## nudge

## April 19, 2009

hiv

*Microarray intensity expression levels for HIV type 1 infection of* CD4+ T-Cell lines

## Description

The "hiv" section of this dataset consists of cDNA from CD4+ T cell lines at 1 hour after infection with HIV-1BRU. There are 4608 genes' expression levels. There are four replicates with balanced dye sways i.e. two of the replicates have sample 1 labeled with red and sample 2 labeled with green and the other two have the opposite labeling scheme. It is useful in testing the specificity and sensitivity of methods identifying differentially expressed genes, since there are 13 genes known to be differentially expressed (HIV-1 genes), called positive controls, identified in the vector of logical values "pos.contr" and 29 genes known not to be differentially expressed (non-human genes), called negative controls, identified in the vector of logical values "neg.contr".

## Usage

data(hiv)

## Format

A matrix containing 4608 observations on eight variables

## Source

Bumgarner Lab

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics, 6, 173-186.

## http://www.biomedcentral.com/1471-2105/6/173

A.B. van't Wout, G. K. Lehrma, S. A. Mikheeva, G. C. O'Keeffe, M. G. Katze, R. E. Bumgarner, G. K. Geiss and J. I. Mullins (2003). Cellular gene expression upon human immunodeficiency virus type 1 infection of CD4+ T-Cell lines. J. Virol. 77, 1392-1402.

#### See Also

pos.contr,neg.contr

#### like

## Description

This data set gives the expression levels for the same sample tested against itself, i.e. one sample split into two and labeled with both red and green dyes. Useful for testing false positive rate of methods for detecting differential expression since there are, theoretically no differentially expressed genes. There are 7680 different genes' expression levels measured.

## Usage

data(like)

## Format

A matrix containing 7680 observations on two variables

#### Source

Bumgarner Lab

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

neg.contr	Indicator of I
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Indicator of known truly non differentially expressed genes in the hiv dataset

#### Description

The "hiv" section of this dataset consists of cDNA from CD4+ T cell lines at 1 hour after infection with HIV-1BRU. There are 4608 genes' expression levels. There are four replicates with balanced dye sways i.e. two of the replicates have sample 1 labeled with red and sample 2 labeled with green and the other two have the opposite labeling scheme. It is useful in testing the specificity of methods identifying differentially expressed genes, since there are 29 genes known not to be differentially expressed (non-human genes), called negative controls, identified in the vector of logical values "neg.contr".

#### Usage

data(hiv)

## Format

A logical vector containing 4608 indicators of known non differential expression

#### norm1a

#### Source

Bumgarner Lab

#### References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

A.B. van't Wout, G. K. Lehrma, S. A. Mikheeva, G. C. O'Keeffe, M. G. Katze, R. E. Bumgarner, G. K. Geiss and J. I. Mullins (2003). Cellular gene expression upon human immunodeficiency virus type 1 infection of CD4+ T-Cell lines. J. Virol. 77, 1392-1402.

## See Also

pos.contr,hiv

normla

Function for normalizing the mean of single replicate log ratios

## Description

This is essentially the same as the lowess normalization suggested in the paper "Statistical methods for identifying differentially expressed genes in replicated cDNA microarrays" by Dudoit et al (2002), except the loess function is used instead of lowess and the recommended span is between 0.6 and 0.8. The normalization is done for each gene by subtracting from its log ratio the loess estimated mean for the log ratio based on the regression of log ratios on log intensities.

#### Usage

norm1a(logratio, logintensity, span = 0.6)

#### Arguments

logratio	A vector or single-column matrix of log (base 2) ratios of gene expressions in two samples.
logintensity	A vector or single-column matrix of log (base 2) total intensities (defined as the product) of gene expressions in the two samples.
span	Proportion of data used to fit the loess regression of the log ratios on the log intensities.

## Value

A vector or single-column matrix of mean normalized log ratios.

## Author(s)

N. Dean and A. E. Raftery

#### References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

S. Dudoit, Y. H. Yang, M. Callow and T. Speed (2002). Statistical methods for identifying differentially expressed genes in replicated cDNA microarray experiments. Stat. Sin. 12, 111-139.

## See Also

norm1b,norm1c,norm1d,norm2c,norm2d

#### Examples

```
data(like)
lR<-log(like[,1],2)-log(like[,2],2)
lI<-log(like[,1],2)+log(like[,2],2)</pre>
```

```
lRnorm<-norm1a(lR,lI)</pre>
```

norm1b

*Function for normalizing the mean and variance (or just the variance) of single replicate log ratios* 

## Description

This performs a robust loess normalization of the variance of the log ratios in a single replicate experiment by regressing the absolute (mean normalized) log ratios on the log intensities and using the fitted values to scale the (mean normalized) log ratio for each gene.

#### Usage

```
norm1b(logratio, logintensity, span1 = 0.6, span2 = 0.2, mean.norm=TRUE)
```

## Arguments

logratio	A vector or single-column matrix of log (base 2) ratios of gene expressions in two samples, if mean.norm is FALSE the log ratios should be already mean normalized.
logintensity	A vector or single-column matrix of log (base 2) total intensities (defined as the product) of gene expressions in the two samples.
spanl	Proportion of data used to fit the loess regression of the log ratios on the log intensities for the mean normalization.
span2	Proportion of data used to fit the loess regression of the absolute (mean normal- ized) log ratios on the log intensities for the variance normalization.
mean.norm	A logical value indicating whether or not a mean normalization should be per- formed prior to the variance normalization.

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#### norm1c

## Value

A vector or single-column matrix of mean and variance normalized log (base 2) ratios of gene expressions in two samples.

#### Author(s)

N. Dean and A. E. Raftery

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

## See Also

norm1a,norm1c,norm1d,norm2c,norm2d

## Examples

data(like)
lR<-log(like[,1],2)-log(like[,2],2)
lI<-log(like[,1],2)+log(like[,2],2)</pre>

lRnorm<-norm1b(lR,lI,mean.norm=TRUE)</pre>

norm1c

Function for normalizing the mean of average-across-replicates log ratios

#### Description

This is essentially the same as the lowess normalization suggested in the paper "Statistical methods for identifying differentially expressed genes in replicated cDNA microarrays" by Dudoit et al (2002), except the loess function and average-across-replicates log ratios are used and the recommended span is between 0.6 and 0.8. The normalization is done for each gene by subtracting from its average-across-replicates log ratio the loess estimated mean for average-across-replicates log ratio based on the loess regression of average-across-replicates log ratios on average-across-replicates log total intensities.

#### Usage

```
norm1c(logratio, logintensity, span = 0.6)
```

## Arguments

logratio	A multiple-column matrix of replicates of log (base 2) ratios of gene expressions
	in two samples.
logintensity	A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions in two samples.
span	Proportion of data used to fit the loess regression of the average-across-replicates log ratios on the average-across-replicates log total intensities

A vector of mean normalized average-across-replicates log ratios.

## Author(s)

N. Dean and A. E. Raftery

#### References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

S. Dudoit, Y. H. Yang, M. Callow and T. Speed (2002). Statistical methods for identifying differentially expressed genes in replicated cDNA microarray experiments. Stat. Sin. 12, 111-139.

## See Also

norm1d,norm1a,norm1b,norm2c,norm2d

#### Examples

```
data(hiv)
lR<-log(hiv[,1:4],2)-log(hiv[,5:8],2)
lI<-log(hiv[,1:4],2)+log(hiv[,5:8],2)</pre>
```

```
lRnorm<-norm1c(lR,lI)</pre>
```

norm1d

Function for normalizing the mean and variance of average-acrossreplicates log ratios

#### Description

This performs a robust normalization of the variance of the (mean normalized) average-acrossreplicates log ratios by scaling the (mean normalized) average-across-replicates log ratio for each gene either by the standard deviation of the log ratios for that gene across replicates (if bigger than its absolute (mean normalized) average-across-replicates log ratio) or scaling by a constant (a quantile of the distribution of standard deviations of log ratios across replicates for all genes whose standard deviation was bigger than their absolute (mean normalized) average-across-replicates log ratio.

## Usage

```
norm1d(logratio, logintensity, span = 0.6, quant = 0.99, dye.swap = FALSE)
```

#### norm1d

## Arguments

A multiple-column matrix of replicates of log (base 2) ratios of gene expressions in two samples.
A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions in two samples.
Proportion of data used to fit the loess regression of the average-across-replicates log ratios on the average-across-replicates log total intensities.
Quantile to be used from the distribution of standard deviations of log ratios across replicates for all genes whose standard deviation was smaller than their absolute (mean normalized) average-across-replicates log ratio.
A logical value indicating whether or not the data consists of a balanced dye swap (if FALSE a mean normalization will be performed prior to the variance normalization rather than a simple averaging across replicates).

## Details

A balanced dye swap is a data set were the data are split into two sets of replicates (same number of replicates in each set) where one set has the reverse dye assignment of the other set.

## Value

A vector of mean and variance normalized average-across-replicates log ratios.

#### Author(s)

N. Dean and A. E. Raftery

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

## See Also

norm1c,norm1a,norm1b,norm2c,norm2d

## Examples

```
data(hiv)
lR<-log(hiv[,1:4],2)-log(hiv[,5:8],2)
lI<-log(hiv[,1:4],2)+log(hiv[,5:8],2)</pre>
```

lRnorm<-norm1d(lR,lI,dye.swap=TRUE)</pre>

norm2c

Function for normalizing the mean of average-across-replicates log ratio differences

## Description

This normalization is used when the two samples (control and treatment, say) are not being directly compared on the slides but instead are being compared to a common reference sample. The quantity of interest for each gene is thus the average difference between control and treatment log ratios. The normalization is done for each gene by subtracting from its average-across-replicates log ratio difference the loess estimated mean for average-across-replicates log ratio difference based on the loess regression of average-across-replicates log ratio differences on average-across-replicates log total intensities.

## Usage

```
norm2c(control.logratio, txt.logratio, control.logintensity, txt.logintensity,
span = 0.6)
```

## Arguments

control.logra	atio	
	A multiple-column matrix of replicates of log (base 2) ratios of gene expressions for the control versus reference slides.	
txt.logratio	A multiple-column matrix of replicates of log (base 2) ratios of gene expressions for the treatment versus reference slides.	
control.logintensity		
	A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions for the control versus reference slides.	
txt.logintensity		
	A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions for the treatment versus reference slides.	
span	Proportion of data used to fit the loess regression of the average-across-replicates log ratio differences on the average-across-replicates log total intensities.	

#### Value

A vector of mean normalized average-across-replicates log ratio differences.

#### Author(s)

N. Dean and A. E. Raftery

#### References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

S. Dudoit, Y. H. Yang, M. Callow and T. Speed (2002). Statistical methods for identifying differentially expressed genes in replicated cDNA microarray experiments. Stat. Sin. 12, 111-139.

#### norm2d

#### See Also

norm2d,norm1a,norm1b,norm1c,norm1d

#### Examples

```
apo<-read.csv("http://www.stat.berkeley.edu/users/terry/zarray/Data/ApoA1/rg_alko_morph.t
header=TRUE)
rownames(apo)<-apo[,1]
apo<-apo[,-1]
apo<-apo+1
lRctl<-log(apo[,c(seq(2,16,2))],2)-log(apo[,c(seq(1,15,2))],2)
lRtxt<-log(apo[,c(seq(18,32,2))],2)-log(apo[,c(seq(17,31,2))],2)
lIctl<-log(apo[,c(seq(18,32,2))],2)+log(apo[,c(seq(17,31,2))],2)
lItxt<-log(apo[,c(seq(18,32,2))],2)+log(apo[,c(seq(17,31,2))],2)</pre>
```

lRnorm<-norm2c(lRctl,lRtxt,lIctl,lItxt)</pre>

norm2d

Function for normalizing the mean and variance of average-acrossreplicates log ratio differences

## Description

This normalization is used when the two samples (control and treatment, say) are not being directly compared on the slides but instead are being compared to a common reference sample. The quantity of interest for each gene is thus the average difference between control and treatment log ratios. This function performs a robust normalization of the variance of the (mean normalized) average-across-replicates log ratio differences by scaling the (mean normalized) average-across-replicates log ratio differences for each gene either by the standard deviation of the log ratio differences for that gene across replicates (if bigger than the absolute (mean normalized) average-across-replicates log ratio difference) or scaling by a constant (a quantile of the distribution of standard deviations of (mean normalized) average-across-replicates log ratio differences for all genes whose standard deviation was bigger than their absolute (mean normalized) average-across-replicates log ratio difference.

#### Usage

```
norm2d(control.logratio, txt.logratio, control.logintensity, txt.logintensity,
span = 0.6, quant = 0.99)
```

#### Arguments

control.logra	atio
	A multiple-column matrix of replicates of log (base 2) ratios of gene expressions for the control versus reference slides.
txt.logratio	A multiple-column matrix of replicates of log (base 2) ratios of gene expressions for the treatment versus reference slides.
control.logintensity	
	A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions for the control versus reference slides.

txt.loginten	sity
	A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions for the treatment versus reference slides.
span	Proportion of data used to fit the loess regression of the average-across-replicates log ratio differences on the average-across-replicates log intensities.
quant	Quantile to be used from the distribution of standard deviations of log ratio dif- ferences across replicates for all genes whose standard deviation was smaller than their absolute (mean normalized) average-across-replicates log ratio differ- ence.

### Value

A vector of mean and variance normalized average-across-replicates log ratio differences.

#### Author(s)

N. Dean and A. E. Raftery

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

S. Dudoit, Y. H. Yang, M. Callow and T. Speed (2002). Statistical methods for identifying differentially expressed genes in replicated cDNA microarray experiments. Stat. Sin. 12, 111-139.

#### See Also

norm2c,norm1a,norm1b,norm1c,norm1d

#### Examples

```
apo<-read.csv("http://www.stat.berkeley.edu/users/terry/zarray/Data/ApoA1/rg_a1ko_morph.t
header=TRUE)
rownames(apo)<-apo[,1]
apo<-apo[,-1]
apo<-apo+1</pre>
```

```
lRctl<-log(apo[, c(seq(2,16,2))],2)-log(apo[, c(seq(1,15,2))],2)
lRtxt<-log(apo[, c(seq(18,32,2))],2)-log(apo[, c(seq(17,31,2))],2)
lIctl<-log(apo[, c(seq(2,16,2))],2)+log(apo[, c(seq(1,15,2))],2)
lItxt<-log(apo[, c(seq(18,32,2))],2)+log(apo[, c(seq(17,31,2))],2)</pre>
```

```
lRnorm<-norm2d(lRctl,lRtxt,lIctl,lItxt)</pre>
```

nudge1	Function for normalizing data, fitting a normal-uniform mixture and estimating probabilities of differential expression in the case where the two samples are being compared directly

## Description

After a mean and variance normalization, a two component mixture model is fitted to the data. The normal component represents the genes that are not differentially expressed and the uniform component represents the genes that are differentially expressed. Posterior probabilities for differential expression are computed from the fitted model.

## Usage

```
nudge1(logratio, logintensity, dye.swap = FALSE, span1 = 0.6, span2 = 0.2,
quant = 0.99, z = NULL, tol = 0.00001,iterlim=500)
```

## Arguments

logratio	A matrix or vector of log (base 2) ratios of intensity expressions in 2 samples, with rows indexing genes and columns (if necessary) indexing replicates.
logintensity	A matrix or vector of total log (base 2) total intensities (defined as the product) of intensity expressions in 2 samples, with rows indexing genes and columns (if necessary) indexing replicates.
dye.swap	A logical value indicating whether or not the data is from a balanced dye-swap. Only used for multiple replicate experiments.
span1	Proportion of data used to fit the loess regression of the (average-across-replicates) log ratios on the (average-across-replicates) log total intensities for the mean normalization.
span2	Proportion of data used to fit the loess regression of the absolute (mean normal- ized) log ratios on the log total intensities for the variance normalization. Only used for single replicate experiments.
quant	Quantile to be used from the distribution of standard deviations of log ratios across replicates for all genes whose standard deviation was smaller than their absolute (mean normalized) average-across-replicates log ratio. Only used for multiple replicate experiments.
Z	An optional 2-column matrix with each row giving a starting estimate for the probability of the gene (in the corresponding row of the log ratio matrix/vector) not being differentially expressed and a starting estimate for the probability of the gene being differentially expressed. Each row should add up to 1.
tol	A scalar tolerance for relative convergence of the loglikelihood.
iterlim	The maximum number of iterations the EM is run for.

#### Details

A balanced dye swap is where a certain number of replicates have a particular dye to sample assigment and the same number of other replicates have the reversed assignment. Note in this case log ratios should be taken with numerators being the same sample and denominators the other sample, i.e. ratios should always be sample i/sample j rather than red dye/green dye for all replicates.

## Value

A list including the following components

pdiff	A vector with the estimated posterior probabilities of being in the group of dif- ferentially expressed genes.
lRnorm	A vector with the normalized (average-across-replicates) log ratios.
mu	The estimated mean of the group of genes that are not differentially expressed.
sigma	The estimated variance of the group of genes that are not differentially expressed.
mixprob	The prior/mixing probability of a gene being in the group of genes that are not differentially expressed.
a	The minimum value of the normalized data.
b	The maximum value of the normalized data.
loglike	The log likelihood for the fitted mixture model.
iter	The number of iterations run by the EM algorithm until either convergence or iteration limit was reached.

## Author(s)

N. Dean and A. E. Raftery

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

S. Dudoit, Y. H. Yang, M. Callow and T. Speed (2002). Statistical methods for identifying differentially expressed genes in replicated cDNA microarray experiments. Stat. Sin. 12, 111-139.

## See Also

nudge2,norm1a,norm1b,norm1c,norm1d,norm2c,norm2d

## Examples

```
data(like)
lR<-log(like[,1],2)-log(like[,2],2)
lI<-log(like[,1],2)+log(like[,2],2)</pre>
```

```
result<-nudge1(lR,lI)
```

data(hiv)
lR<-log(hiv[,1:4],2)-log(hiv[,5:8],2)
lI<-log(hiv[,1:4],2)+log(hiv[,5:8],2)</pre>

result<-nudge1(lR,lI,dye.swap=TRUE)</pre>

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Function for normalizing data, fitting a normal-uniform mixture and estimating probabilities of differential expression in the case where the two samples are being compared indirectly through a common reference sample

## Description

nudge2

After a mean and variance normalization a two component mixture model is fitted to the data. The normal component represents the genes that are not differentially expressed and the uniform component represents the genes that are differentially expressed. Posterior probabilities for differential expression are computed from the fitted model.

## Usage

nudge2(control.logratio, txt.logratio, control.logintensity, txt.logintensity, span1 = 0.2, quant = 0.99, z = NULL, tol = 0.00001,iterlim=500)

## Arguments

control.logra	atio
	A multiple-column matrix of replicates of log (base 2) ratios of gene expressions for the control versus reference slides.
txt.logratio	A multiple-column matrix of replicates of log (base 2) ratios of gene expressions for the treatment versus reference slides.
control.logir	ntensity
_	A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions for the control versus reference slides.
txt.logintens	sitv
5	A multiple-column matrix of replicates of log (base 2) total intensities (defined as the product) of gene expressions for the treatment versus reference slides.
spanl	Proportion of data used to fit the loess regression of the (average-across-replicates) log ratio differences on the (average-across-replicates) log intensities for the mean normalization.
quant	Quantile to be used from the distribution of standard deviations of log ratio dif- ferences across replicates for all genes whose standard deviation was smaller than their absolute (mean normalized) average-across-replicates log ratio differ- ence.
Z	An optional 2-column matrix with each row giving a starting estimate for the probability of the gene (in the corresponding row of the log ratio matrix/vector) not being differentially expressed and a starting estimate for the probability of the gene being differentially expressed. Each row should add up to 1.
tol	A scalar tolerance for relative convergence of the loglikelihood.
iterlim	The maximum number of iterations the EM is run for.

## Value

A list including the following components

pdiff	A vector with the estimated posterior probabilities of being in the group of dif- ferentially expressed genes.
lRnorm	A vector with the normalized (average-across-replicates) log ratio differences.
mu	The estimated mean of the group of genes that are not differentially expressed.
sigma	The estimated variance of the group of genes that are not differentially expressed.
mixprob	The prior/mixing probability of a gene being in the group of genes that are not differentially expressed.
a	The minimum value of the normalized data.
b	The maximum value of the normalized data.
loglike	The log likelihood for the fitted mixture model.
iter	The number of iterations run by the EM algorithm until either convergence or iteration limit was reached.

## Author(s)

N. Dean and A. E. Raftery

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

S. Dudoit, Y. H. Yang, M. Callow and T. Speed (2002). Statistical methods for identifying differentially expressed genes in replicated cDNA microarray experiments. Stat. Sin. 12, 111-139.

## See Also

nudge1,norm2c,norm2d,norm1a,norm1b,norm1c,norm1d

#### Examples

```
apo<-read.csv("http://www.stat.berkeley.edu/users/terry/zarray/Data/ApoA1/rg_a1ko_morph.t
header=TRUE)
rownames(apo)<-apo[,1]
apo<-apo[,-1]</pre>
```

```
apo<-apo+1
```

```
lRctl<-log(apo[,c(seq(2,16,2))],2)-log(apo[,c(seq(1,15,2))],2)
lRtxt<-log(apo[,c(seq(18,32,2))],2)-log(apo[,c(seq(17,31,2))],2)
lIctl<-log(apo[,c(seq(2,16,2))],2)+log(apo[,c(seq(1,15,2))],2)
lItxt<-log(apo[,c(seq(18,32,2))],2)+log(apo[,c(seq(17,31,2))],2)</pre>
```

```
result<-nudge2(lRctl,lRtxt,lIctl,lItxt)</pre>
```

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pos.contr

Indicator of known truly differentially expressed genes in the hiv dataset

## Description

The "hiv" section of this dataset consists of cDNA from CD4+ T cell lines at 1 hour after infection with HIV-1BRU. There are 4608 genes' expression levels. There are four replicates with balanced dye sways i.e. two of the replicates have sample 1 labeled with red and sample 2 labeled with green and the other two have the opposite labeling scheme. It is useful in testing the sensitivity of methods identifying differentially expressed genes, since there are 13 genes known to be differentially expressed (HIV-1 genes), called positive controls, identified in the vector of logical values "pos.contr"

## Usage

data(hiv)

#### Format

A logical vector containing 4608 indicators of known differential expression

## Source

Bumgarner Lab

## References

N. Dean and A. E. Raftery (2005). Normal uniform mixture differential gene expression detection for cDNA microarrays. BMC Bioinformatics. 6, 173-186.

http://www.biomedcentral.com/1471-2105/6/173

A.B. van't Wout, G. K. Lehrma, S. A. Mikheeva, G. C. O'Keeffe, M. G. Katze, R. E. Bumgarner, G. K. Geiss and J. I. Mullins (2003). Cellular gene expression upon human immunodeficiency virus type 1 infection of CD4+ T-Cell lines. J. Virol. 77, 1392-1402.

## See Also

hiv,neg.contr

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