Package: wevid (via r-universe)

September 10, 2024

Type Package

Title Quantifying Performance of a Binary Classifier Through Weight of Evidence

Version 0.6.2.9000

Date 2019-09-12

Description The distributions of the weight of evidence (log Bayes factor) favouring case over noncase status in a test dataset (or test folds generated by cross-validation) can be used to quantify the performance of a diagnostic test (McKeigue (2019), <doi:10.1177/0962280218776989>). The package can be used with any test dataset on which you have observed case-control status and have computed prior and posterior probabilities of case status using a model learned on a training dataset. To quantify how the predictor will behave as a risk stratifier, the quantiles of the distributions of weight of evidence in cases and controls can be calculated and plotted.

Depends R (>= 2.10)

License GPL-3

URL http://www.homepages.ed.ac.uk/pmckeigu/preprints/classify/wevidtutorial.html

LazyLoad yes

Imports ggplot2, mclust, pROC (>= 1.9), reshape2, zoo

Suggests testthat (>= 2.0.0)

ByteCompile TRUE

Encoding UTF-8

RoxygenNote 6.1.1

Repository https://mcol.r-universe.dev

RemoteUrl https://github.com/mcol/wevid

RemoteRef HEAD

RemoteSha 9483c34bb6826c6d6c7bb89de1bb8ddc018963b9

Contents

wevid-package			•	•					•		•		•	•			•	•				2
plotcumfreqs									•													3
plotroc									•													4
plotWdists									•													4
prop.belowthreshold									•													5
recalibrate.p									•													6
summary-densities .									•													6
Wdensities									•													7
weightsofevidence .									•													8
wevid.datasets				•	•	•	•		•			•	•			•	•	•				9
																						10

Index

wevid-package

Quantifying performance of a diagnostic test using the sampling distribution of the weight of evidence favouring case over noncase status

Description

The **wevid** package provides functions for quantifying the performance of a diagnostic test (or any other binary classifier) by calculating and plotting the distributions in cases and noncases of the weight of evidence favouring case over noncase status.

Details

The distributions of the weight of evidence (log Bayes factor) favouring case over noncase status in a test dataset (or test folds generated by cross-validation) can be used to quantify the performance of a diagnostic test.

In comparison with the C-statistic (area under ROC curve), the expected weight of evidence (expected information for discrimination) has several advantages as a summary measure of predictive performance. To quantify how the predictor will behave as a risk stratifier, the quantiles of the distributions of weight of evidence in cases and controls can be calculated and plotted.

This package can be used with any test dataset on which you have observed case-control status and have computed prior and posterior probabilities of case status using a model learned on a training dataset. Therefore, you should have computed on a test dataset (or on test folds used for cross-validation):

- 1. The prior probability of case status (this may be just the frequency of cases in the training data).
- 2. The posterior probability of case status (using the model learned on the training data to predict on the test data).
- 3. The observed case status (coded as 0 for noncases, 1 for cases).

The main function of the package is Wdensities which computes the crude and model-based densities of weight of evidence in cases and controls. Once these are computed, they can be plotted with plotWdists and plotcumfreqs. Summary statistics can be reported with summary.

plotcumfreqs

Author(s)

Paul McKeigue <paul.mckeigue@ed.ac.uk>

References

Paul McKeigue (2019), Quantifying performance of a diagnostic test as the expected information for discrimination: Relation to the C-statistic. *Statistical Methods for Medical Research*, 28 (6), 1841-1851. https://doi.org/10.1177/0962280218776989.

See Also

Useful links:

• http://www.homepages.ed.ac.uk/pmckeigu/preprints/classify/wevidtutorial.html

1	plotcumfreas	Plot the	cumulative	frequency	distributions	in cases	and in	controls
	proteumineqs	1 ioi ine	cumulaive	јгедиенс у	aismounons	in cuses	unu in	comitors

Description

Plot the cumulative frequency distributions in cases and in controls

Usage

```
plotcumfreqs(densities)
```

Arguments

densities Densities object produced by Wdensities.

Value

A ggplot object representing the cumulative frequency distributions of the smoothed densities of the weights of evidence in cases and in controls.

Examples

```
data(cleveland)
densities <- with(cleveland, Wdensities(y, posterior.p, prior.p))
plotcumfreqs(densities)</pre>
```

plotroc

Description

While the crude ROC curve can be non-concave and is generally not smooth, the model-based ROC curve is always concave, as the corresponding densities have been adjusted to be mathematically consistent.

Usage

plotroc(densities)

Arguments

densities Densities object produced by Wdensities.

Value

A ggplot object representing crude and model-based ROC curves.

Examples

```
data(cleveland)
densities <- with(cleveland, Wdensities(y, posterior.p, prior.p))
plotroc(densities)</pre>
```

plotWdists

Plot the distribution of the weight of evidence in cases and in controls

Description

Plot the distribution of the weight of evidence in cases and in controls

Usage

```
plotWdists(densities, distlabels = c("Crude", "Model-based"))
```

Arguments

densities	Densities object produced by Wdensities.
distlabels	Character vector of length 2 to be used to label the crude and the model-based
	curves (in that order).

prop.belowthreshold

Value

A ggplot object representing the distributions of crude and model-based weights of evidence in cases and in controls.

Examples

```
data(cleveland)
densities <- with(cleveland, Wdensities(y, posterior.p, prior.p))
plotWdists(densities)
# Example which requires fitting a mixture distribution
data(fitonly)
densities <- with(fitonly, Wdensities(y, posterior.p, prior.p))
# truncate spike
plotWdists(densities) + ggplot2::scale_y_continuous(limits=c(0, 0.5))
```

prop.belowthreshold	Proportions of cases and controls below a threshold of weight of evi-
	dence

Description

Proportions of cases and controls below a threshold of weight of evidence

Usage

```
prop.belowthreshold(densities, w.threshold)
```

Arguments

densities	Densities object produced by Wdensities.
w.threshold	Threshold value of weight of evidence (natural logs)

Value

Numeric vector of length 2 listing the proportions of controls and cases with weight of evidence below the given threshold.

Examples

```
data(cleveland)
densities <- with(cleveland, Wdensities(y, posterior.p, prior.p))
w.threshold <- log(4) # threshold Bayes factor of 4
prop.belowthreshold(densities, w.threshold)</pre>
```

recalibrate.p

Description

Transforms posterior probabilities to logits, fits a logistic regression model and returns the predictive probabilities from this model.

Usage

```
recalibrate.p(y, posterior.p)
```

Arguments

У	Binary outcome label (0 for controls, 1 for cases).
posterior.p	Vector of posterior probabilities.

Value

Recalibrated posterior probabilities.

summary-densities Summary evaluation of predictive performance

Description

Summary evaluation of predictive performance

Usage

```
## S3 method for class 'Wdensities'
summary(object, ...)
## S3 method for class 'Wdensities'
mean(x, ...)
auroc.crude(densities)
```

auroc.model(densities)

lambda.crude(densities)

lambda.model(densities)

Wdensities

Arguments

object, x, densi	ties
	Densities object produced by Wdensities.
	Further arguments passed to or from other methods. These are currently ignored.

Value

summary returns a data frame that reports the number of cases and controls, the test log-likelihood, the crude and model-based C-statistic and expected weight of evidence Lambda.

mean returns a numeric vector listing the mean densities of the weight of evidence in controls and in cases.

auroc.crude and auroc.model return the area under the ROC curve according to the crude and the model-based densities of weight of evidence, respectively.

lambda.crude and lambda.model return the expected weight of evidence (expected information for discrimination) in bits from the crude and the model-based densities, respectively.

Examples

```
data(cleveland)
densities <- with(cleveland, Wdensities(y, posterior.p, prior.p))
summary(densities)
mean(densities)
auroc.model(densities)
lambda.model(densities)</pre>
```

Wdensities

Compute densities of weights of evidence in cases and controls

Description

The function computes smoothed densities of the weight of evidence in cases and in controls from the crude probabilities, then adjusts them to make them mathematically consistent so that $p(W_ctrl) = exp(-W) p(W_case)$.

Usage

```
Wdensities(y, posterior.p, prior.p, range.xseq = c(-25, 25),
    x.stepsize = 0.01, adjust.bw = 1, recalibrate = TRUE,
    debug = FALSE)
```

Arguments

У	Binary outcome label (0 for controls, 1 for cases).
posterior.p	Vector of posterior probabilities generated by using model to predict on test data.
prior.p	Vector of prior probabilities.
range.xseq	Range of points where the curves should be sampled.
x.stepsize	Distance between each point.
adjust.bw	Bandwidth adjustment for the Gaussian kernel density estimator. By default it is set to 1 (no adjustment), setting it to a value smaller/larger than 1 re- duces/increases the smoothing of the kernel. This argument is ignored if more than one mixture component is identified.
recalibrate	If TRUE (the default) the weights of evidence are calculated after the posterior probabilities have been recalibrated against y using a logistic regression model.
debug	If TRUE, the size of the adjustment is reported.

Details

If the sample distributions in cases and controls support a 2-component mixture model (based on model comparison with BIC) for the densities, this will be detected and a 2-component mixture model will be fitted before adjustment.

Value

A densities object that contains the information necessary to compute summary measures and generate plots.

Examples

```
data(cleveland)
densities <- with(cleveland, Wdensities(y, posterior.p, prior.p))
# Example which requires fitting a mixture distribution
data(fitonly)
densities <- with(fitonly, Wdensities(y, posterior.p, prior.p))</pre>
```

weightsofevidence Calculate weights of evidence in natural log units

Description

Calculate weights of evidence in natural log units

Usage

weightsofevidence(posterior.p, prior.p)

wevid.datasets

Arguments

posterior.p	Vector of posterior probabilities generated by using model to predict on test data.
prior.p	Vector of prior probabilities.

Value

The weight of evidence in nats for each observation.

Examples

```
data(cleveland) # load example dataset
W <- with(cleveland, weightsofevidence(posterior.p, prior.p))</pre>
```

wevid.datasets Example datasets

Description

Example datasets

The wevid package comes with the following dataset:

- cleveland is based on cross-validated prediction of coronary disease in the Cleveland Heart Study (297 observations).
- pima is based on cross-validated prediction of diabetes in Pima Native Americans (768 observations).
- fitonly is based on cross-validated prediction of colorectal cancer from fecal immunochemical test (FIT) only in Michigan (242 observations). As most controls and some cases have have zero values in the FIT test, to fit densities to the sampled values of weight of evidence in controls and cases it is necessary to specify spike-slab mixtures.

Format

Each dataset consists of a data frame with the following variables:

- prior.p: Prior probabilities of case status.
- posterior.p: Posterior probabilities of case status.
- y: Case-control status.

Source

http://www.homepages.ed.ac.uk/pmckeigu/preprints/classify/wevidtutorial.html

Index

* datasets wevid.datasets,9 auroc.crude (summary-densities), 6 auroc.model(summary-densities),6 cleveland (wevid.datasets), 9 fitonly (wevid.datasets), 9 lambda.crude(summary-densities), 6 lambda.model(summary-densities), 6 mean.Wdensities(summary-densities), 6 pima (wevid.datasets), 9 plotcumfreqs, 2, 3 plotroc, 4 plotWdists, 2, 4 prop.belowthreshold, 5 recalibrate.p, 6 summary, 2 ${\tt summary-densities}, {\bf 6}$ summary.Wdensities (summary-densities), 6 Wdensities, 2-5, 7, 7 weightsofevidence, 8 wevid (wevid-package), 2 wevid-package, 2wevid.datasets, 9