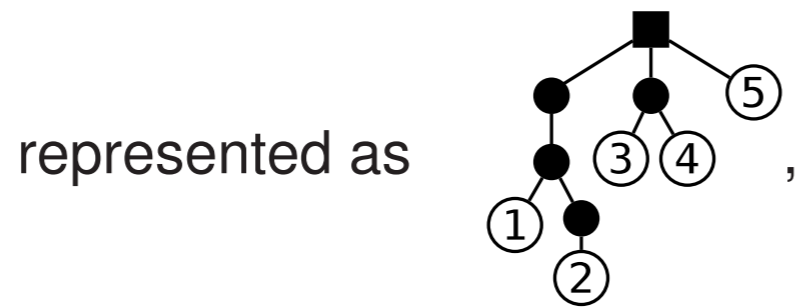


MOTIVATION: LIKELIHOODS WITH AN UNKNOWN ANCESTRAL TREE

Given a set of aligned sequences, e.g.

Sequence 1: ... T ... G ... A ... A ...
Sequence 2: ... T ... G ... A ... G ...
Sequence 3: ... A ... A ... A ... T ... A ...
Sequence 4: ... A ... A ... A ... T ... A ...
Sequence 5: ... A ... A ... A ... A ... A ...



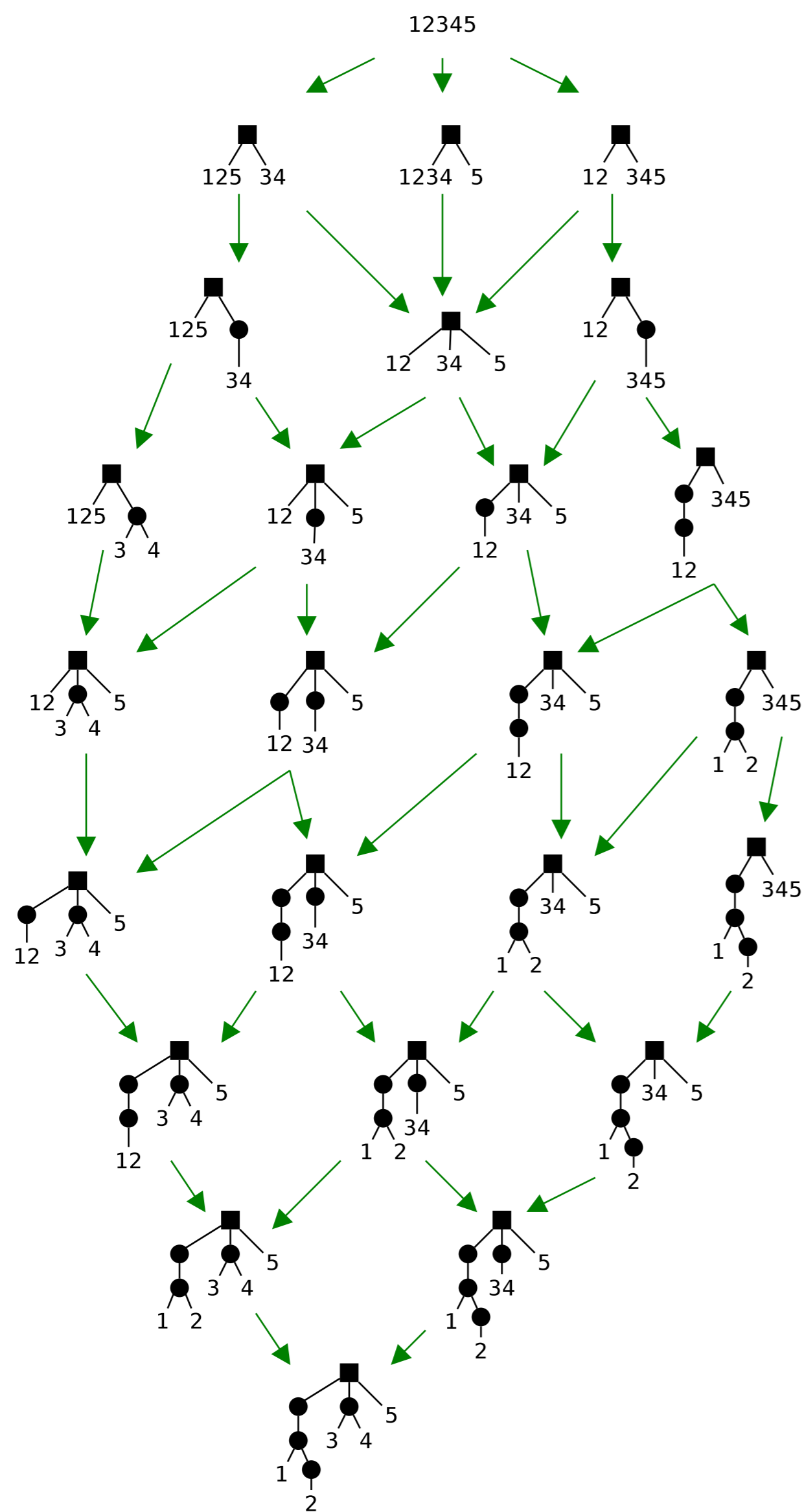
the probability of having evolved from some initial sequence (here ...A...A...A...A...) may be expressed:

$$\mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right) = \sum_{x \in \text{Histories} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right)} \mathbb{P}(x)$$

which in turn may be expressed by conditioning on the most recent event:

$$\mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right) = \frac{1}{\theta + 4} \mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right) + \frac{\theta}{\theta + 4} \frac{1}{5} \mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right).$$

If we apply this conditioning-trick recursively, computing $\mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right)$ reduces to computing a weighted sum over all paths from "12345" to $\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix}$ in the *ancestral graph* below:



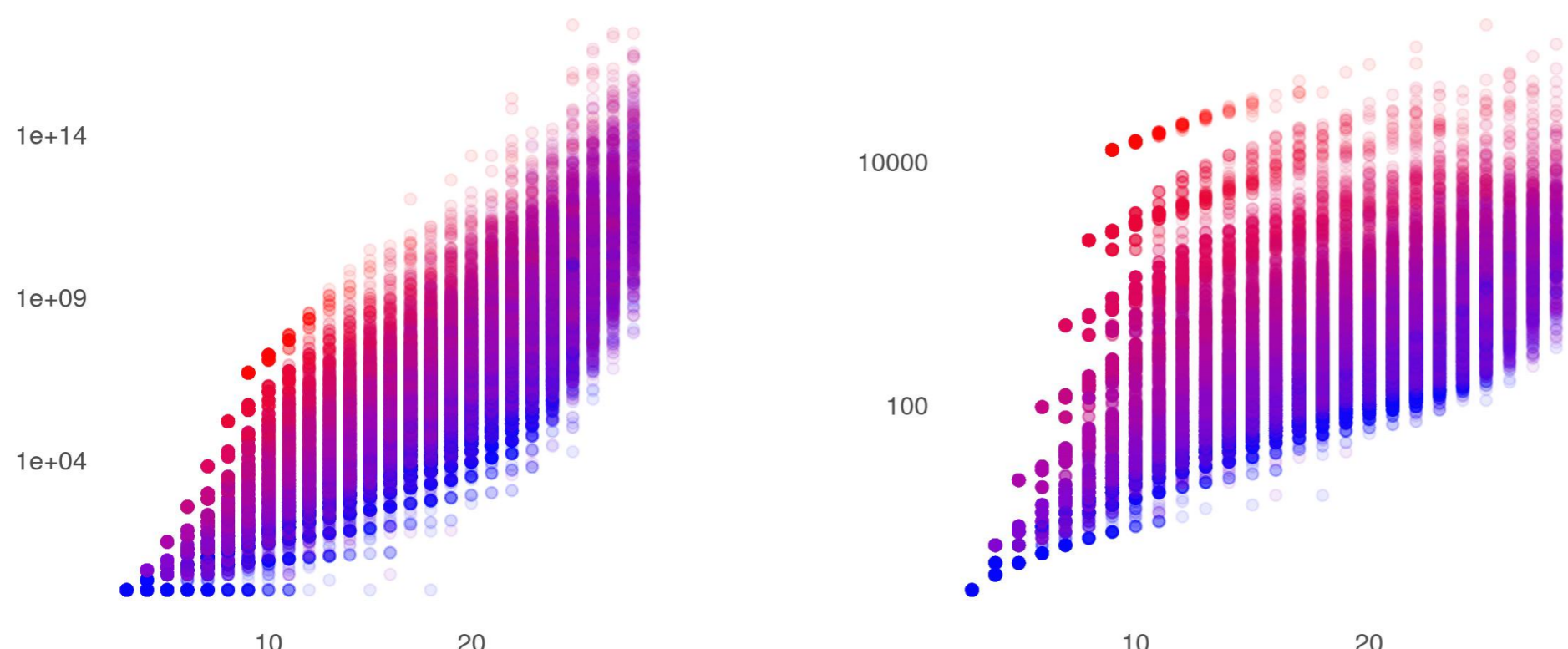
- number of nodes = number of distinct terms in \mathbb{P} -recursion,
- number of paths "12345 \rightarrow ... \rightarrow $\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix}$ " = number of execution paths when evaluating $\mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right)$ via tail-recursion (without memoization/tabling).

CHALLENGE: A GROWING GRAPH OF ANCESTRAL STATES

As the number of sequences n and segregating sites s increases, it quickly becomes computationally intractable to recursively compute exact likelihoods.

Rank $(n + s)$ vs. #paths
colour = max degree (blue=low, red=high)

Rank $(n+s)$ vs. #past states
colour = max degree (blue=low, red=high)

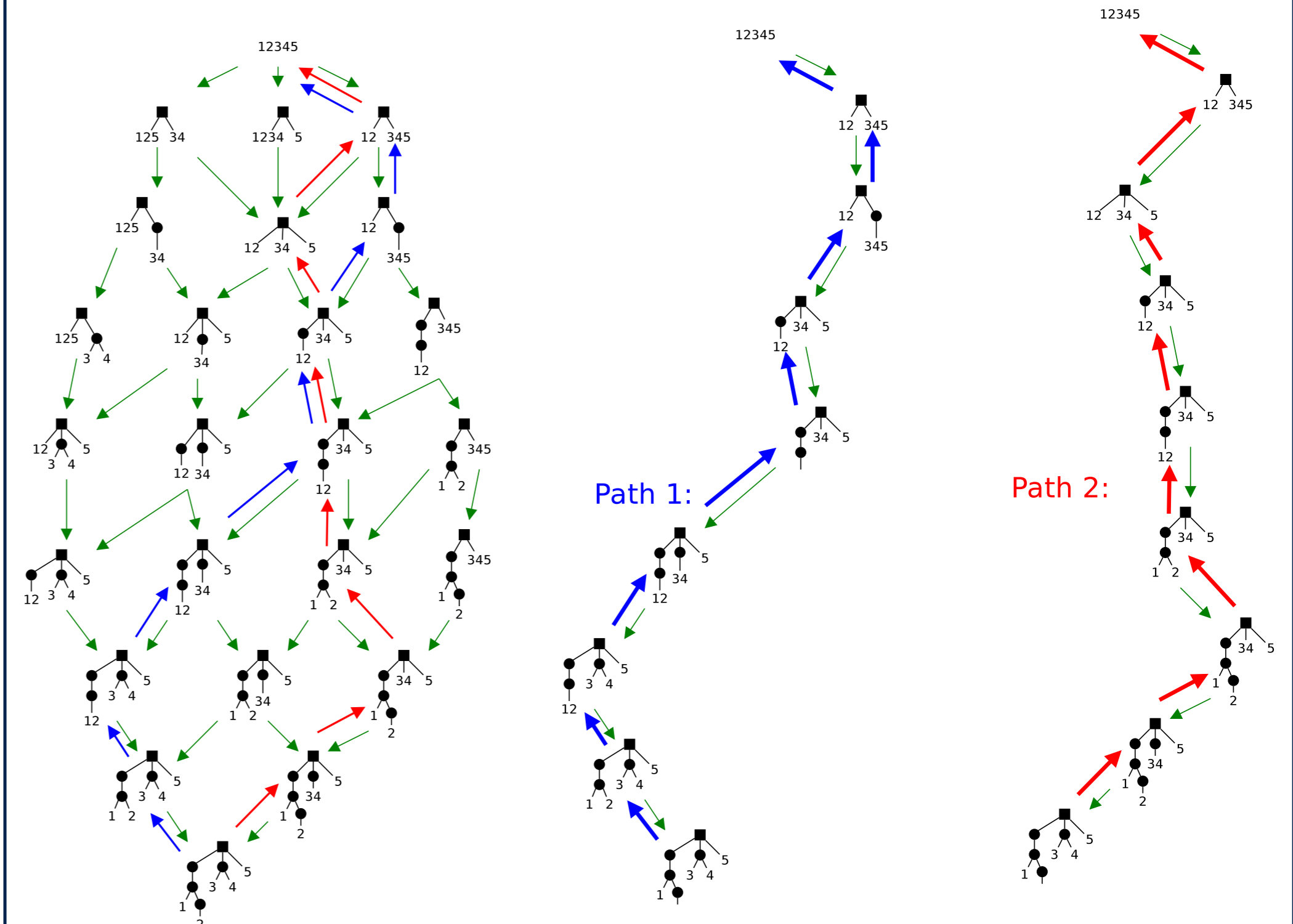


We need methods which do not pre-suppose the ancestral graph, since it is a priori unknown and generating it is as hard as computing likelihoods.

IMPORTANCE SAMPLING OF ANCESTRAL PATHS

We can approximate probabilities of aligned sequences—e.g. $\mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right)$ —by sampling ancestral histories $X_1, \dots, X_N \stackrel{iid}{\sim} \mathbb{Q} \ll \mathbb{P}$ and relying on the following approximation:

$$\mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right) = \sum_{x \in \text{H} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right)} \frac{\mathbb{P}(x)}{\mathbb{Q}(x)} \mathbb{Q}(x) = \mathbb{E}_{X \sim \mathbb{Q}} \left[\frac{\mathbb{P}(X)}{\mathbb{Q}(X)} \right] \approx \frac{1}{N} \sum_{i=1}^N \frac{\mathbb{P}(X_i)}{\mathbb{Q}(X_i)}$$



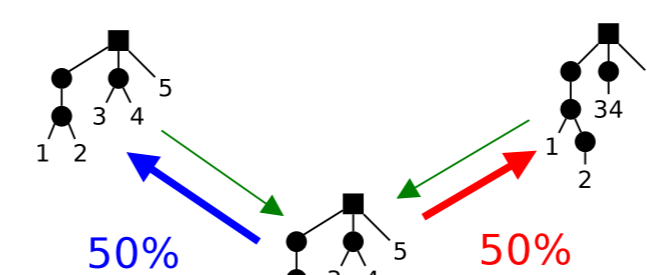
e.g. $\mathbb{P} \left(\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} \right) \approx \frac{1}{2} \left(\frac{\mathbb{P}(\text{Path 1})}{\mathbb{Q}(\text{Path 1})} + \frac{\mathbb{P}(\text{Path 2})}{\mathbb{Q}(\text{Path 2})} \right)$

For this approach to work effectively, \mathbb{Q} should satisfy:

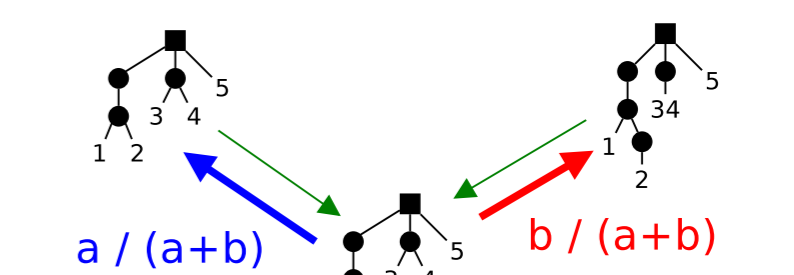
1. \mathbb{Q} must approximate \mathbb{P} well on the space of histories;
2. sampling $X_i \sim \mathbb{Q}$ should be fast;
3. computing $\mathbb{Q}(X_i)$ should be fast.

SEQUENTIAL SAMPLING SCHEMES

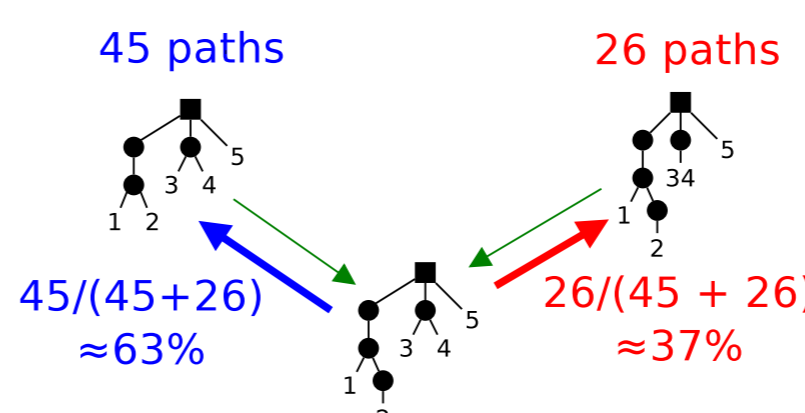
Existing proposal distributions are all sequential: they construct paths step-by-step from the bottom up. They differ by how the next step in a path is sampled.



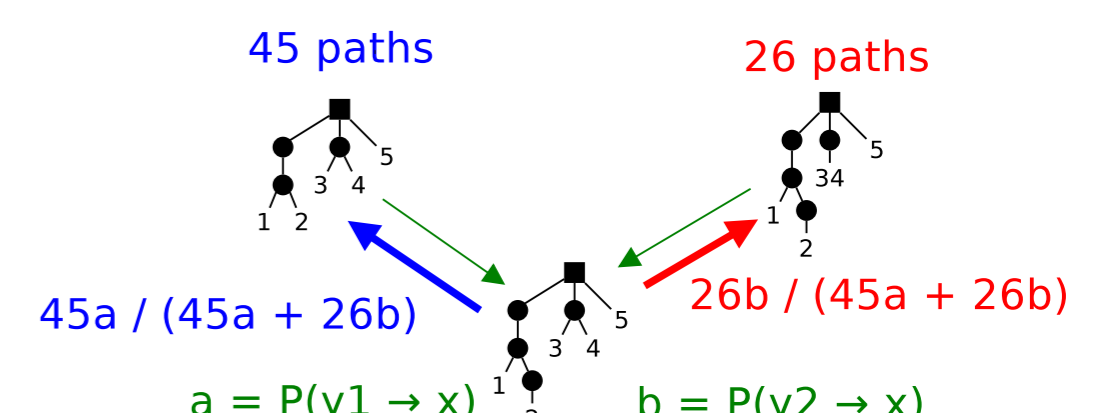
Stephens and Donnelly (S&D)



Griffiths and Taveré (G&T)



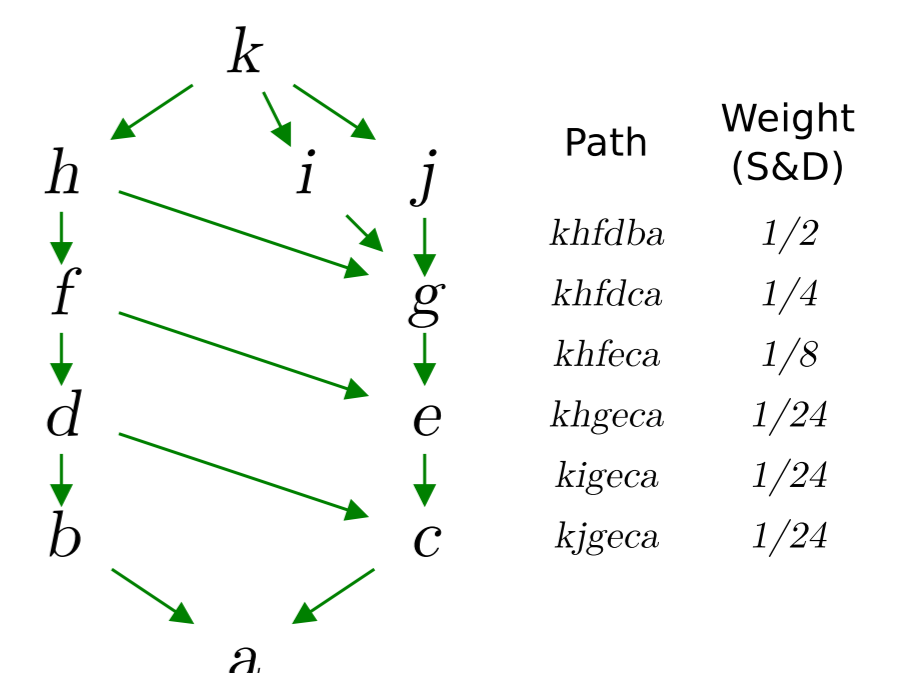
simple combinatorial sampling



G&T + combinatorial correction

PATH DENSITY BIAS AND PATH COUNTING

Any step-by-step scheme which does not penalize choices which "lead to fewer choices down the line", will be biased in favour of low-density regions of path-space, e.g.



To correct for path density bias, we must be able to count ancestral histories effectively (i.e. without generating the ancestral graph), which we do as follows:

$$h(T) = \sum_{S \subseteq [r], 1 \in S} h(\{T_i \mid i \in S\}) h(\{T_i \mid i \notin S\}) \left(\frac{(\sum_{i=1}^r k_i) - 2}{(\sum_{i \in S} k_i) - 1} \right)$$

whereby we here encode rooted unordered trees as nested systems of sets, e.g.

$$\begin{matrix} \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \\ \circ & \circ & \circ & \circ \end{matrix} = \{ \{ \{1, \{2\} \} \}, \{3, 4\}, 5 \}.$$