

# Package ‘pickgene’

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**Title** Adaptive Gene Picking for Microarray Expression Data Analysis

**Description** Functions to Analyze Microarray (Gene Expression) Data.

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**License** GPL (>= 2)

**Imports** graphics, grDevices, MASS, stats, utils

**URL** <http://www.stat.wisc.edu/~yandell/statgen>

**biocViews** Microarray, DifferentialExpression

**NeedsCompilation** no

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em.ggb	<i>EM calculation for Gamma-Gamma-Bernoulli Model</i>
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## Description

The function plots contours for the odds that points on microarray show differential expression between two conditions (e.g. Cy3 and Cy5 dye channels on the same microarray).

**Usage**

```
em.ggb(x, y, theta, start = c(2,1.2,2.7), pprior = 2,
       printit = FALSE, tol = 1e-9, offset = 0 )
```

**Arguments**

x	first condition expression levels
y	second condition expression levels
theta	four parameters a, a0, nu, p
start	starting estimates for theta
pprior	Beta hyperparameter for prob p of differential expression
printit	print iterations if TRUE
tol	parameter tolerance for convergence
offset	offset added to xx and yy before taking log (can help with negative adjusted values)

**Details**

Fit Gamma/Gamma/Bernoulli model (equal marginal distributions) The model has spot intensities  $x \sim \text{Gamma}(a,b)$ ;  $y \sim \text{Gamma}(a,c)$ . The shape parameters b and c are  $\sim \text{Gamma}(a0,nu)$ . With probability p,  $b = c$ ; otherwise  $b \neq c$ . All spots are assumed to be independent.

**Value**

Four parameter vector theta after convergence.

**Author(s)**

Michael Newton

**References**

MA Newton, CM Kendzioriski, CS Richmond, FR Blattner and KW Tsui (2000) "On differential variability of expression ratios: improving statistical inference about gene expression changes from microarray data," *J Computational Biology* 00: 000-000.

**See Also**

[oddsplot](#)

**Examples**

```
## Not run:
em.ggb( x, y )

## End(Not run)
```

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model.pickgene	<i>Create Model Matrix for Orthogonal Contrasts</i>
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**Description**

The function created a model matrix of orthogonal contrasts to be used by pickgene.

**Usage**

```
model.pickgene(faclevel, facnames = letters[seq(length(faclevel))],
               contrasts.fac = "contr.poly", collapse = "+", show =
               NULL, renorm = 1, modeexpr = formula(paste("~",
               paste(facnames, collapse = collapse))),
               contrasts.list = contr.list)
```

**Arguments**

faclevel	vector with number of levels for each factor
facnames	vector of factor names (default = "a", "b", ...)
contrasts.fac	vector of contrast types
collapse	"+" for additive model, "*" for full model with interactions
show	vector of contrast numbers to show (default is all)
renorm	vector to renormalize contrasts (e.g., use $\sqrt{2}$ to turn two-condition contrast into fold change)
modeexpr	model formula
contrasts.list	list of contrasts indexed by facnames

**Details**

Creates a model matrix data frame with first column having all 1's and other columns having contrasts.

**Value**

Result of call to model.matrix

**Author(s)**

Brian Yandell

**See Also**

[model.matrix](#)

**Examples**

```
model.pickgene(c(2,3), c("sex", "genotype"))
```

oddsplot

*Odds Plot for Differential Microarray Expression***Description**

The function plots contours for the odds that points on microarray show differential expression between two conditions (e.g. Cy3 and Cy5 dye channels on the same microarray).

**Usage**

```
oddsplot(x, y, theta, by.level = 10, rotate = FALSE, offset =
         0, main = "", xlab = xlabs, ylab = ylabs, col = NULL,
         cex = c(0.25, 0.75), shrink = FALSE, lims =
         range(c(x, y)))
```

**Arguments**

x	first condition expression levels
y	second condition expression levels
theta	four parameters from em.ggb
by.level	odds plot contours increase by this level
rotate	rotate to average versus ratio if TRUE, otherwise plot conditions against each other
offset	offset for log transform
main	main title for plot
xlab	horizontal axis label (default if Cy3 if rotate is FALSE, Average Intensity otherwise)
ylab	vertical axis label (default if Cy5 if rotate is FALSE, Cy3 / Cy5 otherwise)
col	color of points (if NULL, use black for non-changing points, blue for changing points)
cex	character expansion (use rep(.25, 2) to have all points the same size)
shrink	use shrinkage on expression levels if TRUE (default is FALSE)
lims	limits for plot area

**Details**

Fit Gamma/Gamma/Bernoulli model (equal marginal distributions) The model has spot intensities  $x \sim \text{Gamma}(a,b)$ ;  $y \sim \text{Gamma}(a,c)$ . The shape parameters  $b$  and  $c$  are  $\sim \text{Gamma}(a_0, \nu)$ . With probability  $p$ ,  $b = c$ ; otherwise  $b \neq c$ . All spots are assumed to be independent.

**Value**

Log odds for all points in original order.

**Author(s)**

Michael Newton

**References**

MA Newton, CM Kendzierski, CS Richmond, FR Blattner and KW Tsui (2000) "On differential variability of expression ratios: improving statistical inference about gene expression changes from microarray data," *J Computational Biology* 00: 000-000.

**See Also**

[em.ggb](#)

**Examples**

```
## Not run:  
oddsplot( x, y )  
  
## End(Not run)
```

---

pickgene

*Plot and Pick Genes based on Differential Expression*

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**Description**

The function picks plots the average intensity versus linear contrasts (currently linear, quadratic up to cubic) across experimental conditions. Critical line is determine according to Bonferroni-like multiple comparisons, allowing SD to vary with intensity.

**Usage**

```
pickgene(data, geneID = 1:nrow(data), overalllevel = 0.05,  
         npickgene = -1, marginal = FALSE, rankbased = TRUE,  
         allrank = FALSE, meanrank = FALSE, offset = 0,  
         modelmatrix = model.pickgene(faclevel, facnames,  
         contrasts.fac, collapse, show, renorm), faclevel =  
         ncol(data), facnames =  
         letters[seq(length(faclevel))], contrasts.fac =  
         "contr.poly", show = NULL, main = "", renorm = 1,  
         drop.negative = FALSE, plotit = npickgene < 1, mfrow  
         = c(nr, nc), mfcol = NULL, ylab = paste(shownames,  
         "Trend"), ...)
```

**Arguments**

<code>data</code>	data matrix
<code>geneID</code>	gene identifier (default <code>1:nrow(x)</code> )
<code>overalllevel</code>	overall significance level (default <code>0.05</code> )
<code>npickgene</code>	number of genes to pick (default <code>-1</code> allows automatic selection)
<code>marginal</code>	additive model if TRUE, include interactions if FALSE
<code>rankbased</code>	use ranks if TRUE, log transform if FALSE
<code>allrank</code>	rank all chips together if true, otherwise rank separately
<code>meanrank</code>	show mean abundance as rank if TRUE
<code>offset</code>	offset for log transform
<code>modelmatrix</code>	model matrix with first row all 1's and other rows corresponding to design contrasts; automatically created by call to <code>model.pickgene</code> if omitted
<code>faclevel</code>	number of factor levels for each factor
<code>facnames</code>	factor names
<code>contrasts.fac</code>	type of contrasts
<code>show</code>	vector of contrast numbers to show (default is all)
<code>main</code>	vector of main titles for plots (default is none)
<code>renorm</code>	vector to renormalize contrasts (e.g. use <code>sqrt(2)</code> to turn two-condition contrast into fold change)
<code>drop.negative</code>	drop negative values in log transform
<code>plotit</code>	plot if TRUE
<code>mfrow</code>	<code>par()</code> plot arrangement by rows (default up to 6 per page; set to NULL to not change)
<code>mfcol</code>	<code>par()</code> plot arrangement by columns (default is NULL)
<code>ylab</code>	vertical axis labels
<code>...</code>	parameters for <code>robustscale</code>

**Details**

Infer genes that differentially express across conditions using a robust data-driven method. Adjusted gene expression levels  $A$  are replaced by `qnorm(rank(A))`, followed by `robustscale` estimation of center and spread. Then Bonferroni-style gene by gene tests are performed and displayed graphically.

**Value**

Data frame containing significant genes with the following information:

<code>pick</code>	data frame with picked genes
<code>score</code>	data frame with center and spread for plotting

Each of these is a list with elements for each contrast. The `pick` data frame elements have the following information:

probe	gene identifier
average	average gene intensity
fold1	positive fold change
fold2	negative fold change
pvalue	Bonferroni-corrected p-value

The score data frame elements have the following:

x	mean expression level (antilog scale)
y	contrast (antilog scale)
center	center for contrast
scale	scale (spread) for contrast
lower	lower test limit
upper	upper test limit

### Author(s)

Yi Lin and Brian Yandell

### References

Y Lin, BS Yandell and ST Nadler (2000) "Robust Data-Driven Inference for Gene Expression Microarray Experiments," Technical Report, Department of Statistics, UW-Madison.

### See Also

[pickgene](#)

### Examples

```
## Not run:  
pickgene( data )  
  
## End(Not run)
```

---

robustscale

*Robust Estimation of Median (center) and MAD (scale)*

---

### Description

Smoothing spline estimate of median and mean absolute deviation (MAD).

### Usage

```
robustscale(y, x, nslice=400, corcenter=TRUE, decrease=TRUE)
```

**Arguments**

y	response
x	predictor
nslice	number of slices (should be "large")
corcenter	correct for center
decrease	force MAD to decrease with x

**Details**

This divides data into roughly many `nslice` slices and computes median and mean absolute deviation (`mad`) for each slice. These are then smoothed using `smooth.spline`.

**Value**

Data frame containing significant genes with the following information:

center	estimate of center median
scale	MAD estimate of scale
x	ordered x values for plotting
y	y sorted by x

**Author(s)**

Yi Lin

**See Also**

[mad](#), [smooth.spline](#)

**Examples**

```
## Not run:  
robustscale(y,x)  
  
## End(Not run)
```

---

Simulation.pickgene    *Yi Lin's simulations for microarray analysis*

---

**Description**

Example simulations

**See Also**

[multipickgene](#)

**Examples**

```

### Note: This uses old pickgene
#detail of the model (7-8). (first run does not include meas error \eta_i)
#par(mfrow=c(3,3))
t<-rnorm(10000,4,2)
changes1<-rep(0,10000)
changes1[1:500]<-rnorm(500)
t1<-t+changes1
changes2<-rep(0,10000)
changes2[1:500]<-rnorm(500)
t2<-t+changes2
s<-rnorm(10000,0,0.1)
cx<-3
cy<-2
t1<-t1+rnorm(10000,0,0.1)
t2<-t2+rnorm(10000,0,0.1)
x<-cx*exp(t1)
y<-cy*exp(t2)
#x<-cx*exp(t1)+rnorm(10000,0,50)
#y<-cy*exp(t2)+rnorm(10000,0,40)
xx<-qnorm(rank(x)/(10000+1))
yy<-qnorm(rank(y)/(10000+1))
#hist(x,breaks=100)
#hist(y,breaks=100)
#plot(x,y)
#hist(y[x<=0],breaks=20)
#hist(x[y<=0],breaks=20)
#plot(xx,yy)
topgenepick<-multipickgene( cbind(xx,yy),condi=0:1,geneID=1:10000, d=1,
                           npickgene=500)$pick[[1]]$probe

abchangesrank<-rank((-1)*abs(t1-t2))
count <- rep(NA,500)
for( i in 1:500 ) {
  topipick <- topgenepick[1:i]
  count[i] <- sum( abchangesrank[topipick] <= i )
}

## Figure 2
plot( 1:500, 1:500, type="n",
      xlab="Rank of 500 most changed genes by our procedure",
      ylab="Number similarly ranked by the 'optimal' procedure",
      xaxs="i", yaxs="i" )
lines( 1:500, count, type="s", lty=1, lwd=2 )
abline(0,1)
## Not run: dev.print( hor=F, height=6.5, width=6.5, file="rank1.ps" )

#again, but with the additive noise. (includes \eta_i)
par(mfrow=c(2,2))
t<-rnorm(10000,4,2)
changes1<-rep(0,10000)
changes1[1:500]<-rnorm(500)
t1<-t+changes1

```

```

changes2<-rep(0,10000)
changes2[1:500]<-rnorm(500)
t2<-t+changes2
s<-rnorm(10000,0,0.1)
cx<-3
cy<-2
t1<-t1+rnorm(10000,0,0.1)
t2<-t2+rnorm(10000,0,0.1)
### note that noise is very large here (50,40)
x<-cx*exp(t1)+rnorm(10000,0,50)
y<-cy*exp(t2)+rnorm(10000,0,40)
xx<-qnorm(rank(x)/(10000+1))
yy<-qnorm(rank(y)/(10000+1))
hist(x,breaks=100)
hist(y,breaks=100)
plot(x,y,cex=0.4)
#hist(y[x<=0],breaks=20)
#hist(x[y<=0],breaks=20)
plot(xx,yy,cex=0.4)
## Not run: dev.print( hor=F, height=6.5, width=6.5, file="simudata.ps" )

topgenepick<-multipickgene(cbind(xx,yy),condi=0:1,geneID=1:10000, d=1,
                           npickgene=500)$pick[[1]]$probe
abchangesrank<-rank((-1)*abs(t1-t2))
count <- rep(NA,500)
for( i in 1:500 ) {
  topipick <- topgenepick[1:i]
  count[i] <- sum( abchangesrank[topipick] <= i )
}
par(mfrow=c(1,1)) # figure 4
plot( 1:500, 1:500, type="n",
      xlab="Rank of 500 most changed genes by our procedure",
      ylab="Number similarly ranked by the 'optimal' procedure",
      xaxs="i", yaxs="i" )
lines( 1:500, count, type="s", lty=1, lwd=2 )
abline(0,1)
## Not run: dev.print( hor=F, height=6.5, width=6.5, file="rank2.ps" )

### Figure 5
genepick <- multipickgene( cbind(xx,yy), condi=0:1, geneID=1:10000, d=1)
## Not run: dev.print( hor=F, height=6.5, width=6.5, file="simutest.ps" )$pick[[1]]$probe
npick<-length(genepick$pickedgene)
genepick$pickedgene
npick
count[npick]

```

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