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Contents

1	Introduction					
2	liminaries	3				
	2.1	Citing QuasR	3			
	2.2	Installation	4			
	2.3	Loading of $QuasR$ and other required libraries	4			
	2.4	How to get help	4			
3	Qui	ck Start	6			
	3.1	A brief introduction to R	6			
	3.2	Sample $QuasR$ session	7			
4	Quc	asR Overview	9			
5	Example tasks 1					
	5.1	Create a sample file	11			
	5.2	Create an auxiliary file (optional)	12			
	5.3	Select the reference genome	13			
	5.4	Sequence data pre-processing	14			
6	Exa	imple workflows	L 6			
	6.1	ChIP-seq: Measuring protein-DNA binding and chromatin mod-				
		ifications	16			
		6.1.1 Align reads using the qAlign function	16			
		6.1.2 Create a quality control report	18			
		6.1.3 Alignment statistics	18			
		1 0 0	18			
		6.1.5 Count alignments using qCount	19			
		6.1.6 Create a genomic profile for a set of regions using qProfile 2	21			

		6.1.7 Using a <i>BSgenome</i> package as reference genome	23
	6.2	RNA-seq: Gene expression profiling	24
	6.3	Bis-seq: Measuring DNA methylation	27
	6.4	Allele-specific analysis	31
7	Des	cription of Individual $QuasR$ Functions	35
	7.1	preprocessReads	35
	7.2	qAlign	35
	7.3	qProject class	36
	7.4	qQCReport	37
	7.5	alignmentStats	40
	7.6	qExportWig	41
	7.7	qCount	41
		7.7.1 Determination of overlap	41
		7.7.2 Running modes of qCount	42
	7.8	qProfile	43
	7.9	qMeth	43
8	Sess	sion information	44

1 Introduction

The QuasR package (short for Quantify and annotate short reads in R) integrates the functionality of several R packages (such as IRanges and Rsamtools) and external software (e.g. bowtie, through the Rbowtie package). The package aims to cover the whole analysis workflow of typical ultra-high throughput sequencing experiments, starting from the raw sequence reads, over pre-processing and alignment, up to quantification. A single R script can contain all steps of a complete analysis, making it simple to document, reproduce or share the workflow containing all relevant details.

The current QuasR release supports the analysis of single read and pairedend ChIP-seq (chromatin immuno-precipitation combined with sequencing), RNAseq (gene expression profiling by sequencing of RNA) and Bis-seq (measurement of DNA methylation by sequencing of bisulfite-converted genomic DNA) experiments.

2 Preliminaries

2.1 Citing QuasR

If you use QuasR [1] in your work, you can cite it as follows:

> citation("QuasR")

Please use the QuasR reference below to cite the software itself. If you were using qAlign with Rbowtie as aligner, it can be cited as Langmead et al. (2009) (unspliced alignments) or Au et al. (2010) (spliced alignments).

Anita Lerch, Dimos Gaiditzis and Michael Stadler (2012). QuasR: Quantify and Annotate Short Reads in R. R package version 1.0.9 (unpublished)

Langmead B, Trapnell C, Pop M, Salzberg SL. Ultrafast and memory-efficient alignment of short DNA sequences to the human genome. Genome Biology 10(3):R25 (2009).

Au KF, Jiang H, Lin L, Xing Y, Wong WH. Detection of splice junctions from paired-end RNA-seq data by SpliceMap. Nucleic Acids Research, 38(14):4570-8 (2010).

This free open-source software implements academic research by the authors and co-workers. If you use it, please support the project by citing the appropriate journal articles.

2.2 Installation

QuasR is a package for the R computing environment and it is assumed that you have already installed R. See the R project at http://www.r-project.org. To install the latest version of QuasR, you will need to be using the latest version of R. QuasR is part of the Bioconductor project at http://www.bioconductor.org. To get QuasR together with its dependencies you can use

```
> source("http://www.bioconductor.org/biocLite.R")
> biocLite("QuasR")
```

Bioconductor works on a 6-monthly official release cycle. As with other Bioconductor packages, there are always two versions of *QuasR*. Most users will use the current official release version, which will be installed by biocLite if you are using the current version of R. There is also a developmental version of *QuasR* that includes new features due for the next official release. The developmental version will be installed if you are using the developmental version of R. The official release version always has an even second number (for example 0.2.1), whereas the developmental version has an odd second number (for example 0.3.6).

2.3 Loading of *QuasR* and other required libraries

In order to run the code examples in this vignette, the QuasR library and a few additional libraries need to be loaded:

- > library(QuasR)
- > library(BSgenome)
- > library(Rsamtools)
- > library(rtracklayer)
- > library(GenomicFeatures)
- > library(Gviz)

2.4 How to get help

Most questions about QuasR will hopefully be answered by the documentation or references. If you've run into a question which isn't addressed by the documentation, or you've found a conflict between the documentation and software itself, then there is an active support community which can offer help. The authors of the package (maintainer: Michael Stadler <michael.stadler@fmi.ch>) always appreciate receiving reports of bugs in the package functions or in the documentation. The same goes for well-considered suggestions for improvements. Any other questions or problems concerning QuasR should be sent to the Bioconductor mailing list bioconductor@stat.math.ethz.ch. To subscribe to the mailing list, see https://stat.ethz.ch/mailman/listinfo/bioconductor. Please send requests for general assistance and advice to the mailing list rather than to the individual authors. Users posting to the mailing list for the first

time should read the helpful posting guide at http://www.bioconductor.org/doc/postingGuide.html. Note that each function in QuasR has it's own help page, as described in the section 3.1. Mailing list etiquette requires that you read the relevant help page carefully before posting a problem to the list.

3 Quick Start

3.1 A brief introduction to R

If you already use R and know about its command line interface, just skip this section and continue with section 3.2 on page 7.

The structure of this vignette and in particular this section is based on the excellent user guide of the *limma* package, which we would like to hereby acknowledge. R is a program for statistical computing. It is a command-driven language meaning that you have to type commands into it rather than pointing and clicking using a mouse. In this guide it will be assumed that you have successfully downloaded and installed R from http://www.r-project.org as well as *QuasR* (see section 2.2). A good way to get started is to type

> help.start()

at the R prompt or, if you're using R for Windows, to follow the drop-down menu items $Help \succ Html\ help$. Following the links $Packages \succ QuasR$ from the html help page will lead you to the contents page of help topics for functions in QuasR. Before you can use any QuasR commands you have to load the package by typing

> library(QuasR)

at the R prompt. You can get help on any function in any loaded package by typing ? and the function name at the R prompt, for example

> ?preprocessReads

or equivalently

> help("preprocessReads")

for detailed help on the preprocessReads function. The individual function help pages are especially important for listing all the arguments which a function will accept and what values the arguments can take.

A key to understanding R is to appreciate that anything that you create in R is an *object*. Objects might include data sets, variables, functions, anything at all. For example

will create a variable x and will assign it the value 2. At any stage of your R session you can type

to get a list of all the objects you have created. You can see the contents of any object by typing the name of the object at the prompt. The following command will print out the contents of \mathbf{x} :

We hope that you can use QuasR without having to spend a lot of time learning about the R language itself but a little knowledge in this direction will be very helpful, especially when you want to do something not explicitly provided for in QuasR or in the other Bioconductor packages. For more details about the R language see An Introduction to R which is available from the online help. For more background on using R for statistical analysis see [2].

3.2 Sample QuasR session

This is a quick overview of what an analysis could look like for users preferring to jump right into an analysis. The example uses data that is provided with the *QuasR* package, which is first copied to the current working directory, into a subfolder called "extdata":

```
> file.copy(system.file(package="QuasR", "extdata"), ".", recursive=TRUE)
[1] TRUE
```

The sequence files to be analyzed are listed in sampleFile (see section 5.1 on page 11 for details). The sequence reads will be aligned using bowtie [3] (from the *Rbowtie* package [4]) to a small reference genome (consisting of three short segments from the hg19 human genome assembly, available in full for example in the *BSgenome.Hsapiens.UCSC.hg19* package). Make sure that you have sufficient disk space, both in your R temporary directory (tempdir) as well as to store the resulting alignments (see section 7.2).

```
> sampleFile <- "extdata/samples_chip_single.txt"</pre>
> genomeFile <- "extdata/hg19sub.fa"
> proj <- qAlign(sampleFile, genomeFile)
nodeNames
george2
> proj
Project: qProject
 Options
           : maxHits
                              : 1
             paired
                              : no
             splicedAlignment: FALSE
             bisulfite
                              : no
             snpFile
                              : none
 Aligner
           : Rbowtie v1.0.3 (parameters: -m 1 --best --strata)
 Genome
           : /tmp/Rtmp3ckjsv/Rbuild484015dfa.../hg19sub.fa (file)
 Reads
           : 2 files, 2 samples (fastq format):

    chip_1_1.fq.bz2 Sample1 (phred33)

   2. chip_2_1.fq.bz2 Sample2 (phred33)
```

Genome alignments: directory: same as reads

- 1. chip_1_1_6cdf5473cc09.bam
- 2. chip_2_1_6cdf7f640d43.bam

Aux. alignments: none

The proj object keeps track of all the information of a sequencing experiment, for example where sequence and alignment files are stored, and what aligner and reference genome was used to generate the alignments.

Now that the alignments have been generated, further analyses can be performed. A quality control report is saved to the "extdata/qc_report.pdf" file using the qQCReport function.

> qQCReport(proj, "extdata/qc_report.pdf")

The number of alignments per promoter region is quantified using qCount. Genomic coordinates for promoter regions are imported from a gtf file (annotFile) into the GRanges-object with the name promReg:

- > library(rtracklayer)
- > library(GenomicFeatures)
- > annotFile <- "extdata/hg19sub_annotation.gtf"</pre>
- > txStart <- import.gff(annotFile, format="gtf", asRangedData=FALSE,
- + feature.type="start_codon")
- > promReg <- promoters(txStart, upstream=500, downstream=500)</pre>
- > names(promReg) <- mcols(promReg)\$transcript_name</pre>
- > promCounts <- qCount(proj, query=promReg)</pre>
- > promCounts

-			
	${\tt width}$	Sample1	Sample2
TNFRSF18-003	1000	20	4
TNFRSF18-002	1000	20	4
TNFRSF18-001	1000	20	4
TNFRSF4-001	1000	5	2
SDF4-007	1000	8	2
SDF4-001	1000	8	2
SDF4-002	1000	8	2
SDF4-201	1000	8	2
B3GALT6-001	1000	25	274
RPS7-001	1000	121	731
RPS7-008	1000	121	731
RPS7-009	1000	121	731
RPS7-005	1000	121	731
C3orf10-201	1000	176	496
C3orf10-001	1000	176	496
AC034193.1-201	1000	5	2
VHL-001	1000	61	336
VHL-002	1000	61	336
VHL-201	1000	61	336

4 QuasR Overview

The following scheme shows the major components of QuasR and their relationships:

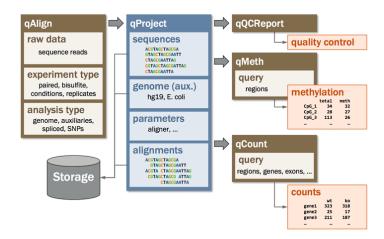


Figure 1: QuasR package overview

QuasR works with data (sequences and alignments, reference genome, etc.) that are stored as files on your storage (the gray cylinder on the lower left of Figure 1). QuasR does not need a database management system, or these files to be named and organized according to a specific scheme.

In order to keep track of directory paths during an analysis, *QuasR* makes use of a qProject object that is returned by the qAlign function, which at the minimum requires two inputs: the name of a samples text file (see section 5.1 for details), and the reference genome for the alignments (see section 5.3).

The qProject object is the main argument passed to subsequent functions such as qQCReport and qCount. The qProject object contains all necessary information on the current project and eliminates the need to repeatedly enter the same information. All functions that work on qProject objects can be recognized by their names starting with the letter q.

Read quantification (apart from quantification of methylation which has its own function qMeth) is done using the qCount function: It counts the alignments in regions of interest (e.g. promoters, genes, exons, etc.) and produces a count table (regions in rows, samples in columns) for further visualization and analysis. The count table can also be used as input to a statistical analysis using packages such as edqeR [5] or DESeq [6].

In summary, a typical QuasR analysis consists of the following steps (some

of them are optional):

- preprocessReads (optional): Remove adapters from start or end of reads, filter out reads of low quality, short length or low complexity (section 5.4 on page 14).
- Prepare *samples file*: List sequence files or alignments, provide sample names (section 5.1 on page 11).
- Prepare auxiliary file (optional): List additional reference sequences for alignment of reads not matching the reference genome (section 5.2 on page 12).
- qAlign: Create qProject object and specify project parameters. Also download BSgenome package, create aligner indices and align reads if not already existing (7.2 on page 35).
- qQCReport (optional): Create quality control report with plots on sequence qualities and alignment statistics (section 7.4 on page 37).
- qExportWig (optional): Export genomic alignments as wiggle tracks for genome browser visualization (section 7.6 on page 41).
- qCount: Quantify alignments in regions of interest (section 7.7 on page 41).

Recurrent example tasks that may be part of any typical analysis are described in section 5 starting on page 11. Example workflows for specific experiment types (ChIP-seq, RNA-seq and Bis-seq) are described in section 6 starting on page 16.

5 Example tasks

5.1 Create a sample file

The sample file is a tab-delimited text file with two or three columns. The first row contains the column names: For a single read experiment, these are 'File-Name' and 'SampleName'; for a paired-end experiment, these are 'File-Name1', 'FileName2' and 'SampleName'. If the first row does not contain the correctly spelled column names, *QuasR* will not accept the samples file. Subsequent rows contain the input sequence files.

Here are examples of such sample files for a single read experiment:

FileName	SampleName
chip_1_1.fq.bz2	Sample1
chip_2_1.fq.bz2	Sample2

and for a paired-end experiment:

FileName1	FileName2	SampleName
rna_1_1.fq.bz2	$rna_1_2.fq.bz2$	Sample1
rna_2_1.fq.bz2	rna_2_2.fq.bz2	Sample2

These example files are also contained in the *QuasR* package and may be used as templates. The path of the files can be determined using:

The columns FileName for single-read, or FileName1 and FileName2 for paired-end experiments contain paths and names to files containing the sequence data. The paths can be absolute or relative to the location of the sample file. This allows combining files from different directories in a single analysis. For each input sequence file, qAlign will create one alignment file and by default store it in the same directory as the sequence file. Already existing alignment files with identical parameters will not be re-created, so that it is easy to reuse the same sequence files in multiple projects without unnecessarily copying sequence files or recreating alignments.

The SampleName column contains sample names for each sequence file. The same name can be used on several lines to indicate multiple sequence files that belong to the same sample (qCount can use this information to automatically combine counts for one sample from multiple files).

Three file formats are supported for input files (but cannot be mixed within a single sample file):

• FASTA files have names that end with '.fa', '.fna' or '.fasta'. They contain only sequences (and no base qualities) and will thus by default be aligned

on the basis of mismatches (the best alignment is the one with fewest mismatches).

- FASTQ files have names that end with '.fq' or '.fastq'. They contain sequences and corresponding base qualities and will be aligned by default using these qualities.
- BAM files have names that end with '.bam'. They can be used if the sequence reads have already been aligned outside of QuasR, and QuasR will only be used for downstream analysis based on the alignments contained in the BAM files. This makes it possible to use alignment tools that are not available within QuasR, but making use of this option comes with a risk and should only be used by experienced users. For example, it cannot be guaranteed any more that certain assumptions made by qCount are fulfilled by the external aligner. In addition, since the data provided to QuasR has already been processed, quality control plots that are based on unprocessed raw data will be missing from the output of qQCReport.

FASTA and FASTQ files can be compressed with gzip, bzip2 or xz (file extensions '.gz', '.bz2' or 'xz', respectively) and will automatically decompressed when necessary.

Consistency of samples within a project

The sample file implicitly defines the type of samples contained in the project: single read or paired-end read, sequences with or without qualities. This type will have a profound impact on the downstream analysis. For example, it controls whether alignments will be performed in single or paired-end mode, either with or without base qualities. That will also determine availability of certain options for quality control and quantification in qQCReport and qCount. For consistency, it is therefore required that all samples within a project have the same type; it is not possible to mix both single and paired-end read samples, or FASTA and FASTQ files in a single project (sample file). If necessary, it may be possible to analyse different types of files in separate QuasR projects and combine the derived results at the end.

5.2 Create an auxiliary file (optional)

By default QuasR aligns reads only to the reference genome. However, it may be interesting to align non-matching reads to further targets, for example to identify contamination from vectors or a different species, or in order to quantify spike-in material not contained in the reference genome. In QuasR, such supplementary reference files are called auxiliary references and can be specified to qAlign using the auxiliaryFile argument (see section 7.2 on page 35 for details). The format of the auxiliary file is similar to the one of the sample file described in section 5.1: It contains two columns with column names 'FileName' and

'AuxName' in the first row. Additional rows contain names and files of one or several auxiliary references in FASTA format.

An example auxiliary file looks like this:

FileName	AuxName
NC_001422.1.fa	phiX174

and is available from your QuasR installation at

> auxFile <- system.file(package="QuasR", "extdata", "auxiliaries.txt")

5.3 Select the reference genome

Sequence reads are primarily aligned against the reference genome. If necessary, QuasR will create an aligner index for the genome. The reference genome can be provided in one of two different formats:

- a string, referring to the name of a BSgenome package:
 - > available.genomes()
 - [1] "BSgenome.Alyrata.JGI.v1"
 - [2] "BSgenome.Amellifera.BeeBase.assembly4"
 - [3] "BSgenome.Amellifera.UCSC.apiMel2"
 - [4] "BSgenome.Athaliana.TAIR.04232008"
 - [5] "BSgenome.Athaliana.TAIR.TAIR9"
 - [6] "BSgenome.Btaurus.UCSC.bosTau3"
 - [7] "BSgenome.Btaurus.UCSC.bosTau4"
 - [8] "BSgenome.Btaurus.UCSC.bosTau6"
 - [9] "BSgenome.Celegans.UCSC.ce10"
 - [10] "BSgenome.Celegans.UCSC.ce2"
 - [11] "BSgenome.Celegans.UCSC.ce6"
 - [12] "BSgenome.Cfamiliaris.UCSC.canFam2"
 - [13] "BSgenome.Cfamiliaris.UCSC.canFam3"
 - [14] "BSgenome.Dmelanogaster.UCSC.dm2"
 - [15] "BSgenome.Dmelanogaster.UCSC.dm3"
 - [16] "BSgenome.Drerio.UCSC.danRer5"
 - [17] "BSgenome.Drerio.UCSC.danRer6"
 - [18] "BSgenome.Drerio.UCSC.danRer7"
 - [19] "BSgenome.Ecoli.NCBI.20080805"
 - [20] "BSgenome.Gaculeatus.UCSC.gasAcu1"
 - [21] "BSgenome.Ggallus.UCSC.galGal3"
 - [22] "BSgenome.Ggallus.UCSC.galGal4"
 - [23] "BSgenome. Hsapiens. UCSC. hg17"
 - [24] "BSgenome. Hsapiens. UCSC. hg18"
 - [25] "BSgenome. Hsapiens. UCSC. hg19"
 - [26] "BSgenome.Mmulatta.UCSC.rheMac2"

```
[27] "BSgenome.Mmusculus.UCSC.mm10"
[28] "BSgenome.Mmusculus.UCSC.mm8"
[29] "BSgenome.Mmusculus.UCSC.mm9"
[30] "BSgenome.Ptroglodytes.UCSC.panTro2"
[31] "BSgenome.Ptroglodytes.UCSC.panTro3"
[32] "BSgenome.Rnorvegicus.UCSC.rn4"
[33] "BSgenome.Rnorvegicus.UCSC.rn5"
[34] "BSgenome.Scerevisiae.UCSC.sacCer1"
[35] "BSgenome.Scerevisiae.UCSC.sacCer2"
[36] "BSgenome.Scerevisiae.UCSC.sacCer2"
[37] "BSgenome.Tgondii.ToxoDB.7.0"

> genomeName <- "BSgenome.Hsapiens.UCSC.hg19"
```

• a file name, referring to a sequence file containing one or several reference sequences (e.g. chromosomes) in FASTA format:

```
> genomeFile <- system.file(package="QuasR", "extdata", "hg19sub.fa")
```

5.4 Sequence data pre-processing

The preprocessReads function can be used to prepare the input sequence files prior to alignment. The function takes one or several sequence files (or pairs of files for a paired-end experiment) in FASTA or FASTQ format as input and produces the same number of output files with the processed reads.

In the following example, we truncate the reads by removing the three bases from the 3'-end (the right side), remove the adapter sequence AAAAAAAAA from the 5'-end (the left side) and filter out reads that, after truncation and adapter removal, are shorter than 14 bases or contain more than 2 N bases:

```
> td <- tempdir()</pre>
> infiles <- system.file(package="QuasR", "extdata",
                          c("rna_1_1.fq.bz2", "rna_2_1.fq.bz2"))
> outfiles <- file.path(td, basename(infiles))</pre>
> res <- preprocessReads(filename = infiles,
                          outputFilename = outfiles,
                          truncateEndBases = 3,
                          Lpattern = "AAAAAAAAA",
+
                          minLength = 14,
                          nBases = 2)
> res
                  rna_1_1.fq.bz2 rna_2_1.fq.bz2
                            3002
                                            3000
totalSequences
                                              463
matchTo5pAdapter
                              466
matchTo3pAdapter
                                0
                                                0
tooShort
                                               91
                              107
```

```
      tooManyN
      0
      0

      lowComplexity
      0
      0

      totalPassed
      2895
      2909

      > unlink(outfiles)
```

preprocessReads returns a matrix with a summary of the pre-processing. The matrix contains one column per (pair of) input sequence files, and contains the total number of reads (totalSequences), the number of reads that matched to the five prime or three prime adapters (matchTo5pAdapter and matchTo3pAdapter), the number of reads that were too short (tooShort), contained too many non-base characters (tooManyN) or were of low sequence complexity (lowComplexity, deactivated by default). Finally, the number of reads that passed the filtering steps is reported in the last row (totalPassed).

In the example below we process paired-end reads, removing all pairs with one or several \mathbb{N} bases. Even if only one sequence in a pair fulfills the filtering criteria, both reads in the pair are removed, thereby preserving the matching order of the sequences in the two files:

```
> td <- tempdir()</pre>
> infiles1 <- system.file(package="QuasR", "extdata", "rna_1_1.fq.bz2")</pre>
> infiles2 <- system.file(package="QuasR", "extdata", "rna_1_2.fq.bz2")</pre>
> outfiles1 <- file.path(td, basename(infiles1))</pre>
> outfiles2 <- file.path(td, basename(infiles2))</pre>
> res <- preprocessReads(filename=infiles1,
                           filenameMate=infiles2,
                           outputFilename=outfiles1,
                           outputFilenameMate=outfiles2,
                           nBases=0)
> res
                  rna_1_1.fq.bz2:rna_1_2.fq.bz2
totalSequences
                                             3002
matchTo5pAdapter
                                               NA
matchTo3pAdapter
                                               NA
                                                0
tooShort
                                                3
tooManvN
                                                0
lowComplexity
totalPassed
                                             2999
> unlink(c(outfiles1,outfiles2))
```

More details on the preprocessReads function can be found in the function documentation (see ?preprocessReads) or in the section 7.1 on page 35.

6 Example workflows

6.1 ChIP-seq: Measuring protein-DNA binding and chromatin modifications

Here we show an exemplary single-end ChIP-seq workflow using a small number of reads from a histone 3 lysine 4 trimethyl (H3K4me3) ChIP-seq experiment. This histone modification is known to locate to genomic regions with a high density of CpG dinucleotides (so called CpG islands); about 60% of mammalian genes have such a CpG island close to their transcript start site. All necessary files are included in the *QuasR* package, and we start the example workflow by copying those files into the current working directly, into a subfolder called "extdata":

```
> file.copy(system.file(package="QuasR", "extdata"), ".", recursive=TRUE)
[1] TRUE
```

6.1.1 Align reads using the qAlign function

We assume that the sequence reads have already been pre-processed as described in section 5.4. Also, a sample file (section 5.1) that lists all sequence files to be analyzed has been prepared. A FASTA file with the reference genome sequence(s) is also available (section 5.3), as well as a auxiliary file for alignment of reads that failed to match the reference genome (section 5.2).

By default, newly generated BAM files will be stored at the location of the input sequence files, which should be writable and have sufficient capacity (an alternative location can be specified using the alignmentsDir argument). Make also sure that you have sufficient temporary disk space for intermediate files in tempdir() (see section 7.2). We start by aligning the reads using qAlign:

```
> sampleFile <- "extdata/samples_chip_single.txt"
> auxFile <- "extdata/auxiliaries.txt"
> genomeFile <- "extdata/hg19sub.fa"
> proj1 <- qAlign(sampleFile, genome=genomeFile, auxiliaryFile=auxFile)
nodeNames
george2
> proj1
Project: qProject
Options
           : maxHits
                             : 1
             paired
                             : no
             splicedAlignment: FALSE
             bisulfite
                             : no
             snpFile
                             : none
           : Rbowtie v1.0.3 (parameters: -m 1 --best --strata)
 Aligner
           : /tmp/Rtmp3ckjsv/Rbuild484015dfa.../hg19sub.fa (file)
 Genome
```

```
Reads : 2 files, 2 samples (fastq format):

1. chip_1_1.fq.bz2 Sample1 (phred33)

2. chip_2_1.fq.bz2 Sample2 (phred33)

Genome alignments: directory: same as reads

1. chip_1_1_6cdf5473cc09.bam

2. chip_2_1_6cdf7f640d43.bam

Aux. alignments: 1 file, directory: same as reads

a. /tmp/Rtmp3ckjsv/Rbuild484015dfa3.../NC_001422.1.fa phiX174

1. chip_1_1_6cdf25e3b15b.bam

2. chip_2_1_6cdf77cbad04.bam
```

qAlign will build alignment indices if they do not yet exist (by default, if the genome and auxiliary sequences are given in the form of FASTA files, they will be stored in the same folder). The qProject object (proj1) returned by qAlign now contains all information about the ChIP-seq experiment: the (optional) project name, the project options, aligner package, reference genome, and at the bottom the sequence and alignment files. For each input sequence file, there will be one BAM file with alignments against the reference genome, and one for each auxiliary target sequence with alignments of reads without genome hits. Our auxFile contains a single auxiliary target sequence, so we expect two BAM files per input sequence file:

```
> list.files("extdata", pattern=".bam$")
[1] "chip_1_1_6cdf25e3b15b.bam" "chip_1_1_6cdf5473cc09.bam"
[3] "chip_2_1_6cdf77cbad04.bam" "chip_2_1_6cdf7f640d43.bam"
```

The BAM file names consist of the base name of the sequence file with an added random string. The random suffix makes sure that newly generated alignment files do not overwrite existing ones, for example of the same reads aligned against an alternative reference genome. Each alignment file is accompanied by two additional files with suffixes ".bai" and ".txt":

```
> list.files("extdata", pattern="^chip_1_1_")[1:3]
[1] "chip_1_1_6cdf25e3b15b.bam"
[2] "chip_1_1_6cdf25e3b15b.bam.bai"
[3] "chip_1_1_6cdf25e3b15b.bam.txt"
```

The ".bai" file is the BAM index used for fast access by genomic coordinate. The ".txt" file contains all the parameters used to generate the corresponding BAM file. Before new alignments are generated, qAlign will look for available ".txt" files in default locations (the directory containing the input sequence file, or the value of alignmentsDir), and read their contents to determine if a compatible BAM file already exists. A compatible BAM file is one with the same reads and genome, aligned using the same aligner and identical alignment parameters. If a

compatible BAM file is not found, or the ".txt" file is missing, qAlign will generate a new BAM file. It is therefore recommended not to delete the ".txt" file - without it, the corresponding BAM file will become unusable for QuasR.

6.1.2 Create a quality control report

QuasR can produce a quality control report in the form of a series of diagnostic plots with details on sequences and alignments (see Figure 1 on page 9). The plots are generated by calling the qQCReport function with the qProject object as argument. qQCReport uses ShortRead [7] internally to obtain some of the quality metrics, and some of the plots are inspired by the FastQC quality control tool by Simon Andrews (http://www.bioinformatics.bbsrc.ac.uk/projects/fastqc/). The plots will be stored into a multipage PDF document defined by the pdfFilename argument, or else shown as individual plot windows on the current graphics device. In order to keep the running time reasonably short, some quality metrics are obtained from a random sub-sample of the sequences or alignments.

> qQCReport(proj1, pdfFilename="extdata/qc_report.pdf")

Currently available plots are described in section 7.4 on page 37 and following.

6.1.3 Alignment statistics

The alignmentStats gets the number of (un-)mapped reads for each sequence file in a qProject object, by reading the BAM file indices, and returns them as a data.frame. The function also works for arguments of type character with one or several BAM file names (for details see section 7.5 on page 40).

> alignmentStats(proj1)

	seqlength	${\tt mapped}$	unmapped
Sample1:genome	95000	2339	258
Sample2:genome	95000	3609	505
Sample1:phiX174	5386	251	7
Sample2:phiX174	5386	493	12

6.1.4 Export genome wig file from alignments

For visualization in a genome browser, alignment coverage along the genome can be exported to a (compressed) wig file using the qExportWig function. The created fixedStep wig file (see http://genome.ucsc.edu/goldenPath/help/wiggle.html for details on the wig format) will contain one track per sample in the qProject object. The resolution is defined using the binsize argument, and if scaling is set to TRUE, read counts per bin are scaled by the total number of aligned reads in each sample to improve comparability:

> qExportWig(proj1, binsize=100L, scaling=TRUE, collapseBySample=TRUE)

6.1.5 Count alignments using qCount

Alignments are quantified using qCount, for example using a GRanges object as a query. In our H3K4me3 ChIP-seq example, we expect the reads to occur around the transcript start site of genes. We can therefore construct suitable query regions using genomic intervals around the start sites of known genes. In the code below, this is achieved with help from the *GenomicFeatures* package: We first create a TranscriptDb object from a ".gtf" file with gene annotation. With the promoters function, we can then create the GRanges object with regions to be quantified. Finally, because most genes consist of multiple overlapping transcripts, we select the first transcript for each gene:

```
> library(GenomicFeatures)
> annotFile <- "extdata/hg19sub_annotation.gtf"
> chrLen <- scanFaIndex(genomeFile)</pre>
 chrominfo <- data.frame(chrom=as.character(seqnames(chrLen)),</pre>
                           length=width(chrLen),
                           is_circular=rep(FALSE, length(chrLen)))
 txdb <- makeTranscriptDbFromGFF(file=annotFile, format="gtf",</pre>
                                   exonRankAttributeName="exon_number",
                                   gffGeneIdAttributeName="gene_name",
+
                                   chrominfo=chrominfo,
                                   dataSource="Ensembl",
                                   species="Homo sapiens")
 promReg <- promoters(txdb, upstream=1000, downstream=500,</pre>
                        columns=c("gene_id","tx_id"))
> gnId <- sapply(mcols(promReg)$gene_id, paste, collapse=",")
> promRegSel <- promReg[ match(unique(gnId), gnId) ]</pre>
> names(promRegSel) <- unique(gnId)
> head(promRegSel)
GRanges with 6 ranges and 2 metadata columns:
                   seqnames
                                    ranges strand
                      <Rle>
                                 <IRanges>
                                            <Rle> |
 ENSG00000176022
                       chr1 [31629, 33128]
                       chr1 [ 6452, 7951]
 ENSG00000186891
                       chr1 [14013, 15512]
 ENSG00000186827
 ENSG00000078808
                       chr1 [31882, 33381]
 ENSG00000171863
                       chr2 [ 1795,
                                     3294]
                                     8659]
 ENSG00000252531
                       chr2 [ 7160,
                           gene_id
                                        tx_id
                   <CharacterList> <integer>
 ENSG00000176022 ENSG00000176022
                                            1
 ENSG00000186891 ENSG00000186891
                                            2
                                            6
 ENSG00000186827 ENSG00000186827
                                            9
 ENSG00000078808 ENSG00000078808
 ENSG00000171863 ENSG00000171863
                                           17
 ENSG00000252531 ENSG00000252531
                                           26
```

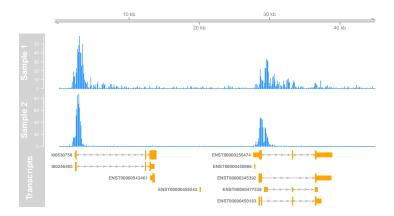
seqlengths: chr1 chr2 chr3 40000 10000 45000

Using promRegSel object as query, we can now count the alignment per sample in each of the promoter windows.

> cnt <- qCount(proj1, promRegSel)</pre> > cnt width Sample1 Sample2 ENSG00000176022 ENSG00000186891 ENSG00000186827 ENSG00000078808 ENSG00000171863 ENSG00000252531 ENSG00000247886 1500 ENSG00000254999 ENSG00000238642 ENSG00000134086 ENSG00000238345 ENSG00000134075

The counts returned by qCount are the raw number of alignments per sample and region, without any normalization for the query region length, or the total number of aligned reads in a sample. As expected, we can find H3K4me3 signal at promoters of a subset of the genes with CpG island promoters, which we can visualize with help of the Gviz package:

```
> gr1 <- import("Sample1.wig.gz", asRangedData=FALSE)
> gr2 <- import("Sample2.wig.gz", asRangedData=FALSE)
> library(Gviz)
> axisTrack <- GenomeAxisTrack()
> dTrack1 <- DataTrack(range=gr1, name="Sample 1", type="h")
> dTrack2 <- DataTrack(range=gr2, name="Sample 2", type="h")
> txTrack <- GeneRegionTrack(txdb, name="Transcripts", showId=TRUE)
> plotTracks(list(axisTrack, dTrack1, dTrack2, txTrack),
+ chromosome="chr3", extend.left=1000)
```



6.1.6 Create a genomic profile for a set of regions using qProfile

Given a set of anchor positions in the genome, qProfile calculates the number of nearby alignments relative to the anchor position, for example to generate a average profile. The neighborhood around anchor positions can be specified by the upstream and downstream argument. Alignments that are upstream of an anchor position will have a negative relative position, and downstream alignments a positive. The anchor positions are all aligned at position zero in the return value.

Anchor positions will be provided to qProfile using the query argument, which takes a GRanges object. The anchor positions correspond to start() for regions on "+" or "*" strands, and to end() for regions on the "-" strand. As mentioned above, we expect H3K4me3 ChIP-seq alignments to be enriched around the transcript start site of genes. We can therefore construct a suitable query object from the start sites of known genes. In the code below, start sites ('start_codon') are imported from a ".gtf" file with the help of the rtracklayer package. In addition, 'strand' and 'gene_name' are also selected for import. Duplicated start sites, e.g. from genes with multiple transcripts, are removed. Finally, all regions are given the name "TSS", because qProfile combines regions with identical names into a single profile.

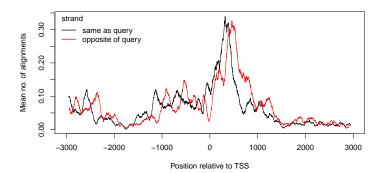
```
<Rle>
                   <IRanges>
                               <Rle> | <character>
TSS
         chr1 [ 6949, 6951]
                                           TNFRSF18
TSS
         chr1 [14505, 14507]
                                            TNFRSF4
        chr1 [29171, 29173]
TSS
                                               SDF4
TSS
         chr1 [32659, 32661]
                                            B3GALT6
         chr2 [ 3200,
                       3202]
                                               RPS7
TSS
         chr3 [ 2386,
TSS
                       2388]
                                            C3orf10
seqlengths:
 chr1 chr2 chr3
   NA
        NA
              NA
```

Alignments around the tssRegions coordinates are counted in a window defined by the upstream and downstream arguments, which specify the number of bases to include around each anchor position. For query regions on "+" or "*" strands, upstream refers to the left side of the anchor position (lower coordinates), while for regions on the "-" strand, upstream refers to the right side (higher coordinates). The following example creates separate profiles for alignments on the *same* and on the *opposite* strand of the regions in query.

```
> prS <- qProfile(proj1, tssRegions, upstream=3000, downstream=3000,
                  orientation="same")
> pr0 <- qProfile(proj1, tssRegions, upstream=3000, downstream=3000,
                  orientation="opposite")
> lapply(prS, "[", , 1:10)
$coverage
-3000 -2999 -2998 -2997 -2996 -2995 -2994 -2993 -2992 -2991
                8
                      8
                            8
                                   8
                                         8
                                               8
                                                     8
                                                            8
$Sample1
-3000 -2999 -2998 -2997 -2996 -2995 -2994 -2993 -2992 -2991
          0
                0
                      0
                            0
                                   0
                                         0
    1
$Sample2
-3000 -2999 -2998 -2997 -2996 -2995 -2994 -2993 -2992 -2991
                      2
                            0
                                   0
```

The counts returned by qProfile are the raw number of alignments per sample and position, without any normalization for the number of query regions or the total number of alignments in a sample per position. To obtain the average number of alignments, we divide the alignment counts by the number of query regions that covered a given relative position around the anchor sites. This coverage is available as the first element in the return value. The shift between same and opposite strand alignments is indicative for the average length of the sequenced ChIP fragments.

```
> prCombS <- do.call("+", prS[-1]) /prS[[1]]
> prComb0 <- do.call("+", pr0[-1]) /pr0[[1]]
> plot(as.numeric(colnames(prCombS)), filter(prCombS[1,], rep(1/100,100)),
+ type='l', xlab="Position relative to TSS", ylab="Mean no. of alignments")
> lines(as.numeric(colnames(prComb0)), filter(prComb0[1,], rep(1/100,100)),
+ type='l', col="red")
> legend(title="strand", legend=c("same as query", "opposite of query"),
+ x="topleft", col=c("black", "red"), lwd=1.5, bty="n", title.adj=0.1)
```



6.1.7 Using a BSgenome package as reference genome

QuasR also allows using of BSgenome packages instead of a FASTA file as reference genome (see section 5.3). To use a BSgenome, the genome argument of qAlign is set to a string matching the name of a BSgenome package, for example "BSgenome.Hsapiens.UCSC.hg19". If that package is not already installed, qAlign will check if it is available from bioconductor.org and download it automatically. The corresponding alignment index will be saved as a new package, named after the original BSgenome package and the aligner used to build the index, for example BSgenome.Hsapiens.UCSC.hg19.Rbowtie.

The code example below illustrates the use of a *BSgenome* reference genome for the same example data as above. Running it for the first time will take several hours in order to build the aligner index:

```
> file.copy(system.file(package="QuasR", "extdata"), ".", recursive=TRUE)
> sampleFile <- "extdata/samples_chip_single.txt"
> auxFile <- "extdata/auxiliaries.txt"
> available.genomes() # list available genomes
> genomeName <- "BSgenome.Hsapiens.UCSC.hg19"
> proj1 <- qAlign(sampleFile, genome=genomeName, auxiliaryFile=auxFile)
> proj1
```

6.2 RNA-seq: Gene expression profiling

In *QuasR*, an analysis workflow for an RNA-seq dataset is very similar to the one described above for a ChIP-seq experiment. The major difference is that here reads are aligned using splicedAlignment=TRUE, which will cause qAlign to align reads with SpliceMap[8], rather than bowtie[3] (both are contained in the *Rbowtie* package). SpliceMap and *QuasR* also support spliced paired-end alignments; the splicedAlignment argument can be freely combined with the paired argument.

We start the example workflow by copying the example data files into the current working directly, into a subfolder called "extdata", and then create spliced alignments using qAlign:

```
> file.copy(system.file(package="QuasR", "extdata"), ".", recursive=TRUE)
> sampleFile <- "extdata/samples_rna_paired.txt"
> genomeFile <- "extdata/hg19sub.fa"</pre>
> proj2 <- qAlign(sampleFile, genome=genomeFile, splicedAlignment=TRUE)
nodeNames
george2
      1
> proj2
Project: qProject
Options
           : maxHits
                             : 1
             paired
                             : fr
             splicedAlignment: TRUE
             bisulfite
                            : no
             snpFile
                             : none
 Aligner
           : Rbowtie v1.0.3 (parameters: -max_intron 400000 -min_intron 20000 -max_multi
 Genome
           : /tmp/Rtmp3ckjsv/Rbuild484015dfa.../hg19sub.fa (file)
Reads
           : 2 pairs of files, 2 samples (fastq format):
   1. rna_1_1.fq.bz2 rna_1_2.fq.bz2 Sample1 (phred33)
   2. rna_2_1.fq.bz2 rna_2_2.fq.bz2 Sample2 (phred33)
Genome alignments: directory: same as reads
   1. rna_1_1_6cdf603fc991.bam
  2. rna_2_1_6cdf6991d795.bam
```

Aligning the reads with splicedAlignment=TRUE is much slower than the default, but will allow to also align reads that cross one or two exon junctions, and thus have a large deletion (the intron) relative to the reference genome.

```
> proj2unspl <- qAlign(sampleFile, genome=genomeFile,
+ splicedAlignment=FALSE)</pre>
```

Aux. alignments: none

```
nodeNames
george2
> alignmentStats(proj2)
               seqlength mapped unmapped
Sample1:genome
                    95000
                            6002
                    95000
                            6000
Sample2:genome
                                         0
> alignmentStats(proj2unspl)
               seqlength mapped unmapped
Sample1:genome
                    95000
                            2258
                                      3746
Sample2:genome
                    95000
                            2652
                                      3348
```

As with ChIP-seq experiments, qCount is used to quantify alignments. For quantification of gene or exon expression levels, qCount can be called with a query of type TranscriptDB, such as the one we constructed in the ChIP-seq workflow above from a ".gtf" file. The argument reportLevel can be used to control if annotated exonic regions should be quantified independently (reportLevel="exon") or non-redundantly combined per gene (reportLevel="gene"):

```
> geneLevels <- qCount(proj2, txdb, reportLevel="gene")
> exonLevels <- qCount(proj2, txdb, reportLevel="exon")
> head(geneLevels)
                width Sample1 Sample2
ENSG00000078808
                 4697
                           708
                                  1076
ENSG00000134075
                  589
                          1201
                                  1322
ENSG00000134086
                 4213
                           282
                                   295
                          2922
ENSG00000171863
                 5583
                                  2224
                                   344
ENSG00000176022
                 2793
                            62
ENSG00000186827
                 1721
                            37
                                     8
> head(exonLevels)
```

	width	Sample1	Sample2
1	2793	62	344
10	187	1	0
11	307	1	0
12	300	9	2
13	493	18	2
14	129	5	0

The values returned by qCount are the number of alignments. Sometimes it is required to normalize for the length of query regions, or the size of the libraries. For example, gene expression levels in the form of RPKM values (reads per kilobase of transcript and million mapped reads) can be obtained as follows:

```
geneRPKM <- t(t(geneLevels[,-1] /geneLevels[,1] *1000)</pre>
                /colSums(geneLevels[,-1]) *1e6)
> geneRPKM
                    Sample1
                                Sample2
ENSG00000078808 21281.168
                             31628.1089
ENSG00000134075 287879.322 309882.9454
ENSG00000134086
                  9450.187
                              9667.4531
                 73891.646
ENSG00000171863
                             54998.2246
ENSG00000176022
                  3134.033
                             17004.7018
ENSG00000186827
                  3035.314
                               641.7866
ENSG00000186891
                  2679.388
                               201.5538
ENSG00000238345
                      0.000
                                 0.0000
ENSG00000238642
                      0.000
                                 0.0000
                      0.000
ENSG00000247886
                                 0.0000
ENSG00000252531
                  5196.924
                              1694.0410
ENSG00000254999 210572.382 220589.4127
```

Please note the RPKM values in our example are higher than what you would usually get for a real RNA-seq dataset. The values here are artificially scaled up because our example data contains reads only for a small number of genes.

Exon-exon junctions can be quantified by setting reportLevel="junction". In this case, qCount will ignore the query argument and scan all alignments for any detected splices, which are returned as a GRanges object: The region start and end coordinates correspond to the first and last bases of the intron, and the counts are returned in the mcols() of the GRanges object. Alignments that are identically spliced but reside on opposite strands will be quantified separately. In an unstranded RNA-seq experiment, this may give rise to two counts for the same intron, one each for the supporting alignments on plus and minus strands.

- > exonJunctions <- qCount(proj2, NULL, reportLevel="junction")
 > exonJunctions
- GRanges with 52 ranges and 2 metadata columns:

	seqnames		ranges	strand	- 1	Sample1	Sample2
	<rle></rle>	<ii< td=""><td>Ranges></td><td><rle></rle></td><td></td><td><numeric></numeric></td><td><numeric></numeric></td></ii<>	Ranges>	<rle></rle>		<numeric></numeric>	<numeric></numeric>
[1]	chr1	[12213,	12321]	+	- 1	1	0
[2]	chr1	[12213,	12321]	-	- 1	2	0
[3]	chr1	[13085,	13371]	-	- 1	1	0
[4]	chr1	[18069,	18837]	+	- 1	8	7
[5]	chr1	[18069,	18837]	-		6	12
[48]	chr1	[14166,	14362]	-		0	1
[49]	chr1	[19009,	19148]	_		0	2
[50]	chr1	[29327,	32271]	_		0	2
[51]	chr1	[4617,	4778]	_		0	1
[52]	chr3	[2504,	5589]	+	- 1	0	2

```
seqlengths:
chr1 chr2 chr3
```

About half of the exon-exon junctions detected in this sample datset correspond to known introns; they tend to be the ones with higher coverage:

```
> knownIntrons <- unlist(intronsByTranscript(txdb))</pre>
> isKnown <- overlapsAny(exonJunctions, knownIntrons, type="equal")
> table(isKnown)
isKnown
FALSE TRUE
   28
> tapply(rowSums(as.matrix(mcols(exonJunctions))),
         isKnown, summary)
$`FALSE`
   Min. 1st Qu.
                 Median
                            Mean 3rd Qu.
                                             Max.
   1.00
           2.00
                    5.00
                           32.18
                                    47.75
                                           177.00
$`TRUE`
   Min. 1st Qu.
                 Median
                            Mean 3rd Qu.
                                             Max.
   1.00
                   10.50
           1.00
                           42.71
                                    60.00
                                           209.00
```

6.3 Bis-seq: Measuring DNA methylation

Sequencing of bisulfite-converted genomic DNA allows detection of methylated cytosines, which in mammalian genomes typically occur un the context of CpG dinucleotides. The treatment of DNA with bisulfite induces deamination of non-methylated cytosines, converting them to uracils. Sequencing and aligning of such bisulfite-converted DNA results in C-to-T mismatches. Both alignment of converted reads, as well as the interpretation of the alignments for calculation of methylation levels require specific approaches and are supported in *QuasR* by qAlign (bisulfite argument, section 7.2) and qMeth (section 7.9), respectively.

We start the analysis by copying the example data files into the current working directly, into a subfolder called "extdata". Then, bisulfite-specific alignment is selected in qAlign by setting bisulfite to "dir" for a directional experiment, or to "undir" for an undirectional Bis-seq experiment:

```
> file.copy(system.file(package="QuasR", "extdata"), ".", recursive=TRUE)
[1] TRUE
> sampleFile <- "extdata/samples_bis_single.txt"
> genomeFile <- "extdata/hg19sub.fa"
> proj3 <- qAlign(sampleFile, genomeFile, bisulfite="dir")</pre>
```

```
nodeNames
george2
      1
> proj3
Project: qProject
Options
           : maxHits
                             : 1
             paired
                             : no
             splicedAlignment: FALSE
             bisulfite
                             : dir
             snpFile
                             : none
 Aligner
           : Rbowtie v1.0.3 (parameters: -k 2 --best --strata -v 2)
           : /tmp/Rtmp3ckjsv/Rbuild484015dfa.../hg19sub.fa (file)
 Genome
 Reads
           : 1 file, 1 sample (fasta format):
   1. bis_1_1.fa.bz2 Sample1
Genome alignments: directory: same as reads
   1. bis_1_1_6cdf3e49cb47.bam
 Aux. alignments: none
```

The resulting alignments are not different from those of non-Bis-seq experiments, apart from the fact that they may contain many C-to-T (or A-to-G) mismatches that are not counted as mismatches when aligning the reads. The number of alignments in specific genomic regions could be quantified using qCount as with ChIP-seq or RNA-seq experiments. For quantification of methylation the qMeth function is used:

```
> meth <- qMeth(proj3, mode="CpGcomb", collapseBySample=TRUE)
> meth
```

GRanges with 3110 ranges and 2 metadata columns:

O		O				
	seqnames	ranges	strand		Sample1_T	Sample1_M
	<rle></rle>	Ranges	<rle></rle>		<integer></integer>	<integer></integer>
[1]	chr1	[19, 20]	*		1	1
[2]	chr1	[21, 22]	*		1	1
[3]	chr1	[54, 55]	*		3	1
[4]	chr1	[57, 58]	*		3	0
[5]	chr1	[80, 81]	*	-	6	5
[3106]	chr3	[44957, 44958]	*		8	7
[3107]	chr3	[44977, 44978]	*		5	3
[3108]	chr3	[44981, 44982]	*		4	3
[3109]	chr3	[44989, 44990]	*		1	1
[3110]	chr3	[44993, 44994]	*	- 1	1	1

seqlengths:

```
chr1 chr2 chr3
40000 10000 45000
```

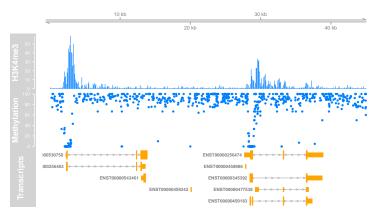
By default, qMeth quantifies methylation for all cytosines in CpG contexts, combining the data from plus and minus strands (mode="CpGcomb"). The results are returned as a GRanges object with coordinates of each CpG, and two metadata columns for each input sequence file in the qProject object. These two columns contain the total number of aligned reads that overlap a given CpG (C-to-C matches or C-to-T mismatches, suffix _T in the column name), and the number of read alignments that had a C-to-C match at that position (methylated events, suffix _M).

Independent of the number of alignments, the returned object will list all CpGs in the target genome including the ones that have zero coverage, unless you set keepZero=FALSE:

```
> chrs <- readDNAStringSet(genomeFile)
> sum(vcountPattern("CG",chrs))
[1] 3110
> length(qMeth(proj3))
[1] 3110
> length(qMeth(proj3, keepZero=FALSE))
[1] 2929
```

The fraction methylation can easily be obtained as the ratio between $_M$ and $_T$ columns:

```
> percMeth <- mcols(meth)[,2] *100 /mcols(meth)[,1]
> summary(percMeth)
   Min. 1st Qu. Median
                                                     NA's
                            Mean 3rd Qu.
                                             Max.
   0.00
          75.00
                  90.91
                           75.45 100.00 100.00
                                                      181
> axisTrack <- GenomeAxisTrack()</pre>
> dTrack1 <- DataTrack(range=gr1, name="H3K4me3", type="h")</pre>
> dTrack2 <- DataTrack(range=meth, data=percMeth,</pre>
                        name="Methylation", type="p")
> txTrack <- GeneRegionTrack(txdb, name="Transcripts", showId=TRUE)
> plotTracks(list(axisTrack, dTrack1, dTrack2, txTrack),
             chromosome="chr3", extend.left=1000)
```



If qMeth is called without a query argument, it will by default return methylation states for each C or CpG in the genome. Using a query argument it is possible to restrict the analysis to specific genomic regions, and if using in addition collapseByQueryRegion=TRUE, the single base methylation states will further be combined for all C's that are contained in the same query region:

```
> qMeth(proj3, query=GRanges("chr1",IRanges(start=31633,width=2)),
        collapseBySample=TRUE)
GRanges with 1 range and 2 metadata columns:
                       ranges strand | Sample1_T Sample1_M
      seqnames
         <Rle>
                     <IRanges> <Rle> | <integer> <integer>
  [1]
          chr1 [31633, 31634]
                                    * |
                                                10
 seqlengths:
    chr1 chr2 chr3
   40000 10000 45000
> qMeth(proj3, query=promRegSel, collapseByQueryRegion=TRUE,
        collapseBySample=TRUE)
GRanges with 12 ranges and 2 metadata columns:
       segnames
                        ranges strand
                                          | Sample1_T Sample1_M
                     <IRanges>
          <Rle>
                                 <Rle>
                                           <numeric> <numeric>
   [1]
                                                             74
           chr1 [31629, 33128]
                                                  426
   [2]
           chr1 [ 6452,
                                                  388
                                                            244
                        7951]
   [3]
           chr1 [14013, 15512]
                                                  627
                                                            560
   [4]
                                                  522
           chr1 [31882, 33381]
                                                            232
   [5]
           chr2 [ 1795,
                          3294]
                                                  997
                                                            539
   [8]
           chr3 [ 1276,
                          2775]
                                                  715
                                                            253
   [9]
           chr3 [19069, 20568]
                                                  253
                                                            204
  [10]
           chr3 [26692, 28191]
                                                  934
                                                            818
           chr3 [26834, 28333]
                                                  934
  [11]
                                                            777
  Γ12]
           chr3 [13102, 14601]
                                                  307
                                                            287
```

```
seqlengths:
chr1 chr2 chr3
40000 10000 45000
```

6.4 Allele-specific analysis

All experiment types supported by *QuasR* (ChIP-seq, RNA-seq and Bis-seq; only alignments to the genome, but not to auxiliaries) can also be analyzed in an allele-specific manner. For this, a file containing genomic location and the two alleles of known sequence polymorphisms has to be provided to the snpFile argument of qAlign. The file is in tab-delimited text format without a header and contains four columns with chromosome name, position, reference allele and alternative allele.

Below is an example of a SNP file, also available from system.file(package="QuasR", "extdata", "hg19sub_snp.txt"):

chr1	8596	G	Α
chr1	18443	G	Α
chr1	18981	C	T
chr1	19341	G	Α

For a given locus, either reference or alternative allele may but does not have to be identical to the sequence of the reference genome (genomeFile in this case). To avoid an alignment bias, all reads are aligned separately to each of the two new genomes, which QuasR generates by injecting the SNPs listed in snpFile into the reference genome. Finally, the two alignment files are combined, only retaining the best alignment for each read. While this procedure takes more than twice as long as aligning against a single genome, it has the advantage to correctly align reads even in regions of high SNP density and has been shown to produce more accurate results.

While combining alignments, each read is classified into one of three groups (stored in the BAM files under the XV tag):

- R: the read aligned better to the **reference** genome
- U: the read aligned equally well to both genomes (unknown origin)
- A: the read aligned better to the alternative genome

Using these alignment classifications, the qCount and qMeth functions will produce three counts instead of a single count; one for each class. The column names will be suffixed by _R, _U and _A.

The examples below use data provided with the *QuasR* package, which is first copied to the current working directory, into a subfolder called "extdata":

```
> file.copy(system.file(package="QuasR", "extdata"), ".", recursive=TRUE)
[1] TRUE
```

The example below aligns the same reads that were also used in the ChIP-seq workflow (section 6.1), but this time using a snpFile:

```
> sampleFile <- "extdata/samples_chip_single.txt"
> genomeFile <- "extdata/hg19sub.fa"</pre>
> snpFile <- "extdata/hg19sub_snp.txt"
> proj1SNP <- qAlign(sampleFile, genome=genomeFile, snpFile=snpFile)
nodeNames
george2
> proj1SNP
Project: qProject
Options
          : maxHits
                            : 1
             paired
                            : no
             splicedAlignment: FALSE
             bisulfite
             snpFile
                             : /tmp/Rtmp3ckjs.../hg19sub_snp.txt
           : Rbowtie v1.0.3 (parameters: -k 2 --best --strata -v 2)
 Aligner
           : /tmp/Rtmp3ckjsv/Rbuild484015dfa.../hg19sub.fa (file)
 Genome
           : 2 files, 2 samples (fastq format):
 Reads
   1. chip_1_1.fq.bz2 Sample1 (phred33)
   2. chip_2_1.fq.bz2 Sample2 (phred33)
Genome alignments: directory: same as reads
   1. chip_1_1_6cdf24e2072b.bam
   2. chip_2_1_6cdfcaf54e1.bam
 Aux. alignments: none
```

Instead of one count per promoter region and sample, qCount now returns three (promRegSel was generated in the ChIP-seq example workflow on page 19):

> head(qCount(proj1, promRegSel))

```
width Sample1 Sample2
ENSG00000176022 1500
                         157
                                  701
ENSG00000186891 1500
                           22
                                    5
ENSG00000186827 1500
                           10
                                    3
ENSG00000078808 1500
                           73
                                  558
ENSG00000171863 1500
                           94
                                  339
ENSG00000252531 1500
                                    9
                           59
```

> head(qCount(proj1SNP, promRegSel))

	${\tt width}$	Sample1_R	Sample1_U	${\tt Sample1_A}$	$Sample2_R$
ENSG00000176022	1500	0	133	0	0
ENSG00000186891	1500	4	16	0	0
ENSG00000186827	1500	2	8	0	0
ENSG00000078808	1500	0	59	0	0
ENSG00000171863	1500	4	78	0	8
ENSG00000252531	1500	3	50	2	0
	Sample	e2_U Sample	e2_A		
ENSG00000176022		559	0		
ENSG00000186891		5	0		
ENSG00000186827		2	0		
ENSG00000078808		432	0		
ENSG00000171863		263	0		
ENSG00000252531		6	0		

The example below illustrates use of a snpFile for Bis-seq experiments. Similarly as for qCount, the count types are labeled by R, U and A. These labels are added to the total and methylated column suffixes _T and _M, resulting in a total of six instead of two counts per feature and sample:

```
> sampleFile <- "extdata/samples_bis_single.txt"
> genomeFile <- "extdata/hg19sub.fa"
> proj3SNP <- qAlign(sampleFile, genomeFile,
+ snpFile=snpFile, bisulfite="dir")
nodeNames
george2</pre>
```

> head(qMeth(proj3SNP, mode="CpGcomb", collapseBySample=TRUE))
GRanges with 6 ranges and 6 metadata columns:

_		_						
	seqnames	ra	anges	strand		Sample1_TR	Sample1	_MR
	<rle></rle>	<ira:< td=""><td>nges></td><td><rle></rle></td><td> </td><td><integer></integer></td><td><integ< td=""><td>er></td></integ<></td></ira:<>	nges>	<rle></rle>		<integer></integer>	<integ< td=""><td>er></td></integ<>	er>
[1]	chr1	[19,	20]	*		0		0
[2]	chr1	[21,	22]	*		0		0
[3]	chr1	[54,	55]	*		0		0
[4]	chr1	[57,	58]	*		0		0
[5]	chr1	[80,	81]	*		0		0
[6]	chr1	[103,	104]	*		0		0
	Sample1_	ΓU Samj	ole1_N	MU Sampl	e1	_TA Sample1	L_MA	

<integer> <integer> <integer> <integer> [1] [2] [3] [4] [5] [6] ___

seqlengths:

chr1 chr2 chr3 40000 10000 45000

7 Description of Individual QuasR Functions

Please refer to the QuasR reference manual or the function documentation (e.g. using ?qAlign) for a complete description of QuasR functions. The descriptions provided below are meant to give and overview over all functions and summarize the purpose of each one.

7.1 preprocessReads

The preprocessReads function can be used to prepare the input sequences before alignment to the reference genome, for example to filter out low quality reads unlikely to produce informative alignments. When working with paired-end experiments, the paired reads are expected to be contained in identical order in two separate files. For this reason, both reads of a pair are filtered out if any of the two reads fulfills the filtering criteria. The following types of filtering tasks can be performed (in the order as listed):

- Truncate reads: remove nucleotides from the start and/or end of each read.
- Trim adapters: remove nucleotides at the beginning and/or end of each read that match to a defined (adapter) sequence.
 The adapter trimming is done by calling trimLRPatterns from the Biostrings package [9].
- 3. Filter out low quality reads: Filter out reads that fulfill any of the filtering criteria (contain more than nBases N bases, are shorter than minLength or have a dinucleotide complexity of less than complexity-times the average complexity of the human genome sequence).

The dinucleotide complexity is calculated in bits as Shannon entropy using the following formula $-\sum_i f_i \cdot \log_2 f_i$, where f_i is the frequency of dinucleotide i (i = 1, 2, ..., 16).

7.2 qAlign

qAlign is the function that generates alignment files in BAM format, for all input sequence files listed in sampleFile (see section 5.1), against the reference genome (genome argument), and for reads that to not match to the reference genome, against one or several auxiliary target sequences (auxiliaryFile, see section 5.2).

The reference genome can be provided either as a FASTA sequence file or as a BSgenome package name (see section 5.3). If a BSgenome package is not found in the installed packages but available from Bioconductor, it will be automatically downloaded.

The alignment program is set by aligner, and parameters by maxHits, paired, splicedAlignment and alignmentParameter. Currently, aligner can

only be set to "Rbowtie", which is a wrapper for bowtie [3] and SpliceMap [8]. SpliceMap will be used if splicedAlignment=TRUE. The alignment strategy is affected by the parameters snpFile (alignments to variant genomes containing sequence polymorphisms) and bisulfite (alignment of bisulfite-converted reads). Finally, clObj can be used to enable parallelized alignment, sorting and conversion to BAM format.

For each input sequence file listed in sampleFile, one BAM file with alignments to the reference genome will be generated, and an additional one for each auxiliary sequence file listed in auxiliaryFile. By default, these BAM files are stored at the same location as the sequence files, unless a different location is specified under alignmentsDir. If compatible alignment files are found at this location, they will not be regenerated, which allows re-use of the same sequencing samples in multiple analysis projects by listing them in several project-specific sampleFiles.

The alignment process produces temporary files, such as decompressed input sequence files or raw alignment files before conversion to BAM format, which can be several times the size of the input sequence files. These temporary files are stored in the directory specified by cacheDir, which defaults to the R process temporary directory returned by tempdir(). The location of tempdir() can be set using environment variables (see ?tempdir).

qAlign returns a qProject object that contains all file names and paths, as well as all alignment parameters necessary for further analysis (see section 7.3 for methods to access the information contained in a qProject object).

7.3 qProject class

The qProject objects are returned by qAlign and contain all information about a sequencing experiment needed for further analysis. It is the main argument passed to the functions that start with a q letter, such as qCount, qQCReport and qExportWig. Some information inside of a qProject object can be accessed by specific methods (in the examples below, x is a qProject object):

- length(x) gets the number of input files.
- genome(x) gets the reference genome as a character(1). The type of genome is stored as an attribute in attr(genome(x), "genomeFormat"): "BSgenome" indicates that genome(x) refers to the name of a BSgenome package, "file" indicates that it contains the path and file name of a genome in FASTA format.
- auxiliaries(x) gets a data.frame with auxiliary target sequences. The data.frame has one row per auxiliary target file, and two columns "File-Name" and "AuxName".
- alignments(x) gets a list with two elements "genome" and "aux". "genome" contains a data.frame with length(x) rows and two columns "FileName" (containing the path to bam files with genomic alignments) and "Sample-Name". "aux" contains a data.frame with one row per auxiliary target

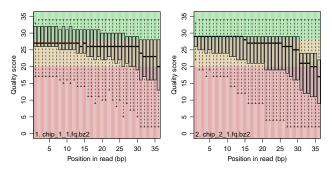
file (with auxiliary names as row names), and length(x) columns (one per input sequence file).

• x[i] returns a qProject object instance with i input files, where i can be an NA-free logical, numeric, or character vector.

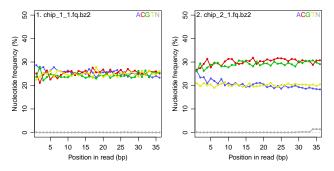
7.4 qQCReport

The qQCReport function samples a random subset of sequences and alignments from each sample or input file and generates a series of diagnostic plots for estimating data quality. The plots below show the currently available plots as produced by the ChIP-seq example in section 6.1 (except for the fragment size distributions which are based on an unspliced alignment of paired-end RNA seq reads):

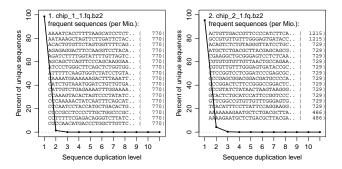
• Quality score boxplot shows the distribution of base quality values as a box plot for each position in the input sequence. The background color (green, orange or red) indicates ranges of high, intermediate and low qualities. The plot is available for fastq only (BAM files may contain base quality information, which is however not used here because reads contained in the BAM file, e.g. aligned reads, may not be a representative sub-sample of all sequenced reads).



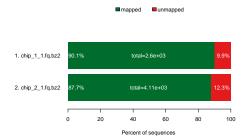
• Nucleotide frequency plot shows the frequency of A, C, G, T and N bases by position in the read. The plot is always available.



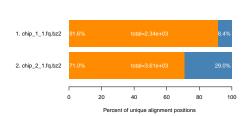
• Duplication level plot shows for each sample the fraction of reads observed at different duplication levels (e.g. once, two-times, three-times, etc.). In addition, the most frequent sequences are listed. The plot is available for fasta or fastq files, but not for bam files, again because contained reads may not be representative for the experiment.



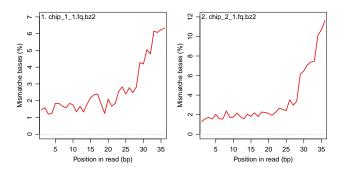
• Mapping statistics shows fractions of reads that were (un)mappable to the reference genome. This plot is available for bam input, i.e. if input is a vector of BAM files, or a qProject with alignment files as returned by qAlign.



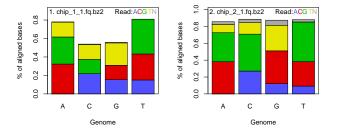
• Library complexity shows fractions of unique read(-pair) alignment positions. Please note that this measure is not independent from the total number of reads in a library, and is best compared between libraries of similar sizes. This plot is available for bam input.



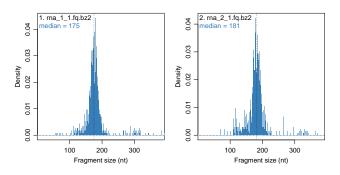
• Mismatch frequency shows the frequency and position (relative to the read sequence) of mismatches in the alignments against the reference genome. The plot is available for bam input.



• Mismatch types shows the frequency of read bases that caused mismatches in the alignments to the reference genome, separately for each genome base. This plot is available for bam input.



• Fragment size shows the distribution of fragment sizes inferred from aligned read pairs. This plot is available for paired-end bam input.



7.5 alignmentStats

alignmentStats is comparable to the "idxstats" function from Samtools; it returns the size of the target sequence, as well as the number of mapped and unmapped reads that are contained in an indexed BAM file. The function works for arguments of type qProject, as well as on a string with one or several BAM file names. There is however a small difference in the two that is illustrated in the following example, which uses the qProject object from the ChIP-seq workflow created on page 23:

- > # using bam files
- > alignmentStats(alignments(proj1)\$genome\$FileName)

alignmentstats(alignments(proji) & genome & File Name)				
seqlength	${\tt mapped}$	unmapped		
95000	2339	258		
95000	3609	505		
lignments(proj1)\$a	aux))		
seqlength	${\tt mapped}$	unmapped		
5386	251	0		
5386	493	0		
ct				
mapped unr	napped			
	seqlength 95000 95000 lignments(p seqlength 5386 5386	seqlength mapped 95000 2339 95000 3609 lignments(proj1)\$a seqlength mapped 5386 251 5386 493		

	seqlength	${\tt mapped}$	unmapped
Sample1:genome	95000	2339	258
Sample2:genome	95000	3609	505
Sample1:phiX174	5386	251	7
Sample2:phiX174	5386	493	12

If calling alignmentStats on the bam files directly as in the first two expressions of the above example, the returned numbers correspond exactly to what you would obtain by the "idxstats" function from Samtools, only that the latter would report them separately for each target sequence, while alignmentStats sums them for each BAM file. These numbers correctly state that there are zero unmapped reads in the auxiliary BAM files. However, if calling alignmentStats on a qProject object, it will report 7 and 12 unmapped reads in the auxiliary BAM files. This is because alignmentStats is aware that unmapped reads are

removed from auxiliary BAM files by QuasR, but can be calculated from the total number of reads to be aligned to the auxiliary target, which equals the number of unmapped reads in the corresponding genomic BAM file.

7.6 qExportWig

qExportWig creates fixed-step wig files (see http://genome.ucsc.edu/goldenPath/help/wiggle.html for format definition) from the genomic alignments contained in a qProject object. The combine argument controls if several input files are combined into a single multi-track wig file, or if they are exported as individual wig files. Alignments of single read experiments can be shifted towards there 3'-end using shift; paired-end alignments are automatically shifted by half the insert size. The resolution of the created wig file is defines by the bin-size argument, and if scaling=TRUE, multiple alignment files in the qProject object are scaled by their total number of reads.

7.7 qCount

qCount is the workhorse for counting alignments that overlap query regions. Usage and details on parameters can be obtained from the qCount function documentation. Two aspects that are of special importance are also discussed here:

7.7.1 Determination of overlap

How an alignment overlap with a query region is defined can be controlled by the following four arguments of qCount:

- selectReadPosition specifies the read base that serves as a reference for overlaps with query regions. The alignment position of that base, eventually after shifting (see below), needs to be contained in the query region for an overlap. selectReadPosition can be set to "start" (the default) or "end", which refer to the biological start (5-end) and end (3-end) of the read. For example, the "start" of a read aligned to the plus strand is the leftmost base in the alignment (the one with the lowest coordinate), and the "end" of a read aligned to the minus strand is also its leftmost base in the alignment.
- shift allows shifting of alignments towards their 3-end prior to overlap determination and counting. This can be helpful to increase resolution of ChIP-seq experiments by moving alignments by half the immuno-precipitated fragment size towards the middle of fragments. shift can either contain "integer" values that specify the shift size, or for paired-end experiments, it can be set to the keyword "halfInsert", which will estimate the true fragment size from the distance between aligned read pairs and shift the alignments accordingly.

- orientation controls the interpretation of alignment strand relative to the strand of the query region. The default value "any" will count all overlapping alignments, irrespective of the strand. This setting is for example used in an unstranded RNA-seq experiment where both sense and antisense reads are generated from an mRNA. A value of "same" will only count the alignments on the same strand as the query region (e.g. in a stranded RNA-seq experiment), and "opposite" will only count the alignments on the opposite strand from the query region (e.g. to quantify anti-sense transcription in a stranded RNA-seq experiment).
- useRead only applies to paired-end experiments and allows to quantify either all alignments (useRead="any"), or only the first (useRead="first") or last (useRead="last") read from each read pair or read group. Note that for useRead="any" (the default), an alignment pair that is fully contained within a query region will contribute two counts to the value of that region.

7.7.2 Running modes of qCount

The features to be quantified are specified by the query argument. At the same time, the type of query selects the mode of quantification. qCount supports three different types of query arguments and implements three corresponding quantification types, which primarily differ in the way they deal with redundancy, such as query bases that are contained in more than one query region. A fourth quantification mode allows counting of alignments supporting exon-exon juctions:

- GRanges query: Overlapping alignments are counted separately for each coordinate region in the query object. If multiple regions have identical names, their counts will be summed, counting each alignment only once even if it overlaps more than one of these regions. Alignments may however be counted more than once if they overlap multiple regions with different names. This mode is for example used to quantify ChIP-seq alignments in promoter regions (see section 6.1 on page 20), or gene expression levels in an RNA-seq experiment (using a 'query' with exon regions named by gene).
- GRangesList query: Alignments are counted and summed for each list element in the query object if they overlap with any of the regions contained in the list element. The order of the list elements defines a hierarchy for quantification: Alignment will only be counted for the first element (the one with the lowest index in the query) that they overlap, but not for any potential further list elements containing overlapping regions. This mode can be used to hierarchically and uniquely count (assign) each alignment to a one of several groups of regions (the elements in the query), for example to estimate the fractions of different classes of RNA in an RNA-seq experiment (rRNA, tRNA, snRNA, snoRNA, mRNA, etc.)

- TranscriptDb query: Used to extract regions from annotation and report alignment counts depending on the value of the reportLevel argument. If reportLevel="exon", alignments overlapping each exon in the query are counted. If reportLevel="gene", alignment counts for all exons of a gene will be summed, counting each alignment only once even if it overlaps multiple annotated exons of a gene. These are useful to calculate exon or gene expression levels in RNA-seq experiments based on the annotation in a TranscriptDB object. If reportLevel="promoter", the promoters function from package GenomicFeatures is used with default arguments to extract promoter regions around transcript start sites, e.g. to quantify alignments inf a ChIP-seq experiment.
- any of the above or NULL for reportLevel="junction": The query argument is ignored if reportLevel is set to "junction", and qCount will count the number of alignments supporting each exon-exon junction detected in any of the samples in proj. The arguments selectReadPosition, shift, orientation, useRead and mask will have no effect in this quantification mode.

7.8 qProfile

The qProfile function differs from qCount in that it returns alignments counts relative to their position in the query region. Except for upstream and downstream, the arguments of qProfile and qCount are the same. This section will describe these two additional arguments; more details on the other arguments are available in section 7.7 and from the qProfile function documentation.

The regions to be profiled are anchored by the biological start position, which are aligned at position zero in the return value. The biological start position is defined as start(query) for regions on the plus strand and end(query) for regions on the minus strand. The anchor positions are extended to the left and right sides by the number of bases indicated in the upstream and downstream arguments.

- upstream indicates the number of bases upstream of the anchor position, which is on the left side of the anchor point for regions on the plus strand and on the right side for regions on the minus strand.
- downstream indicates the number of bases downstream of the anchor position, which is on the left side of the anchor point for regions on the plus strand and on the left side for regions on the minus strand.

Be aware that query regions with a "*" strand are handled the same way as regions on the plus strand.

7.9 qMeth

qMeth is used exclusively for Bis-seq experiments. In contrast to qCount, which counts the number of read alignments per query region, qMeth quantifies the

number of C and T bases per cytosine in query regions, in order to determine methylation status.

gMeth can be run in one of four modes, controlled by the mode argument:

- CpGcomb: Only C's in CpG context are considered. It is assumed that methylation status of the CpG base-pair on both strands is identical. Therefore, the total and methylated counts obtained for the C at position i and the C on the opposite strand at position i + 1 are summed.
- CpG: As with CpGcomb, only C's in CpG context are quantified. However, counts from opposite strand are not summed, resulting in separate output values for C's on both strands.
- allC: All C's contained in query regions are quantified, keeping C's from either strand separate. While this mode allows quantification of non-CpG methylation, it should be used with care, as the large result could use up available memory. In that case, a possible work-around is to divide the region of interest (e.g. the genome) into several regions (e.g. chromosomes) and call qMeth separately for each region.
- var: In this mode, only alignments on the opposite strand from C's are analysed in order to collect evidence for sequence polymorphisms. Methylated C's are hot-spots for C-to-T transitions, which in a Bis-seq experiment cannot be discriminated from completely unmethylated C's. The information is however contained in alignments to the G from the opposite strand: Reads containing a G are consistent with a non-mutated C, and reads with an A support the presence of a sequence polymorphism. qMeth(..., mode="var") returns counts for total and matching bases for all C's on both strands. A low fraction of matching bases is an indication of a mutation and can be used as a basis to identify mutated positions in the studied genome relative to the reference genome. Such positions should not be included in the quantification of methylation.

When using qMeth in a allele-specific quantification (see also section 6.4), cytosines (or CpGs) that overlap a sequence polymorphism will not be quantified.

8 Session information

The output in this vignette was produced under:

```
[3] LC_TIME=en_US.UTF-8 LC_COLLATE=C
```

[5] LC_MONETARY=en_US.UTF-8 LC_MESSAGES=en_US.UTF-8

[7] LC_PAPER=C LC_NAME=C

[9] LC_ADDRESS=C LC_TELEPHONE=C

[11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C

attached base packages:

[1] grid parallel stats graphics grDevices utils

[7] datasets methods base

other attached packages:

[1] BiocInstaller_1.10.3 Gviz_1.4.5

[3] GenomicFeatures_1.12.4 AnnotationDbi_1.22.6

[5] Biobase_2.20.1 rtracklayer_1.20.4

[7] Rsamtools_1.12.4 BSgenome_1.28.0

[9] Biostrings_2.28.0 QuasR_1.0.9

[11] Rbowtie_1.0.3 GenomicRanges_1.12.5

[13] IRanges_1.18.4 BiocGenerics_0.6.0

loaded via a namespace (and not attached):

[1]	DBI_0.2-7	Hmisc_3.12-2	RColorBrewer_1.0-5
[4]	RCurl_1.95-4.1	RSQLite_0.11.4	ShortRead_1.18.0
[7]	XML_3.98-1.1	biomaRt_2.16.0	biovizBase_1.8.1
[10]	bitops_1.0-6	cluster_1.14.4	colorspace_1.2-3
[13]	dichromat_2.0-0	hwriter_1.3	labeling_0.2
[16]	lattice_0.20-23	munsell_0.4.2	plyr_1.8
[19]	rpart_4.1-3	scales_0.2.3	stats4_3.0.1
[22]	stringr_0.6.2	tools_3.0.1	zlibbioc_1.6.0

References

- [1] A. Lerch, D. Gaidatzis, F. Hahne, and M.B. Stadler. Quasr: Quantify and annotate short reads in r. unpublished, 2012.
- [2] P. Dalgaard. Introductory Statistics with R. Springer, 2002.
- [3] B. Langmead, C. Trapnell, M. Pop, and S.L. Salzberg. Ultrafast and memory-efficient alignment of short dna sequences to the human genome. *Genome Biology*, 10(3):R25, 2009.
- [4] F. Hahne, A. Lerch, and M.B. Stadler. Rbowtie: A r wrapper for bowtie and splicemap short read aligners. unpublished, 2012.
- [5] Mark D Robinson, Davis J McCarthy, and Gordon K Smyth. edger: a bioconductor package for differential expression analysis of digital gene expression data. *Bioinformatics*, 26:–1, 2010.

- [6] Simon Anders and Wolfgang Huber. Differential expression analysis for sequence count data. *Genome Biology*, 11:R106, 2010.
- [7] Martin Morgan, Simon Anders, Michael Lawrence, Patrick Aboyoun, Hervé Pagès, and Robert Gentleman. Shortread: a bioconductor package for input, quality assessment and exploration of high-throughput sequence data. *Bioinformatics*, 25:2607–2608, 2009.
- [8] K.F. Au, H. Jiang, L. Lin, Y. Xing, and W.H. Wong. Detection of splice junctions from paired-end rna-seq data by splicemap. *Nucleic Acids Research*, 38(14):4570–4578, 2010.
- [9] H. Pages, P. Aboyoun, R. Gentleman, and S. DebRoy. *Biostrings: String objects representing biological sequences, and matching algorithms.* R package version 2.22.0.