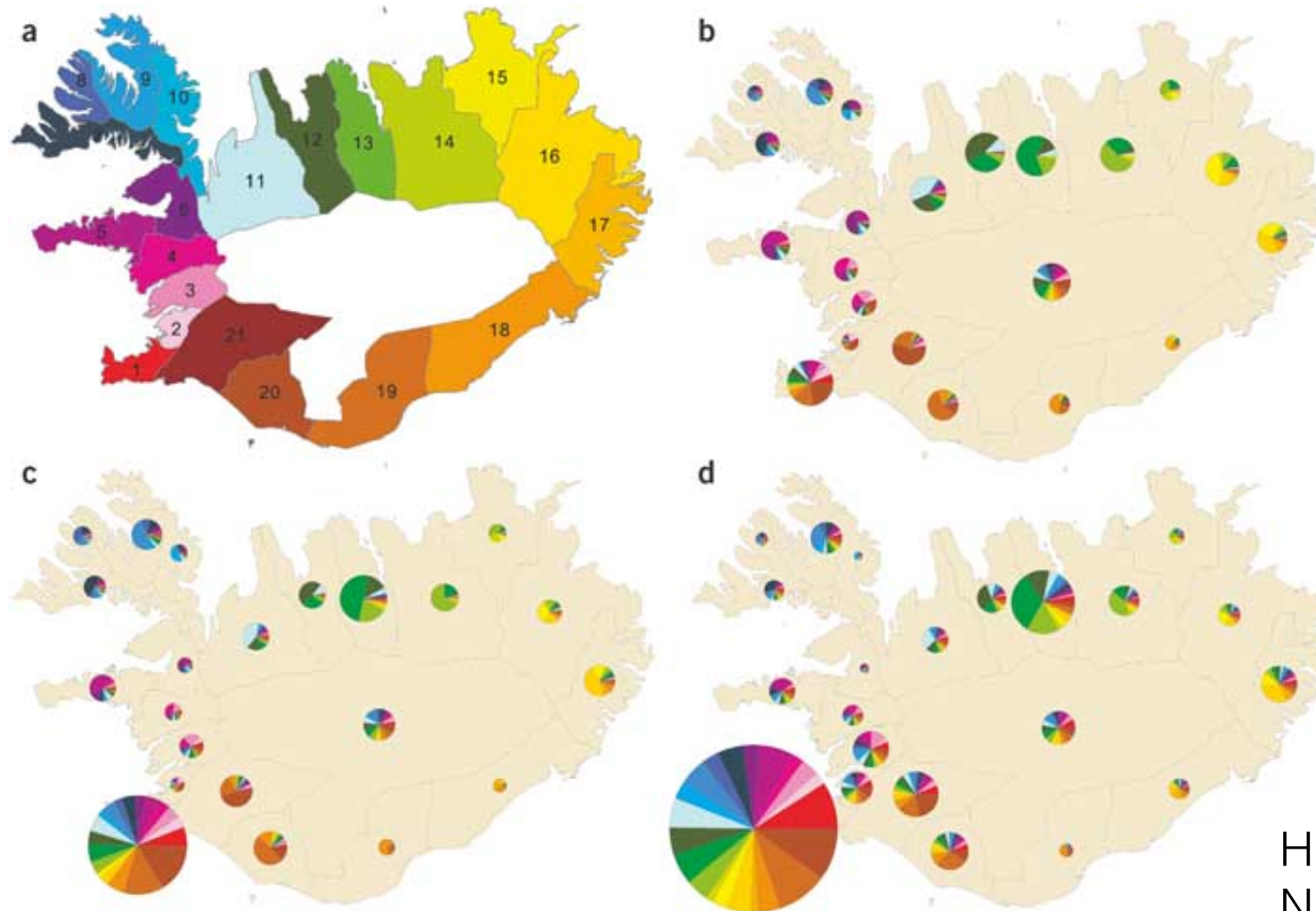


Cryptic Relatedness and fine scale population structure

Learning objectives

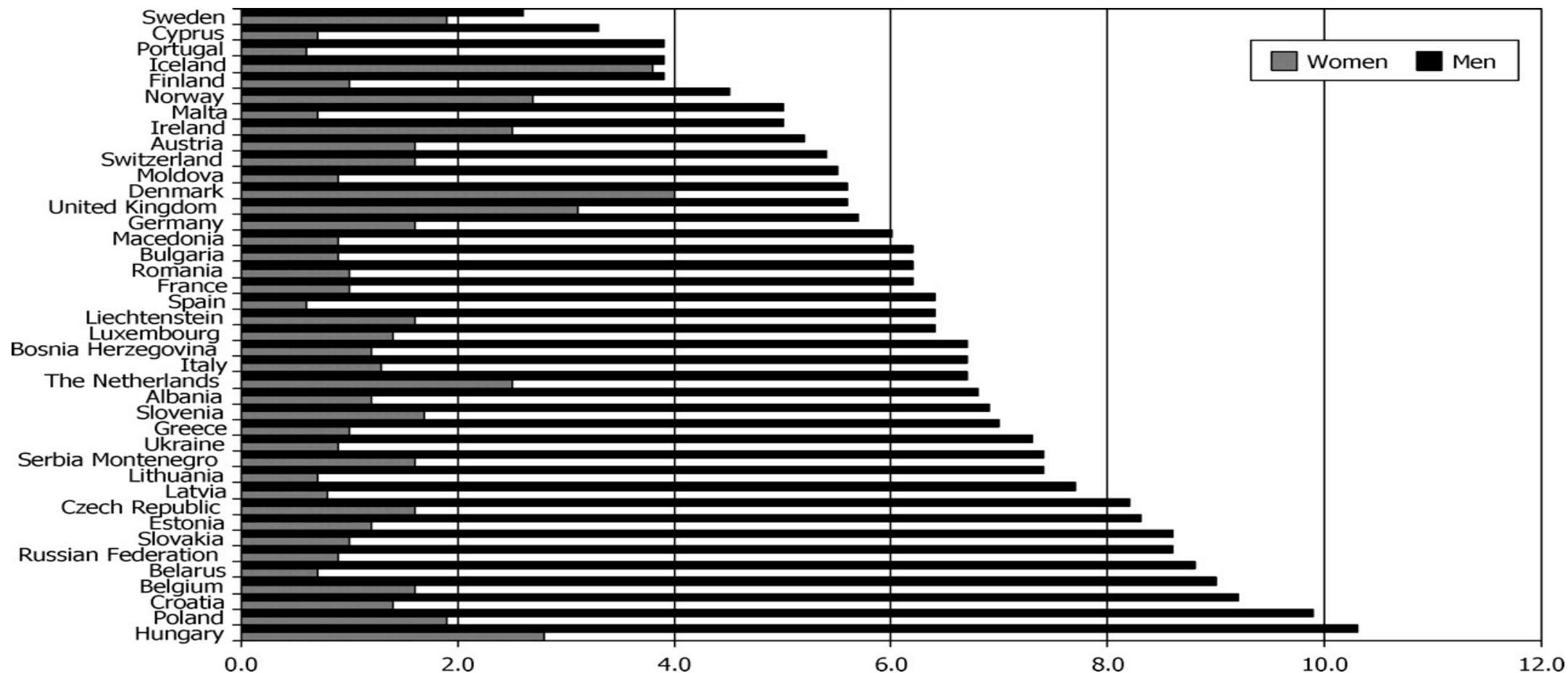
- Define fine scale population structure and cryptic relatedness
- How is it identified
 - Identity-by-descent
 - Rare variation
- Why it can be important for association analyses, especially of rare variants.

Cryptic Population Structure



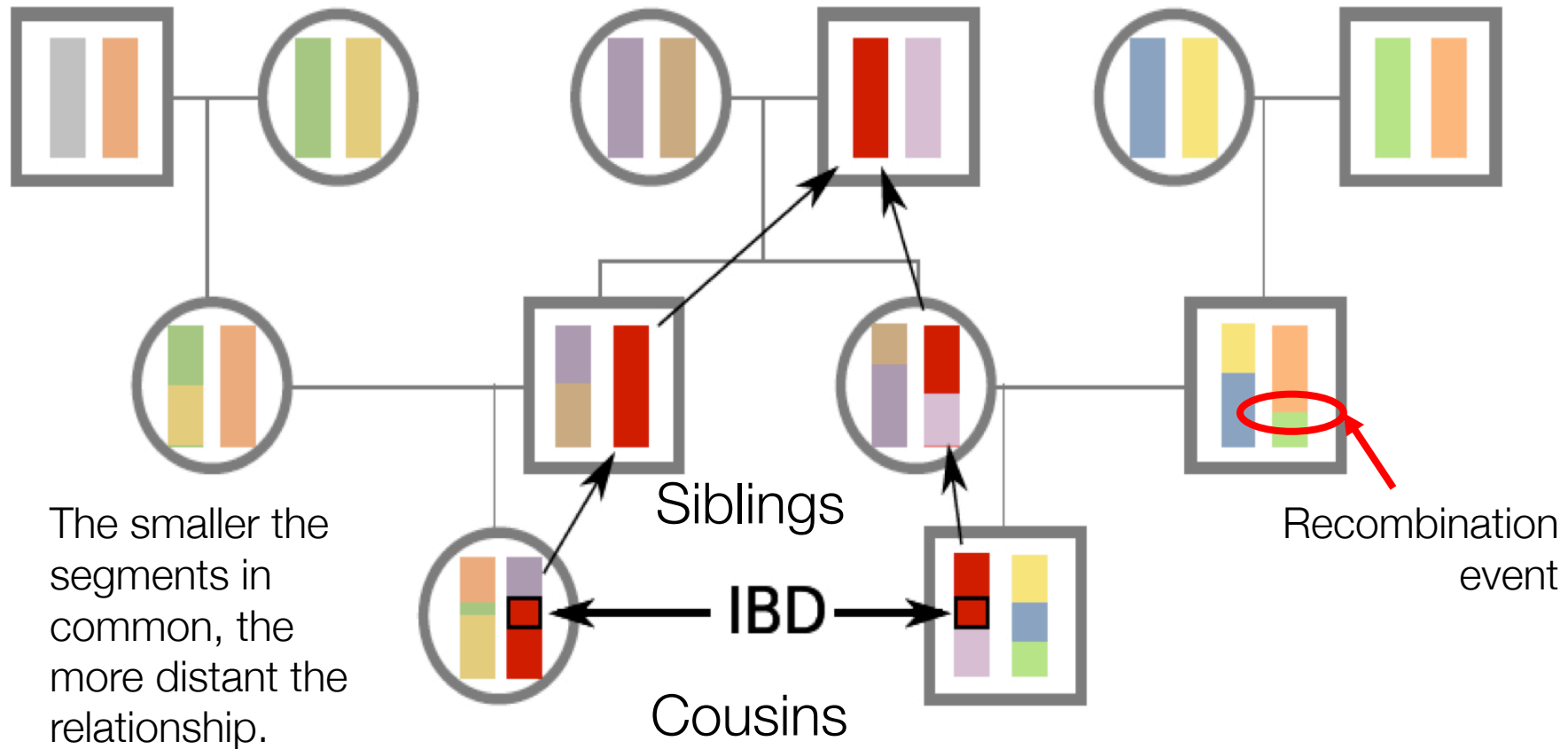
Helgason et al. (2004)
Nature Genet.

Lung Cancer Prevalence in Europe

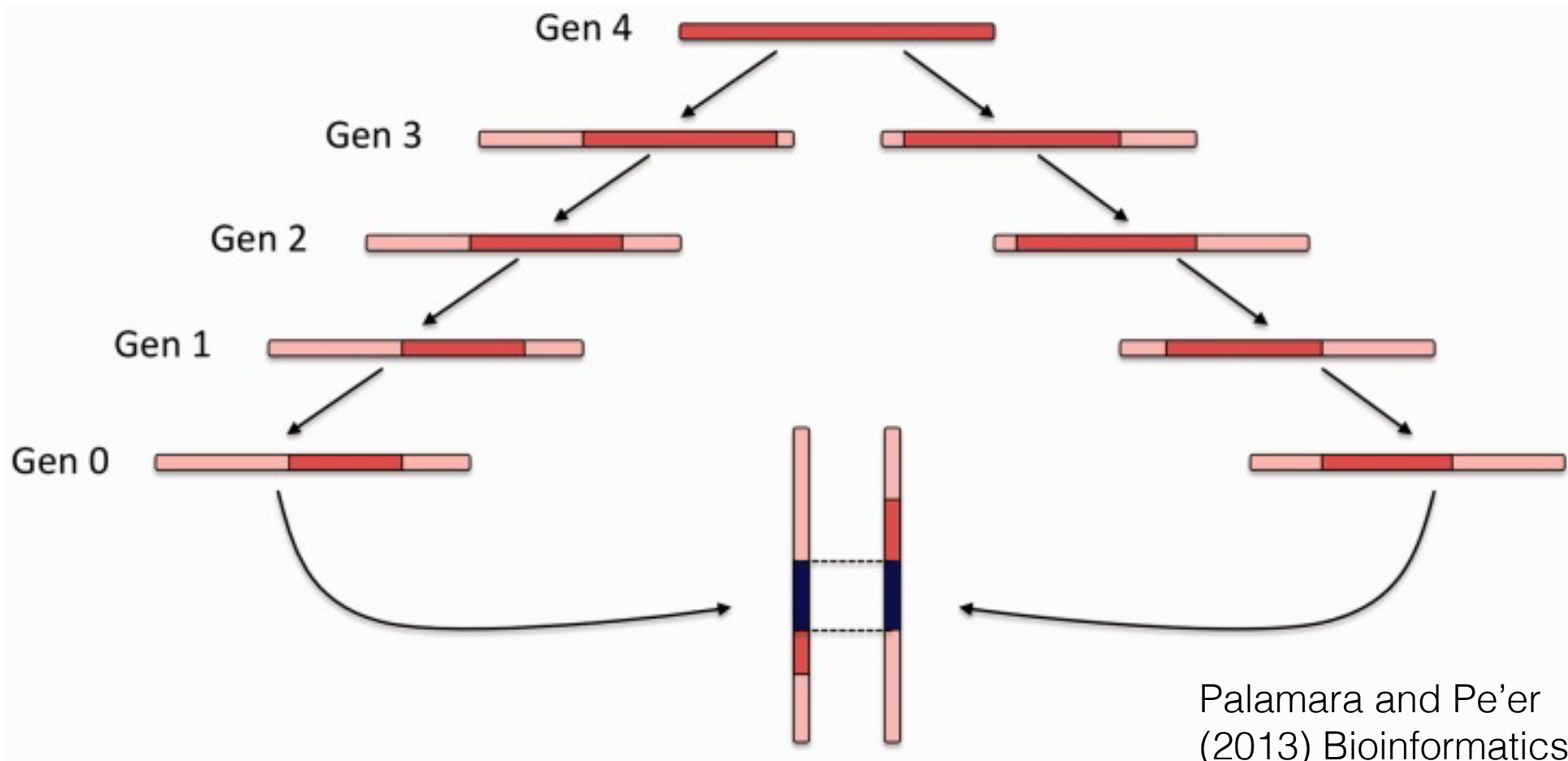


Boyle and Ferlay (2005) Annals of Oncology

Identity by Decent (IBD): A method to find both distant and recent relationships



IBD length is correlated with historical relationships.

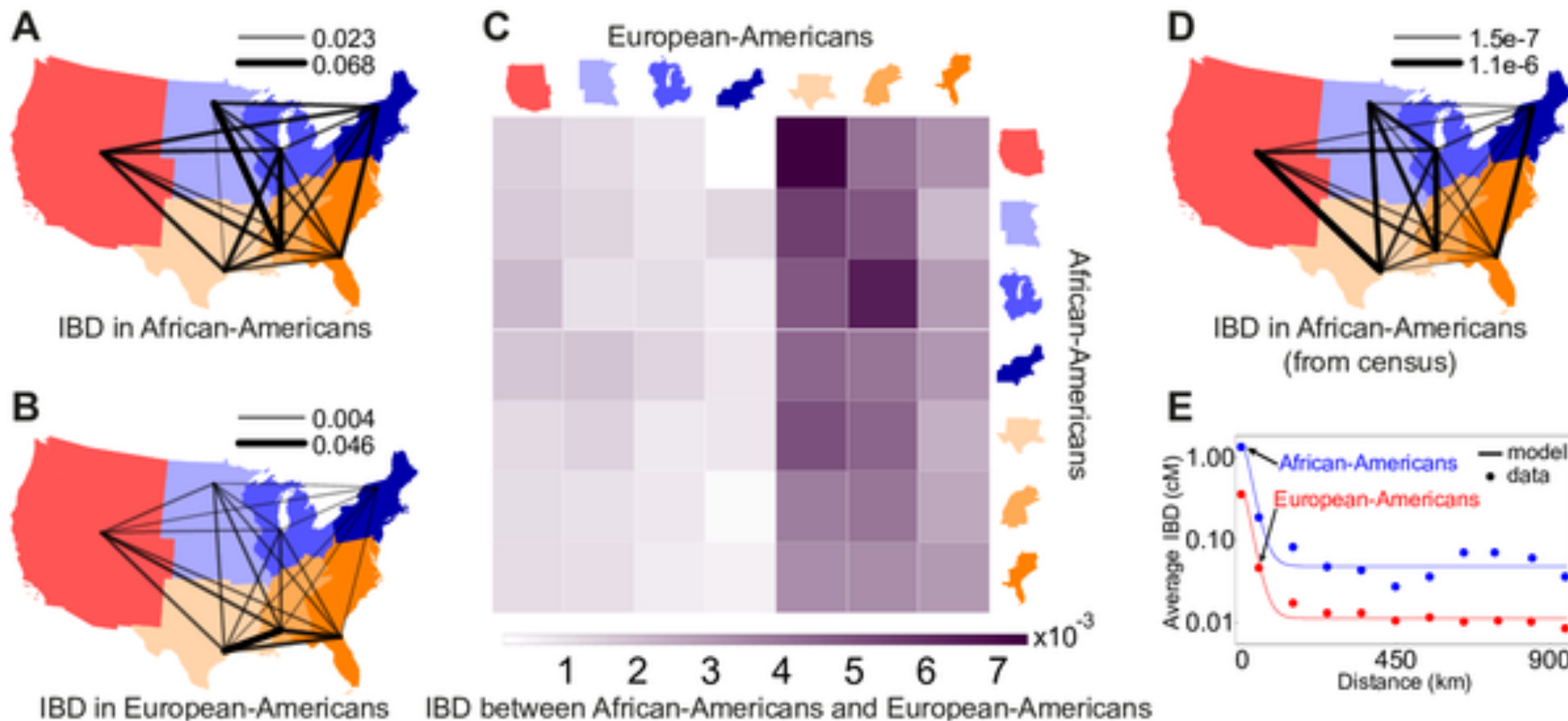


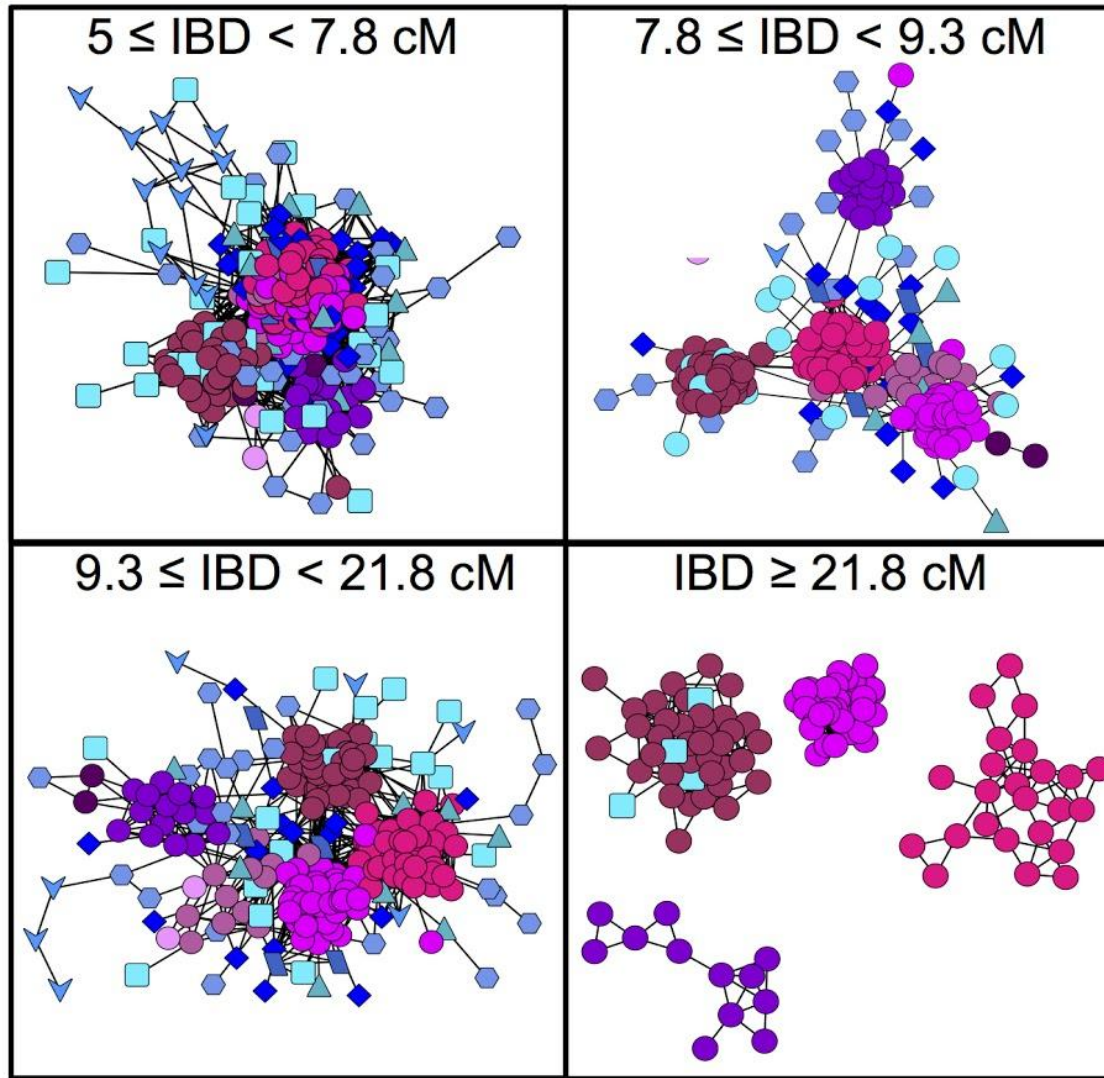
$$E[g|l] \cong \frac{3}{2 * l}$$

Baharian et al. (2016)
PLoS Genet.

Palamara and Pe'er
(2013) Bioinformatics

Pairwise genetic relatedness across



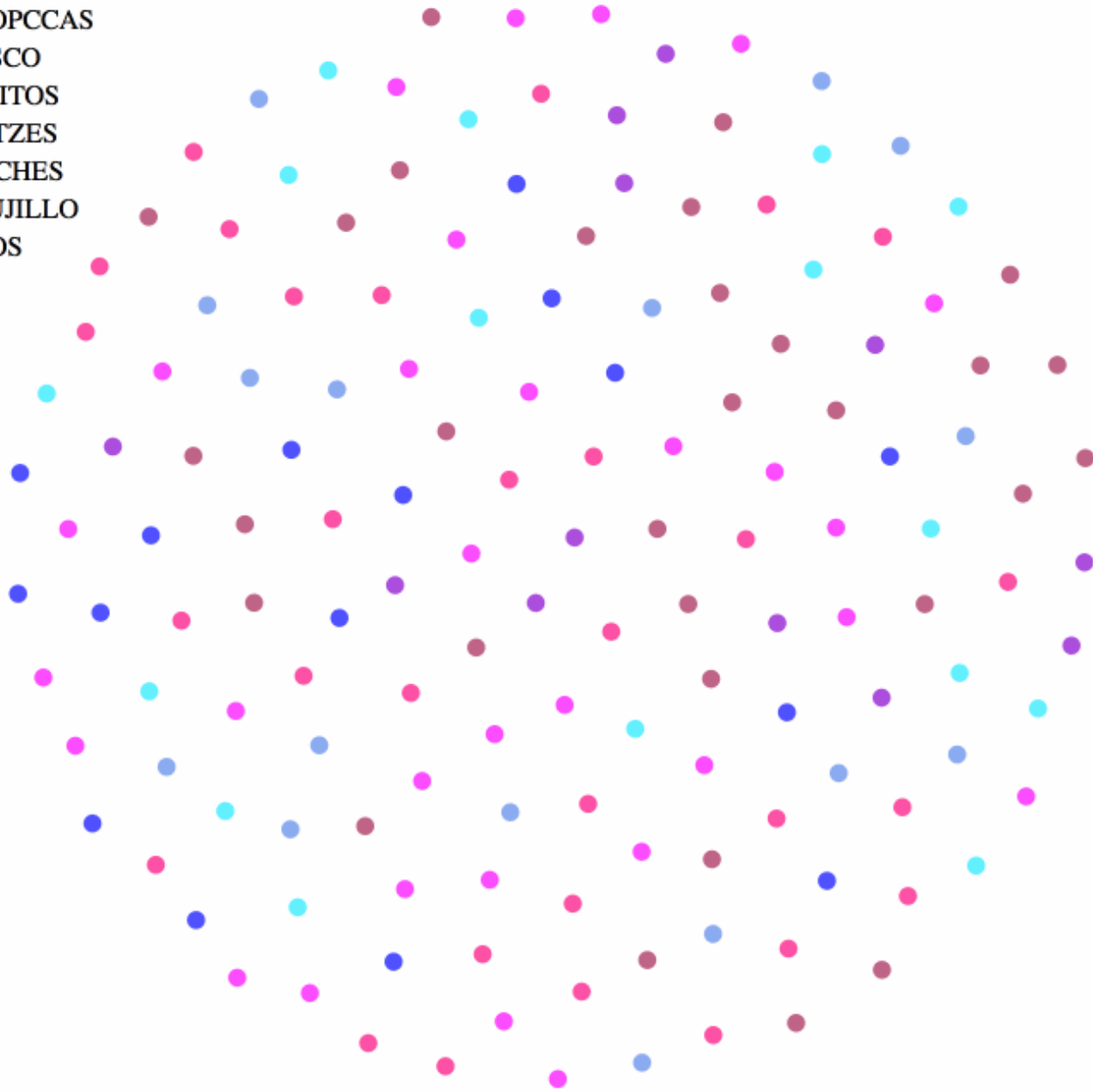
C

Legend for populations (represented by color and shape):

- Trujillo (light blue square)
- Lima (light blue triangle)
- Iquitos (light blue hexagon)
- AP (light blue inverted triangle)
- Puno (dark blue square)
- Cusco (dark blue diamond)
- Chopccas (pink circle)
- Moches (dark red circle)
- Matsig (dark purple circle)
- Nahua (light pink circle)
- Uros (magenta circle)
- Qeros (grey circle)
- Matzes (purple circle)

Identity-by-descent as a means to look at fine-scale structure over time

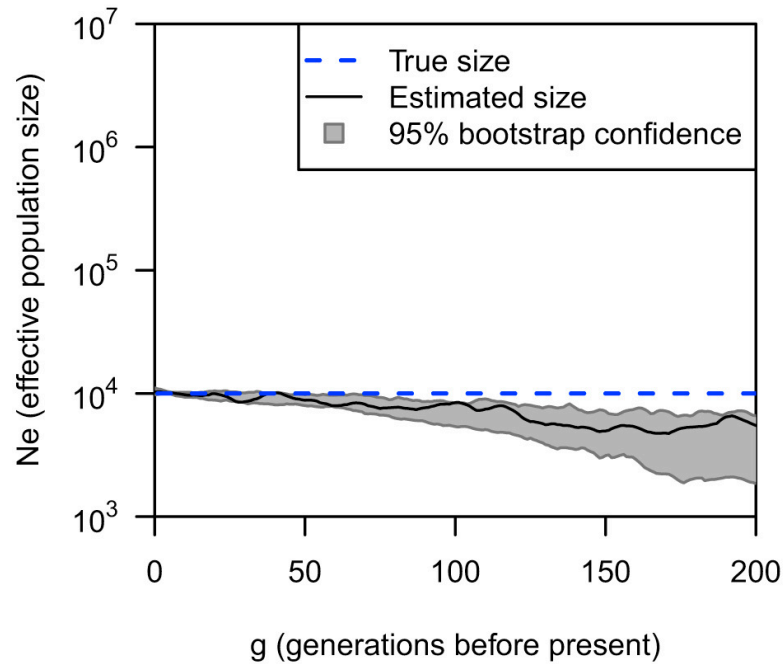
CHOPCCAS
CUSCO
IQUITOS
MATZES
MOCHES
TRUJILLO
UROS



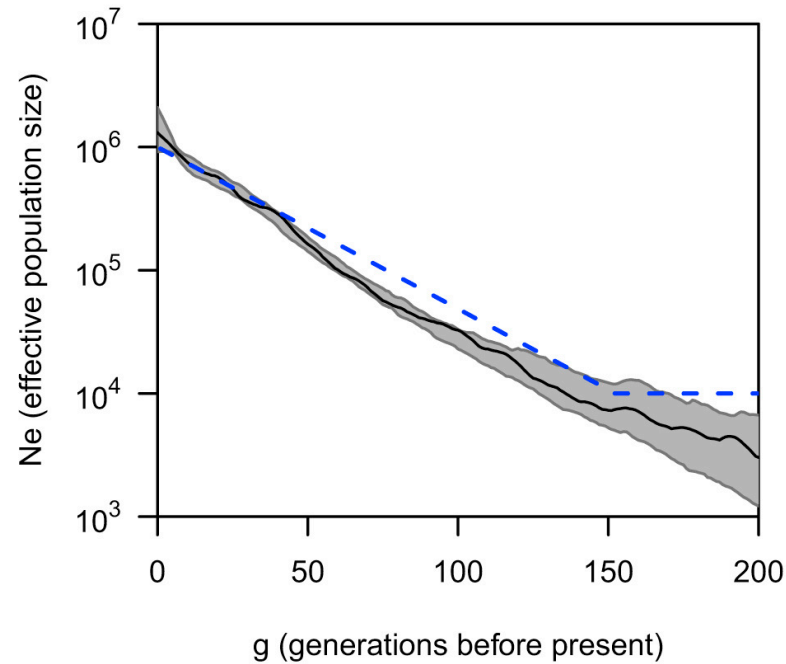
Identity-by-
descent as a
means to look at
fine-scale
structure over
time

IBD can estimate effective population size over time.

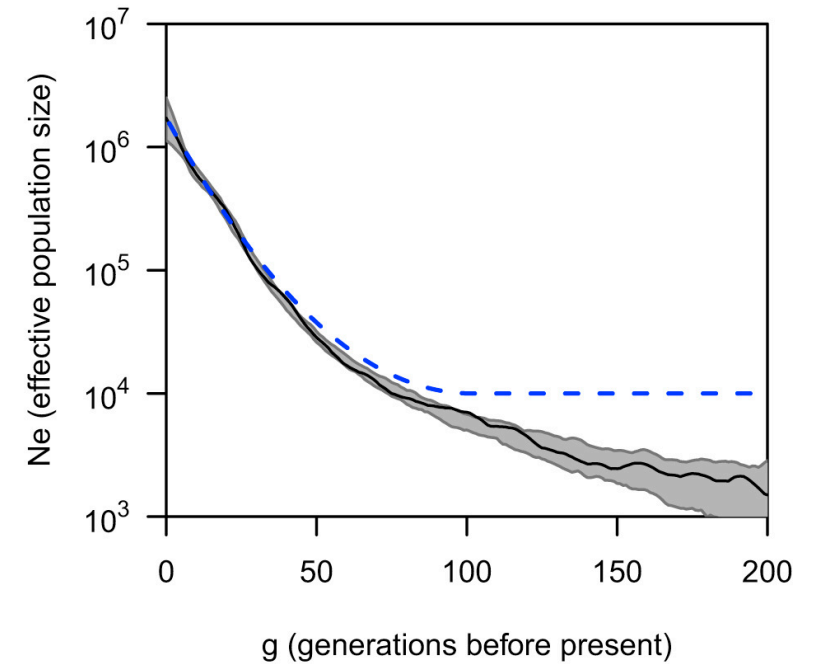
Constant size: SNP array data



Exponential growth: SNP array data



Super-exponential: SNP array data

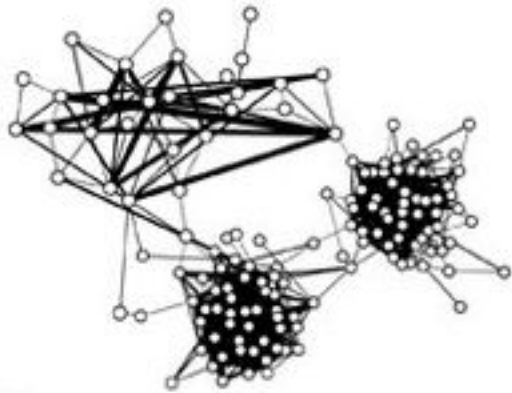


IBD on a large scale

a

Construct network from IBD.

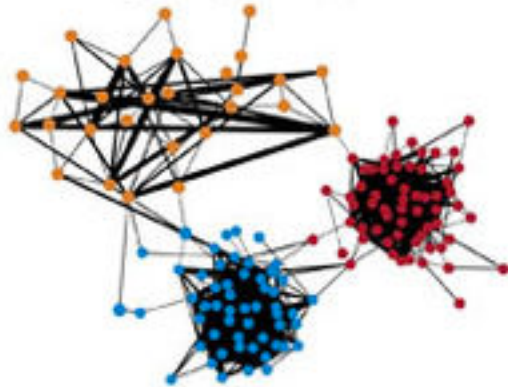
Join vertex pairs (genotyped samples) if $IBD > 12$ cM. Edge weights are a function of total detected IBD.



b

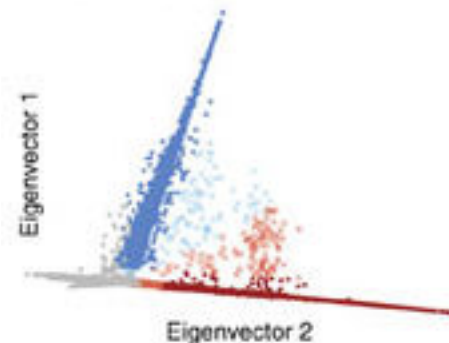
Detect network clusters.

Recursively identify disjoint sets that maximize the modularity of the network. (Here one level of clustering hierarchy is shown.)



c

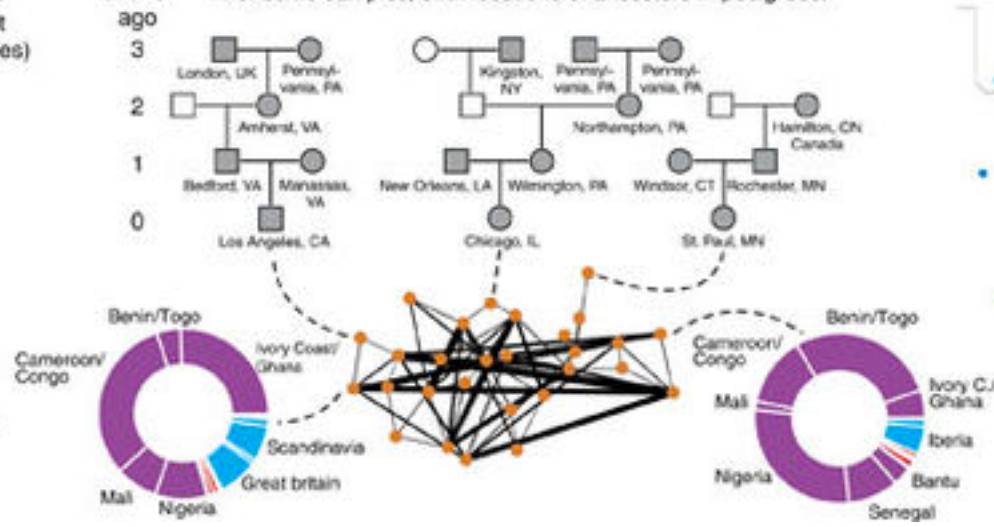
Identify subsets of the clusters that separate in the spectral embedding. Spectral embedding is computed from eigen-decomposition of Laplacian matrix. In the plot below, we identify "stable subsets" (filled circles) of the blue and red clusters.



d

Annotate each cluster with two kinds of data:

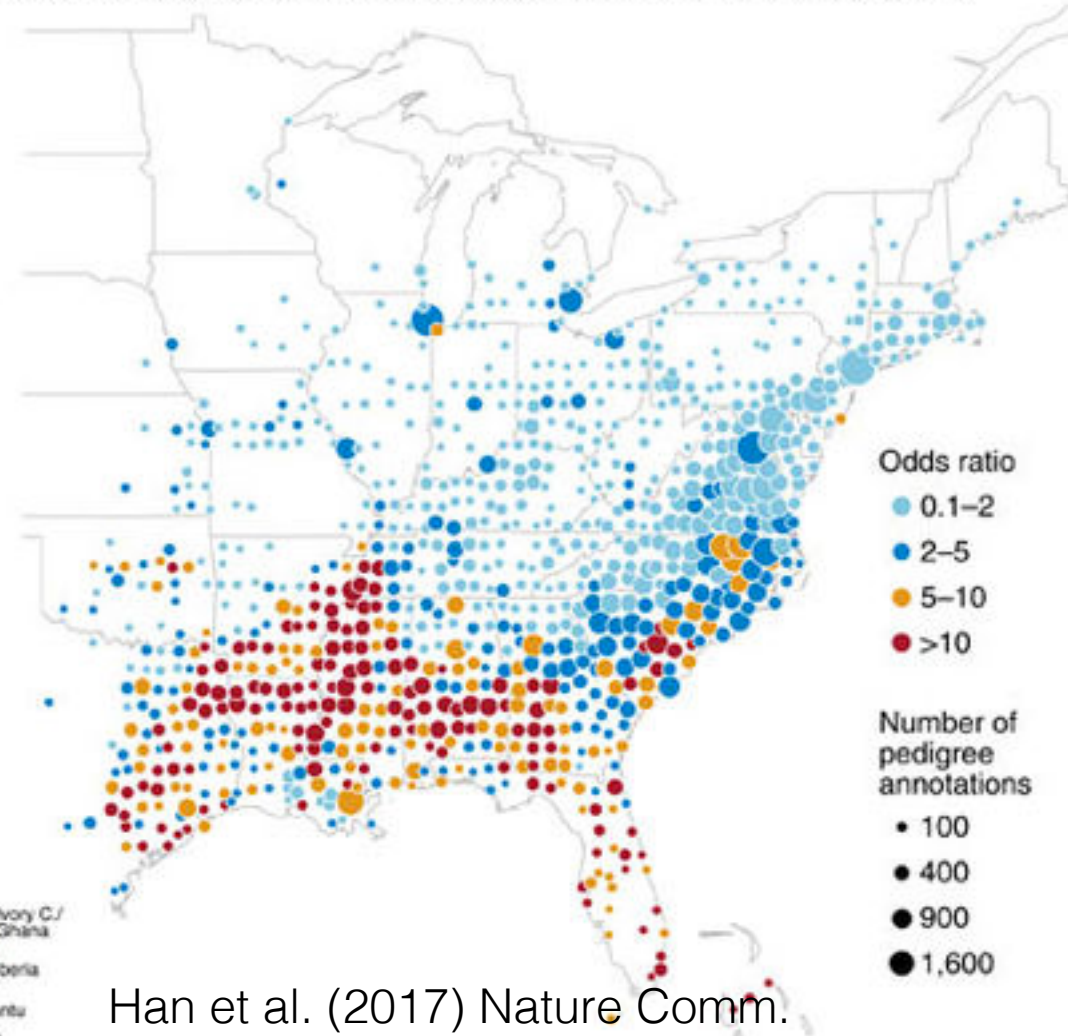
- In all samples, global admixture of 20 populations (donut charts);
- For some samples, birth locations of ancestors in pedigrees.



e

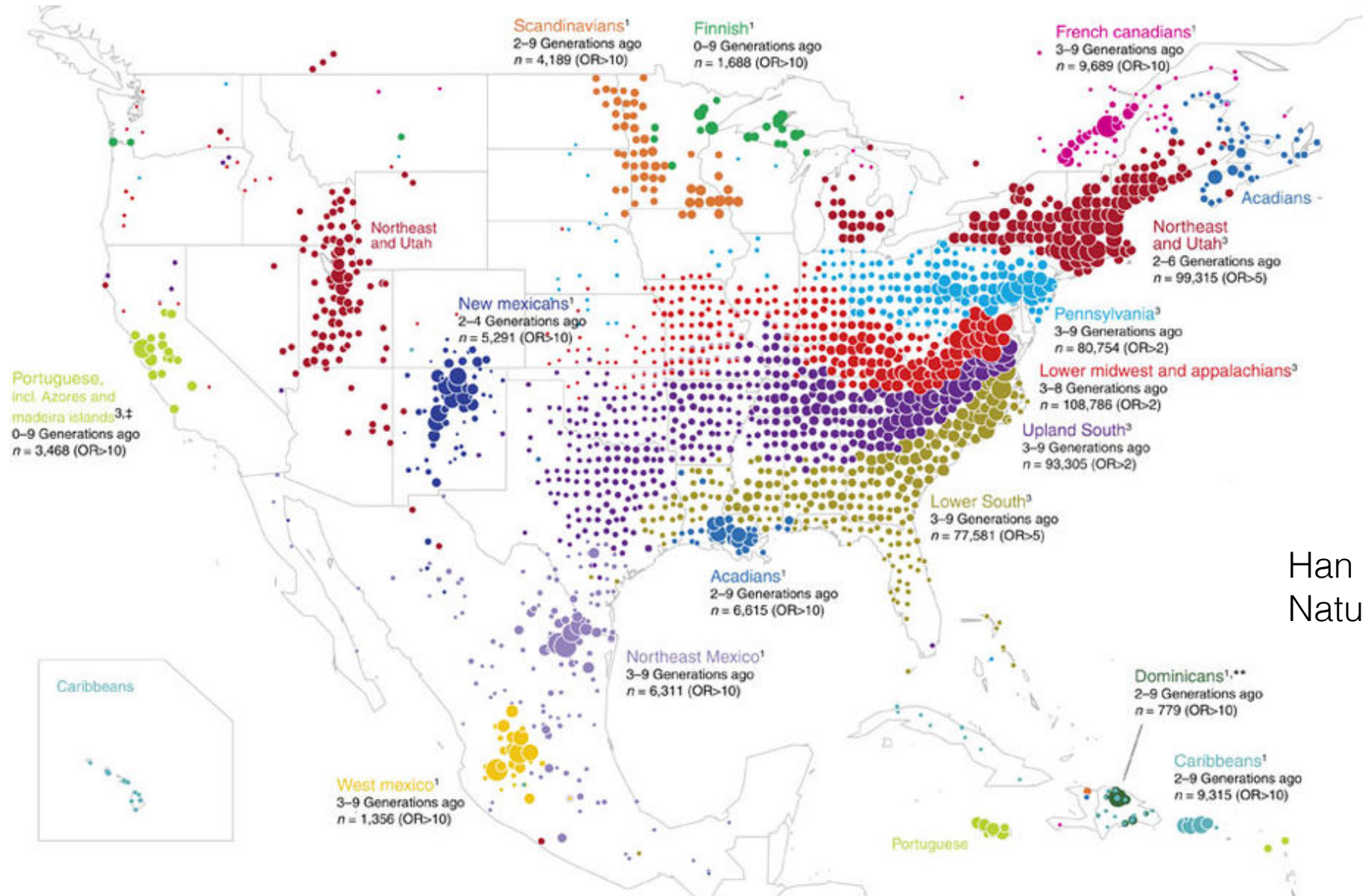
Visualize geographic distribution of ancestral birth locations in each cluster.

Map below shows birth locations of ancestors in the African American cluster. Locations are colored by degree of over-representation (odds ratio), and scaled by number of birth location annotations.



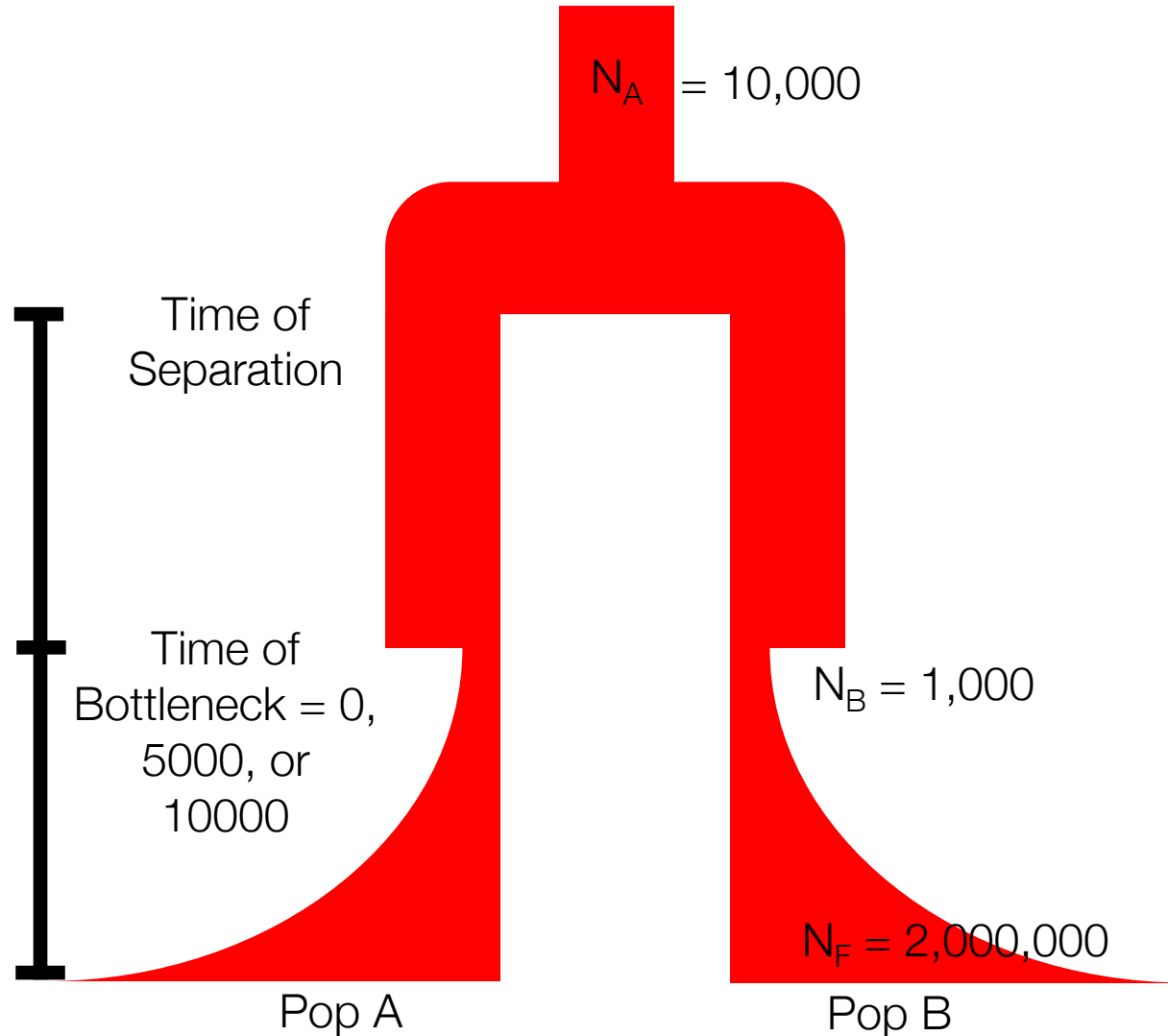
Han et al. (2017) Nature Comm.

IBD on a large scale



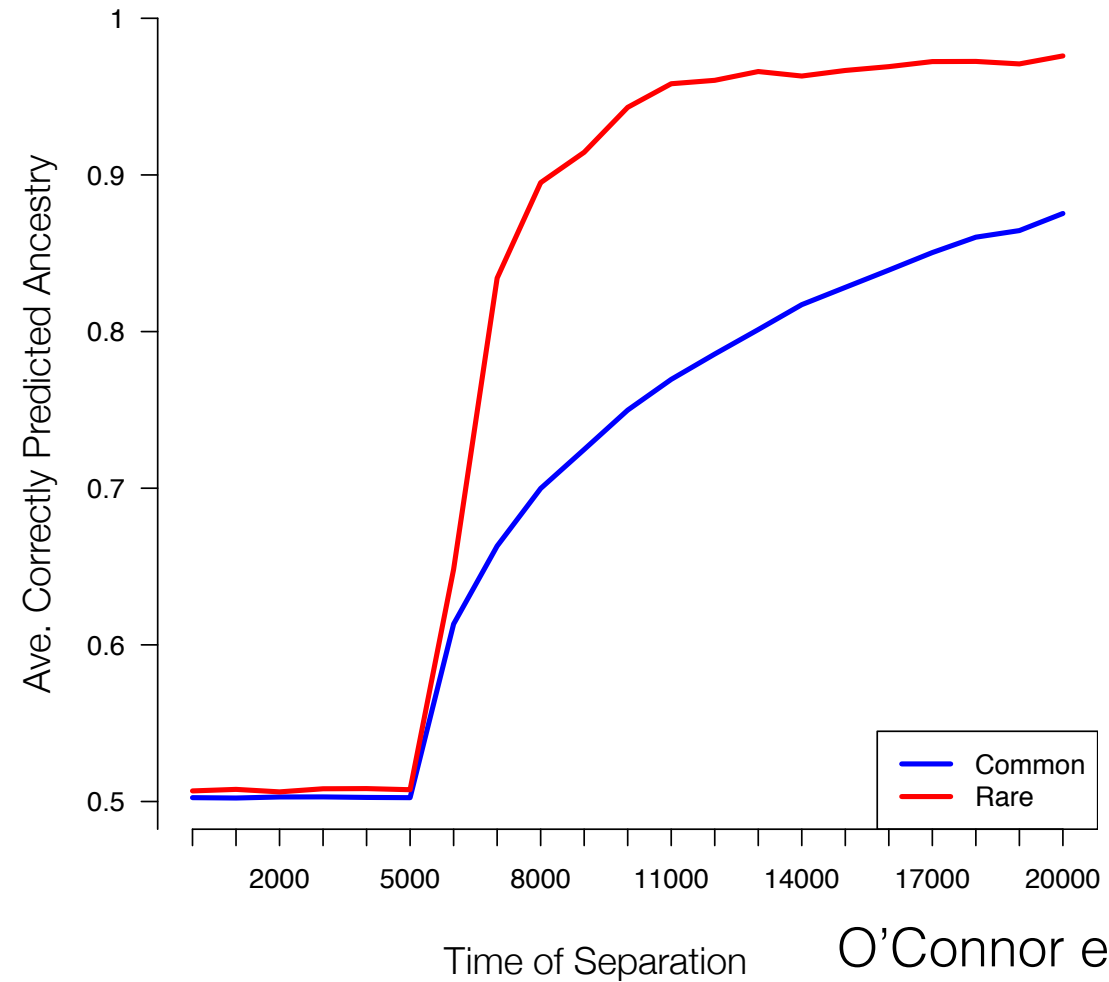
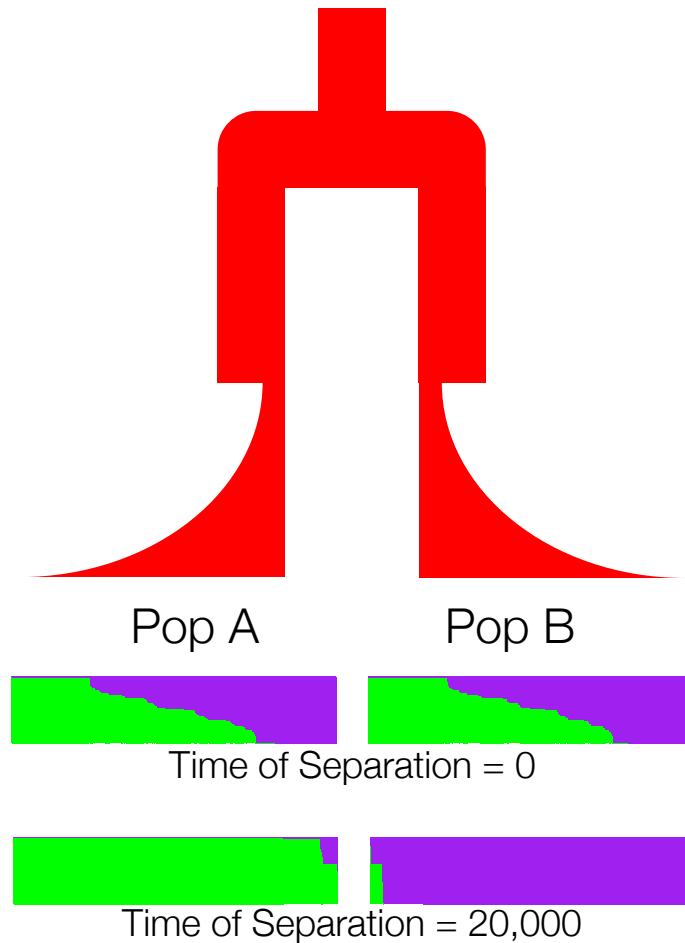
Han et al. (2017)
Nature Comm.

Rare VS Common: Population Structure Simulations



O'Connor et al. (2014)
Mol. Biol. Evol.

Rare VS Common: Assignment of Ancestry Proportions



O'Connor et al. (2014)
Mol. Biol. Evol.

Rare VS Common: Which has Greater Information? And When?

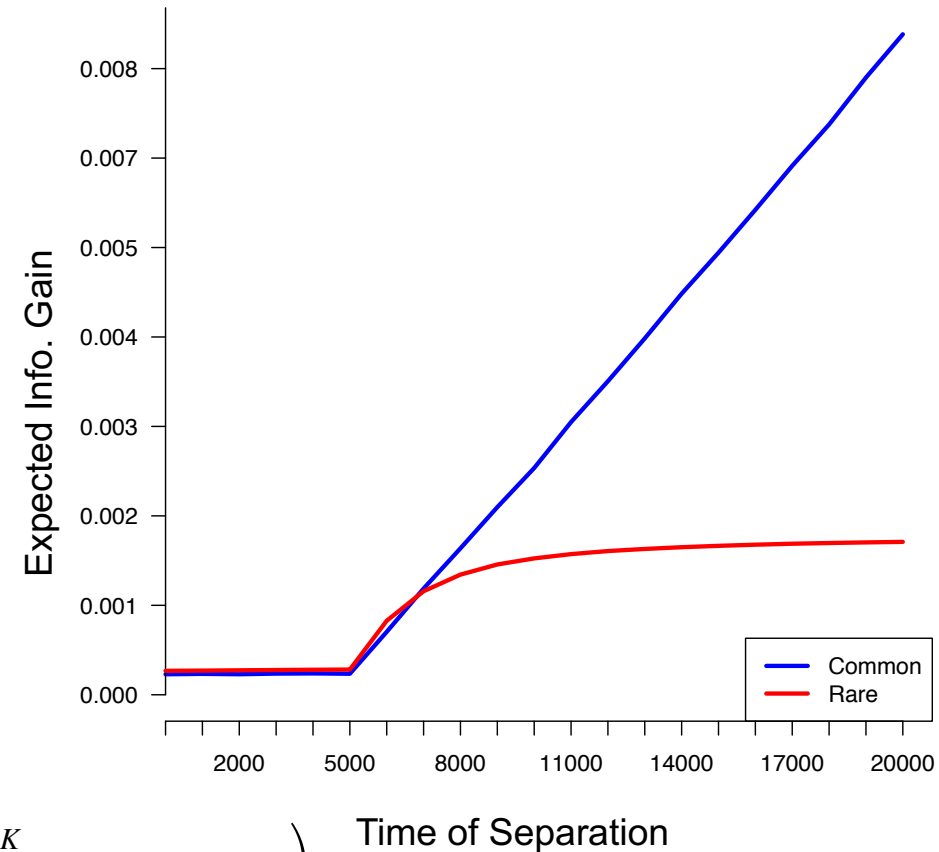
Information Gain: how well a variant can distinguish between populations.
(Rosenberg et al. 2003)

$$I_n(Q; J) = \sum_{j=1}^N \left(-p_j \ln p_j + \sum_{i=1}^K q_i p_{ij} \ln p_{ij} \right)$$

Expected Information Gain

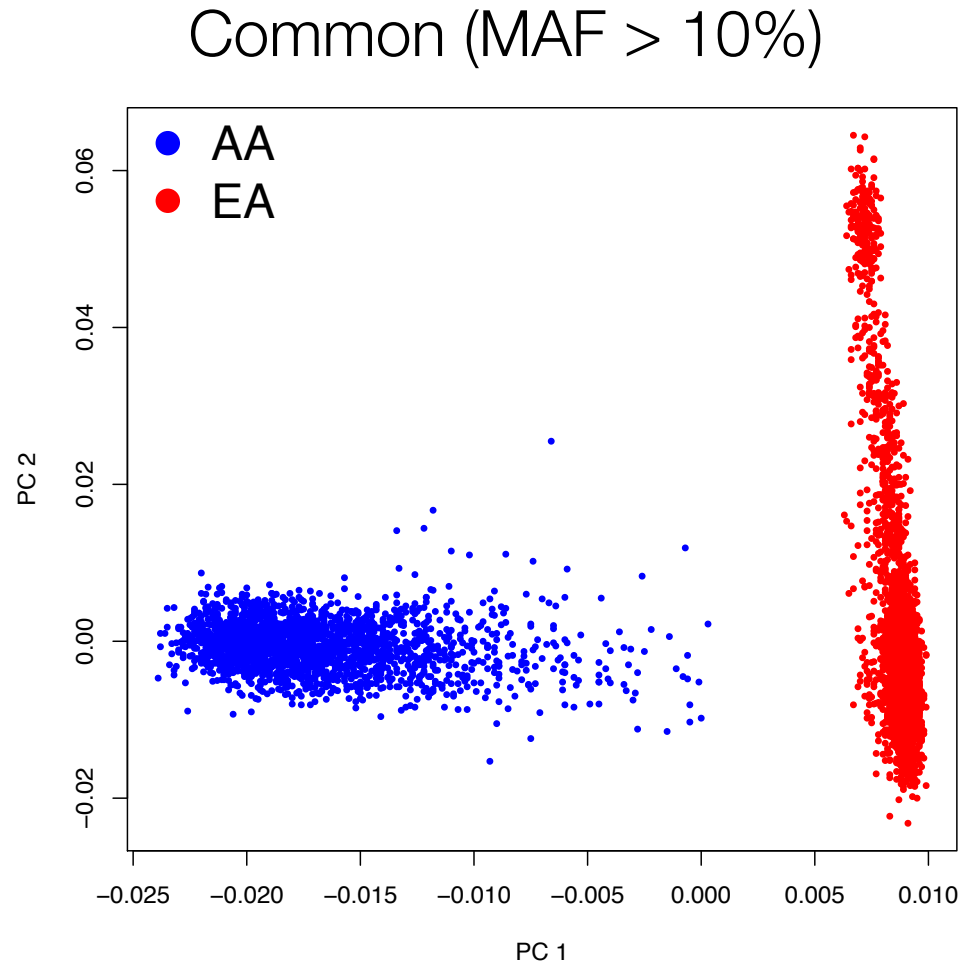
- Calculate for a specific site count
- Correct for missing data
- Weighted average to calculate across a range of frequency (rare or common)

$$E(I_n | C, M) = \sum_{m \in M} \sum_{l=0}^C r_{lm} \times \sum_{j=1}^N \left(-p_{jlm} \ln p_{jlm} + \sum_{i=1}^K q_i p_{ijlm} \ln p_{ijlm} \right)$$



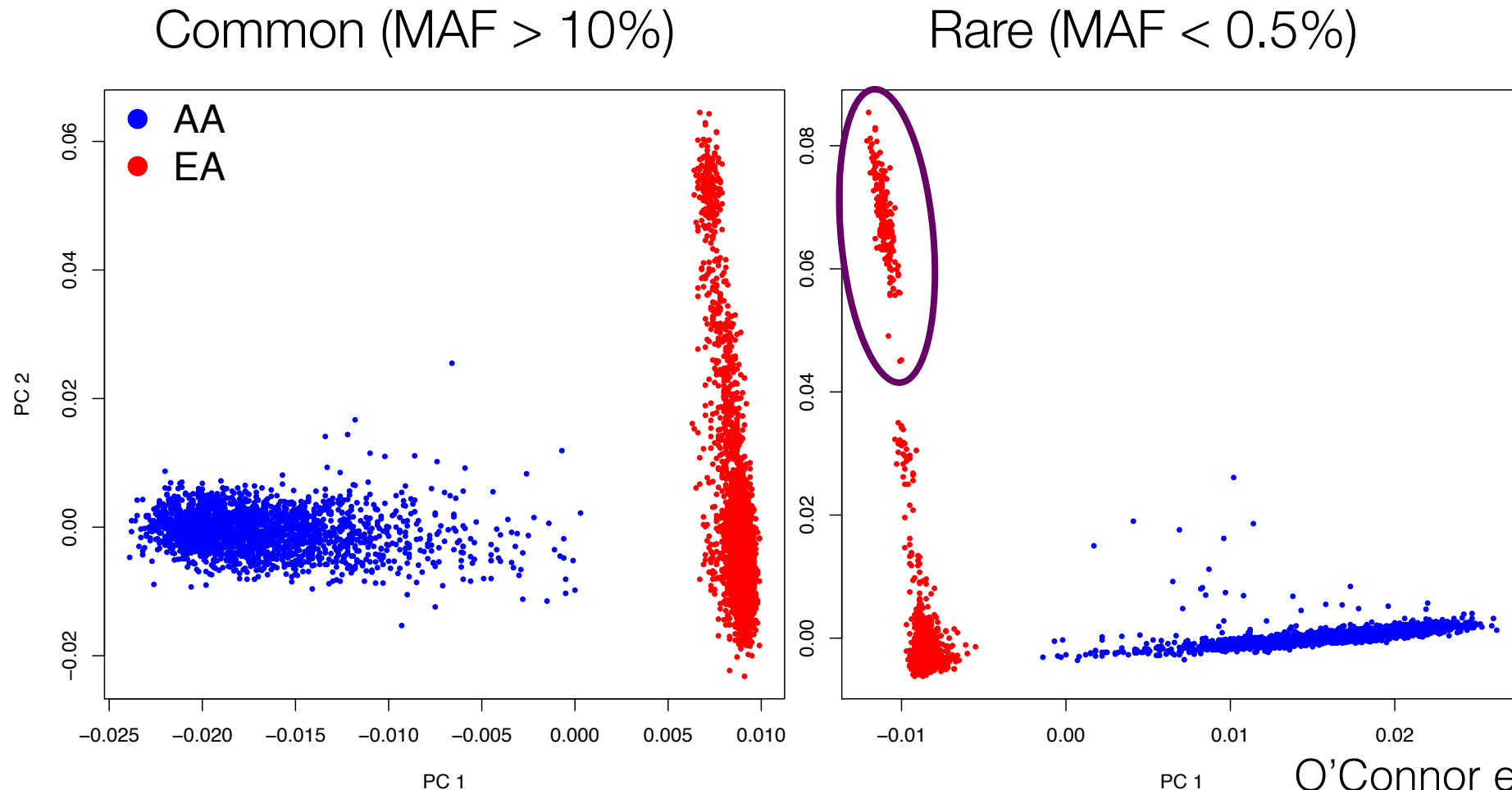
O'Connor et al. (2014)
Mol. Biol. Evol.

Rare Variants Identify Cryptic Populations



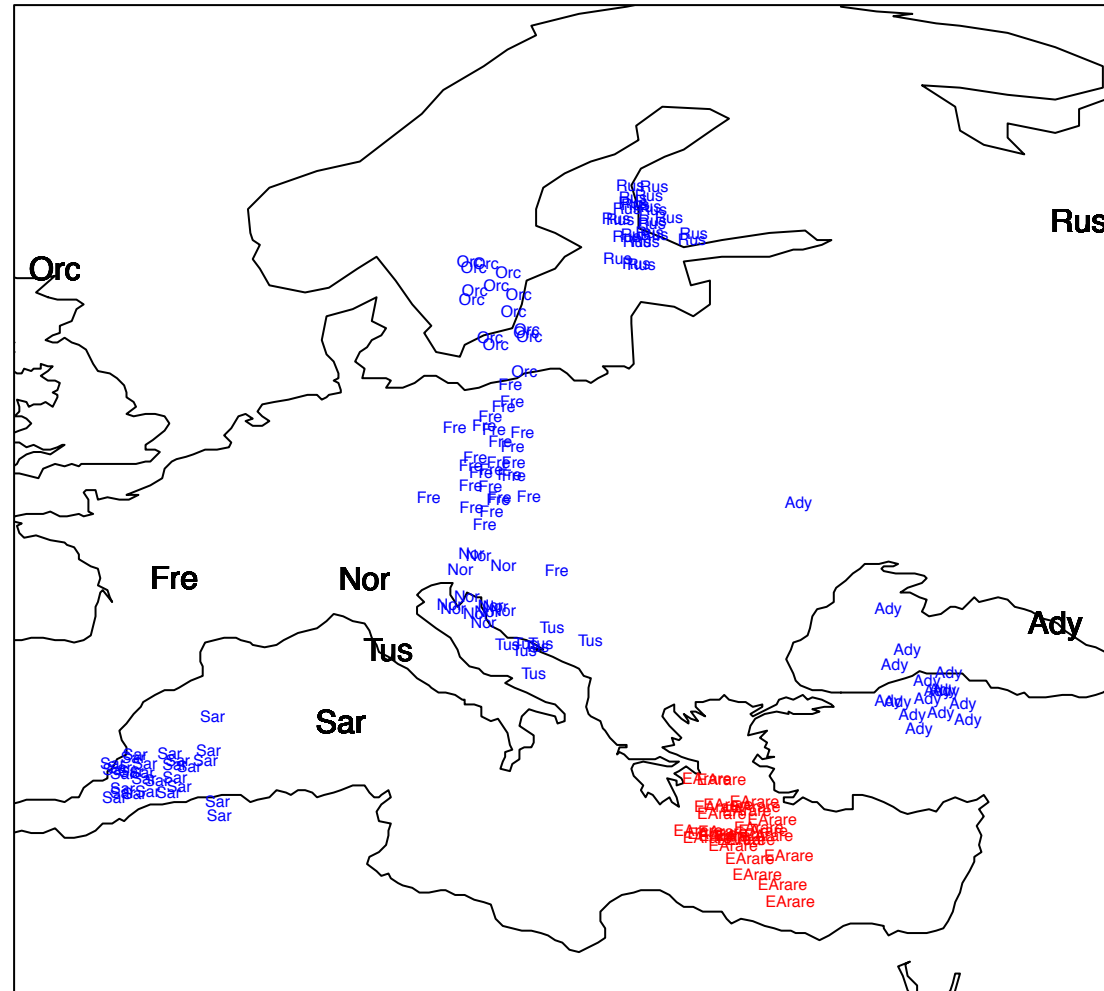
O'Connor et al. (2014)
Mol. Biol. Evol.

Rare Variants Identify Cryptic Populations



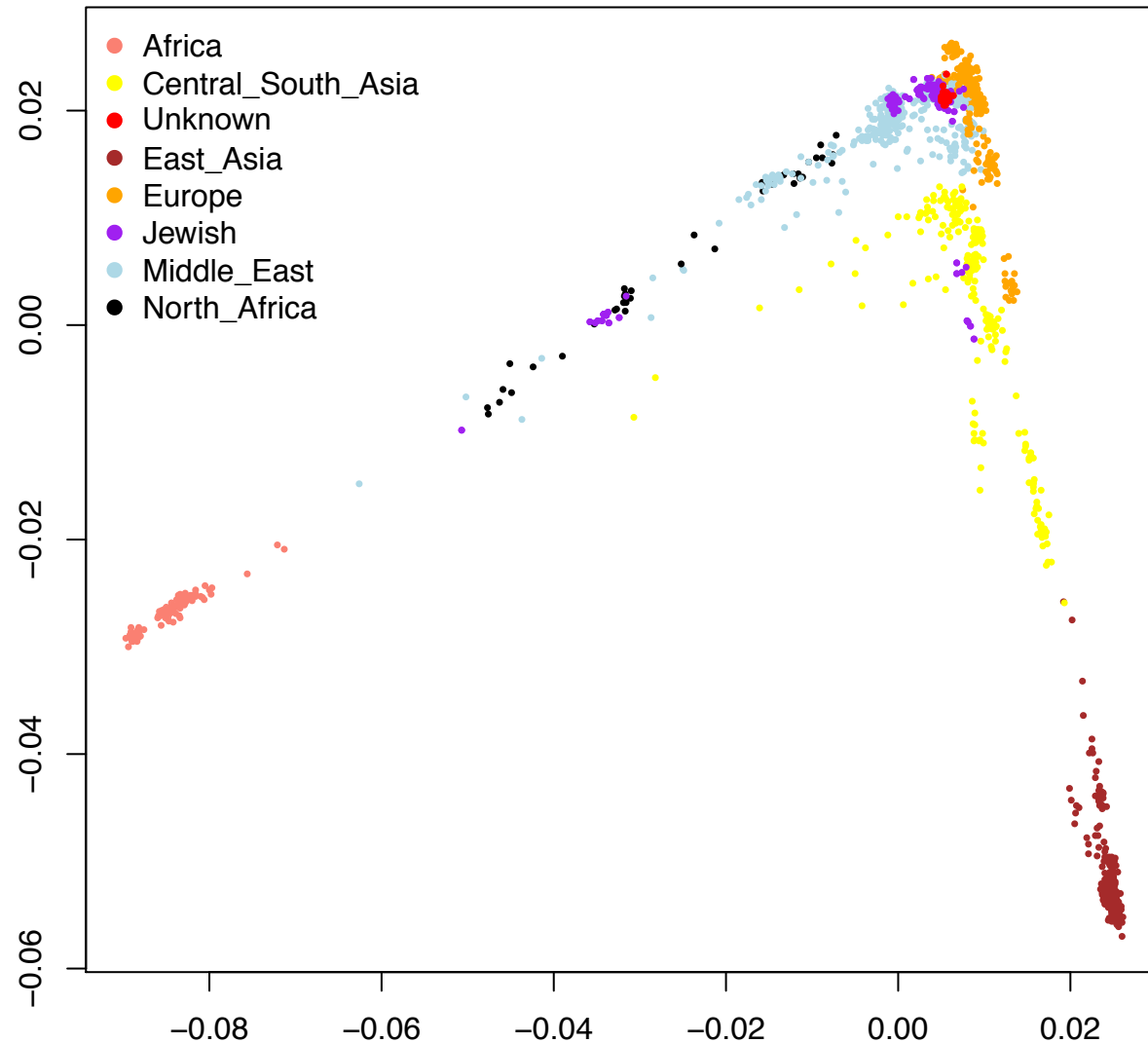
O'Connor et al. (2014)
Mol. Biol. Evol.

What is Their Geographic Ancestry?



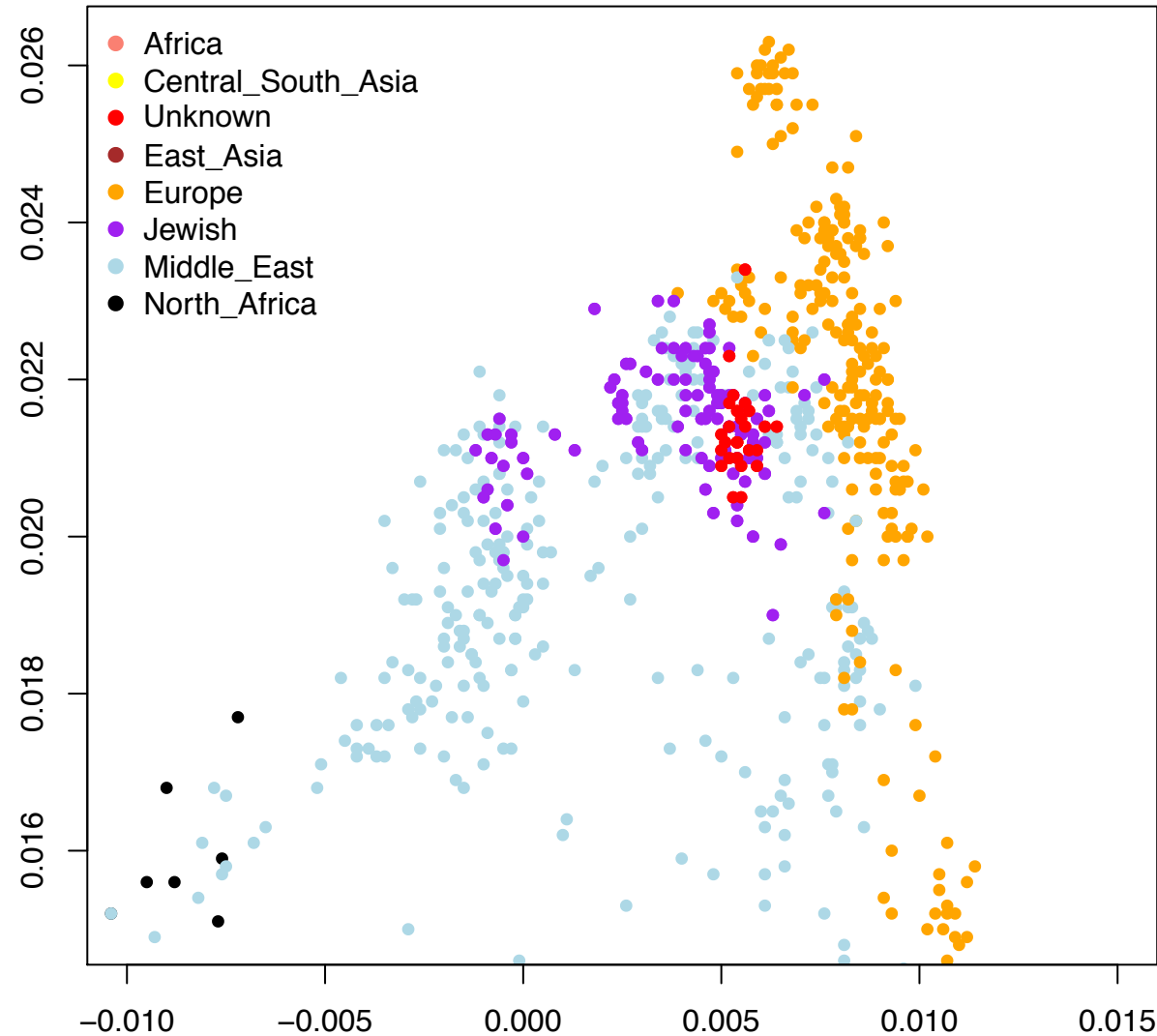
O'Connor et al. (2014)
Mol. Biol. Evol.

PCA of Global Diversity Including Cryptic Population



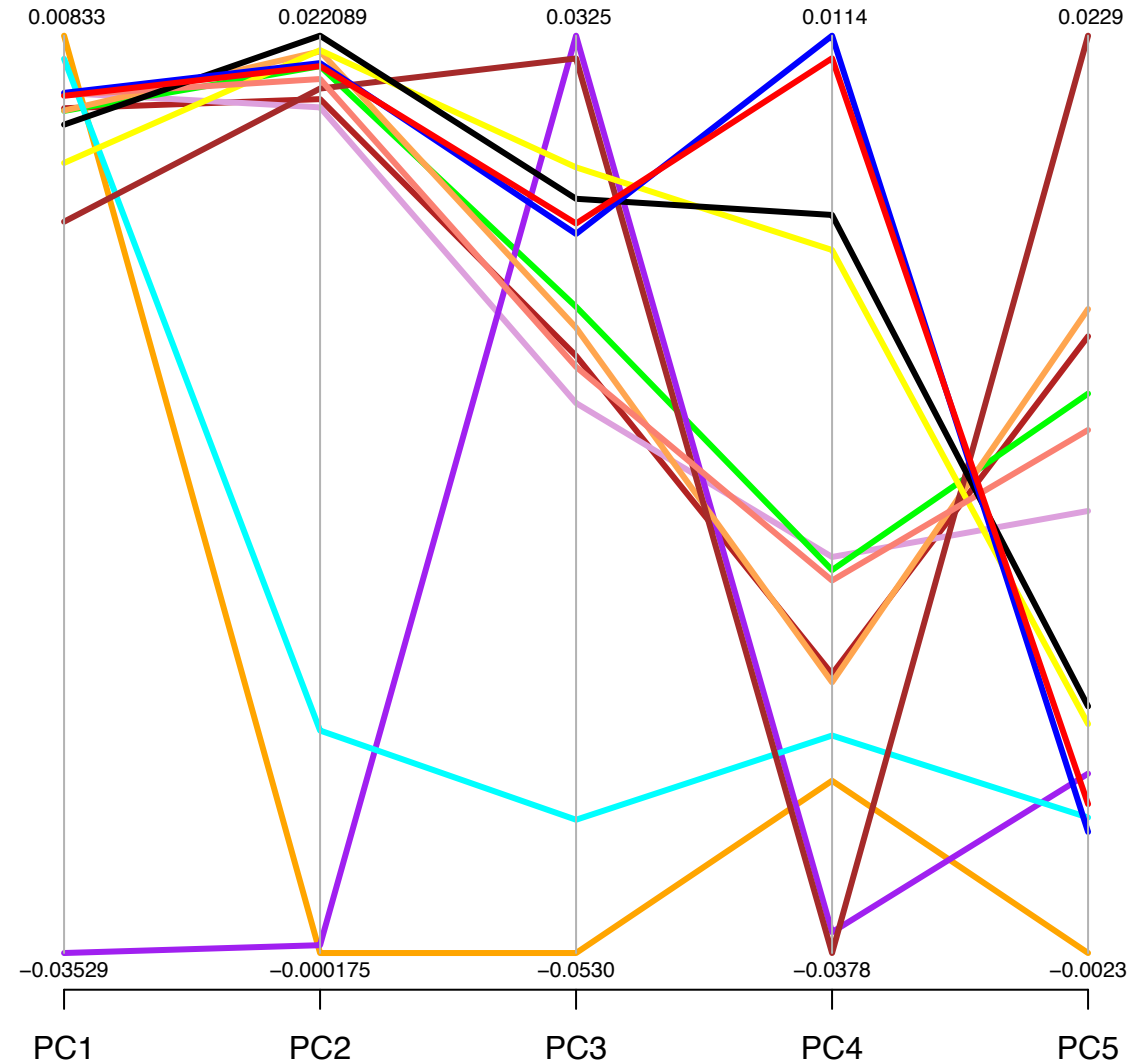
O'Connor et al. (2014)
Mol. Biol. Evol.

PCA of Global Diversity Including Cryptic Population



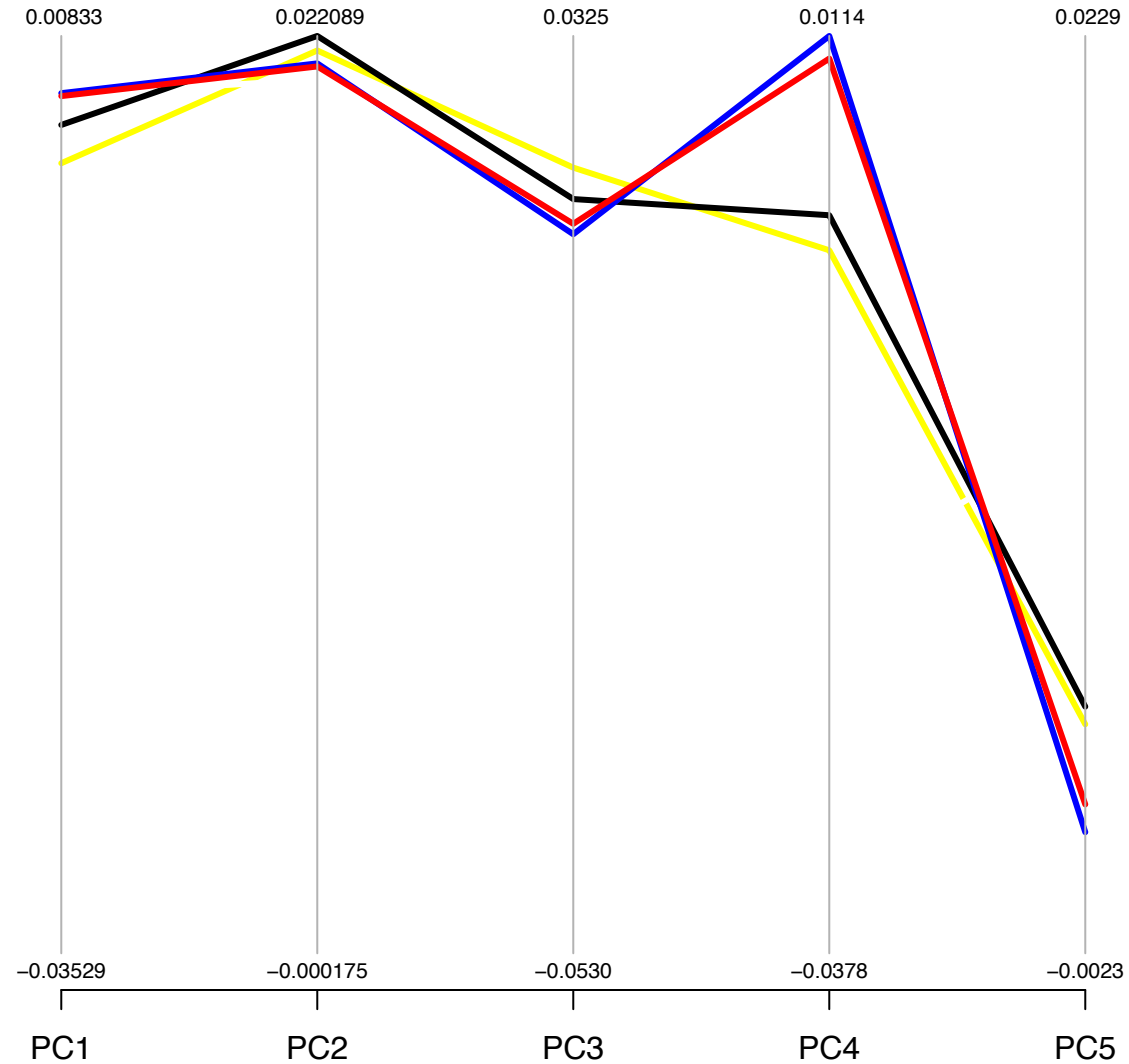
O'Connor et al. (2014)
Mol. Biol. Evol.

Population Average PCA with More Axes



O'Connor et al. (2014)
Mol. Biol. Evol.

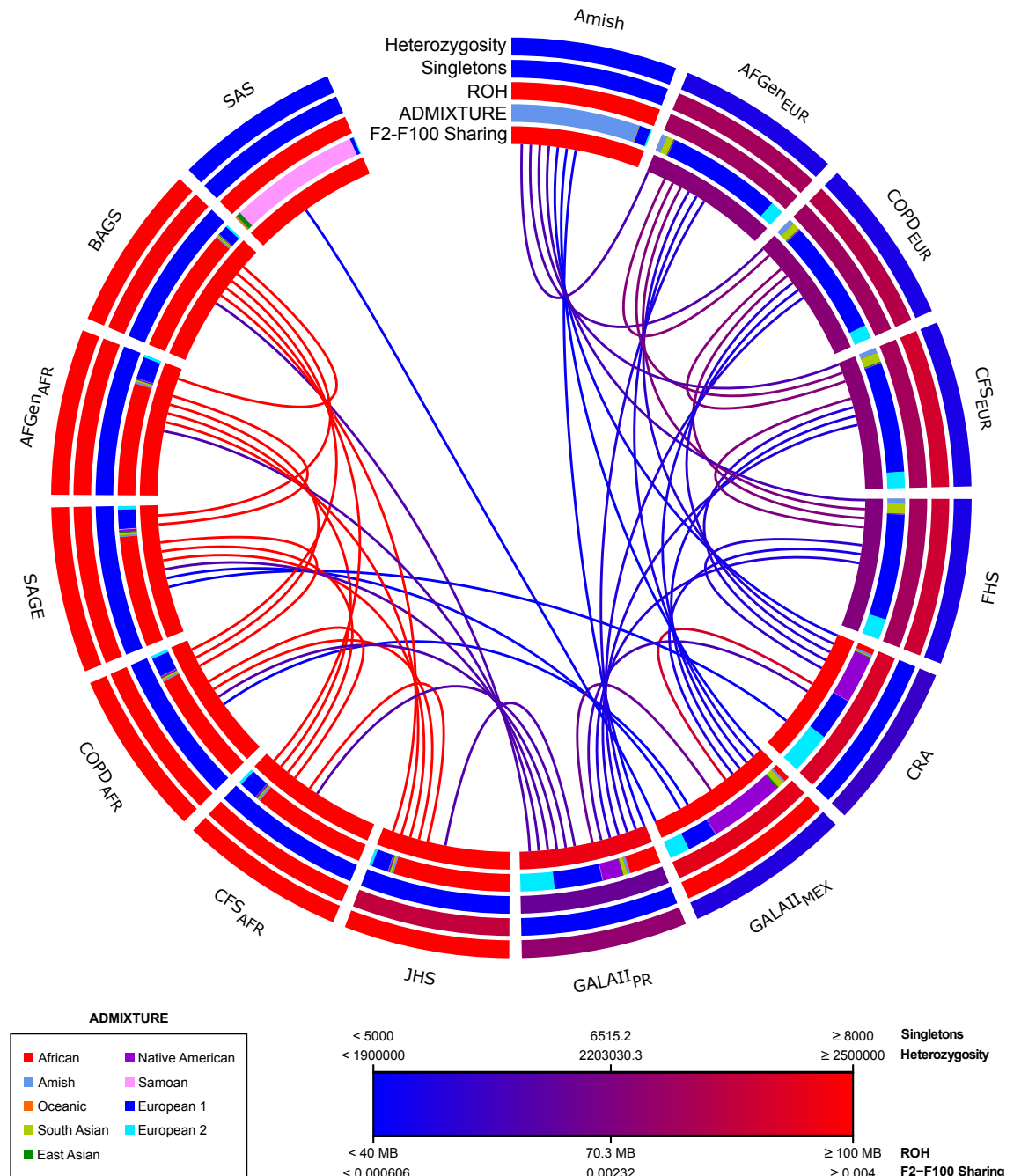
Population Average PCA with More Axes



O'Connor et al. (2014)
Mol. Biol. Evol.

Trans-Omics for Precision Medicine (TOPMed) Cohorts

- $N \cong 18K$
- This data freeze has 15 cohorts, each with 100s of samples
- Predominantly African, Latino, and European American
 - Samoa
 - Amish
- All are well characterized for heart, lung, blood, and sleep phenotypes



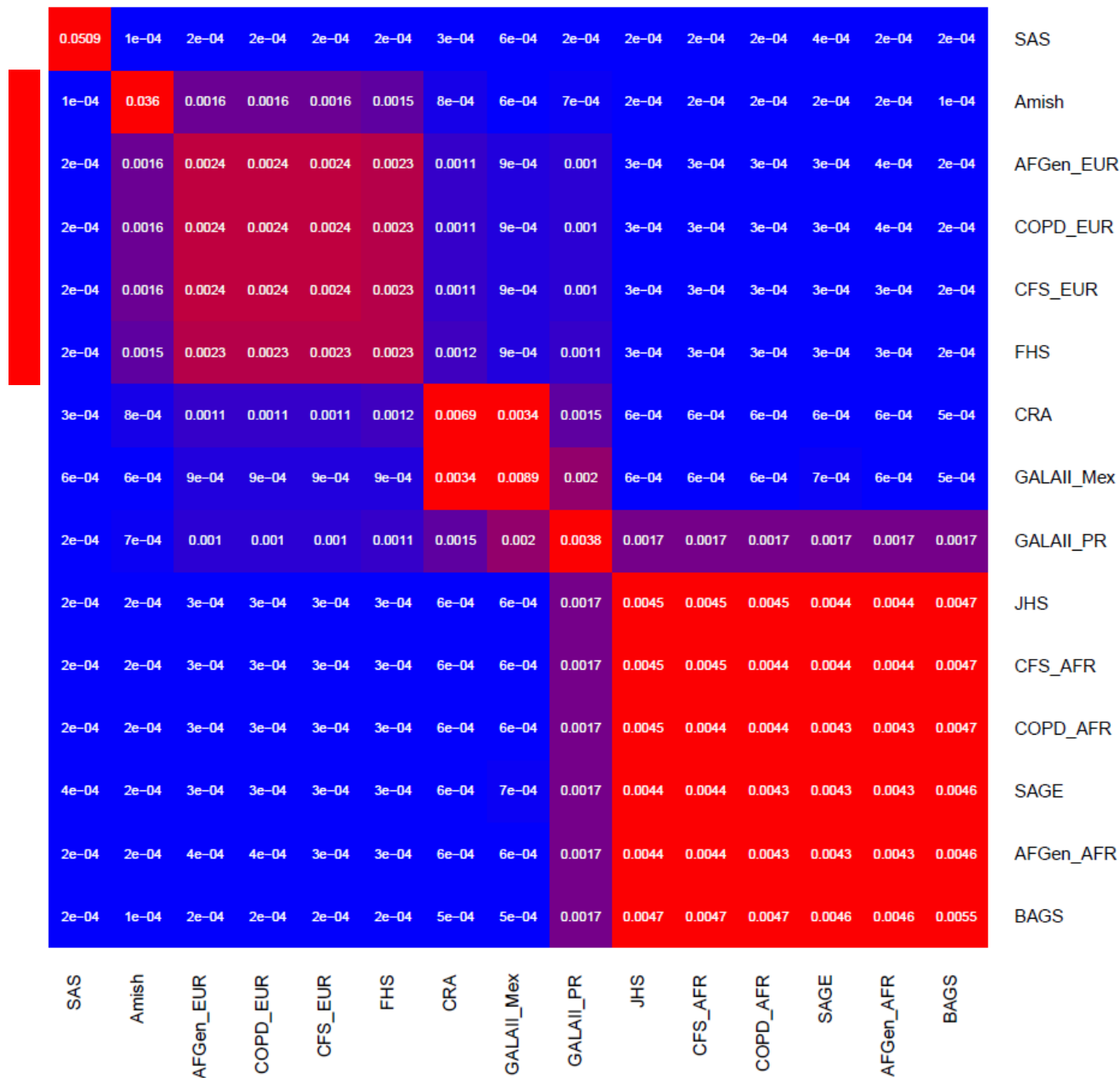
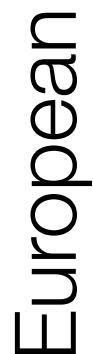
Rare variant sharing across cohorts

- Allele Count 2 to 100
- Corrected for:
 - sample size
 - Genome-wide heterozygosity

0.0509	1e-04	2e-04	2e-04	2e-04	2e-04	3e-04	6e-04	2e-04	2e-04	2e-04	2e-04	4e-04	2e-04	2e-04	SAS
1e-04	0.036	0.0016	0.0016	0.0016	0.0015	8e-04	6e-04	7e-04	2e-04	2e-04	2e-04	2e-04	2e-04	1e-04	Amish
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	4e-04	2e-04	AFGen_EUR
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	4e-04	2e-04	COPD_EUR
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	3e-04	2e-04	CFS_EUR
2e-04	0.0015	0.0023	0.0023	0.0023	0.0023	0.0012	9e-04	0.0011	3e-04	3e-04	3e-04	3e-04	3e-04	2e-04	FHS
3e-04	8e-04	0.0011	0.0011	0.0011	0.0012	0.0069	0.0034	0.0015	6e-04	6e-04	6e-04	6e-04	6e-04	5e-04	CRA
6e-04	6e-04	9e-04	9e-04	9e-04	9e-04	0.0034	0.0089	0.002	6e-04	6e-04	6e-04	7e-04	6e-04	5e-04	GALAII_Mex
2e-04	7e-04	0.001	0.001	0.001	0.0011	0.0015	0.002	0.0038	0.0017	0.0017	0.0017	0.0017	0.0017	0.0017	GALAII_PR
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0045	0.0045	0.0044	0.0044	0.0047	JHS
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0045	0.0044	0.0044	0.0044	0.0047	CFS_AFR
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0044	0.0044	0.0043	0.0043	0.0047	COPD_AFR
4e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	7e-04	0.0017	0.0044	0.0044	0.0043	0.0043	0.0043	0.0046	SAGE
2e-04	2e-04	4e-04	4e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0044	0.0044	0.0043	0.0043	0.0043	0.0046	AFGen_AFR
2e-04	1e-04	2e-04	2e-04	2e-04	2e-04	5e-04	5e-04	0.0017	0.0047	0.0047	0.0047	0.0046	0.0046	0.0055	BAGS
SAS	Amish	AFGen_EUR	COPD_EUR	CFS_EUR	FHS	CRA	GALAII_Mex	GALAII_PR	JHS	CFS_AFR	COPD_AFR	SAGE	AFGen_AFR	BAGS	

Rare
variant
sharing
across
cohorts

- Allele Count 2 to 100
- Corrected for:
 - sample size
 - Genome-wide heterozygosity



Rare variant sharing across cohorts

- Allele Count 2 to 100
- Corrected for:
 - sample size
 - Genome-wide heterozygosity

European

0.0509	1e-04	2e-04	2e-04	2e-04	2e-04	3e-04	6e-04	2e-04	2e-04	2e-04	2e-04	4e-04	2e-04	2e-04	SAS
1e-04	0.036	0.0016	0.0016	0.0016	0.0015	8e-04	6e-04	7e-04	2e-04	2e-04	2e-04	2e-04	2e-04	1e-04	Amish
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	4e-04	2e-04	AFGen_EUR
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	4e-04	2e-04	COPD_EUR
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	3e-04	2e-04	CFS_EUR
2e-04	0.0015	0.0023	0.0023	0.0023	0.0023	0.0012	9e-04	0.0011	3e-04	3e-04	3e-04	3e-04	3e-04	2e-04	FHS
3e-04	8e-04	0.0011	0.0011	0.0011	0.0012	0.0069	0.0034	0.0015	6e-04	6e-04	6e-04	6e-04	6e-04	5e-04	CRA
6e-04	6e-04	9e-04	9e-04	9e-04	9e-04	0.0034	0.0089	0.002	6e-04	6e-04	6e-04	7e-04	6e-04	5e-04	GALAII_Mex
2e-04	7e-04	0.001	0.001	0.001	0.0011	0.0015	0.002	0.0038	0.0017	0.0017	0.0017	0.0017	0.0017	0.0017	GALAII_PR
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0045	0.0045	0.0044	0.0044	0.0047	JHS
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0045	0.0044	0.0044	0.0044	0.0047	CFS_AFR
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0044	0.0044	0.0043	0.0043	0.0047	COPD_AFR
4e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	7e-04	0.0017	0.0044	0.0044	0.0043	0.0043	0.0043	0.0046	SAGE
2e-04	2e-04	4e-04	4e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0044	0.0044	0.0043	0.0043	0.0043	0.0046	AFGen_AFR
2e-04	1e-04	2e-04	2e-04	2e-04	2e-04	5e-04	5e-04	0.0017	0.0047	0.0047	0.0047	0.0046	0.0046	0.0055	BAGS

African

SAS
Amish
AFGen_EUR
COPD_EUR
CFS_EUR
FHS
CRA
GALAII_Mex
GALAII_PR
JHS
CFS_AFR
COPD_AFR
SAGE
AFGen_AFR
BAGS

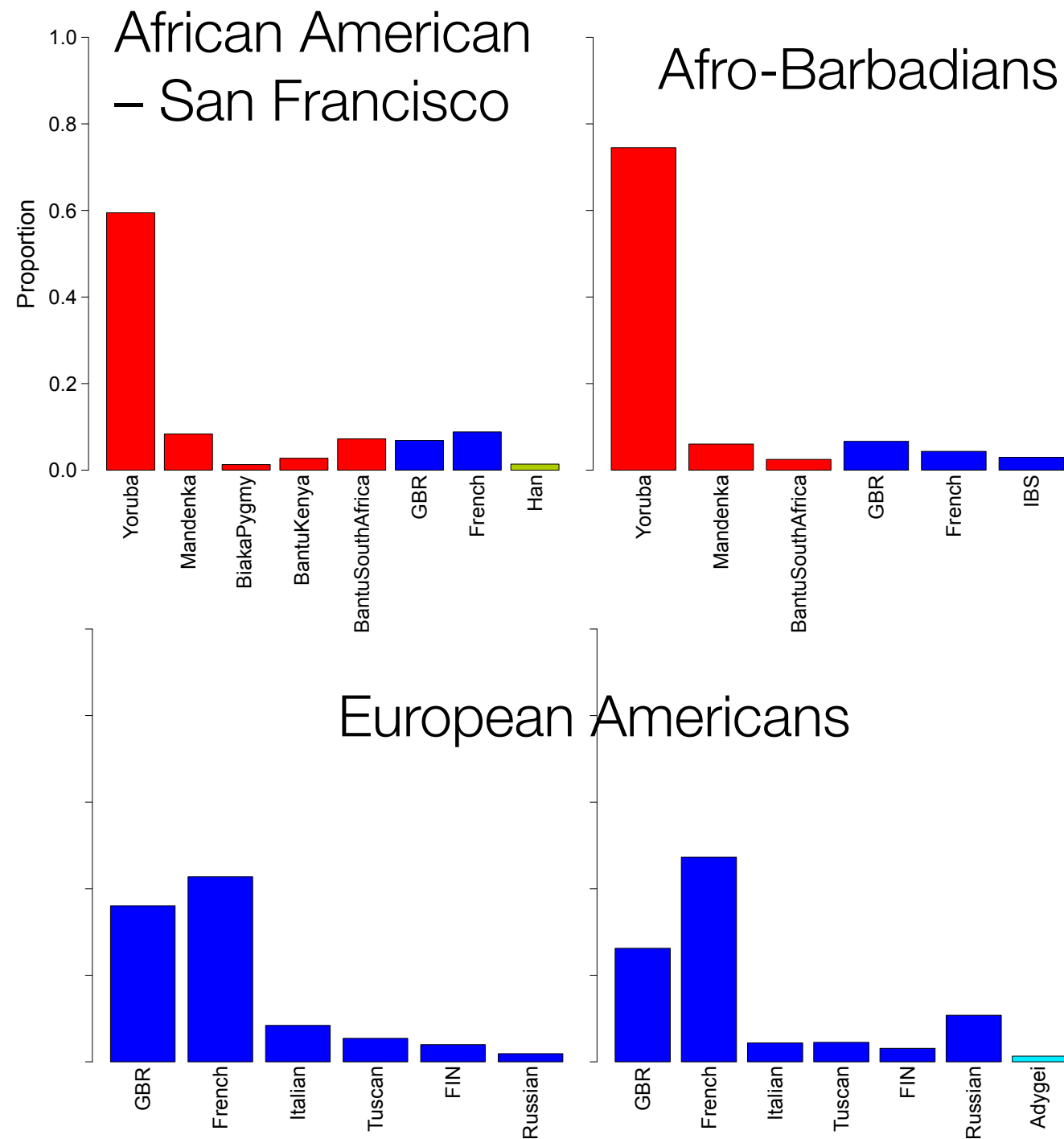
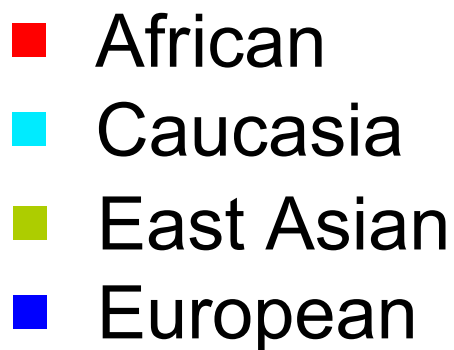
Rare variant sharing across cohorts

- Allele Count 2 to 100
- Corrected for:
 - sample size
 - Genome-wide heterozygosity

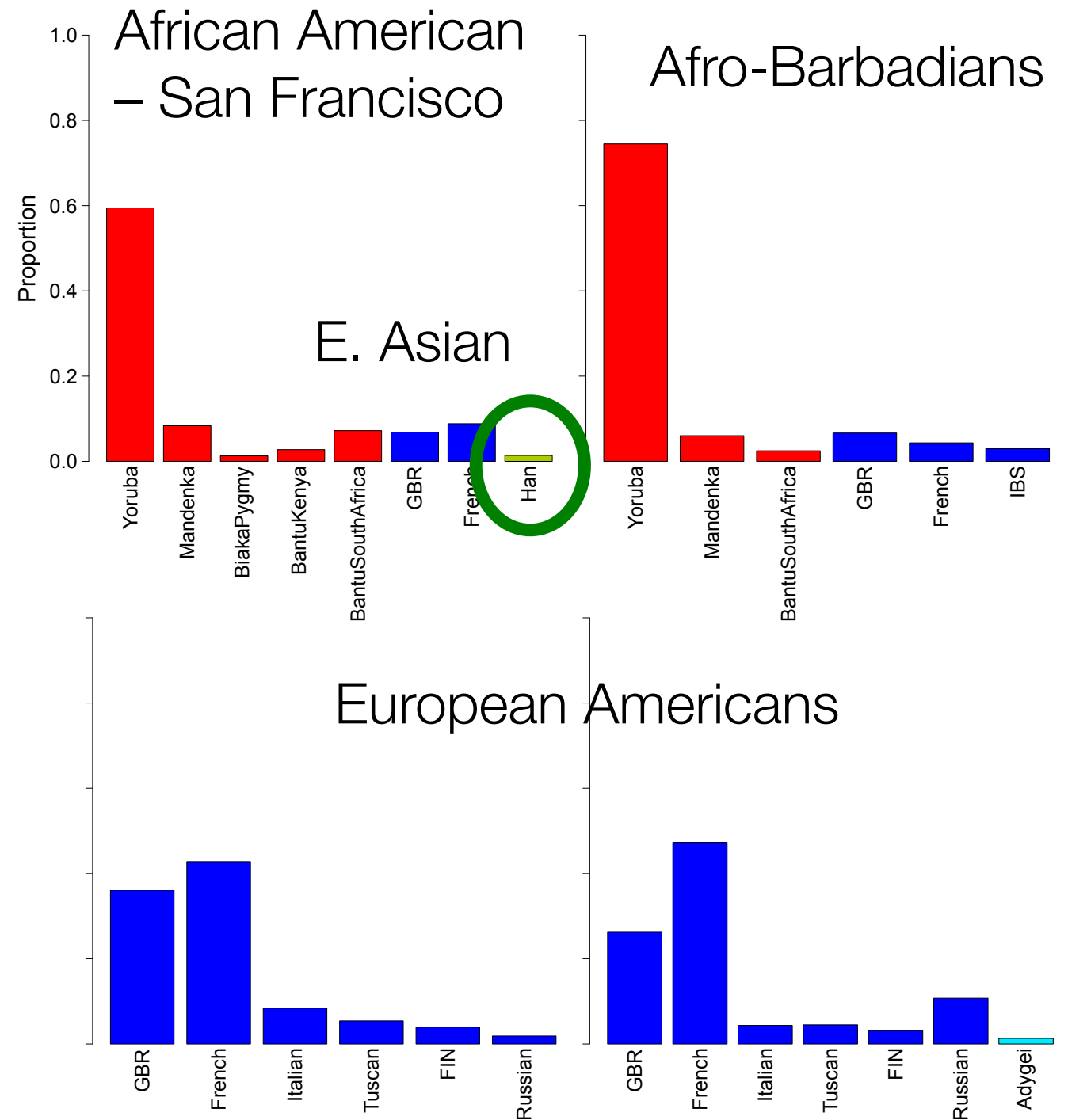
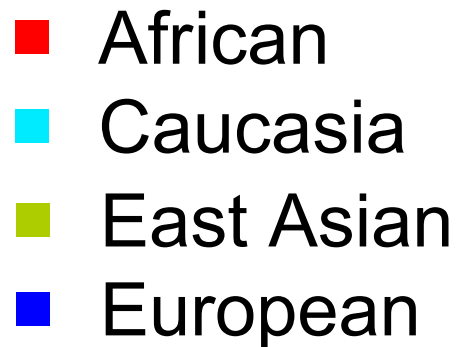
European
Latino
African

	0.0509	1e-04	2e-04	2e-04	2e-04	2e-04	3e-04	6e-04	2e-04	2e-04	2e-04	2e-04	4e-04	2e-04	2e-04	SAS
1e-04	0.036	0.0016	0.0016	0.0016	0.0015	8e-04	6e-04	7e-04	2e-04	2e-04	2e-04	2e-04	2e-04	2e-04	1e-04	Amish
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	3e-04	4e-04	2e-04	AFGen_EUR
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	3e-04	4e-04	2e-04	COPD_EUR
2e-04	0.0016	0.0024	0.0024	0.0024	0.0023	0.0011	9e-04	0.001	3e-04	3e-04	3e-04	3e-04	3e-04	3e-04	2e-04	CFS_EUR
2e-04	0.0015	0.0023	0.0023	0.0023	0.0023	0.0012	9e-04	0.0011	3e-04	3e-04	3e-04	3e-04	3e-04	3e-04	2e-04	FHS
3e-04	8e-04	0.0011	0.0011	0.0011	0.0012	0.0069	0.0034	0.0015	6e-04	6e-04	6e-04	6e-04	6e-04	6e-04	5e-04	CRA
6e-04	6e-04	9e-04	9e-04	9e-04	9e-04	0.0034	0.0089	0.002	6e-04	6e-04	6e-04	7e-04	6e-04	6e-04	5e-04	GALAII_Mex
2e-04	7e-04	0.001	0.001	0.001	0.0011	0.0015	0.002	0.0038	0.0017	0.0017	0.0017	0.0017	0.0017	0.0017	0.0017	GALAII_PR
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0045	0.0045	0.0044	0.0044	0.0044	0.0047	JHS
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0045	0.0044	0.0044	0.0044	0.0044	0.0047	CFS_AFR
2e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0045	0.0044	0.0044	0.0043	0.0043	0.0043	0.0047	COPD_AFR
4e-04	2e-04	3e-04	3e-04	3e-04	3e-04	6e-04	7e-04	0.0017	0.0044	0.0044	0.0043	0.0043	0.0043	0.0043	0.0046	SAGE
2e-04	2e-04	4e-04	4e-04	3e-04	3e-04	6e-04	6e-04	0.0017	0.0044	0.0044	0.0043	0.0043	0.0043	0.0043	0.0046	AFGen_AFR
2e-04	1e-04	2e-04	2e-04	2e-04	2e-04	5e-04	5e-04	0.0017	0.0047	0.0047	0.0047	0.0046	0.0046	0.0046	0.0055	BAGS
SAS	Amish	AFGen_EUR	COPD_EUR	CFS_EUR	FHS	CRA	GALAII_Mex	GALAII_PR	JHS	CFS_AFR	COPD_AFR	SAGE	AFGen_AFR	BAGS		

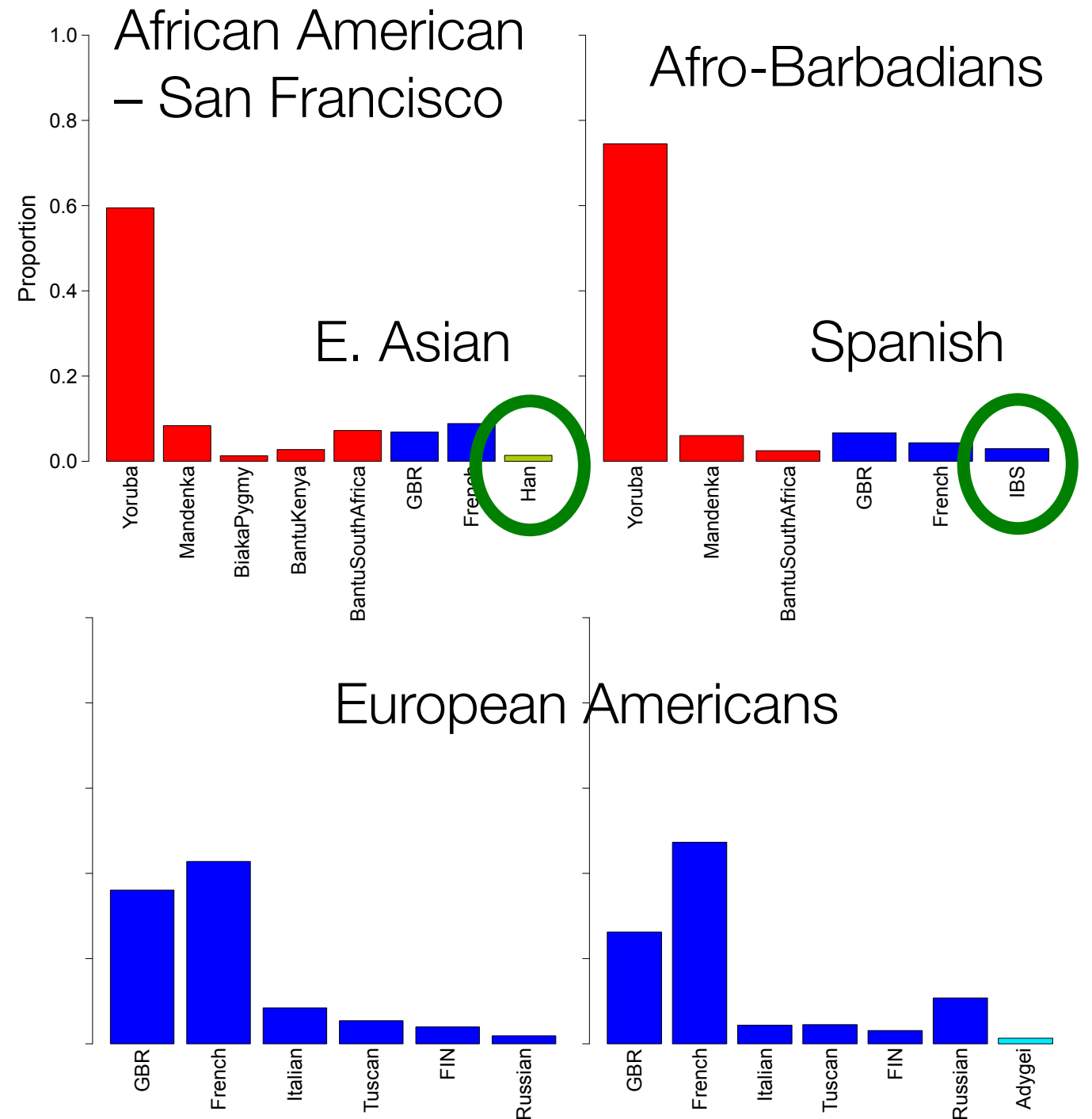
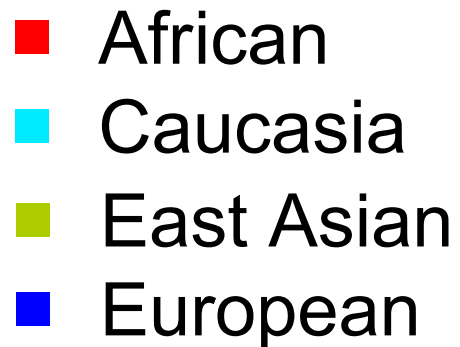
fineStructure analysis of genome-wide ancestry



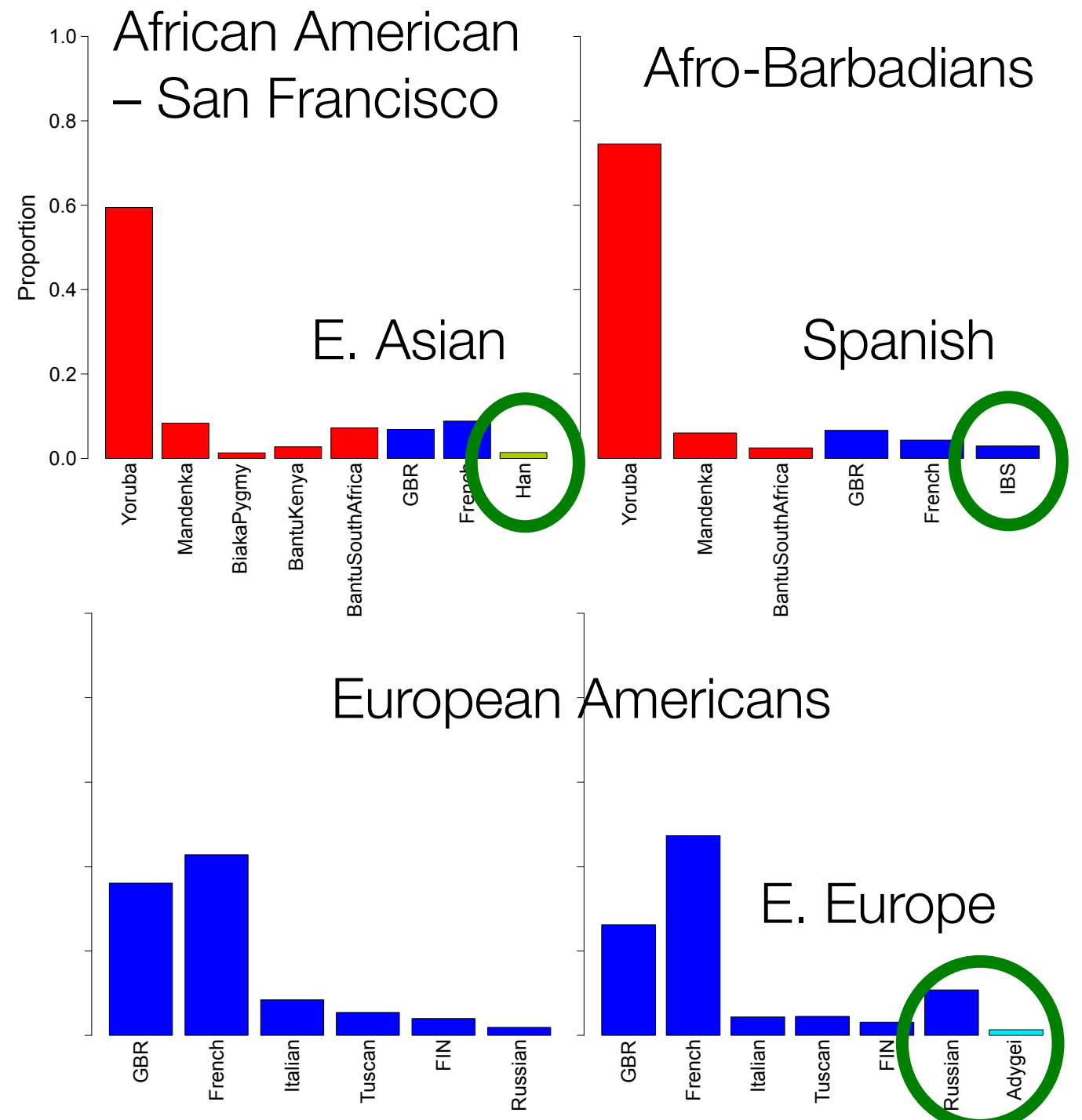
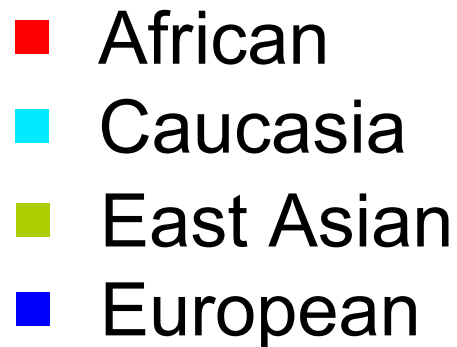
fineStructure analysis of genome-wide ancestry



fineStructure analysis of genome-wide ancestry

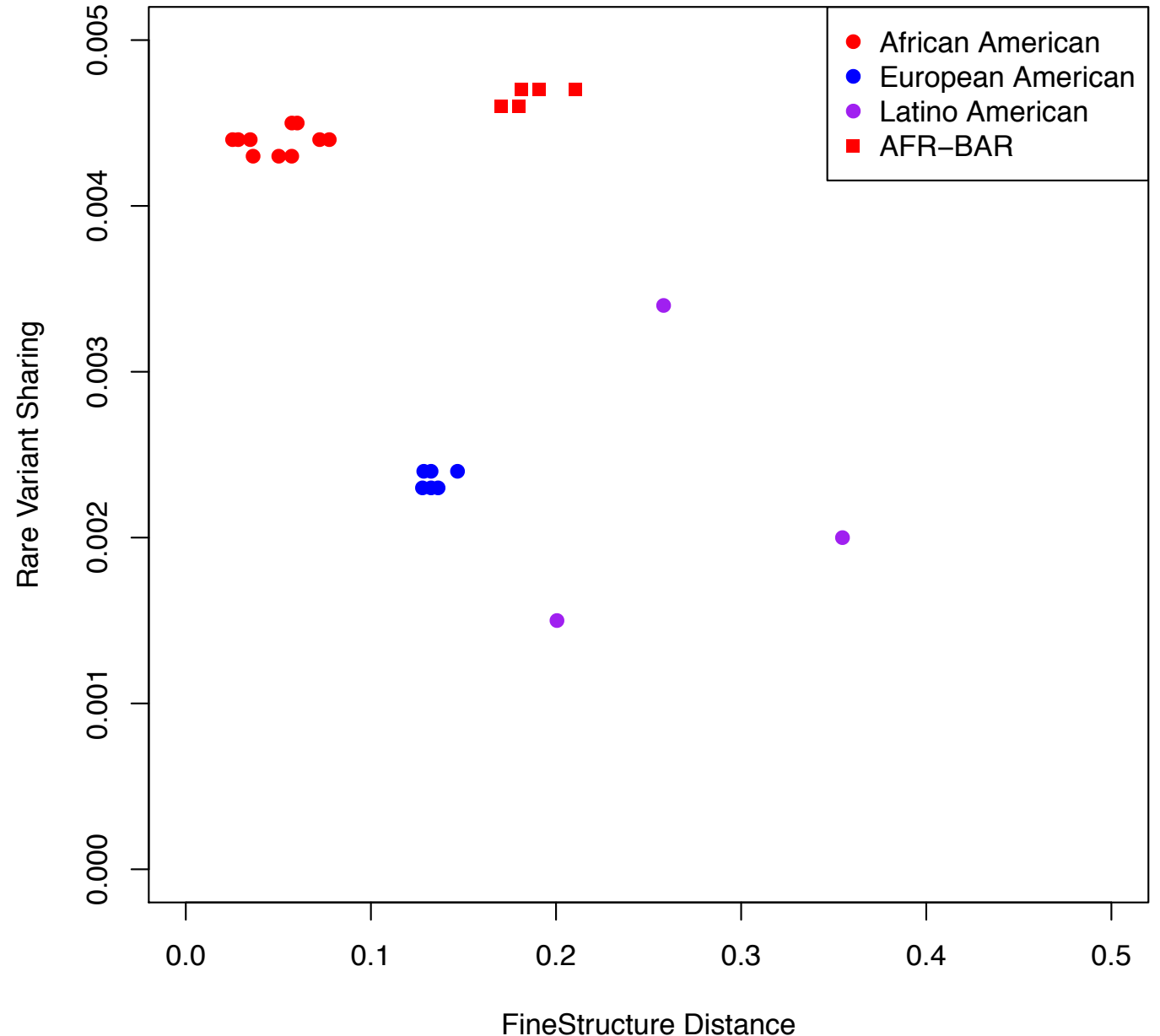


fineStructure analysis of genome-wide ancestry



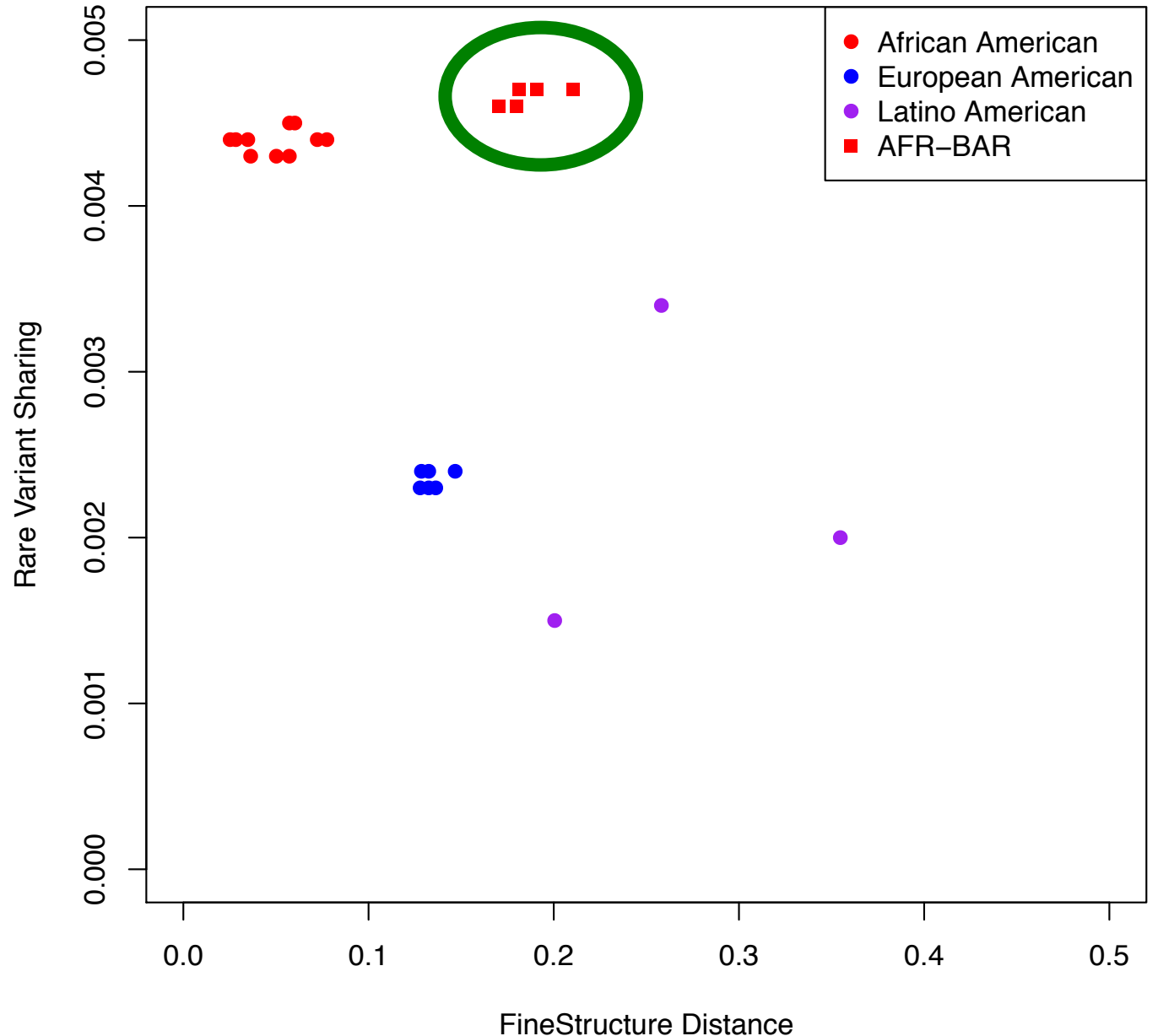
African American's have more homogeneous ancestral proportions

- Calculated Euclidian distance between fineSTRUCTURE proportions
- African American cohorts have the shortest distance and the greatest rare variant sharing

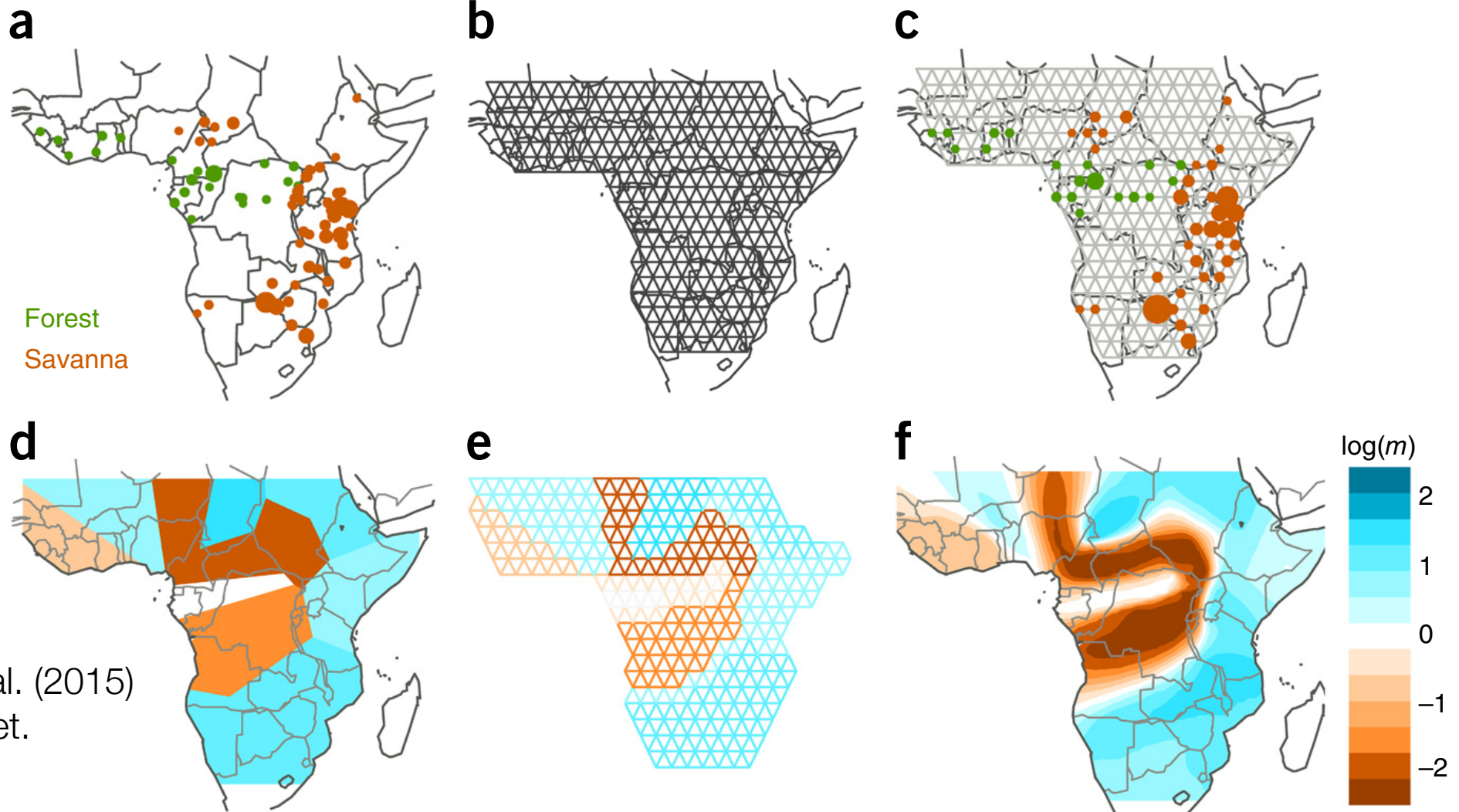


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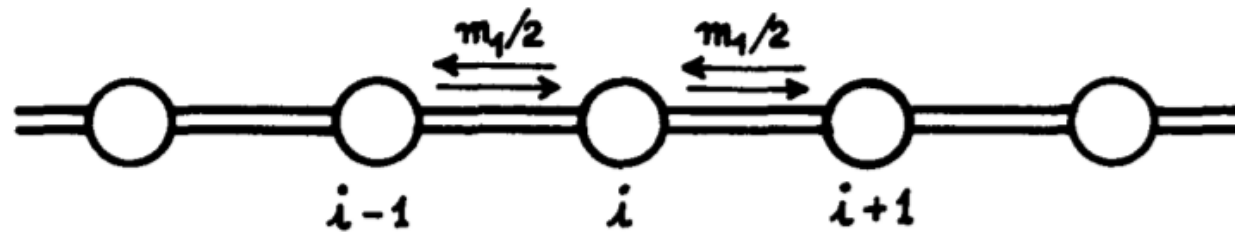


Estimated Effective Migration Surfaces (EEMS)



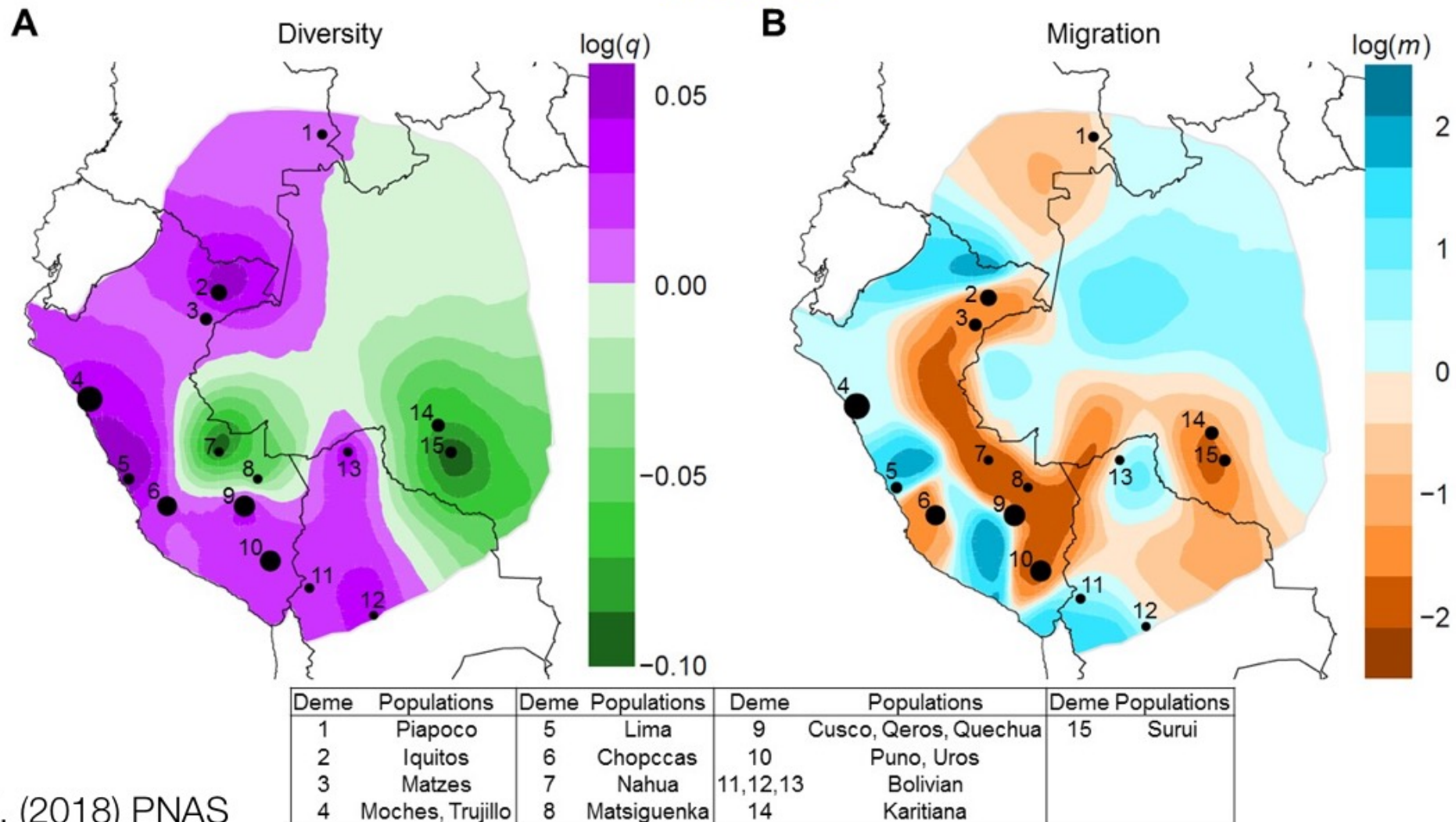
Assumptions: Stepping Stone Model

- Migration can only occur between adjacent demes
- Migration rate between each deme is assumed to be equal

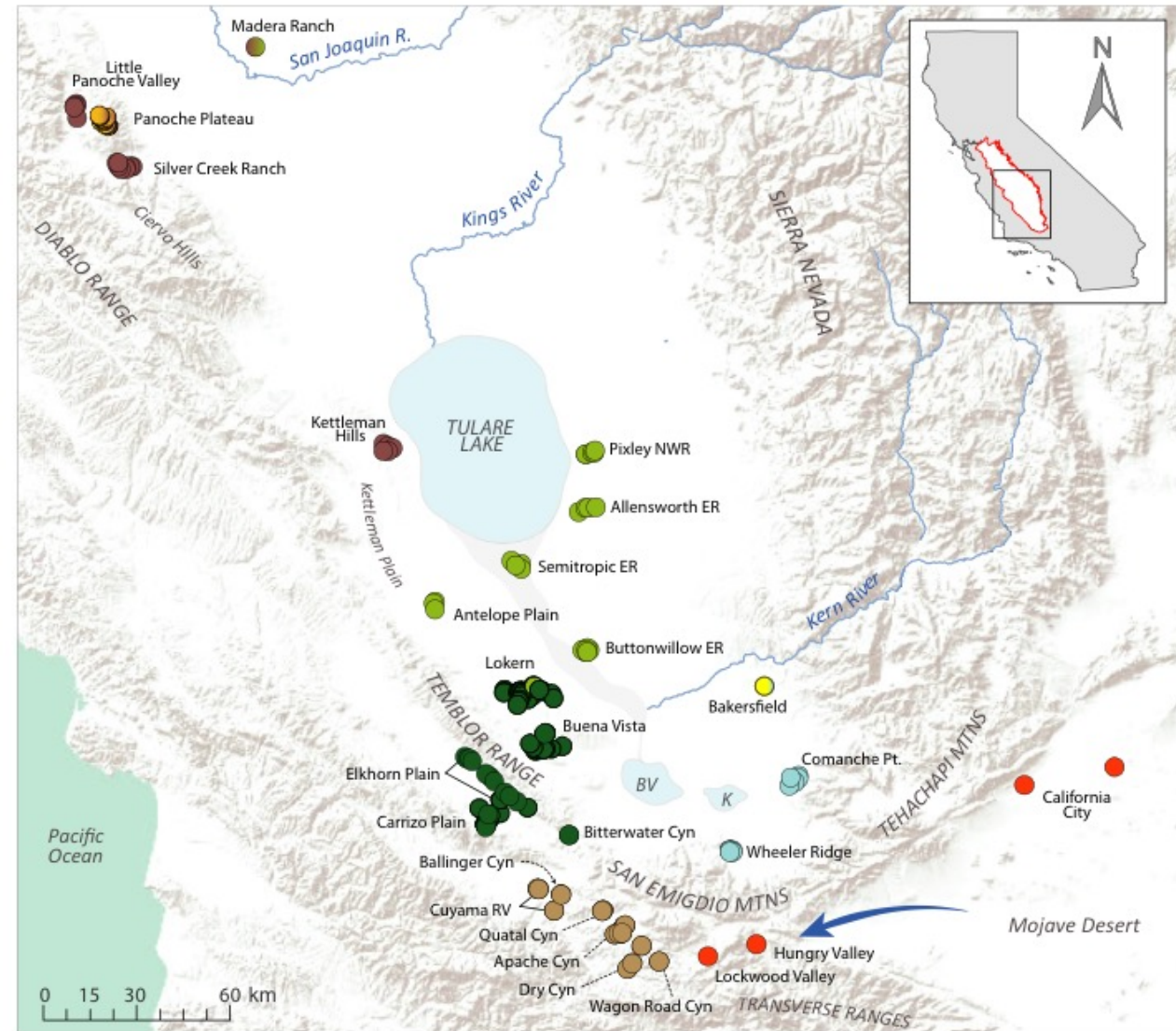
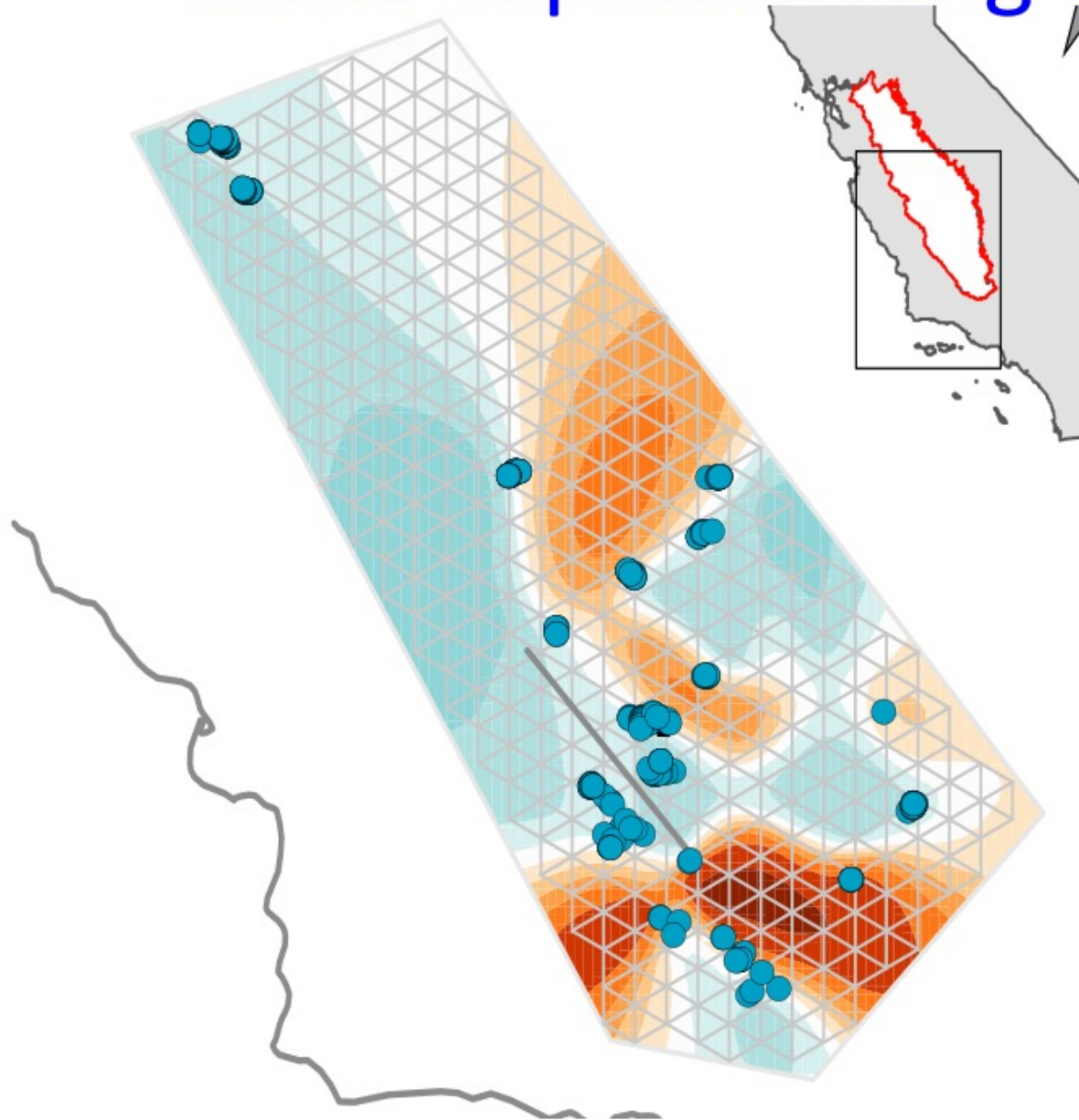


Kimura and Weiss (1964)

EEMS: Migration and diversity within Peru

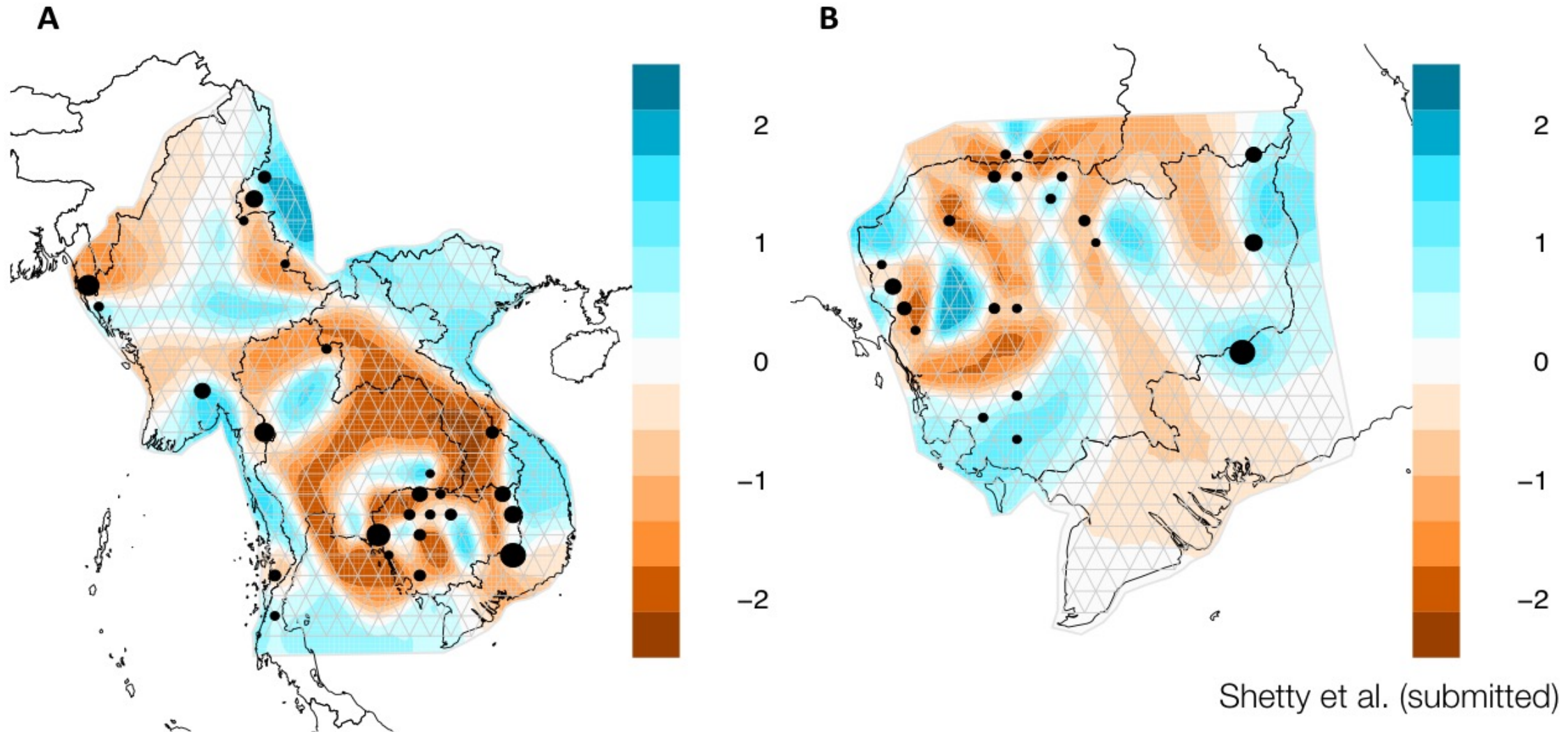


EEMS captures long-term migration patterns

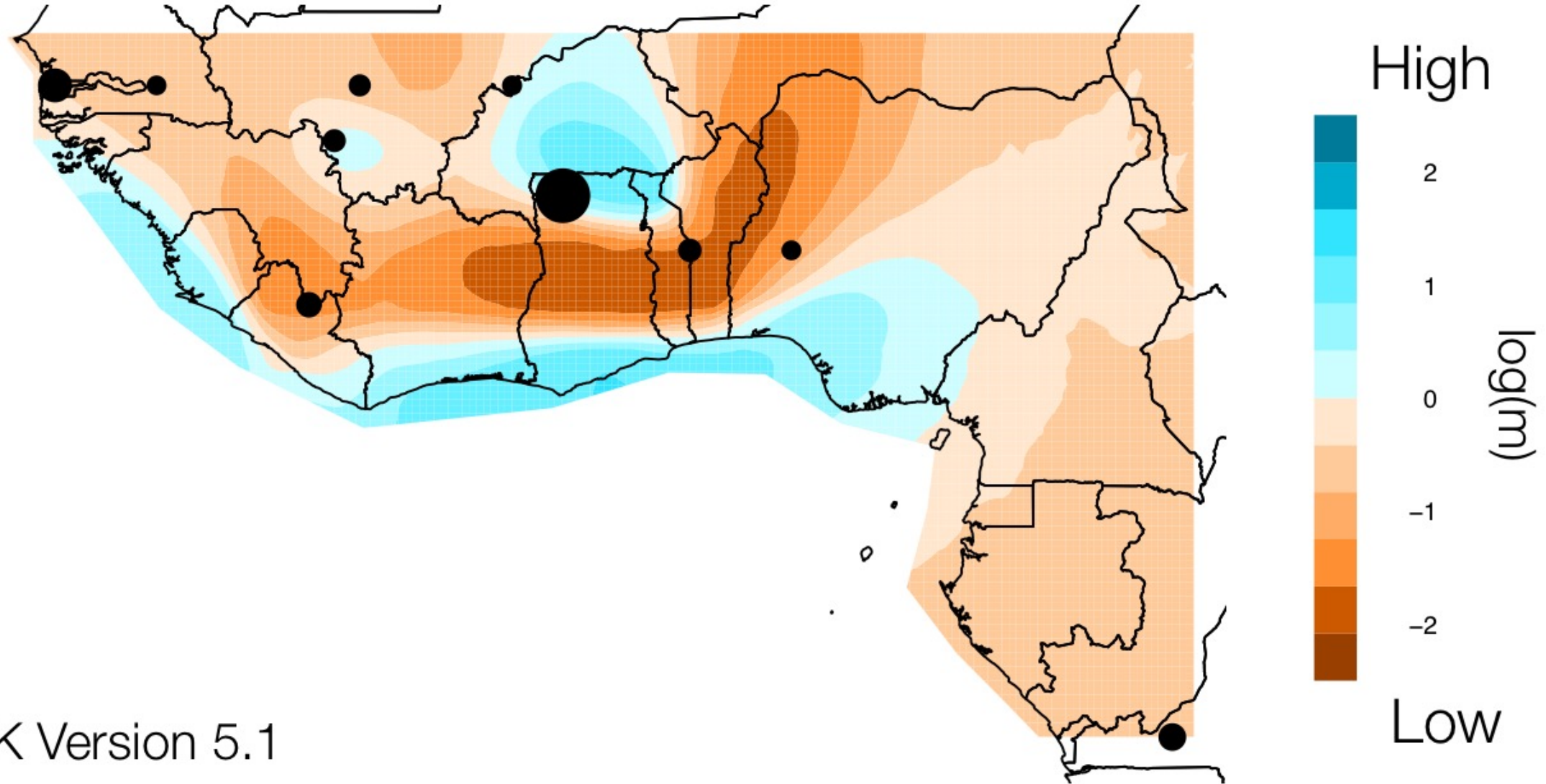


Richmond et al. (2015) Molecular Ecology

EEMS in Malaria Parasites of South East Asia

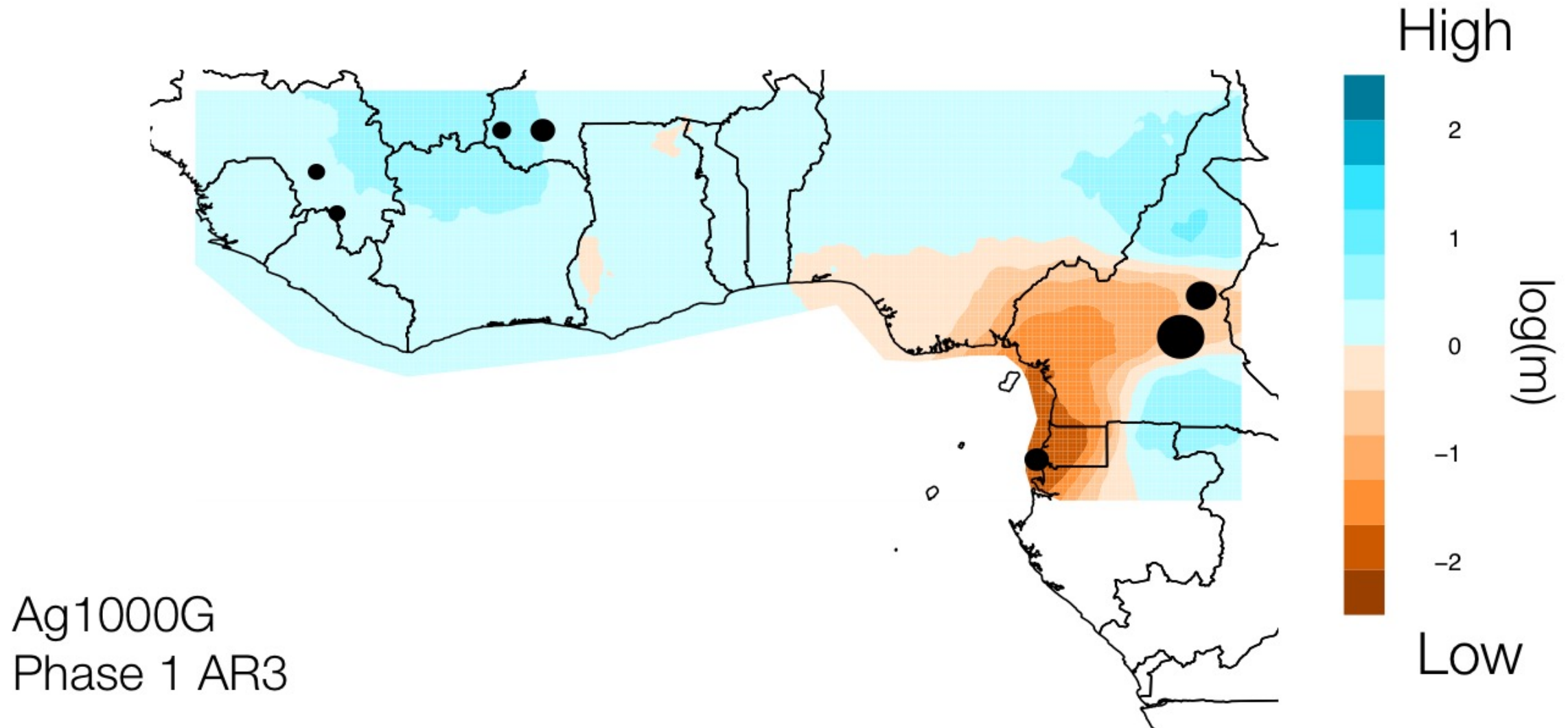


Application to Malaria Parasites in W. Africa

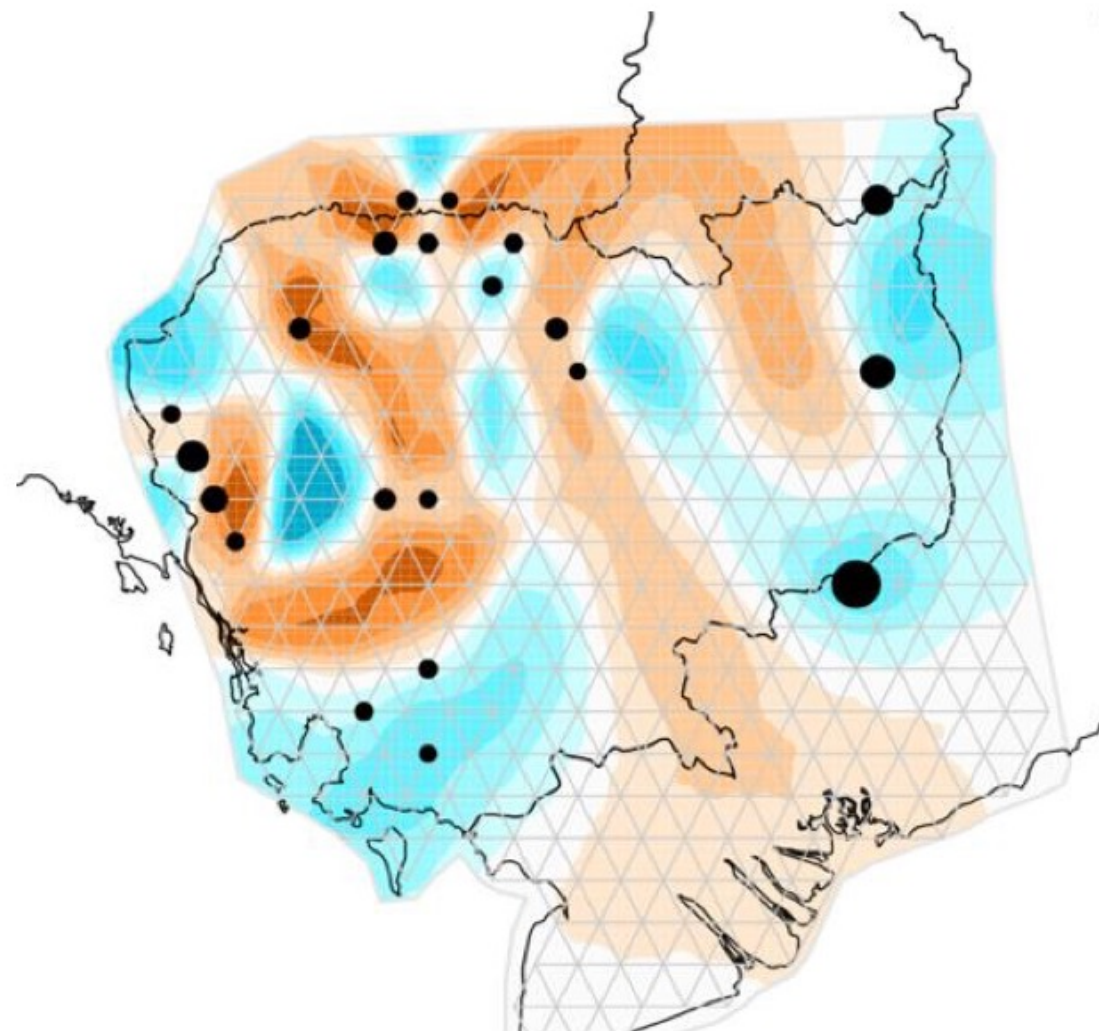
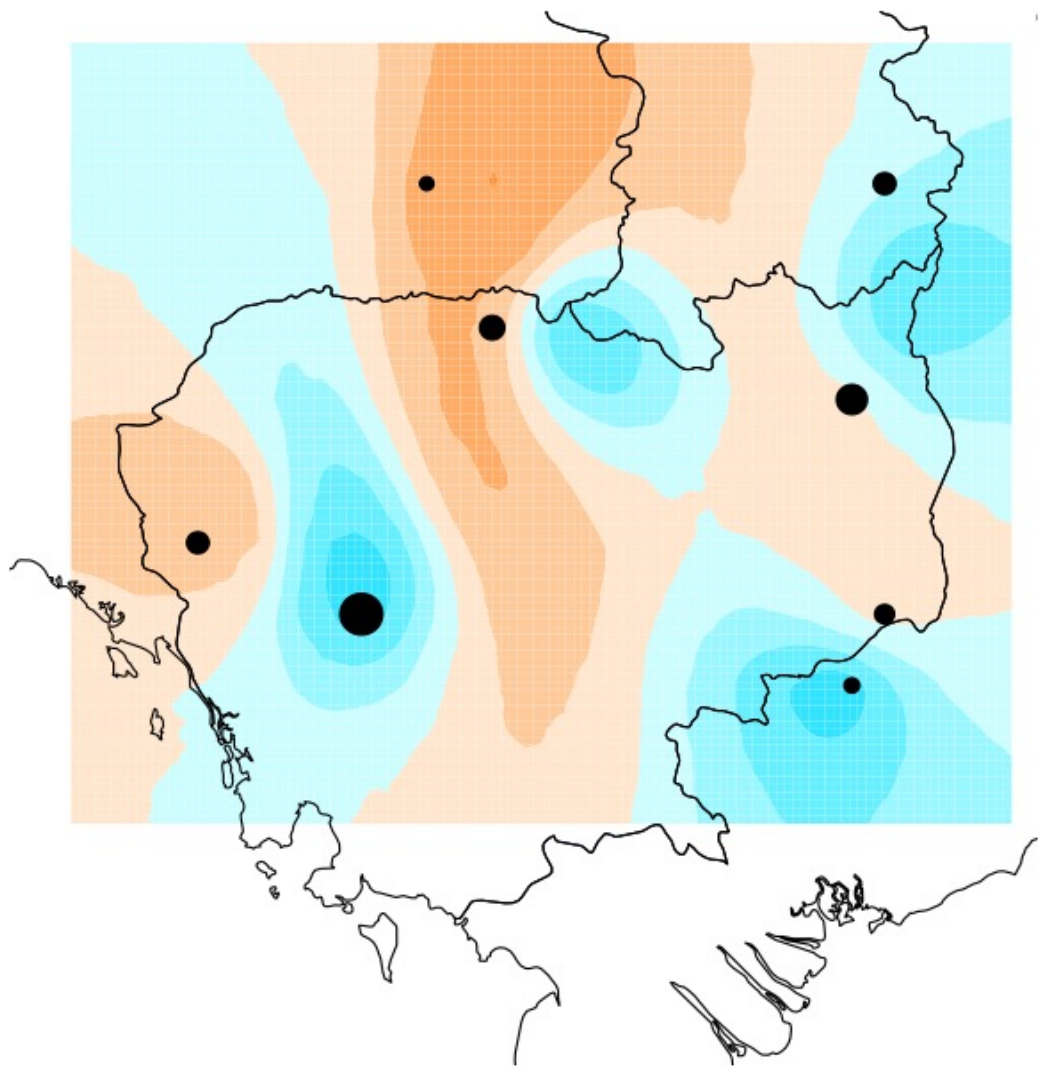


Pf3K Version 5.1

Application to Mosquito in W. Africa



Robustness of Sampling on EEMS



Concluding summary

- Fine-scale population structure is subdivisions of individuals on an ever increasingly granular scale
- Identity-by-descent and sharing of rare variants are a powerful method of identifying recent relationships and can be scaled by time.
- Cryptic population structure arises with extended relationships within a cohort, unknown to the investigators.
- EEMS can visualize migration patterns on a fine-scale illustrating cryptic structure not observed with other methods

Questions?

CHOPCCAS
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MOCHES
TRUJILLO
UROS

