The R package {bigstatsr}: memory- and computation-efficient tools for big matrices stored on disk

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eRum 2018

About

I'm a PhD Student (2016-2019) in **Predictive Human Genetics** in Grenoble.

 $ext{Disease} \sim ext{DNA mutations} + \cdots$



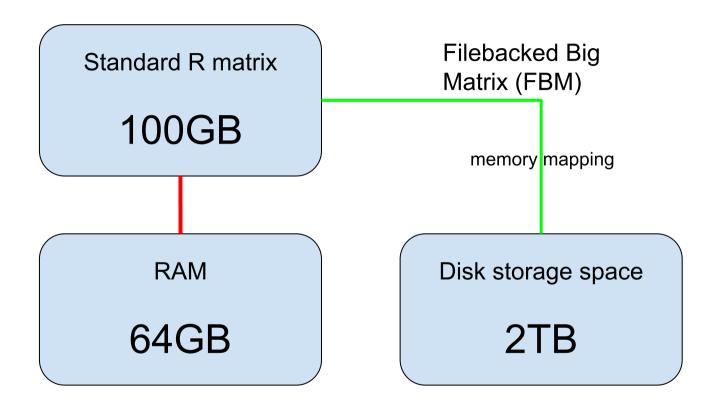
Very large genotype matrices

- previously: 15K x 280K, celiac disease (~30GB)
- currently: 500K x 500K, UK Biobank (~2TB)



But I still want to use R..

The solution I found



FBM is very similar to filebacked.big.matrix from package {bigmemory}.

Similar accessor as R matrices

```
X <- FBM(2, 5, init = 1:10, backingfile = "test")</pre>
X$backingfile
## [1] "/home/privef/Bureau/eRum-2018/test.bk"
X[, 1] ## ok
## [1] 1 2
X[1, ] ## bad
## [1] 1 3 5 7 9
X[] ## super bad
## [,1] [,2] [,3] [,4] [,5]
## [1,] 1 3 5 7 9
## [2,] 2 4 6 8 10
```

Similar accessor as R matrices

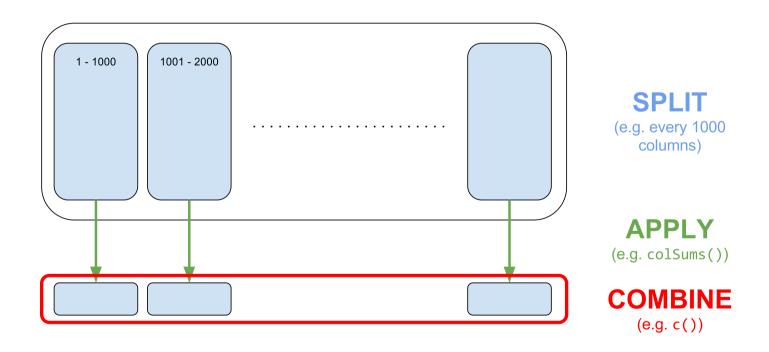
```
colSums(X[]) ## super bad
```

[1] 3 7 11 15 19



Split-(par)Apply-Combine Strategy

Apply standard R functions to big matrices (in parallel)



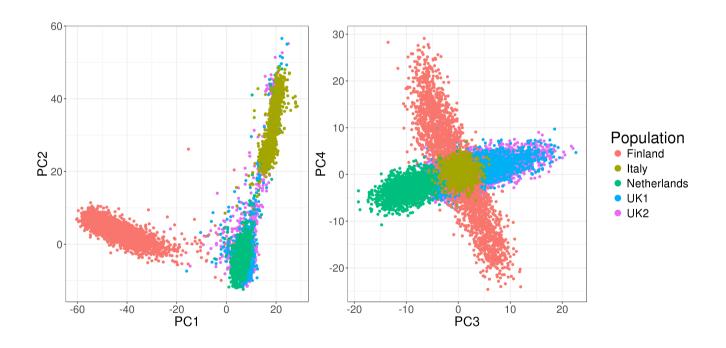
Implemented in big_apply().

Similar accessor as Rcpp matrices

```
// [[Rcpp::depends(BH, bigstatsr)]]
#include <bigstatsr/BMAcc.h>
// [[Rcpp::export]]
NumericVector big_colsums(Environment BM) {
  XPtr<FBM> xpBM = BM["address"];
  BMAcc<double> macc(xpBM);
  size_t n = macc.nrow();
  size_t m = macc.ncol();
  NumericVector res(m);
  for (size_t j = 0; j < m; j++)</pre>
    for (size_t i = 0; i < n; i++)</pre>
      res[j] += macc(i, j);
  return res;
```

Partial Singular Value Decomposition

 $15K \times 100K - 10$ first PCs - 6 cores - 1 min (vs 2h in base R)

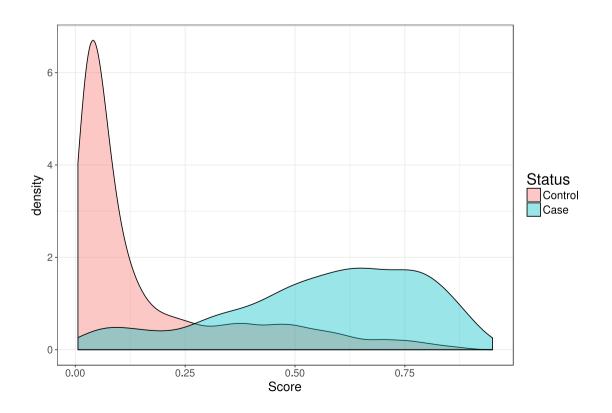


Implemented in big_randomSVD(), powered by R packages {RSpectra} and {Rcpp}.

Sparse linear models

Predicting complex diseases with a penalized logistic regression

 $15K \times 280K - 6$ cores - 2 min



Other functions

- matrix operations
- association of each variable with an output
- plotting functions
- read from text files
- many other functions...

Parallel

- most of the functions are parallelized (memory-mapping makes it easy!)
- you can parallelize you own functions with big_parallelize()

I'm now able
to run algorithms
on 100GB of data
on my computer

R Packages

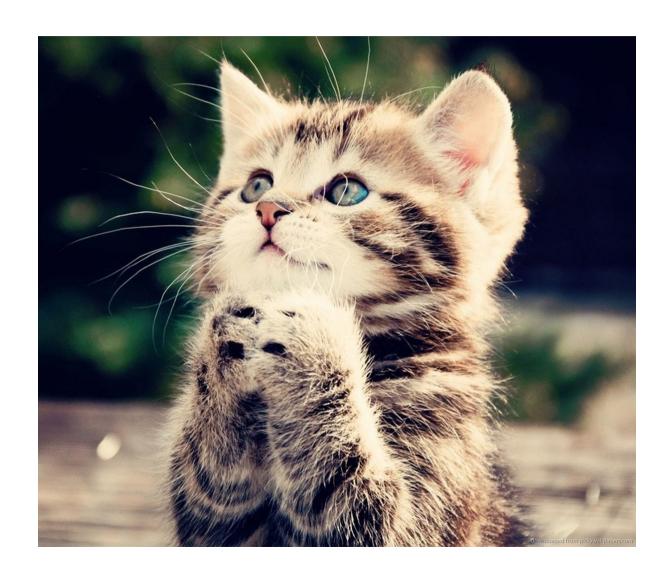
Efficient analysis of large-scale genome-wide data with two R packages: bigstatsr and bigsnpr 3

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Bioinformatics, bty185, https://doi.org/10.1093/bioinformatics/bty185

- {bigstatsr}: to be used by any field of research
- {bigsnpr}: algorithms specific to my field of research

Contributors are welcomed!



Thanks!

Presentation: https://privefl.github.io/eRum-2018/slides.html

Package's website: https://privefl.github.io/bigstatsr/

DOI: 10.1093/bioinformatics/bty185



Slides created via the R package xaringan.