

BioC 2016 Developer Day

Core team updates

Welcome and Project Update

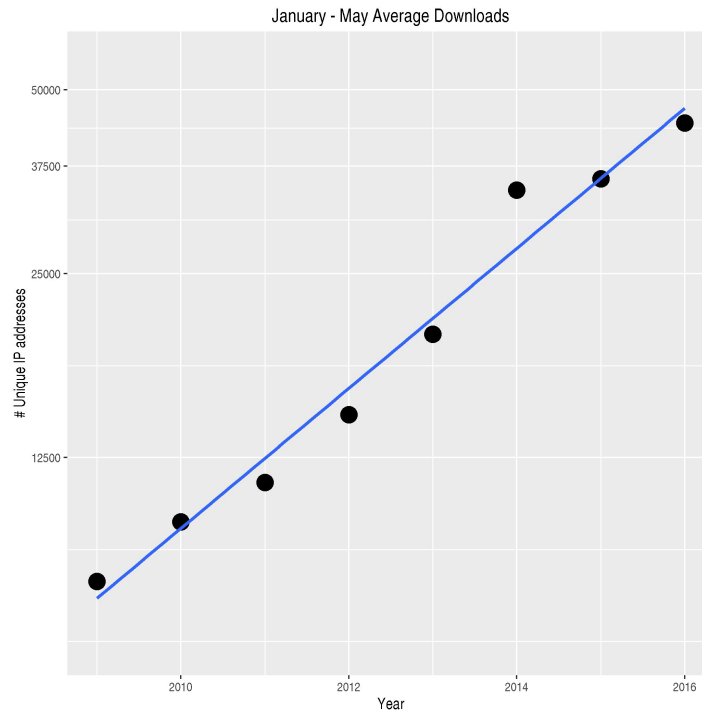
Thanks!



- Pam Jarrett, Ellen Sanders Noonan, Ellen Van Stone
- Susan Holmes, Sean Davis
- Speakers and Workshop presenters
- Bioc developers!

The project since last year

- 2 Releases with 187 new packages
- Lots of activity on the support site
- Steadily growing user base
- Move to Roswell Park



Activities and opportunities

Core team activities

- *GenomicRanges* infrastructure
- *AnntotationHub* and *ExperimentHub*
- *BiocParallel* / *GenomicFiles*
- Progress on *MultiAssayExperiment*
- On-disk / lazy evaluation of large data
- Public new package submissions
- User and developer support

Keeping up with the burgeoning R community

- Package development best practices
- Approaches to version control and testing

Increasingly cloud-based computing

- Efficient access to cloud-based resources
- Participation in cloud-based bioinformatics initiatives
- Computation in the cloud

Career opportunities!

- [Senior Programmer / Analyst](#) -- creative web / system administration / development -- <https://goo.gl/2s26pp>

Acknowledgements

Core team (current & recent): Valerie Obenchain, Hervé Pagès, Dan Tenenbaum, Lori Shepherd, Marcel Ramos, Jim Hester, Jim Java, Brian Long, Sonali Arora, Nate Hayden, Paul Shannon, Marc Carlson

Technical advisory board: Vincent Carey, Wolfgang Huber, Robert Gentleman, Rafael Irizzary, Levi Waldron, Michael Lawrence, Sean Davis, Aedin Culhane

Scientific advisory board: Simon Tavaré (CRUK), Paul Flicek (EMBL/EBI), Simon Urbanek (AT&T), Vincent Carey (Brigham & Women's), Wolfgang Huber (EBI), Rafael Irizzary (Dana Farber), Robert Gentleman (23andMe)



Research reported in this presentation was supported by the National Human Genome Research Institute and the National Cancer Institute of the National Institutes of Health under award numbers U41HG004059 and U24CA180996. The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health.

Lori Shepherd

GenomicFiles / VcfStack / RangedVcfStack

disjoin() in IRanges / GenomicRanges

GenomicFiles

VcfStack / RangedVcfStack

VcfStack / RangedVcfStack

The VcfStack class is a vector of related VCF files, for instance each file representing a separate chromosome. The class helps manage these files as a group.

The RangedVcfStack class extends VcfStack by associating genomic ranges of interest to the collection of VCF files.

VcfStack / RangedVcfStack

`VcfStack(files=NULL, seqinfo=NULL, colData=NULL)`

`files`: A character vector of files paths pointing to VCF files. The character vector must be named, with names correspond to `seqnames` in each VCF file.

`seqinfo`: A `Seqinfo` object describing the levels genome and circularity of each sequence.

`colData`: An optional `DataFrame` describing each sample in the `VcfStack`. When present, row names must correspond to sample names in the VCF file.

`RangedVcfStack(vs=NULL, rowRanges=NULL)`

`vs`: A `VcfStack` object.

`rowRanges`: An optional `GRanges` object associating the genomic ranges of interest to the collection of VCF files. The `seqnames` of `rowRanges` are a subset of `seqnames(vs)`. If missing, a default is created from the `seqinfo` object of the provided `VcfStack`

VcfStack / RangedVcfStack Accessors

- `dim(x)`
- `dimnames(x)`
- `rownames(x)`
- `colnames(x)`

As well as your typical getters and setters for object attributes:

- `files(x)`
- `seqinfo(x)`
- `colData(x)`
- `rowRanges(x)`

VcfStack / RangedVcfStack Methods

`assay(x, i, ...)`

Get matrix of genotype calls from VCF files

`readVcfStack(x, i, j=colnames(x))`

Get content of VCF files in the VcfStack

`show(x)`

Display abbreviated information about VcfStack / RangedVcfStack

- i: indicated which files to read
is a GRanges object, character() vector of seqnames, numeric() vector, logical() vector, or can be missing. For a RangedVcfStack object, assay and readVcfStack will use the associated rowRanges object for i.
- j: indicates which samples to read
can be missing or a character() vector of sample names

VcfStack / RangedVcfStack Subsetting

`x[i, j]`

Get elements from ranges `i` and samples `j` as a `VcfStack` or `RangedVcfStack` object

`x`: is a `VcfStack` or `RangedVcfStack` object

`i`: indicated which files to subset

can be missing, a `character()` vector of seqnames, `numeric()` vector of indexes, or `logical()` vector.
When `x` is a `VcfStack` instance, `i` can also be a `GRanges` object; `seqnames(i)` is then used to subset the files in the `VcfStack`.

`j`: indicated which samples to subset.

can be missing, a `character()` vector of sample names, a `numeric()` vector, or `logical()` vector.

IRanges / GenomicRanges

`disjoin()`

IRanges / GenomicRanges

`disjoin(x, with.revmap=FALSE)`

- Ranges
- RangesList
- CompressedIRangesList

`disjoin(x, with.revmap=FALSE, ignore.strand=FALSE)`

- GenomicRanges
- GRangesList

`with.revmap`

TRUE or FALSE. Should the mapping from output to input ranges be stored in the returned object? If yes, then it is stored as metadata column `revmap` of type `IntegerList`

GenomicRanges 'GRanges' Example

```
> gr <- GRanges(Rle(c("chr1", "chr3"), c(2, 2)),  
                IRanges(c(8,6,8,6),c(11,15,11,15)), names=c("k","l","m","n")),  
                Rle(strand(c("-", "-", "+", "*"))),  
                score=11:14, GC=c(.2,.3,.3,.1))
```

> gr

GRanges object with 4 ranges and 2 metadata columns:

	seqnames	ranges	strand	score	GC
	<Rle>	<IRanges>	<Rle>	<integer>	<numeric>
k	chr1	[8, 11]	-	11	0.2
l	chr1	[6, 15]	-	12	0.3
m	chr3	[8, 11]	+	13	0.3
n	chr3	[6, 15]	*	14	0.1

seqinfo: 2 sequences from an unspecified genome; no
seqlengths

> dgr <- disjoint(gr, with.revmap=TRUE)

> dgr

GRanges object with 5 ranges and 1 metadata column:

	seqnames	ranges	strand	revmap
	<Rle>	<IRanges>	<Rle>	<IntegerList>
[1]	chr1	[6, 7]	-	2
[2]	chr1	[8, 11]	-	1,2
[3]	chr1	[12, 15]	-	2
[4]	chr3	[8, 11]	+	3
[5]	chr3	[6, 15]	*	4

seqinfo: 2 sequences from an unspecified genome; no seqlengths

To Get Original Metadata Values:

```
> revmap <- mcols(dgr)$revmap  
> score <- extractList(mcols(gr)$score, revmap)  
> GC <- extractList(mcols(gr)$GC, revmap)  
> mcols(dgr)$score <- score  
> mcols(dgr)$GC <- GC  
> dgr
```

GRanges object with 5 ranges and 3 metadata columns:

	seqnames	ranges	strand	revmap	score	GC
	<Rle>	<IRanges>	<Rle>	<IntegerList>	<IntegerList>	<NumericList>
[1]	chr1	[6, 7]	-	2	12	0.3
[2]	chr1	[8, 11]	-	1,2	11,12	0.2,0.3
[3]	chr1	[12, 15]	-	2	12	0.3
[4]	chr3	[8, 11]	+	3	13	0.3
[5]	chr3	[6, 15]	*	4	14	0.1

seqinfo: 2 sequences from an unspecified genome; no seqlengths

GenomicRanges 'GRangesList' Example

```
gr <- GRanges(Rle(c("chr1", "chr3"), c(2, 2)),
              IRanges(c(8,6,8,6),c(11,15,11,15)), names=c("k","l","m","n")),
              Rle(strand(c("-", "-", "+", "*"))),
              score=11:14, GC=c(.2,.3,.3,.1))
grl <- GRangesList(gr, gr)
```

> grl

GRangesList object of length 2:

[[1]]

GRanges object with 4 ranges and 2 metadata columns:

seqnames	ranges	strand	score	GC
<Rle>	<IRanges>	<Rle>	<integer>	<numeric>
k	chr1 [8, 11]	-	11	0.2
l	chr1 [6, 15]	-	12	0.3
m	chr3 [8, 11]	+	13	0.3
n	chr3 [6, 15]	*	14	0.1

[[2]]

GRanges object with 4 ranges and 2 metadata columns:

seqnames	ranges	strand	score	GC
k	chr1 [8, 11]	-	11	0.2
l	chr1 [6, 15]	-	12	0.3
m	chr3 [8, 11]	+	13	0.3
n	chr3 [6, 15]	*	14	0.1

seqinfo: 2 sequences from an unspecified genome; no seqlengths

> disjoint(grl, with.revmap=TRUE)

GRangesList object of length 2:

[[1]]

GRanges object with 5 ranges and 1 metadata column:

seqnames	ranges	strand	revmap
<Rle>	<IRanges>	<Rle>	<IntegerList>
[1]	chr1 [6, 7]	-	2
[2]	chr1 [8, 11]	-	1,2
[3]	chr1 [12, 15]	-	2
[4]	chr3 [8, 11]	+	3
[5]	chr3 [6, 15]	*	4

[[2]]

GRanges object with 5 ranges and 1 metadata column:

seqnames	ranges	strand	revmap
[1]	chr1 [6, 7]	-	2
[2]	chr1 [8, 11]	-	1,2
[3]	chr1 [12, 15]	-	2
[4]	chr3 [8, 11]	+	3
[5]	chr3 [6, 15]	*	4

seqinfo: 2 sequences from an unspecified genome; no seqlengths

Valerie Obenchain

ExperimentHub

ExperimentHub

Resource to house curated data from experiments, publications or courses

Similar interface as AnnotationHub except ...

- Parent package documentation
- List resources by package
- Interface with the data through the package or ExperimentHub
- All data stored in AWS S3; no web downloads

ExperimentHub: parent package documentation

```
> library(ExperimentHub)
```

```
> eh = ExperimentHub()  
snapshotDate(): 2016-06-08
```

```
> eset = eh[[100]]
```

```
see ?curatedMetagenomicData and browseVignettes('curatedMetagenomicData') for documentation  
downloading from 'https://experimenthub.bioconductor.org/fetch/100'
```

```
retrieving 1 resource
```

```
|=====| 100%
```

```
> ?curatedMetagenomicData
```

ExperimentHub: list resources by package

```
> head(package(eh), 3)
```

EH1	EH2	EH3
"GSE62944"	"curatedMetagenomicData"	"curatedMetagenomicData"

```
> table(package(eh))
```

curatedMetagenomicData	GSE62944
162	1

ExperimentHub: interface with data via package

```
> eh["EH100"]
ExperimentHub with 1 record
# snapshotDate(): 2016-06-08
# package(): curatedMetagenomicData
# $dataprovder: Human Microbiome Project Consortium
# $species: Homo sapiens
# $title: hmp.r_retroauricular_crease.marker_ab.eset.rda
...

> ?hmp.r_retroauricular_crease.marker_ab.eset ## package man page
> hmp.r_retroauricular_crease.marker_ab.eset() ## loads the data
> hmp.r_retroauricular_crease.marker_ab.eset(metadata = TRUE) ## loads the metadata
```

ExperimentHubData

Information on adding resources to ExperimentHub is found in the ExperimentHubData [vignette](#).

Marcel Ramos

MultiAssayExperiment

MultiAssayExperiment

A package to manage multiple assays on sets of samples or specimens

- A container class for handling overlapping sets of samples
- User-friendly operations (subsetting)
- Mapping scheme for relating samples to participants or experiment results to specimen data
- Set up for common genomic computations across diverse assays
- On-disk representation of data (moving to lazy eval with `HDF5Array`)

Hierarchy of information:

Study

└ Experiment

└ Biological Unit



Datasets will soon be accessible via ExperimentHub

MultiAssayExperiment: Structure Overview

- **MultiAssayExperiment** class
 - **Elist** class and slot - *workhorse container*
 - Any class that has a `[]` bracket method, ``colnames``, ``rownames`` and ``dim``.
 - *RangedRaggedAssay*
 - *SummarizedExperiment, RangedSummarizedExperiment*
 - *ExpressionSet*
 - *matrix*
 - **pData** (of class *DataFrame*) - *specimen description*
 - Each row is a patient or specimen
 - Includes demographics and/or other specimen-wide variables
 - **sampleMap** (of class *DataFrame*) - *mapping scheme*
 - Maps sample identifiers to participants/specimen in a table
 - **metadata** (*ANY* class)
 - Include additional study level information

MultiAssayExperiment: Quick Example

```
> library(MultiAssayExperiment)
```

```
> example("MultiAssayExperiment")
```

```
> myMultiAssayExperiment
```

A "MultiAssayExperiment" object of 3 listed experiments with user-defined names and respective classes.

Containing an "Elist" class object of length 3:

[1] Affy: "ExpressionSet" - 2 rows, 4 columns

[2] Methyl450k: "matrix" - 2 rows, 5 columns

[3] CNVgistic: "RangedRaggedAssay" - 5 rows, 3 columns

To access slots use:

elist() - to obtain the "Elist" of experiment instances

pData() - for the primary/phenotype "DataFrame"

sampleMap() - for the sample availability "DataFrame"

metadata() - for the metadata object of "ANY" class

See also: subsetByAssay(), subsetByRow(), subsetByColumn()

MultiAssayExperiment: Thorough Example

An in-depth example on how to build your own **MultiAssayExperiment** can be found in the package [vignette](#)

Hervé Pagès

Recent developments:

- GPos class
- HDF5Array, DelayedArray

What's next?

GPos

A very light GRanges-like container for storing a set of *positions* along the genome.

Particularly memory-efficient when the object contains long runs of adjacent positions.

Can be put on a SummarizedExperiment object (as rowRanges).

```
> gpos
GPos object with 12162995 positions and 0 metadata columns:
      seqnames      pos strand
      <Rle> <integer> <Rle>
[1]      chrI         1      *
[2]      chrI         2      *
[3]      chrI         3      *
...
[12162993] 2micron     6316     *
[12162994] 2micron     6317     *
[12162995] 2micron     6318     *
-----
seqinfo: 18 sequences (2 circular) from sacCer2 genome
```

All the single positions along the Yeast genome are represented.

```
> object.size(gpos)
14000 bytes
```

GPos

Metadata columns need to be light too.

Good candidates:

- Rle (e.g. coverage)
- DNASTring
- sparse object (e.g. Matrix)
- on-disk object (e.g. HDF5Array)
- ?

Current limitation: length of a GPos object cannot exceed 2^{31} (2 billions).

See ?GPos in the GenomicRanges package for more information.

```
> gpos
GPos object with 12162995 positions and 2 metadata columns:
      seqnames      pos strand |      cov      dna
      <Rle> <integer> <Rle> | <Rle> <DNASTring>
[1]      chrI         1      * |      0      C
[2]      chrI         2      * |      0      C
[3]      chrI         3      * |      0      A
[4]      chrI         4      * |      0      C
[5]      chrI         5      * |      0      A
...      ...      ...      ...      ...
[12162991] 2micron     6314      * |      0      A
[12162992] 2micron     6315      * |      0      A
[12162993] 2micron     6316      * |      0      C
[12162994] 2micron     6317      * |      0      G
[12162995] 2micron     6318      * |      0      A
-----
seqinfo: 18 sequences (2 circular) from sacCer2 genome
```

HDF5Array / DelayedArray

Convenient access and manipulation of HDF5 datasets.

Can be used inside a SummarizedExperiment object (assay data).

A dataset with coverage for 6 samples along Human chr 16:

```
> cov0 <- HDF5Array(tally_file, "/ExampleStudy/16/Coverages")
> cov0
HDF5Array object of 6 x 2 x 90354753 integers:
, , 1
      [,1] [,2]
[1,]    0    0
[2,]    0    0
...     .     .
[5,]    0    0
[6,]    0    0
...
, , 90354753
      [,1] [,2]
[1,]    0    0
[2,]    0    0
...     .     .
[5,]    0    0
[6,]    0    0
```

HDF5Array / DelayedArray

Support delayed operations.

Result is a DelayedArray object.

as `.array()` would **realize it in memory**. Don't do that!

Instead **realize it on disk** (if you really need to) with `writeHDF5Dataset()`.

```
Compute unstranded coverage:
```

```
> pcov <- drop(cov0[ , 1, ]) # delayed
> mcov <- drop(cov0[ , 2, ]) # delayed
> cov <- pcov + mcov        # delayed
> cov
```

```
DelayedMatrix object of 6 x 90354753 integers:
```

```
          [,1]      [,2]      [,3]      . [,90354751]
[1,]          0          0          0      .          0
[2,]          0          0          0      .          0
[3,]          0          0          0      .          0
[4,]          0          0          0      .          0
[5,]          0          0          0      .          0
[6,]          0          0          0      .          0
          [,90354752] [,90354753]
[1,]          0          0
[2,]          0          0
[3,]          0          0
[4,]          0          0
[5,]          0          0
[6,]          0          0
```

HDF5Array / DelayedArray

Block-processing:

- Operations that cannot be delayed (e.g. `rowSums()` or matrix multiplication) process the `DelayedArray` object block-by-block, one block at a time.
- Each block is *realized* (i.e. all delayed operations are executed) and the current operation (e.g. `rowSums`) applied to the result.

See `?DelayedArray` in the `HDF5Array` package for more information.

```
> sum_cov <- rowSums(cov) # block-processing
> sum_cov
[1] 39807797 45246576 18405376 36487401 17218497 36681571

> gc()
      used (Mb) gc trigger (Mb) max used (Mb)
Ncells 2947878 157.5   4703850 251.3  4703850 251.3
Vcells 3765245  28.8   67472700 514.8 58464312 446.1
```

Loading the full dataset at once in memory would use 4 Gb of RAM!

What's next?

- ❖ **HDF5Array:**
 - Support more operations on DelayedArray objects
 - Vignette
 - Integration of HDF5Array to some common workflows (e.g. `summarizeOverlaps`)
- ❖ Support **long Vector derivatives** (e.g. long Rle, long DataFrame, long GRanges, long Hits, long DNASTring, long DNASTringSet, etc). Will require important changes to the internals of several core packages (S4Vectors, IRanges, GenomicRanges, Biostrings, and more...)
- ❖ **On-disk GRanges objects.** Indexed for fast extraction of elements that overlap a set of regions of interest (i.e. `fastSubsetByOverlaps`). Analog to `scanBam` “which” feature. An immediate use case for this is to speed up `snpsByOverlaps`.
- ❖ Support **easy creation of standalone BSgenome objects** (from 2bit, FASTA, and maybe other sources).
- ❖ Maybe other **"genomic Views"** objects (in addition to `BSgenomeViews`).
- ❖ Build system: **incremental builds.**