

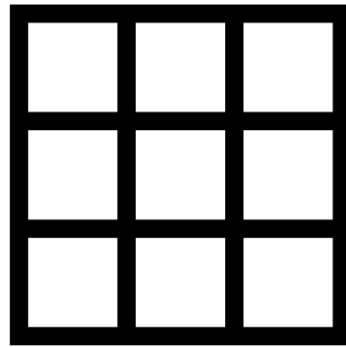
# Hands-on 1 discussion

# Supercomputing with R

# Supercomputing structure

1. Create a **parameter grid** (data frame)
2. Create an **analysis function** that takes in a row of the parameter grid and outputs a result
3. Create a **self-contained job script** to load a chunk of the grid and run the analysis function on several rows in parallel
4. Create a **shell script** to run the R job with SLURM parameters

**grid**



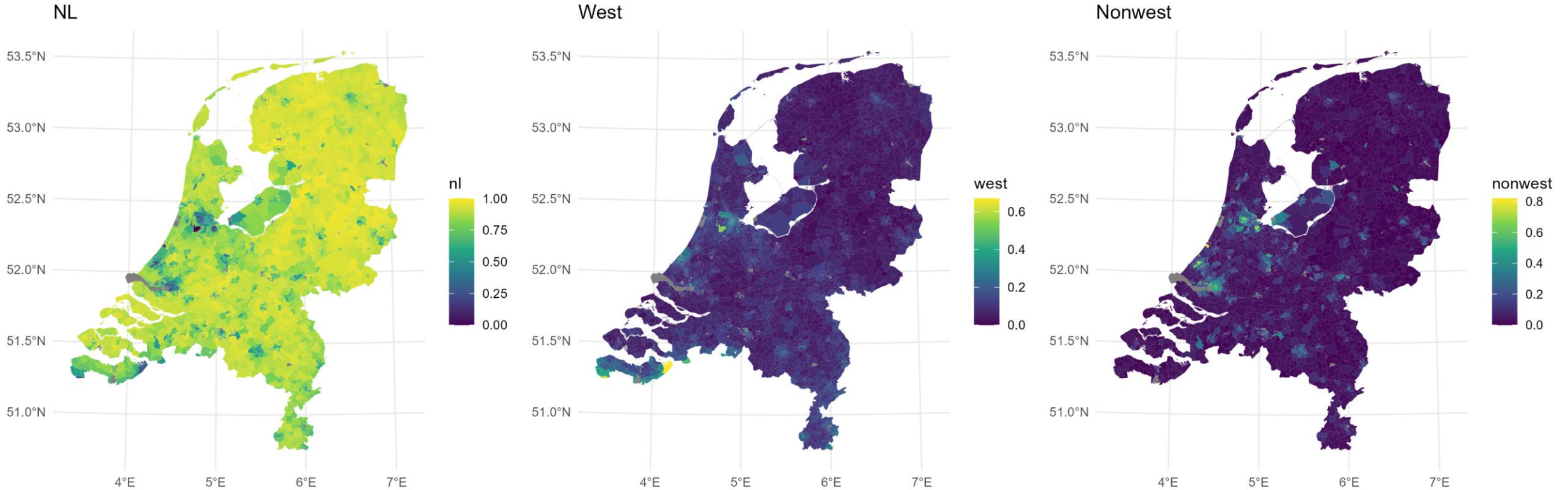
# ABM for the Netherlands

- We have a good ABM implementation now.
- Let's connect our ABM to real data
- Base our **proportion parameter** on population data about neighbourhoods in NL

(fake, illustrative) research question:

**What proportion of non-western migrants is “happy” with different levels of neighbourhood preference  $B_a$ ?**

# ABM for the Netherlands



<https://www.pdok.nl/introductie/-/article/cbs-wijken-en-buurten>

# ABM for the Netherlands

Simple feature collection with 3248 features and 5 fields

Geometry type: MULTIPOLYGON

Dimension: XY

Bounding box: xmin: 13565.4 ymin: 306846.2 xmax: 278026.1 ymax: 619374.9

Projected CRS: Amersfoort / RD New

First 10 features:

	wijkcode	wijknaam	n1	west	nonwest	geom
1	WK001400	Centrum	0.69	0.19	0.12	MULTIPOLYGON (((233335.8 58 ...
2	WK001401	Oud-Zuid	0.75	0.16	0.09	MULTIPOLYGON (((235128 5811 ...
3	WK001402	Oud-West	0.74	0.16	0.10	MULTIPOLYGON (((233335.8 58 ...
4	WK001403	Oud-Noord	0.68	0.14	0.18	MULTIPOLYGON (((234047.5 58 ...
5	WK001404	Oosterparkwijk	0.71	0.13	0.16	MULTIPOLYGON (((234689.3 58 ...
6	WK001405	Zuidoost	0.78	0.14	0.08	MULTIPOLYGON (((239416.9 57 ...
7	WK001406	Helpman e.o.	0.77	0.13	0.10	MULTIPOLYGON (((235885.3 57 ...
8	WK001407	Zuidwest	0.80	0.10	0.10	MULTIPOLYGON (((233581 5794 ...
9	WK001408	Hoogkerk e.o.	0.85	0.09	0.06	MULTIPOLYGON (((231577.2 58 ...
10	WK001409	Nieuw-West	0.68	0.14	0.18	MULTIPOLYGON (((230032.1 58 ...

# Parameter grid

- There are **3248** neighbourhoods in NL
- We will inspect **91** different levels of Ba parameter
- For stability, we want **50** iterations to average over

**$3248 * 91 * 50 = 14\,778\,400$  ABMs to run!**



# Tibbles and nested columns

- In the hands-on, you will go through the grid code
- This is just 1 version / implementation
- There are other ways to create the grid (probably faster, too)
  
- End result: one row per desired result

# Tibbles and nested columns

- Useful function:  
`expand_grid()`

```
expand_grid(  
  theta = c(1, 2, 3),  
  phi = c("A", "B", "C"),  
  rho = c(0.1, 0.15, 0.20)  
)  
  
# A tibble: 27 x 3  
  theta phi    rho  
  <dbl> <chr> <dbl>  
1     1 A     0.1  
2     1 A     0.15  
3     1 A     0.2  
4     1 B     0.1  
5     1 B     0.15  
6     1 B     0.2  
7     1 C     0.1  
8     1 C     0.15  
9     1 C     0.2  
10    2 A     0.1  
# ... with 17 more rows
```

# Tibbles and nested columns

- We will use tibbles with nested columns
- We `unnest_longer()` those nested columns to different rows:

```
# A tibble: 3,248 x 6
  row    nl west nonwest iter      Ba
  <int> <dbl> <dbl> <dbl> <list> <list>
1     1  0.69  0.19  0.12 <int [50]> <dbl [91]>
2     2  0.75  0.16  0.09 <int [50]> <dbl [91]>
3     3  0.74  0.16  0.1  <int [50]> <dbl [91]>
4     4  0.68  0.14  0.18 <int [50]> <dbl [91]>
5     5  0.71  0.13  0.16 <int [50]> <dbl [91]>
6     6  0.78  0.14  0.08 <int [50]> <dbl [91]>
7     7  0.77  0.13  0.1  <int [50]> <dbl [91]>
8     8  0.8   0.1   0.1  <int [50]> <dbl [91]>
9     9  0.85  0.09  0.06 <int [50]> <dbl [91]>
10    10  0.68  0.14  0.18 <int [50]> <dbl [91]>
# ... with 3,238 more rows
```



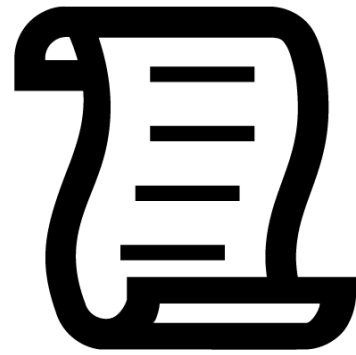
```
# A tibble: 14,778,400 x 6
  row    nl west nonwest iter      Ba
  <int> <dbl> <dbl> <dbl> <int> <dbl>
1     1  0.69  0.19  0.12     1  0.05
2     1  0.69  0.19  0.12     2  0.05
3     1  0.69  0.19  0.12     3  0.05
4     1  0.69  0.19  0.12     4  0.05
5     1  0.69  0.19  0.12     5  0.05
6     1  0.69  0.19  0.12     6  0.05
7     1  0.69  0.19  0.12     7  0.05
8     1  0.69  0.19  0.12     8  0.05
9     1  0.69  0.19  0.12     9  0.05
10    1  0.69  0.19  0.12    10  0.05
# ... with 14,778,390 more rows
```

**More of this in the hands-on later**

# Supercomputing structure

1. Create a **parameter grid** (data frame)
2. Create an **analysis function** that takes in a row of the parameter grid and outputs a result
3. Create a **self-contained job script** to load a chunk of the grid and run the analysis function on several rows in parallel
4. Create a **shell script** to run the R job with SLURM parameters

**script**



# The analysis function

What it does

**Input** a row from the grid

**Output** our quantity of interest  
(proportion of happy nonwestern migrants)

Should be **robust**, i.e., deal with problematic inputs gracefully

- You should spend time testing this, you will literally run this code millions of times

# The analysis function

```
analysis_function ← function(row_idx) {  
  # Get the parameters belonging to this row  
  settings ← as.list(grid_tbl[row_idx,])  
  
  # compute the proportion of happy nonwestern migrants  
  # use trycatch to avoid crashing. This is important otherwise  
  # you will have a lot of problems with underused compute!  
  out ← tryCatch(  
    # this is the expression to evaluate  
    expr = {  
      prop_vec ← c(settings$nl, settings$west, settings$nonwest)  
      res ← abm_cpp(prop = prop_vec, Ba = settings$Ba)  
      return(res$h_prop[3])  
    },  
    # if there is an error, return NA as output!  
    error = function(e) return(NA)  
  )  
  
  return(out)  
}
```



# The R job script

What it does

**Input** a job number

**Output** a file with results from the ABMs and a log file

- Self-contained, runnable from the command-line
- Nice logging capabilities to show where things are going wrong if they do
- Within-node parallelization(!)

# Self-contained R scripts

- You can run R in non-interactive mode (if it's in your environment variables)

```
# getting the arguments from the commandline
args ← commandArgs(trailingOnly = TRUE)
num  ← as.numeric(args[1])

# return random numbers
rnorm(num)
```

```
> Rscript my_script.R 10
```

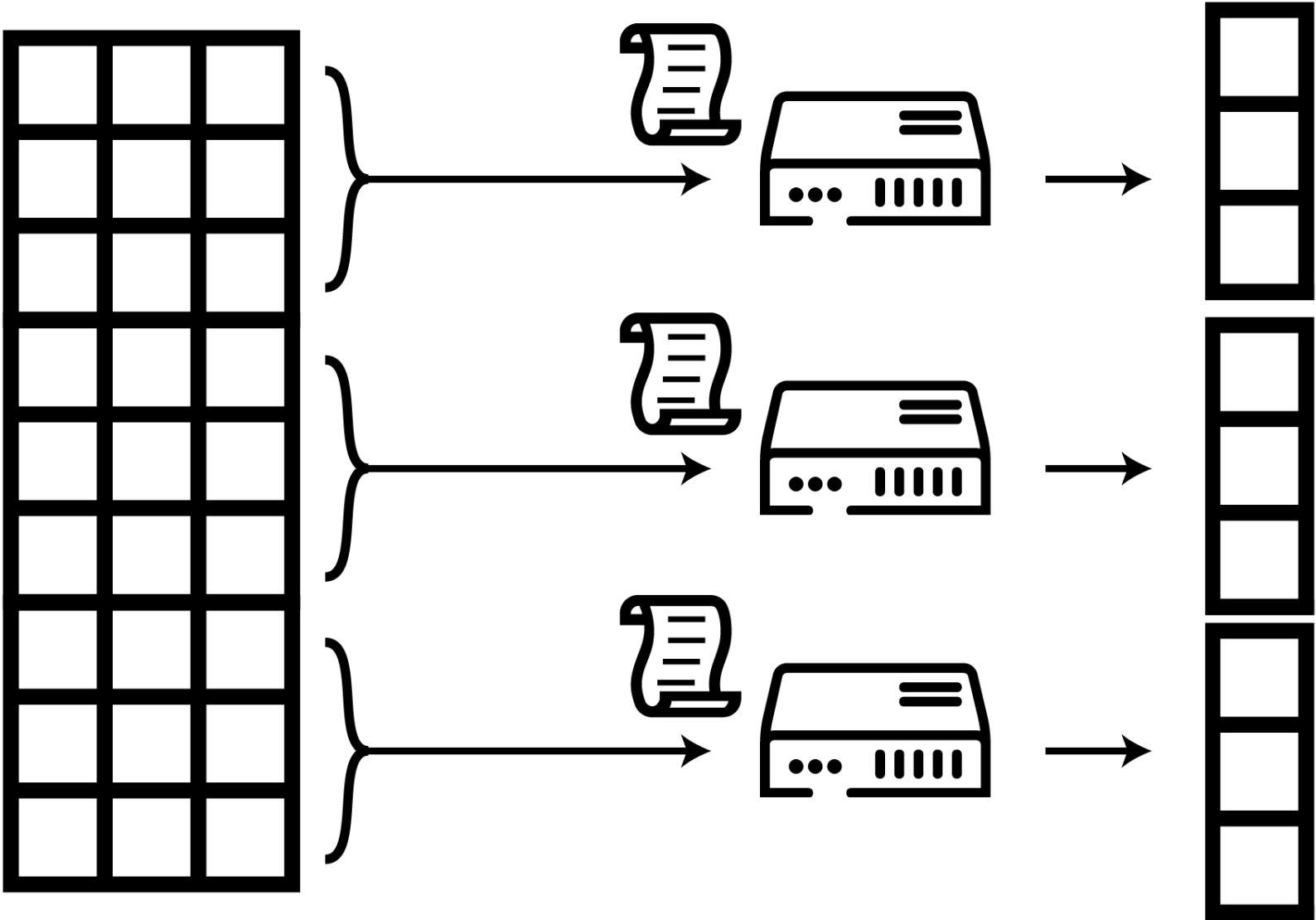
```
[1]  1.01272137 -2.10078427  0.58351622 -0.62444158 -1.22377068 -0.07592772
[7] -0.12156296  2.17437392 -0.18906297  1.78086798
```

# Logging

- There are several options available, e.g., the package **logging**
- For our case, we only need simple print (`cat()`) statements
- SLURM will store the R console output to a file
  
- Include statements in the script about which step is running
- Include timestamps / elapsed time

# Within-node parallelization

- In SLURM, you get (and pay for!) at least 16 cores at a time
- Therefore, your R scripts need within-node parallelization
- Compute results for multiple grid rows at a time
  
- “**Chunking**” your grid



# Within-node parallelization

- Get chunk
- Start cluster / child processes
- Compute chunk results on cluster
- Save output to file (`results_0001.rds`)

# Within-node parallelization

- How big should the chunk be?
- Depends on
  - speed (ABM runs/second/core)
  - number of cores on the node (16?)
  - how long you want each job to take
- How long? Make it manageable, e.g., 30 minutes
  - If something goes wrong (and something will go wrong!) you can rerun in reasonable amount of time
  - Balance manageability and overhead: data loading, Rcpp code compiling, results storing

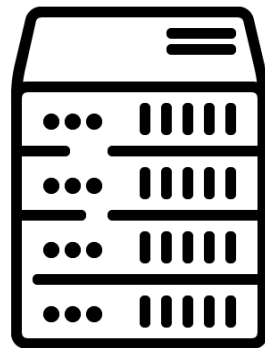
**More of this in the hands-on later**



# Supercomputing structure

1. Create a **parameter grid** (data frame)
2. Create an **analysis function** that takes in a row of the parameter grid and outputs a result
3. Create a **self-contained job script** to load a chunk of the grid and run the analysis function on several rows in parallel
4. Create a **shell script** to run the R job with SLURM parameters

**submit**



# Shell script

- We will use array jobs (as per SLURM terminology)

```
sbatch -a 1-77 my_script.sh
```

- Will queue 77 jobs
- Each job has a different environment variable  
**SLURM\_ARRAY\_TASK\_ID**
- Pass this environment variable to Rscript

# Shell script

```
#!/bin/bash
# Set job requirements
#SBATCH --nodes=1
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=16
#SBATCH --time=00:15:00
#SBATCH --partition=rome
#SBATCH --output=./logs/output.%a.out
#SBATCH --error=./logs/output.%a.err

# Loading modules
module load 2023
module load R-bundle-CRAN/2023.12-foss-2023a

# Run the script
Rscript "04_array_job.R" $SLURM_ARRAY_TASK_ID
```

```
lcur0520@login3:~/ossc_workshop$ sbatch -a 1-77 05_array_job.sh
```

```
Submitted batch job 9185382
```

```
lcur0520@login3:~/ossc_workshop$ squeue -u lcur0520
```

JOBID	PARTITION	NAME	USER	ST	TIME	NODES	NODELIST(REASON)
9185382_[1-77]	normal	05_array	lcur0520	PD	0:00	1	(Resources)

```
lcur0520@login3:~/ossc_workshop$ squeue -u lcur0520
```

JOBID	PARTITION	NAME	USER	ST	TIME	NODES	NODELIST(REASON)
9185382_[4-77]	normal	05_array	lcur0520	PD	0:00	1	(Resources)
9185382_1	normal	05_array	lcur0520	R	0:19	1	r27n17
9185382_2	normal	05_array	lcur0520	R	0:19	1	r25n17
9185382_3	normal	05_array	lcur0520	R	0:19	1	r25n27

```
lcur0520@login3:~/ossc_workshop$ scancel 9185382
```

```
lcur0520@login3:~/ossc_workshop$ squeue -u lcur0520
```

JOBID	PARTITION	NAME	USER	ST	TIME	NODES	NODELIST(REASON)
9185382_6	normal	05_array	lcur0520	CG	0:07	1	r27n6
9185382_7	normal	05_array	lcur0520	CG	0:07	1	r27n7
9185382_5	normal	05_array	lcur0520	CG	0:23	1	r26n7
9185382_4	normal	05_array	lcur0520	CG	0:26	1	r27n31
9185382_1	normal	05_array	lcur0520	CG	0:58	1	r27n17
9185382_2	normal	05_array	lcur0520	CG	0:58	1	r25n17
9185382_3	normal	05_array	lcur0520	CG	0:58	1	r25n27

# Supercomputing structure

1. Create a **parameter grid** (data frame)
2. Create an **analysis function** that takes in a row of the parameter grid and outputs a result
3. Create a **self-contained job script** to load a chunk of the grid and run the analysis function on several rows in parallel
4. Create a **shell script** to run the R job with SLURM parameters

# Hands-on session 2