

# tess3r: a R package for estimating spatial population structure

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# Genotypic Data

## DNA Sequencing Technologies:

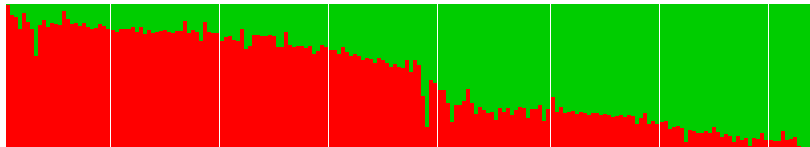
- ▶ SNPs array (*Arabidopsis thaliana* RegMap lines [Horton et al., 2012] : 200k loci of 1 307 individuals)
- ▶ next generation sequencing (1000 Genome project [Consortium et al., 2015] : whole genome of 2504 individuals)

	chr: 1 pos: 657	chr: 1 pos: 3102	chr: 1 pos: 4648
02B6	1	1	1
09A3	1	0	1
12A1	1	1	1
13B5	0	0	0

# Goal: Estimating Ancestral Population Structure

We assume that the genome of each individual come from  $K$  ancestral populations. We want to estimate:

- ▶ **ancestral genotype frequencies** for each locus
- ▶ **ancestral coefficients** for each individual



Several method exist:

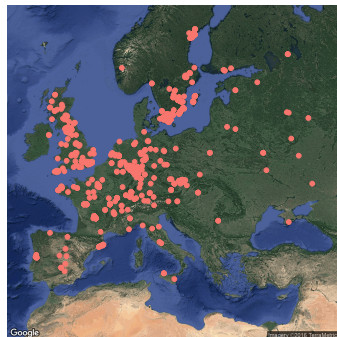
- ▶ **Bayesian**: structure [Pritchard et al., 2000]
- ▶ **optimisation based**: sNMF [Frichot et al., 2014]

These methods are crucial for demographic analysis, medical genetics, conservation genetics or landscape genetics.

# Spatial Data

Several methods incorporate geographic data in prior distributions:

- ▶ TESS 2.3 [Durand et al., 2009]
- ▶ BAPS [Corander et al., 2008]



We propose a new method to estimate spatial population based on an optimization problem.

# Model for the Genotypic Matrix

We write  $G$  the genomic matrix

$$P(G_{i,\ell} = j) = \sum_{k=1}^K Q_{i,k} f_{k,\ell}(j),$$

$$P = QF^T,$$

where  $Q$  is the ancestry coefficient matrix and  $F$  the ancestral genotype frequency matrix. There are such as:

$$Q \succeq 0, \sum_{k=1}^K Q_{i,k} = 1, \forall i \in \{1, \dots, n\}$$

$$F \succeq 0, \sum_{j=0}^d f_{k,\ell}(j) = 1, \forall \ell \in \{1, \dots, L\}.$$

# Optimisation Problem

We construct a weighted graph using spatial data:

$$W_{i,j} = e^{-\frac{\|z_i - z_j\|^2}{\sigma}},$$

where  $z$  are geographic positions.

The loss function is:

$$\begin{aligned} \text{Loss}(Q, F) &= \|X - QF^T\|^2 + \lambda \sum_{i,j}^n W_{i,j} \|Q_i - Q_j\|^2 \\ &= \|X - QF^T\|^2 + \lambda \text{trace}(Q^T \Lambda Q), \end{aligned}$$

where  $\Lambda$  is the graph Laplacian matrix and  $X$  is the data matrix.

# Block-coordinate Descent Scheme

- ▶ The optimisation problem is not convex.
- ▶ It is convex with respect to one of the variables  $Q$  or  $F$  when the other one is fixed.
- ▶ We can use a block-coordinate descent scheme:

```
for  $it \in 1, \dots, itMax$  do  
     $F \leftarrow \arg \min_F f_F(Q, F)$   
     $Q \leftarrow \arg \min_Q f_Q(Q, F)$   
end for
```

# Alternating Projected Least Squares

F optimisation step :

1. Solving least squares problems for  $F$  rows
2. Projecting onto the polygon of  $F$  constraints

Q optimisation step :

1. Solving  $L_2$ -regularized least squares problems for  $Q$  rows
2. Projecting onto the polygon of  $Q$  constraints



# Alternating Projected Least Squares

- ▶ Advantage:

The  $F$  optimisation step requires to solve  $L$  least squares problems of size  $K$ .

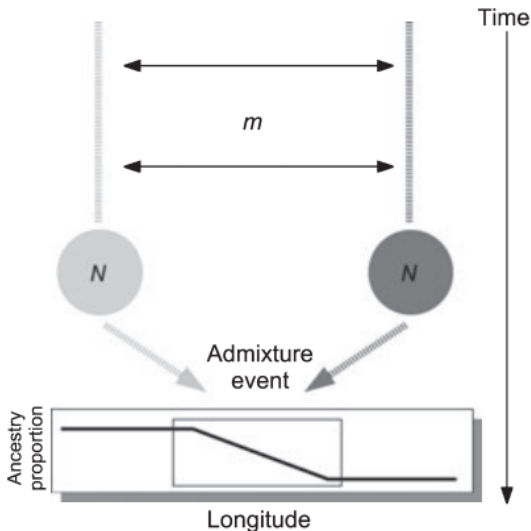
The  $Q$  optimisation step requires to solve  $n$   $L_2$ -regularized least squares problems of size  $K$ .

- ▶ Drawback:

The algorithm is not guaranteed to provide a stationary point.

# Simulations

Simulation of admixed population dataset [François and Durand, 2010]



# Comparison with TESS 2.3

On these simulation TESS3 algorithm performs 30 time better than TESS 2.3

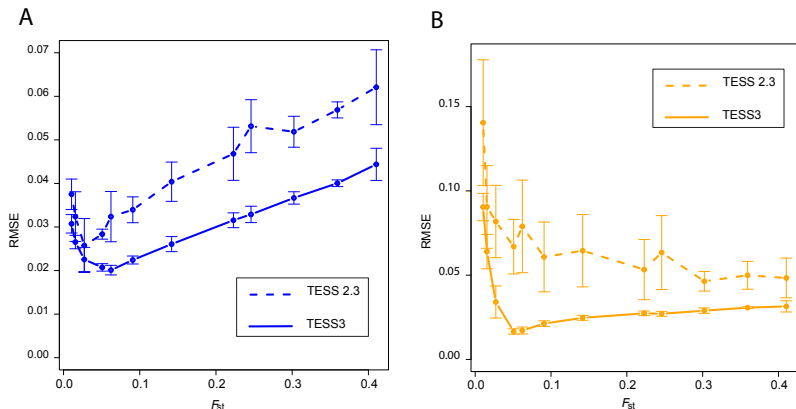


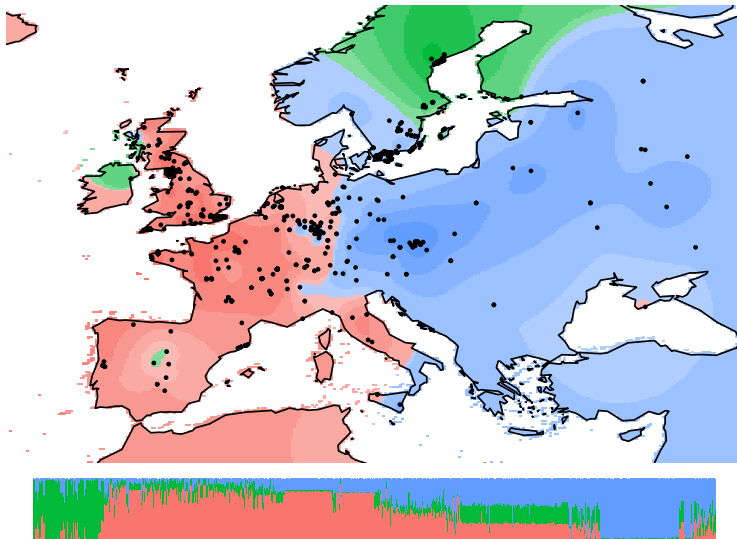
Figure 1: A) RMSEs of  $F$  estimates. B) RMSEs of  $Q$  estimates.

# *Arabidopsis thaliana* RegMap Lines Dataset



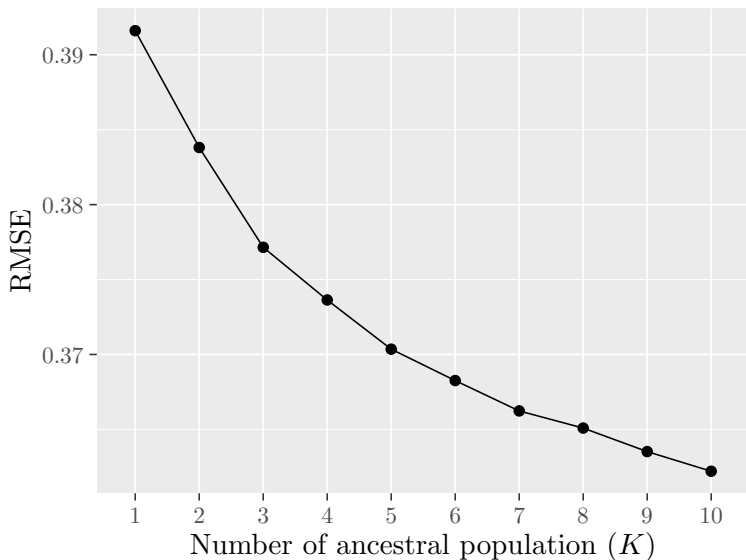
- ▶ 1,307 accessions of *A. thaliana* have been genotyped using the Affymetrix Arabidopsis 250k - SNP chip [Horton et al., 2012]
- ▶ Geographic coordinates for 1,193 of these samples [Anastasio et al., 2011]

# Ancestry Coefficients with $K = 3$ Ancestral Populations



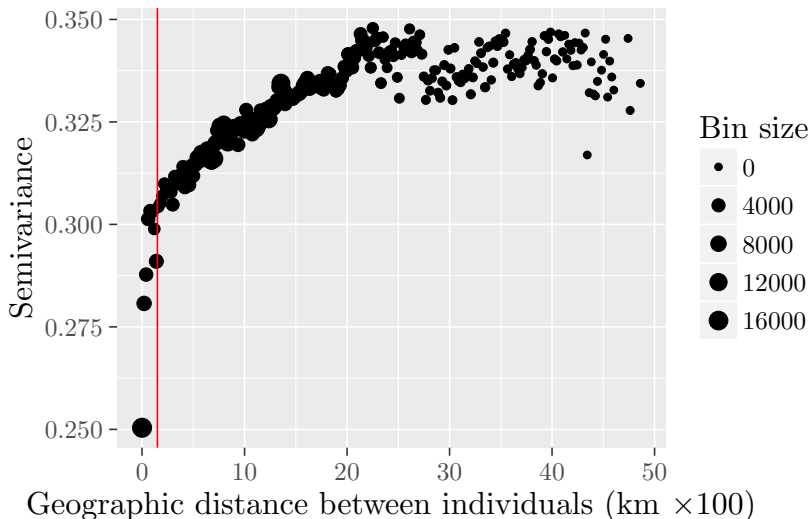
# Selection of the Number of Ancestral Populations

We choose  $K = 6$  ancestral populations.

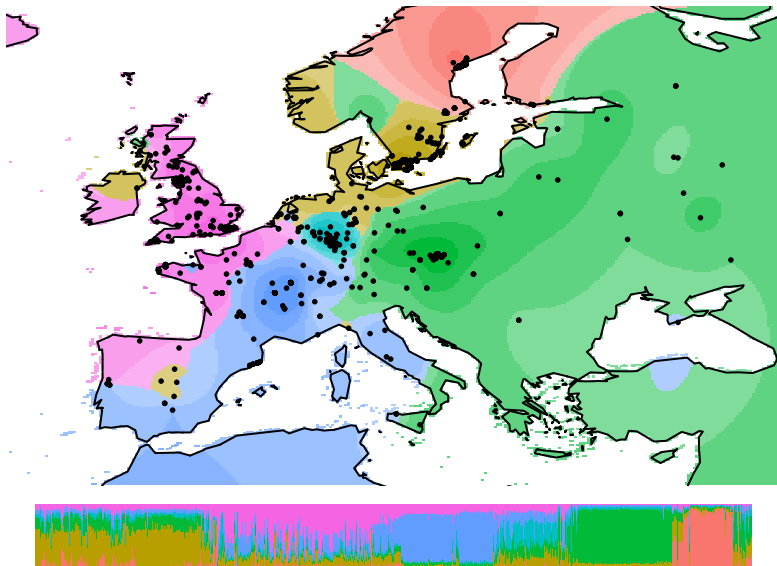


## Selection of the Spatial Autocorrelation Scale

We choose  $\sigma = 1.5$  for the spatial autocorrelation scale.

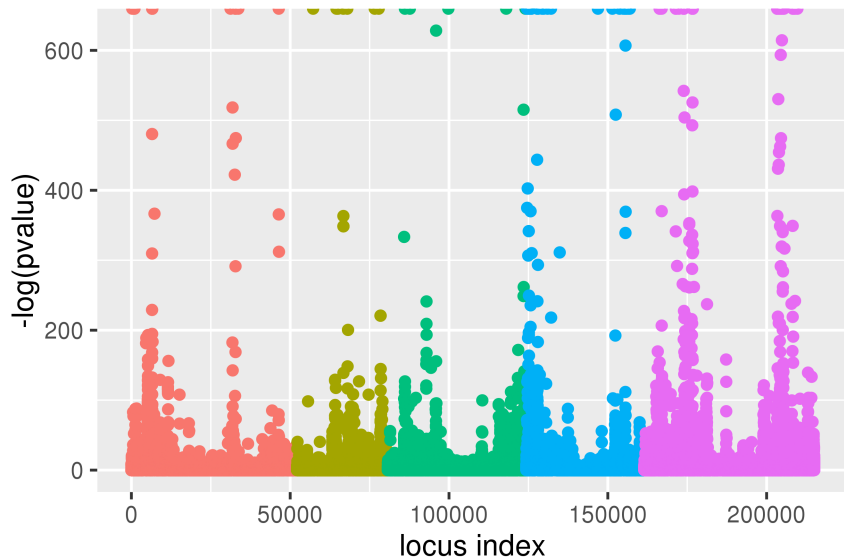


## Ancestry Coefficients for $K = 6$ and $\sigma = 1.5$





# Ancestral Frequency Manhattan Plot for $K = 6$ and $\sigma = 1.5$



## Detecting Adaptive Loci (Preliminary Results)

- ▶ Flowering-related genes detected in the top list: FRIGIDA, FLOWERING LOCUS C (FLC), DELAY OF GERMINATION 1 (DOG1) [Horton et al., 2012].
- ▶ Statistical overrepresentation of genes involved in metabolic processes: 1.08 fold enrichment,  $p\text{-value} < 1e^{-6}$  (PANTHER database).

# Conclusion

- ▶ Estimation and visualisation of spatial population structure.
- ▶ Local adaptation detection.
- ▶ Beta Version available on github.

```
devtools::install_github("cayek/TESS3_encho_sen@master")
```

# Acknowledgments

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- Alison E Anastasio, Alexander Platt, Matthew Horton, Erich Grotewold, Randy Scholl, Justin O Borevitz, Magnus Nordborg, and Joy Bergelson. Source verification of mis-identified arabidopsis thaliana accessions. *The Plant Journal*, 67(3): 554–566, 2011.
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