

Lesson 10: qpAdm

Friday July 27, 2018

9:00 – 11:30 am

Important D-statistic relationships in Haak et al. (2015)

- Class Discussion

Important D-statistic relationships in Haak et al. (2015)

- Three main categories we will focus on:
 - Early Neolithic
 - Middle Neolithic
 - Late Neolithic/Bronze Age

Early Neolithic (LBK)

- $D(\text{EUR, Esperstedt; LBK, Mbuti}) < 0$ and $D(\text{EUR, LBK; Esperstedt, Mbuti}) < 0$
 - Clade with Esperstedt relative to most EUR
- Clade with MN/EN relative to LN/HG
 - $D(\text{EN, LBK; HG, Mbuti}) \sim 0$
 - $D(\text{MN, LBK; HG, Mbuti}) > 0$

Middle Neolithic (Yamnaya)

- $D(\text{EUR, CW/Mot/Kar/Sam; Yamnaya, Mbuti}) < 0$
and $D(\text{EUR no HG, Yamnaya; CW/Mot/Kar/Sam, Mbuti}) < 0$
 - Yamnaya mostly clade with EHG and Corded_Ware
- $D(\text{EN/MN, LN; Yamnaya, Mbuti}) < 0$ but most $D(\text{EN/MN, Yamnaya; LN, Mbuti}) \sim 0$ (but Karsdorf < 0 and some others)
 - $D(\text{LN, Yamnaya; EN/MN, Mbuti}) > 0$
 - What is going on here?

Middle Neolithic (Esperstedt)

- $D(\text{EUR, SpainMN/EN/LBK; Esperstedt, Mbuti}) < 0$ and $D(\text{EUR, Esperstedt; SpainMN/EN/LBK, Mbuti}) < 0$
 - Esperstedt mostly groups with other MN and EN in Europe

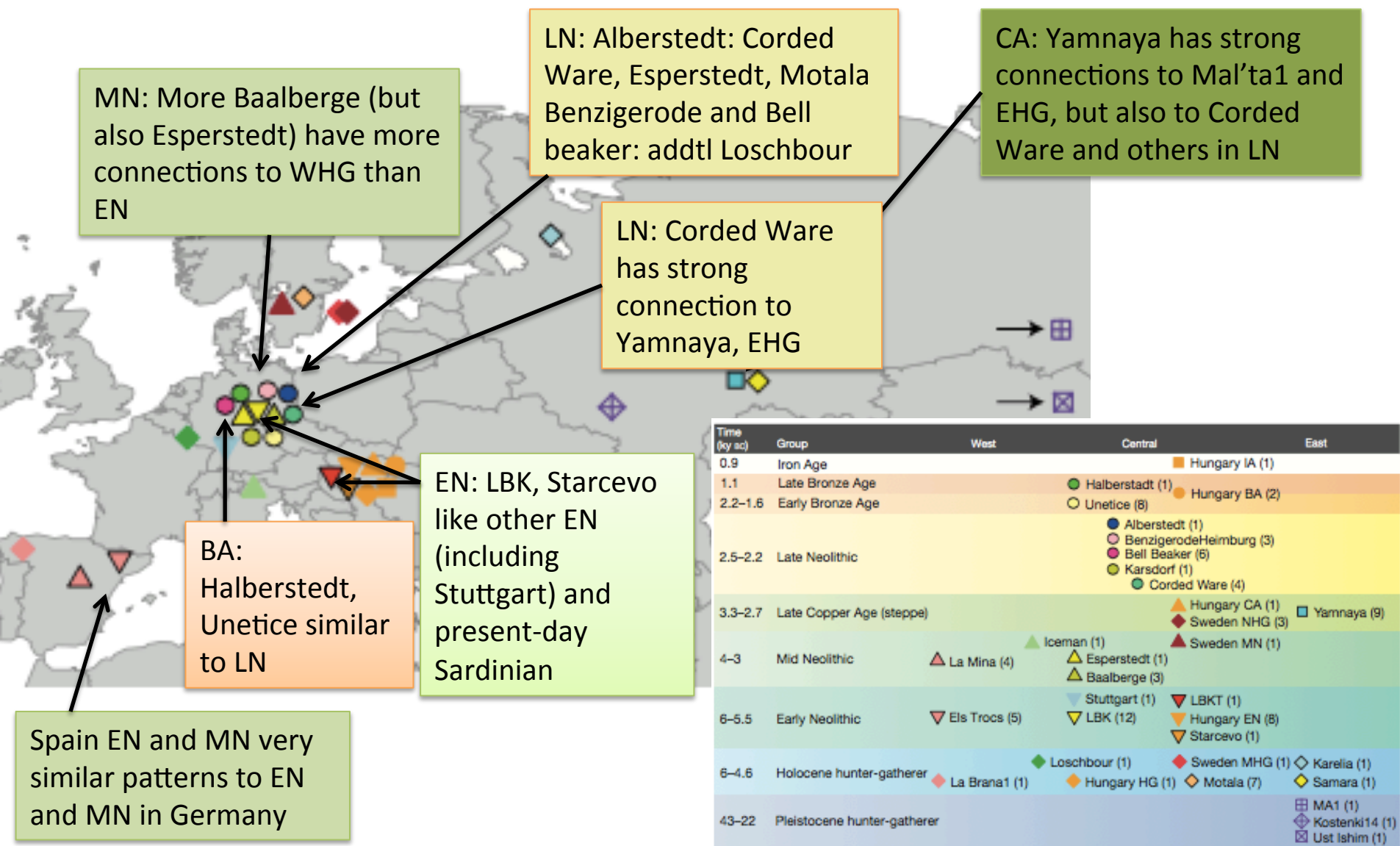
Late Neolithic/Bronze Age (Corded Ware)

- $D(\text{EUR}, \text{Yam/Mot/Kar/Sam}; \text{Corded_Ware}, \text{Mbuti}) < 0$ and $D(\text{EUR no HG}, \text{Corded_Ware}; \text{Yam/Mot/Kar/Sam}, \text{Mbuti}) < 0$
 - Corded_Ware mostly clade with EHG and Yamnaya
- See $D(\text{EUR}, \text{Esperstedt}; \text{Corded Ware}, \text{Mbuti}) < 0$ but $D(\text{EUR}, \text{Corded Ware}; \text{Esperstedt}, \text{Mbuti})$ mostly ≥ 0

Late Neolithic/Bronze Age (Alberstedt)

- $D(\text{EUR, CW/Esper/Mot; Alberstedt, Mbuti}) < 0$
but $D(\text{EN/MN, Alberstedt; CW/Yam/Mot/Kar/Sam, Mbuti}) < 0$
- CW connections (and thus, Yamnaya connections), but also MN connections

Comparing to Haak et al. (2015)



From the Haak et al. paper

instead, the Corded Ware are both the earliest and most strongly differentiated from the Middle Neolithic population.

'Outgroup' f_3 statistics⁶ (Supplementary Information section 7), which measure shared genetic drift between a pair of populations (Extended Data Fig. 1), support the clustering of hunter-gatherers, Early/Middle Neolithic, and Late Neolithic/Bronze Age populations into different groups as in the PCA (Fig. 2a). We also analysed f_4 statistics, which allow us to test whether pairs of populations are consistent with descent from common ancestral populations, and to assess significance using a normally distributed Z score. Early European farmers from the Early and Middle Neolithic were closely related but not identical. This is reflected in the fact that Loschbour, a WHG individual from Luxembourg⁴, shared more alleles with post-4,000 BC European farmers from Germany, Spain, Hungary, Sweden and Italy than with early farmers of Germany, Spain, and Hungary, documenting an increase of hunter-gatherer ancestry in multiple regions of Europe during the course of the Neolithic. The two EHGs form a clade with respect to all other present-day and ancient populations ($|Z| < 1.9$), and MA1 shares more alleles with them ($|Z| > 4.7$) than with other ancient or modern populations, suggesting that they may be a source for the ANE ancestry in present Europeans^{4,12,22} as they are geographically and temporally more proximate than Upper Paleolithic Siberians. The Yamnaya differ from the EHGs by sharing fewer alleles with MA1 ($|Z| = 6.7$) suggesting a dilution of ANE ancestry between 5,000–3,000 BC on the European steppe. This was likely due to admixture of EHGs with a population related to present-day Near Easterners, as the most negative f_3 statistic in the Yamnaya (giving unambiguous evidence of admixture) is observed when we model them as a mixture of EHGs and present-day Near Eastern populations like Armenians ($Z = -6.3$;

Supplementary Information section 7). The Late Neolithic/Bronze Age groups of central Europe share more alleles with Yamnaya than the Middle Neolithic populations do ($|Z| = 12.4$) and more alleles with the Middle Neolithic than the Yamnaya do ($|Z| = 12.5$), and have a negative f_3 statistic with the Middle Neolithic and Yamnaya as references ($Z = -20.7$), indicating that they were descended from a mixture of the local European populations and new migrants from the east. Moreover, the Yamnaya share more alleles with the Corded Ware ($|Z| \geq 3.6$) than with any other Late Neolithic/Early Bronze Age group with at least two individuals (Supplementary Information section 7), indicating that they had more eastern ancestry, consistent with the PCA and ADMIXTURE patterns (Fig. 2).

Modelling of the ancient samples shows that while Karelia is genetically intermediate between Loschbour and MA1, the topology that considers Karelia as a mixture of these two elements is not the only one that can fit the data (Supplementary Information section 8). To avoid biasing our inferences by fitting an incorrect model, we developed new statistical methods that are substantial extensions of a previously reported approach⁴, which allow us to obtain precise estimates of the proportion of mixture in later Europeans without requiring a formal model for the relationship among the ancestral populations. The method (Supplementary Information section 9) is based on the idea that if a Test population has ancestry related to reference populations $\text{Ref}_1, \text{Ref}_2, \dots, \text{Ref}_N$ in proportions $\alpha_1, \alpha_2, \dots, \alpha_N$, and the references are themselves differentially related to a triple of outgroup populations A, B, C , then:

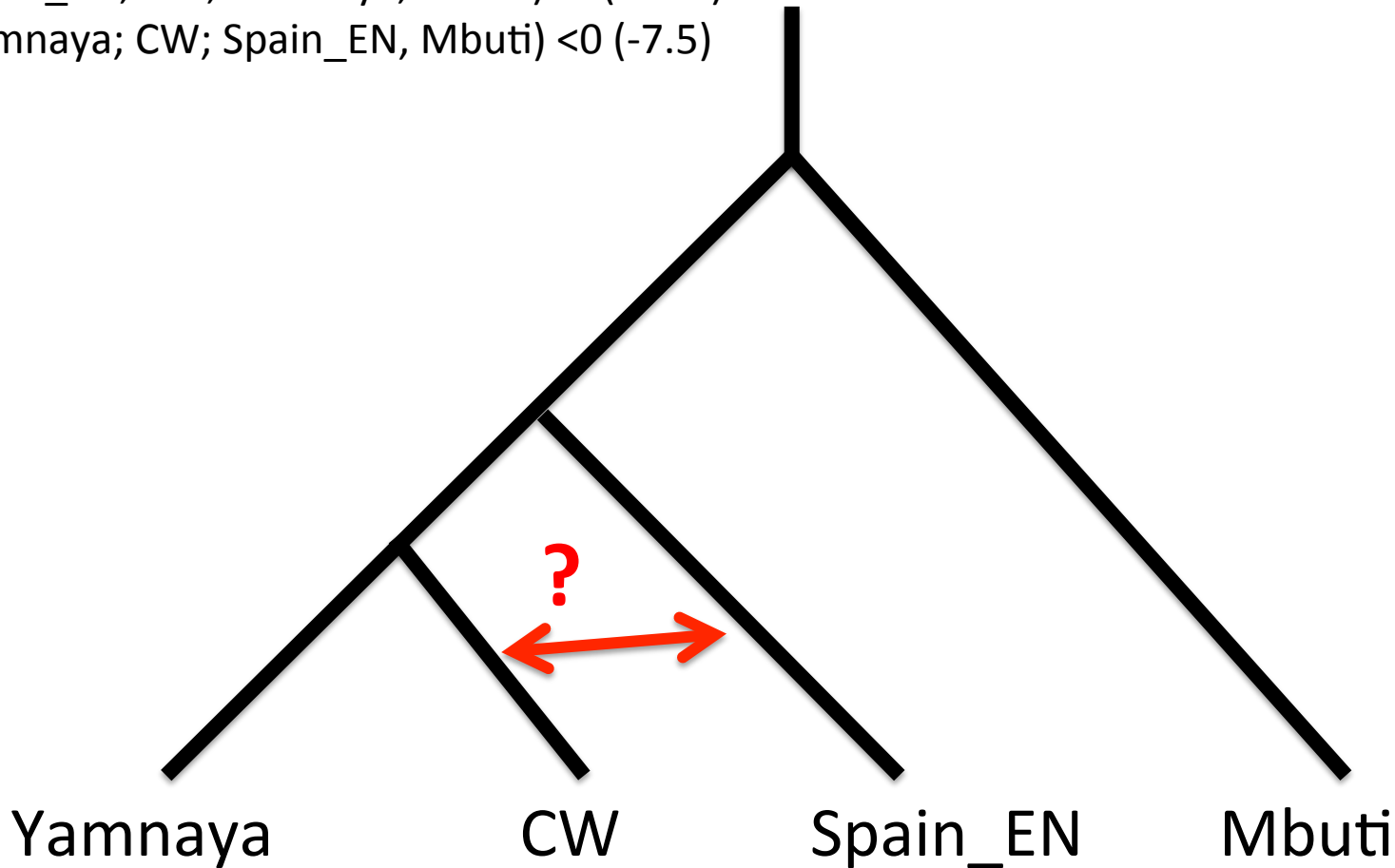
$$f_4(\text{Test}, A; B, C) = \sum_{i=1}^N \alpha_i f_4(\text{Ref}_i, A; B, C)$$

Spain_EN, Yamnaya, Corded Ware

$D(\text{Spain_EN}, \text{Yamnaya}; \text{CW}, \text{Mbuti}) < 0$ (-6.8)

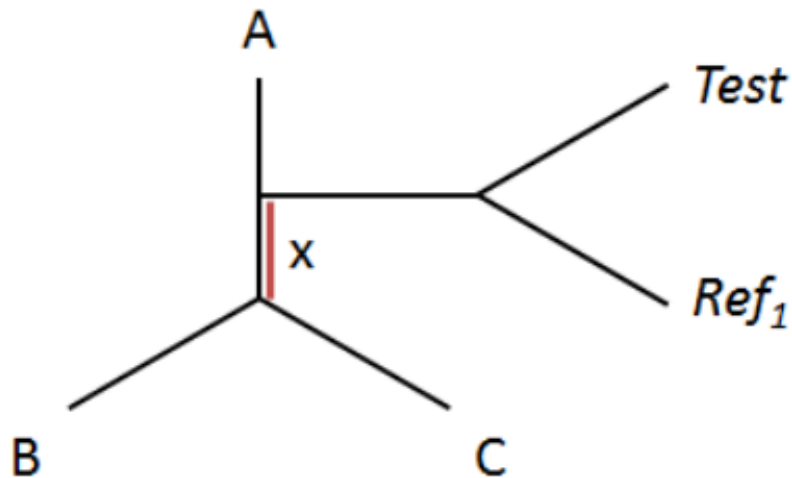
$D(\text{Spain_EN}, \text{CW}; \text{Yamnaya}, \text{Mbuti}) < 0$ (-14.3)

$D(\text{Yamnaya}; \text{CW}; \text{Spain_EN}, \text{Mbuti}) < 0$ (-7.5)



qpWave

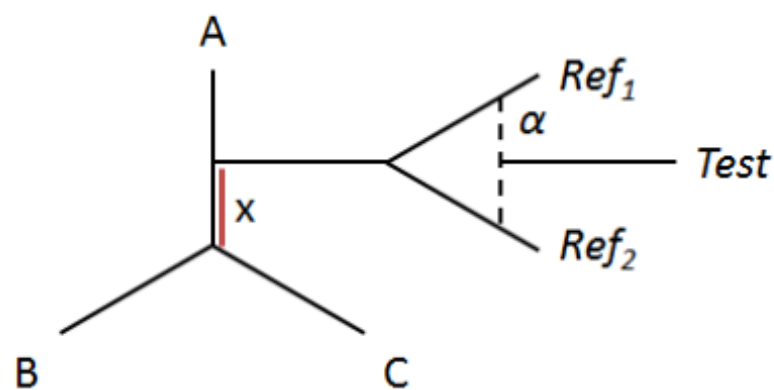
Figure S9.1: If *Test* and *Ref₁* form a clade with respect to the outgroups, then statistics involving either one of them and the outgroups are identical.



	<i>X=Test</i>	<i>X=Ref₁</i>
$f_4(X, A; B, C)$	0	0
$f_4(X, B; A, C)$	x	x
$f_4(X, C; A, B)$	x	x

$$f_4(\text{Test}, A; B, C) = f_4(\text{Ref}_1, A; B, C)$$

(a)

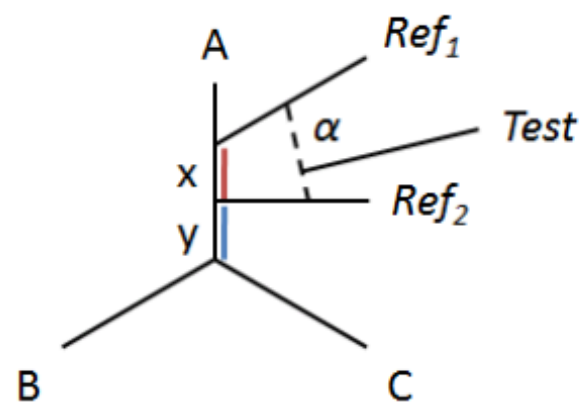


$$f_4(\text{Test or Ref}_1 \text{ or Ref}_2, A; B, C) = 0$$

$$f_4(\text{Test or Ref}_1 \text{ or Ref}_2, B; A, C) = x$$

$$f_4(\text{Test or Ref}_1 \text{ or Ref}_2, C; A, B) = x$$

(b)



$$f_4(\text{Ref}_1, A; B, C) = 0 \quad f_4(\text{Ref}_2, A; B, C) = 0$$

$$f_4(\text{Ref}_1, B; A, C) = x+y \quad f_4(\text{Ref}_2, B; A, C) = y$$

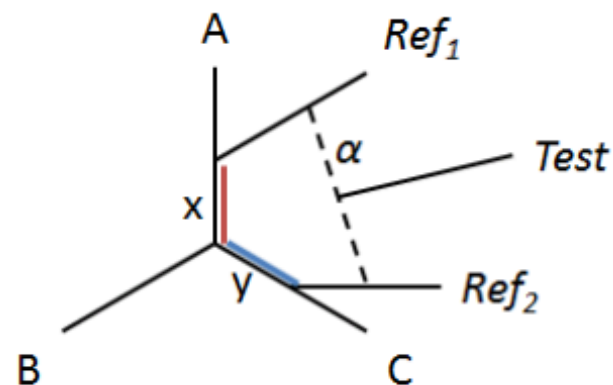
$$f_4(\text{Ref}_1, C; A, B) = x+y \quad f_4(\text{Ref}_2, C; A, B) = y$$

$$f_4(\text{Test}, A; B, C) = 0$$

$$f_4(\text{Test}, B; A, C) = y + \alpha x = (1-\alpha) f_4(\text{Ref}_2, B; A, C) + \alpha f_4(\text{Ref}_1, B; A, C)$$

$$f_4(\text{Test}, C; A, B) = y + \alpha x = (1-\alpha) f_4(\text{Ref}_2, C; A, B) + \alpha f_4(\text{Ref}_1, C; A, B)$$

(c)



$$f_4(\text{Ref}_1, A; B, C) = 0 \quad f_4(\text{Ref}_2, A; B, C) = -y$$

$$f_4(\text{Ref}_1, B; A, C) = x \quad f_4(\text{Ref}_2, B; A, C) = -y$$

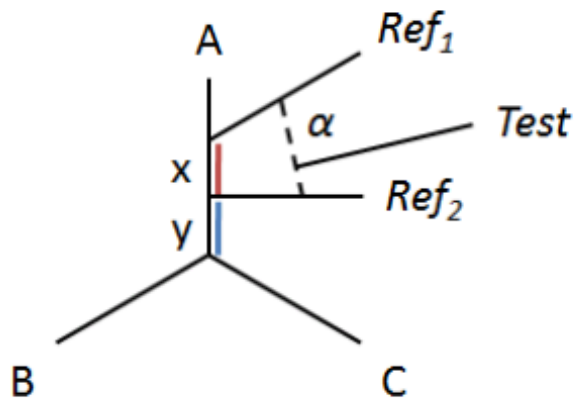
$$f_4(\text{Ref}_1, C; A, B) = x \quad f_4(\text{Ref}_2, C; A, B) = 0$$

$$f_4(\text{Test}, A; B, C) = -(1-\alpha)y = (1-\alpha) f_4(\text{Ref}_2, A; B, C) + \alpha f_4(\text{Ref}_1, A; B, C)$$

$$f_4(\text{Test}, B; A, C) = \alpha x - (1-\alpha)y = \alpha f_4(\text{Ref}_1, A; B, C) + (1-\alpha) f_4(\text{Ref}_2, A; B, C)$$

$$f_4(\text{Test}, C; A, B) = \alpha x = \alpha f_4(\text{Ref}_1, A; B, C) + (1-\alpha) f_4(\text{Ref}_2, A; B, C)$$

(b)



$$f_4(\text{Ref}_1, A; B, C) = 0$$

$$f_4(\text{Ref}_1, B; A, C) = x+y$$

$$f_4(\text{Ref}_1, C; A, B) = x+y$$

$$f_4(\text{Ref}_2, A; B, C) = 0$$

$$f_4(\text{Ref}_2, B; A, C) = y$$

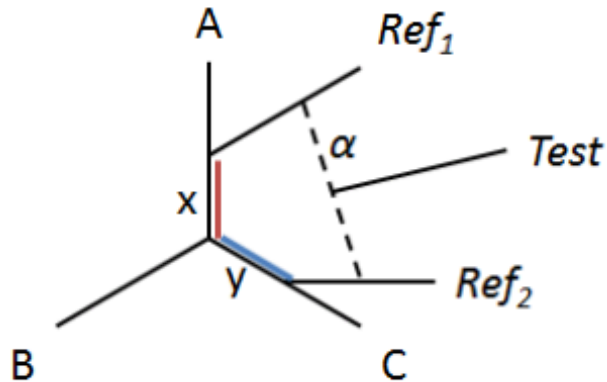
$$f_4(\text{Ref}_2, C; A, B) = y$$

$$f_4(\text{Test}, A; B, C) = 0$$

$$f_4(\text{Test}, B; A, C) = y+\alpha x = (1-\alpha)f_4(\text{Ref}_2, B; A, C) + \alpha f_4(\text{Ref}_1, B; A, C)$$

$$f_4(\text{Test}, C; A, B) = y+\alpha x = (1-\alpha)f_4(\text{Ref}_2, C; A, B) + \alpha f_4(\text{Ref}_1, C; A, B)$$

(c)



$$f_4(\text{Ref}_1, A; B, C) = 0$$

$$f_4(\text{Ref}_1, B; A, C) = x$$

$$f_4(\text{Ref}_1, C; A, B) = x$$

$$f_4(\text{Ref}_2, A; B, C) = -y$$

$$f_4(\text{Ref}_2, B; A, C) = -y$$

$$f_4(\text{Ref}_2, C; A, B) = 0$$

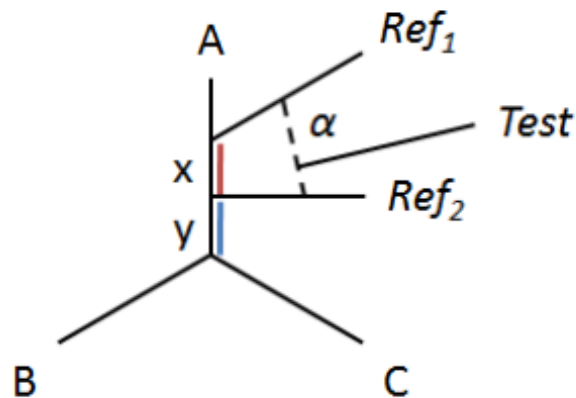
$$f_4(\text{Test}, A; B, C) = -(1-\alpha)y = (1-\alpha)f_4(\text{Ref}_2, A; B, C) + \alpha f_4(\text{Ref}_1, A; B, C)$$

$$f_4(\text{Test}, B; A, C) = \alpha x - (1-\alpha)y = \alpha f_4(\text{Ref}_1, B; A, C) + (1-\alpha)f_4(\text{Ref}_2, B; A, C)$$

$$f_4(\text{Test}, C; A, B) = \alpha x = \alpha f_4(\text{Ref}_1, C; A, B) + (1-\alpha)f_4(\text{Ref}_2, C; A, B)$$

$$f_4(\text{Test}, A; B, C) = \alpha f_4(\text{Ref}_1, A; B, C) + (1-\alpha) f_4(\text{Ref}_2, A; B, C)$$

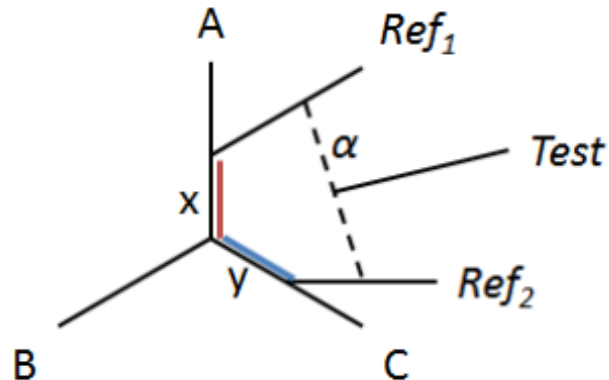
(b)



$$\begin{aligned} f_4(\text{Ref}_1, A; B, C) &= 0 & f_4(\text{Ref}_2, A; B, C) &= 0 \\ f_4(\text{Ref}_1, B; A, C) &= x+y & f_4(\text{Ref}_2, B; A, C) &= y \\ f_4(\text{Ref}_1, C; A, B) &= x+y & f_4(\text{Ref}_2, C; A, B) &= y \end{aligned}$$

$$\begin{aligned} f_4(\text{Test}, A; B, C) &= 0 \\ f_4(\text{Test}, B; A, C) &= y+\alpha x = (1-\alpha)f_4(\text{Ref}_2, B; A, C) + \alpha f_4(\text{Ref}_1, B; A, C) \\ f_4(\text{Test}, C; A, B) &= y+\alpha x = (1-\alpha)f_4(\text{Ref}_2, C; A, B) + \alpha f_4(\text{Ref}_1, C; A, B) \end{aligned}$$

(c)



$$\begin{aligned} f_4(\text{Ref}_1, A; B, C) &= 0 & f_4(\text{Ref}_2, A; B, C) &= -y \\ f_4(\text{Ref}_1, B; A, C) &= x & f_4(\text{Ref}_2, B; A, C) &= -y \\ f_4(\text{Ref}_1, C; A, B) &= x & f_4(\text{Ref}_2, C; A, B) &= 0 \end{aligned}$$

$$\begin{aligned} f_4(\text{Test}, A; B, C) &= -(1-\alpha)y = (1-\alpha)f_4(\text{Ref}_2, A; B, C) + \alpha f_4(\text{Ref}_1, A; B, C) \\ f_4(\text{Test}, B; A, C) &= \alpha x - (1-\alpha)y = \alpha f_4(\text{Ref}_1, B; A, C) + (1-\alpha)f_4(\text{Ref}_2, B; A, C) \\ f_4(\text{Test}, C; A, B) &= \alpha x = \alpha f_4(\text{Ref}_1, C; A, B) + (1-\alpha)f_4(\text{Ref}_2, C; A, B) \end{aligned}$$

$$f_4(\text{Test}, A; B, C) = \alpha f_4(\text{Ref}_1, A; B, C) + (1-\alpha) f_4(\text{Ref}_2, A; B, C)$$

More generally,

$$f_4(\text{Test}; A; B, C) \approx \sum_{i=1}^N \alpha_i f_4(\text{Ref}_i; A; B, C)$$

with $\sum_{i=1}^N \alpha_i = 1$ and $\alpha_i \geq 0, i = 1, \dots, N$.

F4(Test; A, B, C)

- A, B, and C are all outgroups, with m reference sets
- If n outgroups, then $n(n-1 \text{ choose } 2)$ arrangements.
- Matrix of $n(n-1 \text{ choose } 2)$ rows and m columns

$$f_4(\text{Test}; A; B, C) \approx \sum_{i=1}^N \alpha_i f_4(\text{Ref}_i; A; B, C)$$

with $\sum_{i=1}^N \alpha_i = 1$ and $\alpha_i \geq 0, i = 1, \dots, N$. Given a set

t	≈	R	*	α ^T
<div> <div></div> <div>F4(Test; A; B, C)</div> <div></div> </div> <div> <div></div> <div>F4(Test; B; A, C)</div> <div></div> </div> <div> <div></div> <div>F4(Test; C; A, B)</div> <div></div> </div>	≈	<div> <div></div> <div>F4(Ref₁; A; B, C) F4(Ref₂; A; B, C)... F4(Ref_m; A; B, C)</div> <div></div> </div> <div> <div></div> <div>F4(Ref₁; B; A, C) F4(Ref₂; B; A, C)... F4(Ref_m; B; A, C)</div> <div></div> </div> <div> <div></div> <div>F4(Ref₁; C; A, B) F4(Ref₂; C; A, B)... F4(Ref_m; C; A, B)</div> <div></div> </div>		<div> <div></div> <div>α₁</div> <div></div> </div> <div> <div></div> <div>α₂</div> <div></div> </div> <div> <div></div> <div>.</div> <div></div> </div> <div> <div></div> <div>.</div> <div></div> </div> <div> <div></div> <div>.</div> <div></div> </div> <div> <div></div> <div>α_m</div> <div></div> </div>

$$f_4(\text{Test}; A; B, C) \approx \sum_{i=1}^N \alpha_i f_4(\text{Ref}_i; A; B, C)$$

with $\sum_{i=1}^N \alpha_i = 1$ and $\alpha_i \geq 0, i = 1, \dots, N$. Given a set

$$\begin{matrix} \mathbf{t} & \approx & \mathbf{R} & * & \boldsymbol{\alpha}^T \\ \left[\begin{array}{c} F4(\text{Test}; A; B, C) \\ F4(\text{Test}; B; A, C) \\ F4(\text{Test}; C; A, B) \end{array} \right] & \approx & \left[\begin{array}{ccc} F4(\text{Ref}_1; A; B, C) & F4(\text{Ref}_2; A; B, C) & \dots & F4(\text{Ref}_m; A; B, C) \\ F4(\text{Ref}_1; B; A, C) & F4(\text{Ref}_2; B; A, C) & \dots & F4(\text{Ref}_m; B; A, C) \\ F4(\text{Ref}_1; C; A, B) & F4(\text{Ref}_2; C; A, B) & \dots & F4(\text{Ref}_m; C; A, B) \end{array} \right] & * & \left[\begin{array}{c} \alpha_1 \\ \alpha_2 \\ \cdot \\ \cdot \\ \cdot \\ \alpha_m \end{array} \right] \end{matrix}$$

This problem can be solved for α by least squares, minimizing $\|\mathbf{t} - \mathbf{R}\alpha^T\|_2^2$ subject to the constraint that α has non-negative elements and $\|\alpha\|_1 = 1$. We use the implementation of least squares in the *lsqlin* function of Matlab (<http://www.mathworks.com/help/optim/ug/lsqlin.html>) which estimates the vector $\hat{\alpha}$ of the mixture proportions and also reports $\text{resnorm} = \|\mathbf{t} - \mathbf{R}\hat{\alpha}^T\|_2^2$, the squared 2-norm of the residuals.

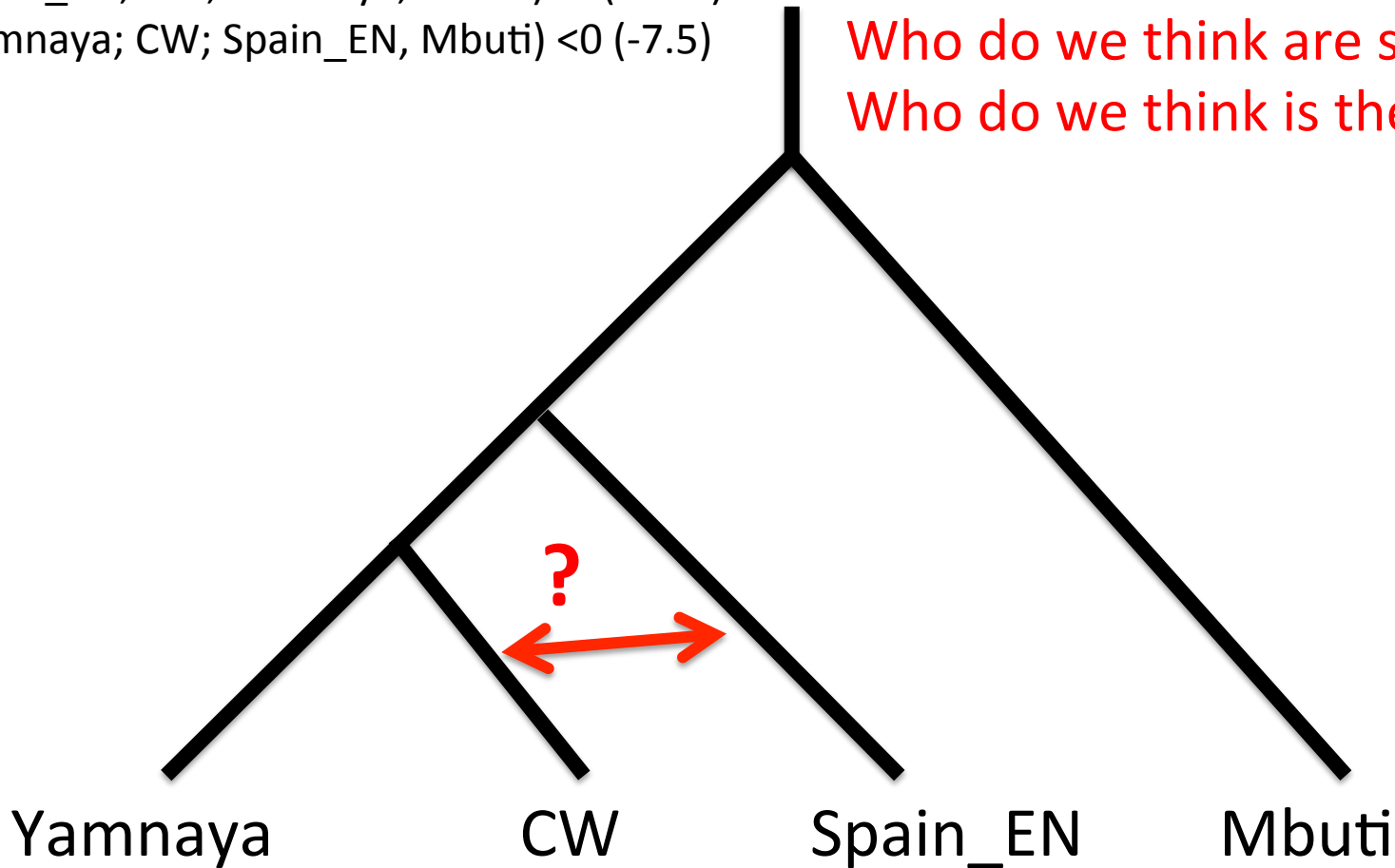
qpAdm: Some Thoughts

- Outgroup choice – very vital!
 - No need to know phylogenetic relationships BUT
 - Cannot all be from same population – if identically related to Test and References, then no leverage to find admixture proportion.
 - No recent admixture from outgroup to Test/Ref
 - Don't make this too big – computationally difficult and poor estimates result.
- qpWave important
 - Assume $T=\text{test}$, $S=\{\text{Ref1}, \text{Ref2}, \dots, \text{Refm}\}$
 - $\text{qpWave}(S)$ must have n streams of ancestry ($r=n-1$)
 - $\text{qpWave}(T, S)$ should not increase rank (n streams of ancestry)
 - In practice, we can try qpAdm without necessarily knowing qpWave conditions (but need $n-1$ streams of ancestry).

Spain_EN, Yamnaya, Corded Ware

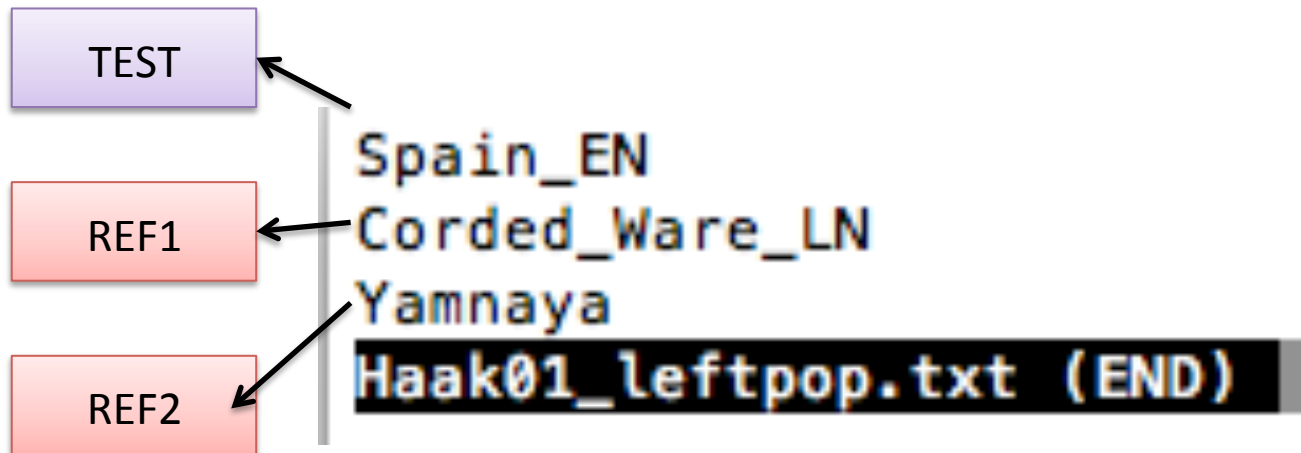
$D(\text{Spain_EN}, \text{Yamnaya}; \text{CW}, \text{Mbuti}) < 0$ (-6.8)
 $D(\text{Spain_EN}, \text{CW}; \text{Yamnaya}, \text{Mbuti}) < 0$ (-14.3)
 $D(\text{Yamnaya}; \text{CW}; \text{Spain_EN}, \text{Mbuti}) < 0$ (-7.5)

Who do we think are sources?
Who do we think is the test?



qpAdm software

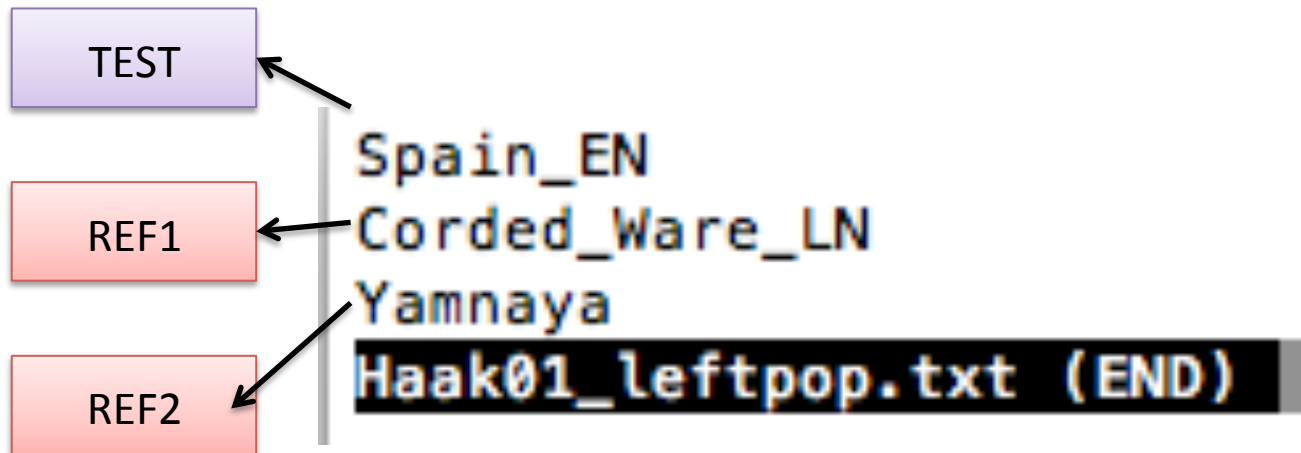
- Just like qpWave, only ORDER MATTERS in the leftpop file.



qpAdm software

- Just like qpWave, only ORDER MATTERS in the leftpop file.

What order do we want?



Results

- qpAdm -p PARFILE > LOGFILE

Corded_Ware_LN
Spain_EN
Yamnaya

```

fixed pat  wt  dof      chisq      tail prob
      00  0   11      10.961      0.446573      0.233      0.767
      01  1   12     169.847     5.13073e-30      1.000      -0.000
      10  1   12      31.394      0.00171472      0.000      1.000
best pat:                00          0.446573      - -
best pat:                10          0.00171472  chi(nested):  20.434 p-value for nested model:
6.17302e-06

```

std. errors: 0.050 0.050

00 is the mixture case.

01 means all ancestry from Ref1

10 means all ancestry from Ref2

1. Look at “tail prob” – if $p > 0.05$, then that model is possible.
2. If ‘00’ is possible, then look at nested model – is ‘00’ better than ‘01’ or ‘10’?
p-value for nested model (p_{nest}) < 0.05 , means you reject ‘10’, so ‘00’ is better than ‘10’.
3. Look at $f_1, f_2 \rightarrow 0.233$ Spain_EN and 0.767 Yamnaya with $SE=0.05$
4. Is $\min(f_1, f_2) - SE > 0$?

Results

- qpAdm -p PARFILE > LOGFILE

```
Corded_Ware_LN
Spain_EN
Yamnaya
```

```
fixed pat  wt  dof      chisq      tail prob      0.233      0.767
      00  0   11      10.961      0.446573
      01  1   12     169.847     5.13073e-30      1.000      -0.000
      10  1   12      31.394      0.00171472      0.000      1.000
best pat:                00          0.446573      -      -
best pat:                10          0.00171472  chi(nested):  20.434 p-value for nested model:
6.17302e-06
```

```
std. errors:      0.050      0.050
```

00 is the mixture case.

01 means all ancestry from Ref1

10 means all ancestry from Ref2

Corded Ware seems to be a mixture of populations related to the Spain_EN and Yamnaya populations.

1. Look at "tail prob" – if $p > 0.05$, then that model is possible.
2. If '00' is possible, then look at nested model – is '00' better than '01' or '10'?
p-value for nested model (p_{nested}) < 0.05 , means you reject '10', so '00' is better than '10'.
3. Look at $f_1, f_2 \rightarrow 0.233$ Spain_EN and 0.767 Yamnaya with $SE=0.05$
4. Is $\min(f_1, f_2) - SE > 0$?

Spain_EN or Yamnaya as Test?

Spain_EN

Corded_Ware_LN

Yamnaya

fixed	pat	wt	dof	chisq	tail prob			
	00	0	11	10.714	0.467501	4.298	-3.298	infeasible
	01	1	12	169.762	5.34075e-30	1.000	-0.000	
	10	1	12	344.858	0	0.000	1.000	
best pat:			00		0	-	-	
best pat:			01	5.34075e-30	chi(nested):	159.047	p-value for nested model:	
				1.8271e-36				
std. errors:				0.925	0.925			

Yamnaya

Spain_EN

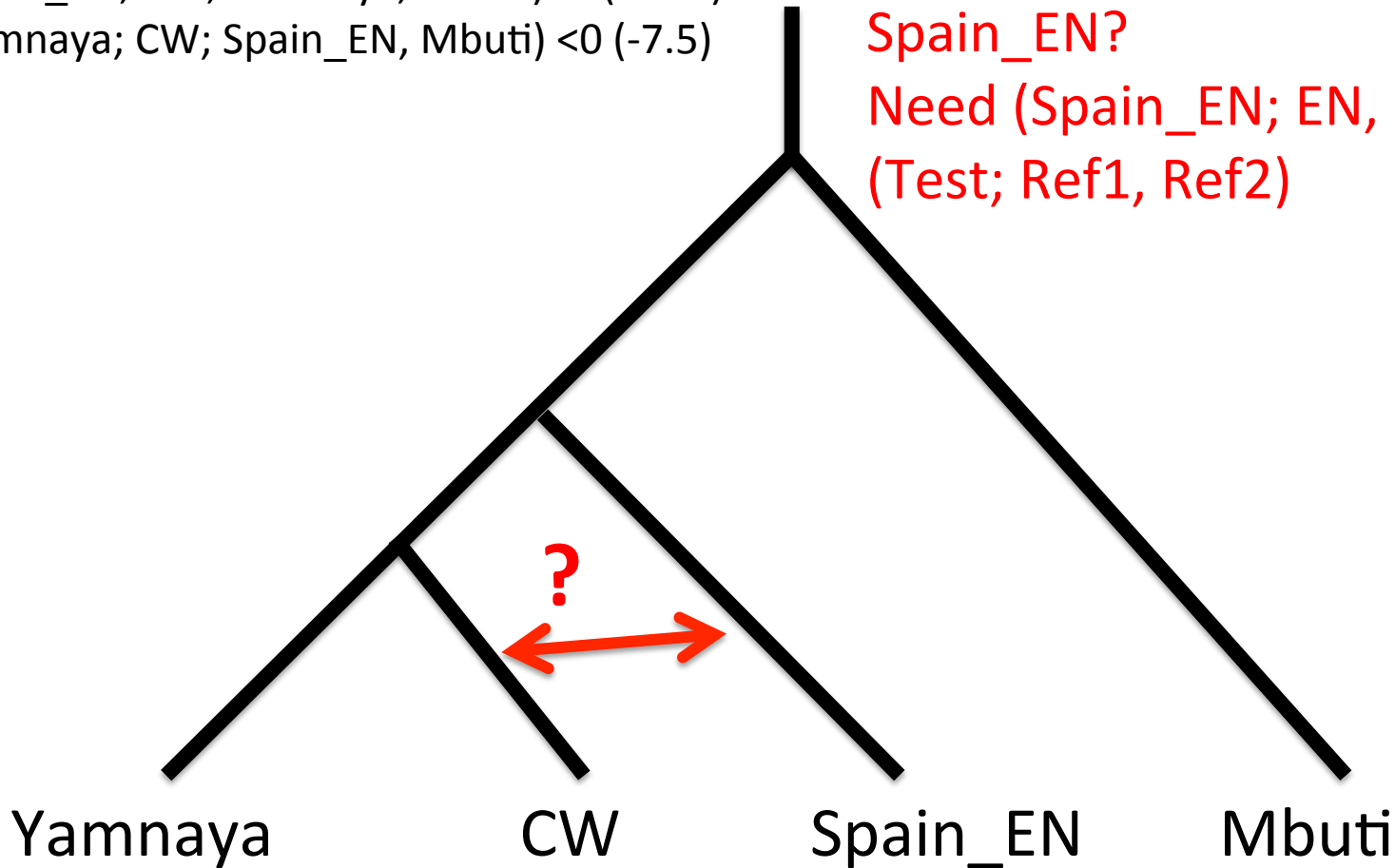
Corded_Ware_LN

fixed	pat	wt	dof	chisq	tail prob			
	00	0	11	10.882	0.453196	-0.305	1.305	infeasible
	01	1	12	345.393	0	1.000	-0.000	
	10	1	12	31.453	0.00167982	0.000	1.000	
best pat:			00		0	-	-	
best pat:			10	0.00167982	chi(nested):	20.571	p-value for nested model:	
				5.74731e-06				
std. errors:				0.084	0.084			

True reciprocal test

$D(\text{Spain_EN}, \text{Yamnaya}; \text{CW}, \text{Mbuti}) < 0$ (-6.8)
 $D(\text{Spain_EN}, \text{CW}; \text{Yamnaya}, \text{Mbuti}) < 0$ (-14.3)
 $D(\text{Yamnaya}; \text{CW}; \text{Spain_EN}, \text{Mbuti}) < 0$ (-7.5)

We can find gene flow into CW.
What about gene flow into Spain_EN?
Need (Spain_EN; EN, CW) as
(Test; Ref1, Ref2)



U	V	W	X	Y	T	U
Spain_MN	Stuttgart	LBKT_EN	LBK_EN	HungaryGamba_EN	Baalberge_MN	Spain_MN
###	-8.5	-3.2	###	###	-9.5	###
###	-3.2	-1.0	###	-4.1	-6.3	###
###	-8.6	-3.5	###	###	-8.3	###
###	-8.8	-3.4	###	-9.1	-6.3	###
###	-5.4	-2.1	-8.7	-5.1	-4.8	-8.8
###	-5.2	-1.6	###	-5.1	-3.3	###
###	-6.5	-2.4	###	-8.1	-6.9	###
-8.9	-2.6	-1.3	-5.9	-2.1	-3.3	-8.1
###	-5.9	-2.5	###	-7.1	-5.1	###
###	-6.0	-2.2	###	-7.1	-7.0	###
-9.6	-5.3	-2.2	-7.8	-6.1	-5.7	-8.7
###	-7.7	-3.4	###	-9.1	-7.4	###
-7.3	-1.7	0.7	-4.9	-2.1	0.1	-8.3
###	###	-5.3	###	###	###	###
###	-7.7	-1.8	###	-8.1	-5.9	###
-6.0	0.2	-0.6	-2.0	0.9	0.3	-5.6
-4.2	1.1	0.6	-1.1	1.8	-0.4	-4.7
-0.5	4.5	1.1	3.3	6.1	1.3	0.7
-5.0	1.9	0.8	-1.2	1.2	nan	-4.8
nan	6.4	2.3	5.6	8.8	0.3	nan
-6.4	nan	0.5	-3.0	0.3	-2.0	-9.2
-2.3	-0.5	nan	-1.0	0.1	0.0	-2.9
-5.6	3.0	1.0	nan	4.9	-1.2	-7.7
-8.8	-0.3	-0.1	-4.9	nan	-0.7	###
-2.1	2.6	0.4	0.6	2.5	-0.9	-2.3
nan	nan	nan	nan	nan	nan	nan
###	-5.6	-2.1	-9.5	-6.1	-1.8	-5.6
-6.7	-3.8	-0.4	-4.9	-3.1	-2.7	-6.3
###	-6.8	-2.6	###	-7.1	-4.5	-8.0
###	-4.8	-1.7	-7.5	-4.1	-1.5	-7.1
###	-7.9	-2.7	###	-9.1	-5.9	###
###	###	-4.0	###	###	-9.3	###
###	###	-4.2	###	###	-8.0	###
###	###	-7.3	###	###	###	###
###	###	-7.0	###	###	###	###
###	###	###	###	###	###	###

Gene flow into Spain_EN?

Test: Spain_EN

Ref1: Corded_Ware

Ref2: Spain_MN

U	V	W	X	Y	T	U
Spain_MN	Stuttgart	LBKT_EN	LBK_EN	HungaryGamba_EN	Baalberge_MN	Spain_MN
###	-8.5	-3.2	###	###	-9.5	###
###	-3.2	-1.0	###	-4.1	-6.3	###
###	-8.6	-3.5	###	###	-8.3	###
###	-8.8	-3.4	###	-9.1	-6.3	###
###	-5.4	-2.1	-8.7	-5.1	-4.8	-8.8
###	-5.2	-1.6	###	-5.1	-3.3	###
###	-6.5	-2.4	###	-8.1	-6.9	###
-8.9	-2.6	-1.3	-5.9	-2.1	-3.3	-8.1

Gene flow into Spain_EN?

Test: Spain_EN

Ref1: Corded_Ware

Ref2: Spain_MN

```

fixed pat  wt  dof      chisq      tail prob
      00   0   11      20.874      0.0346956      -0.084      1.084  infeasible
      01   1   12     175.340           0          1.000      -0.000
      10   1   12      21.887      0.038804          0.000      1.000

best pat:           00           0      - -
best pat:           10      0.038804  chi(nested):      1.013 p-value for nested model:
0.314247

```

-2.3	-0.5	nan	-1.0	0.1
-5.6	3.0	1.0	nan	4.9
-8.8	-0.3	-0.1	-4.9	nan
-2.1	2.6	0.4	0.6	2.5
nan	nan	nan	nan	nan
###	-5.6	-2.1	-9.5	-6.1
-6.7	-3.8	-0.4	-4.9	-3.1
###	-6.8	-2.6	###	-7.1
###	-4.8	-1.7	-7.5	-4.1
###	-7.9	-2.7	###	-9.1
###	###	-4.0	###	###
###	###	-4.2	###	###
###	###	-7.3	###	###
###	###	-7.0	###	###
###	###	###	###	###

0.0	-2.9
-1.2	-7.7
-0.7	###
-0.9	-2.3
nan	nan
-1.8	-5.6
-2.7	-6.3
-4.5	-8.0
-1.5	-7.1
-5.9	###
-9.3	###
-8.0	###
###	###
###	###
###	###

Should try more than one.

Maybe bad source population?

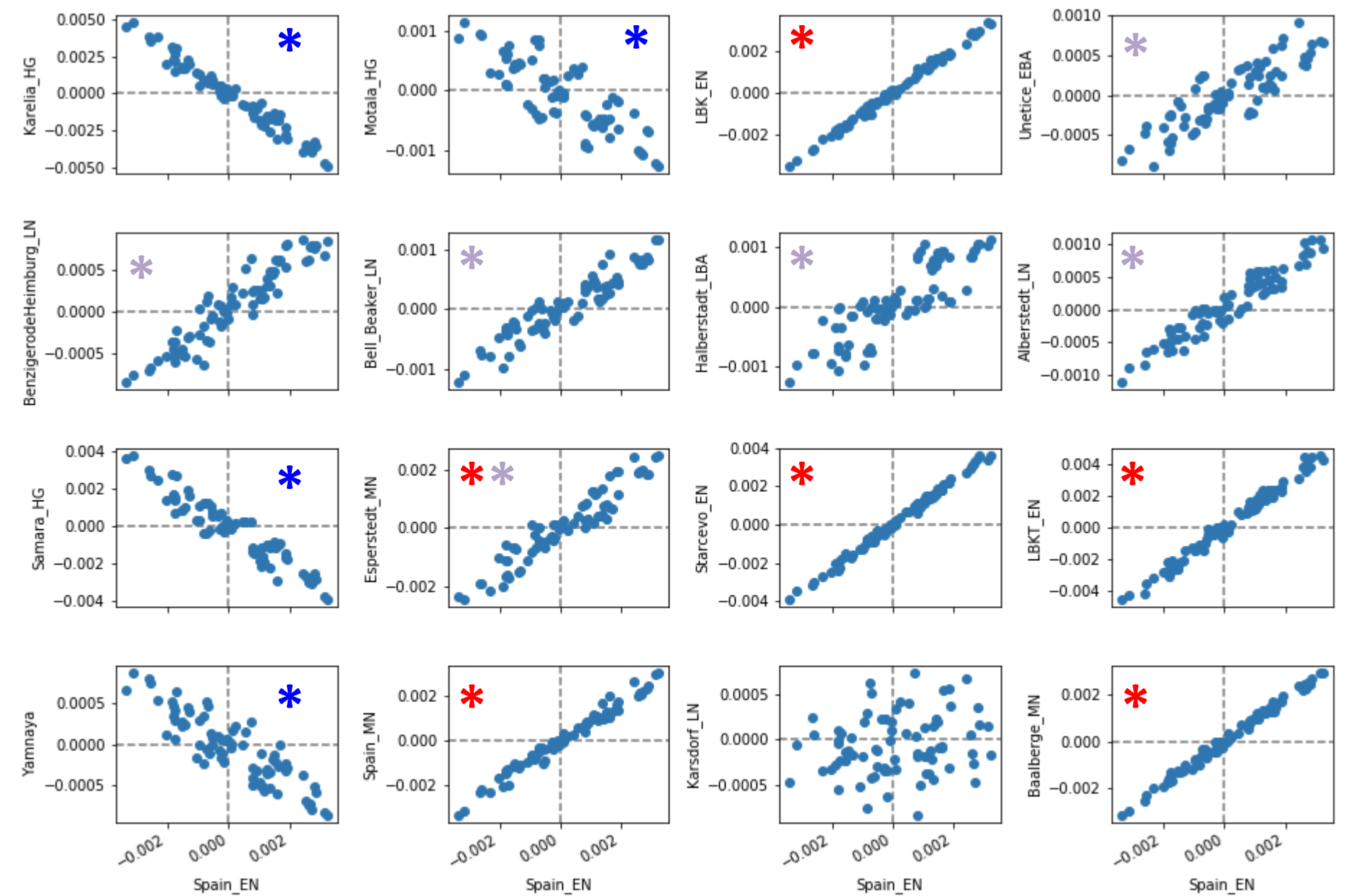
Either way, do not see gene flow into Spain_EN, and with timing, geography, other populations, likely only gene flow into Corded Ware and not other way around.

More intuition, maybe?
F4(Corded Ware, X/Y; Out, Out)

Tight positive – x and y clade
relative to Corded Ware

Negative – admixture!

Loose positive – different
relationship between x and y
with Corded Ware



Class Activity – Try a three reference qpAdm test.

- What are the three sources of European ancestry?

Class Activity – Try a three reference qpAdm test.

- What are the three sources of European ancestry?
- Sources: LBK_EN, Loschbour, Yamnaya
- Test Spain_EN, Esperstedt_MN, Corded_Ware_LN, or Unetice_EBA
 - Try to see if you can figure out how to read the table in the log file for three populations!
 - How much of each type of ancestry do you find?

Three Source Case

fixed	pat	wt	dof	chisq	tail prob				
	000	0	10	12.930	0.227625	0.590	0.685	-0.275	infeasible
	001	1	11	14.080	0.228623	0.707	0.293	0.000	
	010	1	11	16.819	0.113331	0.840	-0.000	0.160	
	100	1	11	15.543	0.158965	0.000	2.144	-1.144	infeasible
	011	2	12	20.569	0.0570605	1.000	-0.000	-0.000	
	101	2	12	34.297	0.000605429	0.000	1.000	-0.000	
	110	2	12	111.478	3.05099e-18	0.000	-0.000	1.000	
best pat:			000		0	-	-		
best pat:			001	0.228623	chi(nested):	1.150	p-value for nested model:		
				0.283597					
best pat:			011	0.0570605	chi(nested):	6.489	p-value for nested model:		
				0.0108521					
std. errors:				0.319	0.844	0.546			

000 is the full mixture case.

010, 001, 100 means two source mixture model.

110, 011, 101 means all ancestry from one source

1. Look at “tail prob” – if $p > 0.05$, then that model is possible.
2. If 000, 010, 001, 100 are possible, then look at nested model – is ‘000’ better than two source models, better than the one source models??

p-value for $p_{\text{nest}} < 0.05$, means reject model in same line, preferring model in above line.

3. Look at f_1, f_2, f_3 .

qpAdm(Spain_EN; LBK_EN, Loschbour, Yamnaya)

fixed	pat	wt	dof	chisq	tail prob				
	000	0	10	9.140	0.51884	0.936	0.093	-0.030	infeasible
	001	1	11	9.396	0.58539	0.948	0.052	0.000	
	010	1	11	9.678	0.55954	0.973	-0.000	0.027	
	100	1	11	23.096	0.0171314	0.000	2.491	-1.491	infeasible
	011	2	12	9.843	0.629774	1.000	-0.000	-0.000	
	101	2	12	79.631	4.85302e-12	0.000	1.000	-0.000	
	110	2	12	339.060	0	0.000	-0.000	1.000	
best pat:			000		0	-	-		
best pat:			001		0.58539	chi(nested):	0.256	p-value for nested model:	
					0.613109				
best pat:			011		0.629774	chi(nested):	0.447	p-value for nested model:	
					0.503992				
std. errors:				0.076	0.163	0.111			

qpAdm(Esperstedt_MN; LBK_EN, Loschbour, Yamnaya)

fixed	pat	wt	dof	chisq	tail prob				
	000	0	10	9.360	0.498287	0.679	0.421	-0.101	infeasible
	001	1	11	10.022	0.528452	0.707	0.293	0.000	
	010	1	11	14.436	0.209782	0.846	0.000	0.154	
	100	1	11	16.502	0.123488	0.000	1.871	-0.871	infeasible
	011	2	12	17.992	0.115945	1.000	-0.000	-0.000	
	101	2	12	32.028	0.00137028	0.000	1.000	0.000	
	110	2	12	121.960	2.5117e-20	0.000	-0.000	1.000	
best pat:			000		0	-	-		
best pat:			001	0.528452	chi(nested):	0.661	p-value for nested model:		
				0.41612					
best pat:			011	0.115945	chi(nested):	7.970	p-value for nested model:		
				0.00475557					
std. errors:				0.150	0.316	0.202			

qpAdm(Corded_Ware_LN; LBK_EN, Loschbour, Yamnaya)

fixed	pat	wt	dof	chisq	tail prob				
	000	0	10	10.210	0.422307	0.172	0.159	0.668	
	001	1	11	24.316	0.0114639	-0.226	1.226	-0.000	infeasible
	010	1	11	11.094	0.435398	0.236	-0.000	0.764	
	100	1	11	13.988	0.233652	0.000	0.440	0.560	
	011	2	12	216.486	0	1.000	-0.000	-0.000	
	101	2	12	25.256	0.0136574	0.000	1.000	0.000	
	110	2	12	31.284	0.00178256	0.000	-0.000	1.000	
best pat:			000	0.422307		-	-		
best pat:			010	0.435398	chi(nested):	0.885	p-value for nested model:		
				0.346901					
best pat:			101	0.0136574	chi(nested):	14.161	p-value for nested model:		
				0.000167793					
std. errors:				0.107	0.236	0.148			

qpAdm(Unetice_LBA; LBK_EN, Loschbour, Yamnaya)

fixed	pat	wt	dof	chisq	tail prob			
	000	0	10	6.449	0.776267	0.275	0.303	0.422
	001	1	11	17.490	0.0941965	0.048	0.952	0.000
	010	1	11	15.390	0.165347	0.395	-0.000	0.605
	100	1	11	15.209	0.173125	-0.000	0.848	0.152
	011	2	12	225.464	0	1.000	-0.000	-0.000
	101	2	12	16.072	0.187958	0.000	1.000	0.000
	110	2	12	92.688	1.48661e-14	0.000	-0.000	1.000
best pat:			000	0.776267	-	-		
best pat:			100	0.173125	chi(nested):	8.760	p-value for nested model:	
				0.00307841				
best pat:			101	0.187958	chi(nested):	0.863	p-value for nested model:	
				0.352902				
std. errors:				0.063	0.129	0.087		