

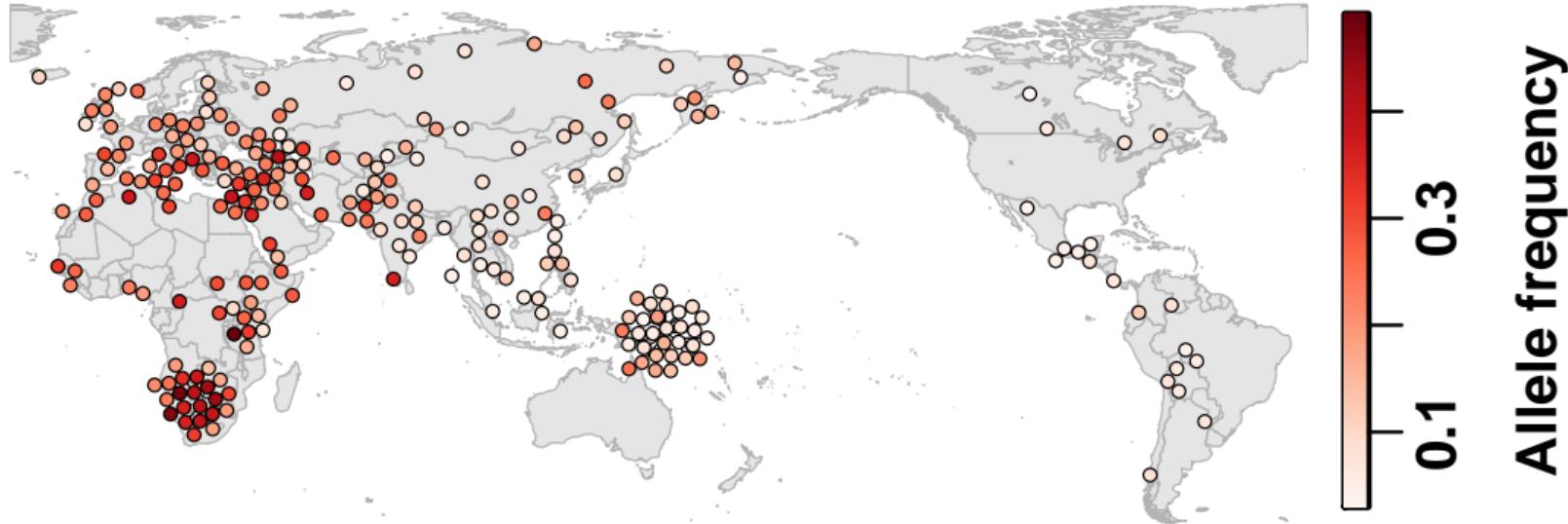
Statistical Genetics Research: Kinship, Bias, Admixture

Alejandro Ochoa

StatGen, Biostatistics & Bioinformatics — Duke University

2022-06-11 — UPGG retreat

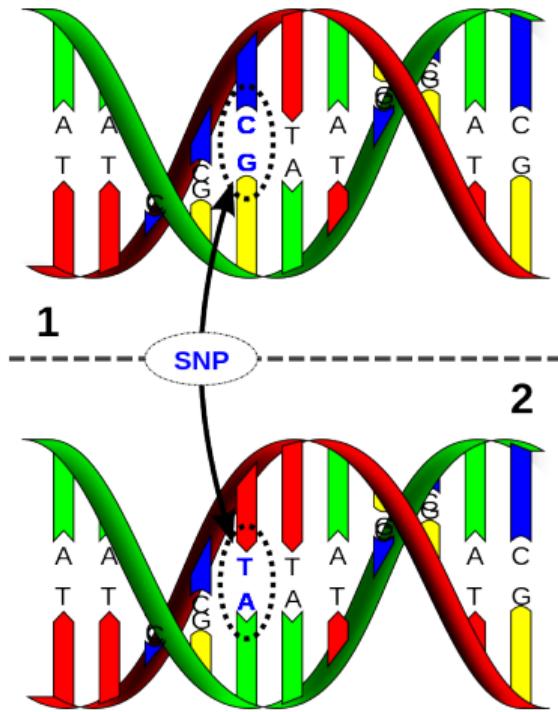
Human genetic structure



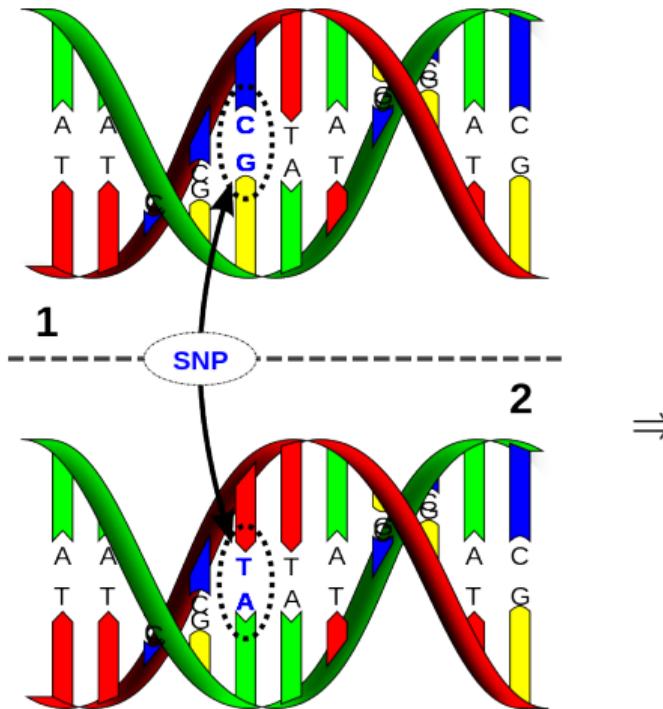
rs17110306; median differentiation given MAF $\geq 10\%$

Why? Migration and isolation, admixture, family structure

Single Nucleotide Polymorphism (SNP) data



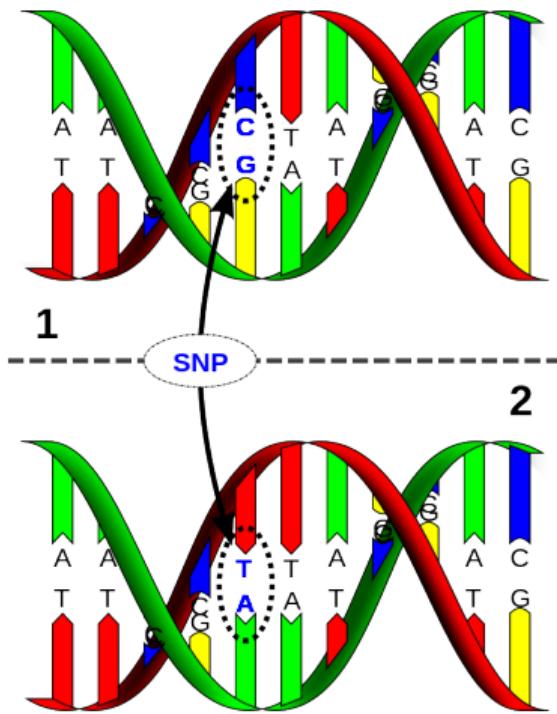
Single Nucleotide Polymorphism (SNP) data



⇒

Genotype	x_{ij}
CC	0
CT	1
TT	2

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Loci	Individuals
	0 2 2 1 1 0 1
	0 2 1 0 1
2 ...	

X

Dependence structure of genotype matrix

Individuals
Loci
0 2 2 1 1 0 1
0 2 1 0 1
2 ...

High-dimensional binomial data

- ▶ No general likelihood function
- ▶ My work: method of moments

X

Dependence structure of genotype matrix

	Individuals						
Loci	0	2	2	1	1	0	1
	0	2	1	0	1		
	2	...					

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High-dimensional binomial data

- ▶ No general likelihood function
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Relatedness / Population structure

- ### ► Dependence between individuals (columns)

Dependence structure of genotype matrix

Individuals	
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0	2 2 1 1 0 1
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High-dimensional binomial data

- ▶ No general likelihood function
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Relatedness / Population structure

- ▶ Dependence between individuals (columns)

Linkage disequilibrium

- ▶ Dependence between loci (rows)

New kinship/GRM estimator for general relatedness

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Kinship model for neutral genotypes $x_{ij} \in \{0, 1, 2\}$:

$$\text{E}[x_{ij}] = 2p_i, \quad \text{Cov}(x_{ij}, x_{ik}) = 4p_i(1 - p_i)\varphi_{jk}.$$

New kinship/GRM estimator for general relatedness

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Standard estimator is **biased**:

$$\hat{p}_i = \frac{1}{2n} \sum_{j=1}^n x_{ij}, \quad \hat{\varphi}_{jk}^{\text{std}} = \frac{1}{m} \sum_{i=1}^m \frac{(x_{ij} - 2\hat{p}_i)(x_{ik} - 2\hat{p}_i)}{4\hat{p}_i(1-\hat{p}_i)} \approx \frac{\varphi_{jk} - \bar{\varphi}_j - \bar{\varphi}_k + \bar{\varphi}}{1 - \bar{\varphi}}.$$

New kinship/GRM estimator for general relatedness

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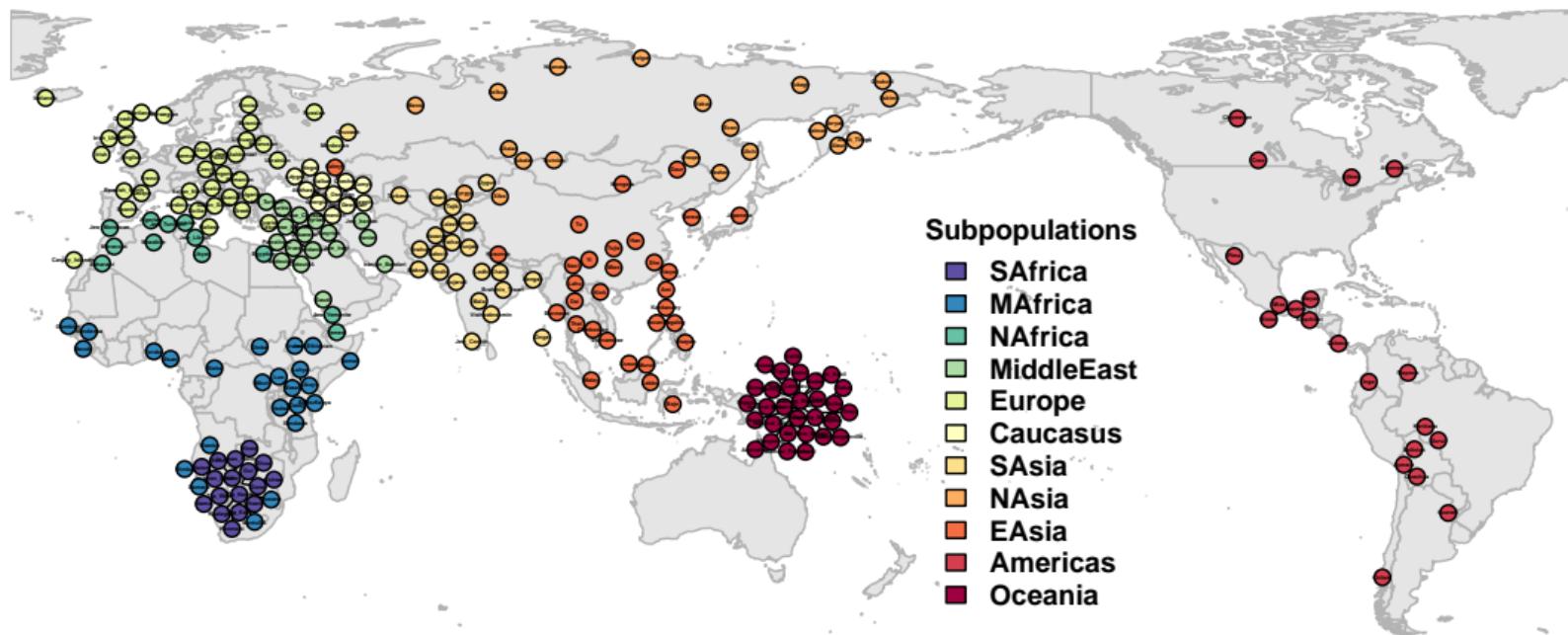
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`popkin`: first unbiased kinship estimator! R package (Ochoa and Storey, 2021)

$$A_{jk} = \frac{1}{m} \sum_{i=1}^m (x_{ij} - 1)(x_{ik} - 1) - 1, \quad \hat{\varphi}_{jk}^{\text{new}} = 1 - \frac{A_{jk}}{\hat{A}_{\min}} \xrightarrow[m \rightarrow \infty]{\text{a.s.}} \varphi_{jk}.$$



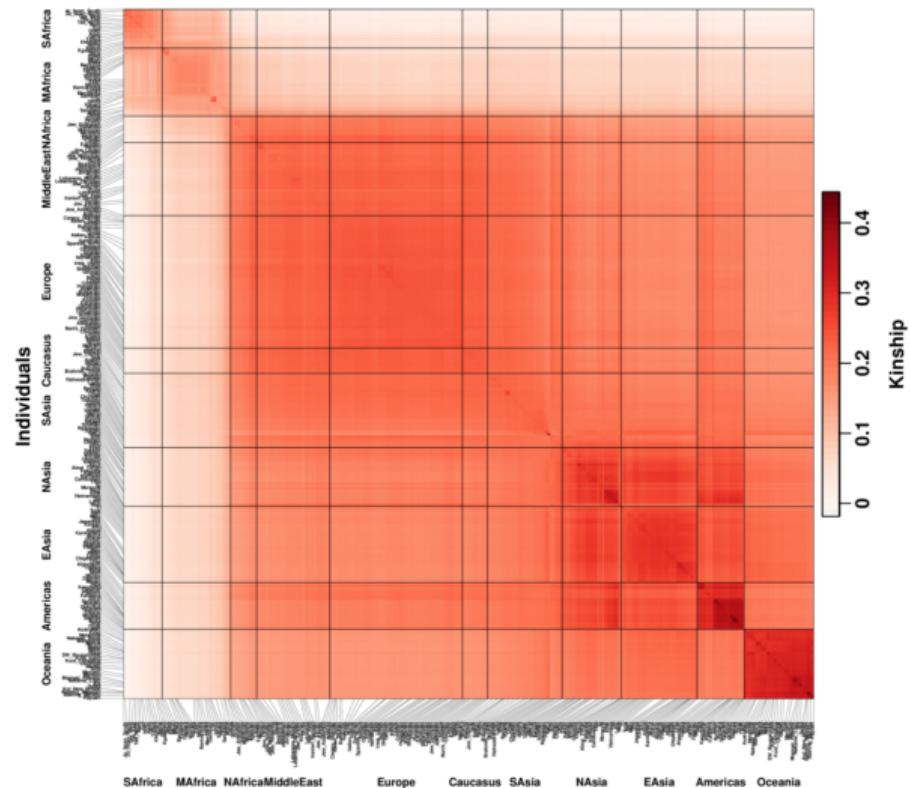
Dataset: Human Origins



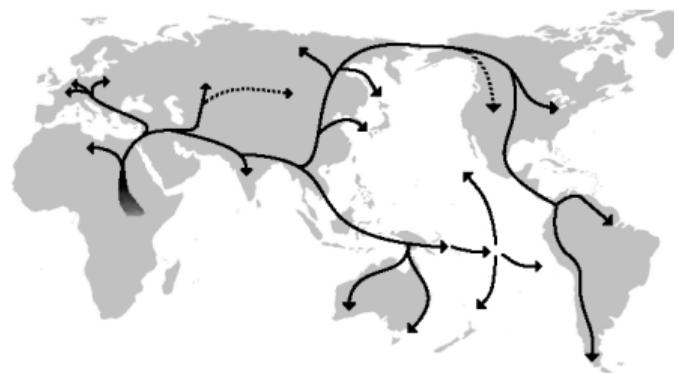
Lazaridis *et al.* (2014), (2016); Skoglund *et al.* (2016)

2,922 indivs. from 243 locs. — 588,091 loci — Array

Kinship matrix of world-wide human population

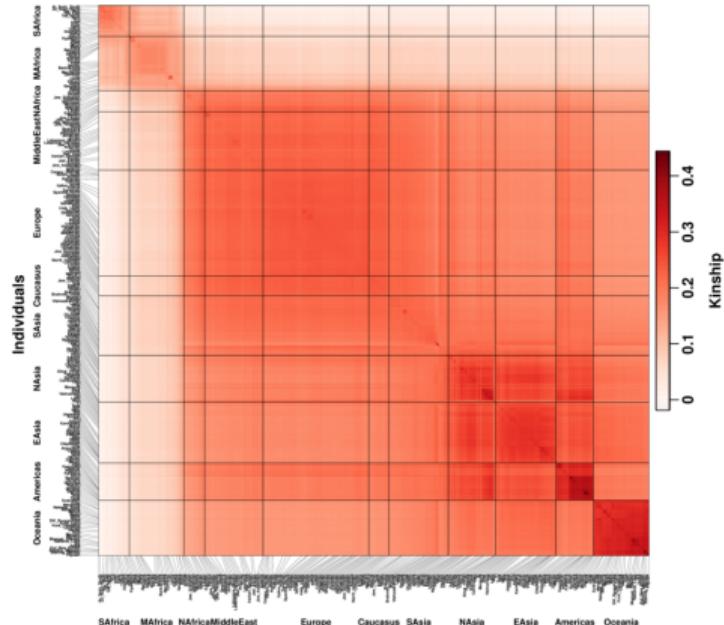


Ochoa and Storey (2019) doi:10.1101/653279

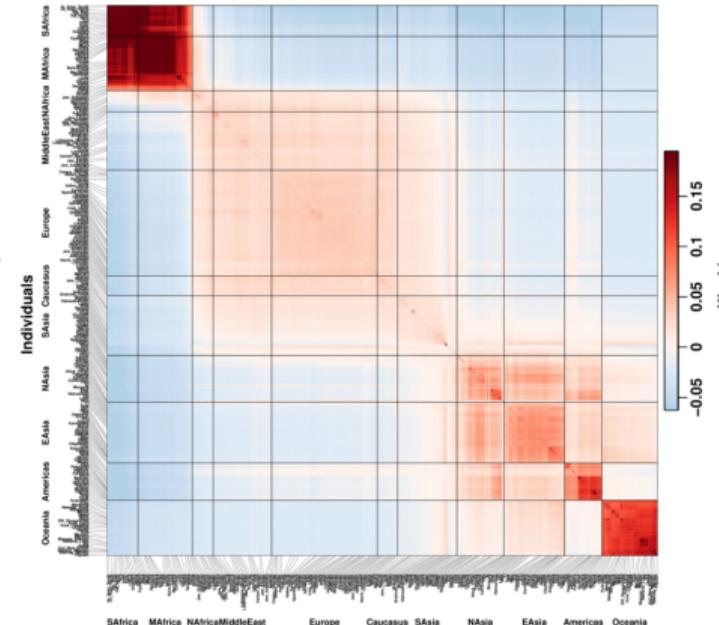


Standard kinship estimator is severely biased

New



Standard



Kinship bias: Consequences? Applications?

- ▶ Genetic association studies
- ▶ Heritability estimation
- ▶ Admixture inference

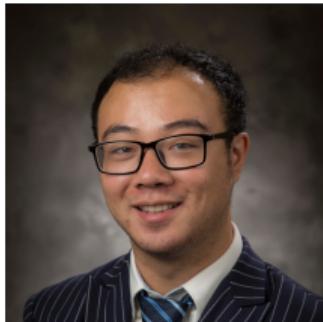
PCA vs LMM in association



Yiqi Yao
MB 2020

BenHealth
Shanghai

PCA vs LMM in association



Association with Principal Components Analysis (PCA)
and Linear Mixed-effects Model (LMM):

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MB 2020

BenHealth
Shanghai

PCA vs LMM in association



Association with Principal Components Analysis (PCA)
and Linear Mixed-effects Model (LMM):

$$\text{PCA : } \mathbf{y} = \mathbf{1}\alpha + \mathbf{x}_i\beta + \mathbf{U}_d\boldsymbol{\gamma}_d + \epsilon,$$

$$\text{LMM : } \mathbf{y} = \mathbf{1}\alpha + \mathbf{x}_i\beta + \mathbf{s} + \epsilon.$$

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\mathbf{U}_d are top d eigenvectors of kinship matrix Φ .
 $\mathbf{s} \sim \text{Normal}(\mathbf{0}, \sigma^2\Phi)$.

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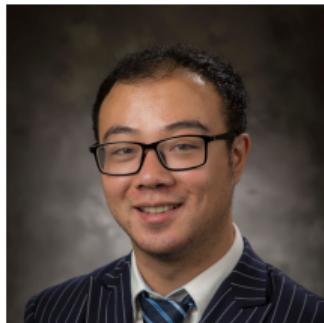
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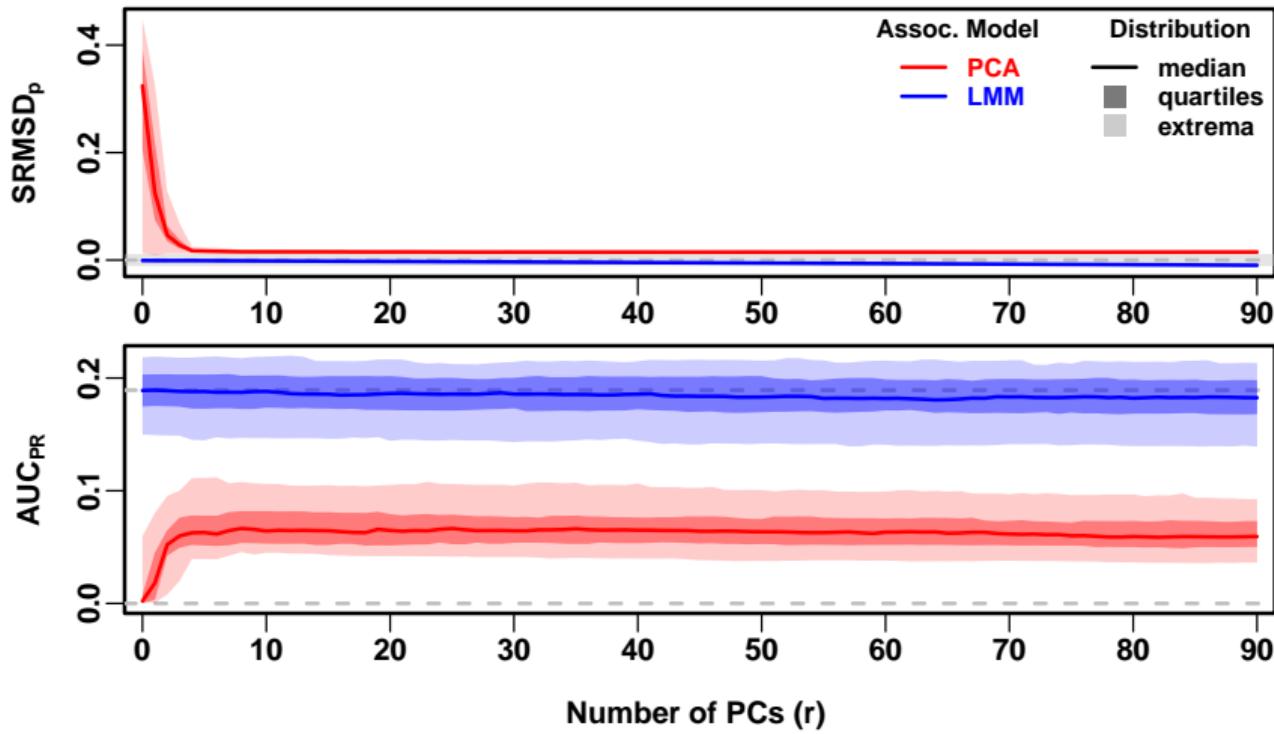
- ▶ PCA is faster but low-dimensional
- ▶ LMM is slower but can model families

PCA < LMM in association for real datasets



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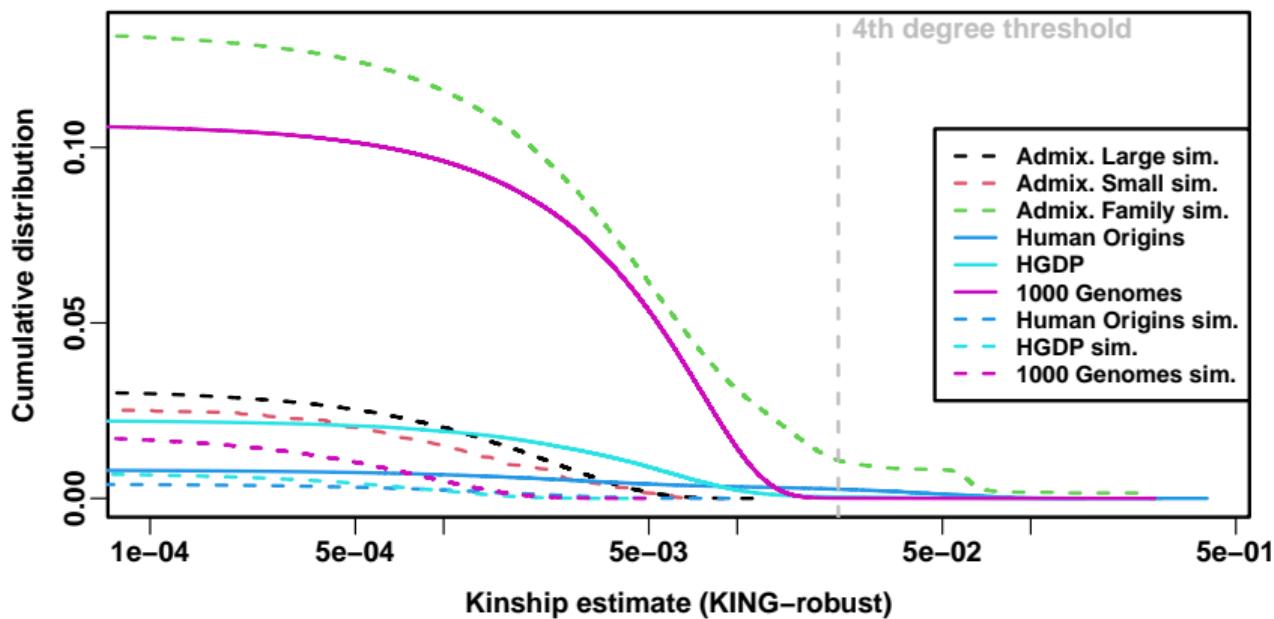
1000 Genomes Project

Numerous distant relatives in real datasets



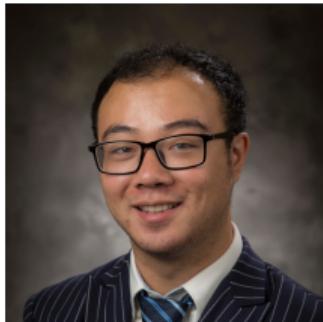
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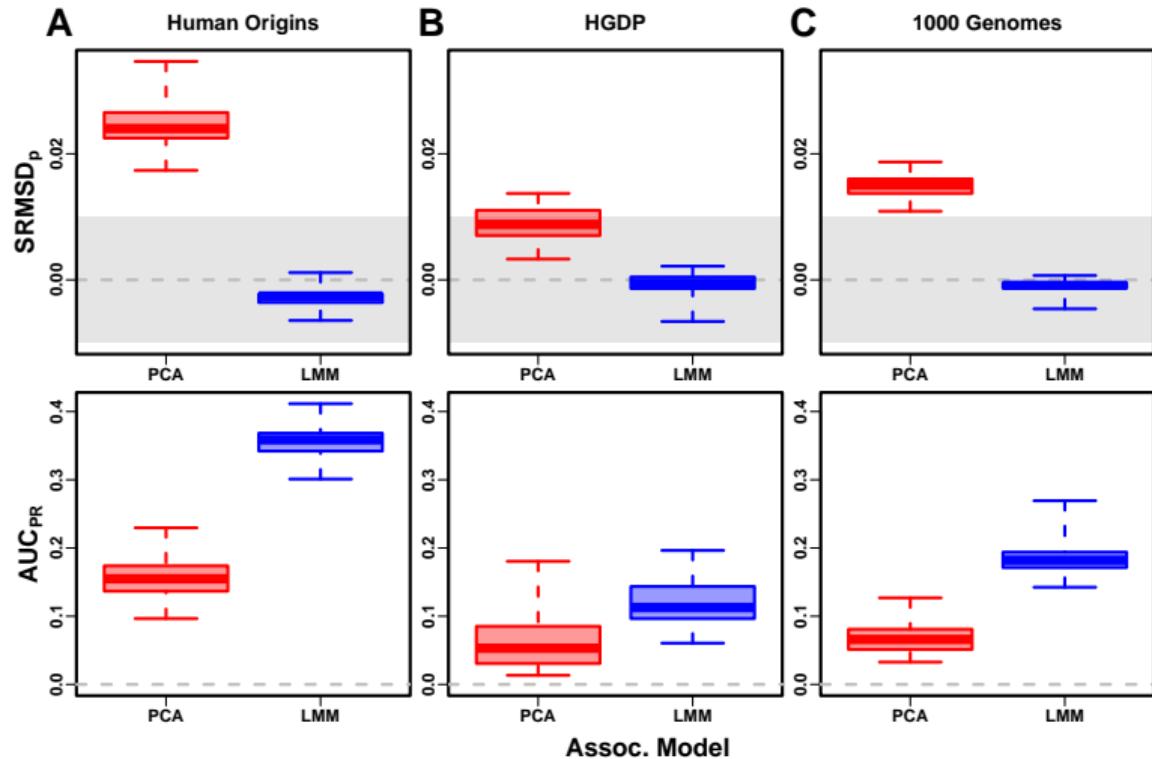
Yao and Ochoa (2022) doi:10.1101/2022.03.25.485885

Numerous distant relatives in real datasets explain PCA < LMM



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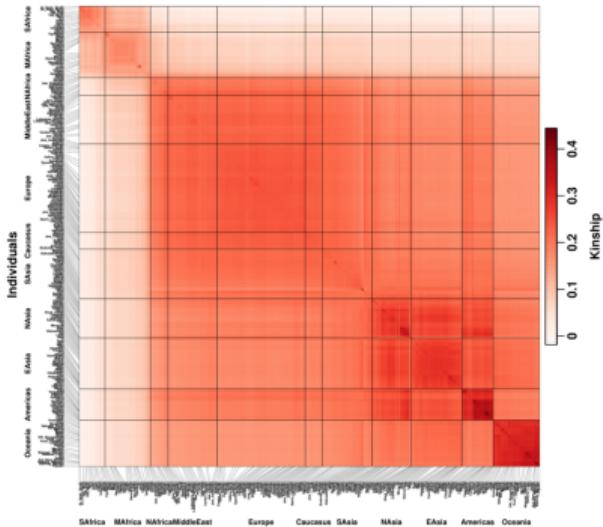


Kinship bias does not affect genetic associations

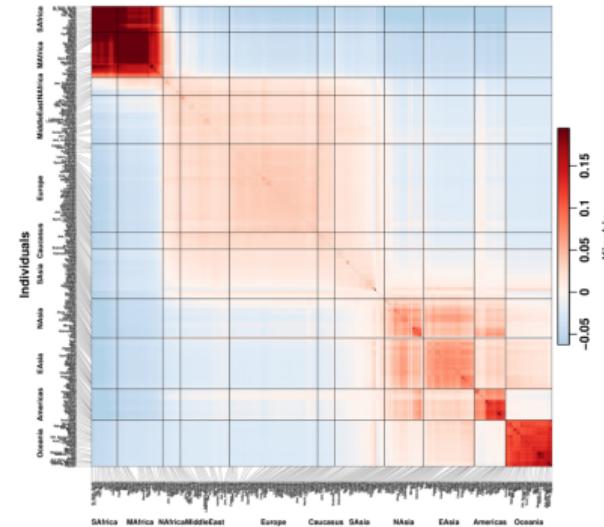


Zhuoran Hou
MB 2021

Now: B&B PhD
student



New popkin
kinship estimator



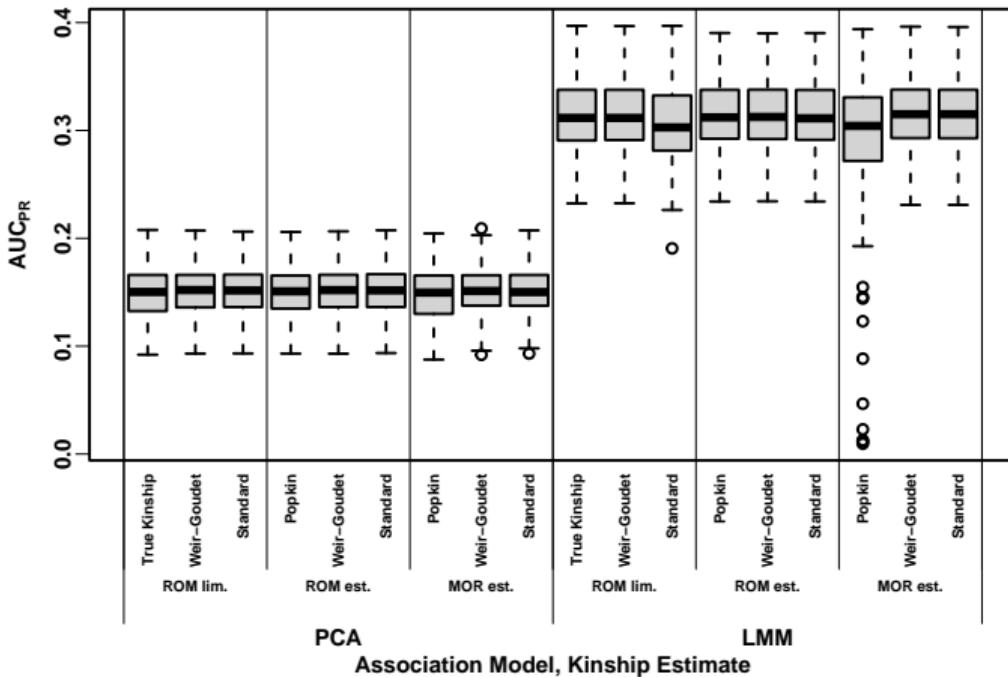
Standard
kinship estimator

Kinship bias does not affect genetic associations



Zhuoran Hou
MB 2021

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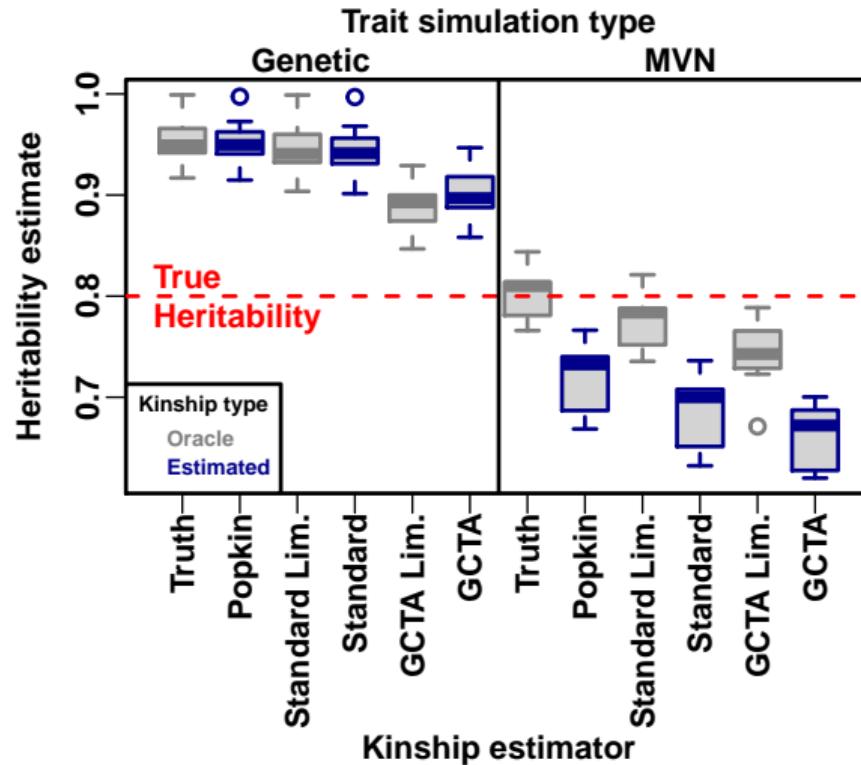
Kinship bias doesn't matter? Proved with linear algebra!

Kinship bias affects heritability estimation



Zhuoran Hou
MB 2021

Now: B&B PhD
student



Nephrotic Syndrome association study



Severe pediatric kidney disease. 1000 cases/1000 controls;
multiethnic; LMM association

Tiffany Tu
CBB PhD
student

With
Gbadegesin lab
(Pediatrics)

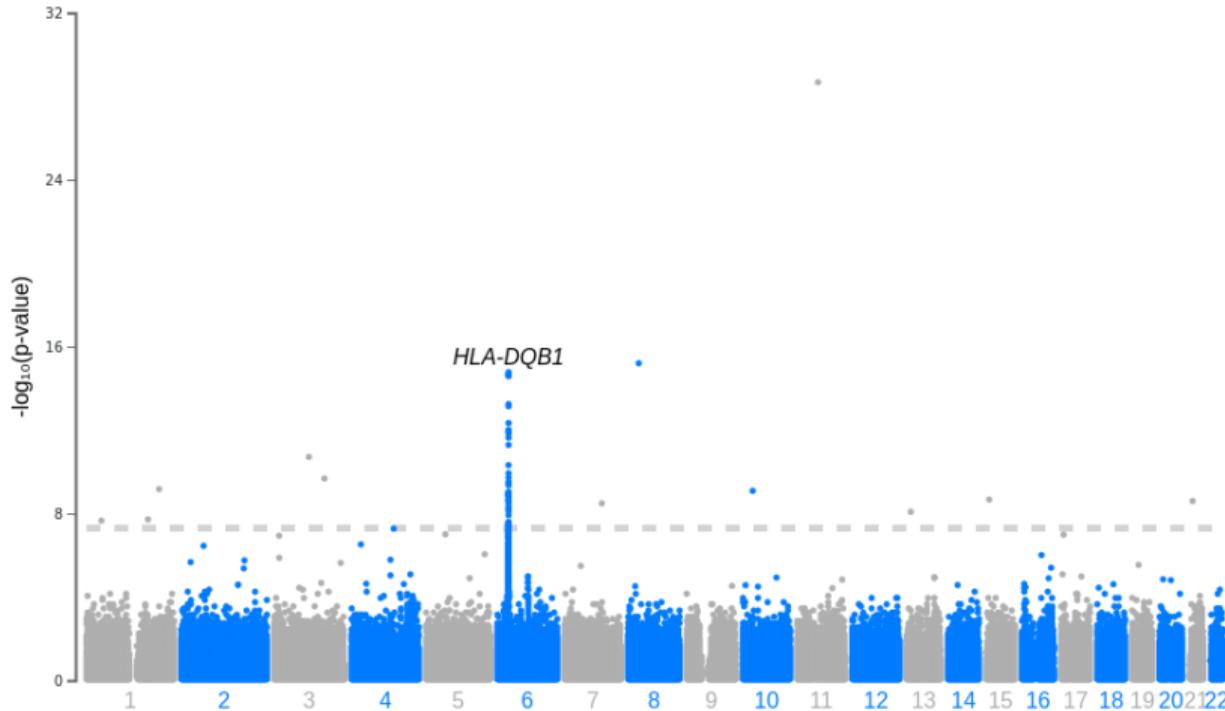
Nephrotic Syndrome association study



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Severe pediatric kidney disease. 1000 cases/1000 controls;
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LIGERA: light genetic robust association



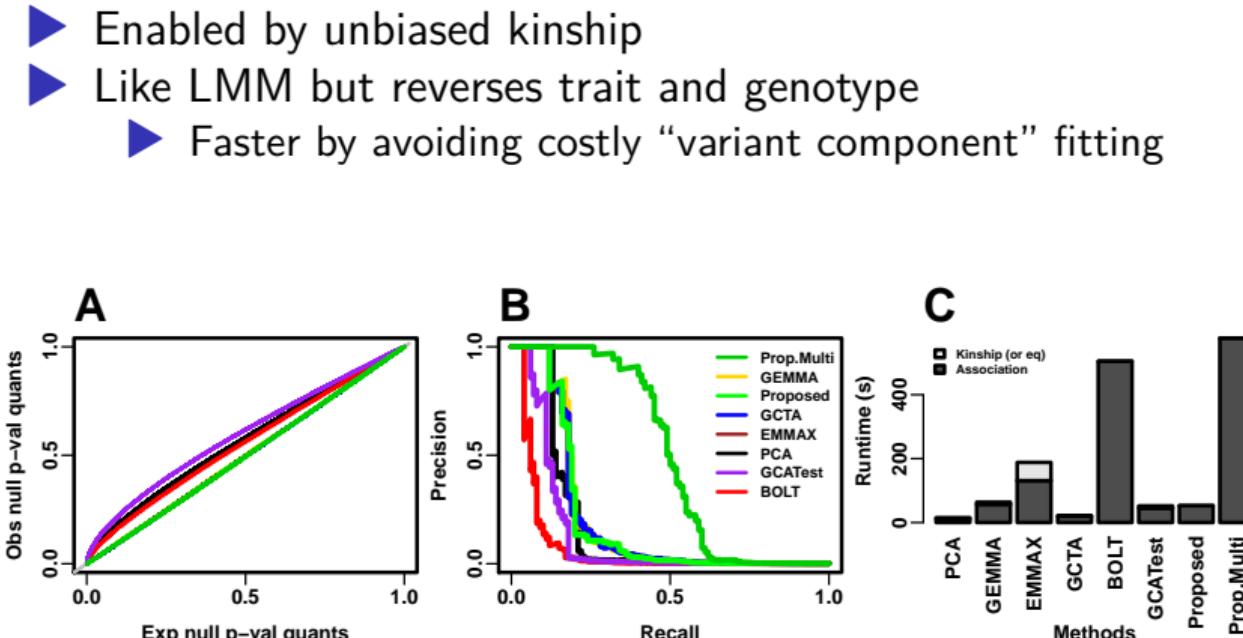
- ▶ Enabled by unbiased kinship
- ▶ Like LMM but reverses trait and genotype
 - ▶ Faster by avoiding costly “variant component” fitting

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LIGERA: light genetic robust association



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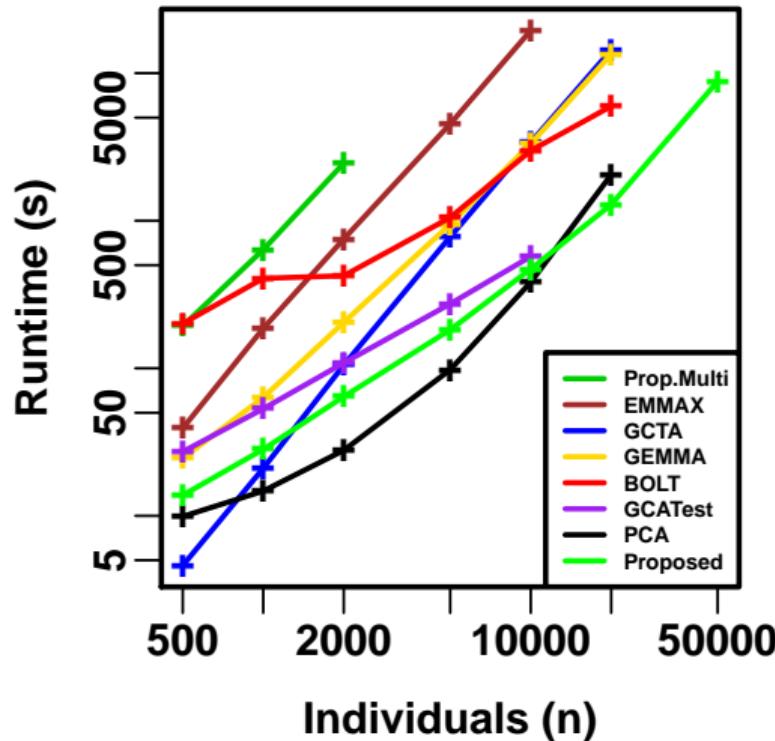


- ▶ Enabled by unbiased kinship
- ▶ Like LMM but reverses trait and genotype
 - ▶ Faster by avoiding costly “variant component” fitting
- ▶ Control of type-I error
- ▶ Increased power with multiscan: polygenic model!

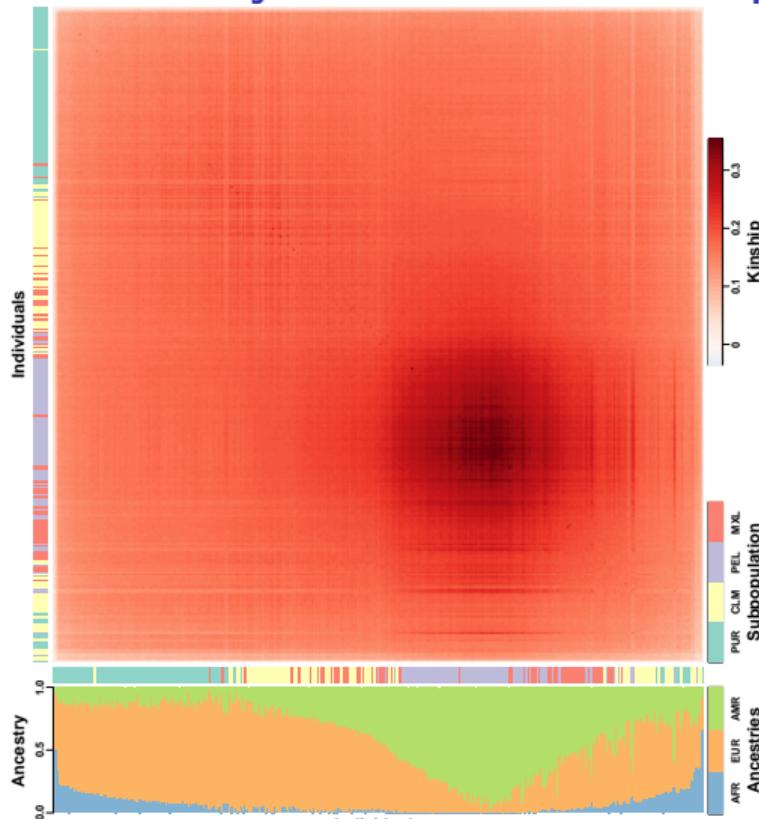
LIGERA: light genetic robust association: scalability



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student



Population kinship driven by admixture in Hispanics



Ochoa and Storey (2019b) doi:10.1101/653279

Kinship under the admixture model



Amika Sood
Postdoc

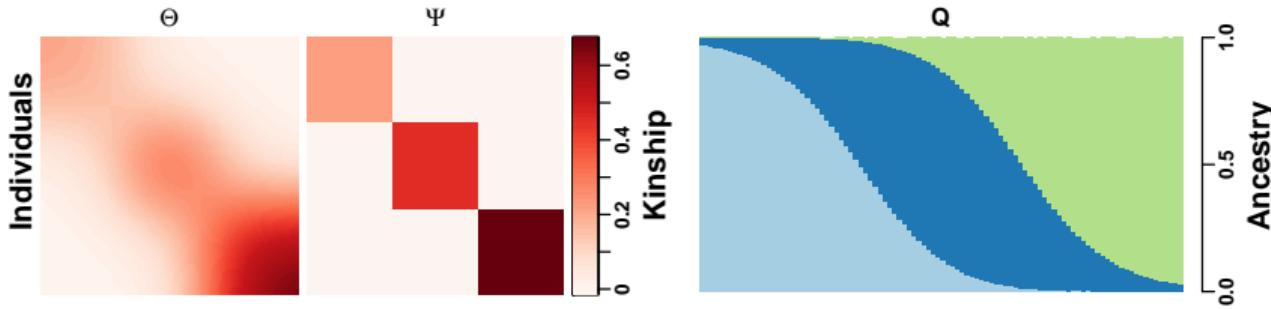
Now: Senior
Researcher
UGA

Kinship under the admixture model



Amika Sood
Postdoc

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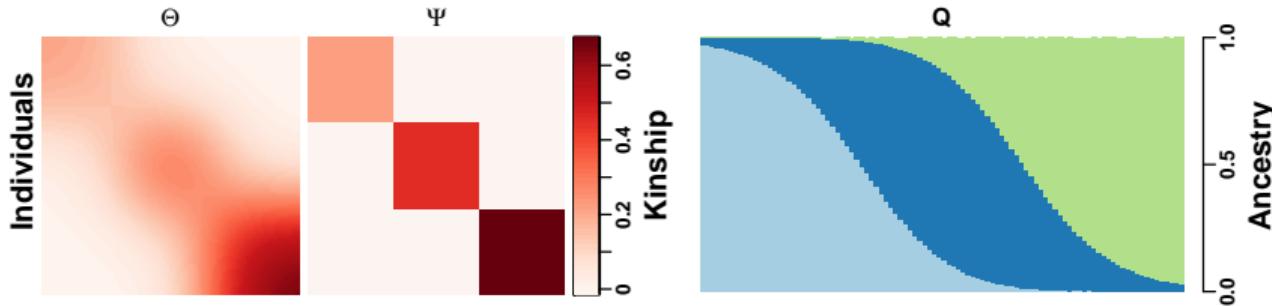


Kinship under the admixture model



Amika Sood
Postdoc

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$$\Theta = Q\Psi Q^\top$$

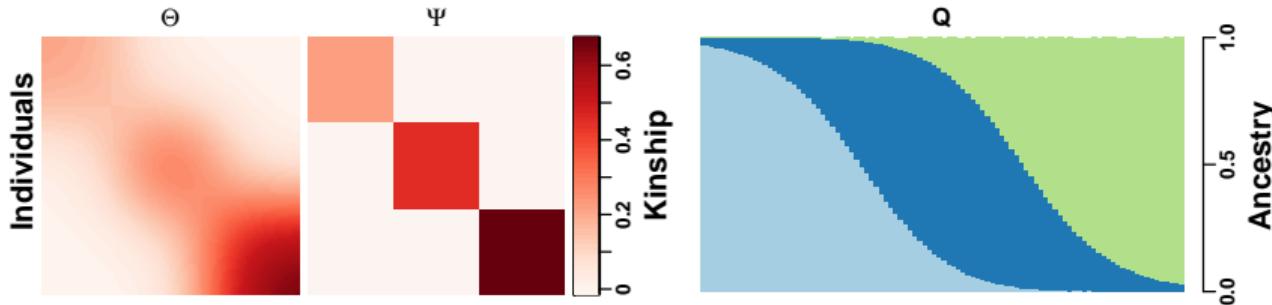
(Only for unbiased kinship)

Kinship under the admixture model



Amika Sood
Postdoc

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$$\Theta = Q\Psi Q^\top$$

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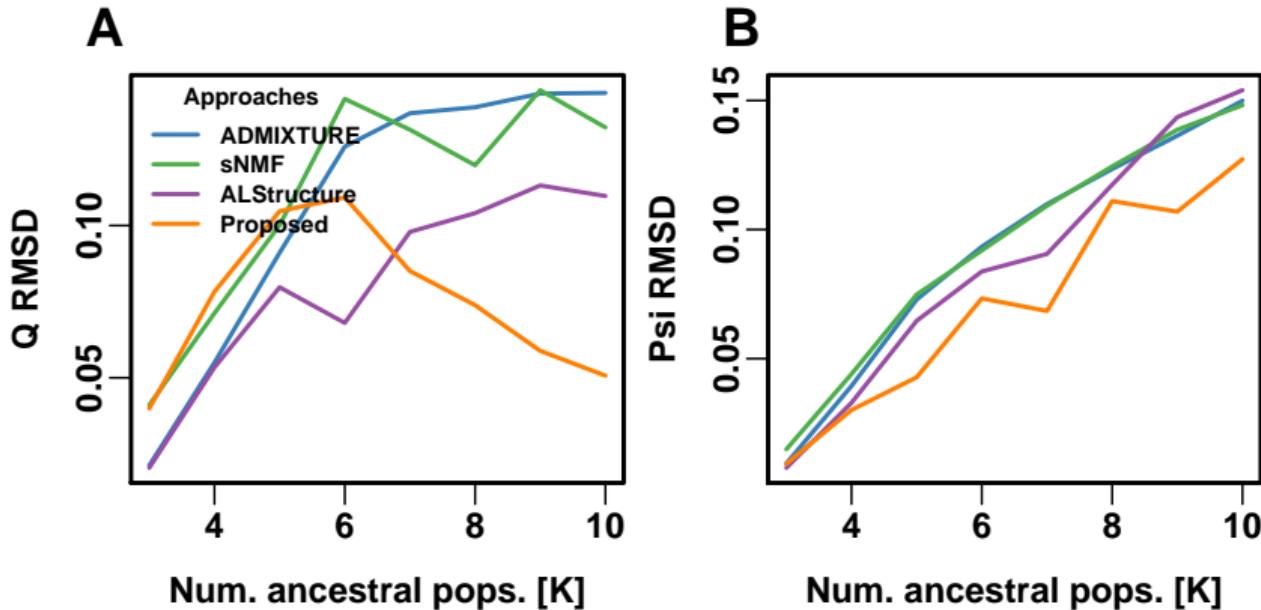
Can we reverse this formula?

AdmixCor: accuracy

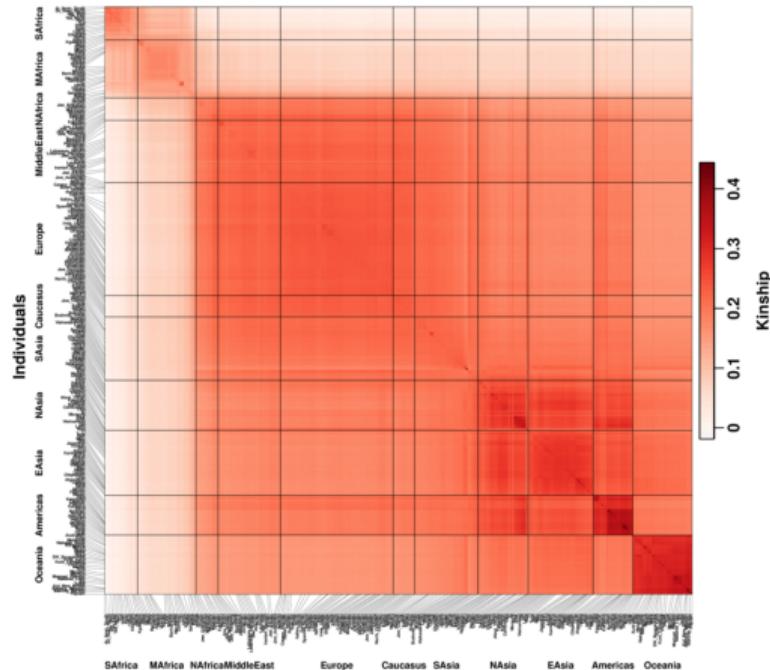


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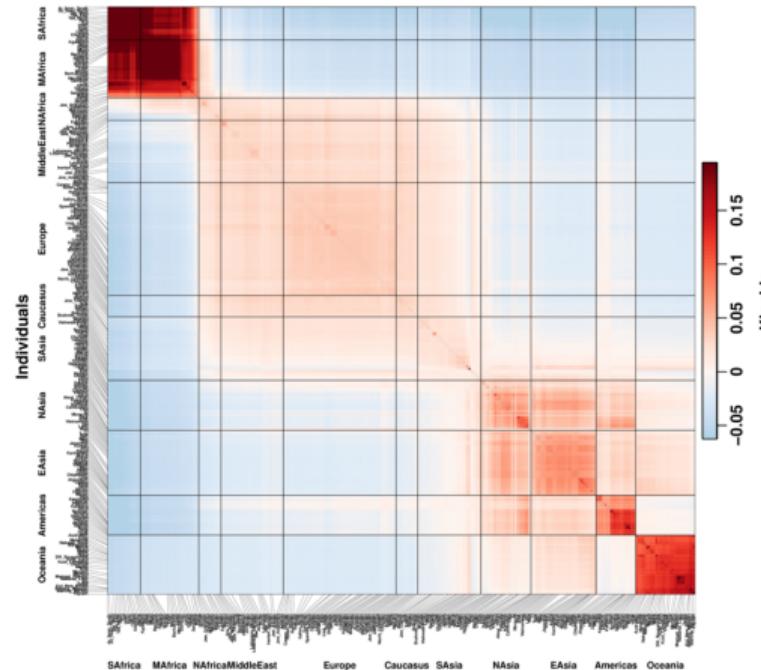
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Unbiased kinship estimates: new models, opportunities



New "popkin"
kinship estimator



Biased "standard"
kinship estimator

Acknowledgments

Ochoa Lab

Amika Sood

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