

Approaches to standardizing parallel evaluation in *Bioconductor*

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Abstract: The *Bioconductor* project represents almost 1000 core and contributed packages for the analysis and comprehension of high-throughput genomic data. The project core provides software infrastructure tailored to our common use cases, including facilities for parallel evaluation via the *BiocParallel* and other packages. *BiocParallel* has had mixed success, simplifying cross-platform compatibility but imperfectly exploiting heterogeneous computational environments and inspiring creative parallel computation.

Outline: *R* / *Bioconductor* for Integrative Analysis

1. The *Bioconductor* project
2. Use cases
3. *BiocParallel* & Friends
4. Critique
5. Prospects

I'm sorry to have left so suddenly. I was taken ill during your talk and had to go home. I am still ill in fact one week later. – Martyn Plummer, President of the R Foundation, 15 Jan 2015.

Bioconductor

Goal Analysis and comprehension of high-throughput genomic data

Focus

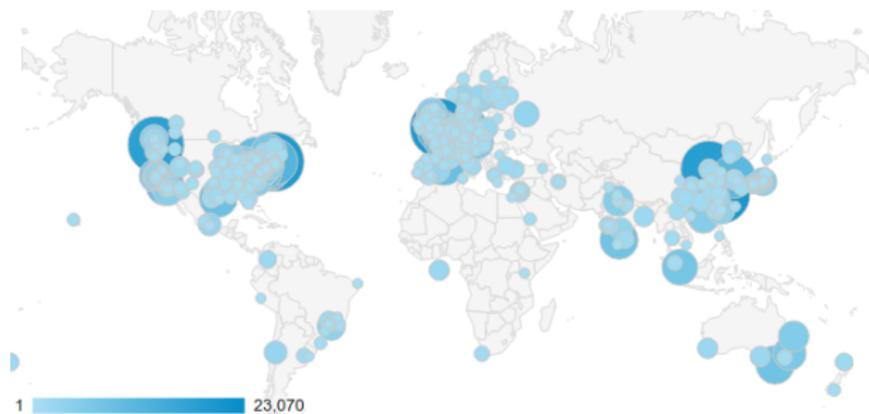
- ▶ Sequencing; RNA-Seq, ChIP-Seq, Variants, ...
- ▶ Expression and other microarrays; flow cytometry; proteomics, imaging

Themes

- ▶ 'Core' and (primarily academic) community contributions.
- ▶ *R* – statistics, visualization, interoperability
- ▶ Reproducible – data structures, scripts, *vignettes*, packages
- ▶ Interoperable – formal classes in 'core' packages
- ▶ Accessible: affordable, transparent, usable

Huber et al., Orchestrating high-throughput genomic analysis with *Bioconductor*. *Nature Methods*: soon!

Project status (December, 2014)



2014 [web site](#) visitors, by city

- ▶ 320,000 unique IP address package downloads / year
- ▶ 1,300 [support site](#) contributors / year, 8,200 visitors / month
- ▶ 10,500 PubMed Central mentions of 'Bioconductor';
≈ 22,000 citations to *Bioconductor* packages
- ▶ Funding from US NIH & NSF, and EC

Use Cases: High-Throughput Sequencing

Questions

- ▶ Which genes are differentially expressed in cancer versus normal tissue?
- ▶ Which transcription factors are regulating gene expression?
- ▶ What single nucleotide polymorphisms (SNPs) are present in a population / associated with a disease?
- ▶ What is the ChIP-seq regulatory signal along a linear genome?

Sample sizes

- ▶ Designed experiments – e.g., 10's or 100's of samples
- ▶ Cohorts – e.g., 100's or 1000's of patients
- ▶ Populations – 1000's - 10000's of individuals

Attributes

- ▶ 10,000's of genes
- ▶ Millions of variants

Use Cases

Patterns

1. Reduction – large idiosyncratic ('BAM') files reduced to e.g., count matrix of $100,000 \times 100$.
2. Intermediate expansion (e.g., pairwise interactions between SNPs. . .) & reduction (. . . reaching statistical significance) – *MatrixEQTL*.
3. Query-like, e.g., predict SNP effects; drill down on subsets
4. 1-dimensional linear dependency

'Academic' work environment

- ▶ Local or shared computer with 10's of cores and moderate memory.
- ▶ Cluster with possibly idiosyncratic batch scheduler.
- ▶ More than 1/2 of our web site visitors are Windows users!

BiocParallel & Friends: Strategies for Large Data

Memory management

- ▶ Restrict input to relevant 'columns'.
- ▶ Select relevant rows.
- ▶ Iterate: read in and operate on successive chunks.

Speed

- ▶ Efficient *R* code – 10-100× speed-up. All gravy.
- ▶ C implementation – 1-5× speed-up. Tedious, error-prone, multiple languages.
- ▶ Parallel evaluation – 2-10× speed-up. Debugging & error recovery; local expertise. Implies memory management.
- ▶ *BiocParallel*, *GenomicFiles*, *Streamer*

Lawrence, M, and Morgan, M. (2014) Scalable Genomics with R and *Bioconductor*. *Statistical Science*, Vol. 29, No. 2, 214-226.

BiocParallel...

How

```
register(MulticoreParam(workers=4)) # stack  
ans <- bplapply(X, FUN, ..., BPPARAM=bpparam())
```

Why

- ▶ Easy(er) cross-platform use – registry of OS-specific back ends.
- ▶ Standardized front end (`bplapply`, `bpvec`, ...) to diverse back-ends (*BiocParallelParam*).
 - ▶ **Multicore**, `snow`, `Rmpi`, **BatchJobs** (reasonable interface to cluster schedulers)
 - ▶ Familiar (?) functional style.
- ▶ Spawned jobs: interactive; direct use of existing code.
- ▶ Registration stack supports coarse-grained nested parallelism

... & Friends

GenomicFiles

- ▶ Manage files underlying many biological applications – references, iteration, restriction, ...
- ▶ Distribution of file references (paths) and shared file system as 'state of the art'

Streamer

- ▶ Compose work flows connecting iterative data input functions through serial and parallel operations to data output.
- ▶ 'Yield' on the stream pulls a chunk of data through the stream.

rhdf5, h5vc

- ▶ Transform idiosyncratic files to intermediate form.
- ▶ Basis for spoke-like down-stream exploration

Critique

- ▶ User *R* code is often very inefficient. 100 – 1000-fold gains in *R*; 1 – 10-fold gain in C.
- ▶ Non-multicore parallel programming is very challenging to support – heterogenous user environments; lack of shared state.
- ▶ Shared memory implies memory management in concert with parallelization – iteration, restriction, sampling.
- ▶ Clunky *R*-level nested parallelism.
- ▶ High-*throughput* computing may be good enough.
- ▶ Interactive debugging and error recovery would be great!
- ▶ The illusion of performance would be a great goal for interactive exploratory analysis
- ▶ SIMD appealing conceptually but less consistent with interactive user expectations.

Prospects

Clouds & virtualization

- ▶ Controlled environment enables advanced configuration, e.g., our *StarCluster* AMI

Grammar of high throughput computing

- ▶ Verbs: what are they? *yield*, *restrict* (columns), *select* (rows), *query*; *map*, *reduce*; *tiles*, *aggregate*, *splice*
- ▶ Lazy exploratory evaluation

```
d %>% restrict() %>% select() %L%  
  (aggregate() %>% display()) %>%  
    aggregate() %>% data.frame
```

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