Package ‘DirichletMultinomial’

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Type Package

Title Dirichlet-Multinomial Mixture Model Machine Learning for Microbiome Data

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Author Martin Morgan <martin.morgan@roswellpark.org>

Maintainer Martin Morgan <martin.morgan@roswellpark.org>

Description Dirichlet-multinomial mixture models can be used to describe variability in microbial metagenomic data. This package is an interface to code originally made available by Holmes, Harris, and Quince, 2012, PLoS ONE 7(2): 1-15, as discussed further in the man page for this package, ?DirichletMultinomial.

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Depends S4Vectors, IRanges

Imports stats4, methods, BiocGenerics

Suggests lattice, parallel, MASS, RColorBrewer, xtable


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Table of contents:

DirichletMultinomial-package ........................................... 2
cvdmggroup ............................................................... 3
data .............................................................................. 4
dmn ................................................................. 5
DMN-class ................................................................. 6
dmngroup ................................................................. 7
DMNGroup-class ......................................................... 8
heatmapdmn ............................................................... 9
model components ........................................................... 10
roc ............................................................................. 11
Utilities ................................................................. 13

Index 15

DirichletMultinomial-package

Dirichlet-Multinomial Mixture Model Machine Learning for Microbiome Data

Description

Dirichlet-multinomial mixture models can be used to describe variability in microbial metagenomic data. This package is an interface to code originally made available by Holmes, Harris, and Quince, 2012, PLoS ONE 7(2): 1-15.

Details

The estimation routine is from the LGPL-licensed (as stated on the corresponding googlecode page) source http://microbedmm.googlecode.com/files/MicrobeDMMv1.0.tar.gz, retrieved 17 February 2012.


Author(s)

Maintainer: Martin Morgan mailto:mtmorgan@fhcrc.org
cvdmngroup

Cross-validation on Dirichlet-Multinomial classifiers.

Description

Run cross-validation on Dirichlet-Multinomial generative classifiers.

Usage

```r
cvdmngroup(ncv, count, k, z, ..., verbose = FALSE, .lapply = parallel::mclapply)
```

Arguments

- `ncv` integer(1) number of cross-validation groups, between 2 and `nrow(count)`.
- `count` matrix of sample x taxon counts, subsets of which are used for training and cross-validation.
- `k` named integer() vector of groups and number of Dirichlet components; e.g., `c(Lean=1, Obese=3)` performs cross-validation for models with `k=1` Dirichlet components for the ‘Lean’ group, `k=3` Dirichlet components for ‘Obese’.
- `z` True group assignment.
- `...` Additional arguments, passed to `dmn` during each cross-validation.
- `verbose` logical(1) indicating whether progress should be reported
- `.lapply` A function used to perform the outer cross-validation loop, e.g., `lapply` for calculation on a single processor, `parallel::mclapply` for parallel evaluation.

Value

A `data.frame` summarizing classifications of test samples in cross-validation groups. Columns are:

- `group` The cross-validation group in which the individual was used for testing.
- `additional columns` Named after classification groups, giving the posterior probability of assignment.

Author(s)

Martin Morgan `mailto:mtmorgan@fhcrc.org`

See Also

`dmn`, `DirichletMultinomial-package`, `vignette("DirichletMultinomial")`
Examples

```r
data(xval)  ## result of following commands
head(xval)

## Not run:
## count matrix
fl <- system.file(package="DirichletMultinomial", "extdata", "Twins.csv")
count <- t(as.matrix(read.csv(fl, row.names=1)))

## phenotype
fl <- system.file(package="DirichletMultinomial", "extdata", "TwinStudy.t")
pheno0 <- scan(fl)
lvls <- c("Lean", "Obese", "Overwt")
pheno <- factor(lvls[pheno0 + 1], levels=lvls)
names(pheno) <- rownames(count)

## subset
keep <- c("Lean", "Obese")
count <- count[pheno]
pheno <- factor(pheno[pheno]

## cross-validation, single Dirichlet component for Lean, 3 for Obese
xval <- cvdmngroup(nrow(count), count, c(Lean=1, Obese=3), pheno,
                   verbose=TRUE, mc.preschedule=FALSE)

## End(Not run)
```

---

**data**  

Data objects used for examples and the vignette

Description

These data objects correspond to steps in a typical work flow, as described in the vignette to this package. fit corresponds to dmn fits to different values of k. bestgroup is the result of the two-group generative classifier. xval summarizes leave-one-out cross validation of the classifier.

Usage

```r
data(fit)
data(bestgrp)
data(xval)
```

Format

- fit is a list of seven `DMN` objects.
- bestgrp is a `DMNGroup` object.
- xval is a data.frame with columns corresponding to the cross-validation group membership and the Lean and Obese posterior probabilities.
dmn

Examples

```r
data(fit); fit[1:2]
plot(sapply(fit, laplace), type="b")
data(bestgrp); bestgrp
data(xval); head(xval, 3)
```

---

**dmn**  
*Fit Dirichlet-Multinomial models to count data.*

---

**Description**

Fit Dirichlet-Multinomial models to a sample x taxon count matrix.

**Usage**

```r
dmn(count, k, verbose = FALSE, seed = runif(1, 0, .Machine$integer.max))
```

**Arguments**

- `count`  
  matrix() of sample x taxon counts.
- `k`  
  integer(1), the number of Dirichlet components to fit.
- `verbose`  
  logical(1) indicating whether progress in fit should be reported.
- `seed`  
  numeric(1) random number seed.

**Details**

This implements Dirichlet-multinomial mixture models describe in the package help page, DirichletMultinomial-package.

**Value**

An object of class dmn, with elements (elements are usually retrieved via functions defined in the package, not directly).

- **GoodnessOfFit**  
  NLE, LogDet, Laplace, AIC, and BIC criteria assessing goodness-of-fit.
- **Group**  
  matrix of dimension samples x k, providing the Dirichlet parameter vectors.
- **Mixture**  
  **Weight** numeric() of length k, with relative weight of each component.
- **Fit**  
  **Lower** matrix() of dimension taxa x k with 95% lower bounds on Dirichlet component vector estimates.
  **Estimate** matrix() of dimension taxa x k with Dirichlet component vector estimates.
  **Upper** matrix() of dimension taxa x k with 95% upper bounds on Dirichlet component vector estimates.

**Author(s)**

Martin Morgan  
mailto:mtmorgan@fhcrc.org
References


See Also

DirichletMultinomial-package, vignette("DirichletMultinomial")

Examples

data(fit)
## k = 1:7; full example in vignette
lplc <- sapply(fit, laplace)
plot(lplc, type="b")
fit[[which.min(lplc)]]

DMN-class

Class "DMN"

Description

Result from fitting a Dirichlet-Multinomial model.

Objects from the Class

Objects can be created by calls to dmn.

Slots

The contents of a slot is usually retrieved via the methods described on the mixture help page.

goodnessOfFit  NLE, LogDet, Laplace, AIC, and BIC criteria assessing goodness-of-fit.
group matrix of dimension samples x k, providing the Dirichlet parameter vectors.
mixture Weight numeric() of length k, with relative weight of each component.
fit Lower matrix() of dimension taxa x k with 95% lower bounds on Dirichlet component vector estimates.
    Estimate matrix() of dimension taxa x k with Dirichlet component vector estimates.
    Upper matrix() of dimension taxa x k with 95% upper bounds on Dirichlet component vector estimates.

Methods

See the mixture help page.

Author(s)

Martin Morgan mailto:mtmorgan@fhcrc.org
\texttt{dmngroup}

See Also
\texttt{dmn, mixture}.

Examples

\begin{verbatim}
data(fit)
fit[[4]]
\end{verbatim}

\texttt{dmngroup} \hspace{1cm} \textit{Dirichlet-Multinomial generative classifiers.}

Description

Fit Dirichlet-Multinomial generative classifiers to groups (rows) within a sample x taxon count matrix.

Usage

\begin{verbatim}
dmngroup(count, group, k, ..., simplify = TRUE, .lapply = parallel::mclapply)
\end{verbatim}

Arguments

- \texttt{count} \hspace{1cm} \texttt{matrix()} of sample x taxon counts.
- \texttt{group} \hspace{1cm} \texttt{factor()} or vector to be coerced to a factor, with as many elements as there are rows in \texttt{count}, indicating the group to which the corresponding sample belongs.
- \texttt{k} \hspace{1cm} \texttt{integer()}, the number(s) of Dirichlet components to fit.
- \texttt{...} \hspace{1cm} Additional arguments, passed to \texttt{dmn}.
- \texttt{simplify} \hspace{1cm} Return only the best-fit model for each group?
- \texttt{lapply} \hspace{1cm} An lapply-like function for application of group x k fits.

Details

This function divided \texttt{count} into groups defined by \texttt{group}, creates all combinations of \texttt{group} x \texttt{k}, and evaluates each using \texttt{dmn}. When \texttt{simplify=TRUE}, the best (Laplace) fit is selected for each group.

Value

An object of class \texttt{dmngroup}, a list of fitted models of class \texttt{dmn}. When \texttt{simplify=TRUE}, elements are named by the group to which they correspond.

Author(s)

Martin Morgan \texttt{mailto:mtmorgan@fhcrc.org}
References

See Also
dmn, DirichletMultinomial-package, vignette("DirichletMultinomial")

Examples
## best fit for groups 'Lean' and 'Obese'; full example in vignette.
## Not run: bestgrp <- dmngroup(count, pheno, k=1:5, verbose=TRUE, mc.preschedule=FALSE)

## End(Not run)
data(bestgrp)
bestgrp
bestgrp[['Obese']]
**heatmapdmn**

See Also

- `mixture`, `DMN`, `SimpleList`.

Examples

```r
data(bestgrp)
bestgrp
bestgrp[[1]]
```

**heatmapdmn**  
Heatmap representation of samples assigned to Dirichlet components.

**Description**

Produce a heat map summarizing count data, grouped by Dirichlet component.

**Usage**

```r
heatmapdmn(count, fit1, fitN, ntaxa = 30, ..., 
transform = sqrt, lblwidth = 0.2 * nrow(count), col = .gradient)
```

**Arguments**

- `count`: A matrix of sample x taxon counts, as supplied to `dmn`.
- `fit1`: An instance of class `dmn`, from a model fit to a single Dirichlet component, \( k=1 \) in `dmn`.
- `fitN`: An instance of class `dmn`, from a model fit to \( N \neq 1 \) components, \( k=N \) in `dmn`.
- `ntaxa`: The \( ntaxa \) most numerous taxa to display counts for.
- `...`: Additional arguments, ignored.
- `transform`: Transformation to apply to count data prior to visualization; this does *not* influence mixture membership or taxonomic ordering.
- `lblwidth`: The proportion of the plot to dedicate to taxonomic labels, as a fraction of the number of samples to be plotted.
- `col`: The colors used to display (possibly transformed, by `transform`) count data, as used by `image`.

**Details**

Columns of the heat map correspond to samples. Samples are grouped by Dirichlet component, with average (Dirichlet) components summarized as a separate wide column. Rows correspond to taxonomic groups, ordered based on contribution to Dirichlet components.

**Author(s)**

Martin Morgan  
mailto:mtmorgan@fhcrc.org
Examples

```r
## counts
fl <- system.file(package="DirichletMultinomial", "extdata",
    "Twins.csv")
count <- t(as.matrix(read.csv(fl, row.names=1)))

## all and best-fit clustering
data(fit)
lplc <- sapply(fit, laplace)
best <- fit[[which.min(lplc)]]
heatmapdmn(count, fit[[1]], best, 30)
```

Description

The accessors `mixture` and `mixturewt` return information about the estimated Dirichlet components of the fitted model. Return values are described in the Values section, below.

Usage

```r
mixture(object, ..., assign=FALSE)
mixturewt(object, ...)
goodnessOfFit(object, ...)
laplace(object, ...)  # S4 method for signature 'DMN'
AIC(object, ..., k = 2)  # S4 method for signature 'DMN'
BIC(object, ...)  # S4 method for signature 'DMN'

fitted(object, ..., scale=FALSE)  # S4 method for signature 'DMN'
predict(object, newdata, ..., logevidence=FALSE)  # S4 method for signature 'DMNGroup'
fitted(object, ...)  # S4 method for signature 'DMNGroup'
predict(object, newdata, ..., assign=FALSE)  # S4 method for signature 'DMNGroup'
summarize(object, ...)  # S4 method for signature 'DMNGroup'
```

Arguments

- `object` An instance of class `dmn`.
- `newdata` A matrix of new sample x taxon data to be fitted to the model of `object`.
Additional arguments, available to methods, when applicable.

assign logical(1) indicating whether the maximum per-sample mixture component should be returned (assign=FALSE), or the full mixture matrix (assign=TRUE).

scale logical(1) indicating whether fitted values should be returned unscaled (default, scaled=FALSE) or scaled by the variability of mixturewt parameter theta.

logevidence logical(1) indicating whether posterior probability (default, logevidence=FALSE) or log evidence logical=TRUE should be returned.

k ignored.

Value

mixture with assign=FALSE returns a matrix of sample x Dirichlet component estimates. With assign=TRUE mixture returns a named vector indexing the maximal Dirichlet component of each sample.

mixturewt returns a matrix with rows corresponding to mixture components, and columns pi (component weight) and theta (component variability). Small values of theta correspond to highly variable components.

goodnessOfFit returns a named numeric vector of measures of goodness of fit.

laplace, AIC, and BIC return the corresponding measures of goodness of fit.

Author(s)

Martin Morgan mailto:mtmorgan@fhcrc.org

Examples

data(fit)
best <- fit[[4]]
mixturewt(best)
head(mixture(best), 3)
head(mixture(best, assign=TRUE), 3)
goodnessOfFit(best)

fl <- system.file(package="DirichletMultinomial", "extdata", "Twins.csv")
count <- t(as.matrix(read.csv(fl, row.names=1)))
data(bestgrp)
bestgrp
head(predict(bestgrp, count))

---

roc Summarize receiver-operator characteristics

Description

Returns a data.frame summarizing the cumulative true- and false-positive probabilities from expected and observed classifications.
Usage

roc(exp, obs, ...)

Arguments

exp logical() vector of expected classifications to a particular group.
obs Predicted probability of assignment to the group identified by TRUE values in
exp. The length of exp and obs must be identical.
...
Additional arguments, available to methods.

Value

A data.frame with columns

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>TruePositive</td>
<td>Cumulative probability of correct assignment.</td>
</tr>
<tr>
<td>FalsePositive</td>
<td>Cumulative probability of incorrect assignment.</td>
</tr>
</tbody>
</table>

Author(s)

Martin Morgan mailto:mtmorgan@fhcrc.org

Examples

library(lattice)

## count matrix
fl <- system.file(package="DirichletMultinomial", "extdata",
"Twins.csv")
count <- t(as.matrix(read.csv(fl, row.names=1)))

## phenotype
fl <- system.file(package="DirichletMultinomial", "extdata",
"TwinStudy.t")
pheno0 <- scan(fl)
lvls <- c("Lean", "Obese", "Overwt")
pheno <- factor(lvls[pheno0 + 1], levels=lvls)
names(pheno) <- rownames(count)

## count data used for cross-validation, and cross-validation
count <- csubset(c("Lean", "Obese"), count, pheno)
data(bestgrp)

## true, false positives from single-group classifier
bst <- roc(pheno[rownames(count)] == "Obese",
predict(bestgrp, count[, "Obese")
head(bst)

## lattice plot
xyplot(TruePositive ~ FalsePositive, bst, type="l",
  xlab="False Positive", ylab="True Positive")
Utilities

Helpful utility functions

Description

csubset creates a subset of a count matrix, based on identity of column phenotypes to a specified value.

Usage

csubset(val, x, pheno, cidx = TRUE)

Arguments

val character(1) specifying the subset of phenotype to select.

x A matrix of counts, with rows corresponding to samples and columns to taxonomic groups.

pheno A character() vector of length equal to the number of rows in count, indicating the phenotype of the corresponding sample.

cidx A logical(1) indicating whether columns (taxa) with zero counts in the count matrix following removal of taxa not satisfying pheno %in% val should be removed. cidx=FALSE removes the 0-count columns.

Value

A matrix of counts, with rows satisfying pheno %in% val and with columns equal either to ncol(x) (when cidx=TRUE) or the number of columns with non-zero counts after row subsetting (cidx=FALSE).

Author(s)

Martin Morgan mailto:mtmorgan@fhcrc.org

Examples

## count matrix
fl <- system.file(package="DirichletMultinomial", "extdata", "Twins.csv")
count <- t(as.matrix(read.csv(fl, row.names=1)))

## phenotype
fl <- system.file(package="DirichletMultinomial", "extdata", "TwinStudy.t")
pheno0 <- scan(fl)
lvls <- c("Lean", "Obese", "Overwt")
pheno <- factor(lvls[pheno0 + 1], levels=lvls)
names(pheno) <- rownames(count)

## subset
dim(count)
sum("Lean" == pheno)
dim(csubset("Lean", count, pheno))
dim(csubset("Lean", count, pheno, cidx=FALSE))
Index

* classes
  DMN-class, 6
  DMNGroup-class, 8
* datasets
  data, 4
* manip
  dmn, 5
  dmngroup, 7
  heatmapdmn, 9
  model components, 10
  Utilities, 13
* package
  DirichletMultinomial-package, 2
* stats
  cvdmngroup, 3
  roc, 11
  AIC, DMN-method (model components), 10
  Annotated, 8
  bestgrp (data), 4
  BIC, DMN-method (model components), 10
  csubset (Utilities), 13
  cvdmngroup, 3
  data, 4
  DirichletMultinomial-package, 2, 3, 5, 6, 8
  DMN, 4, 9
  dmn, 3, 5, 6–10
  DMN-class, 6
  DMNGroup, 4
  dmngroup, 7, 8
  DMNGroup-class, 8
  goodnessOfFit (model components), 10
  heatmapdmn, 9
  image, 9
  laplace (model components), 10
  List, 8
  mixture, 6–9
  mixture (model components), 10
  mixturewt, 11
  mixturewt (model components), 10
  model components, 10
  predict, DMN-method (model components), 10
  predict, DMNGroup-method (model components), 10
  roc, 11
  show, DMN-method (model components), 10
  show, DMNGroup-method (model components), 10
  SimpleList, 8, 9
  summary, DMNGroup-method (model components), 10
  Utilities, 13
  Vector, 8
  xval (data), 4