Package 'entropy'

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Author Jean Hausser and Korbinian Strimmer
Maintainer Korbinian Strimmer <strimmer@uni-leipzig.de></strimmer@uni-leipzig.de>
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Suggests
Description This package implements various estimators of entropy, such as the shrinkage estimator by Hausser and Strimmer, the maximum likelihood and the Millow-Madow estimator, variou Bayesian estimators, and the Chao-Shen estimator. It also offers an R interface to the NSB estimator. Furthermore, it provides functions for estimating mutual information.
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entropy-package The entropy Package

Description

This package implements various estimators of the Shannon entropy. Most estimators in this package can be applied in "small n, large p" situations, i.e. when there are many more bins than counts.

The main function of this package is entropy, which provides a unified interface to various entropy estimators.

If you use this package please cite: Jean Hausser and Korbinian Strimmer. 2009. Entropy inference and the James-Stein estimator, with application to nonlinear gene association networks. J. Mach. Learn. Res. 10: 1469-1484. Available online from http://jmlr.csail.mit.edu/papers/v10/hausser09a.html.

This paper contains a detailed statistical comparison of the estimators available in this package. It also describes the shrinkage entropy estimator entropy.shrink.

Author(s)

Jean Hausser and Korbinian Strimmer (http://strimmerlab.org/)

References

See website: http://strimmerlab.org/software/entropy/

See Also

entropy

entropy

Estimating Entropy From Observed Counts

Description

entropy estimates the Shannon entropy H of the random variable Y from the corresponding observed counts y.

freqs estimates bin frequencies from the counts y.

Usage

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Arguments

У	vector of counts.
method	the method employed to estimate entropy (see Details).
unit	the unit in which entropy is measured. The default is "nats" (natural units). For computing entropy in "bits" set unit="log2".
target	shrinkage target (for "shrink" option).
verbose	verbose option (for "shrink" option).
	option passed on to entropy.NSB.

Details

The entropy function allows to estimate entropy from observed counts by a variety of methods:

- method="ML":maximum likelihood, see entropy.empirical
- method="MM":bias-corrected maximum likelihood, see entropy.MillerMadow
- method="Jeffreys":entropy.Dirichlet with a=1/2
- method="Laplace":entropy.Dirichlet with a=1
- method="SG":entropy.Dirichlet with a=a=1/length(y)
- method="minimax":entropy.Dirichlet with a=sqrt(sum(y))/length(y
- method="CS":see entropy.ChaoShen
- method="NSB":see entropy.NSB
- method="shrink":see entropy.shrink

The freqs function estimates the underlying bin frequencies. Note that estimated frequencies are not available for method="MM", method="CS" and method="NSB". In these instances a vector containing NAs is returned.

Value

entropy returns an estimate of the Shannon entropy.

freqs returns a vector with estimated bin frequencies (if available).

Author(s)

```
Korbinian Strimmer (http://strimmerlab.org).
```

See Also

```
entropy-package
```

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Examples

```
# load entropy library
library("entropy")

# observed counts for each bin
y = c(4, 2, 3, 0, 2, 4, 0, 0, 2, 1, 1)
entropy(y, method="ML")
entropy(y, method="MM")
entropy(y, method="MM")
entropy(y, method="Jeffreys")
entropy(y, method="Laplace")
entropy(y, method="SG")
entropy(y, method="SG")
entropy(y, method="NSB")
entropy(y, method="NSB")
entropy(y, method="shrink")
```

entropy.ChaoShen

Chao-Shen Entropy Estimator

Description

entropy. ChaoShen estimates the Shannon entropy H of the random variable Y from the corresponding observed counts y using the method of Chao and Shen (2003).

Usage

```
entropy.ChaoShen(y, unit=c("log", "log2", "log10"))
```

Arguments

y vector of counts.

unit the unit in which entropy is measured. The default is "nats" (natural units). For

computing entropy in "bits" set unit="log2".

Details

The Chao-Shen entropy estimator (2003) is a Horvitz-Thompson (1952) estimator applied to the problem of entropy estimation, with additional coverage correction as proposed by Good (1953).

Note that the Chao-Shen estimator is not a plug-in estimator, hence there are no explicit underlying bin frequencies.

Value

entropy. ChaoShen returns an estimate of the Shannon entropy.

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Author(s)

Korbinian Strimmer (http://strimmerlab.org).

References

Chao, A., and T.-J. Shen. 2003. Nonparametric estimation of Shannon's index of diversity when there are unseen species in sample. Environ. Ecol. Stat. **10**:429-443.

Good, I. J. 1953. The population frequencies of species and the estimation of population parameters. Biometrika **40**:237-264.

Horvitz, D.G., and D. J. Thompson. 1952. A generalization of sampling without replacement from a finite universe. J. Am. Stat. Assoc. 47:663-685.

See Also

entropy, entropy.shrink, entropy.Dirichlet, entropy.NSB.

Examples

```
# load entropy library
library("entropy")

# observed counts for each bin
y = c(4, 2, 3, 0, 2, 4, 0, 0, 2, 1, 1)

# estimate entropy using Chao-Shen method
entropy.ChaoShen(y)

# compare to empirical estimate
entropy.empirical(y)
```

entropy.Dirichlet

Family of Dirichlet Entropy and Mutual Information Estimators

Description

entropy.Dirichlet estimates the Shannon entropy H of the random variable Y from the corresponding observed counts y by plug-in of Bayesian estimates of the bin frequencies using the Dirichlet-multinomial pseudocount model.

mi.Dirichlet estimates the corresponding mutual information of two random variables.

freqs.Dirichlet computes the Bayesian estimates of the bin frequencies using the Dirichlet-multinomial pseudocount model.

Usage

```
entropy.Dirichlet(y, a, unit=c("log", "log2", "log10"))
mi.Dirichlet(y, a, unit=c("log", "log2", "log10"))
freqs.Dirichlet(y, a)
```

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Arguments

y vector or matrix of counts.

a pseudocount per bin.

unit the unit in which entropy is measured. The default is "nats" (natural units). For

computing entropy in "bits" set unit="log2".

Details

The Dirichlet-multinomial pseudocount entropy estimator is a Bayesian plug-in estimator: in the definition of the Shannon entropy the bin probabilities are replaced by the respective Bayesian estimates of the frequencies, using a model with a Dirichlet prior and a multinomial likelihood.

The parameter a is a parameter of the Dirichlet prior, and in effect specifies the pseudocount per bin. Popular choices of a are:

- a=0:maximum likelihood estimator (see entropy.empirical)
- a=1/2:Jeffreys' prior; Krichevsky-Trovimov (1991) entropy estimator
- a=1:Laplace's prior
- a=1/length(y):Schurmann-Grassberger (1996) entropy estimator
- a=sqrt(sum(y))/length(y):minimax prior

The pseudocount a can also be a vector so that for each bin an individual pseudocount is added.

Value

entropy.Dirichlet returns an estimate of the Shannon entropy.
mi.Dirichlet returns an estimate of the mutual information.

freqs.Dirichlet returns the underlying frequencies.

Author(s)

Korbinian Strimmer (http://strimmerlab.org).

References

Agresti, A., and D. B. Hitchcock. 2005. Bayesian inference for categorical data analysis. Stat. Methods. Appl. **14**:297–330.

Krichevsky, R. E., and V. K. Trofimov. 1981. The performance of universal encoding. IEEE Trans. Inf. Theory **27**: 199-207.

Schurmann, T., and P. Grassberger. 1996. Entropy estimation of symbol sequences. Chaos **6**:41-427.

See Also

entropy, entropy.shrink, entropy.NSB, entropy.ChaoShen, entropy.empirical, entropy.plugin,
mi.plugin.

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Examples

```
# load entropy library
library("entropy")
# observed counts for each bin
y = c(4, 2, 3, 0, 2, 4, 0, 0, 2, 1, 1)
# Dirichlet estimate with a=0
entropy.Dirichlet(y, a=0)
# compare to empirical estimate
entropy.empirical(y)
# Dirichlet estimate with a=1/2
entropy.Dirichlet(y, a=1/2)
# Dirichlet estimate with a=1
entropy.Dirichlet(y, a=1)
# Dirichlet estimate with a=1/length(y)
entropy.Dirichlet(y, a=1/length(y))
# Dirichlet estimate with a=sqrt(sum(y))/length(y)
entropy.Dirichlet(y, a=sqrt(sum(y))/length(y))
# contigency table with counts for two discrete variables
y = rbind(c(1,2,3), c(6,5,4))
# Dirichlet estimate of mutual information (with a=1/2)
mi.Dirichlet(y, a=1/2)
```

entropy.empirical

Empirical Entropy and Mutual Information Estimator

Description

entropy.empirical estimates the Shannon entropy H of the random variable Y from the corresponding observed counts y by plug-in of the empirical frequencies.

mi.empirical computes the empirical mutual information from counts y.

freqs.empirical computes the empirical frequencies from counts y.

Usage

```
entropy.empirical(y, unit=c("log", "log2", "log10"))
mi.empirical(y, unit=c("log", "log2", "log10"))
freqs.empirical(y)
```

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Arguments

У	vector or matrix of counts.
unit	the unit in which entropy is measured. The default is "nats" (natural units). For
	computing entropy in "bits" set unit="log2".

Details

The empirical entropy estimator is a plug-in estimator: in the definition of the Shannon entropy the bin probabilities are replaced by the respective empirical frequencies.

The empirical entropy estimator is the maximum likelihood estimator. If there are many zero counts and the sample size is small it is very inefficient and also strongly biased.

Value

```
entropy.empirical returns an estimate of the Shannon entropy.
mi.empirical returns an estimate of the mutual information.
freqs.empirical returns the underlying frequencies.
```

Author(s)

```
Korbinian Strimmer (http://strimmerlab.org).
```

See Also

```
entropy, entropy.MillerMadow, entropy.plugin, mi.plugin.
```

Examples

```
# load entropy library
library("entropy")

# observed counts for each bin
y = c(4, 2, 3, 0, 2, 4, 0, 0, 2, 1, 1)

# empirical estimate of entropy
entropy.empirical(y)

# contigency table with counts for two discrete variables
y = rbind( c(1,2,3), c(6,5,4) )

# empirical estimate of mutual information
mi.empirical(y)
```

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entropy.MillerMadow

Miller-Madow Entropy Estimator

Description

entropy.MillerMadow estimates the Shannon entropy H of the random variable Y from the corresponding observed counts y using the Miller-Madow correction to the empirical entropy).

Usage

```
entropy.MillerMadow(y, unit=c("log", "log2", "log10"))
```

Arguments

y vector of counts.

unit

the unit in which entropy is measured. The default is "nats" (natural units). For computing entropy in "bits" set unit="log2".

Details

The Miller-Madow entropy estimator (1955) is the bias-corrected empirical entropy estimate.

Note that the Miller-Madow estimator is not a plug-in estimator, hence there are no explicit underlying bin frequencies.

Value

entropy. Miller Madow returns an estimate of the Shannon entropy.

Author(s)

Korbinian Strimmer (http://strimmerlab.org).

References

Miller, G. 1955. Note on the bias of information estimates. Info. Theory Psychol. Prob. Methods **II-B**:95-100.

See Also

```
entropy.empirical
```

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Examples

```
# load entropy library
library("entropy")

# observed counts for each bin
y = c(4, 2, 3, 0, 2, 4, 0, 0, 2, 1, 1)

# estimate entropy using Miller-Madow method
entropy.MillerMadow(y)

# compare to empirical estimate
entropy.empirical(y)
```

entropy.NSB

R Interface to NSB Entropy Estimator

Description

entropy. NSB estimates the Shannon entropy H of the random variable Y from the corresponding observed counts y using the method of Nemenman, Shafee and Bialek (2002).

Note that this function is an R interface to the "nsb-entropy" program. Hence, this needs to be installed separately from http://nsb-entropy.sourceforge.net/.

Usage

```
entropy.NSB(y, unit=c("log", "log2", "log10"), CMD="nsb-entropy")
```

Arguments

y vector of counts.

unit the unit in which entropy is measured. The default is "nats" (natural units). For

computing entropy in "bits" set unit="log2".

CMD path to the "nsb-entropy" executable.

Details

The NSB estimator is due to Nemenman, Shafee and Bialek (2002). It is a Dirichlet-multinomial entropy estimator, with a hierarchical prior over the Dirichlet pseudocount parameters.

Note that the NSB estimator is not a plug-in estimator, hence there are no explicit underlying bin frequencies.

Value

entropy. NSB returns an estimate of the Shannon entropy.

Author(s)

Jean Hausser.

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References

Nemenman, I., F. Shafee, and W. Bialek. 2002. Entropy and inference, revisited. In: Dietterich, T., S. Becker, Z. Gharamani, eds. Advances in Neural Information Processing Systems 14: 471-478. Cambridge (Massachusetts): MIT Press.

See Also

```
entropy, entropy.shrink, entropy.Dirichlet, entropy.ChaoShen.
```

Examples

```
# load entropy library
library("entropy")

# observed counts for each bin
y = c(4, 2, 3, 0, 2, 4, 0, 0, 2, 1, 1)

## Not run:
# estimate entropy using the NSB method
entropy.NSB(y) # 2.187774

## End(Not run)

# compare to empirical estimate
entropy.empirical(y)
```

entropy.plugin

Plug-In Entropy Estimator

Description

entropy.plugin computes the Shannon entropy H of a discrete random variable from the specified bin frequencies.

Usage

```
entropy.plugin(freqs, unit=c("log", "log2", "log10"))
```

Arguments

freqs bin frequencies.

unit the unit in which entropy is measured. The default is "nats" (natural units). For

computing entropy in "bits" set unit="log2".

Details

The Shannon entropy of a discrete random variable is defined as $H = -\sum p(x_i) \log(p(x_i))$, where $p(x_i)$ are the bin probabilities.

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Value

entropy.plugin returns the Shannon entropy.

Author(s)

Korbinian Strimmer (http://strimmerlab.org).

See Also

```
entropy, entropy.empirical, mi.shrink.
```

Examples

```
# load entropy library
library("entropy")

# some frequencies
freqs = c(0.2, 0.1, 0.15, 0.05, 0, 0.3, 0.2)

# and corresponding entropy
entropy.plugin(freqs)
```

entropy.shrink

Shrinkage Entropy and Mutual Information Estimator

Description

entropy.shrink estimates the Shannon entropy H of the random variable Y from the corresponding observed counts y by plug-in of shrinkage estimate of the bin frequencies.

mi.shrink estimates the corresponding mutual information of two random variables.

freq. shrink estimates the bin frequencies from the counts y using a James-Stein-type shrinkage estimator. The default shrinkage target is the uniform, unless otherwise specified.

Usage

```
entropy.shrink(y, unit=c("log", "log2", "log10"), target=1/length(y), verbose=TRUE)
mi.shrink(y, unit=c("log", "log2", "log10"), target=1/length(y), verbose=TRUE)
freqs.shrink(y, target=1/length(y), verbose=TRUE)
```

Arguments

y vector or matrix of counts.

unit the unit in which entropy is measured. The default is "nats" (natural units). For

computing entropy in "bits" set unit="log2".

target the shrinkage target for the frequencies (default: uniform distribution).

verbose report shrinkage intensity.

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Details

The shrinkage estimator is a James-Stein-type estimator. It is essentially a entropy.Dirichlet estimator, where the pseudocount is estimated from the data.

For details see Hausser and Strimmer (2009).

Value

entropy. shrink returns an estimate of the Shannon entropy.

mi.shrink returns an estimate of mutual information.

freqs.shrink returns the underlying frequencies.

In all instances the estimated shrinkage intensity is attached to the returned value in the attribute lambda.freqs.

Author(s)

Korbinian Strimmer (http://strimmerlab.org).

References

Hausser, J., and K. Strimmer. 2009. Entropy inference and the James-Stein estimator, with application to nonlinear gene association networks. J. Mach. Learn. Res. **10**: 1469-1484. Available online from http://jmlr.csail.mit.edu/papers/v10/hausser09a.html.

See Also

entropy, entropy.Dirichlet, entropy.NSB, entropy.ChaoShen, entropy.plugin, mi.plugin.

Examples

```
# load entropy library
library("entropy")

# observed counts for each bin
y = c(4, 2, 3, 0, 2, 4, 0, 0, 2, 1, 1)

# shrinkage estimate
entropy.shrink(y)

# contigency table with counts for two discrete variables
y = rbind( c(1,2,3), c(6,5,4) )

# shrinkage estimate of mutual information
mi.shrink(y)
```

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mi.plugin

Plug-In Mutual Information Estimator

Description

mi.plugin computes the mutual information of of two discrete random variables from the specified bin frequencies.

Usage

```
mi.plugin(freqs2d, unit=c("log", "log2", "log10"))
```

Arguments

freqs2d matrix of bin frequencies.

unit the unit in which entropy is measured. The default is "nats" (natural units). For

computing entropy in "bits" set unit="log2".

Details

The mutual information of two discrete random variables X and Y is defined as MI = H(X) + H(Y) - H(X,Y).

Value

mi.plugin returns the mutual information.

Author(s)

Korbinian Strimmer (http://strimmerlab.org).

See Also

```
entropy.plugin.
```

Examples

```
# load entropy library
library("entropy")

# joint distribution of two discrete variables
freqs2d = rbind( c(0.2, 0.1, 0.15), c(0, 0.3, 0.25) )

# and corresponding mutual information
mi.plugin(freqs2d)
```

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