Package 'SCOPE'

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Type Package

Title A normalization and copy number estimation method for single-cell DNA sequencing

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Author Rujin Wang, Danyu Lin, Yuchaojiang

Maintainer Rujin Wang <rujin@email.unc.edu>

Description Whole genome single-cell DNA sequencing (scDNA-seq) enables characterization of copy number profiles at the cellular level. This circumvents the averaging effects associated with bulk-tissue sequencing and has increased resolution yet decreased ambiguity in deconvolving cancer subclones and elucidating cancer evolutionary history. ScDNA-seq data is, however, sparse, noisy, and highly variable even within a homogeneous cell population, due to the biases and artifacts that are introduced during the library preparation and sequencing procedure. Here, we propose SCOPE, a normalization and copy number estimation method for scDNAseq data. The distinguishing features of SCOPE include: (i) utilization of cell-specific Gini coefficients for quality controls and for identification of normal/diploid cells, which are further used as negative control samples in a Poisson latent factor model for normalization; (ii) modeling of GC content bias using an expectation-maximization algorithm embedded in the Poisson generalized linear models, which accounts for the different copy number states along the genome; (iii) a cross-sample iterative segmentation procedure to identify breakpoints that are shared across cells from the same genetic background. We evaluate performance of SCOPE on real scDNA-seq data sets from cancer genomic studies. Compared to existing methods, SCOPE more accurately estimates subclonal copy number aberrations and is shown to have higher correlation with array-based copy number profiles of purified bulk samples from the same patient. We further demonstrate SCOPE on three recently released data sets using the 10X Genomics single-cell CNV pipeline and show that it can reliably recover 1% of the cancer cells from a background of normal.

Depends R (>= 3.4.3)

Imports CODEX2, Rsamtools, GenomicRanges, IRanges, stats, GenomeInfoDb, BSgenome.Hsapiens.UCSC.hg19, graphics, utils

License GPL-2 LazyData true RoxygenNote 6.1.1 Encoding UTF-8

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Estep.Pois

Expectation step for SCOPE normalization

Description

Expectation step of GC content fitting for SCOPE normalization.

Usage

```
Estep.Pois(Yj, Nj, betatemp, fGCi, vec_pi, min.prop)
```

Arguments

Yj read depth vector for each single cell

Nj a numerire value of total number of reads per cell

betatemp bin-specific bias estimated using negative control samples

fGCi estimated vector of GC content bias fitting from the M-step Mstep. Pois

vec_pi vector of incident rates for CNV events that span bin *i* from the M-step Mstep. Pois min.prop the minmimum of mixture proportion for candidate CNV groups, which serves

as a stopping metric for EM algorithm

Value

A list with components

Z Matrix of optimized missing data to be fed into M-step Mstep. Pois

obs_LL Observed log-Likelihood

keep_going If normal proportion or a mixture proportion reached the minimum value, keep_going = FALSE,

and output NULL, aka this case won't be involved in optimal number of CNV

group selection based upon BIC.

Author(s)

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mataguaras asDNA	Cat no ad a quanta a from six als a all DNA as a quanting	
getcoverage.scDNA	Get read coverage from single-cell DNA sequencing	

Description

Get read coverage for each genomic bin across all single cells from scDNA-seq.

Usage

```
getcoverage.scDNA(bambedObj, mapqthres, mask.ref, seq)
```

Arguments

bambedObj object returned from getbambed
mapqthres mapping quality threshold of reads

mask.ref a GRanges object indicating bad regions/bins, such as segmental duplication

regions and gaps near telomeres/centromeres, which need to be masked prior to

getting coverage

seq the sequencing method to be used. This should be either "paired-end" or "single-

end"

Value

Y Read depth matrix

Author(s)

Rujin Wang <rujin@email.unc.edu>

getfGC.Pois.random Get GC content bias fitting using Expectation-Maximization algorithm

Description

Fit a Poisson generalized linear model to normalize the raw read depth data from single-cell DNA sequencing, without latent factors under the case-control setting. SCOPE implements an EM algorithm with random initialization to unmask null regions. This is a bulit-in function within multi_run.Pois. We don't recommend running this function independently.

Usage

```
getfGC.Pois.random(gcfitj, gctemp, Yj, Nj, betatemp, numGroup,
  verbose.plot = FALSE, gctemp.keep, min.prop)
```

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Arguments

gcfitj vector of bin-specific GC content biases for each single cell

gctemp a vector giving values of bin-specific GC content after gcfitj outlier removal

(for better fitting)

Yj read depth vector for each single cell

Nj a numerire value of total number of reads per cell

betatemp bin-specific bias estimated using negative control samples

numGroup a vector of integers indicating number of CNV groups. Use BIC to select optimal

number of CNV groups. If numGroup = 1, assume all reads are from normal regions so that EM algorithm is not implemented. Otherwise, we assume there is always a CNV group of heterozygous deletion and a group of null region. The

rest groups are representative of different duplication states.

verbose.plot logical, whether to plot GC content bias fitting results using EM and the choice

of optimal CNV group. Default is FALSE

gctemp.keep a vector giving values of bin-specific GC content without gcfitj outlier re-

moval. For better fitting, extreme GC content biases need to be excluded from

EM input. But a post hoc admission of all bins is necessary.

min.prop the minmimum of mixture proportion for candidate CNV groups, which serves

as a stopping metric for EM algorithm

Value

A list with components

BIC BIC for CNV group selection logL Observed log-Likelihood

fGCi EM estimated vector of GC content bias fitting given gctemp

Z Matrix of optimized missing data using EM algorithm

vec_pi Vector of EM estimated incident rate for CNV events that span bin *i*

K Choice of optimal CNV group number based upon BIC

fGCi.keep EM estimated vector of GC content bias fitting given gctemp.keep

Author(s)

Rujin Wang <rujin@email.unc.edu>

getmapp	Compute mappability
getillapp	сотрые таррадии

Description

Compute mappability for each bin. Note that scDNA sequencing is whole-genome amplification and the mappability score is essential to determine variable binning method. Mappability track for 100-mers on the GRCh37/hg19 human reference genome from ENCODE is pre-saved. Compute the mean of mappability scores that overlapped reads map to bins, weighted by the width of mappability tracks on the genome reference. Use liftOver utility to calculate mappability for hg38, which is presaved as well.

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Usage

```
getmapp(ref, genome = NULL)
```

Arguments

ref GRanges object returned from getbambed

genome by default, genome = BSgenome. Hsapiens. UCSC. hg19. To calculate mappa-

bility for hg38, specify genome = BSgenome.Hsapiens.UCSC.hg38

Value

mapp Vector of mappability for each bin/target

Author(s)

Rujin Wang <rujin@email.unc.edu>

getsampQC

Get QC metrics for single cells

Description

Perform QC step on single cells.

Usage

```
getsampQC(bambedObj)
```

Arguments

bambedObj object returned from getbambed

Value

QCmetric A matrix containing total number/proportion of reads, total number/proportion

of mapped reads, total number/proportion of mapped non-duplicate reads, and

number/proportion of reads with mapping quality greater than 20

Author(s)

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logsumexp

Logarithm of summation of exponentials

Description

Computes the logarithm of summation of numeric exponentials.

Usage

```
logsumexp(xx)
```

Arguments

XX

a numeric value or vector

Value

A numeric value giving natural logarithm of summation of exponentials

Author(s)

Rujin Wang <rujin@email.unc.edu>

mapp_hg19

GRanges with mappability scores for hg19

Description

GRanges object specifying target positions with mappabilities across the whole genome.

Usage

```
data(mapp_hg19)
```

Format

A GRanges object with 21591667 ranges and 1 metadata column of mappability scores

Details

GRanges of mappability track for 100-mers on the GRCh37/hg19 human reference genome from ENCODE.

Value

GRanges object with mappabilities for hg19

Author(s)

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References

http://rohsdb.cmb.usc.edu/GBshape/cgi-bin/hgFileUi?db=hg19&g=wgEncodeMapability

mapp_hg38

GRanges with mappability scores for hg38

Description

GRanges object specifying target positions with mappabilities across the whole genome.

Usage

```
data(mapp_hg38)
```

Format

A GRanges object with 21584930 ranges and 1 metadata column of mappability scores

Details

Use liftOver utility to convert hg19 coordinates to hg38

Value

GRanges object with mappabilities for hg38

Author(s)

```
Rujin Wang <rujin@email.unc.edu>
```

References

http://hgdownload.cse.ucsc.edu/goldenPath/hg19/liftOver/

Mstep.Pois

Maximization step for SCOPE normalization

Description

Maximization step of GC content fitting for SCOPE normalization.

Usage

```
Mstep.Pois(Z, gcfitj, gctemp)
```

Arguments

Z	matrix of optimized missing data from the E-step Estep.Pois
gcfitj	vector of bin-specific GC content biases for each single cell
gctemp	a vector giving values of GC content for each bin after quality control

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Value

A list with components

vec_pi Vector of incident rate for CNV events that span bin i, which is to be fed into

E-step Estep. Pois

fGCi Estimated vector of GC content bias fitting

Author(s)

Rujin Wang <rujin@email.unc.edu>

multi_run.Pois Get the optimal GC content bias fitting using Expectation-

Maximization algorithm

Description

SCOPE implements an EM algorithm with random initialization to unmask null regions. Adopt BIC to choose optimal GC content bias fitting among multiple runs of getfGC.Pois.random. This is a bulit-in function within normalize_codex2_ns_noK_EM_random.

Usage

```
multi_run.Pois(gcfitj, gctemp, Yj, Nj, betatemp, numGroup, rerun,
  verbose.plot = FALSE, qc.thres = 5e-05, min.prop = 0.002)
```

Arguments

gcfitj vector of bin-specific GC content biases for each single cell

gctemp a vector giving values of bin-specific GC content after gcfitj outlier removal

(for better fitting)

Yj read depth vector for each single cell

Nj a numerire value of total number of reads per cell

betatemp bin-specific bias estimated using negative control samples

numGroup a vector of integers indicating number of CNV groups. Use BIC to select optimal

number of CNV groups. If numGroup = 1, assume all reads are from normal regions so that EM algorithm is not implemented. Otherwise, we assume there is always a CNV group of heterozygous deletion and a group of null region. The

rest groups are representative of different duplication states.

rerun specify the number of running EM algorithm with random initialization

verbose.plot logical, whether to plot the optimal GC content bias fitting results using EM and

the choice of optimal CNV group. Default is FALSE

qc. thres the lower bound of bin-specific GC content bias threshold

min.prop the minmimum of mixture proportion for candidate CNV groups, which serves

as a stopping metric for EM algorithm

Value

A list with components for the optimal GC content bias fitting using EM

BIC BIC for CNV group selection logL Observed log-Likelihood

fGCi EM estimated vector of GC content bias fitting given gctemp

Z Matrix of optimized missing data using EM algorithm

vec_pi Vector of EM estimated incident rate for CNV events that span bin i

K Choice of optimal CNV group number based upon BIC

fGCi.keep EM estimated vector of GC content bias fitting given gctemp.keep

Author(s)

Rujin Wang <rujin@email.unc.edu>

normalize_codex2_ns_noK

Normalization of read depth without latent factors under the casecontrol setting

Description

Assuming that all reads are from diploid regions, fit a Poisson generalized linear model to normalize the raw read depth data from single-cell DNA sequencing, without latent factors under the case-control setting.

Usage

```
normalize_codex2_ns_noK(Y_qc, gc_qc, K = 1, norm_index, N)
```

Arguments

Y_qc read depth matrix after quality control

gc_qc vector of GC content for each bin after quality control

K default is K = 1

norm_index indices of normal/diploid cells

N library size factor, which is computed from the genome-wide read depth data

Value

A list with components

Yhat A list of normalized read depth matrix

GC.hat A list of estimated GC content bias matrix

beta.hat A list of estimated bin-specific bias vector

Author(s)

```
normalize_codex2_ns_noK_EM_random
```

Normalization of read depth without latent factors using Expectation-Maximization algorithm under the case-control setting

Description

Fit a Poisson generalized linear model to normalize the raw read depth data from single-cell DNA sequencing, without latent factors under the case-control setting. Model GC content bias using an expectation-maximization algorithm, which accounts for the different copy number states.

Usage

```
normalize_codex2_ns_noK_EM_random(Y_qc, gc_qc, K = 1, norm_index, N,
numGroup, qc.thres = 5e-05, min.prop = 0.002)
```

Arguments

Y_qc read depth matrix after quality control

gc_qc vector of GC content for each bin after quality control

K default is K = 1

norm_index indices of normal/diploid cells

N library size factor, which is computed from the genome-wide read depth data

numGroup a vector of integers indicating number of CNV groups. Use BIC to select optimal

number of CNV groups. If numGroup = 1, assume all reads are from normal regions so that EM algorithm is not implemented. Otherwise, we assume there is always a CNV group of heterozygous deletion and a group of null region. The

rest groups are representative of different duplication states.

qc. thres the lower bound of bin-specific GC content bias threshold

min.prop the minmimum of mixture proportion for candidate CNV groups, which serves

as a stopping metric for EM algorithm

Value

A list with components

Yhat A list of normalized read depth matrix with EM fGC.hat A list of EM estimated GC content bias matrix beta.hat A list of EM estimated bin-specific bias vector

Author(s)

segmentCBScs 11

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Description

SCOPE offers a cross-sample Poisson likelihood-based recursive segmentation, enabling shared breakpoints across cells from the same genetic background.

Usage

```
segmentCBScs(Y, Yhat, sampname, ref, lmax, mode, segment.CODEX2 = FALSE)
```

Arguments

Y raw read depth matrix after quality control procedure

Yhat normalized read depth matrix

sampname vector of sample names

ref GRanges object after quality control procedure

1max maximum CNV length in number of bins returned

mode format of returned copy numbers, which can be either "integer" or "fraction".

"integer" is recommended for scDNA-seq data.

segment.CODEX2 logical, whether to perform individual segmentation. Default is FALSE.

Value

A list with components

poolcall Cross-sample CNV callings indicating shared breakpoints

finalcall Final cross-sample segmented callset of CNVs with genotyping results

finalcall_CODEX2

Final individual segmented callset of CNVs with genotyping results

image.orig A matrix giving logarithm of normalized z-scores

image.seg A matrix of logarithm of estimated copy number over 2

Author(s)

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