

# Package ‘spp’

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**Title** ChIP-Seq Processing Pipeline

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**Author** Peter K

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**Suggests** methods

**LinkingTo** Rcpp, BH (>= 1.66)

**OS\_type** unix

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**Description** Analysis of ChIP-seq and other functional sequencing data [Kharchenko PV (2008) <DOI:10.1038/nbt.1508>].

**License** GPL-2

**LazyLoad** yes

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spp-package*ChIP-seq (Solexa) Processing Pipeline*

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**Description**

A set of routines for reading short sequence alignments, calculating tag density, estimates of statistically significant enrichment/depletion along the chromosome, identifying point binding positions (peaks), and characterizing saturation properties related to sequencing depth.

**Details**

Package:	spp
Type:	Package
Version:	1.11
Date:	2012-06-20
License:	What license is it under?
LazyLoad:	yes

See example below for typical processing sequence.y

**Author(s)**

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**References**

Kharchenko P., Tolstorukov M., Park P. "Design and analysis of ChIP-seq experiments for DNA-binding proteins." Nature Biotech. doi:10.1038/nbt.1508

**Examples**

```
## Not run:
# load the library
library(spp);

## The following section shows how to initialize a cluster of 8 nodes for parallel processing
## To enable parallel processing, uncomment the next three lines,
##and comment out "cluster<-NULL";
## see "snow" package manual for details.
#library(snow)
#cluster <- makeCluster(2);
#invisible(clusterCall(cluster,source,"routines.r"));
cluster <- NULL;

# read in tag alignments
chip.data <- read.eland.tags("chip.eland.alignment");
input.data <- read.eland.tags("input.eland.alignment");

# get binding info from cross-correlation profile
# strange gives the possible range for the size of the protected region;
# strange should be higher than tag length; making the upper
#boundary too high will increase calculation time
#
# bin - bin tags within the specified number of basepairs to speed up calculation;
# increasing bin size decreases the accuracy of the determined parameters
binding.characteristics <- get.binding.characteristics(chip.data,
  srangle=c(50,500),
  bin=5,
  cluster=cluster);

# plot cross-correlation profile
pdf(file="example.crosscorrelation.pdf",width=5,height=5)
par(mar = c(3.5,3.5,1.0,0.5), mgp = c(2,0.65,0), cex = 0.8);
plot(binding.characteristics$cross.correlation,
  type='l',
  xlab="strand shift",
  ylab="cross-correlation");
abline(v=binding.characteristics$peak$x,lty=2,col=2)
dev.off();
```

```

# select informative tags based on the binding characteristics
chip.data <- select.informative.tags(chip.data,binding.characteristics);
input.data <- select.informative.tags(input.data,binding.characteristics);

# restrict or remove positions with anomalous number of tags relative
# to the local density
chip.data <- remove.local.tag.anomalies(chip.data);
input.data <- remove.local.tag.anomalies(input.data);

# output smoothed tag density (subtracting re-scaled input) into a WIG file
# note that the tags are shifted by half of the peak separation distance
smoothed.density <- get.smoothed.tag.density(chip.data,
                                               control.tags=input.data,
                                               bandwidth=200,
                                               step=100,
                                               tag.shift=round(binding.characteristics$peak$x/2));

writewig(smoothed.density,
          "example.density.wig","Example smoothed, background-subtracted tag density");

rm(smoothed.density);

# output conservative enrichment estimates
# alpha specifies significance level at which confidence intervals will be estimated
enrichment.estimates <- get.conservative.fold.enrichment.profile(chip.data,
                                                               input.data,
                                                               fws=2*binding.characteristics$whs,
                                                               step=100,
                                                               alpha=0.01);

writewig(enrichment.estimates,
          "example.enrichment.estimates.wig",
          "Example conservative fold-enrichment/depletion estimates shown on log2 scale");
rm(enrichment.estimates);

# binding detection parameters
# desired FDR. Alternatively, an E-value can be supplied to the
#method calls below instead of the fdr parameter
fdr <- 1e-2;
# the binding.characteristics contains the optimized half-size for binding detection window
detection.window.halfsize <- binding.characteristics$whs;

# determine binding positions using wtd method
bp <- find.binding.positions(signal.data=chip.data,
                               control.data=input.data,
                               fdr=fdr,
                               method=tag.wtd,
                               whs=detection.window.halfsize,
                               cluster=cluster)

# alternatively determined binding positions using lwcc

```

```
# method (note: this takes longer than wtd)
# bp <- find.binding.positions(signal.data=chip.data,control.data=input.data,
# fdr=fdr,method=tag.lwcc,whs=detection.window.halfsize,cluster=cluster)

print(paste("detected",sum(unlist(lapply(bp$npl,function(d) length(d$x))))),"peaks"));

# output detected binding positions
output.binding.results(bp,"example.binding.positions.txt");

# -----
# the set of commands in the following section illustrates methods for saturation analysis
# these are separated from the previous section, since they are highly CPU intensive
# -----

# determine MSER
# note: this will take approximately 10-15x the amount of time the initial
# binding detection did
# The saturation criteria here is 0.99 consistency in the set of binding
# positions when adding 1e5 tags.
# To ensure convergence the number of subsampled chains (n.chains) should be higher (80)
mser <- get.mser(chip.data,
  input.data,
  step.size=1e5,
  test.agreement=0.99,
  n.chains=8,
  cluster=cluster,
  fdr=fdr,
  method=tag.wtd,
  whs=detection.window.halfsize)

print(paste("MSER at a current depth is",mser));

# note: an MSER value of 1 or very near one implies that the set of
# detected binding positions satisfies saturation criteria without
# additional selection by fold-enrichment ratios. In other words,
# the dataset has reached saturation in a traditional sense (absolute saturation).

# interpolate MSER dependency on tag count
# note: this requires considerably more calculations than the previous
# steps (~ 3x more than the first MSER calculation)
# Here we interpolate MSER dependency to determine a point at which MSER of 2 is reached
# The interpolation will be based on the difference in MSER at the
# current depth, and a depth at 5e5 fewer tags (n.steps=6);
# evaluation of the intermediate points is omitted here to
# speed up the calculation (excluded.steps parameter)
# A total of 7 chains is used here to speed up calculation,
# whereas a higher number of chains (50) would give good convergence
msers <- get.mser.interpolation(chip.data,
  input.data,
  step.size=1e5,
  test.agreement=0.99,
  target.fold.enrichment=2,
```

```

n.chains=7,
n.steps=6,
excluded.steps=c(2:4),
cluster=cluster,
fdr=fdr,
method=tag.wtd,
whs=detection.window.halfsize)

print(paste("predicted sequencing depth =",
round(unlist(lapply(msers,function(x) x$prediction))/1e6,5)," million tags"))

# note: the interpolation will return NA prediction if the dataset
# has reached absolute saturation at the current depth.
# note: use return.chains=T to also calculate random chains
# returned under msers$chains field) - these can be passed back as
# "get.mser.interpolation( ..., chains=msers$chains)" to calculate
# predictions for another target.fold.enrichment value
# without having to recalculate the random chain predictions.

## stop cluster if it was initialized
#stopCluster(cluster);

## End(Not run)

```

**add.broad.peak.regions***Calculate chromosome-wide profiles of smoothed tag density***Description**

Looks for broader regions of enrichment associated with the determined peak positions, adds them to the \$npl data as \$rs, \$re columns.

**Usage**

```
add.broad.peak.regions(chip.tags,input.tags,bp,window.size=500,z.thr=2)
```

**Arguments**

chip.tags	signal chromosome tag coordinate vectors (e.g. output of <a href="#">select.informative.tags</a> )
input.tags	optionall control (input) tags
bp	output of find.binding.positions call
window.size	window size to be used in calculating enrichment
z.thr	Z-score corresponding to the Poisson ratio threshold used to flag significantly enriched windows

**Value**

A structure identical to bp (binding.positions) with two additional columns added (rs and re) corresponding to start and end of the associated significantly enriched region. If no region was associated with a particular peak, NAs values are reported.

---

densum	<i>Do Something</i>
--------	---------------------

---

**Description**

Densum

**Usage**

```
densum(vin, bw = 5, dw = 3, match.wt.f = NULL, return.x = T,
       from = min(vin), to = max(vin), step = 1, new.code = T)
```

**Arguments**

vin	Parameter
bw	Parameter
dw	Parameter
match.wt.f	Parameter
return.x	Parameter
from	Parameter
to	Parameter
step	Parameter
new.code	Parameter

**Value**

Some sum

**Examples**

```
## Not run:
##### Should be DIRECTLY executable !! -----
### ==> Define data, use random,
###-or do help(data=index) for the standard data sets.

## The function is currently defined as
function (vin, bw = 5, dw = 3, match.wt.f = NULL, return.x = T,
          from = min(vin), to = max(vin), step = 1, new.code = T)
{
  tc <- table(vin[vin >= from & vin <= to])
```

```

pos <- as.numeric(names(tc))
storage.mode(pos) <- "double"
tc <- as.numeric(tc)
storage.mode(tc) <- "double"
n <- length(pos)
if (!is.null(match.wt.f)) {
  tc <- tc * match.wt.f(pos)
}
rng <- c(from, to)
if (rng[1] < 0) {
  stop("range extends into negative values")
}
if (range(pos)[1] < 0) {
  stop("position vector contains negative values")
}
storage.mode(n) <- storage.mode(rng) <- storage.mode(bw)
  <- storage.mode(dw) <- storage.mode(step) <- "integer"
spos <- rng[1]
storage.mode(spos) <- "double"
dlength <- floor((rng[2] - rng[1])/step) + 1
if (dlength < 1) {
  stop("zero data range")
}
if (new.code) {
  storage.mode(step) <- storage.mode(dlength)
    <- storage.mode(bw) <- storage.mode(dw) <- "integer"
  dout <- .Call("ccdensum", pos, tc, spos, bw, dw, dlength,
    step)
}
else {
  dout <- numeric(dlength)
  storage.mode(dout) <- "double"
  storage.mode(dlength) <- "integer"
  .C("cdensum", n, pos, tc, spos, bw, dw, dlength, step,
    dout)
}
if (return.x) {
  return(list(x = c(rng[1], rng[1] + step * (dlength -
    1)), y = dout, step = step))
}
else {
  return(dout)
}
}

## End(Not run)

```

## Description

Given the signal and optional control (input) data, determine location of the statistically significant point binding positions. If the control data is not provided, the statistical significance can be assessed based on tag randomization. The method also provides options for masking regions exhibiting strong signals within the control data.

## Usage

```
find.binding.positions(signal.data, f=1,e.value = NULL, fdr = NULL, masked.data = NULL,
control.data = NULL, whs = 200, min.dist = 200, window.size = 4e+07, cluster = NULL,
debug = T, n.randomizations = 3, shuffle.window = 1, min.thr = 2, topN = NULL,
tag.count.whs = 100, enrichment.z = 2, method = tag.wtd, tec.filter = T,
tec.window.size = 10000, tec.z = 5, tec.masking.window.size=tec.window.size,
tec.poisson.z=5,tec.poisson.ratio=5, tec = NULL, n.control.samples = 1,
enrichment.scale.down.control =F, enrichment.background.scales = c(1, 5, 10),
use.randomized.controls = F, background.density.scaling = T,
mle.filter = F,min.mle.threshold = 1, ...)
```

## Arguments

	~~ tag data ~~
	signal tag vector list
<del>signal.olddata</del>	optional control (input) tag vector list
f	Fraction of signal read subsampled. Default=1, i.e. no subsampling ~ position stringency criteria ~~
e.value	E-value defining the desired statistical significance of binding positions.
fdr	FDR defining statistical significance of binding positions
topN	instead of determining statistical significance thresholds, return the specified number of highest-scoring positions ~~ other params ~~
whs	window half-sized that should be used for binding detection (e.g. determined from cross-correlation profiles)
masked.data	optional set of coordinates that should be masked (e.g. known non-unique regions)
min.dist	minimal distance that must separate detected binding positions. In case multiple binding positions are detected within such distance, the position with the highest score is returned.
window.size	size of the window used to segment the chromosome during calculations to reduce memory usage.
cluster	optional snow cluster to parallelize the processing on
debug	debug mode, whether to print debug messages
min.thr	minimal score requirement for a peak

**background.density.scaling**  
 If TRUE, regions of significant tag enrichment will be masked out when calculating size ratio of the signal to control datasets (to estimate ratio of the background tag density). If FALSE, the dataset ratio will be equal to the ratio of the number of tags in each dataset.

**tec** tec  
**n.control.samples** n.control.samples  
**enrichment.scale.down.control** enrichment.scale.down.control  
**...** additional parameters should be the same as those passed to the `lwcc.prediction`  
 ~~ randomized controls ~~

**n.randomizations** number of tag randomizations that should be performed (when the control data is not provided)

**use.randomized.controls** Use randomized tag control, even if `control.data` is supplied.

**shuffle.window** during tag randomizations, tags will be split into groups of `shuffle.window` and will be maintained together throughout the randomization.  
 ~~ fold-enrichment confidence intervals ~~

**tag.count.whs** half-size of a window used to assess fold enrichment of a binding position

**enrichment.z** Z-score used to define the significance level of the fold-enrichment confidence intervals

**enrichment.background.scales** In estimating the peak fold-enrichment confidence intervals, the background tag density is estimated based on windows with half-sizes of  $2 \times \text{tag.count.whs} \times \text{enrichment.background.scales}$

**method** either `tag.wtd` for WTD method, or `tag.lwcc` for MTC method

**mle.filter** If turned on, will exclude predicted positions whose MLE enrichment ratio (for any of the background scales) is below a specified `min.mle.threshold`

**min.mle.threshold** MLE enrichment ratio threshold that each predicted position must exceed if `mle.filter` is turned on.  
 ~~ masking regions of significant control enrichment ~~

**tec.filter** Whether to mask out the regions exhibiting significant enrichment in the control data in doing other calculations. The regions are identified using Poisson statistics within sliding windows, either relative to the scaled signal (`tec.z`), or relative to randomly-distributed expectation (`tec.poisson.z`).

**tec.window.size** size of the window used to determine significantly enriched control regions

**tec.masking.window.size** size of the window used to mask the area around significantly enriched control regions

**tec.z** Z-score defining statistical stringency by which a given window is determined to be significantly higher in the input than in the signal, and masked if that is the case.

**tec.poisson.z** Z-score defining statistical stringency by which a given window is determined to be significantly higher than the tec.poisson.ratio above the expected uniform input background.

**tec.poisson.ratio**

Fold ratio by which input must exceed the level expected from the uniform distribution.

### Value

- **npl** A per-chromosome list containing data frames describing determined binding positions.  
Column description
  - **x**: position
  - **y**: score
  - **evalue**: E-value
  - **fdr**: FDR. For peaks higher than the maximum control peak, the highest dataset FDR is reported
  - **enr**: lower bound of the fold-enrichment ratio confidence interval. This is the estimate determined using scale of 1. Estimates corresponding to higher scales are returned in other enr columns with scale appearing in the name.
  - **enr.mle**: enrichment ratio maximum likely estimate
- **thr**: info on the chosen statistical threshold of the peak scores

### Examples

```
## Not run:
# find binding positions using WTD method, 200bp half-window size,
# control data, 1
bp <- find.binding.positions(signal.data=chip.data,
  control.data=input.data,
  fdr=0.01,
  method=tag.wtd,
  whs=200);

# find binding positions using MTC method, using 5 tag randomizations,
# keeping pairs of tag positions together (shuffle.window=2)
bp <- find.binding.positions(signal.data=chip.data,
  control.data=input.data,
  fdr=0.01,method=tag.lwcc,
  whs=200,
  use.randomized.controls=T,
  n.randomizations=5,
  shuffle.window=2)

# print out the number of determined positions
print(paste("detected",sum(unlist(lapply(bp$npl,function(d) length(d$x))))),"peaks"));

## End(Not run)
```

**get.binding.characteristics**

*Calculate characteristics of observed DNA-binding signal from cross-correlation profiles*

**Description**

The methods calculates strand cross-correlation profile to determine binding peak separation distance and approximate window size that should be used for binding detection. If quality scores were given for the tags, which quality bins improve the cross-correlation pattern.

**Usage**

```
get.binding.characteristics(data, strange = c(50, 500), bin = 5,
  cluster = NULL, debug = F, min.tag.count = 1000,
  acceptance.z.score = 3, remove.tag.anomalies = T,
  anomalies.z = 5, accept.all.tags=F)
```

**Arguments**

<code>data</code>	Tag/quality data: output of <code>read.eland.tags</code> or similar function
<code>strange</code>	A range within which the binding peak separation is expected to fall. Should be larger than probe size to avoid artifacts.
<code>bin</code>	Resolution (in basepairs) at which cross-correlation should be calculated. <code>bin=1</code> is ideal, but takes longer to calculate.
<code>cluster</code>	optional snow cluster for parallel processing
<code>debug</code>	whether to print debug messages
<code>min.tag.count</code>	minimal number of tags on the chromosome to be considered in the cross-correlation calculations
<code>acceptance.z.score</code>	A Z-score used to determine if a given tag quality bin provides significant improvement to the strand cross-correlation
<code>remove.tag.anomalies</code>	Whether to remove singular tag count peaks prior to calculation. This is recommended, since such positions may distort the cross-correlation profile and increase the necessary computational time.
<code>anomalies.z</code>	Z-score for determining if the number of tags at a given position is significantly higher about background, and should be considered an anomaly.
<code>accept.all.tags</code>	Whether tag alignment quality calculations should be skipped and all available tags should be accepted in the downstream analysis.

**Value**

<code>cross.correlation</code>	Cross-correlation profile as an \$x/\$y data.frame
<code>peak</code>	Position (\$x) and height (\$y) of automatically detected cross-correlation peak.
<code>whs</code>	Optimized window half-size for binding detection (based on the width of the cross-correlation peak)
<code>quality.bin.acceptance</code>	A list structure, describing the effect of inclusion of different tag quality bins on cross-correlation, and a resolution on which bins should be considered.
	<ul style="list-style-type: none"> <li>• <code>informative.bins</code>: A boolean vector indicating whether the inclusion of tags from the tag quality bin specified in the name attribute significantly increases cross-correlation profile near the peak.</li> <li>• <code>quality.cc</code>: A list giving the cross-correlation profile after the inclusion of the tags from different quality bins</li> </ul>

`get.broad.enrichment.clusters`  
*Determine broad clusters of enrichment*

**Description**

Scan chromosomes with a pre-defined window size, comparing scaled ChIP and input tag counts to see if their ratio exceeds that expected from a Poisson process (normalized for dataset size).

**Usage**

```
get.broad.enrichment.clusters(signal.data,
  control.data,
  window.size=1e3,
  z.thr=3,
  tag.shift=146/2,
  background.density.scaling = F,
  ...)
```

**Arguments**

<code>signal.data</code>	chip.tags, foreground tag vector list
<code>control.data</code>	input.tags, background tag vector list
<code>window.size</code>	window size to be used for tag counting
<code>z.thr</code>	Z-score to be used as a significance threshold
<code>tag.shift</code>	number of base pairs by which positive and negative tag coordinates should be shifted towards each other (half of binding peak separation distance)

`background.density.scaling`

If TRUE, regions of significant tag enrichment will be masked out when calculating size ratio of the signal to control datasets (to estimate ratio of the background tag density). If FALSE, the dataset ratio will be equal to the ratio of the number of tags in each dataset.

... additional parameters should be the same as those passed to the `find.significantly.enriched.region`.

## Value

A list of elements corresponding to chromosomes, with each element being an `$s/$e/$rv` data.frame giving the starting, ending positions and the log2 enrichment estimate for that region.

`get.conservative.fold.enrichment.profile`

*Estimate minimal fold enrichment/depletion along the chromosomes*

## Description

The method provides a statistical assessment of enrichment/depletion along the chromosomes. To assess tag density enrichment/depletion, a sliding window of a specified size (`fws`) is used to calculate the density of the foreground tags (`ftl`). Multiple, typically larger windows are used to estimate background tag (`btl`) density around the same location. The densities are compared as ratios of two Poisson processes to estimate lower bound of foreground enrichment, or upper bound of foreground depletion. If multiple window sizes were used to estimate the background tag density, the most conservative one is chosen for each point.

## Usage

```
get.conservative.fold.enrichment.profile(ftl,
  btl,
  fws,
  bwsl = c(1, 5, 25, 50) * fws,
  step = 50,
  tag.shift = 146/2,
  alpha = 0.05,
  use.most.informative.scale = F,
  quick.calculation = T,
  background.density.scaling = T,
  bg.weight = NULL,
  posl= NULL,
  return.mle = F)
```

## Arguments

<code>ftl</code>	foreground tag vector list
<code>btl</code>	background tag vector list
<code>fws</code>	foreground window size

bws1	background window scales. The size(s) of background windows will be fws*bws1.
step	spacing between positions at which the enrichment/depletion is evaluated
tag.shift	number of basepairs by which positive and negative tag coordinates should be shifted towards eachother (half of binding peak separation distance)
alpha	desired level of statistical significance
use.most.informative.scale	for each position, instead of evaluating enrichment ratio bounds for all background window scales, choose the one with the highest observed density to speed up the calculations
quick.calculation	Use square root transformation method instead of a Bayesian method. This speeds up the calculation considerably and is turned on by default.
background.density.scaling	If TRUE, regions of significant tag enrichment will be masked out when calculating size ratio of the signal to control datasets (to estimate ratio of the background tag density). If FALSE, the dataset ratio will be equal to the ratio of the number of tags in each dataset.
bg.weight	optional weight by which the background density should be multiplied for scaling. If not supplied, the weight is calculated based on the ratio of the reduced ChIP to input dataset sizes.
pos1	pos1
return.mle	return.mle

### Value

A list of elements corresponding to chromosomes, with each element being an `x/y` data.frame giving the position and the log2 conservative estimate of enrichment/depletion fold ratios around that position. Use [writewig](#) to output the structure to a WIG file.

### References

R.M.Price, D.G. Bonett "Estimating the ratio fo two Poisson rates", Comp. Stat & Data Anal. 32(2000) 345

### See Also

[get.smoothed.tag.density](#)

### Examples

```
## Not run:
enrichment.estimates <- get.conservative.fold.enrichment.profile(chip.data,
  input.data,
  fws=2*binding.characteristics$whs,
  step=100,
  alpha=0.01);
writewig(enrichment.estimates,
```

```

"example.enrichment.estimates.wig",
"Example conservative fold-enrichment/depletion estimates shown on log2 scale");

## End(Not run)

```

### get.conservative.fold.enrichment.profile2

*Return Conservative fold enrichment profile controlling for input and  
a single background scale*

## Description

Returns a conservative upper/lower bound profile (log2) given signal tag list, background tag list and window scales controlling for input, and supporting only a single background scale.

Novel version of get.conservative.fold.enrichment.profile() supporting a single background scale.

## Usage

```
get.conservative.fold.enrichment.profile2(ftl1, ftl2, btl1, btl2, fws,
                                         bws = 1 * fws, step = 50, tag.shift = 146/2,
                                         alpha = 0.05, background.density.scaling = T,
                                         bg.weight1 = NULL, bg.weight2 = NULL,
                                         posl = NULL, return.mle = F)
```

## Arguments

ftl1	Parameter
ftl2	Parameter
btl1	Parameter
btl2	Parameter
fws	Parameter
bws	Parameter
step	Parameter
tag.shift	Parameter
alpha	Parameter
background.density.scaling	Parameter
bg.weight1	Parameter
bg.weight2	Parameter
posl	Parameter
return.mle	Parameter

**Value**

A list of elements corresponding to chromosomes, with each element being an `x/y` data.frame giving the position and associated log2 signal/control enrichment estimate.

**See Also**

[get.smoothed.enrichment.mle](#)

**Examples**

```
## Not run:
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (ftl1, ftl2, btl1, btl2, fws, bws = 1 * fws, step = 50,
         tag.shift = 146/2, alpha = 0.05, background.density.scaling = T,
         bg.weight1 = NULL, bg.weight2 = NULL, pos1 = NULL, return.mle = F)
{
  ftl1 <- ftl1[unlist(lapply(ftl1, length)) > 2]
  chrl <- names(ftl1)
  names(chrl) <- chrl
  if (!is.null(pos1)) {
    chrl <- chrl[chrl %in% names(pos1)]
  }
  if (is.null(bg.weight1)) {
    bg.weight1 <- dataset.density.ratio(ftl1, btl1,
                                         background.density.scaling = background.density.scaling)
  }
  if (is.null(bg.weight2)) {
    bg.weight2 <- dataset.density.ratio(ftl2, btl2,
                                         background.density.scaling = background.density.scaling)
  }
  lapply(chrl, function(chr) {
    x <- binomial.proportion.ratio.bounds(abs(ftl1[[chr]] +
                                              tag.shift), abs(btl1[[chr]] + tag.shift), abs(ftl2[[chr]] +
                                              tag.shift), abs(btl2[[chr]] + tag.shift), fws = fws,
                                              bws = bws, step = step, bg.weight1 = bg.weight1,
                                              bg.weight2 = bg.weight2, alpha = alpha, pos = if (is.null(pos1)) {
                                                NULL
                                              }
                                              else {
                                                pos1[[chr]]
                                              })
    ps <- rep(0, length(x$mle))
    vi <- which(!is.na(x$lb) & x$lb > 0)
    ps[vi] <- x$lb[vi]
    vi <- which(!is.na(x$ub) & x$ub < 0)
    ps[vi] <- x$ub[vi]
    if (is.null(pos1)) {
      if (return.mle) {
```

```

        return(data.frame(x = x$x, y = ps, mle = x$mle,
                           lb = x$lb, ub = x$ub))
    }
    else {
        return(data.frame(x = x$x, y = ps))
    }
}
else {
    if (return.mle) {
        return(data.frame(x = pos1[[chr]], y = ps, mle = x$mle,
                           lb = x$lb, ub = x$ub))
    }
    else {
        return(data.frame(x = pos1[[chr]], y = ps))
    }
}
})
}

## End(Not run)

```

get.mser

*Calculate minimal saturated enrichment fold ratio*

## Description

Determine if the dataset has reached absolute saturation, or otherwise find minimal fold enrichment ratio above which the detection of peaks has stabilized enough to meet the saturation criteria.

## Usage

```
get.mser(signal.data,
         control.data,
         n.chains = 5,
         step.size = 1e+05,
         chains = NULL,
         cluster = NULL,
         test.agreement = 0.99,
         return.chains = F,
         enrichment.background.scales = c(1),
         n.steps = 1, ...)
```

## Arguments

signal.data	signal tag vector list
control.data	control tag vector list
n.chains	number of dataset subsamples to use

step.size	subsampling step describing the saturation criteria. The criteria requires the set of detected binding sites to be stable (as described by the test.agreement param) when the number of tags in the dataset is reduced by step.size. The value can either be an integer above one, in which case it specifies a fixed number of tags, or a real value below one, in which case it specifies the fraction of tags that should be removed (e.g. 0.1 will remove 10)
test.agreement	Fraction of the detected peaks that should agree between the full and subsampled datasets.
chains	optional parameter, giving pre-calculated chains
cluster	optional snow cluster to parallelize processing
return.chains	whether subsampled dataset results should be returned as well
enrichment.background.scales	one or multiple window scales at which the background tag density should be assessed. See enrichment.background.scales in <a href="#">find.binding.positions</a> . If multiple scales are provided, multiple MSER estimates will be returned.
...	additional parameters should be the same as those passed to the <a href="#">find.binding.positions</a>
n.steps	n.steps

### Value

A single, or multiple (if multiple enrichment.background.scales were provided) MSER value. A value of 1 or very close to it implies that the dataset has reached absolute saturation based on the given criteria.

## get.mser.interpolation

*Interpolate MSER dependency on the tag count*

### Description

MSER generally decreases with increasing sequencing depth. This function interpolates the dependency of MSER on tag counts as a log-log linear function. The log-log fit is used to estimate the depth of sequencing required to reach desired target.fold.enrichment.

### Usage

```
get.mser.interpolation(signal.data,
  control.data,
  target.fold.enrichment = 5,
  n.chains = 10,
  n.steps = 6,
  step.size = 1e+05,
  chains = NULL,
  test.agreement = 0.99,
  return.chains = F,
  enrichment.background.scales = c(1),
  excluded.steps = c(seq(2, n.steps - 2)), ...)
```

## Arguments

<code>signal.data</code>	signal chromosome tag vector list
<code>control.data</code>	control chromosome tag vector list
<code>target.fold.enrichment</code>	target MSER for which the depth should be estimated
<code>n.steps</code>	number of steps in each subset chain.
<code>step.size</code>	Either number of tags or fraction of the dataset size, see <code>step.size</code> parameter for <a href="#">get.mser</a> .
<code>test.agreement</code>	Fraction of the detected peaks that should agree between the full and subsampled datasets. See <code>test.agreement</code> parameter for <a href="#">get.mser</a>
<code>n.chains</code>	number of random subset chains
<code>chains</code>	optional structure of pre-calculated chains (e.g. generated by an earlier call with <code>return.chains=T</code> ).
<code>return.chains</code>	whether to return peak predictions calculated on random chains. These can be passed back using <code>chains</code> argument to skip subsampling/prediction steps, and just recalculate the depth estimate for a different MSER.
<code>enrichment.background.scales</code>	see <code>enrichment.background.scales</code> parameter for <a href="#">get.mser</a>
<code>excluded.steps</code>	Intermediate subsampling steps that should be excluded from the chains to speed up the calculation. By default, all intermediate steps except for first two and last two are skipped. Adding intermediate steps improves interpolation at the expense of computational time.
<code>...</code>	additional parameters are passed to <a href="#">get.mser</a>

## Details

To simulate sequencing growth, the method calculates peak predictions on random chains. Each chain is produced by sequential random subsampling of the original data. The number of steps in the chain indicates how many times the random subsampling will be performed.

## Value

Normally returns a list, specifying for each backgroundscale:

<code>prediction</code>	estimated sequencing depth required to reach specified target MSER
<code>log10.fit</code>	linear fit model, a result of <code>lm()</code> call

If `return.chains=T`, the above structure is returned under `interpolation` field, along with `chains` field containing results of [find.binding.positions](#) calls on subsampled chains.

---

get.smoothed.enrichment.mle*Calculate chromosome-wide profiles of smoothed enrichment estimate*

---

**Description**

Given signal and control tag positions, the method calculates log2 signal to control enrichment estimates (maximum likelihood) for each chromosome, based on the smoothed tag density profile (see [get.smoothed.tag.density](#)).

**Usage**

```
get.smoothed.enrichment.mle(signal.tags,
  control.tags,
  tag.shift = 146/2,
  background.density.scaling = F,
  pseudocount = 1,
  bg.weight = NULL,
  rngl = NULL,
  chrl = NULL,
  ...)
```

**Arguments**

signal.tags	signal chromosome tag coordinate vectors (e.g. output of <a href="#">select.informative.tags</a> )
control.tags	control (input) tags
tag.shift	number of base pairs by which positive and negative tag coordinates should be shifted towards eachother (half of binding peak separation distance)
background.density.scaling	background.density.scaling
pseudocount	pseudocount value to be added to tag density - defaults to 1
bg.weight	optional weight by which the background density should be multiplied for scaling. If not supplied, the weight is calculated based on the ratio of the reduced ChIP to input dataset sizes.
rngl	rngl
chrl	chrl
...	additional parameters should be the same as those passed to the <a href="#">get.smoothed.tag.density</a> , such as for example bandwidth (default value 150) and step (default value 50). see appropriate reference for <a href="#">get.smoothed.tag.density</a> for details.

**Value**

A list of elements corresponding to chromosomes, with each element being an \$x/\$y data.frame giving the position and associated log2 signal/control enrichment estimate.

**See Also**[writewig](#)**Examples**

```
## Not run:
# get smoothed enrichment estimate profile using 500bp bandwidth at
# 50bp steps
smoothed.M <- get.smoothed.enrichment.mle(chip.data,bandwidth=500,step=50);
writewig(smoothed.M,"example.smoothedM.wig","Example smoothed log2 intensity ratio estimate");

## End(Not run)
```

`get.smoothed.enrichment.mle2`

*Calculate background input controlled chromosome-wide profiles of smoothed enrichment estimate*

**Description**

Given signal and control tag positions, the method calculates log2 signal to control enrichment estimates (maximum likelihood) for each chromosome, based on the smoothed tag density profile (see [get.smoothed.tag.density](#)).

**Usage**

```
get.smoothed.enrichment.mle2(signal.tags1, control.tags1,
                             signal.tags2, control.tags2,
                             tag.shift = 146/2, background.density.scaling = F,
                             pseudocount = 1, bg.weight1 = NULL,
                             bg.weight2 = NULL,
                             rng1 = NULL, chrl = NULL, ...)
```

**Arguments**

<code>signal.tags1</code>	Parameter
<code>control.tags1</code>	Parameter
<code>signal.tags2</code>	Parameter
<code>control.tags2</code>	Parameter
<code>tag.shift</code>	Parameter
<code>background.density.scaling</code>	Parameter
<code>pseudocount</code>	Parameter
<code>bg.weight1</code>	Parameter
<code>bg.weight2</code>	Parameter

rngl	Parameter
chr1	Parameter
...	Parameter

**Value**

A list of elements corresponding to chromosomes, with each element being an \$x/\$y data.frame giving the position and associated log2 signal/control enrichment estimate.

**See Also**

[get.smoothed.enrichment.mle](#)

**Examples**

```
## Not run:
##### Should be DIRECTLY executable !! -----
##### ==> Define data, use random,
##### or do help(data=index) for the standard data sets.

## The function is currently defined as
function (signal.tags1, control.tags1, signal.tags2, control.tags2,
  tag.shift = 146/2, background.density.scaling = F, pseudocount = 1,
  bg.weight1 = NULL, bg.weight2 = NULL, rngl = NULL, chr1 = NULL,
  ...)
{
  if (is.null(chr1)) {
    chr1 <- intersect(names(signal.tags1), names(signal.tags2))
    names(chr1) <- chr1
  }
  if (is.null(rngl)) {
    rngl <- lapply(chr1, function(chr) range(c(range(abs(signal.tags1[[chr]] +
      tag.shift)), range(abs(signal.tags2[[chr]] + tag.shift)))))
  }
  else {
    chr1 <- names(rngl)
    names(chr1) <- chr1
  }
  ssd1 <- get.smoothed.tag.density(signal.tags1, rngl = rngl,
    ..., scale.by.dataset.size = F)
  ssd2 <- get.smoothed.tag.density(signal.tags2, rngl = rngl,
    ..., scale.by.dataset.size = F)
  csd1 <- get.smoothed.tag.density(control.tags1, rngl = rngl,
    ..., scale.by.dataset.size = F)
  csd2 <- get.smoothed.tag.density(control.tags2, rngl = rngl,
    ..., scale.by.dataset.size = F)
  if (is.null(bg.weight1)) {
    bg.weight1 <- dataset.density.ratio(signal.tags1, control.tags1,
      background.density.scaling = background.density.scaling)
  }
  if (is.null(bg.weight2)) {
    bg.weight2 <- dataset.density.ratio(signal.tags2, control.tags2,
```

```

        background.density.scaling = background.density.scaling)
    }
cmle <- lapply(chrl, function(chr) {
  d <- ssd1[[chr]]
  d$y <- log2(ssd1[[chr]]$y + pseudocount * bg.weight1) -
    log2(csd1[[chr]]$y + pseudocount) - log2(bg.weight1) -
    log2(ssd2[[chr]]$y + pseudocount * bg.weight2) +
    log2(csd2[[chr]]$y + pseudocount) + log2(bg.weight2)
  return(d)
})
}

## End(Not run)

```

**get.smoothed.tag.density***Calculate chromosome-wide profiles of smoothed tag density***Description**

Given tag positions, the method calculates for each chromosome a tag density profile, smoothed by the Gaussian kernel. If the optional control tags are provided, the difference between ChIP and control tag density is returned.

**Usage**

```
get.smoothed.tag.density(signal.tags,
  control.tags = NULL,
  bandwidth = 150,
  bg.weight = NULL,
  tag.shift = 146/2,
  step = round(bandwidth/3),
  background.density.scaling = T,
  rngl = NULL,
  scale.by.dataset.size = F)
```

**Arguments**

<code>signal.tags</code>	signal chromosome tag coordinate vectors (e.g. output of <a href="#">select.informative.tags</a> )
<code>control.tags</code>	optional control (input) tags
<code>bandwidth</code>	standard deviation of the Gaussian kernel
<code>bg.weight</code>	optional weight by which the background density should be multiplied for scaling. If not supplied, the weight is calculated based on the ratio of the reduced ChIP to input dataset sizes.
<code>tag.shift</code>	Distance by which the positive and negative strand tags should be shifted towards eachother. This normally corresponds to the half of the cross-correlation peak position (e.g. <code>get.binding.characteristics()\$peak\$x/2</code> )

step            The distance between the regularly spaced points for which the values should be calculated.

background.density.scaling  
 If TRUE, regions of significant tag enrichment will be masked out when calculating size ratio of the signal to control datasets (to estimate ratio of the background tag density). If FALSE, the dataset ratio will be equal to the ratio of the number of tags in each dataset.

rngl            rngl

scale.by.dataset.size  
 scale.by.dataset.size

**Value**

A list of elements corresponding to chromosomes, with each element being an \$x/\$y data.frame giving the position and associated tag density. Use [writelnwig](#) to output the structure to a WIG file.

**See Also**

[writelnwig](#)

**Examples**

```
## Not run:
smoothed.density <- get.smoothed.tag.density(chip.data,
  control.tags=input.data,
  bandwidth=200,
  step=100,
  tag.shift=round(binding.characteristics$peak$x/2));

writelnwig(smoothed.density,
  "example.density.wig",
  "Example smoothed, background-subtracted tag density");

## End(Not run)
```

**output.binding.results**

*Write out determined binding peaks into a text file table*

**Description**

Writes out determined binding positions into a text file. The file will contain a table with each row corresponding to a detected position, with the following columns

- chr: chromosome or target sequence
- pos: position of detected binding site on the chromosome/sequence
- score: a score reflecting magnitude of the binding

- **Evalue:** E-value corresponding to the peak magnitude
- **FDR:** FDR corresponding to the peak magnitude
- **enrichment.lb:** lower bound of the fold-enrichment ratio
- **enrichment.mle:** maximum likelihood estimate of the fold-enrichment ratio

## Usage

```
output.binding.results(results, filename)
```

## Arguments

<b>results</b>	output of the <a href="#">find.binding.positions</a>
<b>filename</b>	file name

<b>points_within</b>	<i>Find points within</i>
----------------------	---------------------------

## Description

`points_within` substitutes the deprecated function "points.within"

## Usage

```
points_within(x, fs, fe, return.list = F, return.unique = F,
              sorted = F, return.point.counts = F, ...)
```

## Arguments

<b>x</b>	Parameter
<b>fs</b>	Parameter
<b>fe</b>	Parameter
<b>return.list</b>	Parameter
<b>return.unique</b>	Parameter
<b>sorted</b>	Parameter
<b>return.point.counts</b>	Parameter
<b>...</b>	Parameter

## Value

Parameter

## Examples

```

## Not run:
##### Should be DIRECTLY executable !! -----
##### ==> Define data, use random,
##### or do help(data=index) for the standard data sets.

## The function is currently defined as
function (x, fs, fe, return.list = F, return.unique = F, sorted = F,
          return.point.counts = F, ...)
{
  if (is.null(x) | length(x) < 1) {
    return(c())
  }
  if (!sorted) {
    ox <- rank(x, ties.method = "first")
    x <- sort(x)
  }
  se <- c(fs, fe)
  fi <- seq(1:length(fs))
  fi <- c(fi, -1 * fi)
  fi <- fi[order(se)]
  se <- sort(se)
  storage.mode(x) <- storage.mode(fi) <- storage.mode(se) <- "integer"
  if (return.unique) {
    iu <- 1
  }
  else {
    iu <- 0
  }
  if (return.list) {
    il <- 1
  }
  else {
    il <- 0
  }
  if (return.point.counts) {
    rpc <- 1
  }
  else {
    rpc <- 0
  }
  storage.mode(iu) <- storage.mode(il) <- storage.mode(rpc) <- "integer"
  result <- .Call("points_withinC", x, se, fi, il, iu, rpc)
  if (!sorted & !return.point.counts) {
    result <- result[ox]
  }
  return(result)
}

## End(Not run)

```

`read.arachne.tags`      *Read in Arachne tags*

## Description

Read in ARACHNE Tag file

## Usage

```
read.arachne.tags(filename, fix.chromosome.names = F)
```

## Arguments

<code>filename</code>	filename
<code>fix.chromosome.names</code>	do we fix chromosome names

## Value

A list like element

## See Also

Fill in later

## Examples

```
## Not run:
##### Should be DIRECTLY executable !!
#### ==> Define data, use random,
#### or do help(data=index) for the standard data sets.

## The function is currently defined as
function (filename, fix.chromosome.names = F)
{
  tl <- lapply(.Call("read_arachne_long", path.expand(filename)),
    function(d) {
      xo <- order(abs(d$t))
      d$t <- d$t[xo]
      d$n <- d$n[xo]
      d$l <- d$l[xo]
      return(d)
    })
  if (fix.chromosome.names) {
    names(tl) <- gsub("\.fa", "", names(tl))
  }
  return(list(tags = lapply(tl, function(d) d$t), quality = lapply(tl,
    function(d) d$n), length = lapply(tl, function(d) d$l)))
}
```

---

```
## End(Not run)
```

---

<code>read.bam.tags</code>	<i>Read BAM alignment file</i>
----------------------------	--------------------------------

---

### Description

Reads in aligned reads from BAM file. Note: no split (non-unique) alignments should be reported in the BAM file.

### Usage

```
read.bam.tags(filename, read.tag.names = F, fix.chromosome.names = F)
```

### Arguments

filename	BAM file
read.tag.names	Whether the tag names should be read in
fix.chromosome.names	Whether to remove ".fa" from the end of the sequence names

### Value

tags	A vector of 5' tag coordinates, with negative values corresponding to tags mapped to the negative strand.
quality	Number of mismatches
names	Tag names, if read.tag.names was set

---

<code>read.bin.maqmap.tags</code>	<i>Read MAQ binary alignment map file</i>
-----------------------------------	---

---

### Description

Reads in MAQ binary map alignment result file

### Usage

```
read.bin.maqmap.tags(filename, read.tag.names = F, fix.chromosome.names = T)
```

### Arguments

filename	MAQ map output file (binary)
read.tag.names	Whether the tag names should be read in
fix.chromosome.names	Whether to remove ".fa" from the end of the sequence names

**Value**

<code>tags</code>	A vector of 5' tag coordinates, with negative values corresponding to tags mapped to the negative strand.
<code>quality</code>	Number of mismatches
<code>names</code>	Tag names, if <code>read.tag.names</code> was set

`read.bowtie.tags`      *Read bowtie text alignment output file*

**Description**

Reads in bowtie alignment results in text format

**Usage**

```
read.bowtie.tags(filename, read.tag.names = F, fix.chromosome.names = F)
```

**Arguments**

<code>filename</code>	bowtie text output file
<code>read.tag.names</code>	Whether the tag names should be read in
<code>fix.chromosome.names</code>	Whether to remove ".fa" from the end of the sequence names

**Value**

<code>tags</code>	A vector of 5' tag coordinates, with negative values corresponding to tags mapped to the negative strand.
<code>quality</code>	Number of mismatches
<code>names</code>	Tag names, if <code>read.tag.names</code> was set

`read.eland.tags`      *Read eland output file*

**Description**

Reads in ELAND output file, returning 5'-end tag coordinates and number of mismatches associated with each mapped tag.

**Usage**

```
read.eland.tags(filename,
  read.tag.names = F,
  fix.chromosome.names = T,
  max.eland.tag.length = -1,
  extended=F,
  multi = F)
```

**Arguments**

filename	ELAND output file
read.tag.names	Whether the tag names should be read in
fix.chromosome.names	Whether to remove ".fa" from the end of the sequence names
max.eland.tag.length	Specifies max length of the tag sequence considered by ELAND. This needs to be specified if the tags are longer than the sequences considered by ELAND during alignment.
extended	Whether the file is written out in "extended" format provided in GA pipeline 1.0.
multi	Whether the file is written in "multi" format, showing multiple alignments of the reads

**Value**

tags	A vector of 5' tag coordinates, with negative values corresponding to tags mapped to the negative strand.
quality	Number of mismatches
names	Tag names, if <code>read.tag.names</code> was set

`read.helicos.tags`      *Read in helicos tags*

**Description**

Read in Helicos tags

**Usage**

```
read.helicos.tags(filename, read.tag.names = F,
  fix.chromosome.names = F, include.length.info = T)
```

**Arguments**

```

filename      filename
read.tag.names Read in tag names
fix.chromosome.names
                  Do we fix chromosome names
include.length.info
                  include length information

```

**Value**

A list like structure

**See Also**

Fill in later

**Examples**

```

## Not run:
##### Should be DIRECTLY executable !! -----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (filename, read.tag.names = F, fix.chromosome.names = F,
           include.length.info = T)
{
  if (read.tag.names) {
    rtn <- as.integer(1)
  }
  else {
    rtn <- as.integer(0)
  }
  tl <- lapply(.Call("read_helicostabf", path.expand(filename),
    rtn), function(d) {
    xo <- order(abs(d$t))
    d$t <- d$t[xo]
    d$n <- d$n[xo]
    d$l <- d$l[xo]
    if (read.tag.names) {
      d$s <- d$s[xo]
    }
    return(d)
  })
  if (fix.chromosome.names) {
    names(tl) <- gsub("\.fa", "", names(tl))
  }
  if (read.tag.names) {
    return(list(tags = lapply(tl, function(d) d$t), quality = lapply(tl,
      function(d) d$n), length = lapply(tl, function(d) d$l),
      names = lapply(tl, function(d) d$s)))
  }
}

```

```

    }
  else {
    return(list(tags = lapply(tl, function(d) d$t), quality = lapply(tl,
      function(d) d$n), length = lapply(tl, function(d) d$l)))
  }
}

## End(Not run)

```

**read.maqmap.tags***Read MAQ text alignment output file***Description**

Reads in MAQ alignment results in text format (that results from "maq mapview" command.)

**Usage**

```
read.maqmap.tags(filename, read.tag.names = F, fix.chromosome.names = T)
```

**Arguments**

<code>filename</code>	MAQ text output file
<code>read.tag.names</code>	Whether the tag names should be read in
<code>fix.chromosome.names</code>	Whether to remove ".fa" from the end of the sequence names

**Value**

<code>tags</code>	A vector of 5' tag coordinates, with negative values corresponding to tags mapped to the negative strand.
<code>quality</code>	Number of mismatches
<code>names</code>	Tag names, if <code>read.tag.names</code> was set

**read.meland.tags***Read modified BED tag alignment file that contains variable match length information***Description**

Reads in an extended BED tag alignment file. An example line given below: 49 . U1 . 1 . . 23 chr2 -234567  
The line above specifies a 23-bp portion of the tag tag with id 49 was aligned with 1 mismatch to the negative strand of chr2 at position 234567.

**Usage**

```
read.meland.tags(filename, read.tag.names = F, fix.chromosome.names = T)
```

**Arguments**

filename	name of the extended BED file
read.tag.names	whether to read in tag names
fix.chromosome.names	whether to remove ".fa" from the sequence name ends.

**Value**

tags	A vector of 5' tag coordinates, with negative values corresponding to tags mapped to the negative strand.
quality	Quality expressed as a float x.y, where x is tag.length - aligned.tag.portion.length, and y is the number of mismatches (must be less than 10).
names	Tag names, if read.tag.names was set

**read.short.arachne.tags**

*Read in ARACHNE short tags*

**Description**

Read in short arachne reads

**Usage**

```
read.short.arachne.tags(filename, fix.chromosome.names = F)
```

**Arguments**

filename	filename
fix.chromosome.names	Fix chromosome names

**Details**

Not necessary

**Value**

A list like structure

**Note**

No Notes

**Author(s)**

Peter Kharchenko

**References**

spp by Peter Kharchenko

**See Also**

Nothing to see here

**Examples**

```
## Not run:
##### Should be DIRECTLY executable !! -----
#### ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (filename, fix.chromosome.names = F)
{
  tl <- lapply(.Call("read_arachne", path.expand(filename)),
    function(d) {
      xo <- order(abs(d$t))
      d$t <- d$t[xo]
      d$n <- d$n[xo]
      return(d)
    })
  if (fix.chromosome.names) {
    names(tl) <- gsub("\.fa", "", names(tl))
  }
  return(list(tags = lapply(tl, function(d) d$t), quality = lapply(tl,
    function(d) d$n)))
}
## End(Not run)
```

**read.tagalign.tags**      *Read in tagalign tags*

**Description**

Fill in later

**Usage**

```
read.tagalign.tags(filename, fix.chromosome.names = T, fix.quality = T)
```

**Arguments**

filename	Filename of tag file
fix.chromosome.names	chromosome names
fix.quality	fix quality

**Details**

...

**Value**

a list like structure

**Note**

Needs further editing

**Author(s)**

Peter K.

**References**

spp by kharchenko

**See Also**

Buh!

**Examples**

```
## Not run:
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (filename, fix.chromosome.names = T, fix.quality = T)
{
  tl <- lapply(.Call("read_tagalign", path.expand(filename)),
    function(d) {
      xo <- order(abs(d$t))
      d$t <- d$t[xo]
      d$n <- d$n[xo]
      if (fix.quality) {
        if (min(d$n) < 0.5) {
          d$n = ceiling(1000/4^d$n)
        }
        break.vals <- unique(sort(c(0, unique(d$n))))
      }
    })
  return(list(tl = tl, break.vals = break.vals))
}
```

```

d$n <- length(break.vals) - 1 - cut(d$n, breaks = break.vals,
  labels = F)
}
return(d)
})
if (fix.chromosome.names) {
  names(tl) <- gsub("\.fa", "", names(tl))
}
return(list(tags = lapply(tl, function(d) d$t), quality = lapply(tl,
  function(d) d$n)))
}

## End(Not run)

```

**remove.local.tag.anomalies**

*Restrict or remove positions with too many tags relative to local background.*

**Description**

In Solexa ChIP-seq experiments some anomalous positions contain extremely high number of tags at the exact coordinates. The function scans the chromosomes, determining local tag density based on a provided `window.size`, doing two types of corrections: 1. removing all tags from positions that exceed local density by `eliminate.fold`; 2. reducing the tag count at positions exceeding `cap.fold` to the maximal allowed count. The statistical significance of counts exceeding either of these two threshold densities is calculated based on Poisson model, with confidence interval determined by the `z.threshold` Z-score parameter.

**Usage**

```
remove.local.tag.anomalies(tags,
  window.size = 200,
  eliminate.fold = 10,
  cap.fold = 4,
  z.threshold = 3)
```

**Arguments**

<code>tags</code>	Chromosome-list of tag vectors
<code>window.size</code>	Size of the window used to assess local density. Increasing the window size considerably beyond the size of the binding features will result in flattened profiles, with bound positions exhibiting a difference of just 1 tag beyond the background.
<code>eliminate.fold</code>	Threshold defining fold-over background density above which the position is considered anomalous and removed completely.
<code>cap.fold</code>	Threshold fold-over background density above which the position is capped to the maximum statistically likely given local tag density

`z.threshold` Z-score used to assess significance of a given position exceeding either of the two density thresholds.

### Value

A modified chromosome-wise tag vector list.

### Note

~~further notes~~ Increasing window.size to very large values will result in flat profiles similar to those described by Zhang et al. "Model-based Analysis of ChIP-Seq (MACS)." Genome Biol. 2008 Sep 17;9(9):R137.

### References

~put references to the literature/web site here ~

`select.informative.tags`

*Choose informative tags*

### Description

For datasets with tag alignment quality information (e.g. number of mismatches for Eland alignments), `get.binding.characteristics` determines whether inclusion of tags from each specific quality bin improves the cross-correlation profile. The present function is then used to actually select these informative tags, discarding all other information, including quality scores that are not used in further processing.

### Usage

`select.informative.tags(data, binding.characteristics)`

### Arguments

<code>data</code>	Full alignment data (a list with \$tags and \$quality elements)
<code>binding.characteristics</code>	result of a <code>get.binding.characteristics</code> call. If NULL value is supplied, all tags will be accepted.

### Value

A chromosome-wise tag list. Each element of the list corresponds to a chromosome and is a numeric vector of 5' tag coordinates, with sign designating DNA strand. This form of tag data is used for most of the other processing.

---

`write.broadpeak.info`    *Write out determined broad enrichment regions using broadPeak format*

---

## Description

Writes out broad regions of enrichment determined by the `get.broad.enrichment.clusters` method in a broadPeak format.

## Usage

```
write.broadpeak.info(bp, fname)
```

## Arguments

<code>bp</code>	broadpeak.results, output of the <a href="#">get.broad.enrichment.clusters</a>
<code>fname</code>	file name

---

`write.narrowpeak.binding`

*Write out determined binding peaks using narrowPeak format*

---

## Description

Writes out determined binding positions into a narrowPeak file. The region will correspond to associated broad enrichment region, if such were added using `add.broad.peak.regions` method. Otherwise the region size will be determined using margin (which defaults to the window half size that was used to determine binding positions) Note: since v1.13, FDR is written out in -log10() scale.

## Usage

```
write.narrowpeak.binding(bd, fname, margin=bd$whs, npeaks)
```

## Arguments

<code>bd</code>	output of the <a href="#">find.binding.positions</a>
<code>fname</code>	file name
<code>margin</code>	explicit value of the margin to be used if the broad region information is absent (defaults to peak detection window half-size)
<code>npeaks</code>	optionally, limit the output to the specified number of top peaks

**writewig***A function to save a list of chromosome-wise x/y data frames into a WIG file format.*

---

## Description

Takes a list that contains an \$x and \$y data.frame for a number of chromosomes and writes it out to a WIG BED style format.

## Usage

```
writewig(dat, fname, feature, threshold = 5, zip = F)
```

## Arguments

<b>dat</b>	Chromosome coordinate-value data. dat is a list, each member of a list is a data frame with \$x and \$y columns containing chromosome positions and associated values. The names of the list elements correspond to the chromosomes.
<b>fname</b>	Filename to which the output should be written
<b>feature</b>	Data description to be incorporated into the WIG header
<b>threshold</b>	Optional threshold to be saved in the WIG file
<b>zip</b>	Wheter to invoke a zip program to compress the file

## See Also

~~objects to See Also as [help](#), ~~

## Examples

```
## Not run:
data <- list("chr1"=data.frame(x=c(100,130,200),y=c(1.2,4.0,2.3)));
writewig(data,"filename");

## End(Not run)
```

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