

# Package ‘ramwas’

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**Type** Package

**Title** Fast Methylome-Wide Association Study Pipeline for Enrichment Platforms

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**Maintainer** Andrey A Shabalin <andrey.shabalin@gmail.com>

**Description** A complete toolset for methylome-wide association studies (MWAS). It is specifically designed for data from enrichment based methylation assays, but can be applied to other data as well. The analysis pipeline includes seven steps:  
(1) scanning aligned reads from BAM files,  
(2) calculation of quality control measures,  
(3) creation of methylation score (coverage) matrix,  
(4) principal component analysis for capturing batch effects and detection of outliers,  
(5) association analysis with respect to phenotypes of interest while correcting for top PCs and known covariates,  
(6) annotation of significant findings, and  
(7) multi-marker analysis (methylation risk score) using elastic net. Additionally, RaMWAS include tools for joint analysis of methylation and genotype data. This work is published in Bioinformatics, Shabalin et al. (2018) <doi:10.1093/bioinformatics/bty069>.

**URL** <https://bioconductor.org/packages/ramwas/>

**BugReports** <https://github.com/andreysbabalin/ramwas/issues>

**License** LGPL-3

**LazyLoad** yes

**NeedsCompilation** yes

**Depends** R (>= 3.3.0), methods, filematrix

**VignetteBuilder** knitr

**Suggests** knitr, rmarkdown, pander, BiocStyle, BSgenome.Ecoli.NCBI.20080805

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**biocViews** DNAMethylation, Sequencing, QualityControl, Coverage, Preprocessing, Normalization, BatchEffect, PrincipalComponent, DifferentialMethylation, Visualization

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**Author** Andrey A Shabalin [aut, cre] (<<https://orcid.org/0000-0003-0309-6821>>),  
Shaunna L Clark [aut],  
Mohammad W Hattab [aut],  
Karolina A Aberg [aut],  
Edwin J C G van den Oord [aut]

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<i>ramwas-package</i>	<i>Fast Methylome-wide Association Study Pipeline for Enrichment Platforms</i>
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## Description

RaMWAS provides a complete toolset for methylome-wide association studies (MWAS). It is specifically designed for data from enrichment based methylation assays, but can be applied to other methylomic data as well. The analysis pipeline includes seven steps: (1) scanning aligned reads from BAM files, (2) calculation of quality control measures, (3) creation of methylation score (coverage) matrix, (4) principal component analysis for capturing batch effects and detection of outliers, (5) association analysis with respect to phenotypes of interest while correcting for top PCs and known covariates, (6) annotation of significant findings, and (7) multi-marker analysis (methylation risk score) using elastic net.

## Details

Package: ramwas  
 Type: Package  
 License: LGPL-3  
 LazyLoad: yes  
 Depends: methods

## Author(s)

Andrey A Shabalín <andrey.shabalín@gmail.com>

Maintainer: Andrey A Shabalín <andrey.shabalín@gmail.com>

## See Also

See vignettes: `browseVignettes("ramwas")`.

cachedRDSload                      *Cached Loading of RDS Files*

---

### Description

Loads an .rds file `rdsfilename` using `readRDS` and returns the loaded object. The object is also saved in a cache so that repeated calls of the function with the same filename return the same object instantaneously.

### Usage

```
cachedRDSload(rdsfilename)
```

### Arguments

`rdsfilename`      Name of the RDS file.

### Details

The cached object is stored in a private package environment.

### Value

Returns the object loaded with `readRDS` from `rdsfilename` at this or a previous call of the function.

### Author(s)

Andrey A Shabalin <andrey.shabalin@gmail.com>

### Examples

```
### Change filename to hg19 CpGset

filename = system.file("extdata", "qc_sample.rds", package = "ramwas")

time1 = system.time( {obj1 = cachedRDSload(filename)} )
time2 = system.time( {obj1 = cachedRDSload(filename)} )

cat("First loading time:",time1[3],"seconds","\n")
cat("Second loading time:",time2[3],"seconds","\n")
```

---

findBestNpvs                      *Quickly Find N Smallest P-values in a Long Vector*

---

### Description

Finding top, say, 100 p-values out of millions can be slow. This function does it much faster than the usual application of `order(pv)[1:N]`.

### Usage

```
findBestNpvs(pv, n)
```

**Arguments**

pv                    Vector of p-values.  
n                     Number of best p-values to select.

**Details**

The function is a faster analog of `sort(order(pv)[1:N])`

**Value**

Return a vector of positions of the smallest N p-values in pv.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See [order](#).

**Examples**

```
pv = runif(1000)^10
n = 100

# Faster version
topSites1 = findBestNpvs(pv, n)

# Slow alternative
topSites2 = sort(order(pv)[1:n])

# The results must match
stopifnot(all( topSites1 == topSites2 ))
```

---

get

*Functions for Access to Data, MWAS Results, and Location Information*

---

**Description**

Functions for access to data, MWAS results, and location information.

Function `getLocation`s obtains the location information for all variables (CpGs).

Function `getMWASandLocations` obtains both MWAS results and location information in a single data frame.

Functions `getDataByLocation` and `getMWASrange` return the data (coverage) and MWAS results for the selected set of variables (CpGs).

**Usage**

```

getLocations(x)
getMWAS(x)
getMWASandLocations(x)
getMWASrange(x, chr, start, end)
getDataByLocation(x, chr, start, end)

```

**Arguments**

x	Name of directory or list of RaMWAS parameters as described in the "RW6_param.Rmd" vignette. Try: vignette("RW6_param", "ramwas").
	If a directory name is provided, it must point to <ul style="list-style-type: none"> <li>• Data (coverage) directory (parameter <code>dircoveragenorm</code>) for <code>getDataByLocation</code> and <code>getLocations</code></li> <li>• MWAS directory (parameter <code>dirmwas</code>) for <code>getMWAS</code>, <code>getMWASandLocations</code>, and <code>getMWASrange</code></li> </ul>
chr	Chromosome name or number.
start	Start position of the genomic region of interest.
end	End position of the genomic region of interest.

**Details**

The functions return the MWAS results and/or locations.

**Value**

Function `getLocations` returns a data frame with

chr	Chromosome
start	Start position
end	End position

Function `getMWAS` returns a data frame with

cor	coverage - phenotype correlation
t.test	t-statistic
p.value	p-value
q.value	q-value (FDR)

If the outcome variable was categorical, columns `cor` and `t.test` are replaced with `R.squared` and `F.test`.

Functions `getMWASandLocations` and `getMWASrange` return a data frame with elements of output of both `getLocations` and `getMWAS`

Function `getDataByLocation` returns a list with

locations	Chromosomal location information for located variables
matrix	Data (coverage) matrix for the selected locations

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
## Not run:
# Extract locations using parameter vector
getLocations(param)

# Extract locations using directory name
getLocations("/data/myMWAS")

# Extract MWAS using parameter vector
getMWAS(param)

# Extract MWAS using directory name
getMWAS("/data/myMWAS")

# Extract MWAS using parameter vector
getMWASandLocations(param)

# Extract MWAS using directory name
getMWASandLocations("/data/myMWAS")

# Extract MWAS for a region
getMWASrange(param, 1, 123321, 223321)

# Chromosome can be character
getMWASrange(param, "chr1", 123321, 223321)

# Extract data for a region
getDataByLocation(param, 1, 123321, 223321)

# Chromosome can be character
getDataByLocation(param, "chr1", 123321, 223321)

## End(Not run)
```

---

getCpGset

*Construct CpG set for a Reference Genome*

---

**Description**

Finds all CpGs in a reference genome.

## Usage

```
getCpGsetCG(genome)
getCpGsetALL(genome)
```

## Arguments

genome            A [BSgenome](#) object or a character vector with genome sequences.

## Details

The `getCpGsetCG` function searches for all CG pairs in the genome.  
The `getCpGsetALL` function also works for genomes with injected SNPs.

## Value

Returns a list with CpG coordinates for each genome sequence.

## Author(s)

Andrey A Shabalin <andrey.shabalin@gmail.com>

## Examples

```
### Using a BSgenome input

library(BSgenome.Ecoli.NCBI.20080805)
cpgset = getCpGsetCG(BSgenome.Ecoli.NCBI.20080805)

print("First 10 CpGs in NC_008253:")
print(cpgset$NC_008253[1:10])

### Using a character vector input

genome = list(
  chr1 = "AGCGTTTTTCATTCTGACTGCAACGGGCYR",
  chr2 = "AAAAAACGCCTTAGTAAGTGATTTTCGYR")
cpgset1 = getCpGsetCG(genome)
cpgset2 = getCpGsetALL(genome)

print("Pure CG coordinates in the toy genome:")
print(cpgset1)

print("CG coordinates in the toy genome possible with SNPs:")
print(cpgset2)
```

## Description

Injects SNPs from a VCF count file into a DNA sequence.



**Usage**

```
injectSNPsMAF(gensequence, frqcount, MAF = 0.01)
```

**Arguments**

gensequence	A string or <a href="#">DNAString</a> of the DNA sequence.
frqcount	File name of the allele count file produced by <code>vcftools</code> with <code>--counts</code> parameter. Alternatively, the file content can be provided as a character vector (see <a href="#">readLines</a> ).
MAF	SNPs with minor allele frequency at or above MAF are injected.

**Value**

Returns a string with the genome sequence with SNPs injected.

**Author(s)**

Andrey A Shabalin <[andrey.shabalin@gmail.com](mailto:andrey.shabalin@gmail.com)>

**See Also**

See [injectSNPs](#) for the standard analog function without MAF filtering.

**Examples**

```
gensequence1 = "AAAACAAAA"
frqcount = c(
  "CHROM\tPOS\tN_ALLELES\tN_CHR\t{ALLELE:COUNT}",
  "1\t6\t2\t1000\tA:400\tG:600",
  "1\t7\t2\t1000\tA:800\tC:200",
  "1\t9\t2\t1000\tA:900\tG:100")
MAF = 0.01

gensequence2 = injectSNPsMAF(gensequence1, frqcount, MAF)

### No CpGs without SNPs
show(gensequence1)
getCpGsetCG(gensequence1)

### SNPs create 1 CpG
show(gensequence2)
getCpGsetALL(gensequence2)
```

---

insilicoFASTQ

---

*Construct FASTQ File for In-silico Alignment Experiment*


---

**Description**

Creates a FASTQ file with all fragments of `fraglength` bp long.

**Usage**

```
insilicoFASTQ(con, gensequence, fraglength)
```

**Arguments**

con	A <a href="#">connection</a> object or a character string naming the output file. If the name ends with ".gz", a compressed file is created. An empty string can be used to output to the console.
gensequence	A string or <a href="#">DNAString</a> of the DNA sequence.
fraglength	Fragment length.

**Details**

The function a FASTQ file with all fragments of fraglength bp long from the forward strand of the DNA sequence.

**Value**

Returns a list with CpG coordinates for each genome sequence.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**Examples**

```
## There are four 4 bp fragments in a 7 basepair sequence:
insilicoFASTQ(con="", gensequence = "ABCDEFGG", fraglength=4)
```

---

isAbsolutePath      *Check if Path is Absolute.*

---

**Description**

Check whether a path is relative or absolute.

**Usage**

```
isAbsolutePath(path)
```

**Arguments**

path	Path to be tested.
------	--------------------

**Details**

The function is designed to work with both Windows and Unix paths.

**Value**

TRUE if the path is absolute, FALSE otherwise.

**Note**

This function improves upon the analog function in R.utils package. For instance, "~hi" is not an absolute path.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See also [makefullpath](#).

**Examples**

```
isAbsolutePath( "C:/123" ) # TRUE
isAbsolutePath( "~/123" ) # FALSE
isAbsolutePath( "~/123" ) # TRUE
isAbsolutePath( "/123" ) # TRUE
isAbsolutePath( "\\123" ) # TRUE
isAbsolutePath( "asd\\123" ) # FALSE
isAbsolutePath( "a\\123" ) # FALSE
```

---

madeBED

*Export MWAS results in BED format.*

---

**Description**

Functions for exporting MWAS results in BED format files.

Function `madeBED` saves MWAS findings in BED format for all variables (CpGs), while `madeBEDrange` selects only variables on a given chromosome between given locations.

Functions `madeBEDgraph` and `madeBEDgraphRange` do the same, but create a file in BedGraph format.

**Usage**

```
madeBED(x, filename)
madeBEDrange(x, filename, chr, start, end)
madeBEDgraph(x, filename)
madeBEDgraphRange(x, filename, chr, start, end)
```

**Arguments**

<code>x</code>	Name of MWAS directory (parameter <code>dirmwas</code> ) or list of RaMWAS parameters as described in the "RW6_param.Rmd" vignette. Try: <code>vignette("RW6_param", "ramwas")</code> .
<code>filename</code>	Name of the BED file to create. If file exists, it's overwritten.
<code>chr</code>	Chromosome name or number.
<code>start</code>	Start position of the genomic region of interest.
<code>end</code>	End position of the genomic region of interest.

**Details**

The function returns the MWAS results with locations.

**Value**

Returns a data.frame with BED file content:

chrom	Chromosome
chromStart	Start position
chromEnd	End position
name	Empty name column. BED format only
score	p-value

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
## Not run:
# Extract BED file using parameter vector
madeBED(param, "file.bed")
madeBEDrange(param, "file.bed", 1, 123321, 223321)

# Extract BED file using directory name
madeBED("/data/myMWAS", "file.bed")
madeBEDrange("/data/myMWAS", "file.bed", 1, 123321, 223321)

## End(Not run)
```

---

makefullpath

*Combine Path and Filename into Filename with Path*

---

**Description**

Combine a path with a filename into filename with path.

**Usage**

```
makefullpath(path, filename)
```

**Arguments**

path	Path, relative to which the filename is expected to be. Can be absolute, relative, or NULL.
filename	Can be just filename, include relative path, or include absolute path.

**Details**

Function returns filename if it includes absolute path or if path is NULL.

**Value**

Filename with the path included.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See also [isAbsolutePath](#).

**Examples**

```
makefullpath("dir1/dir2", "file.txt")
# "dir1/dir2/file.txt"

makefullpath("dir1/dir2", "dir3/file.txt")
# "dir1/dir2/dir3/file.txt"

# Path is ignored if the filename already includes absolute path

makefullpath("dir1/dir2", "/file.txt")
# "/file.txt"

makefullpath("dir1/dir2", "C:/file.txt")
# "C:/file.txt"
```

---

manhattan

*Fast Manhattan plot for Large Number of P-values*

---

**Description**

The function `manPlotFast` creates a Manhattan plot.

The function `manPlotPrepare` extracts necessary information from a vector of p-values sufficient for creating a Manhattan plot.

It optimized to work quickly even for tens of millions of p-values.

**Usage**

```
manPlotPrepare(
  pvalues,
  chr,
  pos,
  ismlog10 = FALSE,
  chrmargins = 5e6)

manPlotFast(
  man,
  ylim = NULL,
  colorSet = c('steelblue4', "#2C82D1", "#4CB2D1"),
  yaxmax = NULL,
  lwd = 3,
  axistep = 2,
  cex = 1)
```

**Arguments**

pvalues	Vector of p-values. As is (if <code>ismlog10 = FALSE</code> ) or minus log10 transformed (if <code>ismlog10 = TRUE</code> ).
chr, pos	Vectors indicating the chromosomes and genomic positions (in basepairs) for each p-value in pvalues.
ismlog10	Specifies whether the provides p-values (pvalues parameter) are minus log10 transformed ( $-\log_{10}(pv)$ )
chrmargins	The plot margins at the ends of chromosomes (in basepairs).
man	Object returned by <code>manPlotPrepare</code> .
ylim	Numeric vectors of length 2, giving the y coordinate range. Exactly as in <a href="#">Plotting Parameters</a> .
colorSet	Colors of points, rotating over chromosomes. Points for first chromosome have color <code>colorSet[1]</code> , next <code>colorSet[2]</code> , etc. Once the colors are exhausted, the colors are reused from the beginning.
yaxmax	Maximum reach of the y axis.
lwd	The line width. As in <a href="#">Graphics Parameters</a> .
axistep	Distance between axis label ticks for y axis.
cex	The size of Manhattan plot points. As in <a href="#">Graphics Parameters</a> .

**Details**

The function `manPlotFast` creates Manhattan plot.  
It requires the use of the function `manPlotPrepare` which extracts the necessary information from a vector of p-values sufficient for creating Manhattan plot.  
The resulting object is many times smaller than the vector of p-values.

**Value**

This function `manPlotPrepare` returns an object with information for creating Manhattan plot.

**Note**

The plot has no title. To add a title use `title`.

**Author(s)**

Andrey A Shabalin <[andrey.shabalin@gmail.com](mailto:andrey.shabalin@gmail.com)>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
# Simulate data (9 chromosomes, million tests each)
chr = rep(paste0('chr',1:9), each = 1e6)
pos = rep(1:1e6, 9)
pv = runif(9e6)^1.1

# Extract the Manhattan plot info
```

```
man = manPlotPrepare(pv, chr, pos, chrmargins = 1000)

# Create Manhattan plot
manPlotFast(man)
title("Manhattan plot")

# Size of p-values before extraction of Manhattan plot info
object.size(list(pv, chr, pos))

# Size of the Manhattan plot info object
object.size(man)
```

---

`mat2cols`*Split a Matrix into Column Vectors*

---

### Description

Internal function for splitting a matrix into column vectors.

### Usage

```
mat2cols(x)
```

### Arguments

x                    A matrix.

### Value

List of matrix columns.

### Author(s)

Andrey A Shabalin <andrey.shabalin@gmail.com>

### See Also

See vignettes: `browseVignettes("ramwas")`.

### Examples

```
# Sample data
data = matrix(1:12, nrow = 3)

# Split it
mat2cols(data)
```

---

orthonormalizeCovariates

*Orthonormalize Covariates*

---

### Description

Takes a matrix of data frame with covariates, adds a constant covariate (optional), and orthonormalizes the set.

### Usage

```
orthonormalizeCovariates(cvrt, modelhasconstant)
```

### Arguments

`cvrt`                    A matrix or data frame with covariates (one column per covariate).  
`modelhasconstant`       Set to TRUE to add a constant covariate into the set before normalization.

### Details

Factor variables are split into dummy variables before orthonormalization.  
The operation is performed via QR decomposition ([qr](#)).

### Value

Returns a matrix with orthogonal columns with unit length, whose columns spans the same space as the covariates plus a constant (if `modelhasconstant` is TRUE).

### Note

This function is used in several parts of the pipeline.

### Author(s)

Andrey A Shabalin <[andrey.shabalin@gmail.com](mailto:andrey.shabalin@gmail.com)>

### Examples

```
# Sample matrix of covariates
covariates = data.frame(a = 1:12, b = 12:1)

# Orthonormalizing Covariates
cvrtqr = orthonormalizeCovariates(covariates, modelhasconstant = TRUE)

# Checking the results (round to ignore rounding errors)
print( round(crossprod(cvrtqr),15) )

# Stop if not orthonormal
stopifnot(all.equal( crossprod(cvrtqr), diag(ncol(cvrtqr)) ))

# Example with a factor variable
groups = data.frame(gr = c("a","a","a","b","b","b","c","c","c"))
orthonormalizeCovariates(groups)
```



---

parameterDump	<i>Save Parameters in a Text File</i>
---------------	---------------------------------------

---

**Description**

Saves parameters in a text file, prioritizing those listed in topline.

**Usage**

```
parameterDump(dir, param, topline = NULL)
```

**Arguments**

dir	Directory to save the parameters to. The file is named "UsedSettings.txt".
param	A list with RaMWAS parameters. Or any list in general. For detailed description of all available parameters run: browseVignettes("ramwas").
topline	Names of the elements in param to save first (top of the file).

**Details**

This function is used internally by multiple RaMWAS functions to record parameters used to run the analysis.

**Value**

The function creates a file and returns nothing.

**Note**

This function is not intended to be run by the user.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: browseVignettes("ramwas")

**Examples**

```
param = ramwasParameters(  
  number = 123123,  
  integer = 312L,  
  textline = "Hi there",  
  characterVector = c("Hi", "Hi again", "Bye"),  
  dataframe = data.frame(a = 1:12, b = 12:1)  
)  
  
thedir = tempdir()  
  
parameterDump(thedir, param, c("integer", "characterVector"))
```

```
cat( readLines( paste0(thedir, "/UsedSettings.txt") ), sep = "\n")  
file.remove( paste0(thedir, "/UsedSettings.txt") )
```

---

parameterPreprocess     *Preprocess Pipeline Parameter List.*

---

## Description

Fill in missing parameters with default values, read supporting data files, make relative directory path parameters absolute.

## Usage

```
parameterPreprocess(param)
```

## Arguments

param	List with RaMWAS parameters. For detailed description of all available parameters run: <code>browseVignettes("ramwas")</code> .
-------	---------------------------------------------------------------------------------------------------------------------------------------

## Details

A number of common preprocessing steps necessary for parameters of multiple pipeline parts are combined in this function. The actions include

- Fill in default values for all missing parameters.
- Set *bamnames* parameter to the content *filebamlist* file (if *bamnames* was not set).
- Set *bam2sample* parameter to processed content of *filebam2sample* file (if *bam2sample* was not set).
- Set *covariates* parameter to the data frame from *filecovariates* file (if *covariates* was not set).
- Check parameters for consistency, i.e. that *modelcovariates* include only names of columns in *covariates*.
- Check that files *filecpgset* and *filenoncpgset* exist if the parameters are set.

## Value

Returns preprocessed list of parameters.

## Note

This function is not intended to be run by the user.

## Author(s)

Andrey A Shabalin <andrey.shabalin@gmail.com>

## See Also

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
param = ramwasParameters(  
  dirproject = "."  
)  
  
param2 = parameterPreprocess(param)  
  
print(param2)
```

---

parametersFromFile     *Scan Parameters From a R Code File*

---

**Description**

The pipeline parameters can be stored in a simple file, formatted as R code. The `parametersFromFile` function transforms them into a parameter list used by RaMWAS steps.

**Usage**

```
parametersFromFile(.parameterfile)
```

**Arguments**

`.parameterfile` Name of the file with the parameters set as R variables. See the example below.

**Details**

Variables with names starting with period (.) are ignored.

**Value**

Returns the list with all the variables set in the file.

**Note**

The file `.parameterfile` is executed as R code, so use only trusted parameter files.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```

filename = tempfile()

# Create a file with lines
# dirproject = "."
# modelcovariates = c("Age", "Sex")

writelines(
  con = filename,
  text = c(
    "dirproject = \".\" ",
    "modelcovariates = c(\"Age\", \"Sex\")"
  )
)

# Scan the file into a list
param = parametersFromFile(filename)

# Show the list
print(param)

file.remove(filename)

```

---

pipeline

*RaMWAS: High Level Pipeline Functions*

---

**Description**

These functions provide a simple way to run all steps of RaMWAS pipeline.

**Usage**

```

ramwas1scanBams(param)
pipelineProcessBam(bamname, param)
ramwas2collectqc(param)
ramwas3normalizedCoverage(param)
ramwas4PCA(param)
ramwas5MWAS(param)
ramwas6annotateTopFindings(param)
ramwas7ArunMWASes(param)
ramwas7BrunElasticNet(param)
ramwas7CplotByNCpGs(param)
ramwas7riskScoreCV(param)
ramwasSNPs(param)

```

**Arguments**

param	List with RaMWAS parameters. For detailed description of all available parameters run: <code>browseVignettes("ramwas")</code> .
bamname	Name of the BAM file to process. Can be absolute or relative to <code>dirbam</code> parameter (in param list).

**Details**

See vignettes for details: `browseVignettes("ramwas")`.

**Value**

Function `pipelineProcessBam` returns "OK. <bamname>" if no error occurred. Otherwise, returns text with error. Other functions return nothing.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
param = ramwasParameters(  
  dirbam = "/project/bams",  
  dirproject = "/project",  
  filebamlist = "000_list_of_files.txt",  
  scoretag = "AS",  
  minscore = 100,  
  cputhreads = 4,  
  filecpgset = "/RaMWAS/hg19_1kG_MAF_0.01_chr1-22_bowtie2_75bp.rds",  
  filenoncpgset = "/RaMWAS/hg19_1kG_MAF_0.01_chr1-22_bowtie2_75bp_nonCpG.rds",  
  maxrepeats = 3,  
  maxfragmentsize = 250,  
  minfragmentsize = 75,  
  filebam2sample = "000_list_of_files.txt",  
  filecovariates = "Covariates.txt",  
  modelcovariates = c("Age", "Sex"),  
  modeloutcome = "CellType",  
  modelPCs = 1,  
  cvnfolds = 10,  
  mmncpgs = 1000,  
  mmalpha = 0  
)  
  
## Not run:  
ramwas1scanBams(param)  
ramwas2collectqc(param)  
ramwas3normalizedCoverage(param)  
ramwas4PCA(param)  
ramwas5MWAS(param)  
ramwas6annotateTopFindings(param)  
ramwas7riskScoreCV(param)  
## End(Not run)
```

---

plotCV *Plotting Functions used in Cross Validation Analysis (Methylation Risk Score).*

---

### Description

The function `plotPrediction` plots cross validation predictions of a phenotype against true values of the phenotype with multiple summary stats in the title.

The function `plotCVcors` plots the predictive power (correlations) across predictions using various numbers of markers.

The function `plotROC` plots an ROC (Receiver operating characteristic) curve for predictions of a binary outcome.

### Usage

```
plotPrediction(
  param,
  outcome,
  forecast,
  cpgs2use,
  main,
  dfFull = NULL)
```

```
plotCVcors(cl, param)
```

```
plotROC(outcome, forecast)
```

### Arguments

<code>param</code>	List of parameters as described in the "RW6_param.Rmd" vignette. Try: <code>vignette("RW6_param", "ramwas")</code> . Only <code>modeloutcome</code> , <code>cvnfolds</code> and <code>mmalpha</code> elements are used.
<code>outcome</code>	Values of a phenotype. Must be binary for <code>plotROC</code> .
<code>forecast</code>	Predictions for the phenotype.
<code>cpgs2use</code>	Number of variables used for prediction (for the legend).
<code>main</code>	Part of the title (summary stats are added beneath).
<code>dfFull</code>	Number of degrees of freedom for the significance testing. Default is: <code>length(forecast) - 2</code>
<code>cl</code>	List with three elements: <ul style="list-style-type: none"> <li><code>x</code> - vector with the number of variables used for prediction</li> <li><code>corp</code> - Pearson correlations between the predictions and the true value of the phenotype.</li> <li><code>cors</code> - Spearman correlations between the predictions and the true value of the phenotype.</li> </ul>

### Details

The `plotROC` and `plot` has no title.

To add a title use `title`.

**Value**

The plotROC returns the area under the curve (AUC) for the ROC.

The plotPrediction function returns the list of calculated statistics printed in the title.

The plotCVcors returns nothing (NULL).

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: browseVignettes("ramwas").

**Examples**

```
# Sample data
n = 300
param = list(modeloutcome = "Age", mmalpha = 0, cvnfolders = 5)
outcome = rnorm(n, mean = 50, sd = 20)
forecast = outcome + rnorm(n, mean = 0, sd = 20)
cpgs2use = 1000
main = "Prediction success (simulated data)"

# Plot phenotype-prediction plot
plotPrediction(
  param,
  outcome,
  forecast,
  cpgs2use,
  main)

# Artificial data for plotCVcors()
cl = list(
  x = c(50, 100, 200, 500, 1000),
  corp = c(0.1, 0.6, 0.7, 0.85, 0.8),
  cors = c(0.1, 0.6, 0.7, 0.85, 0.8) + rnorm(5)*0.1)

# Plot prediction performance by the number of markers
plotCVcors(cl, param)

# Make the outcome binary for ROC plot
outcome = (outcome > 50)

# Plot ROC curve and calculate the AUC
plotROC(outcome, forecast)
```

---

plotFragmentSizeDistributionEstimate

*Estimate and plot Fragment Size Distribution.*

---

**Description**

RaMWAS functions for estimation and plotting of the fragment size distribution.

**Usage**

```
estimateFragmentSizeDistribution(frdata, seqLength)
plotFragmentSizeDistributionEstimate(
  frdata,
  estimate,
  col1 = "blue",
  col2 = "red")
```

**Arguments**

frdata	Distribution of distances from the starts of isolated reads to the respective CpGs.
seqLength	The length of sequenced part of the fragments. The fragments are assumed to not be smaller than seqLength.
estimate	Fragment size distribution estimate.
col1	Color of frdata points.
col2	Color of estimate curve.

**Value**

The function `estimateFragmentSizeDistribution` returns the estimate of the fragment size distribution.

**Note**

If the length of `frdata` is equal to `seqLength`, the fragments are assumed to all be of length `seqLength`.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
# Simulate data
x = 0:250
truemean = 1 - pnorm(x, mean = 150, sd = 50)
frdata = rpois(n = length(x), lambda = truemean*300)

# Estimate fragment size distribution
estimate = estimateFragmentSizeDistribution(frdata, seqLength = 50)

# Plot fragment size distribution estimate
plotFragmentSizeDistributionEstimate(frdata, estimate)
```



---

plotPC	<i>Plot Principal component (PC) Values (variation explained) and PC vectors (loadings)</i>
--------	---------------------------------------------------------------------------------------------

---

### Description

The function `plotPCvalues` plots PC values (variation explained).  
The function `plotPCvectors` plots PC vectors (loadings).

### Usage

```
plotPCvalues(values, n = 40, ylim = NULL, col = "blue")  
plotPCvectors(eigenvector, i, col = "blue1")
```

### Arguments

<code>values</code>	Vector of PC values.
<code>n</code>	Number of top PCs to plot.
<code>ylim</code>	Numeric vectors of length 2, giving the y coordinate range. Exactly as in <a href="#">Plotting Parameters</a> .
<code>col</code>	Color of the plotted points.
<code>eigenvector</code>	The i-th eigenvector. See <a href="#">eigen</a> .
<code>i</code>	Indicates loadings of which PC to plot.

### Value

This function creates a PC plot and returns nothing (NULL).

### Author(s)

Andrey A Shabalin <andrey.shabalin@gmail.com>

### See Also

See vignettes: `browseVignettes("ramwas")`.

### Examples

```
# Sample data  
# for 1000 observations and 10 samples  
m = 1000  
n = 10  
data = matrix(rnorm(n*m), nrow = m)  
  
# Covariance and eigenvalue decomposition  
covmat = crossprod(data)  
e = eigen(covmat)  
  
# Plot PC values  
plotPCvalues(e$values)
```

```
# Plot PC vectors
plotPCvectors(e$vector[,1], 1)
plotPCvectors(e$vector[,2], 2)
```

---

processCommandLine      *Scan Parameters From Command Line*

---

## Description

The pipeline parameters can be provided via command line.

For example:

```
R pipeline.r dirproject="/project" maxrepeats=0 modeloutcome="Age"
```

Each command line argument is treated as an R statement.

All variables defined this way are collected in a list which is returned.

## Usage

```
processCommandLine(.arg = NULL)
```

## Arguments

`.arg`                      Vector of command line parameters. Obtained from [commandArgs](#) if omitted.

## Details

If a command line argument defines variable "fileparam", it is assumed to be a filename, and the file with this name is scanned for extra pipeline parameters, as by [parametersFromFile](#).

## Value

Returns the list with all the variables set by the statement in the command line.

## Note

Variables with names starting with period (.) are ignored.

## Author(s)

Andrey A Shabalin <[andrey.shabalin@gmail.com](mailto:andrey.shabalin@gmail.com)>

## See Also

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
filename = tempfile()

# Assume command line with two components:
# dirproject="."
# modelcovariates=c("Age", "Sex")

arg = c(
  "dirproject = \".\""",
  "modelcovariates = c(\"Age\", \"Sex\")")

# Process the command line
param = processCommandLine(arg)

# Show the list
print(param)
```

---

pvalue2qvalue

*Calculate Benjamini-Hochberg q-values*

---

**Description**

Calculate Benjamini-Hochberg q-values for a vector of p-values.

**Usage**

```
pvalue2qvalue(pv, n = length(pv))
```

**Arguments**

pv	Vector of p-values.
n	If pv has only top p-values from a bigger set, n should indicate the number of tests performed.

**Details**

The q-values can be slightly conservative compared to other popular q-value calculation methods.

**Value**

Return a vector of q-values matching p-values in pv.

**Note**

The function runs faster if the vector pv is sorted.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**Examples**

```
pv = runif(20)^2

qv = pvalue2qvalue(pv)
```

---

QCs

*Quality Control Measures*


---

**Description**

RaMWAS calculates a number of QC measures for each BAM and saves them in R .rds files.

For full description of the QC measures and the plotting options run  
vignette("RW3\_BAM\_QCs")

**Usage**

```
qcmean(x)
## S3 method for class 'NULL'
qcmean(x)
## S3 method for class 'qcChrX'
qcmean(x)
## S3 method for class 'qcChrY'
qcmean(x)
## S3 method for class 'qcCoverageByDensity'
qcmean(x)
## S3 method for class 'qcEditDist'
qcmean(x)
## S3 method for class 'qcEditDistBF'
qcmean(x)
## S3 method for class 'qcFrwrev'
qcmean(x)
## S3 method for class 'qcHistScore'
qcmean(x)
## S3 method for class 'qcHistScoreBF'
qcmean(x)
## S3 method for class 'qcIsoDist'
qcmean(x)
## S3 method for class 'qcLengthMatched'
qcmean(x)
## S3 method for class 'qcLengthMatchedBF'
qcmean(x)
## S3 method for class 'qcNonCpGreads'
qcmean(x)

## S3 method for class 'qcHistScore'
plot(x, samplename="", xstep = 25, ...)
## S3 method for class 'qcHistScoreBF'
plot(x, samplename="", xstep = 25, ...)
## S3 method for class 'qcEditDist'
plot(x, samplename="", xstep = 5, ...)
```

```
## S3 method for class 'qcEditDistBF'
plot(x, samplename="", xstep = 5, ...)
## S3 method for class 'qcLengthMatched'
plot(x, samplename="", xstep = 25, ...)
## S3 method for class 'qcLengthMatchedBF'
plot(x, samplename="", xstep = 25, ...)
## S3 method for class 'qcIsoDist'
plot(x, samplename="", xstep = 25, ...)
## S3 method for class 'qcCoverageByDensity'
plot(x, samplename="", ...)
```

### Arguments

x	The QC object. See the examples below.
samplename	Name of the sample for plot title.
xstep	The distance between x axis ticks.
...	Parameters passed to the underlying <a href="#">plot</a> or <a href="#">barplot</a> function.

### Value

Function `qcmean` returns one value summary of most QC measures. Run `vignette("RW3_BAM_QCs")` for description of values returned by it.

### Author(s)

Andrey A Shabalin <[andrey.shabalin@gmail.com](mailto:andrey.shabalin@gmail.com)>

### See Also

See vignettes: `browseVignettes("ramwas")`.

### Examples

```
# Load QC data from a sample project
filename = system.file("extdata", "bigQC.rds", package = "ramwas")
qc = readRDS(filename)$qc

## The number of BAM files
cat("N BAMs:", qc$nbams)

## Total number of reads in the BAM file(s)
cat("Reads total:", qc$reads.total)

## Number of reads aligned to the reference genome
cat("Reads aligned:", qc$reads.aligned, "\n")
cat("This is ", qc$reads.aligned / qc$reads.total * 100,
    "% of all reads", sep="")

## Number of reads that passed minimum score filter and are recorded
cat("Reads recorded:", qc$reads.recorded, "\n")
cat("This is ", qc$reads.recorded / qc$reads.aligned * 100,
    "% of aligned reads", sep="")
```

```

## Number of recorded reads aligned to
## the forward and reverse strands respectively
cat("Reads on forward strand:", qc$frwrev[1], "\n")
cat("Reads on reverse strand:", qc$frwrev[2], "\n")
cat("Fraction of reads on forward strand:", qcmean(qc$frwrev), "\n")

## Distribution of the read scores
cat("Average alignment score:", qcmean(qc$hist.score1), "\n")
cat("Average alignment score, no filter:", qcmean(qc$bf.hist.score1), "\n")
par(mfrow=c(1,2))
plot(qc$hist.score1)
plot(qc$bf.hist.score1)

## Distribution of the length of the aligned part of the reads
cat("Average aligned length:", qcmean(qc$hist.length.matched), "\n")
cat("Average aligned length, no filter:",
    qcmean(qc$bf.hist.length.matched), "\n")
par(mfrow = c(1,2))
plot(qc$hist.length.matched)
plot(qc$bf.hist.length.matched)

## Distribution of edit distance between
## the aligned part of the read and the reference genome
cat("Average edit distance:", qcmean(qc$hist.edit.dist1), "\n")
cat("Average edit distance, no filter:", qcmean(qc$bf.hist.edit.dist1), "\n")
par(mfrow = c(1,2))
plot(qc$hist.edit.dist1)
plot(qc$bf.hist.edit.dist1)

## Number of reads after removal of duplicate reads
cat("Reads without duplicates:", qc$reads.recorded.no.repeats, "\n")
cat("This is ", qc$reads.recorded.no.repeats / qc$reads.recorded * 100,
    "% of aligned reads", "\n", sep="")
cat("Fraction of reads on forward strand (with duplicates):",
    qcmean(qc$frwrev), "\n")
cat("Fraction of reads on forward strand (without duplicates):",
    qcmean(qc$frwrev.no.repeats), "\n")

## Number of reads away from CpGs
cat("Non-CpG reads:", qc$cnt.nonCpG.reads[1], "\n")
cat("This is ", qcmean(qc$cnt.nonCpG.reads)*100, "% of recorded reads",
    sep="")

## Average coverage of CpGs and non-CpGs
cat("Summed across", qc$nbams, "bams", "\n")
cat("Average CpG coverage:", qc$avg.cpg.coverage, "\n")
cat("Average non-CpG coverage:", qc$avg.noncpg.coverage, "\n")
cat("Enrichment ratio:", qc$avg.cpg.coverage / qc$avg.noncpg.coverage)

## Coverage around isolated CpGs
plot(qc$hist.isolated.dist1)

## Fraction of reads from chrX and chrY
cat("ChrX reads: ", qc$chrX.count[1], ", which is ",
    qcmean(qc$chrX.count)*100, "% of total", sep=" ", "\n")
cat("ChrY reads: ", qc$chrY.count[1], ", which is ",
    qcmean(qc$chrY.count)*100, "% of total", sep=" ", "\n")

```

```
## Coverage vs. CpG density
cat("Highest coverage is observed at CpG density of",
    qcmean(qc$avg.coverage.by.density)^2)
plot(qc$avg.coverage.by.density)
```

---

qqPlotFast

*Fast QQ-plot for Large Number of P-values*


---

## Description

Function qqPlotFast creates a QQ-plot with a confidence band and an estimate of inflation factor lambda. It optimized to work quickly even for tens of millions of p-values.

## Usage

```
qqPlotPrepare(
  pvalues,
  ntests = NULL,
  ismlog10 = FALSE)
qqPlotFast(
  x,
  ntests = NULL,
  ismlog10 = FALSE,
  ci.level = 0.05,
  ylim = NULL,
  newplot = TRUE,
  col = "#D94D4C",
  cex = 0.5,
  yaxmax = NULL,
  lwd = 3,
  axistep = 2,
  col.band = "#ECA538",
  makelegend = TRUE,
  xlab = expression(
    paste("\u2013", " log"[10]*"(", italic("P"), ")"), null)),
  ylab = expression(
    paste("\u2013", " log"[10]*"(", italic("P"), ")"), observed)))
```

## Arguments

pvalues	Vector of p-values. As is (if ismlog10 = FALSE) or minus log10 transformed (if ismlog10 = TRUE).
ntests	If only significant p-values are provided, the total number of tests performed. By default ntests is equal to the length of pvalues.
ismlog10	Specifies whether the provides p-values (pvalues parameter) are minus log10 transformed ( $-\log_{10}(pv)$ )
x	Either a vector of p-values, as in qqPlotPrepare, or the object returned by qqPlotPrepare.
ci.level	Significance level of the confidence band. Set to NULL avoid plotting the confidence band.

ylim	Numeric vectors of length 2, giving the y coordinate range. Exactly as in <a href="#">Plotting Parameters</a> .
newplot	If TRUE, the function creates a new plot window.
col	The QQ-plot curve color.
col.band	Confidence band curve color.
cex	The size of QQ-plot points. As in <a href="#">Graphics Parameters</a> .
lwd	The line width. As in <a href="#">Graphics Parameters</a> .
axistep	Distance between axis label ticks for both axis.
yaxmax	Maximum reach of the y axis.
makelegend	If true, add legend to the plot.
xlab, ylab	Axis labels. As in <a href="#">plot function</a> .

### Details

The function `qqPlotFast` creates a QQ-plot.

The function `qqPlotPrepare` extracts the necessary information from a vector of p-values sufficient for creating QQ-plot.

The resulting object is many times smaller than the vector of p-values.

### Value

The function `qqPlotPrepare` returns an object with the necessary information from a vector of p-values sufficient for creating QQ-plot.

### Note

The plot has no title. To add a title use `title`.

### Note

The function works faster if the p-values are sorted.

### Author(s)

Andrey A Shabalin <[andrey.shabalin@gmail.com](mailto:andrey.shabalin@gmail.com)>

### See Also

See vignettes: `browseVignettes("ramwas")`.

### Examples

```
# Million p-values
n = 1e6

# Null p-values
pv = runif(n)

# QQ-plot should be nearly diagonal
qqPlotFast(pv)
title("QQ-plot")
```



```

# Size of p-values before extraction of QQ-plot info
object.size(pv)

# Extract the QQ-plot info
qq = qqPlotPrepare(pv)

# Size of the QQ-plot info object
object.size(qq)

# Create QQ-plot, it is the same
qqPlotFast(qq)

# Create QQ-plot with plotting parameters
qqPlotFast(qq, ylim = c(0,10), yaxmax = 9, axistep = 3, lwd = 3, cex = 1)

```

---

```

ramwas0createArtificialData
      Create Artificial Data Set

```

---

## Description

Creates a set of artificial BAM files and supplementary files which can be used to test run the pipeline. The BAMs contain reads aligned only to one human chromosome, with methylation effects embedded for simulated age and case-control status.

## Usage

```

ramwas0createArtificialData(dir,
                             nsamples = 20,
                             nreads = 1e6,
                             ncpgs = 500e3,
                             randseed = 18090212,
                             threads = 1)

```

## Arguments

<code>dir</code>	Directory for generated RaMWAS project files and BAMs.
<code>nsamples</code>	Number of samples/BAMs to create.
<code>nreads</code>	Number of reads in each BAM file.
<code>ncpgs</code>	Number of CpGs in the generated genome (with a single chromosome).
<code>randseed</code>	Random number generator seed for consistency of the output.
<code>threads</code>	Number of CPU cores to use for data generation.

## Details

The function generates a number of files within `dir` directory.

1. `bam_list.txt` - list of created BAM files. To be used in `filebamlist` and `filebam2sample` parameters in the pipeline.

2. covariates.txt - table with age and sex status covariates. For use in filecovariates parameter in the pipeline.
3. Single\_chromosome.rds - CpG location file with the selected chromosome only.
4. bams - directory with all the BAM files.

The generated BAMs have 600 CpGs affected by sex, namely fully methylated or not methylated at all, depending on sex. The methylation level of 1% of all CpGs is affected by age. The methylation of those CpGs is equal to  $\text{age}/100$  or  $1-\text{age}/100$ . The age is generated randomly in the range from 20 to 80.

### Value

The function creates multiple files but returns no value.

### Author(s)

Andrey A Shabalin <andrey.shabalin@gmail.com>

### See Also

See vignettes: `browseVignettes("ramwas")`.

### Examples

```
### Location for the artificial project
dr = paste0(tempdir(), "/simulated_project")

ramwas0createArtificialData(
  dr,
  nsamples = 4,
  nreads = 10e3,
  ncpGs = 1e3)

# Artificial project files created in:
dr
# The generated files are:"
as.matrix(list.files(dr, recursive=TRUE))

### Clean up
unlink(paste0(dr,"/*"), recursive=TRUE)
```

---

ramwasAnnotateLocations

*Extract Biomart Annotation for a Vector of Locations.*

---

### Description

Calls biomart annotation database for a vector of locations and assigns the tracks to the locations.

### Usage

```
ramwasAnnotateLocations(param, chr, pos)
```

**Arguments**

param	List of parameters as described in the "RW6_param.Rmd" vignette. Try: vignette("RW6_param", "ramwas").
chr	A vector of chromosome names or numbers.
pos	A vector of genomic locations on the chromosomes.

**Details**

This function is used internally by RaMWAS annotation step.

**Value**

An annotation table, on line per supplied location.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: browseVignettes("ramwas")

**Examples**

```
bihost = "grch37.ensembl.org"
bimart = "ENSEMBL_MART_ENSEMBL"
bidataset = "hsapiens_gene_ensembl"
biattributes = c("hgnc_symbol", "entrezgene", "strand")
bifilters = list(with_hgnc_trans_name = TRUE)
biflank = 0

param = ramwasParameters(
  bihost = bihost,
  bimart = bimart,
  bidataset = bidataset,
  biattributes = biattributes,
  bifilters = bifilters,
  biflank = biflank)

# Test a location
chr = "chr1"
pos = 15975530
## Not run:
ramwasAnnotateLocations(param, chr, pos)

## End(Not run)
```

---

`ramwasParameters`*Function for Convenient Filling of the RaMWAS Parameter List.*

---

### Description

RaMWAS parameter vector which is used by major functions of the pipeline is a regular R list and setting it does not require a special function. However, using this function makes it much simpler in RStudio as the names and role of every parameter is showed in the RStudio IDE.

### Usage

```
ramwasParameters(  
  dirproject,  
  dirfilter,  
  dirrbam,  
  dirrqc,  
  dirqc,  
  dircoveragenorm,  
  dirtemp,  
  dirpca,  
  dirmwas,  
  dircv,  
  dirbam,  
  filebamlist,  
  bamnames,  
  filebam2sample,  
  bam2sample,  
  filecpgset,  
  filenoncpgset,  
  filecovariates,  
  covariates,  
  cputhreads,  
  diskthreads,  
  usefilelock,  
  scoretag,  
  minscore,  
  maxrepeats,  
  minavgcpgcoverage,  
  minnonzerosamples,  
  buffersize,  
  doublesize,  
  modelcovariates,  
  modeloutcome,  
  modelPCs,  
  modelhasconstant,  
  qqplottitle,  
  toppvthreshold,  
  mmncpgs,  
  mmalpha,  
  cvnfolds,  
  bihost,
```

```

    bimart,
    bidataset,
    biattributes,
    bifilters,
    biflank,
    fileSNPs,
    dirSNPs,
    ...)

```

### Arguments

dirproject	The project directory. Default is current directory. Files specified by "file*" parameters are looked for here, unless they have full path specified.
dirfilter	By default, the same as "dirproject". All files created by RaMWAS are created within this directory. If the user wants to test different read filtering rules, they can dirfilter to TRUE. This will set it to something like "Filter_MAPQ_4", where "MAPQ" is the BAM field used for filtering and "4" is the threshold.
dirrbam	Directory where RaMWAS saves RaMWAS raw data files (read start locations) after scanning BAMs. It is "rds_rbam" by default and located in "dirfilter".
dirrqc	Directory where RaMWAS saves QC files in R format after scanning BAMs. It is "rds_qc" by default and located in "dirfilter".
dirqc	Directory where RaMWAS saves QC plots and text files (BAM QC info) after scanning BAMs. It is "qc" by default and located in "dirfilter".
dircoveragenorm	Directory where RaMWAS saves coverage matrix at Step 3 of the pipeline. It is "coverage_norm_123" by default (123 is the number of samples) and located in "dirfilter".
dirtemp	Directory where RaMWAS stores temporary files during construction of coverage matrix at Step 3 of the pipeline. It is "temp" by default and located in "dircoveragenorm". For better performance it can be set to a location on a different hard drive than "dircoveragenorm".
dirpca	Directory where RaMWAS saves results of PCA analysis at Step 4 of the pipeline. It is "PCA_12_cvrts_0b0a0c" by default and located in "dircoveragenorm", where 12 is the number of covariates regressed out and "0b0a0c" is a unique code to differentiate different sets of 12 covariates.
dirmwas	Directory where RaMWAS saves results of MWAS analysis at Step 5 of the pipeline. It is "Testing_age_7_PCs" by default and located in "dirpca", where "age" is the phenotype being tested and "7" is number of top PCs included in the model.
dircv	Directory where RaMWAS saves results of Methylation Risk Score analysis at Step 7 of the pipeline. It is "CV_10_folds" by default and located in "dirmwas", where 10 is number of folds in N-fold cross validation.
dirbam	Location of BAM files. If not absolute, it is considered to be relative to "dirproject".

filebamlist	If defined, must point to a text file with one BAM file name per line. BAM file names may include path, relative to "dirbam" or absolute.
bamnames	A character vector with BAM file names. Not required if "filebamlist" is specified. BAM file names may include path, relative to "dirbam" or absolute.
filebam2sample	Allows multiple BAMs contain information about common sample. Must point to a file with lines like "sample1=bam1,bam2,bam3".
bam2sample	Allows multiple BAMs contain information about common sample. Not required if "filebam2sample" is specified. Must be a list like <code>list(sample1 = c("bam1", "bam2", "bam3"), sample2 = "bam2")</code>
filecpgset	Name of the file storing a set of CpGs.
filenoncpgset	If defined, must point to a file storing vetted locations away from any CpGs.
filecovariates	Name of the file containing phenotype and covariates for the available samples. If the file has extension ".csv", it is assumed to be comma separated, otherwise - tab separated.
covariates	Data frame with phenotype and covariates for the available samples. Not required if "filecovariates" is specified.
cputhreads	Maximum number of CPU intensive tasks running in parallel. Set to the number of CPU cores by default.
diskthreads	Maximum number of disk intensive tasks running in parallel. Set to 2 by default.
usefilelock	If TRUE, parallel jobs are prevented from simultaneous access to file matrices. Can improve performance on some systems.
scoretag	Reads from BAM files are filtered by this tag. The "minscore" parameter defines the minimum admissible score.
minscore	Reads from BAM files with score "scoretag" below this are excluded.
maxrepeats	Duplicate reads (reads with the same start position and direction) in excess of this limit are removed.
minavgcpgcoverage	CpGs with average coverage below this threshold are removed.
minnonzerosamples	CpGs with fraction of samples with non-zero coverage below this threshold are removed.
buffersize	Coverage matrix transposition is performed using buffers of this size. Larger "buffersize" improves speed of Step 3 of the pipeline, but requires more memory. Default is 1e9, i.e. 1 GB.
doublesize	The coverage matrix is stored with this number of bytes per value. Set to 8 for full (double) precision. Set to 4 to use single precision and create 50% smaller coverage filematrix.
modelcovariates	Names of covariates included in PCA and MWAS.
modeloutcome	Name of the outcome variable for MWAS.
modelPCs	Number of principal components accounted for in MWAS.

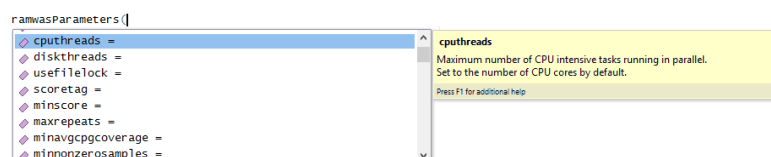
modelhasconstant	By default, the tested linear model includes a constant. To exclude it, set "modelhasconstant" parameter to FALSE.
qqplottitle	The title of the QQ-plot produced by MWAS (step 4 of the pipeline).
toppvthreshold	Determines the number of top MWAS results saved in text file. If it is 1 or smaller, it defines the p-value threshold. If larger than 1, it defines the exact number of top results.
mmncpgs	Parameter for multi-marker elastic net cross validation (MRS). Defines the number of top CpGs on which to train the elastic net. Can be set of a vector of multiple values, each is tested separately.
mmalpha	Parameter for multi-marker elastic net cross validation (MRS). Elastic net mixing parameter alpha. Set to 0 by default.
cvnfolds	Parameter for multi-marker elastic net cross validation (MRS). The number of folds in the N-fold cross validation.
bihost	Parameter for BiomaRt annotation (Step 6 of the pipeline). BioMart host site. Set to "grch37.ensembl.org" by default.
bimart	Parameter for BiomaRt annotation (Step 6 of the pipeline). BioMart database name, see <a href="#">listMarts</a> . Set to "ENSEMBL_MART_ENSEMBL" by default.
bidataset	Parameter for BiomaRt annotation (Step 6 of the pipeline). BioMart data set, see <a href="#">listDatasets</a> . Set to "hsapiens_gene_ensembl" by default.
biattributes	Parameter for BiomaRt annotation (Step 6 of the pipeline). BioMart attributes of interest, see <a href="#">listAttributes</a> . Set to c("hgnc_symbol", "entrezgene", "strand") by default.
bifilters	Parameter for BiomaRt annotation (Step 6 of the pipeline). BioMart filters (if any), see <a href="#">listFilters</a> . Set to list(with_hgnc_transcript_name=TRUE) by default ignore genes without names.
biflank	Parameter for BiomaRt annotation (Step 6 of the pipeline). Allowed distance between CpGs and genes or other annotation track elements. Set to 0 by default, requiring direct overlap.
fileSNPs	Name of the filematrix with genotype (SNP) data. The filematrix dimensions must match the coverage matrix.
dirSNPs	Directory where RaMWAS saves the results of joint methylation-genotype analysis.
...	Any other named parameters can be added here.

## Details

The function simply collects all the parameters in a list.

The main benefit of the function is that the user does not need to memorize the names of RaMWAS parameters.

Here is how it helps in RStudio:



**Value**

List with provided parameters.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
ramwasParameters(dirproject = ".", cputhreads = 4)
```

---

rowcolSumSq

*Form Row and Column Sums of Squares*

---

**Description**

Form row and column sums of squares for numeric matrices. The functions are introduced as faster analogs of `rowSums(x^2)` and `colSums(x^2)` calls.

**Usage**

```
rowSumsSq(x)  
colSumsSq(x)
```

**Arguments**

`x` Numeric matrix.

**Details**

The function is implemented in C for better performance.

**Value**

Return a vector of sums of values in each row/column for matrix `x` (`rowSumsSq/colSumsSq`).

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See [rowSums](#) and [colSums](#) for simple (not squared) row/column sums.



**Examples**

```
x = matrix( 1:99, 9, 11)

# Calculate sums of squared elements in each row
rsum2 = rowSumsSq(x)

# Compare with alternative calculation
stopifnot( all.equal( rsum2, rowSums(x^2) ))

# Calculate sums of squared elements in each column
csum2 = colSumsSq(x)

# Compare with alternative calculation
stopifnot( all.equal( csum2, colSums(x^2) ))
```

---

rwDataClass-class      *Class for Accessing Data (Coverage) Matrix*

---

**Description**

This class is a wrapper for accessing the data (coverage) matrix. It automatically subsets the samples to those listed in the covariates. Data access function imputes missing values and can residualize the variables.

**Extends**

rwDataClass is a reference classes (see [envRefClass](#)).

**Fields**

**fmdata:** Filematrix object for the data matrix.  
 Not intended to be accessed directly.

**samplenames:** Vector of sample names.

**nsamples:** Number of samples

**ncpgs:** Number of variables (CpG sites) in the data matrix.

**ndatarows:** Number of variables in the data matrix (may be bigger than the number of samples).

**rowsubset:** Indices of samples in the data matrix.

**cvrtqr:** Matrix of orthonormalized covariates.

**Methods**

**initialize(param = NULL, getPCs = TRUE, lockfile = NULL):** Create the data access class.  
 'param' should contain the RaMWAS parameter vector.  
 'getPCs' indicates if the covariate set should include Principal components.  
 'lockfile' is the 'lockfile' parameter used in accessing the data filematrix.

**open(param = NULL, getPCs = TRUE, lockfile = NULL):** The same as 'initialize' method, but for already created object.

**close():** Clears the object. Closes the filematrix.

**getDataRez(colset, resid = TRUE):** Extracts data for variables indexed by 'colset'.  
 The data is residualized unless resid = FALSE.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`

**Examples**

```
# Create an empty rwDataClass
data = new("rwDataClass")

## Not run:
# Connect to the data
data$open(param)

# Create a rwDataClass and connect to the data

data = new("rwDataClass", param = param)

# close the object
data$close()

## End(Not run)
```

---

subsetData

*Subset a data matrix and locations*

---

**Description**

Subset a data (coverage) matrix and corresponding matrix of locations to a specified set of locations.

**Usage**

```
subsetCoverageDirByLocation(x, chr, start, targetdir)
```

**Arguments**

x	Name of data (coverage) directory or list of RaMWAS parameters as described in the "RW6_param.Rmd" vignette. Try: <code>vignette("RW6_param", "ramwas")</code> .
chr	Vector of chromosome names or numbers.
start	Start positions of the CpGs of interest.
targetdir	Directory name for the new (subset) data matrix and locations.

**Value**

The function returns nothing.

**Author(s)**

Andrey A Shabalin <andrey.shabalin@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

**Examples**

```
x = "/data/myCoverageMatrix"
chr = c("chr1", "chr2", "chr3")
start = c(12345, 123, 12)
targetdir = "/data/subsetCoverageMatrix"

## Not run:
subsetCoverageDirByLocation(x, chr, start, targetdir)

## End(Not run)
```

---

testPhenotype	<i>Test the Phenotype of Interest for Association with Methylation Coverage.</i>
---------------	----------------------------------------------------------------------------------

---

**Description**

An internal, function for fast association testing. It tests the phenotype of interest for association with methylation coverage (columns of the data parameter).

**Usage**

```
testPhenotype(phenotype, data1, cvrtqr)
```

**Arguments**

phenotype	Vector with phenotype. Can be numerical, character, or factor vector.
data1	Matrix with data (normalized coverage), one variable (CpG) per column.
cvrtqr	Orthonormalized covariates, one covariate per column. See <a href="#">orthonormalizeCovariates</a> .

**Details**

The testing is performed using matrix operations and C/C++ code, emplying an approach similar to that in **MatrixEQTL**.

**Value**

If the phenotype is numerical, the output is a list with

correlation	Correlations between residualized phenotype and data columns.
tstat	Corresponding T-statistics
pvalue	Corresponding P-values
nVarTested	Always 1
dfFull	Number of degrees of freedom of the T-test

If the phenotype is a factor (or character)

Rquared	R-squared for the residualized ANOVA F-test.
Fstat	Corresponding F-test
pvalue	Corresponding P-values
nVarTested	First number of degrees of freedom for the F-test. Equal to the number of factor levels reduced by 1
dfFull	Second number of degrees of freedom for the F-test.

**Note**

This function is used in several parts of the pipeline.

**Author(s)**

Andrey A Shabalín <andrey.shabalín@gmail.com>

**See Also**

See vignettes: `browseVignettes("ramwas")`.

Also check [orthonormalizeCovariates](#).

**Examples**

```
### Generate data inputs
# Random data matrix with signal in the first column
data = matrix(runif(30*5), nrow = 30, ncol = 5)
data[,1] = data[,1] + rep(0:2, each = 10)

# Two random covariates
cvrt = matrix(runif(2*30), nrow = 30, ncol = 2)
cvrtqr = orthonormalizeCovariates(cvrt)

### First, illustrate with numerical phenotype
# Numerical, 3 value phenotype
phenotype = rep(1:3, each = 10)

# Test for association
output = testPhenotype(phenotype, data, cvrtqr)

# Show the results
print(output)

# Comparing with standard R code for the first variable
summary(lm( data[,1] ~ phenotype + cvrt ))

### First, illustrate with numerical phenotype
# Categorical, 3 group phenotype
phenotype = rep(c("Normal", "Sick", "Dead"), each = 10)

# Test for association
output = testPhenotype(phenotype, data, cvrtqr)
```

```
# Show the results  
print(output)
```

```
# Comparing with standard R code for the first variable  
anova(lm( data[,1] ~ cvrt + phenotype ))
```

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