

# Counting reads with summarizeOverlaps

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## 1 Introduction

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This vignette illustrates how reads mapped to a genome can be counted with `summarizeOverlaps`. Different "modes" of counting are provided to resolve reads that overlap multiple features. The built-in count modes are fashioned after the "Union", "IntersectionStrict", and "IntersectionNotEmpty" methods found in the HTSeq package by Simon Anders (see references).

## 2 A First Example

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In this example reads are counted from a list of BAM files and returned in a matrix for use in further analysis such as those offered in [DESeq](#) and [edgeR](#).

```
> library(GenomicAlignments)
> library(DESeq)
> library(edgeR)
> fls <- list.files(system.file("extdata", package="GenomicAlignments"),
+   recursive=TRUE, pattern="*bam$", full=TRUE)
> features <- GRanges(
+   seqnames = c(rep("chr2L", 4), rep("chr2R", 5), rep("chr3L", 2)),
+   ranges = IRanges(c(1000, 3000, 4000, 7000, 2000, 3000, 3600, 4000,
+     7500, 5000, 5400), width=c(rep(500, 3), 600, 900, 500, 300, 900,
+     300, 500, 500)), "-"),
+   group_id=c(rep("A", 4), rep("B", 5), rep("C", 2)))
> olap <- summarizeOverlaps(features, fls)
```

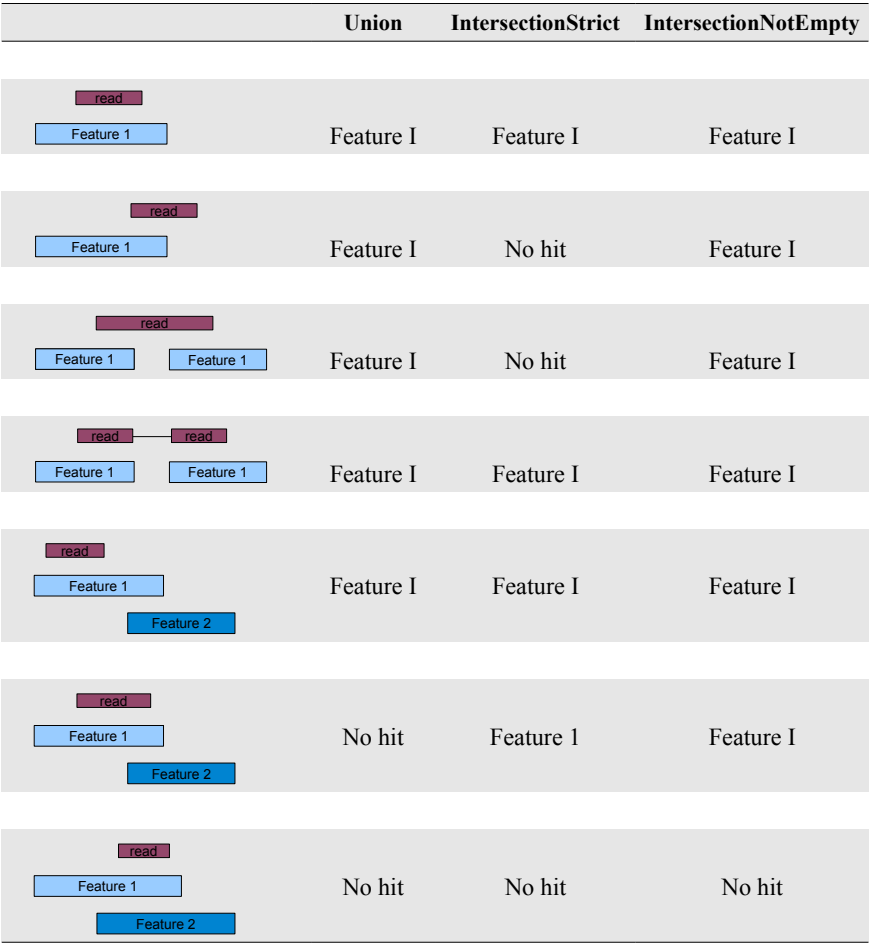
```
> deseq <- newCountDataSet(assay(olap), rownames(colData(olap)))  
> edger <- DGEList(assay(olap), group=rownames(colData(olap)))
```

By default, the `summarizeOverlaps` function iterates through files in 'chunks' and with files processed in parallel. For finer-grain control over memory consumption, use the `BamFileList` function and specify the `yieldSize` argument (e.g., `yieldSize=1000000`) to determine the size of each 'chunk' (smaller chunks consume less memory, but are a little less efficient to process). For controlling the number of processors in use, use `BiocParallel::register` to use an appropriate back-end, e.g., in linux or Mac to process on 6 cores of a single machine use `register(MulticoreParam(workers=6))`; see the [BiocParallel](#) vignette for further details.

### 3 Counting Modes

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The modes of "Union", "IntersectionStrict" and "IntersectionNotEmpty" provide different approaches to resolving reads that overlap multiple features. Figure 1 illustrates how both simple and gapped reads are handled by the modes. Note that a read is counted a maximum of once; there is no double counting. For additional detail on the counting modes see the `summarizeOverlaps` man page.



\* Picture reproduced from HTSeq web site :  
<http://www-huber.embl.de/users/anders/HTSeq/doc/count.html>

Figure 1: Counting Modes

## 4 Counting Features

Features can be exons, transcripts, genes or any region of interest. The number of ranges that define a single feature is specified in the `features` argument.

When annotation regions of interest are defined by a single range a *GRanges* should be used as the `features` argument. With a *GRanges* it is assumed that each row (i.e., each range) represents a distinct feature. If `features` was a *GRanges*

of exons, the result would be counts per exon.

When the region of interest is defined by one or more ranges the *features* argument should be a *GRangesList*. In practice this could be a list of exons by gene or transcripts by gene or other similar relationships. The count result will be the same length as the *GRangesList*. For a list of exons by genes, the result would be counts per gene.

The combination of defining the features as either *GRanges* or *GRangesList* and choosing a counting mode controls how *summarizeOverlaps* assigns hits. Regardless of the mode chosen, each read is assigned to at most a single feature. These options are intended to provide flexibility in defining different biological problems.

This next example demonstrates how the same read can be counted differently depending on how the *features* argument is specified. We use a single read that overlaps two ranges, *gr1* and *gr2*.

```
> rd <- GAlignments("a", seqnames = Rle("chr1"), pos = as.integer(100),
+   cigar = "300M", strand = strand("+"))
> gr1 <- GRanges("chr1", IRanges(start=50, width=150), strand="+")
> gr2 <- GRanges("chr1", IRanges(start=350, width=150), strand="+")
```

When provided as a *GRanges* both *gr1* and *gr2* are considered distinct features. In this case none of the modes count the read as a hit. Mode *Union* discards the read because more than 1 feature is overlapped. *IntersectionStrict* requires the read to fall completely within a feature which is not the case for either *gr1* or *gr2*. *IntersectionNotEmpty* requires the read to overlap a single unique disjoint region of the features. In this case *gr1* and *gr2* do not overlap so each range is considered a unique disjoint region. However, the read overlaps both *gr1* and *gr2* so a decision cannot be made and the read is discarded.

```
> gr <- c(gr1, gr2)
> data.frame(union = assay(summarizeOverlaps(gr, rd)),
+   intStrict = assay(summarizeOverlaps(gr, rd,
+   mode="IntersectionStrict")),
+   intNotEmpty = assay(summarizeOverlaps(gr, rd,
+   mode="IntersectionNotEmpty"))))
```

	reads	reads.1	reads.2
1	0	0	0
2	0	0	0

Next we count with features as a *GRangesList*; this is list of length 1 with 2 elements. Modes *Union* and *IntersectionNotEmpty* both count the read for the single feature.

```
> grl <- GRangesList(c(gr1, gr2))
> data.frame(union = assay(summarizeOverlaps(grl, rd)),
+   intStrict = assay(summarizeOverlaps(grl, rd,
+   mode="IntersectionStrict")),
+   intNotEmpty = assay(summarizeOverlaps(grl, rd,
+   mode="IntersectionNotEmpty"))))
```

	reads	reads.1	reads.2
1	1	0	1

In this more complicated example we have 7 reads, 5 are simple and 2 have gaps in the CIGAR. There are 12 ranges that will serve as the features.

```
> group_id <- c("A", "B", "C", "C", "D", "D", "E", "F", "G", "G", "H", "H")
> features <- GRanges(
+   seqnames = Rle(c("chr1", "chr2", "chr1", "chr1", "chr2", "chr2",
+   "chr1", "chr1", "chr2", "chr2", "chr1", "chr1")),
+   strand = strand(rep("+", length(group_id))),
+   ranges = IRanges(
+   start=c(1000, 2000, 3000, 3600, 7000, 7500, 4000, 4000, 3000, 3350, 5000, 5400),
+   width=c(500, 900, 500, 300, 600, 300, 500, 900, 150, 200, 500, 500)),
```

```

+   DataFrame(group_id)
+ )
> reads <- GAlignments(
+   names = c("a", "b", "c", "d", "e", "f", "g"),
+   seqnames = Rle(c(rep(c("chr1", "chr2"), 3), "chr1")),
+   pos = as.integer(c(1400, 2700, 3400, 7100, 4000, 3100, 5200)),
+   cigar = c("500M", "100M", "300M", "500M", "300M", "50M200N50M", "50M150N50M"),
+   strand = strand(rep.int("+", 7L)))
>

```

Using a *GRanges* as the features all 12 ranges are considered to be different features and counts are produced for each row,

```

> data.frame(union = assay(summarizeOverlaps(features, reads)),
+   intStrict = assay(summarizeOverlaps(features, reads,
+   mode="IntersectionStrict")),
+   intNotEmpty = assay(summarizeOverlaps(features, reads,
+   mode="IntersectionNotEmpty")))

```

	reads	reads.1	reads.2
1	1	0	1
2	1	1	1
3	0	0	0
4	0	0	0
5	0	1	1
6	0	0	0
7	0	0	0
8	0	0	0
9	0	0	0
10	0	0	0
11	0	1	1
12	0	0	0

When the data are split by group to create a *GRangesList* the highest list-levels are treated as different features and the multiple list elements are considered part of the same features. Counts are returned for each group.

```

> lst <- split(features, mcols(features)[["group_id"]])
> length(lst)

```

```
[1] 8
```

```

> data.frame(union = assay(summarizeOverlaps(lst, reads)),
+   intStrict = assay(summarizeOverlaps(lst, reads,
+   mode="IntersectionStrict")),
+   intNotEmpty = assay(summarizeOverlaps(lst, reads,
+   mode="IntersectionNotEmpty")))

```

	reads	reads.1	reads.2
A	1	0	1
B	1	1	1
C	1	0	1
D	1	1	1
E	0	0	0
F	0	0	0
G	1	1	1
H	1	1	1

If desired, users can supply their own counting function as the mode argument and take advantage of the infrastructure for

counting over multiple BAM files and parsing the results into a *RangedSummarizedExperiment* object. See `?BamViews-class` or `?BamFile-class` in the *Rsamtools* package.

## 5 pasilla Data

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In this exercise we count the *pasilla* data by gene and by transcript then create a *CountDataSet*. This object can be used in differential expression methods offered in the *DESeq* package.

### 5.1 source files

Files are available through NCBI Gene Expression Omnibus (GEO), accession number GSE18508. <http://www.ncbi.nlm.nih.gov/projects/geo/query/acc.cgi?acc=GSE18508>. SAM files can be converted to BAM with the `asBam` function in the *Rsamtools* package. Of the seven files available, 3 are single-reads and 4 are paired-end. Smaller versions of untreated1 (single-end) and untreated2 (paired-end) have been made available in the *pasillaBamSubset* package. This subset includes chromosome 4 only.

`summarizeOverlaps` is capable of counting paired-end reads in both a `BamFile`-method (set argument `singleEnd=TRUE`) or a `GAlignmentPairs`-method. For this example, we use the 3 single-end read files,

- treated1.bam
- untreated1.bam
- untreated2.bam

Annotations are retrieved as a GTF file from the ENSEMBL web site. We download the file our local disk, then use *Rtracklayer*'s `import` function to parse the file to a *GRanges* instance.

```
> library(rtracklayer)
> fl <- paste0("ftp://ftp.ensembl.org/pub/release-62/",
+             "gtf/drosophila_melanogaster/",
+             "Drosophila_melanogaster.BDGP5.25.62.gtf.gz")
> gffFile <- file.path(tempdir(), basename(fl))
> download.file(fl, gffFile)
> gff0 <- import(gffFile)
```

Subset on the protein-coding, exon regions of chromosome 4 and split by gene id.

```
> idx <- mcols(gff0)$source == "protein_coding" &
+       mcols(gff0)$type == "exon" &
+       seqnames(gff0) == "4"
> gff <- gff0[idx]
> ## adjust seqnames to match Bam files
> seqlevels(gff) <- paste("chr", seqlevels(gff), sep="")
> chr4genes <- split(gff, mcols(gff)$gene_id)
```

### 5.2 counting

The `param` argument can be used to subset the reads in the bam file on characteristics such as position, unmapped or paired-end reads. Quality scores or the "NH" tag, which identifies reads with multiple mappings, can be included as metadata columns for further subsetting. See `?ScanBamParam` for details about specifying the `param` argument.

```
> param <- ScanBamParam(
+   what='qual',
+   which=GRanges("chr4", IRanges(1, 1e6)),
```

```
+          flag=scanBamFlag(isUnmappedQuery=FALSE, isPaired=NA),
+          tag="NH")
```

We use `summarizeOverlaps` to count with the default mode of "Union". If a `param` argument is not included all reads from the BAM file are counted.

```
> fls <- c("treated1.bam", "untreated1.bam", "untreated2.bam")
> path <- "pathToBAMFiles"
> bamlst <- BamFileList(fls)
> genehits <- summarizeOverlaps(chr4genes, bamlst, mode="Union")
```

A `CountDataSet` is constructed from the counts and experiment data in *pasilla*.

```
> expdata = MIAME(
+   name="pasilla knockdown",
+   lab="Genetics and Developmental Biology, University of
+     Connecticut Health Center",
+   contact="Dr. Brenton Graveley",
+   title="modENCODE Drosophila pasilla RNA Binding Protein RNAi
+     knockdown RNA-Seq Studies",
+   pubMedIds="20921232",
+   url="http://www.ncbi.nlm.nih.gov/projects/geo/query/acc.cgi?acc=GSE18508",
+   abstract="RNA-seq of 3 biological replicates of from the Drosophila
+     melanogaster S2-DRSC cells that have been RNAi depleted of mRNAs
+     encoding pasilla, a mRNA binding protein and 4 biological replicates
+     of the the untreated cell line.")
> design <- data.frame(
+   condition=c("treated", "untreated", "untreated"),
+   replicate=c(1,1,2),
+   type=rep("single-read", 3),
+   countfiles=path(colData(genehits)[,1]), stringsAsFactors=TRUE)
> geneCDS <- newCountDataSet(
+   countData=assay(genehits),
+   conditions=design)
> experimentData(geneCDS) <- expdata
> sampleNames(geneCDS) = colnames(genehits)
```

If the primary interest is to count by transcript instead of by gene, the annotation file can be split on transcript id.

```
> chr4tx <- split(gff, mcols(gff)$transcript_id)
> txhits <- summarizeOverlaps(chr4tx, bamlst)
> txCDS <- newCountDataSet(assay(txhits), design)
> experimentData(txCDS) <- expdata
```

## 6 References

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<http://www-huber.embl.de/users/anders/HTSeq/doc/overview.html> <http://www-huber.embl.de/users/anders/HTSeq/doc/count.html>