

# Large-scale data analysis in R



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# Getting started

## Initial setup:

- Wifi
- Power outlets
- RCC cluster access
- Yubikeys (`rcc-guest-xxxx`)
- Helper(s)
- Handout / exercises
- Pace, questions (e.g., keyboard shortcuts)

# Getting started

1. Download workshop packet.

```
$ cd ~  
$ wget https://tinyurl.com/yad63z9c \   
    -O temp.tar.gz  
$ tar zxvf temp.tar.gz  
$ cd R-large-scale  
$ cd code
```

2. Add these three lines to your `.bashrc` file.

```
# SLURM.  
export SACT_FORMAT="jobid,partition,user,account%12,allocpus,node%12,elapsed,totalcpu,maxRSS,ReqM"  
export SQUEUE_FORMAT="%13i %12j %10P %10u %12a %8T %9r %10l %.11L %5D %4C %8m %N"
```

# Getting started

## 3. Start up three midway2 sessions.

```
midway2
File Edit Options Buffers Tools Imenu-S ESS Help
# You will find that you need more than 2 GB of memory (this is the
# default). Try requesting 14 GB instead:
#
#   sinteractive --partition=broadwl --time=2:00:00 --mem=14G
#
# SET UP ENVIRONMENT
# -----
# These are some functions that we will use here and in other parts of
# the workshop.
source("functions.R")

# For running this code with Rscript, I always load the methods
# package just to be safe.
library(methods)

# Set the sequence of pseudorandom numbers (useful for making sure
# that the randomized PCA results are reproducible).
set.seed(1)

# LOAD REGMAP DATA
# -----
# NOTES:
#
# - Introduce RegMap A. thaliana data set. (Good opportunity for
#   interaction.)
#
# - If you don't request enough memory, the process will be killed
#   after running the load command.
#
cat("Loading RegMap data.\n")
load("../data/regmap.RData")

# INSPECT GENOTYPE DATA
# -----
# A few useful commands for inspecting the data we just loaded:
#
--:--F1  pca.regmap.R  Top L1  Git-master  (ESS[S] [none] Rox) -----
Loading vc-git...done
```

```
Shell
$ ls
)
Desktop          git              pubsw
ExtremeDeconvolution_1.3.tar.gz  intel            rcc
MASTClassic_0.933.tar.gz        matlab          redundans-dev.tar.gz
R                             matlab_simple.m redundans.tar.gz
R-large-scale-packet          matlab_simple.sbatch regmap
R-large-scale-packet.tar.gz    midway2_batch.sh rstudio-server
R_libs                       notes.txt       rstudio-test-jb
SIMLR_1.4.0.tar.gz           output         run_jupyter_notebook.sh
anaconda3                 pardiso        scratch-midway
azure-cli.tar.gz           pardiso.lic    test1.ipynb
bin                       partis.txt     test2.ipynb
data.tar.gz              peer_1.0.tar.gz thindrives
downloads             privatemodules trash
emacs                  public_html
$
```

```
Shell
1  [||] 5.8% 15 [|||||100.0%] 29 [ 0.0%] 43 [ 0.0%]
2  [||] 5.1% 16 [|||||76.2%] 30 [ 0.0%] 44 [|||||52.8%]
3  [||] 3.8% 17 [|||||100.0%] 31 [ 0.6%] 45 [|||||95.7%]
4  [||] 4.3% 18 [ 0.0%] 32 [||] 3.1% 46 [|||||100.0%]
5  [||] 9.4% 19 [|||||44.4%] 33 [ 0.0%] 47 [||] 1.2%]
6  [||] 3.1% 20 [|||||93.8%] 34 [ 0.0%] 48 [ 0.0%]
7  [||] 2.5% 21 [|||||73.3%] 35 [ 0.6%] 49 [|||||61.1%]
8  [||||] 31.1% 22 [||] 4.4% 36 [||] 6.9% 50 [|||||75.2%]
9  [||] 5.0% 23 [||] 3.1% 37 [ 0.0%] 51 [|||||68.1%]
10 [||] 5.1% 24 [|||||43.1%] 38 [||||] 13.9% 52 [||] 3.1%]
11 [||] 4.5% 25 [|||||33.5%] 39 [ 0.0%] 53 [|||||84.0%]
12 [||] 3.1% 26 [|||||35.6%] 40 [||] 3.1% 54 [|||||65.2%]
13 [||||] 10.6% 27 [|||||100.0%] 41 [ 0.0%] 55 [||] 0.6%]
14 [||] 1.3% 28 [|||||50.6%] 42 [||||| 31.4%] 56 [||] 1.9%]
Mem [|||||171G/251G] Tasks: 1885; 14 running
Swp [ 0K/0K] Load average: 19.41 18.81 17.92
Uptime: 54 days, 19:58:26
```

# Getting started

## 4. Request a compute node.

```
$ screen -S rcc_workshop
$ sinteractive --partition=broadwl \
  --time=2:00:00 --account=rcc-guest \
  --reservation=rworkshop2
$ echo $HOSTNAME
```

## 5. Start up R programming environment.

```
$ module load R/3.4.1
$ R
R> getwd( )
```

# Aims of workshop

- ~~Do an analysis of a large-scale data set in R.~~
- Develop some useful skills for large-scale data analysis in R *within a high-performance computing environment*, and apply these skills to a medium-scale data set.

# Outline

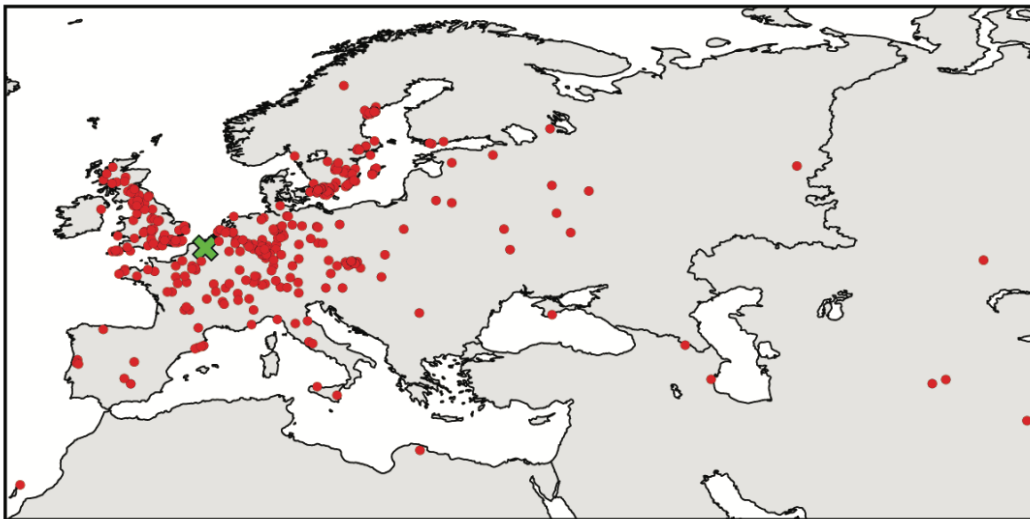
- Initial setup.
- Brief introduction.
- Part 1: Warm-up with PCA of RegMap data.
- Part 2: Implementing multithreaded computation in R for analysis of genetic adaptation to climate.
- Brief recap.

# Part 1: PCA of the RegMap data

## Adaptation to Climate Across the *Arabidopsis thaliana* Genome

Angela M. Hancock,<sup>1</sup> Benjamin Brachi,<sup>2</sup> Nathalie Faure,<sup>2</sup> Matthew W. Horton,<sup>1</sup>  
Lucien B. Jarymowycz,<sup>1</sup> F. Gianluca Sperone,<sup>1</sup> Chris Toomajian,<sup>3</sup> Fabrice Roux,<sup>2</sup> Joy Bergelson<sup>1\*</sup>

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1. Inspect RegMap data.
2. Run PCA interactively and assess time & memory needs.
3. Examine PCA results.
4. Run PCA using Rscript.
5. Run PCA using SLURM job engine, and inspect output.



# ***Exercise: Run PCA using SLURM job engine, and inspect output.***

```
sbatch pca.sbatch
```

- Useful commands while job is running:

```
squeue --user=<cneitd> | less -S  
ssh midway2-xxxx
```

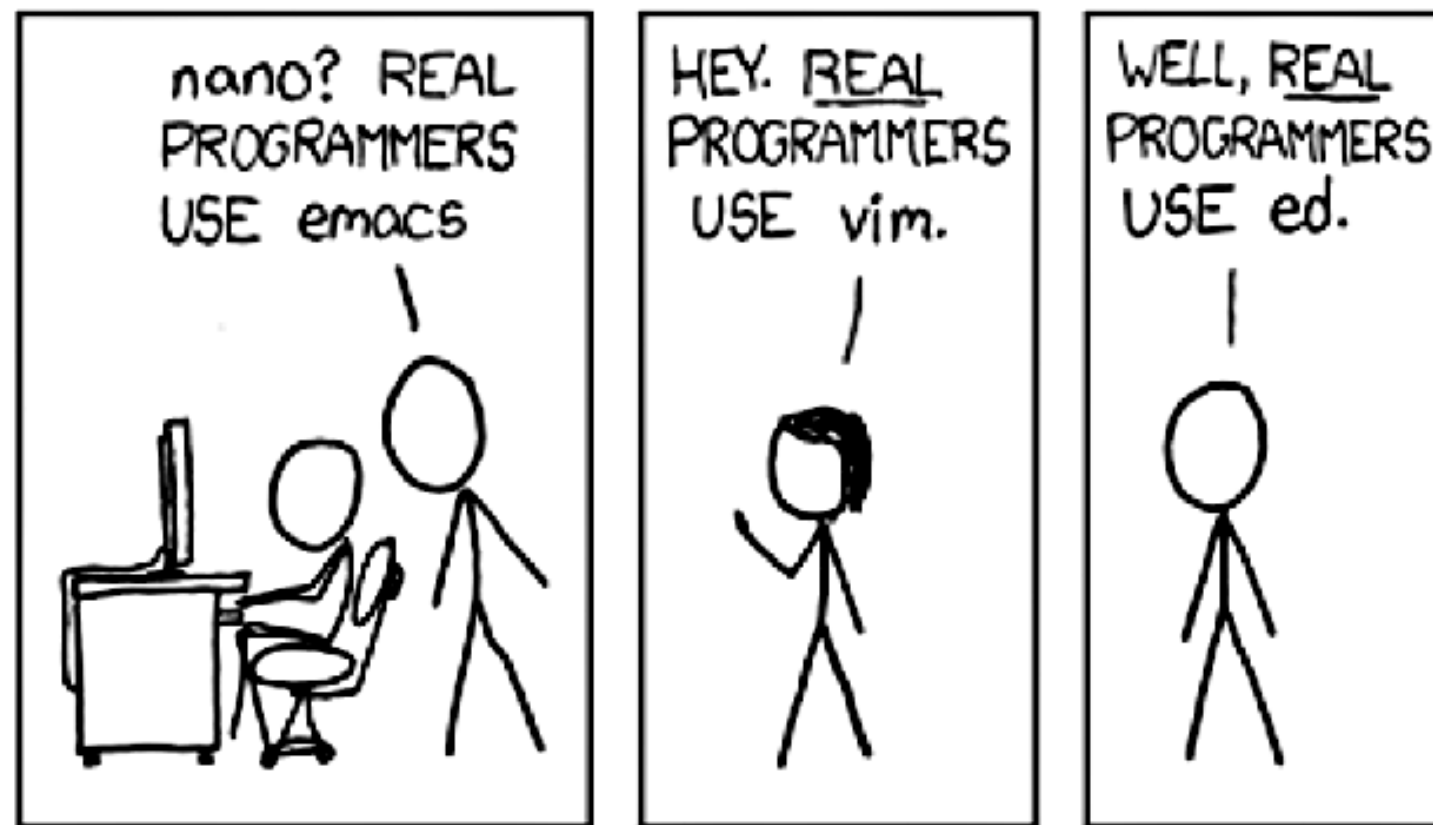
- Useful commands after job has completed:

```
htop --user=<cnetid>  
sacct --user=<cnetid> --units=G | less -S  
less ../output/pca_err.txt  
less ../output/pca_out.txt
```

# Part 2: Genetic analysis of climate variables in RegMap data

1. Inspect climate (“phenotype”) data.
2. Run analysis interactively and assess time & memory needs.
3. Experiment with multithreaded matrix operations to improve computation time.
4. Experiment with parallel computation of the weights using `mclapply`.
5. Automate `climate.R` analysis using command-line arguments.
6. Run analysis using SLURM job engine, and inspect output.
7. Use SLURM job engine to automate analysis of all 48 climate variables.

There is no best tool—use  
*whatever works for you.*



# Some general advice

1. `help(package = cool_package)`.
2. Use **midway2**, not `midway1`.
3. Email [help@rcc.uchicago.edu](mailto:help@rcc.uchicago.edu) R help on the RCC cluster.
4. Learn to avoid loops as much as possible;  
e.g., use `apply()`, `lapply()`, `tapply()`,  
`do.call()`.
5. Document your setup—start with  
`sessionInfo()`.

# After the workshop

- Feedback.
- Suggestions for future R workshop topics (e.g., Rcpp).