



# CS 68: BIOINFORMATICS

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Swarthmore College  
Spring 2018



# Outline: Feb 26

- BWT and FM-Index runtime
- Recap UPGMA
- Neighbor-Joining (NJ)

## Notes:

- Office hours TODAY 3-5pm
- Create “cheat-sheet” for midterm
- Lab 5 released right after midterm

BWT and FM-Index runtime

# BWT and FM-Index runtime

- Building the FM-Index: dominated by sorting the rotations (cyclic permutations). There are actually linear time algorithms for this, but we will assume a standard sorting algorithm so  $O(G \log G)$  where  $G$  is the length of the reference.
- Creating  $M$ ,  $occ$ , and  $A$  are all linear.
- Pattern matching from Lab 4?

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- Creating  $M$ ,  $occ$ , and  $A$  are all linear.
- Pattern matching:  $O(n * L)$ 
  - *Linear in the length of the pattern ( $L$ )*
  - *Linear in the number of patterns/reads ( $n$ )*
  - *Constant in the length of the genome ( $G$ )*

# Recap UPGMA

# Recap questions: discuss with a partner

- 1) How do we define a tree metric?
- 2) True or False: every dissimilarity map is a tree metric.
- 3) How do we define an ultrametric? (both theoretically and intuitively)
- 4) True or False: every tree metric is an ultrametric.
- 5) What biological assumption(s) are we making when creating ultrametric trees like those produced by UPGMA?
- 6) What two biological factors might make ultrametric trees an unrealistic assumption?

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A dissimilarity map  $\delta$  is a tree metric if the sum of the edge weights between  $A$  and  $B$  is equal to  $\delta(A,B)$ .

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3) How do we define an ultrametric? (both theoretically and intuitively)

- The distance from the root to each leaf is the same.
- 3-point condition: For all distinct  $A,B,C$ ,  $\delta(A,B) \leq \max\{\delta(A,C), \delta(B,C)\}$
- Intuitively this means out of these three distances, two are equal and one is less.

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- Mutation rates differ significantly across species.
- Natural selection (both positive and negative) can change the tempo of evolution.

# Bonus questions

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- What is the runtime of UPGMA in terms of the number of samples  $n$ ?

During each iteration we must do  $O(n^2)$  work to compute the new matrix of distances. We merge two nodes each iteration, so we have  $O(n)$  iterations total. This gives us a runtime of  $O(n^3)$ , which can be improved by reusing some distances from the previous iteration.

Next phylogenetic tree method:  
Neighbor-Joining (NJ)

# Notes about UPGMA vs NJ

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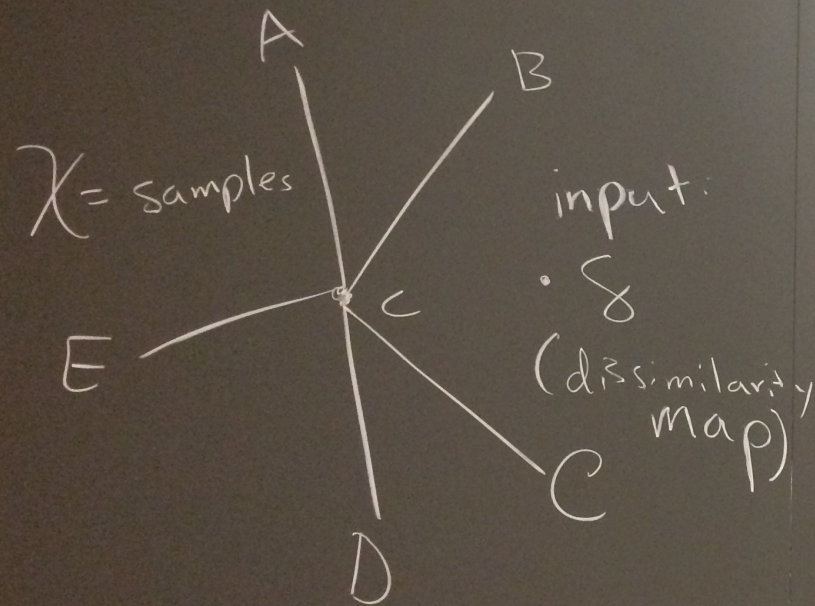
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- NJ creates unrooted trees (direction of evolution is not apparent on all branches), while UPGMA creates rooted trees.
- NJ is much better for representing multi-species evolution and in general creates more realistic trees that better approximate the original dissimilarity map.

# Neighbor-Joining



## Initialization

- Create a "star tree"
- $N_c = \text{Set of neighbors of } c$   
(ex.  $N_c = \{A, B, C, D, E\}$ )
- $n = |N_c|$  (ex.  $n = 5$ )
- representation of distances  
 $d = \delta$

Iterate

(a) choose  $f \neq g$  that  
minimize  $Q$ -criteria

$$Q(i, j) = (n-2) \cdot d(i, j) - \underbrace{S_i - S_j}$$

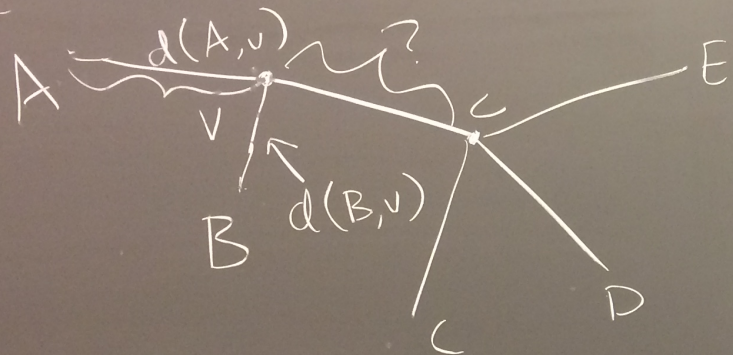
$$S_i = \sum_{k \in N_c} d(i, k)$$

how far  
 $i \neq j$  are  
from everything  
else

must compute  $Q$  for  
all pairs in  $N_c$

$$O(n^2)$$

(b) form new vertex  $v$



$v \in N_c$ , but not  $f, g$

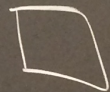
$d[F][v]$   $\ell_F$

$$d(f, v) = \underbrace{\frac{1}{2} d(f, g)}_{\text{distance matrix}} + \underbrace{\frac{1}{2(n-2)} [S_f - S_g]}_{\text{UPGMA}}$$

different  $f \neq g$   
are w.r.t.  
the other  
samples

$$d(g, v) = \frac{1}{2} d(f, g) + \frac{1}{2(n-2)} [S_g - S_f]$$

$d$	$f$	$g$	$v$
$f$			
$g$			
$v$			



$$d(f,v) + d(g,v) = d(f,g) \quad \checkmark$$

$$(c) \quad \forall i \in N_c$$

$$d(i,v) = \frac{1}{2} [d(f,i) - d(f,v)^{\star}] + \frac{1}{2} [d(g,i) - d(g,v)^{\star}]$$