# Package 'ruv' 

August 30, 2019
Title Detect and Remove Unwanted Variation using Negative Controls
Description Implements the 'RUV' (Remove Unwanted Variation) algorithms. These algorithms attempt to adjust for systematic errors of unknown origin in high-dimensional data. The algorithms were originally developed for use with genomic data, especially microarray data, but may be useful with other types of high-dimensional data as well. These algorithms were proposed in Gagnon-Bartsch and Speed (2012) [doi:10.1093/nar/gkz433](doi:10.1093/nar/gkz433), GagnonBartsch, Jacob and Speed (2013), and Molania, et. al. (2019) [doi:10.1093/nar/gkz433](doi:10.1093/nar/gkz433). The algorithms require the user to specify a set of negative control variables, as described in the references. The algorithms included in this package are 'RUV-2', 'RUV-4', 'RUV-inv', 'RUVrinv', 'RUV-I', and RUV-III', along with various supporting algorithms.
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Author Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)
Maintainer Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)
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ruv-package Detect and Remove Unwanted Variation using Negative Controls

## Description

Implements the 'RUV' (Remove Unwanted Variation) algorithms. These algorithms attempt to adjust for systematic errors of unknown origin in high-dimensional data. The algorithms were originally developed for use with genomic data, especially microarray data, but may be useful with other types of high-dimensional data as well. These algorithms were proposed in GagnonBartsch and Speed (2012) [doi:10.1093/nar/gkz433](doi:10.1093/nar/gkz433), Gagnon-Bartsch, Jacob and Speed (2013), and Molania, et. al. (2019) [doi:10.1093/nar/gkz433](doi:10.1093/nar/gkz433). The algorithms require the user to specify a set of negative control variables, as described in the references. The algorithms included in this package are 'RUV-2', 'RUV-4', 'RUV-inv', 'RUV-rinv', 'RUV-I', and RUV-III', along with various supporting algorithms.

## Details

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| :--- | :--- |
| Type: | Package |
| Version: | 0.9 .7 .1 |
| Date: | $2019-08-30$ |
| License: | GPL |
| LazyLoad: | yes |
| URL: | http://www-personal.umich.edu/~johanngb/ruv/ |

## Note

Additional resources can be found at http://www-personal.umich.edu/~johanngb/ruv/.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Gagnon-Bartsch, J.A. and T.P. Speed (2012). Using control genes to correct for unwanted variation in microarray data. Biostatistics. [doi:10.1093/biostatistics/kxr034](doi:10.1093/biostatistics/kxr034)
Gagnon-Bartsch, J.A., L. Jacob, and T.P. Speed (2013). Removing Unwanted Variation from High Dimensional Data with Negative Controls. Technical report. Available at: http://statistics.berkeley.edu/techreports/820

Molania, R., J. A. Gagnon-Bartsch, A. Dobrovic, and T. P. Speed (2019). A new normalization for the Nanostring nCounter gene expression assay. Nucleic Acids Research. [doi:10.1093/nar/gkz433](doi:10.1093/nar/gkz433)

## See Also

RUV2, RUV4, RUVinv, RUVrinv, variance_adjust, RUVI, RUVIII

```
collapse.replicates Collapse Replicates
```


## Description

This function is intended for use in conjunction with RUVIII, specifically when using the average=TRUE option. When using the average=TRUE option, the adjusted data matrix has only one row for each set of replicates. In other words, each set of replicate rows in the orginal data matrix is "collapsed" into a single row in the adjusted data matrix. This function similarly collapses the rows of a dataframe of covariates. Only covariates that are constant within each set of replicates are retained.

## Usage

collapse.replicates(df, M)

## Arguments

df
M

A dataframe.
The replicate structure. See RUVIII for details.

## Value

A sub-dataframe of df.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

See Also
RUVIII
design.matrix Design Matrix

## Description

Creates a design matrix.

## Usage

design.matrix(a, name = "X", remove.collinear = TRUE, include.intercept $=$ TRUE)

## Arguments

a Object from which to create a design matrix. Can be a vector, matrix, factor, or dataframe.
name $\quad$ Name of the design matrix. Used to name columns that aren't already named (e.g. X1, X2, etc.)
remove.collinear
Will remove columns that are collinear, to ensure the design matrix is full rank.
include.intercept
Add an intercept to the matrix if one is not included already.

## Details

Numerical vectors are not modified. Factors are converted to dummy variables. Character vectors are converted to factors, and then to dummy variables.

## Value

A matrix.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

```
getK Get K
```


## Description

Finds an often-suitable value of K for use in RUV-4.

## Usage

$\operatorname{get} \mathrm{K}(\mathrm{Y}, \mathrm{X}, \mathrm{ctl}, \mathrm{Z}=1$, eta $=$ NULL, include.intercept $=$ TRUE, fullwo $=$ NULL, cutoff $=$ NULL, method="select", l=1, inputcheck = TRUE)

## Arguments

Y
$X \quad$ The factor(s) of interest. A $m$ by $p$ matrix, where $m$ is the number of samples and $p$ is the number of factors of interest. Note that $X$ should be only a single column, i.e. $p=1$; if X has more than one column, only column 1 will be used (see below).
ctl An index vector to specify the negative controls. Either a logical vector of length $n$ or a vector of integers.

Z Any additional covariates to include in the model, typically a m by q matrix. Factors and dataframes are also permissible, and converted to a matrix by design.matrix. Alternatively, may simply be 1 (the default) for an intercept term. May also be NULL.
eta Gene-wise (as opposed to sample-wise) covariates. These covariates are adjusted for by RUV-1 before any further analysis proceeds. Can be either (1) a matrix with n columns, (2) a matrix with n rows, (3) a dataframe with n rows, (4) a vector or factor of length $n$, or (5) simply 1 , for an intercept term.
include.intercept
Applies to both Z and eta. When Z or eta (or both) is specified (not NULL) but does not already include an intercept term, this will automatically include one. If only one of $Z$ or eta should include an intercept, this variable should be set to FALSE, and the intercept term should be included manually where desired.
fullwo Can be included to speed up execution. Is returned by previous calls of getK, RUV4, RUVinv, or RUVrinv (see below).
cutoff Specify an alternative cut-off. Default is the (approximate) 90th percentile of the distribution of the first singular value of an $m$ by $n$ gaussian matrix.

$$
\begin{array}{ll}
\text { method } & \begin{array}{l}
\text { Can be set to either leave1out, fast, or select. leave1out is the preferred } \\
\text { method but may be slow, fast is an approximate method that is faster but may } \\
\text { provide poor results if n_c is not much larger than m, and select (the default) } \\
\text { tries to choose for you. }
\end{array} \\
1 & \begin{array}{l}
\text { Which column of } X \text { to use in the getK algorithm. }
\end{array} \\
\text { inputcheck } \quad \begin{array}{l}
\text { Perform a basic sanity check on the inputs, and issue a warning if there is a } \\
\text { problem. }
\end{array}
\end{array}
$$

## Value

A list containing
$k \quad$ the estimated value of $k$
cutoff The cutoff value used
sizeratios A measure of the relative sizes of the rows of alpha.
fullwo Can be used to speed up future calls of RUV4.

## Warning

This value of K will not be the best choice in all cases. Moreover, it will often be a poor choice of K for use with RUV2. See Gagnon-Bartsch and Speed (2012) for commentary on estimating k.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.
Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

RUV4

## Examples

```
## Create some simulated data
m = 50
n = 10000
nc=1000
p = 1
k = 20
ctl = rep(FALSE, n)
ctl[1:nc] = TRUE
X = matrix(c(rep(0,floor(m/2)), rep(1,ceiling(m/2))), m, p)
beta = matrix(rnorm(p*n), p, n)
```

```
beta[,ctl] = 0
W = matrix (rnorm(m*k),m,k)
alpha \(=\operatorname{matrix}(r n o r m(k * n), k, n)\)
epsilon \(=\) matrix \((r n o r m(m * n), m, n)\)
Y = X\%*\%beta + W\%*\%alpha + epsilon
\#\# Run getK
temp \(=\operatorname{getK}(\mathrm{Y}, \mathrm{X}, \mathrm{ctl})\)
K = temp\$k
```

get_empirical_variances
Get empirical variances

## Description

This method implements the method of empirical variances as described in Gagnon-Bartsch, Jacob, and Speed (2013). This function is normally called from the function variance_adjust, and is not normally intended for stand-alone use.

## Usage

get_empirical_variances(sigma2, betahat, bin $=10$, rescaleconst $=$ NULL)

## Arguments

sigma2 Estimates of sigma^2
betahat Estimates of beta
bin The bin size
rescaleconst The expected value of the average of the smallest bin -1 of bin independent chi-square random variables. This can be specified to save computational time (otherwise, it is calculated by simulation).

## Value

A vector of the empirical variances.

## Author(s)

Johann Gagnon-Bartsch

## References

Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

variance_adjust
google_search Google Search URL

## Description

Converts a string to URL for a goolge search of that string.

## Usage

google_search(a)

## Arguments

a
A string.

## Value

A string that is a URL.
inputcheck1 Input Check One

## Description

Performs a basic sanity check on the arguments passed to RUV-2, RUV-4, RUV-inv, and RUV-rinv.

## Usage

inputcheck1(Y, X, Z, ctl, check.na=TRUE)

## Arguments

Y
$X \quad$ The factor(s) of interest. A $m$ by $p$ matrix, where $m$ is the number of samples and p is the number of factors of interest. Very often $\mathrm{p}=1$.
Z
ctl The negative controls. A logical vector of length $n$.
check.na
The data. A $m$ by $n$ matrix, where $m$ is the number of samples and $n$ is the number of features.

Z Any additional covariates to include in the model. Either a m by q matrix of covariates, or simply 1 (the default) for an intercept term.

Whether to check for missing values.

品

## Value

Returns NULL. The function is only called to check for problems in the arguments and to issue warnings if any problems are found.

## See Also

> RUV2, RUV4, RUVinv, RUVrinv
invvar Inverse Method Variances

## Description

Estimate the features' variances using the inverse method. This function is usually called from RUVinv and not normally intended for stand-alone use.

## Usage

invvar(Y, ctl, XZ = NULL, eta = NULL, lambda = NULL,
invsvd = NULL)

## Arguments

$\mathrm{Y} \quad$ The data. A $m$ by $n$ matrix, where $m$ is the number of samples and $n$ is the number of features.
ctl The negative controls. A logical vector of length $n$.
XZ A m by $(p+q)$ matrix containing both the factor $(\mathrm{s})$ of interest $(\mathrm{X})$ and known covariates (Z).
eta Gene-wise (as opposed to sample-wise) covariates. These covariates are adjusted for by RUV-1 before any further analysis proceeds. A matrix with $n$ columns.
lambda Ridge parameter. If specified, the ridged inverse method will be used.
invsvd Can be included to speed up execution. Generally used when calling invvar many times with different values of lambda.

## Value

A list containing
sigma2 Estimates of the features' variances. A vector of length n .
df The "effective degrees of freedom"
invsvd Can be used to speed up future calls of invvar.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

See Also
RUVinv, RUVrinv

```
projectionplotvariables
```

Projection Plot Variables

## Description

Calculates the variables necessary to produce a projection plot.

## Usage

projectionplotvariables(Y, X, W)

## Arguments

$\mathrm{Y} \quad$ The data. A m by n matrix, where m is the number of samples and n is the number of features.
$\mathrm{X} \quad$ A m by p matrix containing the factor(s) of interest.
$W \quad$ A m by k matrix containing the estimated unwanted factors.

## Details

Typically intended for internal use, and called after adjustment for known covariates ( Z ).

## Value

A list containing

| byx | Regression coefficients from regressing Y on X. |
| :--- | :--- |
| bwx | Regression coefficients from regressing W on X. |
| projectionplotalpha |  |
| A reparameterization of alpha. |  |
| projectionplotW | A reparameterization of W. |

randinvvar
(Randomization) Inverse Method Variances

## Description

Estimate the features' variances using a stochastic version of the inverse method. This function is usually called from RUVinv and not normally intended for stand-alone use.

## Usage

randinvvar Y , ctl, XZ = NULL, eta $=$ NULL, lambda $=$ NULL, iterN $=1 \mathrm{e}+05$ )

## Arguments

Y
ctl The negative controls. A logical vector of length $n$.
XZ A m by $(\mathrm{p}+\mathrm{q})$ matrix containing both the factor(s) of interest $(\mathrm{X})$ and known covariates (Z).
eta Gene-wise (as opposed to sample-wise) covariates. These covariates are adjusted for by RUV-1 before any further analysis proceeds. A matrix with n columns.
lambda Ridge parameter. If specified, the ridged inverse method will be used.
iterN The number of random "factors of interest" to generate.

## Value

A list containing
sigma2 Estimates of the features' variances. A vector of length n .
df The "effective degrees of freedom"

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

RUVinv, RUVrinv, invvar

```
replicate.matrix
```


## Replicate (Mapping) Matrix

## Description

For use with RUVIII, generates a mapping matrix that describes the replicate structure.

## Usage

replicate.matrix(a, burst=NULL, return.factor=FALSE, name="M", sep="_", burstsep = "_")

## Arguments

a
burst A character vector, containing the names of factor levels to be "burst." When a factor level is burst, the corresponding observations are no longer replicates; they become singletons.
return.factor Return a factor instead of the mapping matrix. This may be useful in two situations: (1) When the input is a mapping matrix, and it is desired to convert it back to a factor; (2) When making repeated calls to replicate.matrix in order to define the replicates in several steps. Example of (2): Suppose there are 4 patients and 3 sample dates. We wish to designate as replicates observations that have the same patient ID and sample date, but only for the first two sample dates; none of the observations in the third sample date should be considered replicates. We would first call replicate.matrix using only the sampleDate factor, bursting the third level, and returning another factor. We would then call replicate.matrix again, this time with a dataframe containing patientID and the bursted sampleDate. See below for example code.
name $\quad$ Name of the mapping matrix. Used to name columns that aren't already named (e.g. M1, M2, etc.)
sep Text separating the level names of different factors when they are combined.
burstsep Text appended to factor level names when bursting a factor. This text is then followed by a number. Example: if the factor level to be burst is "June29", and burstsep is the default value of "_", then the new levels will be "June29_1", "June29_2", etc.

## Value

A matrix or a factor, depending on the value of return.factor.

## Warning

Be sure to change the default values of sep and burstsep if there is any risk of factor level naming conflicts (e.g. if existing factors already have level names like "patient_1", "patient_2", etc.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## See Also

RUVIII

## Examples

```
# Define patientID and sampleDate
patientID = paste("patient", rep(1:4, each=6), sep="")
#print(patientID)
sampleDate = paste("June", rep(c(12,17, 29), 8), sep="")
#print(sampleDate)
# Create a mapping matrix, where every unique
# patientID / sampleDate combination define a set of replicates
M = replicate.matrix(data.frame(patientID, sampleDate))
#print(M)
# Convert M back to a factor
M = replicate.matrix(M, return.factor=TRUE)
#print(M)
# Create a factor for sampleDate, but burst the third date
temp = replicate.matrix(sampleDate, burst="June29", return.factor=TRUE)
#print(temp)
# Create a mapping matrix as described above in the description of return.factor
M = replicate.matrix(data.frame(temp, patientID))
#print(M)
```

```
residop Residual Operator
```


## Description

Applies the residual operator of a matrix B to a matrix A.

## Usage

residop(A, B)

## Arguments

A A matrix with $m$ rows.
B Another matrix with $m$ rows.

## Details

The columns of B must be linearly independent.

## Value

The matrix A after projection into the orthogonal complement of the column space of B.
RUV2 Remove Unwanted Variation, 2-step

## Description

The RUV-2 algorithm. Estimates and adjusts for unwanted variation using negative controls.

## Usage

RUV2(Y, X, ctl, k, Z=1, eta=NULL, include.intercept=TRUE, fullW=NULL, svdyc=NULL, do_projectionplot=TRUE, inputcheck=TRUE)

## Arguments

$Y \quad$ The data. A $m$ by $n$ matrix, where $m$ is the number of samples and $n$ is the number of features.

X
The factor(s) of interest. A m by p matrix, where m is the number of samples and $p$ is the number of factors of interest. Very often $p=1$. Factors and dataframes are also permissible, and converted to a matrix by design.matrix.
ctl An index vector to specify the negative controls. Either a logical vector of length n or a vector of integers.
$k \quad$ The number of unwanted factors to use. Can be 0 .
Z Any additional covariates to include in the model, typically a m by q matrix. Factors and dataframes are also permissible, and converted to a matrix by design.matrix. Alternatively, may simply be 1 (the default) for an intercept term. May also be NULL.
eta Gene-wise (as opposed to sample-wise) covariates. These covariates are adjusted for by RUV-1 before any further analysis proceeds. Can be either (1) a matrix with $n$ columns, (2) a matrix with $n$ rows, (3) a dataframe with n rows, (4) a vector or factor of length $n$, or (5) simply 1 , for an intercept term.
include.intercept
Applies to both Z and eta. When Z or eta (or both) is specified (not NULL) but does not already include an intercept term, this will automatically include one. If only one of $Z$ or eta should include an intercept, this variable should be set to FALSE, and the intercept term should be included manually where desired.
fullw Can be included to speed up execution. Is returned by previous calls of RUV2 (see below).
svdyc Can be included to speed up execution. For internal use; please use fullw instead.
do_projectionplot
Calculate the quantities necessary to generate a projection plot.
inputcheck Perform a basic sanity check on the inputs, and issue a warning if there is a problem.

## Details

Implements the RUV-2 algorithm as described in Gagnon-Bartsch and Speed (2012), using the SVD as the factor analysis routine. Unwanted factors W are estimated using control genes. Y is then regressed on the variables $\mathrm{X}, \mathrm{Z}$, and W .

## Value

A list containing

| betahat | The estimated coefficients of the factor(s) of interest. A p by n matrix. |
| :---: | :---: |
| sigma2 | Estimates of the features' variances. A vector of length n. |
| t | t statistics for the factor(s) of interest. A p by n matrix. |
| p | P -values for the factor(s) of interest. A p by n matrix. |
| Fstats | F statistics for testing all of the factors in X simultaneously. |
| Fpvals | P -values for testing all of the factors in X simultaneously. |
| multiplier | The constant by which sigma2 must be multiplied in order get an estimate of the variance of betahat |
| df | The number of residual degrees of freedom. |
| W | The estimated unwanted factors. |
| alpha | The estimated coefficients of W. |
| byx | The coefficients in a regression of Y on X (after both Y and X have been "adjusted" for Z). Useful for projection plots. |
| bwx | The coefficients in a regression of W on X (after X has been "adjusted" for Z). Useful for projection plots. |
| X | $X$. Included for reference. |
| k | $k$. Included for reference. |
| ctl | ctl . Included for reference. |
| Z | Z. Included for reference. |
| eta | eta. Included for reference. |

fullw Can be used to speed up future calls of RUV2.
projectionplotW
A reparameterization of W useful for projection plots.
projectionplotalpha
A reparameterization of alpha useful for projection plots.
include.intercept
include. intercept. Included for reference.
method Character variable with value "RUV2". Included for reference.

## Note

Additional resources can be found at http://www-personal.umich.edu/~johanngb/ruv/.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.
Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

RUV4, RUVinv, RUVrinv, variance_adjust

## Examples

```
## Create some simulated data
m = 50
n = 10000
nc = 1000
p = 1
k = 20
ctl = rep(FALSE, n)
ctl[1:nc] = TRUE
X = matrix(c(rep(0,floor(m/2)), rep(1,ceiling(m/2))), m, p)
beta = matrix(rnorm(p*n), p, n)
beta[,ctl] = 0
W = matrix(rnorm(m*k),m,k)
alpha = matrix(rnorm(k*n),k,n)
epsilon = matrix(rnorm(m*n),m,n)
Y = X%*%beta + W%*%alpha + epsilon
## Run RUV-2
fit = RUV2(Y, X, ctl, k)
## Get adjusted variances and p-values
fit = variance_adjust(fit)
```

```
RUV4 Remove Unwanted Variation, 4-step
```


## Description

The RUV-4 algorithm. Estimates and adjusts for unwanted variation using negative controls.

## Usage

RUV4(Y, X, ctl, k, Z = 1, eta = NULL, include.intercept=TRUE,
fullW0=NULL, inputcheck=TRUE)

## Arguments

$\mathrm{Y} \quad$ The data. A m by n matrix, where m is the number of samples and n is the number of features.
$X \quad$ The factor(s) of interest. A $m$ by $p$ matrix, where $m$ is the number of samples and $p$ is the number of factors of interest. Very often $p=1$. Factors and dataframes are also permissible, and converted to a matrix by design.matrix.
ctl An index vector to specify the negative controls. Either a logical vector of length $n$ or a vector of integers.
$k \quad$ The number of unwanted factors to use. Can be 0 .
Z Any additional covariates to include in the model, typically a m by q matrix. Factors and dataframes are also permissible, and converted to a matrix by design.matrix. Alternatively, may simply be 1 (the default) for an intercept term. May also be NULL.
eta Gene-wise (as opposed to sample-wise) covariates. These covariates are adjusted for by RUV-1 before any further analysis proceeds. Can be either (1) a matrix with $n$ columns, (2) a matrix with $n$ rows, (3) a dataframe with n rows, (4) a vector or factor of length $n$, or (5) simply 1 , for an intercept term.
include.intercept
Applies to both Z and eta. When Z or eta (or both) is specified (not NULL) but does not already include an intercept term, this will automatically include one. If only one of $Z$ or eta should include an intercept, this variable should be set to FALSE, and the intercept term should be included manually where desired.
fullwo Can be included to speed up execution. Is returned by previous calls of RUV4, RUVinv, or RUVrinv (see below).
inputcheck Perform a basic sanity check on the inputs, and issue a warning if there is a problem.

## Details

Implements the RUV-4 algorithm as described in Gagnon-Bartsch, Jacob, and Speed (2013), using the SVD as the factor analysis routine. Unwanted factors W are estimated using control genes. Y is then regressed on the variables $\mathrm{X}, \mathrm{Z}$, and W .

## Value

A list containing

| betahat | The estimated coefficients of the factor(s) of interest. A p by n matrix. |
| :---: | :---: |
| sigma2 | Estimates of the features' variances. A vector of length $n$. |
| t | $t$ statistics for the factor(s) of interest. A p by n matrix. |
| p | P -values for the factor(s) of interest. A p by n matrix. |
| Fstats | $F$ statistics for testing all of the factors in X simultaneously. |
| Fpvals | P -values for testing all of the factors in X simultaneously. |
| multiplier | The constant by which sigma2 must be multiplied in order get an estimate of the variance of betahat |
| df | The number of residual degrees of freedom. |
| W | The estimated unwanted factors. |
| alpha | The estimated coefficients of W. |
| byx | The coefficients in a regression of Y on X (after both Y and X have been "adjusted" for Z). Useful for projection plots. |
| bwx | The coefficients in a regression of W on X (after X has been "adjusted" for Z). Useful for projection plots. |
| X | $X$. Included for reference. |
| k | $k$. Included for reference. |
| ctl | ctl . Included for reference. |
| Z | Z. Included for reference. |
| eta | eta. Included for reference. |
| fullwo | Can be used to speed up future calls of RUV4. |
| include.intercept |  |
|  | include. intercept. Included for reference. |
| method | Character variable with value "RUV4". Included for reference. |

## Note

Additional resources can be found at http://www-personal.umich.edu/~johanngb/ruv/.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.

Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

RUV2, RUVinv, RUVrinv, variance_adjust

## Examples

```
## Create some simulated data
m = 50
n = 10000
nc = 1000
p = 1
k = 20
ctl = rep(FALSE, n)
ctl[1:nc] = TRUE
X = matrix(c(rep(0,floor(m/2)), rep(1,ceiling(m/2))), m, p)
beta = matrix(rnorm(p*n), p, n)
beta[,ctl] = 0
W = matrix(rnorm(m*k),m,k)
alpha = matrix(rnorm(k*n),k,n)
epsilon = matrix(rnorm(m*n),m,n)
Y = X%*%beta + W%*%alpha + epsilon
## Run RUV-4
fit = RUV4(Y, X, ctl, k)
## Get adjusted variances and p-values
fit = variance_adjust(fit)
```

RUVI
RUV-I

## Description

The RUV-I algorithm. Generally used as a preprocessing step to RUV-2, RUV-4, RUV-inv, RUVrinv, or RUVIII. RUV1 is an alias of (identical to) RUVI.

## Usage

RUVI(Y, eta, ctl, include.intercept = TRUE)
RUV1 (Y, eta, ctl, include.intercept = TRUE)

## Arguments

Y The data. A $m$ by $n$ matrix, where $m$ is the number of samples and $n$ is the number of features.
eta Gene-wise (as opposed to sample-wise) covariates. A matrix with n columns.
ctl The negative controls. A logical vector of length $n$.
include.intercept
Add an intercept term to eta if it does not include one already.

## Details

Implements the RUV-I algorithm as described in Gagnon-Bartsch, Jacob, and Speed (2013). Most often this algorithm is not used directly, but rather is called from RUV-2, RUV-4, RUV-inv, or RUV-rinv. Note that RUV1 and RUVI are two different names for the same (identical) function.

## Value

An adjusted data matrix (i.e., an adjusted Y)

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.
Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

RUV2, RUV4, RUVinv, RUVrinv, RUVIII
RUVIII RUV-III

## Description

Globally adjust data matrix using both negative controls and replicates.

## Usage

RUVIII(Y, M, ctl, k = NULL, eta = NULL, include.intercept = TRUE, average $=$ FALSE, fullalpha $=$ NULL, return.info $=$ FALSE, inputcheck $=$ TRUE)

## Arguments

$\mathrm{Y} \quad$ The data. A $m$ by $n$ matrix, where $m$ is the number of observations and $n$ is the number of features.
M The replicate structure. Represented internally as a mapping matrix. The mapping matrix has $m$ rows (one for each observation), and each column represents a set of replicates. The ( $\mathrm{i}, \mathrm{j}$ )-th entry of the mapping matrix is 1 if the i -th observation is in replicate set j , and 0 otherwise. Each observation must be in exactly one set of replicates (some replicate sets may contain only one observation), and thus each row of $M$ must sum to 1 . M may be the mapping matrix itself. Alternatively, M may be a vector, factor, or dataframe, in which case it is converted to the mapping matrix by the replicate.matrix function.

| ctl | An index vector to specify the negative controls. Either a logical vector of length <br> $n$ or a vector of integers. |
| :--- | :--- |
| k | The number of unwanted factors to use. Can be 0 , in which case no adjustment <br> is made. Can also be NULL (the default value), in which case the maximum <br> possible value of k is used; note that in this case no singular value decomposition <br> is necessary and execution is faster. |
| etaGene-wise (as opposed to sample-wise) covariates. These covariates are ad- <br> justed for by RUV-1 before any further analysis proceeds. Can be either (1) a <br> matrix with $n$ columns, (2) a matrix with n rows, (3) a dataframe with n rows, <br> (4) a vector or factor of length n, or (5) simply 1, for an intercept term. |  |
| include.intercept |  |
| When eta is specified (not NULL) but does not already include an intercept term, |  |
| this will automatically include one. |  |

## Value

If codereturn.info is TRUE, a list is returned that contains:
new $\quad$ The adjusted data matrix.
M The replicate mapping matrix. Included for reference.
fullalpha Can be used to speed up future calls to RUVIII

Otherwise, if return.info is FALSE, only the adjusted data matrix is returned.

## Note

Additional resources can be found at http://www-personal.umich.edu/~johanngb/ruv/.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## Description

The RUV-inv algorithm. Estimates and adjusts for unwanted variation using negative controls.

## Usage

RUVinv(Y, X, ctl, Z=1, eta=NULL, include.intercept=TRUE, fullW0=NULL, invsvd=NULL, lambda=NULL, randomization=FALSE, iterN=100000, inputcheck=TRUE)

## Arguments

$Y \quad$ The data. A $m$ by $n$ matrix, where $m$ is the number of samples and $n$ is the number of features.
$X \quad$ The factor(s) of interest. A m by p matrix, where $m$ is the number of samples and $p$ is the number of factors of interest. Very often $p=1$. Factors and dataframes are also permissible, and converted to a matrix by design. matrix.
ctl An index vector to specify the negative controls. Either a logical vector of length $n$ or a vector of integers.

Z
Any additional covariates to include in the model, typically a m by q matrix. Factors and dataframes are also permissible, and converted to a matrix by design.matrix. Alternatively, may simply be 1 (the default) for an intercept term. May also be NULL.
eta Gene-wise (as opposed to sample-wise) covariates. These covariates are adjusted for by RUV-1 before any further analysis proceeds. Can be either (1) a matrix with $n$ columns, (2) a matrix with $n$ rows, (3) a dataframe with n rows, (4) a vector or factor of length $n$, or (5) simply 1 , for an intercept term.
include.intercept
Applies to both $Z$ and eta. When $Z$ or eta (or both) is specified (not NULL) but does not already include an intercept term, this will automatically include one. If only one of $Z$ or eta should include an intercept, this variable should be set to FALSE, and the intercept term should be included manually where desired.
fullwo Can be included to speed up execution. Is returned by previous calls of RUV4, RUVinv, or RUVrinv (see below).
invsvd Can be included to speed up execution. Generally used when calling RUV(r)inv many times with different values of lambda. Is returned by previous calls of RUV(r)inv (see below).
lambda Ridge parameter. If specified, the ridged inverse method will be used.
randomization Whether the inverse-method variances should be computed using randomly generated factors of interest (as opposed to a numerical integral).

```
iterN The number of random "factors of interest" to generate (used only when ran-
    domization=TRUE).
inputcheck Perform a basic sanity check on the inputs, and issue a warning if there is a
    problem.
```


## Details

Implements the RUV-inv algorithm as described in Gagnon-Bartsch, Jacob, and Speed (2013).

## Value

A list containing
betahat The estimated coefficients of the factor(s) of interest. A p by n matrix.
sigma2 Estimates of the features' variances. A vector of length $n$.
$t \quad t$ statistics for the factor(s) of interest. A p by $n$ matrix.
$\mathrm{p} \quad \mathrm{P}$-values for the factor(s) of interest. A p by n matrix.
Fstats $\quad F$ statistics for testing all of the factors in $X$ simultaneously.
Fpvals $\quad P$-values for testing all of the factors in $X$ simultaneously.
multiplier The constant by which sigma2 must be multiplied in order get an estimate of the variance of betahat
df The number of residual degrees of freedom.
W The estimated unwanted factors.
alpha The estimated coefficients of W.
byx The coefficients in a regression of Y on X (after both Y and X have been "adjusted" for Z ). Useful for projection plots.
bwx The coefficients in a regression of W on X (after X has been "adjusted" for Z). Useful for projection plots.
$X \quad X$. Included for reference.
$\mathrm{k} \quad \mathrm{k}$. Included for reference.
ctl ctl. Included for reference.
Z Z. Included for reference.
eta eta. Included for reference.
fullwo Can be used to speed up future calls of RUV4.
lambda lambda. Included for reference.
invsvd Can be used to speed up future calls of RUV(r)inv.
include.intercept
include.intercept. Included for reference.
method Character variable with value "RUVinv". Included for reference.

## Note

Additional resources can be found at http://www-personal.umich.edu/~johanngb/ruv/.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.
Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

RUV2, RUV4, RUVrinv, variance_adjust, invvar

## Examples

```
## Create some simulated data
m = 50
n = 10000
nc = 1000
p = 1
k = 20
ctl = rep(FALSE, n)
ctl[1:nc] = TRUE
X = matrix(c(rep(0,floor(m/2)), rep(1,ceiling(m/2))), m, p)
beta = matrix(rnorm(p*n), p, n)
beta[,ctl] = 0
W = matrix(rnorm(m*k),m,k)
alpha = matrix(rnorm(k*n),k,n)
epsilon = matrix(rnorm(m*n),m,n)
Y = X%*%beta + W%*%alpha + epsilon
## Run RUV-inv
fit = RUVinv(Y, X, ctl)
## Get adjusted variances and p-values
fit = variance_adjust(fit)
```

RUVrinv

Remove Unwanted Variation, ridged inverse method

## Description

The RUV-rinv algorithm. Estimates and adjusts for unwanted variation using negative controls.

## Usage

RUVrinv(Y, X, ctl, Z=1, eta=NULL, include.intercept=TRUE, fullwo=NULL, invsvd=NULL, lambda=NULL, k=NULL, l=NULL, randomization=FALSE, iterN=100000, inputcheck=TRUE)

## Arguments

$\mathrm{Y} \quad$ The data. A $m$ by $n$ matrix, where $m$ is the number of samples and $n$ is the number of features.
$X \quad$ The factor(s) of interest. A $m$ by $p$ matrix, where $m$ is the number of samples and $p$ is the number of factors of interest. Very often $p=1$. Factors and dataframes are also permissible, and converted to a matrix by design.matrix.
ctl An index vector to specify the negative controls. Either a logical vector of length n or a vector of integers.
Z
Any additional covariates to include in the model, typically a m by q matrix. Factors and dataframes are also permissible, and converted to matrix by design.matrix. Alternatively, may simply be 1 (the default) for an intercept term. May also be NULL.
eta Gene-wise (as opposed to sample-wise) covariates. These covariates are adjusted for by RUV-1 before any further analysis proceeds. Can be either (1) a matrix with n columns, (2) a matrix with n rows, (3) a dataframe with n rows, (4) a vector or factor of length $n$, or (5) simply 1 , for an intercept term.
include.intercept
Applies to both $Z$ and eta. When $Z$ or eta (or both) is specified (not NULL) but does not already include an intercept term, this will automatically include one. If only one of $Z$ or eta should include an intercept, this variable should be set to FALSE, and the intercept term should be included manually where desired.
fullwo Can be included to speed up execution. Is returned by previous calls of RUV4, RUVinv, or RUVrinv (see below).
invsvd Can be included to speed up execution. Generally used when calling RUV(r)inv many times with different values of lambda. Is returned by previous calls of RUV(r)inv (see below).
lambda Ridge parameter. If unspecified, an appropriate default will be used.
$\mathrm{k} \quad$ When calculating the default value of lambda, a call to RUV4 is made. This parameter specifies the value of k to use. Otherwise, an appropriate default k will be used.
$1 \quad$ If lambda and k are both NULL, then k must be estimated using the getK routine. The getK routine only accepts a single-column $X$. If $p>1,1$ specifies which column of $X$ should be used in the getK routine.
randomization Whether the inverse-method variances should be computed using randomly generated factors of interest (as opposed to a numerical integral).
iterN The number of random "factors of interest" to generate (used only when randomization=TRUE).
inputcheck Perform a basic sanity check on the inputs, and issue a warning if there is a problem.

## Details

Implements the RUV-rinv algorithm as described in Gagnon-Bartsch, Jacob, and Speed (2013). This function is essentially just a wrapper to RUVinv, but with a little extra code to calculate the default value of lambda.

## Value

## A list containing

$$
\text { betahat } \quad \text { The estimated coefficients of the factor(s) of interest. A p by } \mathrm{n} \text { matrix. }
$$

sigma2 Estimates of the features' variances. A vector of length $n$.
$t \quad t$ statistics for the factor(s) of interest. A p by n matrix.
$\mathrm{p} \quad \mathrm{P}$-values for the factor(s) of interest. A p by n matrix.
Fstats $\quad F$ statistics for testing all of the factors in $X$ simultaneously.
Fpvals $\quad P$-values for testing all of the factors in $X$ simultaneously.
multiplier The constant by which sigma2 must be multiplied in order get an estimate of the variance of betahat
df The number of residual degrees of freedom.
W The estimated unwanted factors.
alpha The estimated coefficients of W.
byx The coefficients in a regression of Y on X (after both Y and X have been "adjusted" for Z). Useful for projection plots.
bwx The coefficients in a regression of W on X (after X has been "adjusted" for Z). Useful for projection plots.
$X \quad X$. Included for reference.
$k \quad k$. Included for reference.
ctl ctl. Included for reference.
Z
eta eta. Included for reference.
fullwo Can be used to speed up future calls of RUV4.
lambda lambda. Included for reference.
invsvd Can be used to speed up future calls of RUV(r)inv.
include.intercept
include. intercept. Included for reference.
method Character variable with value "RUVinv". Included for reference. (Note that RUVrinv is simply a wrapper to RUVinv, hence both return "RUVinv" as the method.)

## Note

Additional resources can be found at http://www-personal.umich.edu/~johanngb/ruv/.

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.

Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

RUV2, RUV4, RUVinv, variance_adjust, invvar, getK

## Examples

```
## Create some simulated data
m = 50
n = 10000
nc = 1000
p = 1
k = 20
ctl = rep(FALSE, n)
ctl[1:nc] = TRUE
X = matrix(c(rep(0,floor(m/2)), rep(1,ceiling(m/2))), m, p)
beta = matrix(rnorm(p*n), p, n)
beta[,ctl] = 0
W = matrix(rnorm(m*k),m,k)
alpha = matrix(rnorm(k*n),k,n)
epsilon = matrix(rnorm(m*n),m,n)
Y = X%*%beta + W%*%alpha + epsilon
## Run RUV-rinv
fit = RUVrinv(Y, X, ctl)
## Get adjusted variances and p-values
fit = variance_adjust(fit)
```

ruv_cancorplot

RUV Canonical Correlation Plot

## Description

Canonical correlation plot

## Usage

ruv_cancorplot(Y, X, ctl, W1 = NULL, W2 = NULL)

## Arguments

Y The data matrix. Rows are observations and columns are features (e.g. genes).
$X \quad$ Factor(s) of interest. Can be a vector, factor, matrix, or dataframe. Must have the same length (or number of rows) as the number of row of Y.
ctl Index of negative controls.
W1 Optional. The left singular vectors of Y. Can be included to speed up execution.
W2 Optional. The left singular vectors of Y[,ctl]. Can be included to speed up execution.

## Details

Plots, as a function of $k$, the square of of the first canonical correlation of $X$ and the first $k$ left singular vectors of $Y$ (and also, similarly, $Y[, c t l]$ ).

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch

```
ruv_ecdf RUV P-value Empirical CDF Plot
```


## Description

Plots an ECDF of p-values returned by a call to ruv_summary

## Usage

ruv_ecdf(fit, X.col = "all", power = 1, uniform.lines = 0)

## Arguments

fit The results of a call to ruv_summary.
X.col Which column of the X matrix to make the plot for, i.e. which factor's p-values to plot. Can be either an integer or a character string. Or, if "all" (the default), use the F-test p-values.
power A power transformation of the $x$ and $y$ axes. For example, set to $1 / 2$ for a squareroot transformation. This can help to see the behavior of the ECDF near 0.
uniform. lines A vector of values between 0 and 1 , or NULL. If specified, light gray lines will be drawn, showing (locally) uniform distributions.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch
ruv_hist
RUV P-value Histogram Plot

## Description

Plots a histogram of p-values returned by a call to ruv_summary

## Usage

ruv_hist(fit, X.col = "all", breaks $=c(0,0.001,0.01,0.05, \operatorname{seq}(0.1,1$, by $=0.1))$ )

## Arguments

fit The results of a call to ruv_summary.
X.col Which column of the X matrix to make the plot for, i.e. which factor's p-values to plot. Can be either an integer or a character string. Or, if "all" (the default), use the F-test p-values.
breaks Breakpoints of the histogram.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch

## Description

Projection plot of an RUV regression fit.

## Usage

ruv_projectionplot(fit, X.col = 1, factor = "gradient", adjusted = TRUE)

## Arguments

fit The results of a call to ruv_summary.
X.col Which column of the X matrix to make the plot for. Can be either an integer or a character string.
factor Which unwanted factor to use (horizontal axis). Must be either an integer or the character string "gradient".
adjusted Whether the plot should be adjusted for unwanted factors other than the one being plotted. Not relevant when factor = "gradient".

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch
ruv_rankplot
RUV Rank Plot

## Description

A plot showing the number of positive controls to be found within the N top-ranked features, as a function of $N$. The ranking of the features is by p-value.

## Usage

ruv_rankplot(fit, pctl, X.col = "all", uniform.lines = 0)

## Arguments

$$
\begin{array}{ll}
\text { fit } & \text { The results of a call to ruv_summary. } \\
\text { pctl } & \begin{array}{l}
\text { Either an integer or character string specifying which column of fit\$C to be } \\
\text { used as positive controls. (Must be a logical vector). Alternatively, may some } \\
\text { other index vector specifying the positive controls; importantly, in this case, the } \\
\text { index vector must index the features as they are sorted in fit } \$ C .
\end{array} \\
\text { X.col } & \begin{array}{l}
\text { Which column of the X matrix to make the plot for. Can be either an integer or } \\
\text { a character string. Or, if "all" (the default), use the F-test p-values. }
\end{array} \\
\text { uniform.lines } \begin{array}{l}
\text { A vector of values between 0 and 1, or NULL. If specified, light gray lines will } \\
\text { be drawn, showing (locally) uniform distributions. }
\end{array}
\end{array}
$$

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch
ruv_residuals RUV Residuals

## Description

Calculate the residuals or adjusted data matrix of an RUV2 or RUV4 fit.

## Usage

ruv_residuals(fit, type=c("residuals", "adjusted.Y"), subset_and_sort=TRUE)

## Arguments

fit The results of a call to ruv_summary.
type Whether to compute residuals or an adjusted data matrix. Caution; see details below.
subset_and_sort
Whether to subset and sort the features, as in ruv_summary.

## Details

This function will return either the residuals or an adjusted data matrix. The residuals are the result of removing all factors (wanted and unwanted), whereas the adjusted data matrix is the result of removing only the unwanted factors.
The residuals can be useful for diagnostics, e.g. in producing a residual RLE plot. The adjusted data matrix may also be useful for diagnostics, but typically should *not* be used for any additional downstream analyses. The adjusted data matrix can suffer from overfitting, which can be severe, especially when k is large, and this can produce artificially "good" results in downstream analyses.
If an adjusted data matrix for use in downstream analyses is desired, see RUVIII.

## Value

Either a matrix of residuals, or an adjusted data matrix.

## Author(s)

Johann Gagnon-Bartsch

See Also
RUV2, RUV4, ruv_summary, RUVIII
ruv_rle RUV RLE Plot

## Description

An RLE (Relative Log Expression) Plot

## Usage

ruv_rle $(Y$, rowinfo $=N U L L$, probs $=c(0.05,0.25,0.5,0.75,0.95), y l i m=c(-0.5,0.5))$

## Arguments

Y The data matrix. Rows are observations and columns are features (e.g. genes).
rowinfo A dataframe of information about the observations. Should have the same number of rows as Y. This information will be included in the ggplot, and can be used for setting aesthetics such as color.
probs The percentiles used to construct the boxplots. By default, whiskers are drawn to the 5th and 95th percentiles. Note that this is non-standard for boxplots.
ylim Limits of the y axis. Defaults to $(-0.5,0.5)$ so that the plots are always on the same scale and can be easily compared.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch

## References

Irizarry, R. A., Bolstad, B. M., Collin, F., Cope, L. M., Hobbs, B., and Speed, T. P. (2003). Summaries of Affymetrix GeneChip probe level data. Nucleic acids research, 31(4):e15.

```
ruv_scree RUV Scree Plot
```


## Description

A scree plot (on the log scale)

## Usage

ruv_scree( $Y=$ NULL, $Z=1, Y . s v d=N U L L)$

## Arguments

Y The data matrix. Rows are observations and columns are features (e.g. genes). If not specified, Y.svd must be specified instead (which is faster).
Z Any variables to regress out of Y as a preprocessing step. May simply be 1 (the default) for an intercept term, i.e. the columns of Y are mean centered. May also be NULL.
Y.svd The SVD of Y, as returned by the svd function.

## Details

Because 0 cannot be plotted on a log scale, if any singular values are equal to 0 , they will be changed to the minimum non-zero singular value and plotted in red. Exception: singular values that are 0 as a result of regressing out Z are simply not plotted.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch

```
ruv_shiny RUV Shiny App
```


## Description

A Shiny App that allows quick exploration of a dataset using RUV methods.

## Usage

```
ruv_shiny(Y, rowinfo, colinfo, options = list(port = 3840))
```


## Arguments

$$
\begin{array}{ll}
\text { Y } & \text { The data matrix. Rows are observations and columns are features (e.g. genes). } \\
\text { rowinfo } & \begin{array}{l}
\text { A dataframe of information about the observations. Should have the same num- } \\
\text { ber of rows as Y. Should contain at least one column that can be used as either a } \\
\text { factor of interest or to define replicates. }
\end{array} \\
\text { colinfo } & \begin{array}{l}
\text { A dataframe of information about the observations. Should have the same num- } \\
\text { ber of rows as Y. Should contain at least one column that is a logical vector that } \\
\text { can be used to define negative controls. }
\end{array} \\
\text { options } & \text { A list of options to pass to the shinyApp function. }
\end{array}
$$

## Value

None. Calls shinyApp.

## Author(s)

Johann Gagnon-Bartsch
ruv_summary RUV Summary

## Description

Post-process and summarize the results of call to RUV2, RUV4, RUVinv, or RUVrinv.

## Usage

ruv_summary(Y, fit, rowinfo=NULL, colinfo=NULL, colsubset=NULL, sort.by="F.p", var.type=c("ebayes", "standard", "pooled"), p.type=c("standard", "rsvar", "evar"), min.p.cutoff=10e-25)

## Arguments

Y
fit
rowinfo A matrix or dataframe containing information about the rows (samples). This information is included in the summary that is returned.
colinfo A matrix or dataframe containing information about the columns (features, e.g. genes). This information is included in the summary that is returned.
colsubset A vector indexing the features of interest. Only only data on these features will be returned.
sort.by An index variable; which column of C (see below) should be used to sort the features. The default is "F.p", meaning that features will be sorted by the F-test p -value. If NULL, features will note be sorted.
var.type Which type of estimate for sigma2 should be used from the call to variance_adjust? The options are "ebayes", "standard", or "pooled." See variance_adjust for details.
p.type Which type of p-values should be used from the call to variance_adjust? The options are "standard", "rsvar", or "evar".
min.p.cutoff p-values below this value will be changed and set equal to this value. Useful for plotting p-values on a log scale.

## Details

This function post-processes the results of a call to RUV2/4/inv/rinv and then nicely summarizes the output. The post-processing step primarily consists of a call to variance_adjust, which computes various adjustments to variances, t-statistics, and and p-values. See variance_adjust for details. The var. type and p.type options determine which of these adjustments are used. An additional postprocessing step is that the column means of the $Y$ matrix are computed, both before and after the call to RUV1 (if eta was specified).

After post-processing, the results are summarized into a list containing 4 objects: 1) the data matrix $Y ; 2$ ) a dataframe $R$ containing information about the rows (samples); 3) a dataframe $C$ containing information about the columns (features, e.g. genes), and 4) a list misc of other information returned by RUV2/4/inv/rinv.

Finally, if colsubset is specified, then $C$ is subset to include only the features of interest (as are the relevant entries of misc that are used to compute projection plots). If sort.by is specified, the features will also be sorted.

## Value

A list containing:
$Y \quad$ The original data matrix.
$R \quad$ A dataframe of row-wise information, including $X, Z$, and any other data passed in with rowinfo

C A dataframe of column-wise information, including p-values, estimated regression coefficients, estimated variances, column means, an index of the negative controls, and any other data passed in with colinfo.
misc A list of additional information returned by RUV2/4/inv/rinv

## Author(s)

Johann Gagnon-Bartsch

## See Also

RUV2, RUV4, RUVinv, RUVrinv, variance_adjust

## Description

A plot composed of a grid of several subplots created by ruv_svdplot

## Usage

$$
\begin{aligned}
\text { ruv_svdgridplot } & \text { Y.data, Y.space }=\text { NULL, rowinfo }=\text { NULL, colinfo }=N U L L, k=1: 3, \mathrm{Z}=1 \text {, } \\
& \text { left.additions }=\text { NULL, right.additions }=\text { NULL, } \\
& \text { factor.labels }=\text { paste("S.V.", k)) }
\end{aligned}
$$

## Arguments

Y.data The data matrix. Rows are observations and columns are features (e.g. genes).
Y.space Either a data matrix of the same dimension as Y.data, or the SVD of such a matrix, as returned by the svd function. The singular vectors of this matrix define the space in which Y.data will be plotted. If NULL, Y.data itself is used.
rowinfo A dataframe of information about the observations. Should have the same number of rows as Y. This information will be included in the ggplots, and can be used for setting aesthetics such as color.
colinfo A dataframe of information about the observations. Should have a number of rows equal to the number of columns of Y. This information will be included in the ggplots, and can be used for setting aesthetics such as color.
$k \quad$ A numeric vector of the singular vectors to be plotted. Typically integers, but fractional values can also be specified. For example, a value of 2.5 corresponds to the linear combination (singular vector 2 ) + (singular vector 3 ), rescaled to have unit length. Similarly, a value of 2.2 corresponds to the (rescaled) linear combination $8^{*}$ (singular vector 2$)+2 *$ (singular vector 3 ), and -2.2 corresponds to the (rescaled) linear combination $8 *$ (singular vector 2$)-2 *$ (singular vector 3). Note that the vectors defined by 2.2 and -2.8 are orthogonal to each other, as are those defined by 2.3 and -2.7 , etc.
Z Any variables to regress out of Y.data as a preprocessing step. May simply be 1 (the default) for an intercept term, i.e. the columns of $Y$ are mean centered. May also be NULL. Similarly for Y. space, unless Y. space is already an SVD.
left.additions A list of additions to the ggplots of the left singular vectors. Can be used to set aesthetics such as color, etc.
right.additions
A list of additions to the ggplots of the right singular vectors. Can be used to set aesthetics such as color, etc.
factor.labels The factor labels.

## Details

Plots of the left singular vectors are shown on the left, and plots of the right singular vectors are shown on the right. The diagonal shows squares with side lengths proportional to the singular values.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch
ruv_svdplot RUVSVD Plot

## Description

A generalization of a PC (principal component) plot.

## Usage

ruv_svdplot(Y.data, Y.space $=$ NULL, $\operatorname{info}=$ NULL, $k=c(1,2), Z=1$, left $=$ TRUE)

## Arguments

Y.data The data matrix. Rows are observations and columns are features (e.g. genes).
Y.space Either a data matrix of the same dimension as Y.data, or the SVD of such a matrix, as returned by the svd function. The singular vectors of this matrix define the space in which Y.data will be plotted. If NULL, Y.data itself is used.
info Additional data to be included in the ggplot, which can be used for setting aesthetics such as color. Converted to a dataframe, which should have a number of rows equal to the number of rows of $Y$. data (if left=TRUE) or the number of columns of Y . data (if left=FALSE).
k
A numeric vector of length 2. The singular vectors to be plotted. Typically integers, but fractional values can also be specified. For example, a value of 2.5 corresponds to the linear combination (singular vector 2 ) + (singular vector 3), rescaled to have unit length. Similarly, a value of 2.2 corresponds to the (rescaled) linear combination $8 *($ singular vector 2$)+2 *$ (singular vector 3 ), and -2.2 corresponds to the (rescaled) linear combination $8 *$ (singular vector 2 ) $-2^{*}$ (singular vector 3 ). Note that the vectors defined by 2.2 and -2.8 are orthogonal to each other, as are those defined by 2.3 and -2.7 , etc.
Z
Any variables to regress out of Y. data as a preprocessing step. May simply be 1 (the default) for an intercept term, i.e. the columns of Y are mean centered. May also be NULL. Similarly for Y. space, unless Y. space is already an SVD.
left Plot the left singular vectors (if TRUE) or the right singular vectors (if FALSE).

## Details

When $Y$. space $=$ NULL and $Z=1$ and the values of $k$ are integers, this is a standard PC plot.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch
ruv_varianceplot RUV Variance Plot

## Description

A scatter plot of (squared) coefficient estimates against variance estimates.

## Usage

ruv_varianceplot(fit, X.col = 1, power = 1/4)

## Arguments

fit The results of a call to ruv_summary.
X.col Which column of the X matrix to make the plot for. Can be either an integer or a character string.
power Power transformation of the x and y axes. Default is fourth root.

## Details

A black curve is also plotted, showing the estimated variances of the coefficient estimates.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch

```
ruv_volcano RUV Volcano Plot
```


## Description

A scatter plot of negative $\log \mathrm{p}$-values against coefficient estimates, commonly known as a volcano plot

## Usage

ruv_volcano(fit, X.col = 1)

## Arguments

fit The results of a call to ruv_summary.
X.col Which column of the X matrix to make the plot for. Can be either an integer or a character string.

## Value

A ggplot.

## Author(s)

Johann Gagnon-Bartsch
sigmashrink Empirical Bayes shrinkage estimate of sigma^2

## Description

This function (re)implements the empirical bayes shrinkage estimate of Smyth (2004), which is also implemented in the Limma package. This function is normally called from the function variance_adjust, and is not normally intended for stand-alone use.

## Usage

sigmashrink(s2, d)

## Arguments

s2
"Standard" estimates of sigma^2
d
"Standard" degrees of freedom of the residuals

## Value

A list containing
sigma2 Estimates of sigma^2 using the empirical bayes shrinkage method of Smyth (2004)
df Estimate of degrees of freedom using the empirical bayes shrinkage method of Smyth (2004)

## Author(s)

Johann Gagnon-Bartsch [johanngb@umich.edu](mailto:johanngb@umich.edu)

## References

Linear models and empirical bayes methods for assessing differential expression in microarray experiments. Smyth, 2004.
Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.
Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.

## See Also

variance_adjust

```
variance_adjust Adjust Estimated Variances
```


## Description

Calculate rescaled variances, empirical variances, etc. For use with RUV model fits.

## Usage

variance_adjust(fit, ctl.idx = NULL, ebayes = TRUE, pooled=TRUE, evar = TRUE, rsvar $=$ TRUE, bin $=10$, rescaleconst $=$ NULL)

## Arguments

fit A RUV model fit (a list), as returned by RUV2 / RUV4 / RUVinv / RUVrinv
ctl.idx An index vector to specify the negative controls for use with the rescaled variances method. If unspecified, by default fit\$ctl is used.
ebayes A logical variable. Should empirical bayes variance estimates be calculated?
pooled A logical variable. Should pooled variance estimates be calculated?
evar A logical variable. Should empirical variance estimates be calculated?

| rsvar | A logical variable. Should rescaled variance estimates be calculated? |
| :--- | :--- |
| bin | The bin size to use when calculating empirical variances. |
| rescaleconst | Can be used to speed up execution. See get_empirical_variances. |

## Value

An RUV model fit (a list). In addition to the elements of the list returned by RUV2 / RUV4 / RUVinv / RUVrinv, the list will now contain:
sigma2.ebayes Estimates of sigma^2 using the empirical bayes shrinkage method of Smyth (2004)
df.ebayes Estimate of degrees of freedom using the empirical bayes shrinkage method of Smyth (2004)
sigma2. pooled Estimate of sigma^2 pooled (averaged) over all genes
df.pooled Degrees of freedom for pooled estimate
varbetahat "Standard" estimate of the variance of betahat varbetahat.rsvar
"Rescaled Variances" estimate of the variance of betahat
varbetahat.evar
"Empirical Variances" estimate of the variance of betahat
varbetahat.ebayes
"Empirical Bayes" estimate of the variance of betahat
varbetahat.rsvar.ebayes
"Rescaled Empirical Bayes" estimate of the variance of betahat
varbetahat. pooled
"Pooled" estimate of the variance of betahat
varbetahat.rsvar.pooled
"Rescaled pooled" estimate of the variance of betahat
varbetahat.evar.pooled
Similar to the above, but all genes used to determine the rescaling, not just control genes
p.rsvar $\quad$ P-values, after applying the method of rescaled variances
p.evar P-values, after applying the method of empirical variances
p.ebayes P-values, after applying the empirical bayes method of Smyth (2004)
p.pooled P-values, after pooling variances
p.rsvar.ebayes P-values, after applying the empirical bayes method of Smyth (2004) and the method of rescaled variances
p.rsvar. pooled P -values, after pooling variances and the method of rescaled variances
p.evar. pooled Similar to the above, but all genes used to determine the rescaling, not just control genes
Fpvals.ebayes $F$ test p-values, after applying the empirical bayes method of Smyth (2004)
Fpvals. pooled $F$ test $p$-values, after pooling variances
p.BH FDR-adjusted p-values

```
Fpvals.BH FDR-adjusted p-values (from F test)
p.rsvar.BH FDR-adjusted p-values (from p.rsvar)
p.evar.BH FDR-adjusted p-values (from p.evar)
p.ebayes.BH FDR-adjusted p-values (from p.ebayes)
p.rsvar.ebayes.BH
FDR-adjusted p-values (from p.rsvar.ebayes)
Fpvals.ebayes.BH
    FDR-adjusted F test p-values (from Fpvals.ebayes)
p.pooled.BH FDR-adjusted p-values (from p.pooled)
p.rsvar. pooled. BH
    FDR-adjusted p-values (from p.rsvar.pooled)
p.evar.pooled.BH
    FDR-adjusted p-values (from p.evar.pooled)
Fpvals.pooled.BH
FDR-adjusted F test p-values (from Fpvals.pooled)
```


## Author(s)

Johann Gagnon-Bartsch

## References

Using control genes to correct for unwanted variation in microarray data. Gagnon-Bartsch and Speed, 2012. Available at: http://biostatistics.oxfordjournals.org/content/13/3/539.full.
Removing Unwanted Variation from High Dimensional Data with Negative Controls. GagnonBartsch, Jacob, and Speed, 2013. Available at: http://statistics.berkeley.edu/tech-reports/820.
Linear models and empirical bayes methods for assessing differential expression in microarray experiments. Smyth, 2004.

## See Also

RUV2, RUV4, RUVinv, RUVrinv, get_empirical_variances, sigmashrink

## Examples

```
## Create some simulated data
m = 50
n = 10000
nc = 1000
p = 1
k = 20
ctl = rep(FALSE, n)
ctl[1:nc] = TRUE
X = matrix(c(rep(0,floor(m/2)), rep(1,ceiling(m/2))), m, p)
beta = matrix(rnorm(p*n), p, n)
beta[,ctl] = 0
W = matrix(rnorm(m*k),m,k)
```

variance_adjust

```
alpha = matrix(rnorm(k*n),k,n)
epsilon = matrix(rnorm(m*n),m,n)
Y = X%*%beta + W%*%alpha + epsilon
## Run RUV-inv
fit = RUVinv(Y, X, ctl)
## Get adjusted variances and p-values
fit = variance_adjust(fit)
```


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