# Package 'Clomial'

July 15, 2025

Type Package

Title Infers clonal composition of a tumor

Version 1.45.0
<b>Date</b> 2018-05-31
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<b>Depends</b> R (>= 2.10), matrixStats
Imports methods, permute
<b>Description</b> Clomial fits binomial distributions to counts obtained from Next Gen Sequencing data of multiple samples of the same tumor. The trained parameters can be interpreted to infer the clonal structure of the tumor.
License GPL (>= 2)
biocViews Genetics, GeneticVariability, Sequencing, Clustering, MultipleComparison, Bayesian, DNASeq, ExomeSeq, TargetedResequencing, ImmunoOncology
LazyLoad yes
git_url https://git.bioconductor.org/packages/Clomial
git_branch devel
git_last_commit 484d465
git_last_commit_date 2025-04-15
Repository Bioconductor 3.22
Date/Publication 2025-07-15
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Clomial-package

Fits a binomial model to data from multiple samples of a single tumor to infer its clonal decomposition.

### **Description**

Clomial fits binomial distributions to counts obtained from Next Gen Sequencing data of multiple samples of the same tumor. The trained parameters can be interpreted to infer the clonal structure of the tumor.

### **Details**

Package: Clomial
Type: Package
Version: 0.99.0
Date: 2014-02-11
License: GPL (>= 2)

The main function is Clomial() which requires 2 matrices Dt and Dc among its inputs. They contain the counts of the alternative allele, and the total number of processed reads, accordingly. Their rows correspond to the genomic loci, and their columns correspond to the samples. Several models should be trained using different initial values to escape from local optima, and the best one in terms of the likelihood can be chosen by choose.best() function.

### Author(s)

Habil Zare and Alex Hu

Maintainer: Habil Zare <zare@u.washington.edu>

#### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., PLoS Computational Biology 10.7 (2014): e1003703.

### See Also

Clomial, choose.best, Clomial.iterate, Clomial.likelihood, compute.bic, breastCancer

```
set.seed(1)
data(breastCancer)
Dc <- breastCancer$Dc
Dt <- breastCancer$Dt
ClomialResult <-Clomial(Dc=Dc,Dt=Dt,maxIt=20,C=4,doParal=FALSE,binomTryNum=2)
chosen <- choose.best(models=ClomialResult$models)
M1 <- chosen$bestModel
print("Genotypes:")
print(round(M1$Mu))
print("Clone frequencies:")
print(M1$P)</pre>
```

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breastCancer

Breast cancer data for clonal decomposition.

### **Description**

Counts data from multiple samples of a single primary breast cancer obtained by deep, next-generation sequencing. The file is consist of two matrices Dt and Dc which contain the counts of the alternative alleles, and the total number of counts on each genomic loci for every tumor samples, accordingly.

### Usage

```
data(breastCancer)
```

#### **Format**

A list containing 2 matrices.

### **Details**

Each matrix contains counts of reads mapped to 17 genomic loci for 12 tumor samples where the column A5-2 corresponds to the normal sample.

#### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

### See Also

Clomial

#### **Examples**

data(breastCancer)
breastCancer\$Dt

choose.best

Chooses the best trained Clomial model.

### Description

Given the output of Clomial function, the likelihoods of all models are compared, and the best model is determined.

### Usage

```
choose.best(models, U = NULL, PTrue = NULL, compareTo = NULL, upto =
"All", doTalk=FALSE)
```

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### **Arguments**

models The models trained by Clomial function.

U The optional genotype matrix used for comparison.

PTrue The optional clone frequency matrix used for comparison.

compareTo The index of the model against which all other models are compared. Set to

NULL to disable.

upto The models with index less than this value are considered. Set to "All" to include

every model.

doTalk If TRUE, information on number of analyzed models is reported.

#### **Details**

If compareTo, U, and PTrue are NULL no comparison will be done, and the function runs considerably faster.

#### Value

A list will be made with the following entries:

err A list with 2 entries; err\$P and err\$U the vectors of clonal frequency errors, and

genotype errors, accordingly.

Li A vector of the best obtained log-likelihood for each model.

bestInd The index of the best model in terms of log-likelihood.

comparison If compareTo is not NULL, the result of comparison with the corresponding

model is reported.

bestModel The best model in terms of log-likelihood.

seconds A vector of the time taken, in seconds, to train each model.

### Note

When the number of assumed clones, C, is greater than 6, the comparison will be time taking because all possible permutations of clones should be considered. The running time will be slowed down by C!.

### Author(s)

Habil Zare

### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

### See Also

Clomial, Clomial.likelihood, Clomial.iterate

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### **Examples**

```
set.seed(4)
data(breastCancer)
Dc <- breastCancer$Dc
Dt <- breastCancer$Dt
ClomialResult <-Clomial(Dc=Dc,Dt=Dt,maxIt=20,C=4,doParal=FALSE,binomTryNum=5)
chosen <- choose.best(models=ClomialResult$models)
M1 <- chosen$bestModel
print("Genotypes:")
round(M1$Mu)
print("Clone frequencies:")
M1$P
bestInd <- chosen$bestInd
plot(chosen$Li,ylab="Log-likelihood",type="1")
points(x=bestInd,y=chosen$Li[bestInd],col="red",pch=19)</pre>
```

Clomial

Fits several binomial models to data from multiple samples of a single tumor.

### Description

Using EM, trains several models using different initial values to escape from local optima. The best one in terms of the likelihood can be later chosen by choose.best() function.

### Usage

```
Clomial(Dt = NULL, Dc = NULL, DcDtFile = NULL, C, doParal=FALSE,
outPrefix = NULL, binomTryNum = 1000, maxIt = 100, llCutoff = 0.001,
jobNamePrefix = "Bi", qstatWait = 2, fitBinomJobFile = NULL,
jobShare = 10, ignoredSample = c(), fliProb=0.05, conservative=TRUE,
doTalk=FALSE)
```

### **Arguments**

Dt	A matrix which contains the counts of the alternative allele where rows correspond to the genomic loci, and columns correspond to the samples.
Dc	A matrix which contains the counts of the total number of mapped reads where rows correspond to the genomic loci, and columns correspond to the samples.
DcDtFile	A file from which the data can optionally be loaded. It should contain the matrices Dc and Dt.
С	The assumed number of clones.
doParal	Boolean where TRUE means, in Linux, models with different initialization are trained in parallel on a cluster using qsub.
outPrefix	A prefix for the path to save the results.
binomTryNum	The number of models trained using different initialization.
maxIt	The maximum number of EM iterations.
11Cutoff	EM iterations stops if the relative improvement in the log-likelihood is not more than this threshold.

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jobNamePrefix If run in parallel, this prefix will be used to name the jobs on the cluster.

qstatWait The waiting time between qstat commands to assess the number of running and

waiting jobs.

fitBinomJobFile

If run in parallel, this is the script which loads data, trains a model using a

random initialization, and saves the results.

jobShare If run in parallel, the job\_share option of qsub determines the priority of jobs

over other submitted jobs.

ignoredSample A vector of indices of samples which will be ignored in training. Used by experts

only to measure the stability of the results.

fliProb A "flipping probability" used for noise injection which can be disabled when

fliProb=0. After the first EM iteration, each entry of the matrix Mu such as m may change to 1-m with this probability. This probability decreases on subse-

quent iterations.

conservative Boolean where TRUE means noise will be injected only if likelihood is im-

proved after an EM iteration, otherwise the original Mu matrix will be used for

the next iteration. For expert use only.

doTalk If TRUE, information on the EM optimization iterations is reported.

#### **Details**

The likelihood of the model, given the hidden variables and the parameters, can be computed based on a combination of binomial distributions. In each EM iteration, the likelihood is increased, however, due to presence of local optima, several models should be tried using different random initialization. For higher number of assumed clones, C, the parameter binomTryNum should be increased because the dimension of the search space grows linearly with C.

### Value

Returns a list containing the entry called models, which is a list of the length equal to binomTryNum where each element is a trained model. For each trained model, Mu models the matrix of genotypes, where rows and columns correspond to genomic loci and clones, accordingly. Also, P is the matrix of clonal frequency where rows and columns correspond to clones and samples, accordingly. The first column of P corresponds to the normal clone. The history of Mu, P, and the log-likelihood over iterations is saved in lists Ps, Mus, and Likelihoods, accordingly.

#### Note

The parallel mode works only in Linux, and when qsub and qstat commands are available on a cluster.

### Author(s)

Habil Zare

#### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

### See Also

Clomial, choose.best, Clomial.iterate, compute.bic, breastCancer

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#### **Examples**

```
set.seed(1)
data(breastCancer)
Dc <- breastCancer$Dc
Dt <- breastCancer$Dt
ClomialResult <-Clomial(Dc=Dc,Dt=Dt,maxIt=20,C=4,binomTryNum=2)
chosen <- choose.best(models=ClomialResult$models)
M1 <- chosen$bestModel
print("Genotypes:")
round(M1$Mu)
print("Clone frequencies:")
M1$P</pre>
```

Clomial.generate.data Generates simulated data to test performance of Clomial algorithm.

### **Description**

Data sets are simulated based on binomial distribution using random parameters for the model. The accuracy of the EM procedure can be estimated by comparing the inferred parameters vs. the known ones which were used to generate the data.

### Usage

```
Clomial.generate.data(N, C, S, averageCoverage, mutFraction,
doSample1Normal = FALSE,erroRate=0,doCheckDc=TRUE)
```

### **Arguments**

N The number of genomic loci.

C The number of clones.
S The number of samples.

averageCoverage

The average coverage over each loci, each sample.

mutFraction Should be in range 0-1. Each loci in every sample can be mutated with this

probability.

doSample1Normal

If TRUE, no contamination with the tumor content is allowed for the normal sample. I.e. the first column of the generated P matrix will start with 1, and the

rest of its entries will be equal to 0.

erroRate The sequencing noise can be simulated by assigning a positive value to this

parameter, which is the probability of reading a normal allele as the alternative

allele, and vica versa.

doCheckDc If TRUE, generating with be repeated until no row of Dc is all zeros to guarantee

all loci have positive coverage in at least one sample.

### **Details**

See the reference below for details.

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#### Value

A list will be made with the following entries:

Dc A matrix of simulated coverage for all loci and samples.

Dt A matrix of alternative allele counts for all loci and samples.

Ptrue The true clone frequency matrix used for generating the data.

U The true genotype matrix used for generating the data.

Likelihood The log-likelihood of the model with the true parameters.

Phi The matrix of the second parameters of the binomial distributions; each entry is

the probability that a read contains the variant allele at a locus in a sample.

### Author(s)

Habil Zare

#### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

### See Also

```
Clomial, Clomial.likelihood
```

### **Examples**

```
set.seed(1)
simulated <- Clomial.generate.data(N=20, C=4, S=10,
   averageCoverage=1000, mutFraction=0.1)
simulated$Dc</pre>
```

Clomial.iterate

Runs EM iterations until convergence of the Clomial model.

### **Description**

Given the data and the initial values for the model parameters, runs EM iterations until convergence of the Clomial model.

### Usage

```
Clomial.iterate(Dt, Dc, Mu, P, maxIt=100, U = NULL, PTrue = NULL, llCutoff = 10^(-3), computePFunction = compute.P.reparam, doSilentOptim = TRUE, doTalk = TRUE, doLog = TRUE, debug = FALSE, noiseReductionRate = 0.01, fliProb=0.05,conservative=TRUE)
```

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### **Arguments**

maxIt The maximum number of EM iterations.

Dt A matrix which contains the counts of the alternative allele where rows corre-

spond to the genomic loci, and columns correspond to the samples.

Dc A matrix which contains the counts of the total number of mapped reads where

rows correspond to the genomic loci, and columns correspond to the samples.

Mu The initial value for the Mu matrix which models the genotypes, where rows

and columns correspond to genomic loci and clones, accordingly.

P The initial matrix of clonal frequency where rows and columns correspond to

clones and samples, accordingly.

U The true value for Mu, used for debugging purposes only.

PTrue The true value for P, used for debugging purposes only.

11Cutoff EM iterations stops if the relative improvement in the log-likelihood is not more

than this threshold.

computePFunction

The function used for updating P. For advanced development use only.

doSilentOptim If TRUE, the optimization massages will not be reported.

doTalk If FALSE, the function will be run in silent mode.

doLog Highly recommended to set to TRUE. Then, the computations will be done in

log space to avoid numerical issues.

debug If TRUE, the debug mode will be turned on.

noiseReductionRate

The noise will be reduce by this rate after each EM iteration.

fliProb A "flipping probability" used for noise injection which can be disabled when

fliProb=0. After the first EM iteration, each entry of the matrix Mu such as m may change to 1-m with this probability. This probability decreases on subse-

quent iterations.

conservative Boolean where TRUE means noise will be injected only if likelihood is im-

proved after an EM iteration, otherwise the original Mu matrix will be used for

the next iteration. For expert use only.

### **Details**

Injecting noise can be done by assigning a positive value to fliProb, and can be disabled by fliProb=0. Noise injection is recommended for training models with a high number of clones (>4).

#### Value

A list will be made with the following entries:

Qs The history of matrices containing the posterior Q values.

Ps The history of P matrices.

Mus The history of Mu matrices.

Mu The value of Mu after convergence.

P The value of P after convergence.

11Cutoff The threshold used to decide convergence.

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LRatio The final relative improvement in the log likelihood which lead to convergence.

Likelihoods The history of log-likelihoods.

fliProb The final value of fliProb used for noise injection.

timeTaken An object of class "difftime" which reports the total computational time for EM iterations.

endTaken An object of class "POSIXct" (see DateTimeClasses) which reports the time EM

iterations finished.

### Author(s)

Habil Zare

#### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

#### See Also

```
Clomial, Clomial, breastCancer
```

```
set.seed(1)
## Getting data:
data(breastCancer)
Dc <- breastCancer$Dc</pre>
Dt <- breastCancer$Dt</pre>
freq1 <- Dt/Dc</pre>
N <- nrow(Dc)
S <- ncol(Dc)
Cnum <- 4 ## assumed number of clones.
## Random initialization:
random1 <- runif(n=N*(Cnum-1),min=rowMins(freq1)*0.9,max=rowMaxs(freq1)*1.1)</pre>
random1[random1>1] <- 1</pre>
random1[random1<0] <- 0</pre>
Mu <- matrix(random1,N,Cnum-1)</pre>
Mu <- cbind( matrix(0,N,1), Mu )
rownames(Mu) <- rownames(Dc)</pre>
colnames(Mu) <- paste("C",1:Cnum,sep="")</pre>
P <- matrix(runif(Cnum*S),Cnum,S)</pre>
rownames(P) <- colnames(Mu)</pre>
colnames(P) <- colnames(Dc)</pre>
## Normalizing P:
for( t in 1:S){
 s \leftarrow sum(P[,t])
P[,t] \leftarrow P[,t]/s
}##End for.
## Running EM:
model1 <- Clomial.iterate(Dt=Dt, Dc=Dc, Mu=Mu, P=P)</pre>
print("Genotypes:")
round(model1$Mu)
print("Clone frequencies:")
model1$P
```

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Clomial.likelihood	Computes the complete data log-likelihood of a Clomial model.
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### **Description**

Computes the expected complete data log-likelihood of a Clomial model over all possible values of the hidden variables.

### Usage

```
Clomial.likelihood(Dc, Dt, Mu, P)
```

### **Arguments**

Dt	A matrix which contains the counts of the alternative allele where rows correspond to the genomic loci, and columns correspond to the samples.
Dc	A matrix which contains the counts of the total number of mapped reads where rows correspond to the genomic loci, and columns correspond to the samples.
Mu	The matrix which models the genotypes, where rows and columns correspond to genomic loci and clones, accordingly.
Р	The matrix of clonal frequency where rows and columns correspond to clones and samples, accordingly.

#### **Details**

By assuming that the genomic loci and the samples are independent given the model parameters, the computation is simplified by first summing over the samples for a locus, and then summing over all the loci. This strategy avoids exploring the exponentially huge probability space.

### Value

A list will be made with the following entries:

11 The expectation of complete log-likelihood over the hidden variables.

11S A vector of computed log-likelihoods at all loci.

### Note

The likelihood is computed assuming the heterozygosity is 2.

### Author(s)

Habil Zare

### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

### See Also

```
Clomial, choose.best, compute.bic, breastCancer
```

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### **Examples**

```
set.seed(1)
data(breastCancer)
Dc <- breastCancer$Dc
Dt <- breastCancer$Dt
ClomialResult <-Clomial(Dc=Dc,Dt=Dt,maxIt=20,C=4,doParal=FALSE,binomTryNum=1)
model1 <- ClomialResult$models[[1]]
likelihood <- Clomial.likelihood(Dc=Dc, Dt=Dt, Mu=model1$Mu, P=model1$P)$11
print(likelihood)</pre>
```

Clomial1000

Pre-computed results of Clomial.

### **Description**

Pre-computed results of Clomial function are provided for demo purposes. It contains 1000 trained models on counts data from multiple samples of a single primary breast cancer obtained by deep, next-generation sequencing.

### Usage

```
data(Clomial1000)
```

#### **Format**

Clomial1000[["models"]] is the list of trained models.

### Details

Each model is the output of Clomial.iterate() function on the breastCancer data assuming there are 4 clones.

#### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

#### See Also

```
Clomial, Clomial.iterate, choose.best, breastCancer
```

```
data(Clomial1000)
chosen <- choose.best(models=Clomial1000$models)
M1 <- chosen$bestModel
print("Genotypes:")
round(M1$Mu)
print("Clone frequencies:")
M1$P
bestInd <- chosen$bestInd
plot(chosen$Li,ylab="Log-likelihood",type="l")
points(x=bestInd,y=chosen$Li[bestInd],col="red",pch=19)</pre>
```

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### **Description**

Computes the Bayesian Information Criterion (BIC) for a Clomial model, which might be useful to estimate the number of clones. A "significantly" smaller BIC is usually interpreted as a better fit to the data.

### Usage

```
compute.bic(Dc, Dt, Mu, P)
```

#### **Arguments**

Dt	A matrix which contains the counts of the alternative allele where rows correspond to the genomic loci, and columns correspond to the samples.
Dc	A matrix which contains the counts of the total number of mapped reads where rows correspond to the genomic loci, and columns correspond to the samples.
Mu	The matrix which models the genotypes, where rows and columns correspond to genomic loci and clones, accordingly.
Р	The matrix of clonal frequency where rows and columns correspond to clones and samples, accordingly.

### **Details**

The Bayesian Information Criterion (BIC) for a model is computed by subtracting the expected log-likelihood times 2, from the number of free parameters of the model times logarithm of the total number of observations. For a Clomial model, we have BIC = (NC+SC-S)log(sum(Dc))-2L, where L is the likelihood, N is the number of genomic loci, C is the assumed number of clones, S is the number of samples, and sum(Dc) is the total number of observed reads.

### Value

A list will be made with the following entries:

bic The BIC value. aic The AIC value.

obsNum The total number of observed reades.

#### Note

Theoretically, a method such as the Bayesian information criterion (BIC) or the Akaike information criterion (AIC) may be applied to estimate the number of clones. However, in practice, the outcome of such approaches should be interpreted with great caution because some of the underlying assumptions of the statistical analysis may not be necessarily true for a given model. For example, while a "small" improvement in the BIC is generally considered as a sign to stop making the model more complicated, making such decisions is very objective, and requires relying on thresholds with little statistical basis.

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### Author(s)

Habil Zare

### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

### See Also

Clomial

### **Examples**

```
set.seed(1)
data(breastCancer)
Dc <- breastCancer$Dc
Dt <- breastCancer$Dt
bics <- c()
Clomial3 <-Clomial(Dc=Dc,Dt=Dt,maxIt=20,C=3,doParal=FALSE,binomTryNum=1)
model3 <- Clomial3$models[[1]]
bics[3] <- compute.bic(Dc=Dc,Dt=Dt, Mu=model3$Mu, P=model3$P)$bic
Clomial4 <-Clomial(Dc=Dc,Dt=Dt,maxIt=20,C=4,doParal=FALSE,binomTryNum=1)
model4 <- Clomial4$models[[1]]
bics[4] <- compute.bic(Dc=Dc,Dt=Dt, Mu=model4$Mu, P=model4$P)$bic
print(bics) ## 4 is a better estimate for the number of clones.</pre>
```

compute.errors

Computes the error of a Clomial model.

### **Description**

Given the true genotype and frequency matrices, finds the permutation of genotypes matrix which best matches the true genotypes and returns the corresponding errors.

### Usage

```
compute.errors(Mu, U, P, PTrue)
```

### **Arguments**

Mu	The matrix which models the genotypes, where rows and columns correspond to genomic loci and clones, accordingly.
U	The true genotype matrix defined similar to Mu.
Р	The matrix of clonal frequency where rows and columns correspond to clones and samples, accordingly.
PTrue	The true clonal frequency matrix defined similar to P.

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#### **Details**

Computing the error is useful for estimating the performance of inference on simulated, and for comparing different trained models. Genotype and frequency errors are defined as the normalized 11-error in reconstructing the genotype, and the clone frequency matrices, accordingly, where by normalized 11-error we mean the sum of absolute values of an error matrix divided by the size of the matrix.

#### Value

A list will be made with the following entries:

UError The 11-error of the genotype matrix normalized by the size of matrix.

discretizedUError

The 11-error of the rounded genotype matrix, i.e. the number of mismatching genotypes, normalized by the size of matrix

PErrorAbsolute The normalized 11-error of the clone frequency matrix.

PErrorRelative Each entry of the error clone frequency matrix is normalized by the corresponding entry in PTrue, and then the normalized 11 norm is computed.

#### Note

The use of UError and PErrorAbsolute is recommended. Computing the error is not feasible for more than 7 clones because the number of all possible permutations is factorial in the number of clones which grows super fast. Such input will trigger an error message.

### Author(s)

Habil Zare

### References

Inferring clonal composition from multiple sections of a breast cancer, Zare et al., Submitted.

### See Also

Clomial

```
set.seed(1)
data(breastCancer)
Dc <- breastCancer$Dc
Dt <- breastCancer$Dt
bics <- c()
ClomialResult <-Clomial(Dc=Dc,Dt=Dt,maxIt=20,C=3,doParal=FALSE,binomTryNum=2)
model1 <- ClomialResult$models[[1]]
model2 <- ClomialResult$models[[2]]
## Comparing 2 trained models:
compute.errors(Mu=model1$Mu,U=model2$Mu,P=model1$P,PTrue=model2$P)</pre>
```

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