(class, dataset, score = "loglik", blacklist = NULL, root = NULL)
```

**Arguments**

<code>class</code>	character
<code>dataset</code>	data frame
<code>score</code>	character
<code>blacklist</code>	character matrix
<code>root</code>	character

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cv	<i>Stratified cross validation estimate of predictive accuracy.</i>
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### Description

Stratified cross validation estimate of predictive accuracy.

### Usage

```
cv(x, dataset, k, dag, smooth = NULL)
```

### Arguments

x	List of <a href="#">bnc_bn</a> . The classifiers to evaluate.
dataset	The data frame on which to evaluate the classifiers.
k	An integer. The number of folds.
dag	A logical. Whether to repeat structure learning on each fold or just parameter learning.
smooth	A nonnegative numeric. The smoothing value for Bayesian parameter estimation.

### Value

A numeric vector. The predictive accuracy of each classifier in x.

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family	<i>Gets the parents of a node in the graph</i>
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### Description

Gets the parents of a node in the graph

### Usage

```
family(x, g)
```

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learn_params	<i>Learn the parameters of a Bayesian network structure.</i>
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### Description

lpawnb weights the features' CPTs according to the AWNB method.

### Usage

```
awnb(class, dataset, trees = 10, bootstrap_size = 0.5)
```

```
lp(x, dataset, smooth)
```

```
lpawnb(x, dataset, smooth, trees, bootstrap_size)
```

### Arguments

dataset	The data frame from which to estimate network parameters.
trees	An integer. The number of bootstrap samples to generate.
bootstrap_size	A numeric. The size of the bootstrap subsample, relative to the size of dataset (given in [0,1]).
x	a <a href="#">bnc_dag</a> object. The Bayesian network structure.
smooth	A nonnegative numeric. The smoothing value for Bayesian parameter estimation.

### References

Mark Hall (2004). A decision tree-based attribute weighting filter for naive Bayes. *Knowledge-based Systems*, **20**(2), 120-126.

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nb	<i>Returns a naive Bayes network structure.</i>
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### Description

Returns a naive Bayes network structure.

### Usage

```
nb(class, dataset = NULL, features = NULL)
```

### Arguments

class	A character. Name of the class variable.
dataset	The data frame from which to learn the classifier.
features	A character vector. The names of the features. This argument is ignored if dataset is provided.

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predict.bnc_bn	<i>Predicts class label or class posterior probability distributions.</i>
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### Description

Predicts class label or class posterior probability distributions.

### Usage

```
## S3 method for class 'bnc_bn'
predict(object, newdata, prob = FALSE, ...)
```

### Arguments

object	A <a href="#">bnc_bn</a> object.
newdata	A data frame containing observations whose class has to be predicted.
prob	A logical. Whether class posterior probability should be returned.

### Details

Ties are resolved randomly. Inference is much slower if newdata contains NAs.

### Value

If prob=FALSE, then returns a length- $N$  factor with the same levels as the class variable in  $x$ , where  $N$  is the number of rows in newdata. Each element of the factor is the most likely class for the corresponding row in newdata. If prob=TRUE, returns a  $N$  by  $C$  numeric matrix, where  $C$  is the number of classes; each row corresponds to the class posterior of the instance.

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tan_chowliu	<i>Learns a tree augmented naive Bayes classifier (TAN).</i>
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### Description

Learns a one-dependence Bayesian classifier using Chow-Liu's algorithm. The structure is learned so that either likelihood, the BIC or AIC scores are maximized; the first option corresponds to the well-known tree augmented naive Bayes (Friedman et al., 1997).

### Usage

```
tan_cl(class, dataset, score = "loglik", blacklist = NULL, root = NULL)
```

**Arguments**

class	A character. Name of the class variable.
dataset	The data frame from which to learn the classifier.
score	A character. The score to be maximized. 'loglik', 'bic', and 'aic' return the maximum likelihood, maximum BIC and maximum AIC tree/forest, respectively.
blacklist	A character matrix. Edges that may be blacklisted from the resulting structure.
root	A character. The feature to be used as root of the augmenting tree. Only one feature can be supplied, even in case of an augmenting forest. This argument is optional.

**Details**

Edges with negative BIC or AIC weights are blacklisted from the final structure. The structure thus might be a forest rather than a tree.

**References**

Friedman N, Geiger D and Goldszmidt M (1997). Bayesian network classifiers. *Machine Learning*, **29**, pp. 131–163.

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to_grain	<i>To grain</i>
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**Description**

To grain

**Usage**

to\_grain(x)

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voting	<i>Congress Voting Data Set.</i>
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**Description**

Congress Voting Data Set.

**Format**

A data.frame with 17 columns and 435 rows.

**Source**

<http://sourceforge.net/projects/weka/files/datasets/UCIandStatLib/uci-20070111.tar.gz>

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 wrapper

*Learns Bayesian network classifiers in a wrapper fashion.*


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### Description

bsej is the backward *sequential elimination and joining* algorithm whereas fssj is the *forward sequential selection and joining* algorithms for learning a semi-naive Bayes classifier (Pazzani, 1996). tanhc Learns a tree augmented naive Bayes with a greedy hill-climbing search. tanhc is the super-parent variant of tanhc.

### Usage

```
fssj(class, dataset, k, epsilon = 0.01, smooth = 0.01)
```

```
bsej(class, dataset, k, epsilon = 0.01, smooth = 0.01)
```

```
tanhc(class, dataset, k, epsilon = 0.01, smooth = 0.01)
```

```
tanhc_sp(class, dataset, k, epsilon = 0.01, smooth = 0.01)
```

### Arguments

class	A character. Name of the class variable.
dataset	The data frame from which to learn the classifier.
epsilon	A numeric. Minimum absolute improvement required to keep searching.
smooth	A nonnegative numeric. The smoothing value for Bayesian parameter estimation.

### References

Pazzani M (1996). Constructive induction of Cartesian product attributes. In *Proceedings of the Information, Statistics and Induction in Science Conference (ISIS-1996)*, pp. 66-77

Koegh E and Pazzani M (2002). Learning the structure of augmented Bayesian classifiers. In *International Journal on Artificial Intelligence Tools*, **11**(4), pp. 587-601.



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