

Evolution: Deeper into phylogenetics

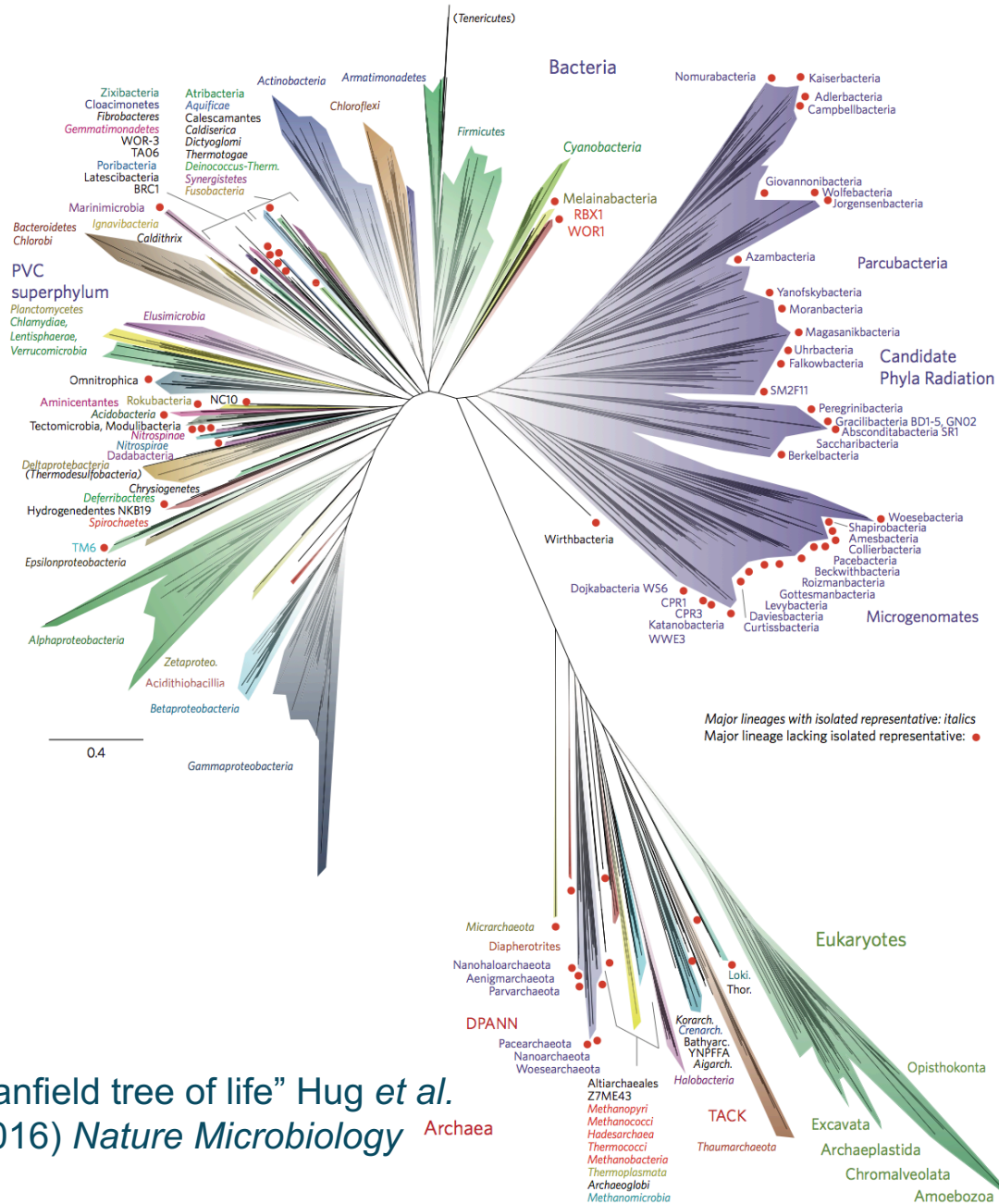
Andrew Rominger

Omidyar Postdoctoral Fellow

Santa Fe Institute



- How to infer a phylogeny
- Why the deep past is difficult



“Banfield tree of life” Hug *et al.* (2016) *Nature Microbiology*

Inferring a phylogeny

Molecular Phylogenetics Protocol



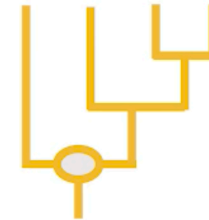
1) Collect homologous protein sequences



2) Multiple sequence alignment

3) Phylogeny estimation*

- Maximum parsimony
- Distance methods
- Maximum likelihood
- Bayesian phylogenetics



4) Test the reliability of phylogenetic estimate(s)*

5) Interpretation and application of phylogenies



Inferring a phylogeny

Molecular Phylogenetics Protocol



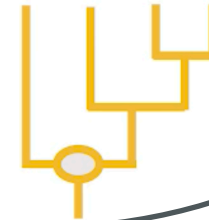
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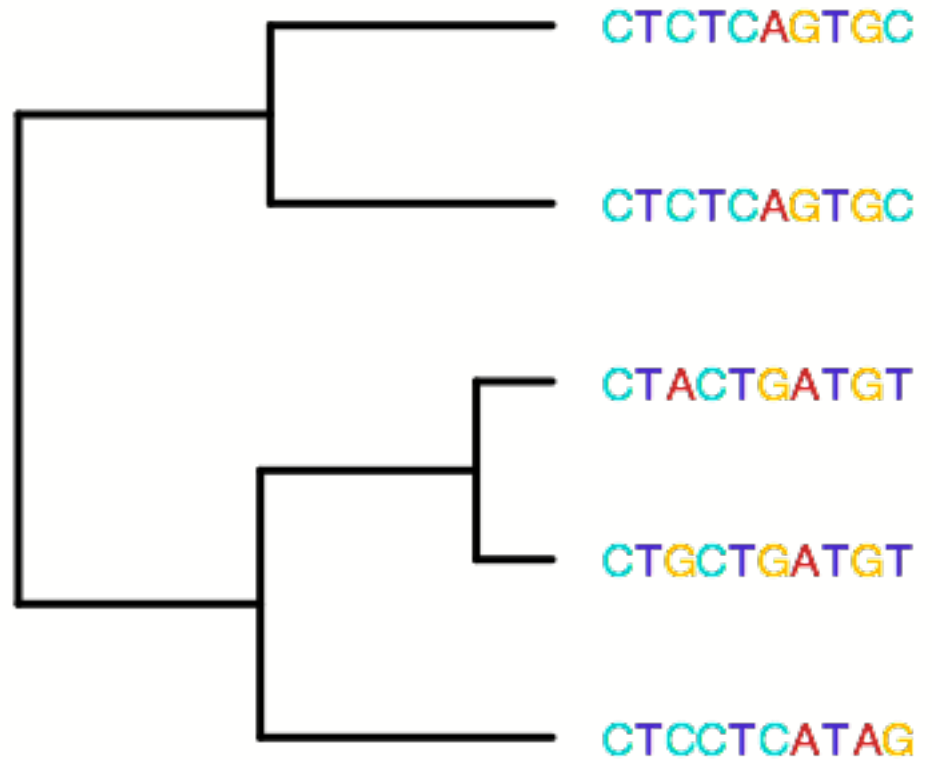
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Inferring a phylogeny

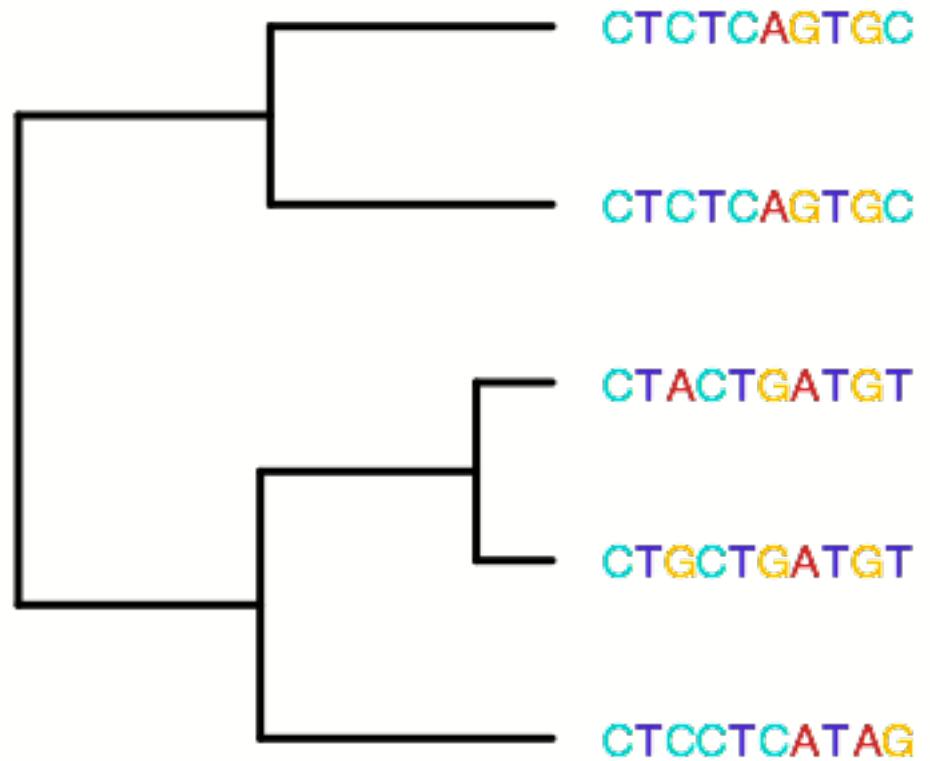
Goal: find the process leading to the data



Inferring a phylogeny

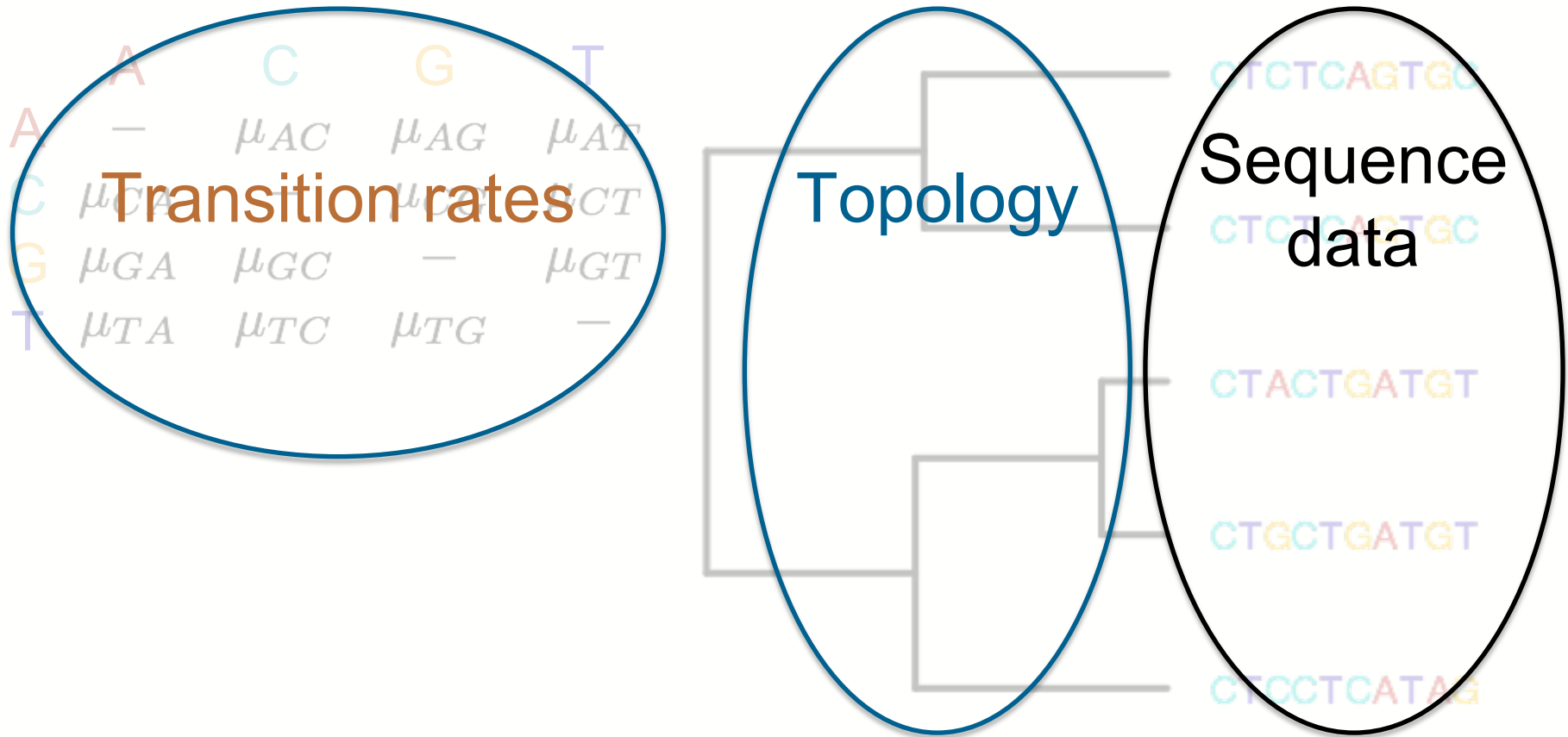
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	A	C	G	T
A	—	μ_{AC}	μ_{AG}	μ_{AT}
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Inferring a phylogeny

Goal: find the process leading to the data



Inferring a phylogeny

Goal: find the process leading to the data

Likelihood:

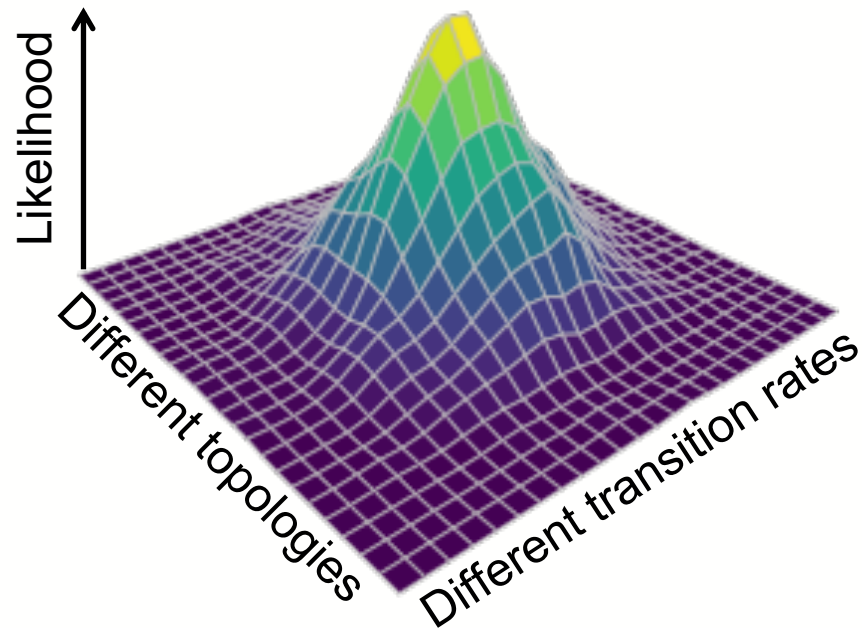
$$P(\text{Sequence data} \mid \text{Transition rates, Topology})$$

Inferring a phylogeny

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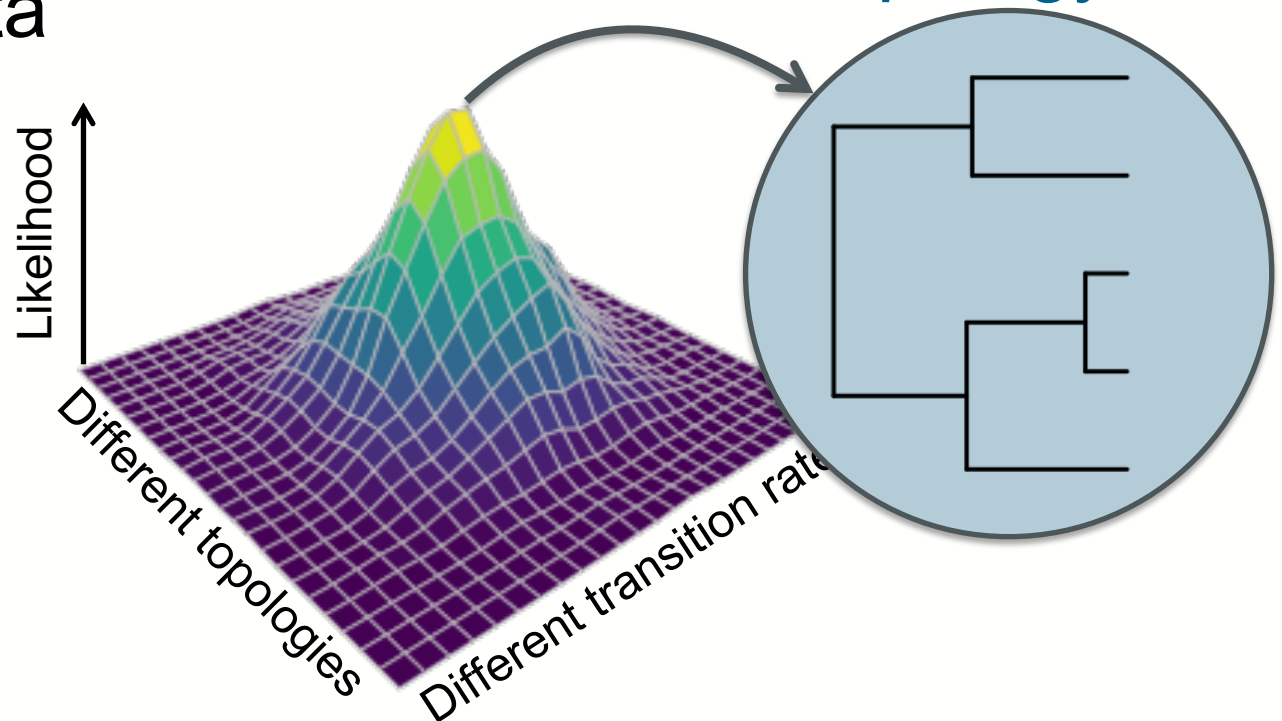


Inferring a phylogeny

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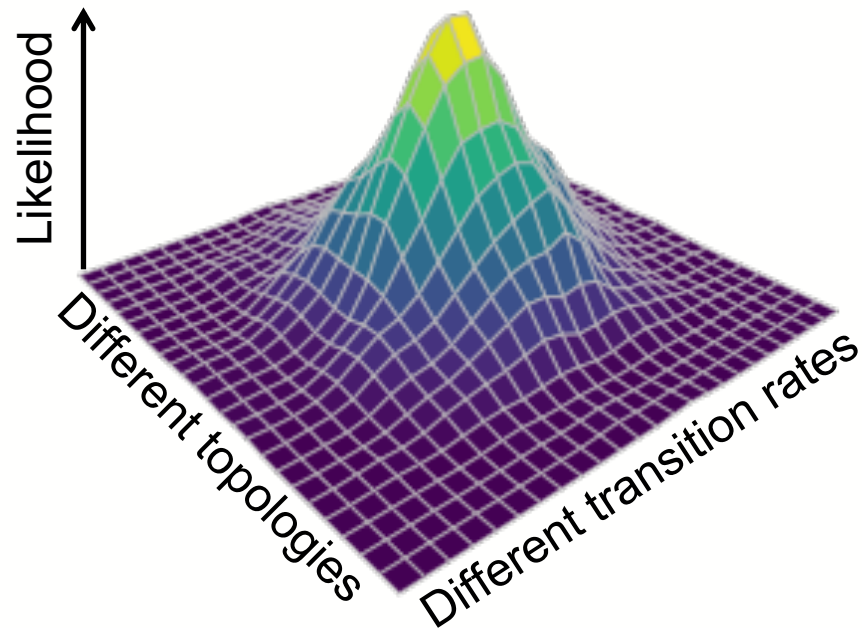
$$P(\text{Sequence data} \mid \text{Transition rates, Topology})$$



Inferring a phylogeny

Goal: find the process leading to the data

Bayesian inference: traversing the complex topology and rate matrix space lends itself to Bayesian algorithms



Inferring a phylogeny

Goal: find the process leading to the data

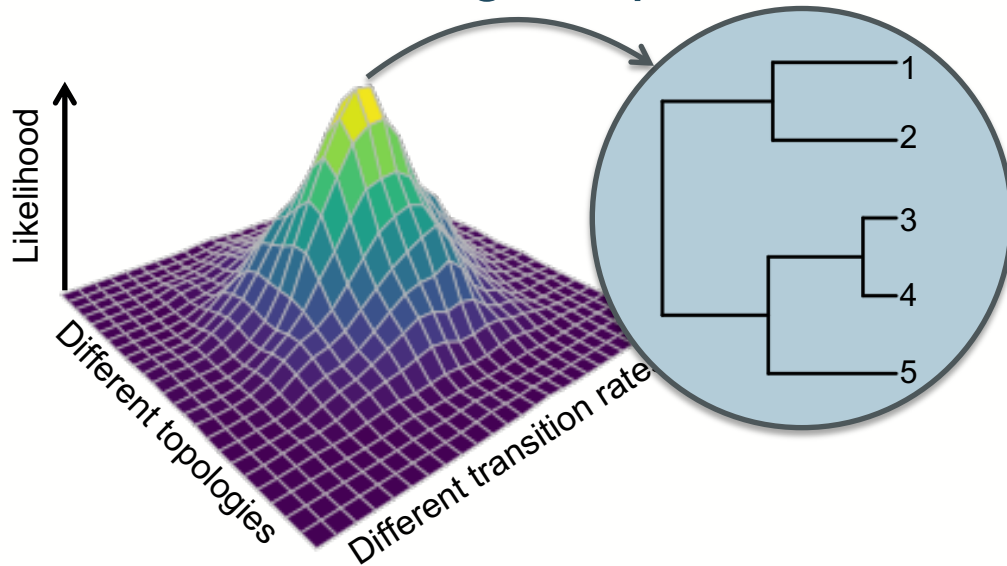
Information on parsimony and distance-based methods are left for the Further Reading section

Inferring a phylogeny

Caveats: inferring the past is hard

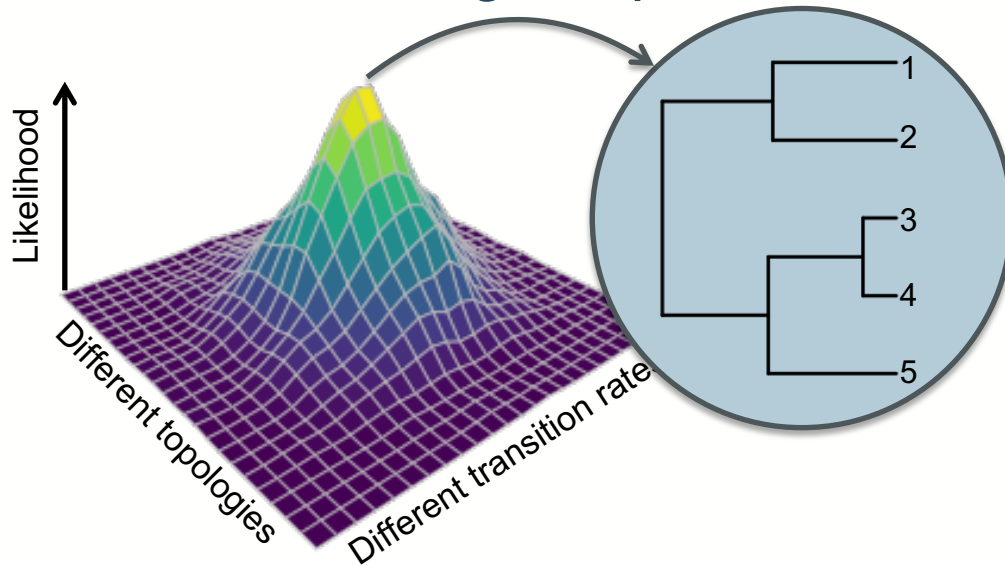
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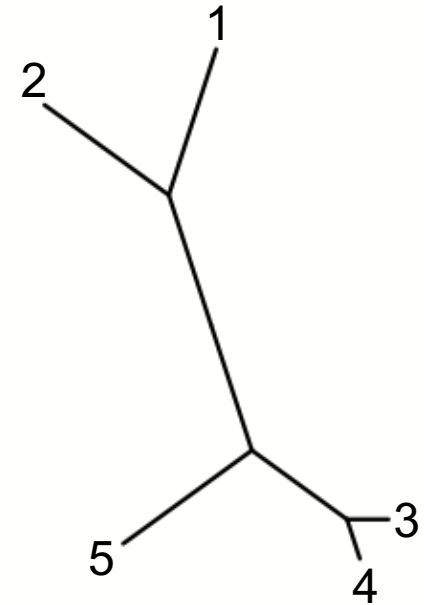


Inferring a phylogeny

Caveats: inferring the past is hard



“Unrooted” phylogeny arises because mutation process is reversible in our models



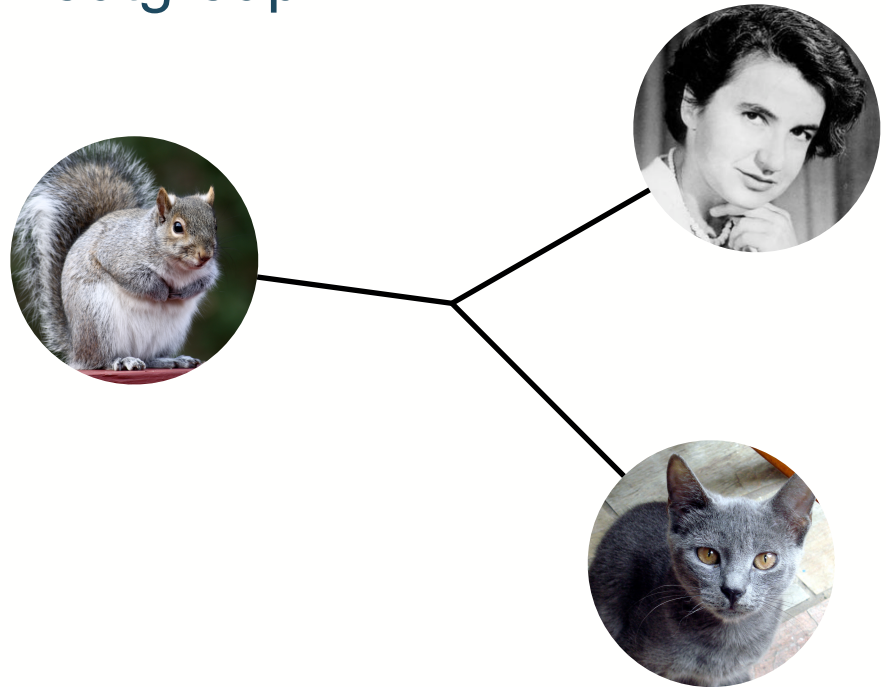
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Inferring a phylogeny

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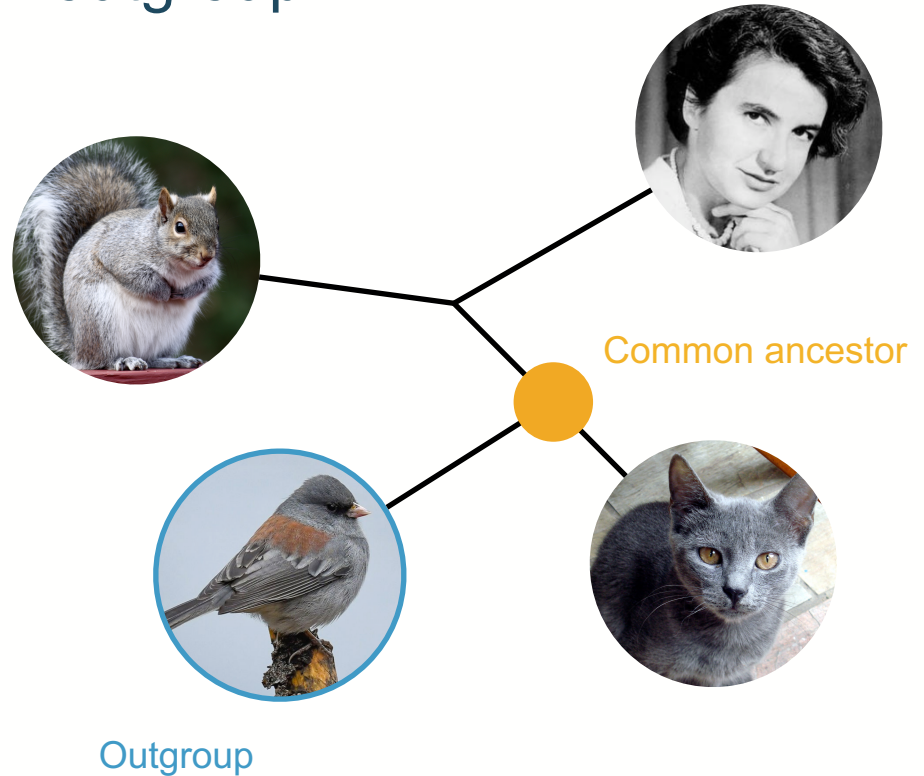
- Rooting the tree with an outgroup



Inferring a phylogeny

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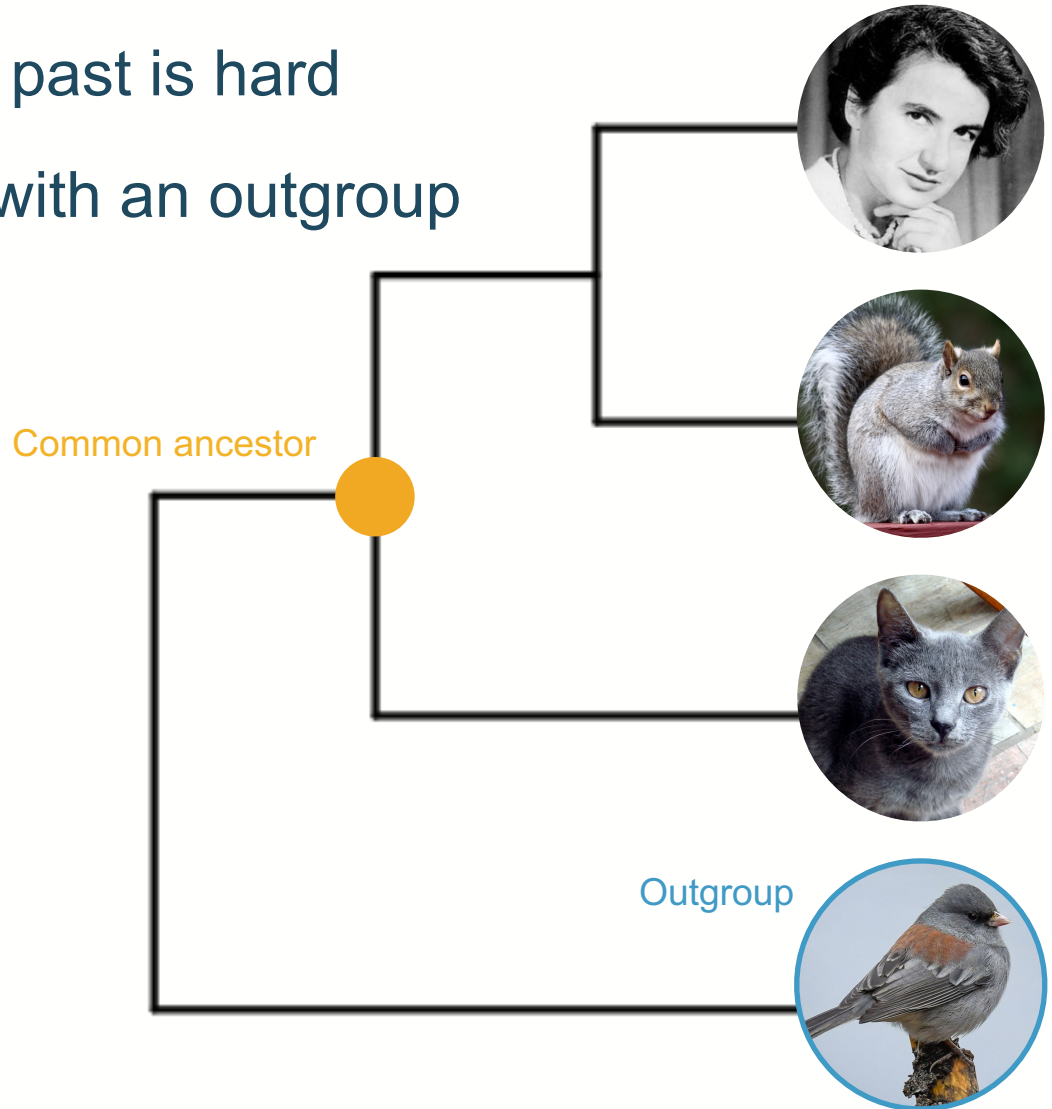
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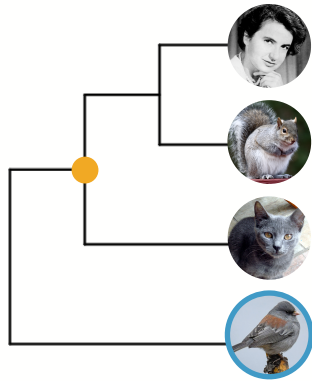
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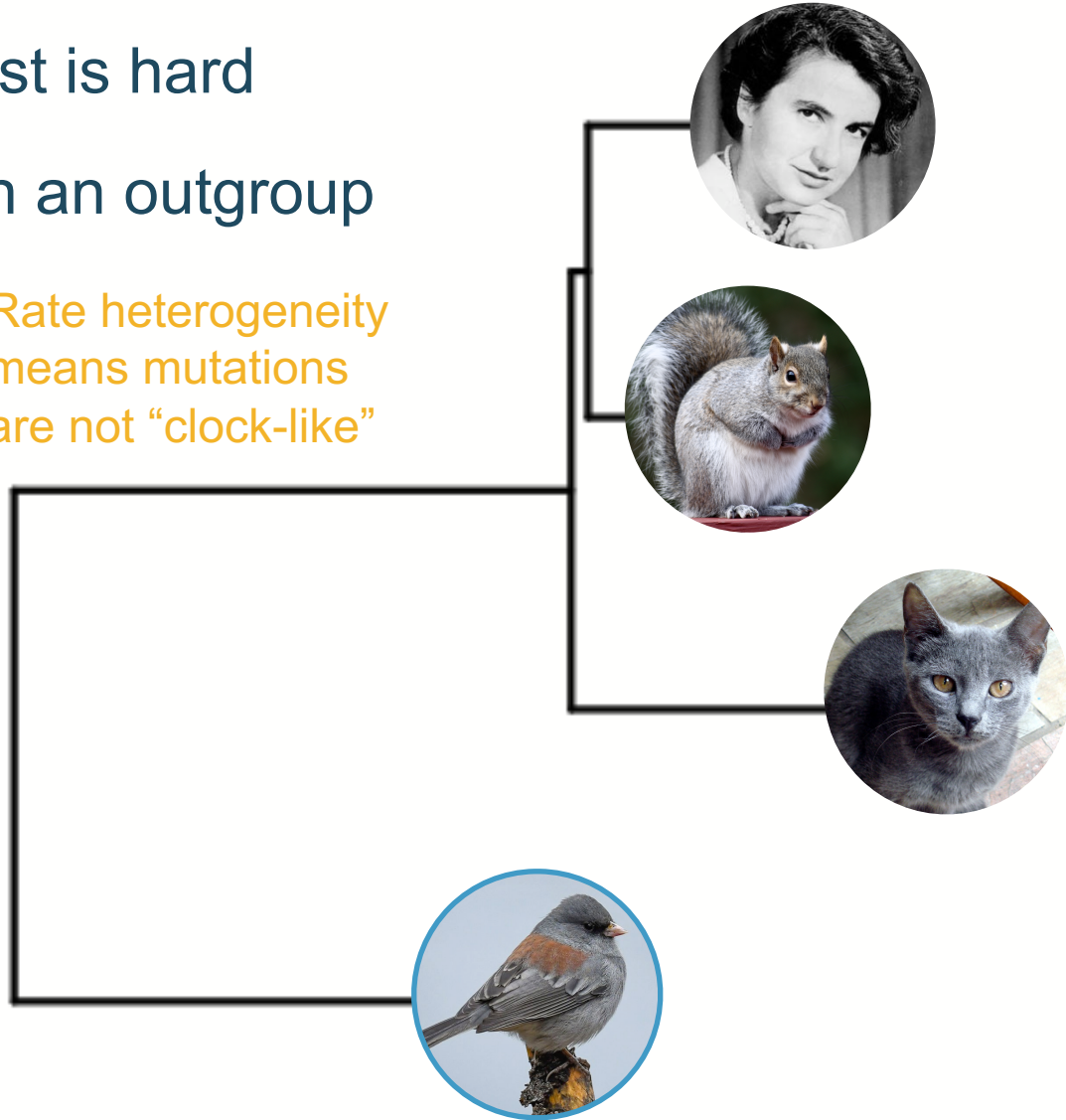
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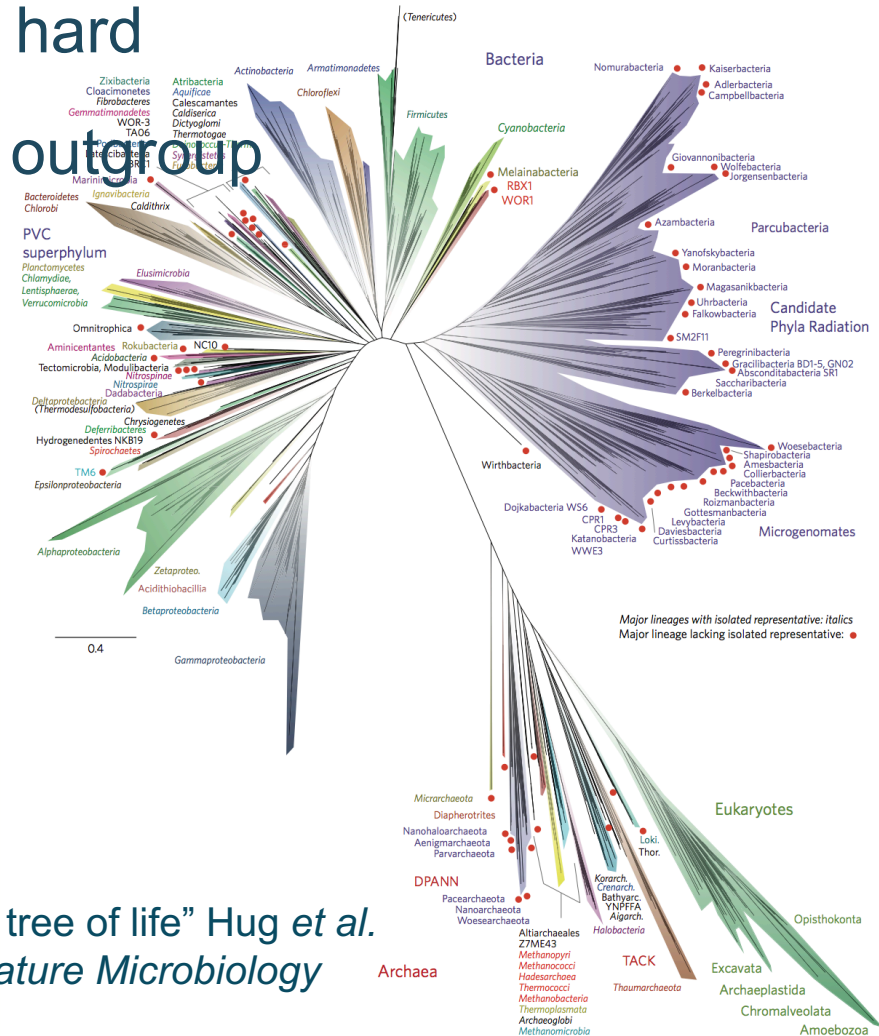
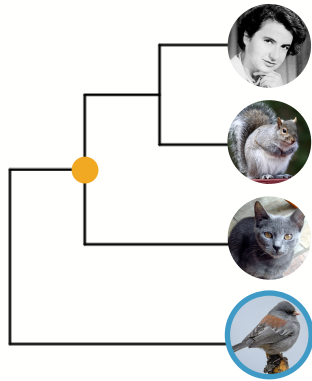
Rate heterogeneity
means mutations
are not “clock-like”



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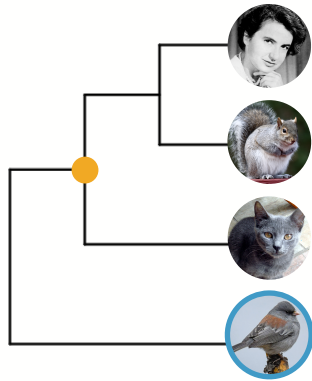


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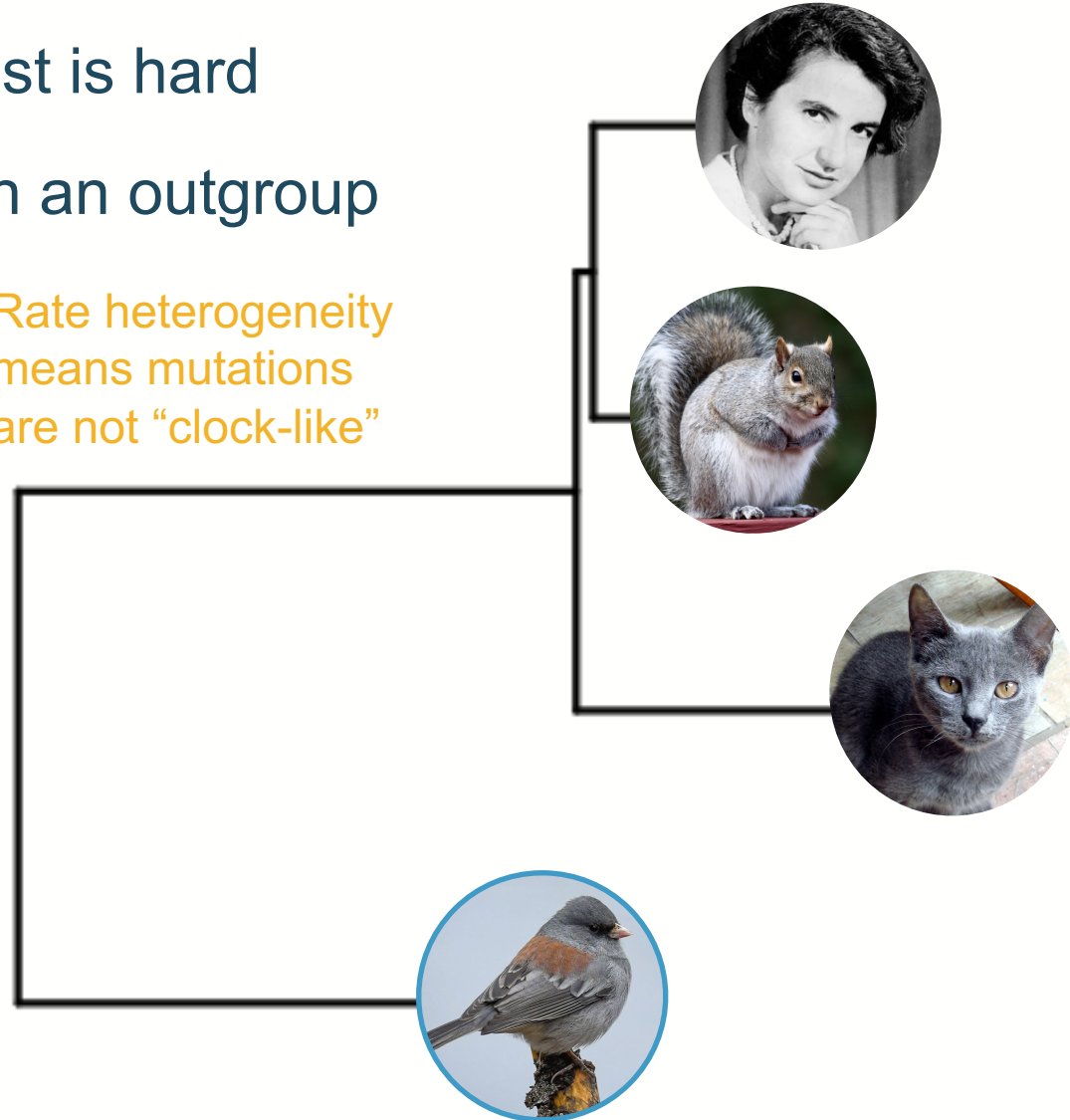
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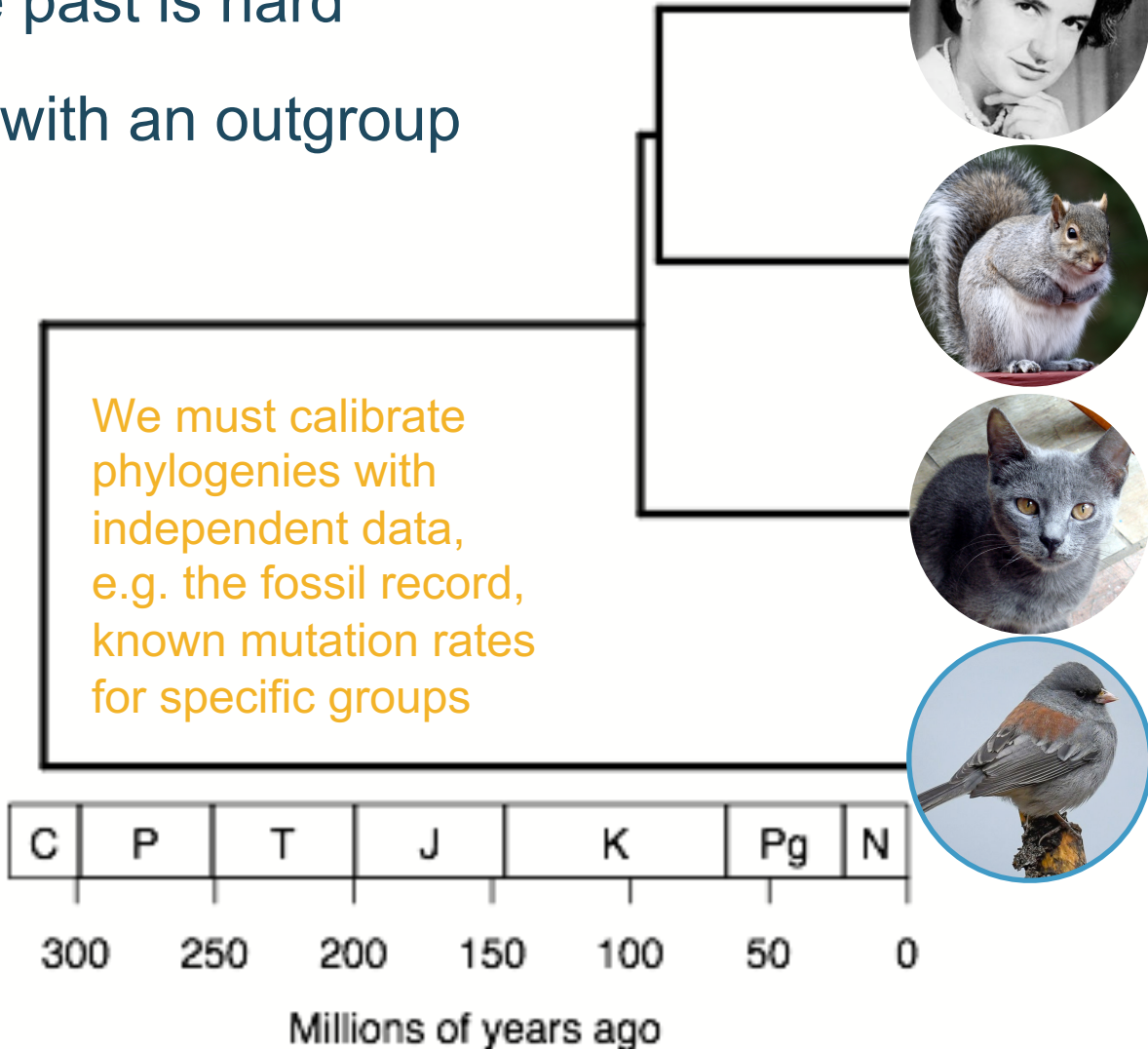
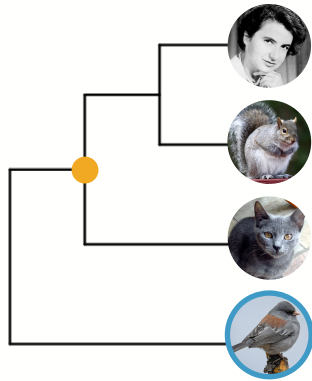
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Inferring a phylogeny

Caveats: inferring the past is hard

- Rooting the tree with an outgroup
- Time calibration



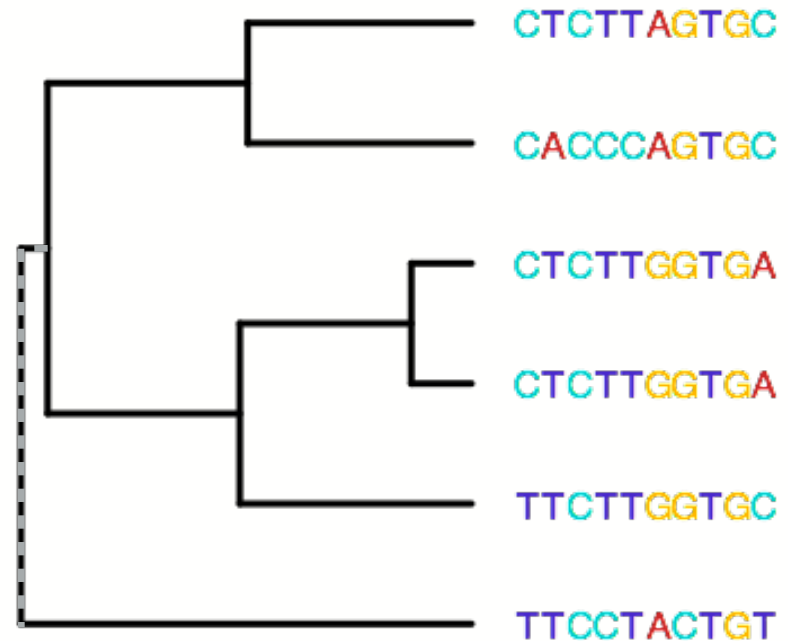
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Long branches “attract” each other



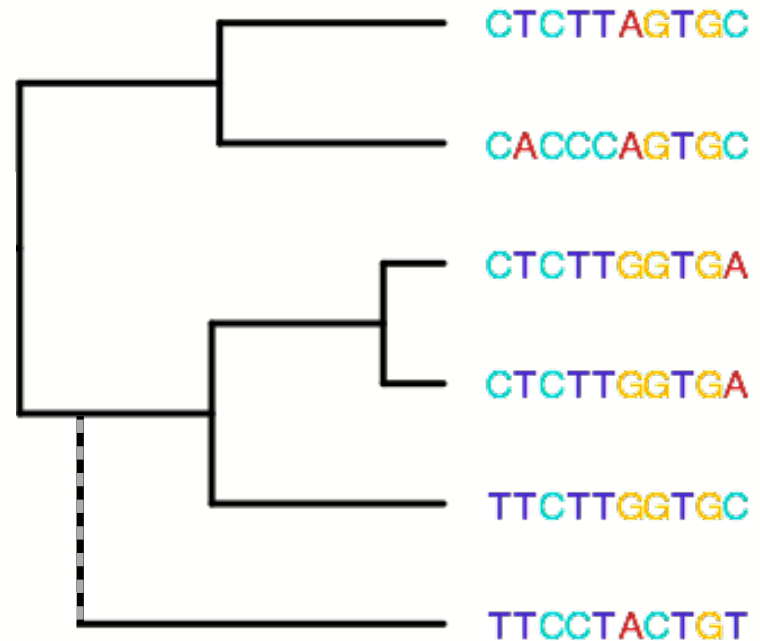
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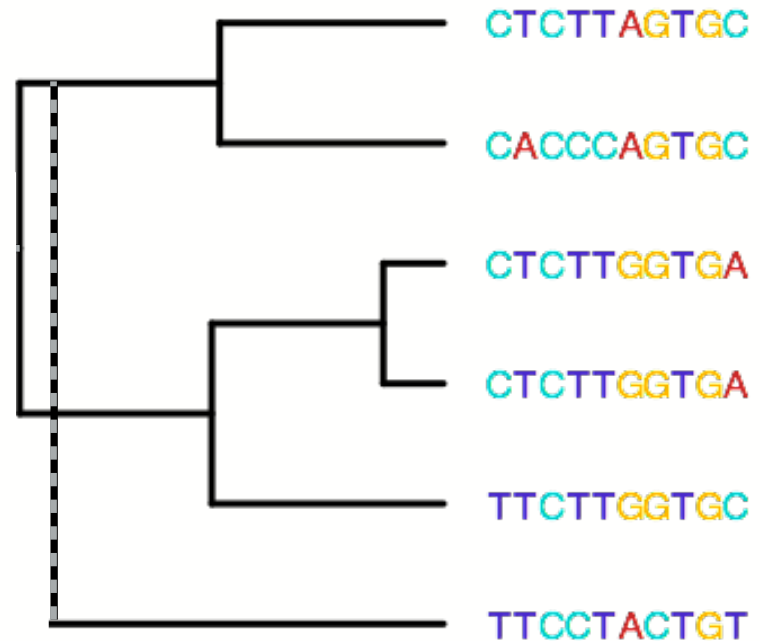
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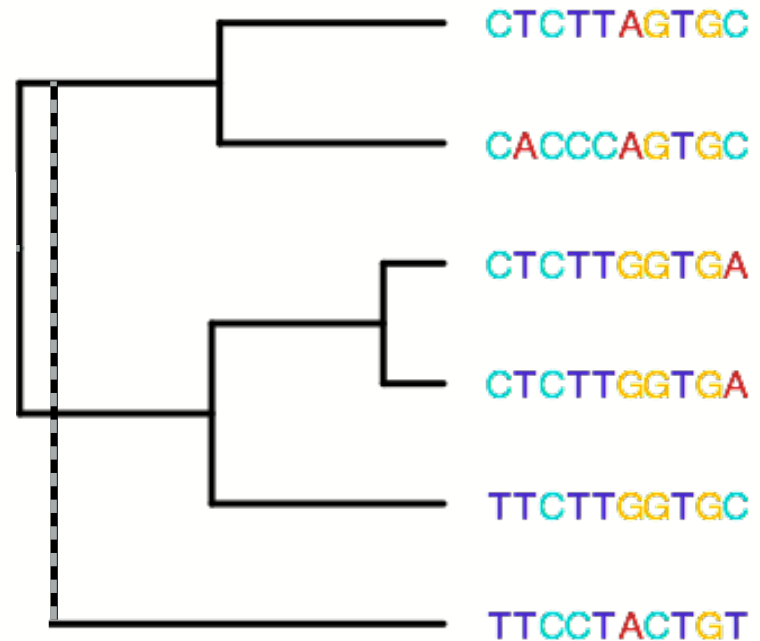
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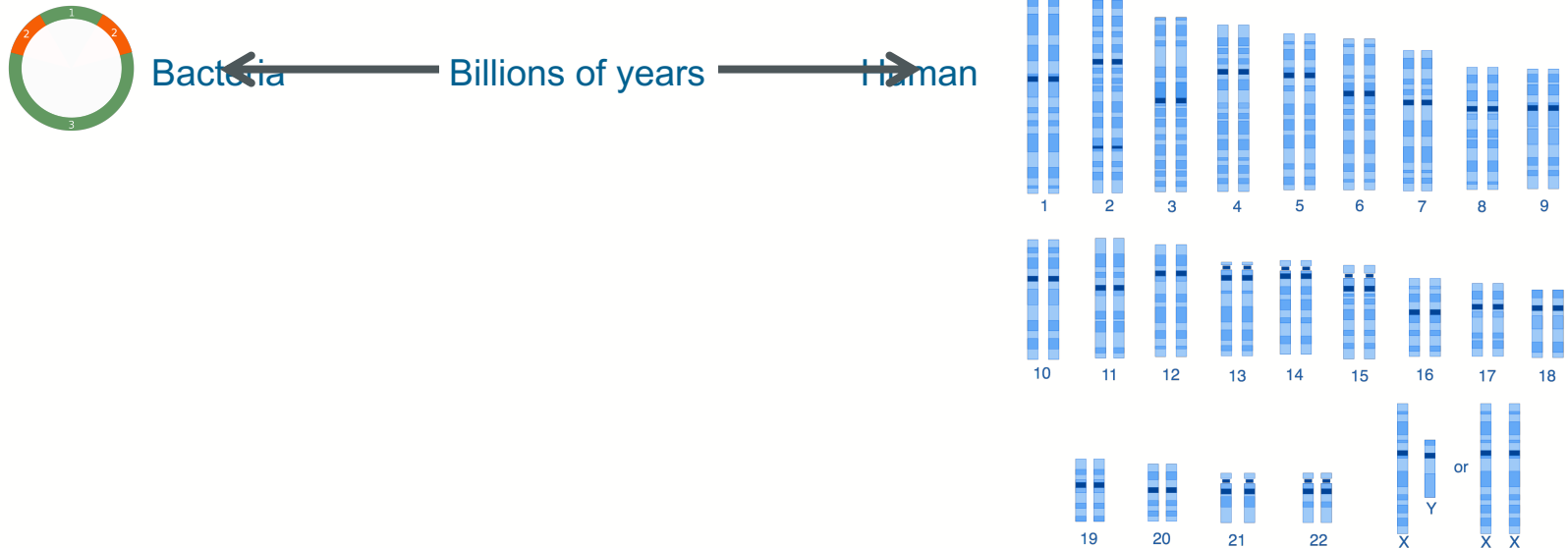
- Rooting the tree with an outgroup
- Time calibration
- Long branch attraction



Inferring a phylogeny

Caveats: inferring the past is hard (especially in deep time)

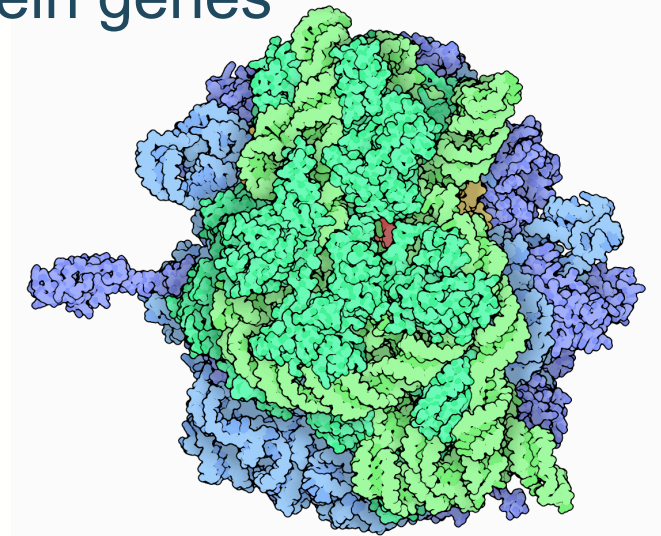
- Rooting the tree with an outgroup
- Time calibration
- Long branch attraction
- What genetic information goes back to LUCA?



Inferring a phylogeny

Caveats: inferring the past is hard (especially in deep time)

- Rooting the tree with an outgroup
- Time calibration
- Long branch attraction
- What genetic information goes back to LUCA?
 - Ribosomal RNA and protein genes



David Goodsell

References and Suggested Reading

Huelsenbeck, J. P., & Ronquist, F. (2001). MRBAYES: Bayesian inference of phylogenetic trees. *Bioinformatics*, 17, 754-755.

Huelsenbeck, J. P., & Crandall, K. A. (1997). Phylogeny estimation and hypothesis testing using maximum likelihood. *Annual Review of Ecology and Systematics*, 28, 437-466.

Hug, L. A., Baker, B. J., Anantharaman, K., ... & Banfield, J. F. (2016). A new view of the tree of life. *Nature Microbiology*, 1, 16048.

Philippe, H., Zhou, Y., Brinkmann, H., Rodrigue, N., & Delsuc, F. (2005). Heterotachy and long-branch attraction in phylogenetics. *BMC Evolutionary Biology*, 5, 50.

Quast, C., Pruesse, E., Yilmaz, P., ... & Glöckner, F. O. (2012). The SILVA ribosomal RNA gene database project: improved data processing and web-based tools. *Nucleic Acids*