## Models for learning about history from genetics

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## How Can We Infer Geography and History from Gene Frequencies?

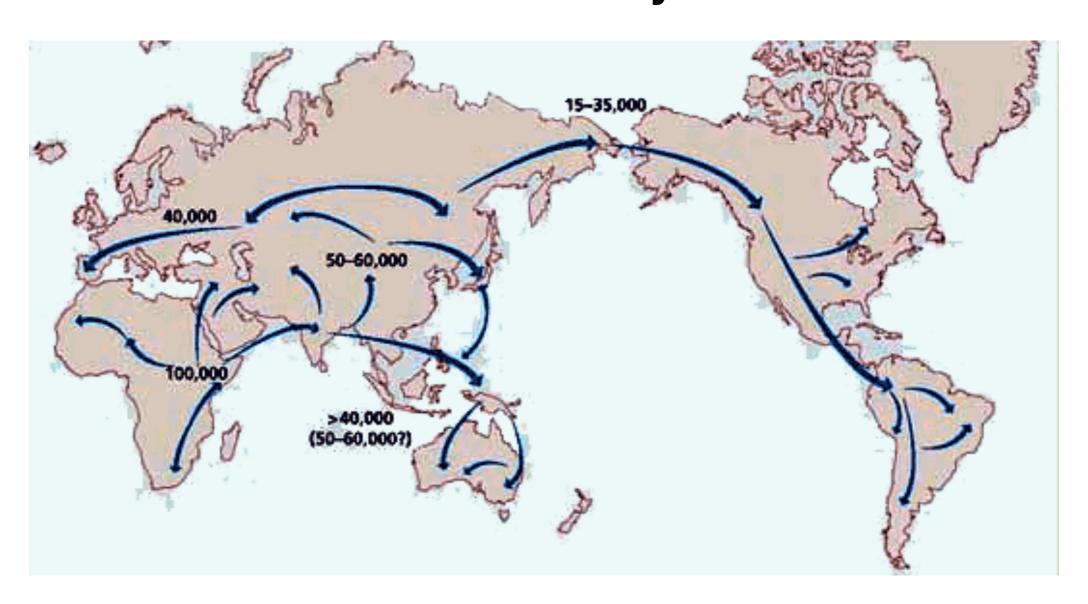
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In principle, patterns of migration and historical branching events can be reconstructed from gene frequency data, but we still lack most of the techniques necessary to do this. This is a fairly clearly defined problem with a variety of interesting subcases. It appears likely to raise interesting mathematical and statistical questions. The amount of data potentially available is very large. The problem is reviewed in this paper, but no new solutions are proposed.

# Humans expanded out of Africa to occupy nearly the entire globe over the last 60-100k years

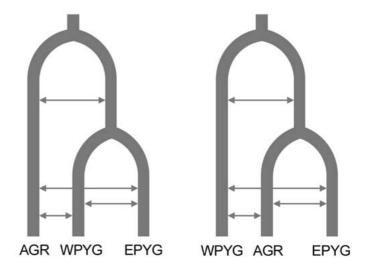


Details in this figure are are vague. How can we figure out population sizes, timings of population movements, etc?

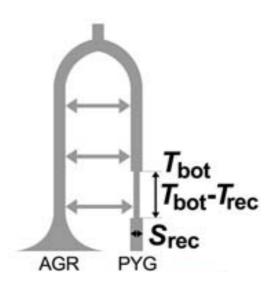
Humans are intensively studied, yet we still know only vague details. What about other species?

### What can we hope to learn?

- Topology
  - what is the branching structure of populations?



- Demography
  - when did demographic events occur?
  - what were population sizes?



# What sources of information do we have?

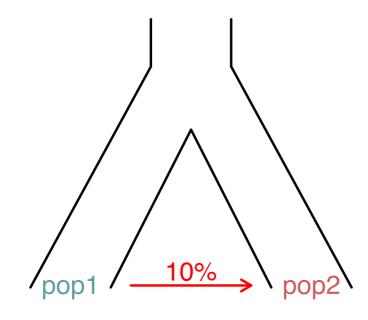
 Allele frequencies--more closely related population have more similar allele frequencies. E.g. clustering algorithms (STRUCTURE/PCA), tree-building algorithms

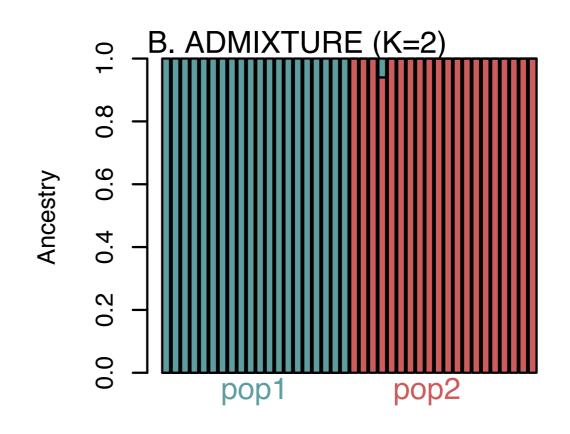
Linkage disequilibrium--influenced by mixture between populations.
 E.g. local ancestry inference, ROLLOFF

 Mutations--shared rare mutations between populations indicate shared history. E.g. mtDNA trees (not going to cover this)

# Caution: all methods can be misinterpreted

A. Simulated demography

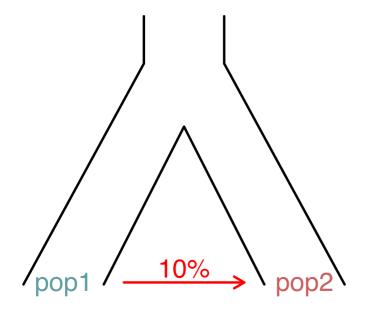


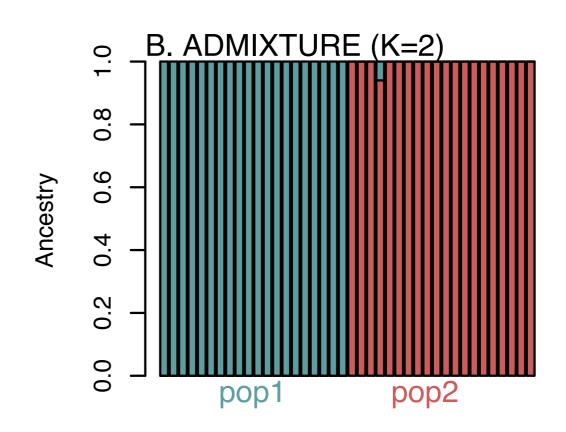


Why does ADMIXTURE not identify admixture?

# Caution: all methods can be misinterpreted

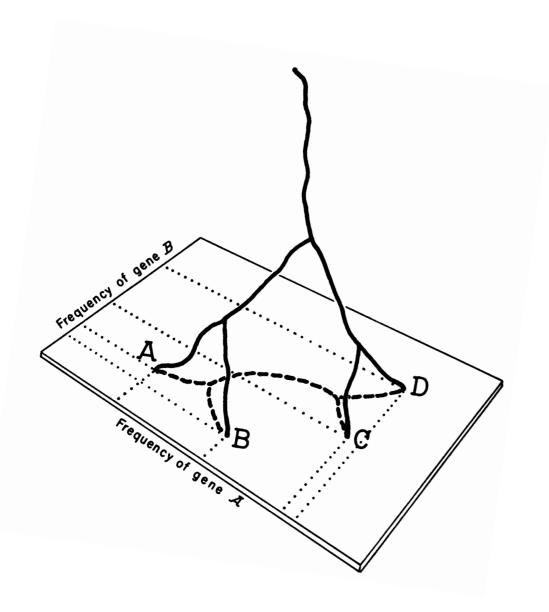
A. Simulated demography





 Need somewhat-more model based methods for inferring/testing tree topologies

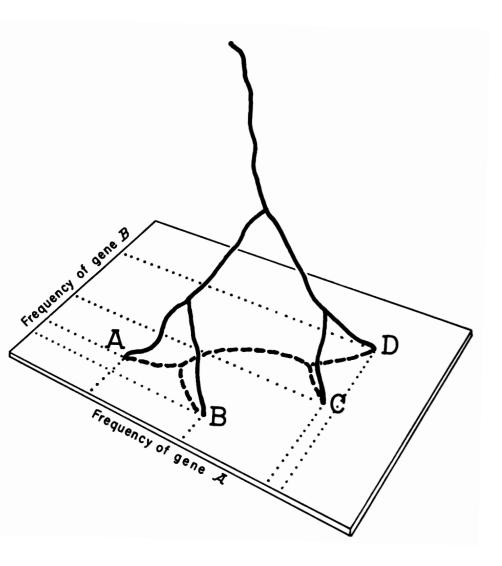
# Inferring and testing topologies



Cavalli-Sforza and Edwards (1967)
 point out that allele frequencies drift
 randomly, might be used to
 reconstruct population trees

• What type of model?

#### Cavalli-Sforza and Edwards (1967)



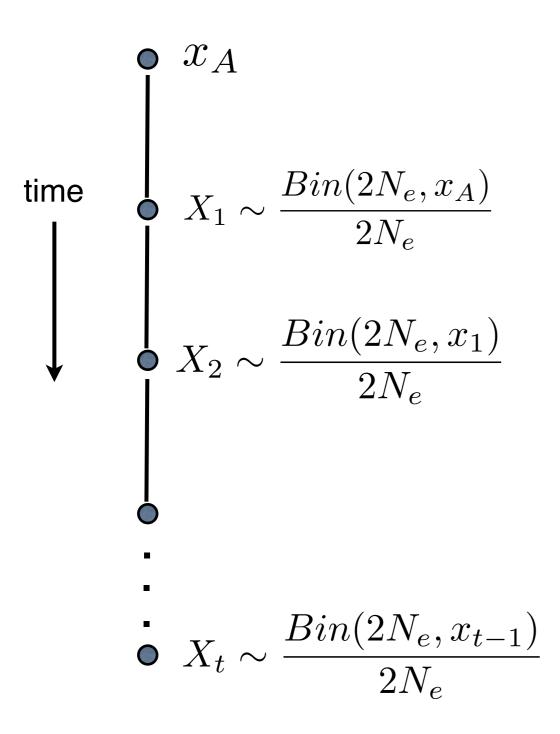
 The first proposal for treating inference of population phylogenies as a statistical problem

Consider an allele with frequency x

What is its frequency t generations later?

 Can be written down analytically (Kimura 1955), but an approximation is useful

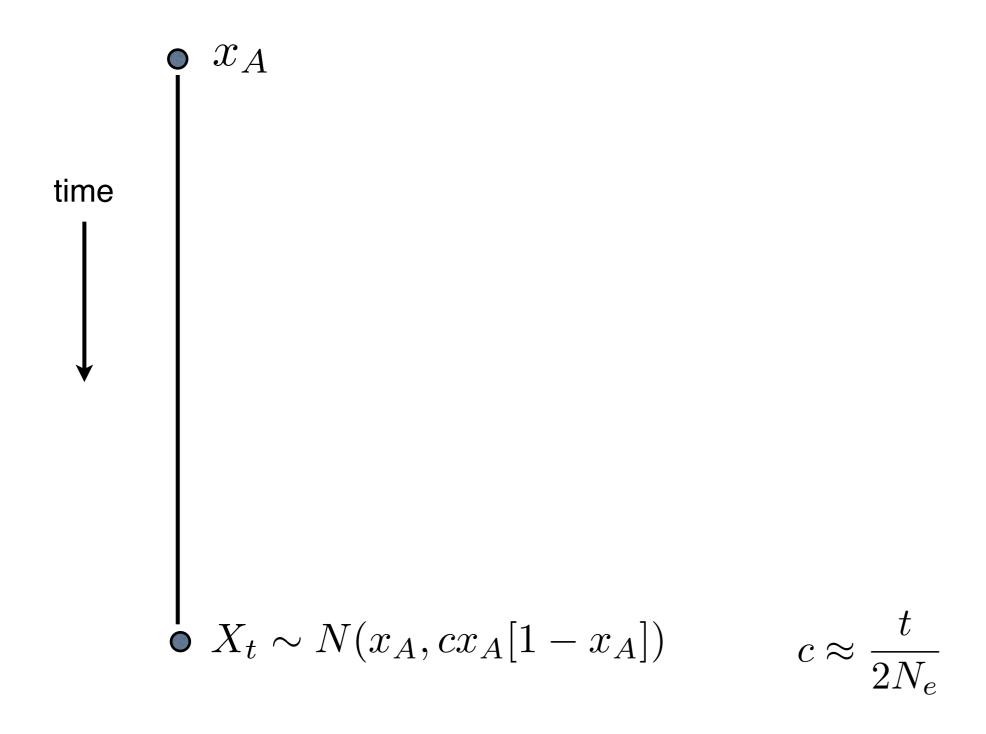
#### Normal approximation to drift



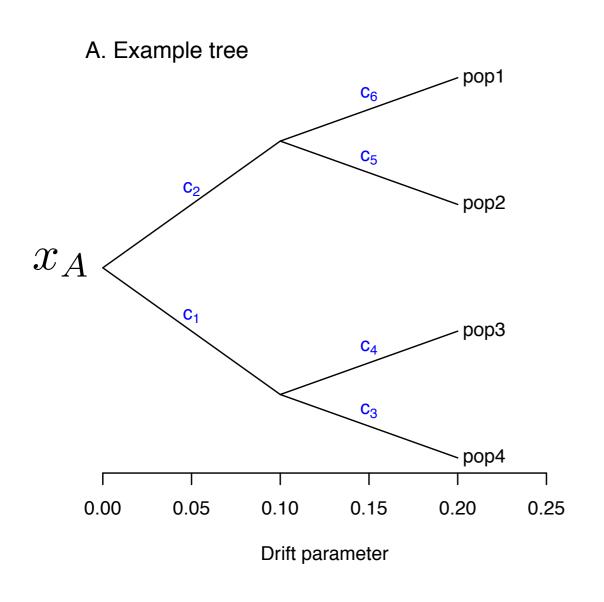
#### Normal approximation to drift

time 
$$X_1 \sim N\left(x_A, \frac{x_A[1-x_A]}{2N_e}\right)$$
 
$$X_2 \sim N\left(x_1, \frac{x_1[1-x_1]}{2N_e}\right)$$
 
$$\vdots$$
 
$$X_t \sim N\left(x_{t-1}, \frac{x_{t-1}[1-x_{t-1}]}{2N_e}\right)$$

#### Normal approximation to drift



## Natural way of learning about tree structure from allele frequencies

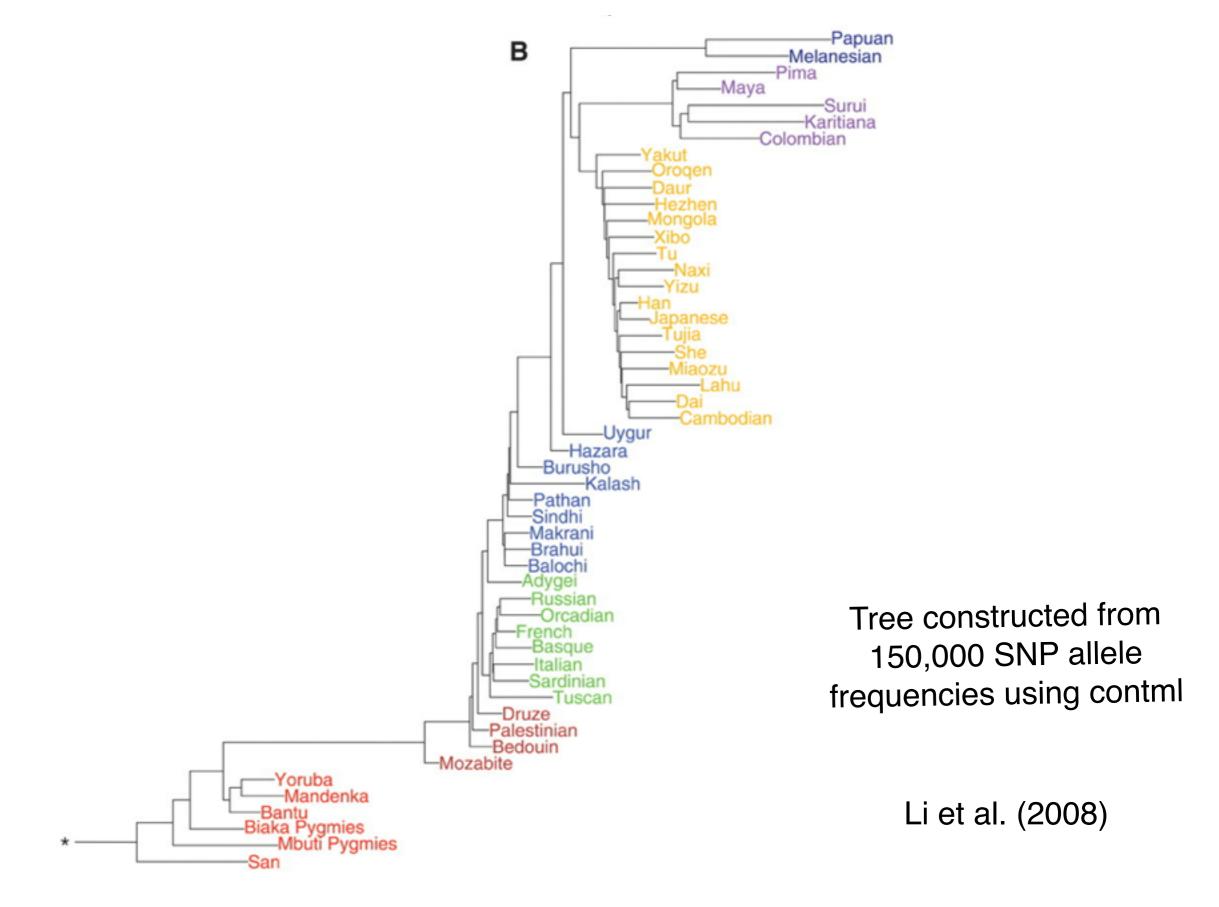


B. Covariance matrix for example tree

| pop1 | C <sub>6</sub> + C <sub>2</sub> | C <sub>2</sub>                    | 0                               | 0                               |  |
|------|---------------------------------|-----------------------------------|---------------------------------|---------------------------------|--|
| pop2 | <b>c</b> <sub>2</sub>           | C <sub>5</sub> + C <sub>2</sub> 0 |                                 | 0                               |  |
| pop3 | 0                               | 0                                 | C <sub>4</sub> + C <sub>1</sub> | C <sub>1</sub>                  |  |
| pop4 | 0                               | 0                                 | <b>c</b> <sub>2</sub>           | C <sub>3</sub> + C <sub>1</sub> |  |
|      | pop1                            | pop2                              | pop3                            | pop4                            |  |

$$[X_1, X_2, ...] \sim MVN(x_A, V)$$

## Application to humans



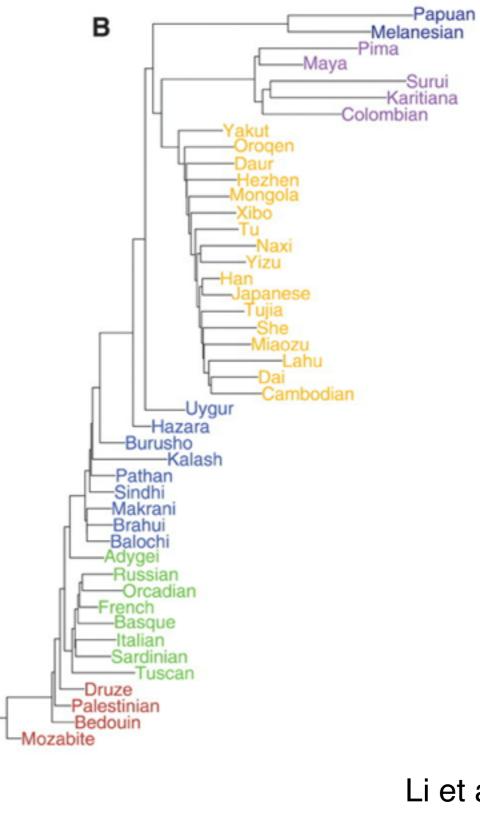
### Application to humans

#### Advantages of tree-building:

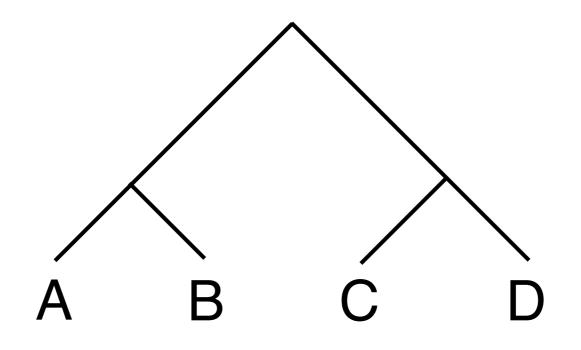
- Fast, useful summary of population relationships
- Uses all populations

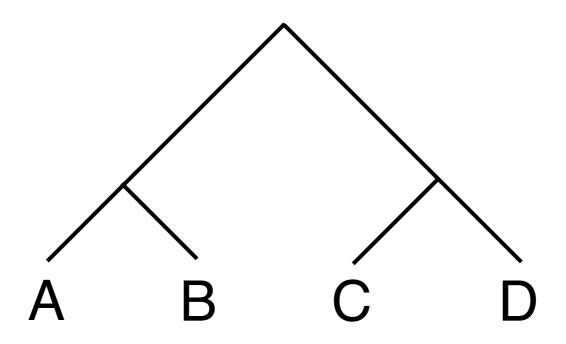
#### Disadvantages:

- Ignores migration, which makes interpretation difficult
- No way of judging goodness of fit



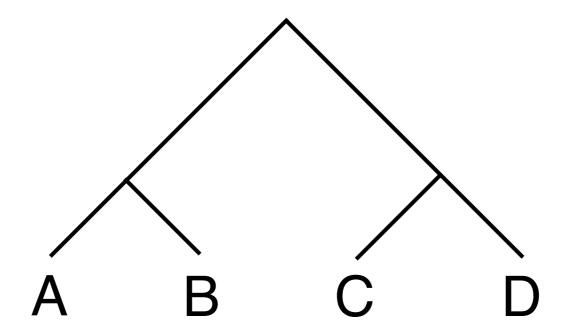
Li et al. (2008)





Four-population test: consider the following statistic, averaged over all SNPs in a genome

$$f_4 = (f_A - f_B)(f_C - f_D)$$

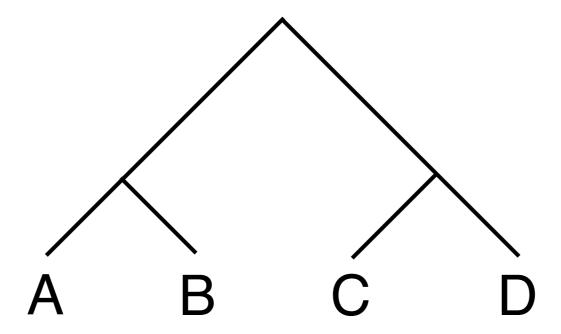


Four-population test: consider the statistic

$$f_4 = (f_A - f_B)(f_C - f_D)$$

This is equivalent to the following expression in terms of covariances:

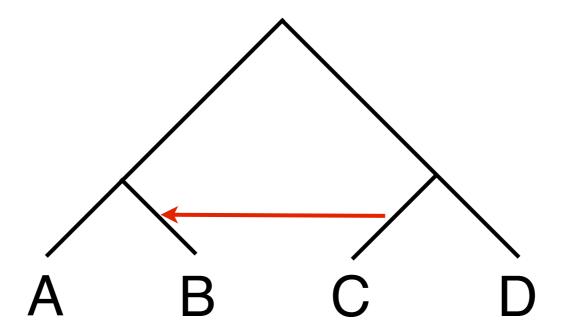
$$f_4 = V_{AC} - V_{BC} - V_{AD} + V_{BD}$$



Four-population test: consider the statistic

$$f_4 = V_{AC} - V_{BC} - V_{AD} + V_{BD}$$

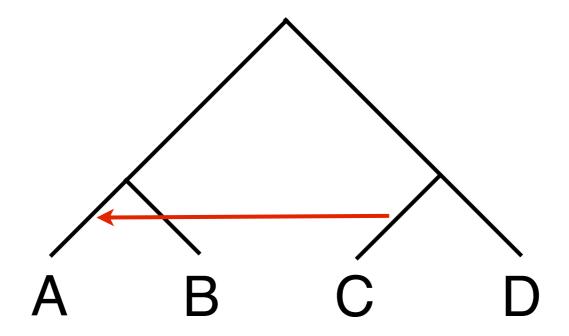
In the absence of migration (i.e. if the tree is correct), the expected value of this statistic is 0



Four-population test: consider the statistic

$$f_4 = V_{AC} - \boxed{V_{BC}} - V_{AD} + V_{BD}$$

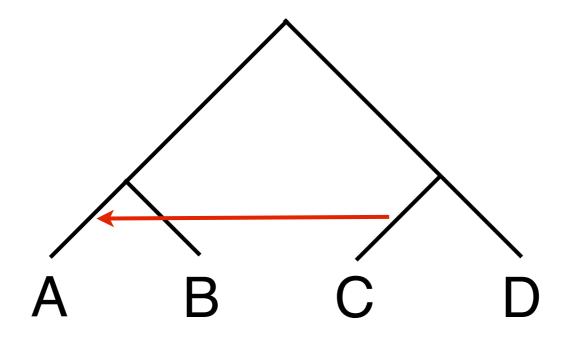
Negative value: gene flow between (populations related to) B and C or A and D



Four-population test: consider the statistic

$$f_4 = V_{AC} - V_{BC} - V_{AD} + V_{BD}$$

Positive value: gene flow between (populations related to) A and C or B and D

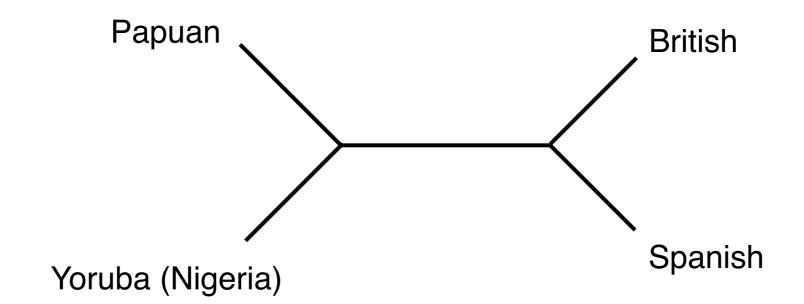


How to test significance? Compute statistic many times, dropping out large genomic regions, compute standard error in estimate, compute Z-score

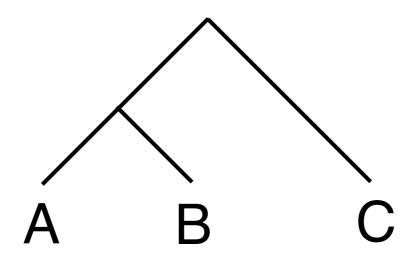
(This is a standard statistical technique known as the jackknife)

### Application to humans

Consider the following unrooted tree



- Does this tree work?
- Test (P-Y)(B-S), get value greater than 0, with a Z-score around 12 (p-value negligible)
- What is the interpretation?

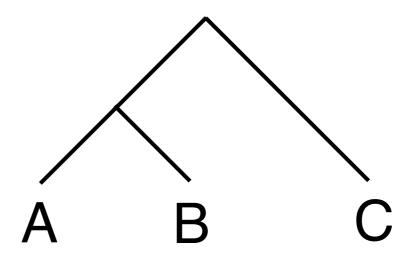


Four-population tests are often hard to interpret. Consider a three-population test for admixture in population A:

$$f_3 = (f_A - f_B)(f_A - f_C)$$

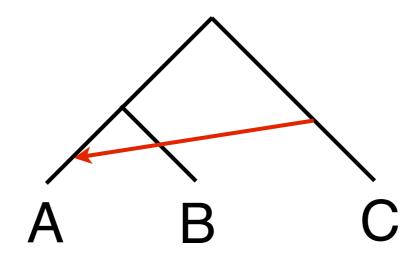
Which is equivalent to:

$$f_3 = V_{AA} - V_{AB} - V_{AC} + V_{BC}$$



Three population test: in the absence of admixture in population A, this statistic is necessarily greater than zero

$$f_3 = V_{AA} - V_{AB} - V_{AC} + V_{BC}$$

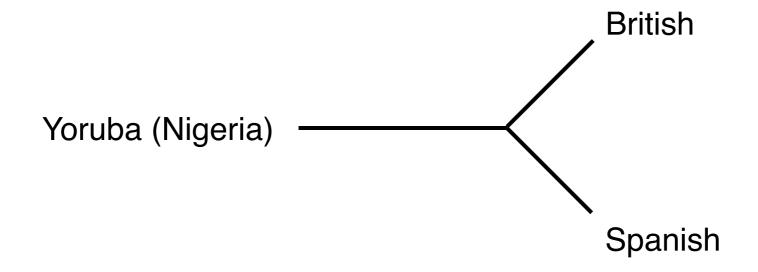


Three population test: in the presence of admixture in population A, this statistic can be less than zero

$$f_3 = V_{AA} - V_{AB} - \boxed{V_{AC} + V_{BC}}$$

#### Application to humans

Consider the following unrooted tree



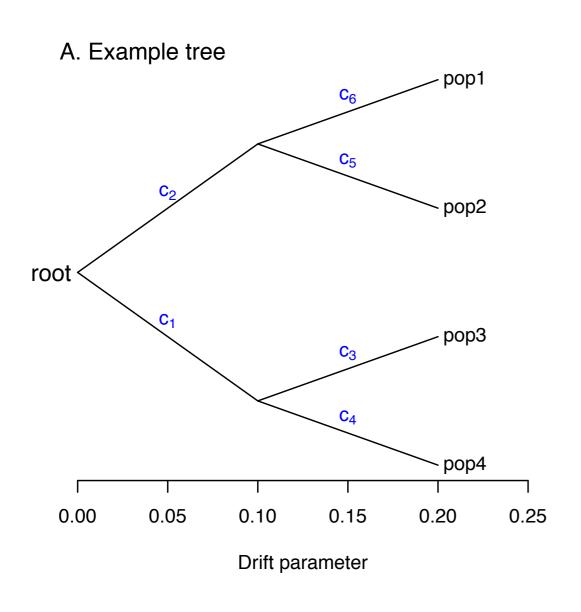
- Does this tree work?
- Test Spanish for admixture; ie. calculate (S-Y)(S-B). Negative three-population test; Z-score around -20 (p-value negligible)
- What is the interpretation?

# Can we incorporate migration into trees?

 Trees of populations can be constructed efficiently for many populations, three- and four-population tests indicate places where a tree fails

Is there a model that can handle many populations with migration?

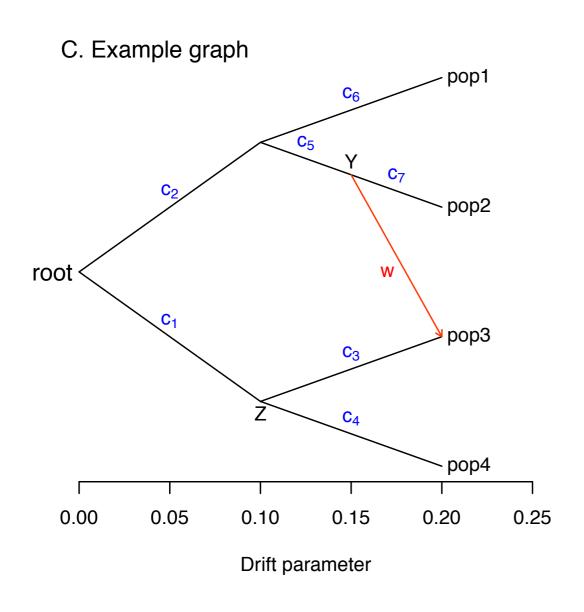
## How are the allele frequencies in different populations related?



#### B. Covariance matrix for tree in A.

| pop1 | C <sub>2</sub> + C <sub>6</sub> | C <sub>2</sub>                  | 0                               | 0                               |  |
|------|---------------------------------|---------------------------------|---------------------------------|---------------------------------|--|
| pop2 | C <sub>2</sub>                  | C <sub>2</sub> + C <sub>5</sub> | 0                               | 0                               |  |
| pop3 | 0                               | 0                               | C <sub>1</sub> + C <sub>3</sub> | C <sub>1</sub>                  |  |
| pop4 | 0                               | 0                               | C <sub>1</sub>                  | C <sub>1</sub> + C <sub>4</sub> |  |
| ,    | pop1                            | pop2                            | рор3                            | pop4                            |  |

## How are the allele frequencies in different populations related?



D. Covariance matrix for graph in C.

| pop1 | C <sub>2</sub> + C <sub>6</sub> | C <sub>2</sub>                            | WC <sub>2</sub>                                | 0                               |  |
|------|---------------------------------|---|--|---------------------------------|--|
| pop2 | C <sub>2</sub>                  | $c_2 + c_5 + c_7$ $w(c_2 + c_5)$          |  | 0                               |  |
| рорЗ | WC <sub>2</sub>                 | $\mathbf{W}(\mathbf{C}_2 + \mathbf{C}_5)$ | $w^{2}(c_{2}+c_{5})$ $+(1-w)^{2}(c_{1}+c_{3})$ | (1-w)c <sub>1</sub>             |  |
| pop4 | 0                               | 0   | (1-w)c <sub>1</sub>                            | C <sub>1</sub> + C <sub>4</sub> |  |
| ,    | pop1                            | pop2                                      | pop3   | pop4                            |  |

Pickrell and Pritchard (2012)

See also Patterson et al. (2012)

## Fit the observed matrix to the one predicted by the tree/graph

#### Graph-based prediction

Observed data

| pop1 | C <sub>2</sub> + C <sub>6</sub> | C <sub>2</sub>                                   | wc <sub>2</sub>                                    | 0                               | pop1 | $\mathbf{\hat{W}}_{11}$ | $\mathbf{\hat{W}}_{12}$       |      |      |
|------|---------------------------------|--|--|---------------------------------|------|-------------------------|-------------------------------|------|------|
| pop2 | C <sub>2</sub>                  | C <sub>2</sub> + C <sub>5</sub> + C <sub>7</sub> | $\mathbf{W}(\mathbf{C}_2 + \mathbf{C}_5)$          | 0                               | pop2 | $\mathbf{\hat{W}}_{21}$ | $oxed{\hat{\mathbf{W}}_{22}}$ | •••  |      |
| pop3 | WC <sub>2</sub>                 | $W(C_2 + C_5)$                                   | $w^{2}(c_{2}+c_{5})$<br>+ $(1-w)^{2}(c_{1}+c_{3})$ | (1-w)c <sub>1</sub>             | рорЗ |                         |                               |      |      |
| pop4 | 0                               | 0  | (1-w)c <sub>1</sub>                                | C <sub>1</sub> + C <sub>4</sub> | pop4 |                         |                               |      |      |
| ·    | pop1                            | pop2   | pop3   | pop4                            |      | pop1                    | pop2                          | pop3 | pop4 |

Algorithm: find a graph that best fits the data

## How to quantify fit?

#### Graph-based prediction (W)

#### Observed data

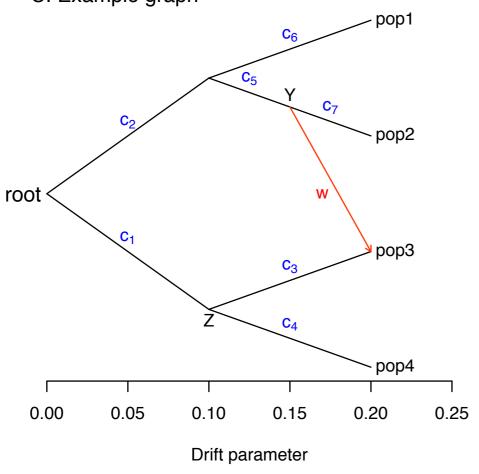
| pop1 | <b>C</b> <sub>2</sub> + <b>C</b> <sub>6</sub> | C <sub>2</sub>                                   | WC <sub>2</sub>                                    | 0                               | pop1 | $\mathbf{\hat{W}}_{11}$     | $\mathbf{\hat{W}}_{12}$   |      |      |
|------|---|--|--|---------------------------------|------|-----------------------------|---------------------------|------|------|
| pop2 | C <sub>2</sub>                                | C <sub>2</sub> + C <sub>5</sub> + C <sub>7</sub> | $\mathbf{W}(\mathbf{C}_2 + \mathbf{C}_5)$          | 0                               | pop2 | $old \hat{\mathbf{W}}_{21}$ | $oldsymbol{\hat{W}}_{22}$ | •    |      |
| рор3 | WC <sub>2</sub>                               | $\mathbf{W}(\mathbf{C}_2 + \mathbf{C}_5)$        | $w^{2}(c_{2}+c_{5})$<br>+ $(1-w)^{2}(c_{1}+c_{3})$ | (1-w)c <sub>1</sub>             | pop3 |                             |                           |      |      |
| pop4 | 0   | 0  | (1-w)c <sub>1</sub>                                | C <sub>1</sub> + C <sub>4</sub> | pop4 |                             |                           |      |      |
| ·    | pop1  | pop2   | рорЗ   | pop4                            | ·    | pop1                        | pop2                      | рорЗ | pop4 |

#### Composite likelihood:

$$l(\hat{W}|W) = \sum_{i=0}^{m} \sum_{j=i}^{m} N(\hat{W}_{ij}|W_{ij}, \hat{\sigma}_{ij}^{2})$$

#### **Estimation**





D. Covariance matrix for graph in C.

| pop1 | C <sub>2</sub> + C <sub>6</sub> | <b>c</b> <sub>2</sub>                     | WC <sub>2</sub>                                | 0                               |
|------|---------------------------------|---|--|---------------------------------|
| pop2 | C <sub>2</sub>                  | $C_2 + C_5 + C_7$ $W(C_2 + C_5)$          |  | 0                               |
| рорЗ | wc <sub>2</sub>                 | $\mathbf{W}(\mathbf{C}_2 + \mathbf{C}_5)$ | $w^{2}(c_{2}+c_{5})$ $+(1-w)^{2}(c_{1}+c_{3})$ | (1-w)c <sub>1</sub>             |
| pop4 | 0                               | 0   | (1-w)c <sub>1</sub>                            | C <sub>1</sub> + C <sub>4</sub> |
|      | pop1                            | pop2                                      | рор3   | pop4                            |

$$C_{11} = c_6 + c_2$$

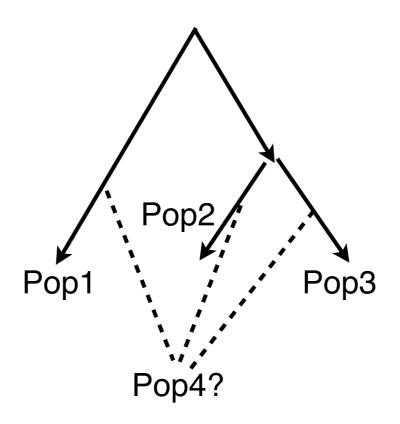
$$C_{12} = c_2$$

$$C_{12} = c_2$$
$$C_{13} = wc_2$$

Fix w, solve c's by (non-negative) least squares Search over w to get MLEs for a given topology

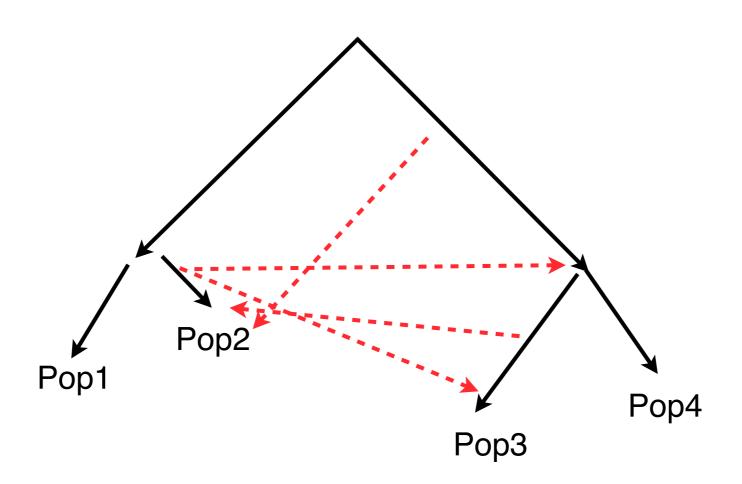
#### Estimation

1. Estimate tree without migration



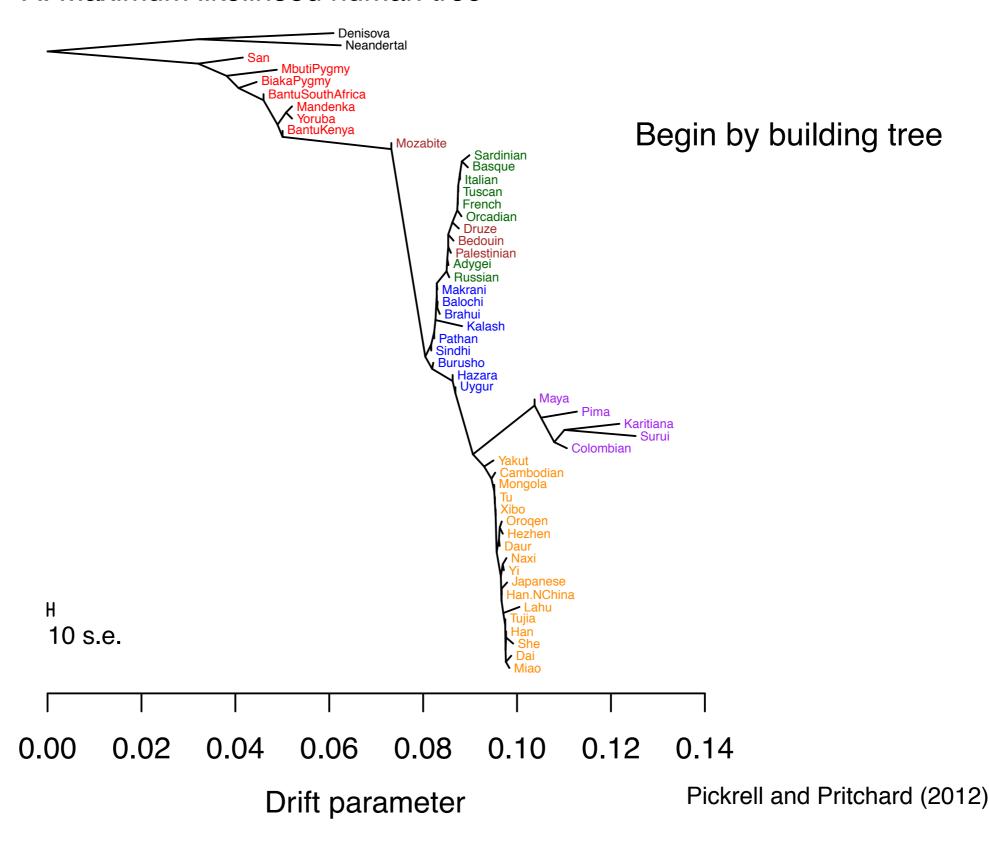
#### Estimation

2. Find poorly fitted populations, try migration events



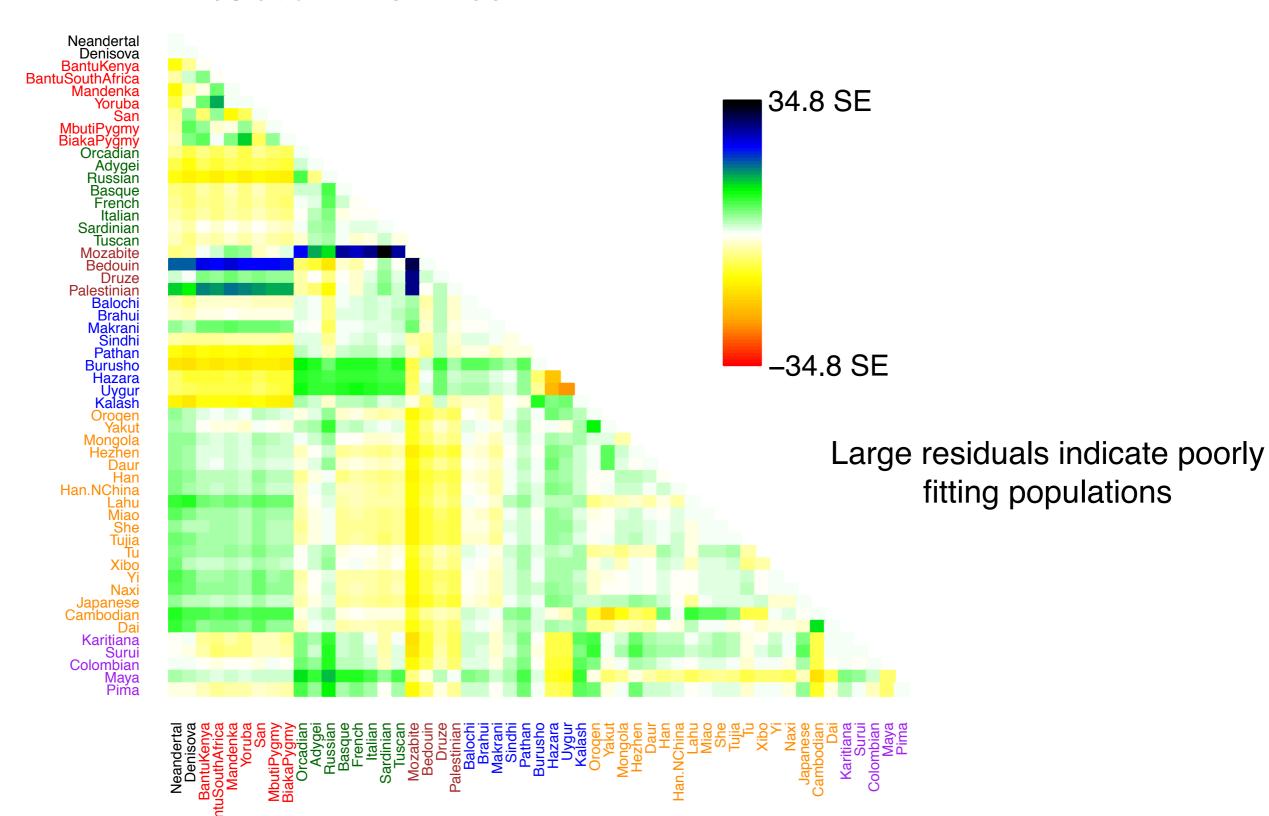
#### Application 1: worldwide sample of humans

#### A. Maximum likelihood human tree



#### Application 1: worldwide sample of humans

#### B. Residual fit from tree



#### Application 1: worldwide sample of humans

