## Package 'dbnorm'

January 30, 2021

Title Drift Across-Batches Normalization and Visualization

**Description** dbnorm includes several functions applicable in a large-scale Metabolomics analysis as well as other big data. Notably, it includes distinct functions for processing of data and estimation of missing values, functions for batch effect correction based on using several statistical models together with various graphical checks and visualize inference. By evaluating model performances from multiple perspectives, sample batch and metabolic features and by comprehensive visualization of data structure, "dbnorm" enabled inclusive comparison among the integrated methods.

Version 0.2.2

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**Encoding** UTF-8

License GPL-3

**Depends** R (>= 3.5)

URL https://github.com/NBDZ/dbnorm

**Imports** sva ,ber(<= 4.0), NormalizeMets(<= 0.25),factoextra, ggplot2

## **Suggests**

AUC, base, Biobase, BiocParallel, devtools, DiffCorr, e1071, edgeR, fs, GGally, ggfortify, graphics, grDevices, installr, knitr, limma, MASS, MetNorm, pcaMethods, plotly, processx, RGtk2, rmarkdown, stats, statTarget, tibble, usethis, utils

LazyData yes

RoxygenNote 7.1.1

## **R** topics documented:

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#### **Description**

It is a function in dbnorm, a package in R. This function allows users to effectively remove variance associated with batch from data by applying L/S method and partial bagging with Bootstrap samples of size n=150, as explain by M.Giordan. In fact, in *dbnormBagging*, we included various types of graphical check for visualization of data point, sample-wise and feature-wise. Considering a single function as a correction algorithm, we aimed to fasten computational processing of big data.

Notably, using this function users applied unsupervised learning algorithm to visualize the most variance explained by the two first components in the different set of samples analyzed in the entire experiment in the raw and corrected data. In parallel, linear association of feature (variable) and batch level has been estimated and visualized by a correlation plot. In fact, estimated *Adjusted-R2* is considered to define the percentage of variance in a dependent variable estimated by independent variable (batch) in a original data (Raw data) and in corrected data. Besides, for quick notification about the performance of the applied model a maximum variability detected in either of datasets is reported as a score. This score notifies the consistency of model performance for all detected features (variables).

## Usage

dbnormBagging(m)

## **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch levels in the first column.

#### **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and /or emvd implemented in the 'dbnorm' package. Input must be normalized and transformed prior.

## Value

Several graphs compiled into a **PDF** file are a *PCA* score plot, *Scree* plot and a *correlation* plot estimated for raw and corrected data. Also, the *RLA* plot for each dataset visualized in the **Viewer** panel in the **rstudio** console.

Files saved as **csv** in the working directory are a dataset corrected by the applied model. Also, a two column matrix for Adjusted R-Square raw and corrected dataset and a table summarizing the maximum score.

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#### References

M.Giordan (2013) < DOI:10.1007/s12561-013-9081-1> https://link.springer.com/article/10.1007/s12561-013-9081-1

## **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5)),1)
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
dbnormBagging(m)
## End(Not run)</pre>
```

dbnormBer

Drift Across Batch Normalization via ber- model and visualization

#### **Description**

It is a function in dbnorm, a package in R. This function allows users to effectively remove variance associated with batch from data by applying L/S method as explain by M.Giordan. In fact, in *dbnormBer*, we included various types of graphical check for visualization of data point, sample-wise and feature-wise. Considering a single function as a correction algorithm, we aimed to fasten computational processing of big data.

Notably, using this function users applied unsupervised learning algorithm to visualize the most variance explained by the two first components in the different set of samples analyzed in the entire experiment in the raw and corrected data. In parallel, linear association of feature (variable) and batch level has been estimated and visualized by a correlation plot. In fact, estimated *Adjusted-R2* is considered to define the percentage of variance in a dependent variable estimated by independent variable (batch) in a original data (Raw data) and in corrected data. Besides, for quick notification about the performance of the applied model a maximum variability detected in either of datasets is reported as a score. This score notifies the consistency of model performance for all detected features (variables).

## Usage

dbnormBer(m)

#### **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch levels in the first column.

#### **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and /or emvd implemented in the 'dbnorm' package. Input must be normalized and transformed prior.

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#### Value

Several graphs compiled into a **PDF** file are a *PCA* score plot, *Scree* plot and a *correlation* plot estimated for raw and corrected data. Also, the *RLA* plot for each dataset visualized in the **Viewer** panel in the **rstudio** console.

Files saved as **csv** in the working directory are a dataset corrected by the applied model. Also, a two column matrix for Adjusted R-Square raw and corrected dataset and a table summarizing the maximum score.

#### References

M.Giordan (2013) < DOI:10.1007/s12561-013-9081-1> https://link.springer.com/article/10.1007/s12561-013-9081-1

## **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5)),1)
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
dbnormBer(m)
## End(Not run)</pre>
```

dbnormNPcom

Drift Across Batch Normalization via non-Parametric ComBat model and visualization

#### **Description**

It is a function in dbnorm, a package in R. This function allows you adjust the data for signal drift across multiple batches or batch effect using non-parametric ComBat methods (see "ComBat" in "sva", a package in bioconductor). Including single method in *dbnormNPcom*, we aimed to fasten the computational processing of big data. This function includes various types of graphical check for visualization of data point, sample-wise and feature-wise.

Notably, using this function users applied unsupervised learning algorithm to visualize the most variance explained by the two first components in the different set of samples analyzed in the entire experiment in the raw and corrected data. In parallel, linear association of feature (variable) and batch level has been estimated and visualized by a correlation plot. In fact, estimated *Adjusted-R2* is considered to define the percentage of variance in a dependent variable estimated by independent variable (batch) in a original data (Raw data) and in corrected data. Besides, for quick notification about the performance of the applied model a maximum variability detected in either of datasets is reported as a score. This score notifies the consistency of model performance for all detected features (variables).

#'

## Usage

dbnormNPcom(m)

## Arguments

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch in the first column.

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#### **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and /or emvd, functions implemented in 'dbnorm' package. Input must be normalized and transformed prior.

#### Value

Several graphs compiled into a **PDF** file are a *PCA* score plot, *Scree* plot and a *correlation* plot for raw and corrected dataset. Also, the *RLA* plot for each dataset visualized in the **Viewer** panel in the **rstudio** console.

Files saved as **csv** in the working directory are a dataset corrected by the applied model. Also, a two column matrix for Adjusted R-Square raw and corrected dataset and a table summarizing the maximum score.

#### References

```
Johnson et al.(2007) http://www.ncbi.nlm.nih.gov/pubmed/16632515
Leek et al. (2012) https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3307112/
```

#### **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5)),1)
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
dbnormNPcom(m)
## End(Not run)</pre>
```

dbnormPcom

Drift Across Batch Normalization via Parametric- ComBat model and visualization

#### **Description**

It is a function in dbnorm, a package in R. This function allows you adjust the data for signal drift across multiple batches or batch effect applying emphEmpirical Bayes method (see "ComBat" in "sva", a package in bioconductor). Including single method in *dbnormPcom*, we aimed to fasten the computational processing of big data. This function includes various types of graphical check for visualization of data point, sample-wise and feature-wise.

Notably, using this function users applied unsupervised learning algorithm to visualize the most variance explained by the two first components in the different set of samples analyzed in the entire experiment in the raw and corrected data. In parallel, linear association of feature (variable) and batch level has been estimated and visualized by a correlation plot. In fact, estimated *Adjusted-R2* is considered to define the percentage of variance in a dependent variable estimated by independent variable (batch) in a original data (Raw data) and in corrected data. Besides, for quick notification about the performance of the applied model a maximum variability detected in either of datasets is reported as a score. This score notifies the consistency of model performance for all detected features (variables).

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## Usage

```
dbnormPcom(m)
```

#### **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch levels in the first column.

#### Details

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and /or emvd implemented in dbnorm package. Input must be normalized and transformed prior.

#### Value

Several graphs compiled into a **PDF** file are a *PCA* score plot, *Scree* plot and a *correlation* plot for raw and corrected data. Also, the *RLA* plots for each dataset visualized in the **Viewer** panel in the **rstudio** console.

Files saved as **csv** in the working directory are a dataset corrected by the applied model. Also, a two column matrix for Adjusted R-Square raw and corrected dataset and a table summarizing the maximum score.

#### References

Johnson et al., (2007) < DOI:10.1093/biostatistics/kxj037 > http://www.ncbi.nlm.nih.gov/pubmed/16632515 Leek et al., (2012) < DOI:10.1093/bioinformatics/bts034> https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3307112/

#### **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5)),1)
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
dbnormPcom(m)
## End(Not run)</pre>
```

dbnormSCORE

Adjusted Coefficient Of Determination for a data normalized for signal drift across batches

## **Description**

It is a function in dbnorm, a package in R. This function gives a quick notification about the performance of the compiled statistical models namely two-stage regression procedure (see functions such as "ber" and ber-bg using partial bagging model with n=150 bootstrap samples) and/or empirical Bayes methods in two setting of parametric and non-parametric (see "ComBat" in "sva", a package in bioconductor), on accommodation of batch effect. Using this function users will estimate values of adjusted coefficient of determination (Adjusted R- Squared) which address the dependency of each feature (variable) to the batch order in each dataset. Immediately, a score calculated based on the maximum variability estimated by the regression analysis is reported and presented in graph.

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This score notifies the consistency of a model performance for the detected features (variables), facilitating quick comparison of the models for selecting one of those models, which is more appropriate to the data structure.

### Usage

```
dbnormSCORE(m)
```

#### **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch levels in the first column.

#### **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by emvf or emvd, functions implemented in dbnorm package. Input data must be normalized prior.

#### Value

Several graphs compiled into a **PDF** file which are a *correlation* plot for each of applied models, a grouped *barplot* presenting the maximum variability associated with batch levels in the raw and the corrected datasets.

Files saved as **csv** in the working directory are a dataset corrected via either of applied models. Also, a two column matrix for Adjusted R-Square for raw and corrected datasets and a table summarizing the score values presented in *barplot*.

## References

```
Giordan (2013) < DOI:10.1007/s12561-013-9081-1 > https://link.springer.com/article/10.1007/s12561-013-9081-1
```

Johnson et al., (2007) < DOI: 10.1093/biostatistics/kxj037 > http://www.ncbi.nlm.nih.gov/pubmed/16632515 Leek et al., (2012) < DOI:10.1093/bioinformatics/bts034 > https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3307112/

## Examples

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5)),1)
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
dbnormSCORE(m)
## End(Not run)</pre>
```

emvd

Estimation of missing value data-based

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## **Description**

This is a function in the dbnorm, a package in R. It returns to a matrix of data in which missing values are estimated by the lowest detected value in the entire experiment. By this function, all NA values are replaced by Zero values, that of being ultimately replaced by the lowest value detected in the experiment. Ultimately, data matrix is transposed to restore original structure.

## Usage

emvd(m)

#### **Arguments**

m

An array or a matrix

## **Details**

empty entries are not allowed

## Value

A matrix with estimated missing value.

## **Examples**

```
m<- data.frame(x1=c(50,NA,6,10,30),x2=c(2,8,NA,15,0)) emvd(m)
```

emvf

Estimation of missing value feature-based

## Description

This is a function in the dbnorm, a package in R. This function returns to a matrix of data in which missing values (Zero and/or NA values) are estimated. By this function, all Zero values are first replaced by NA values, which are then replaced by the lowest detected value on the column margin.

## Usage

emvf(m)

## Arguments

m

An array or a matrix

## **Details**

empty entries are not allowed

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#### Value

A matrix with estimated missing value.

#### **Examples**

```
m < - data.frame(x1=c(50,NA,6,10,30),x2=c(2,8,NA,15,0))
emvf(m)
```

hclustdbnorm

Hierarchical clustering analysis of original data and corrected data for batch effect It is a function in dbnorm, a package in R. This function allows users to evaluate dissimilarity between identical samples (quality control replicates or analytical replicates) analyzed in different batches, prior and after correction using, ber, ber\_bg and parametric and non-parametric ComBat . Pearson distance and average method for clustering were considered.

#### **Description**

Hierarchical clustering analysis of original data and corrected data for batch effect It is a function in dbnorm, a package in R. This function allows users to evaluate dissimilarity between identical samples (quality control replicates or analytical replicates) analyzed in different batches, prior and after correction using, ber, ber\_bg and parametric and non-parametric ComBat . Pearson distance and average method for clustering were considered.

## Usage

hclustdbnorm(m)

#### **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch levels in the first column.

## Details

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and /or emvd implemented in dbnorm package. Input must be normalized and transformed prior.

#### Value

Hierarchical clustering tree for original data (Raw data) and after correction, saved in a single **PDF** file in a working directory and series of **.csv** files includes distance values saved in temporary directory.

#### References

Johnson et al., (2007) < DOI:10.1093/biostatistics/kxj037 > http://www.ncbi.nlm.nih.gov/pubmed/16632515 Leek et al., (2012) < DOI:10.1093/bioinformatics/bts034> https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3307112/

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### **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5)),1)
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
hclustdbnorm(m)
## End(Not run)</pre>
```

ProfPlotBagging

Profile Plot of Features (variables) in corrected data using berbagging model

## **Description**

It is a function in the dbnorm, a package in R. This function allows users to remove variance associated with batch using "partial bagging" model with n=150 bootstrap samples (see also "ber\_bg" function and its package in R). This function visualize the result for global profile of each feature across batches via (*Scatter* plot), (*Violin* plot) and (*Density* (*or pdf*) plot).

#### Usage

ProfPlotBagging(m)

#### **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch in the first column.

## Details

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and/ or emvd implemented in the dbnorm. Input must be normalized and transformed prior.

#### Value

Original and adjusted datasets in **csv** format together with the series of profile plot for the variables( features) in the sample sets analyzed in the entire experiment provided by the *Scatter* plot, *Violin* plot and *pdf* plot compiled into **PDF** file.

## References

M.Giordan (2013) < DOI:10.1007/s12561-013-9081-1 > https://link.springer.com/article/10.1007/s12561-013-9081-1

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#### **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5),1))
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
ProfPlotBagging(m)
## End(Not run)</pre>
```

ProfPlotBer

Profile Plot of Features (variables) in ber- corrected data

## **Description**

It is a function in the dbnorm, a package in R. This function allows users to adjust the data for batch effect via location-scale (L/S) model (see also "ber" function and its package in R). This function visualize the result for global profile of each feature across batches via (*Scatter* plot), (*Violin* plot) and (*Density (or pdf)* plot).

#### **Usage**

ProfPlotBer(m)

## **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch in the first column.

#### **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and/ or emvd implemented in the dbnorm. Input must be normalized and transformed prior.

#### Value

Original and adjusted sets of data in **csv** format together with the series of profile plot for the variables(features) in the sample sets analyzed in the entire experiment provided by the *Scatter* plot, *Violin* plot and *pdf* plot compiled into **PDF** file.

## References

M.Giordan (2013) < DOI:10.1007/s12561-013-9081-1 > https://link.springer.com/article/10.1007/s12561-013-9081-1

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#### **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5),1))
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
profplotBer(m)
## End(Not run)</pre>
```

ProfPlotComNPara

Profile Plot of Features (variables) in corrected data via Non-Parametric ComBat

## **Description**

It is a function in the dbnorm, a package in R. This function allows users to adjust the data for batch effect using non-parametric *Empirical Bayes* approach (see "ComBat" in "sva", a package in bioconductor). *profplotpcom* visualize the result for global profile of each feature across batches via (*Scatter* plot), (*Violin* plot) and (*Density (or pdf)* plot).

## Usage

ProfPlotComNPara(m)

#### Arguments

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch in the first column.

## **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and/ or emvd implemented in the dbnorm. Input must be normalized and transformed prior.

#### Value

Original and adjusted datasets in **csv** format together with the series of profile plot of the features (variables) in the sample sets provided by *Scatter* plot, *Violin* plot and *pdf* plot compiled into a **PDF** file.

## References

Johnson et al., (2007) < DOI:10.1093/biostatistics/kxj037 > http://www.ncbi.nlm.nih.gov/pubmed/16632515 Leek et al., (2012) < DOI:10.1093/bioinformatics/bts034> https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3307112/

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#### **Examples**

```
## Not run:
batch<- rep(gl(2,3,labels=(1:2)),2)
y<- matrix(rnorm(6000), nrow=12)
m<- data.frame (batch,y)
ProfPlotComNPara(m)
## End(Not run)</pre>
```

ProfPlotComPara

Profile Plot of Features (variables) in corrected data via Parametric ComBat

## **Description**

It is a function in the dbnorm, a package in R. This function allows users to adjust the data for batch effect using parametric *Empirical Bayes* approach (see "ComBat" in "sva", a package in bioconductor). *profplotpcom* visualize the result for global profile of each feature across batches via (*Scatter* plot), (*Violin* plot) and (*Density (or pdf)* plot).

#### Usage

ProfPlotComPara(m)

#### **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch in the first column.

## **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by the functions such as emvf and/ or emvd implemented in the 'dbnorm'. Input must be normalized and transformed prior.

#### Value

Original and adjusted sets of data in **csv** format together with the series of profile plot for the features(variables) in the sample sets provided by the *Scatter* plot, *Violin* plot and *pdf* plot compiled into a **PDF** file.

## References

Johnson et al., (2007) < DOI:10.1093/biostatistics/kxj037 > http://www.ncbi.nlm.nih.gov/pubmed/16632515 Leek et al., (2012) < DOI:10.1093/bioinformatics/bts034> https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3307112/

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### **Examples**

```
## Not run:
batch<- rep(gl(2,3,labels=(1:2)),2)
y<- matrix(rnorm(6000), nrow=12)
m<- data.frame (batch,y)
ProfPlotComPara(m)
## End(Not run)</pre>
```

ProfPlotraw

Profile Plot of Features (variables) in original data (Raw data)

## **Description**

It is a function in the dbnorm This function informs you about the presence of across batch signal drift or batch effect in the raw data determined by the shifted probability density function plots (*pdf* plots) of features (variables) detected in an experiment.

## Usage

```
ProfPlotraw(m)
```

#### **Arguments**

m

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch level in the first column.

## **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by functions such as emvf or emvd Compiled in the 'dbnorm' package. Input must be normalized and transformed prior.

## Value

Original dataset in **csv** format together with the series of profile plot for the features (variables) in the sample sets analyzed in the entire experiment provided by the *Scatter* plot, *Violin* plot and *pdf* plot compiled into **PDF** file.

## **Examples**

```
## Not run:
batch<- rep(gl(5,10,labels = c(1:5),1))
y<- matrix(rnorm(5000),nrow=50)
m<-data.frame(batch,y)
ProfPlotraw(m)
## End(Not run)</pre>
```

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Visodbnorm

Visualization and normalization of signal drift across batches

#### **Description**

This function performs batch effect adjustment via three statistical models, namely two-stage procedure as described by M. Giordan (2013) < DOI:10.1007/s12561-013-9081-1> (see also "ber" and ber\_bg ) and/or empirical Bayes methods in two setting of parametric and non-parametric as described by Johnson et al., (2007) < DOI: 10.1093/biostatistics/kxj037 > (see "comBat" in "sva", a package in bioconductor). Meanwhile, the graphical inferences in the context of unsupervised learning algorithms create visual inspection to inform users about the spatial separation of the sample sets analyzed in the different analytical runs alongside the distribution of the features (variables) in the sample sets and across multiple batches. For bagging model, partial bagging model with n=150 bootstrap samples is considered.

## Usage

Visodbnorm(f)

#### Arguments

f

A data frame in which rows define the independent experiments (samples) and columns the features (variables), with the batch in the first column.

## **Details**

Zero and NA values are not allowed. Optionally missing value can be imputed by emvf and /or emvd, functions implemented in the dbnorm package. Input data must be normalized prior.

#### Value

Three datasets, adjusted by either of applied statistical algorithms prepared in **csv** and together with series of plot such as *PCA* plot and *Scree plot* compiled into a **PDF** file are saved in the working directory. *RLA* plots are represented in the **Viewer** panel of **rstudio**.

#### References

```
M.Giordan (2013) < DOI:10.1007/s12561-013-9081-1 > https://link.springer.com/article/10.1007/s12561-013-9081-1 
 Johnson et al., (2007) < DOI:10.1093/biostatistics/kxj037 > http://www.ncbi.nlm.nih.gov/pubmed/16632515
```

#### **Examples**

```
## Not run:
Visdbnorm
batch<- rep(gl(5,10,labels = c(1:5)),1)
y<- matrix(rnorm(5000),nrow=50)
f<-data.frame(batch,y)
Visodbnorm(f)
## End(Not run)</pre>
```

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