

Inference on growing trees

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Inference, Model Selection, and the Combinatorics of Growing Trees

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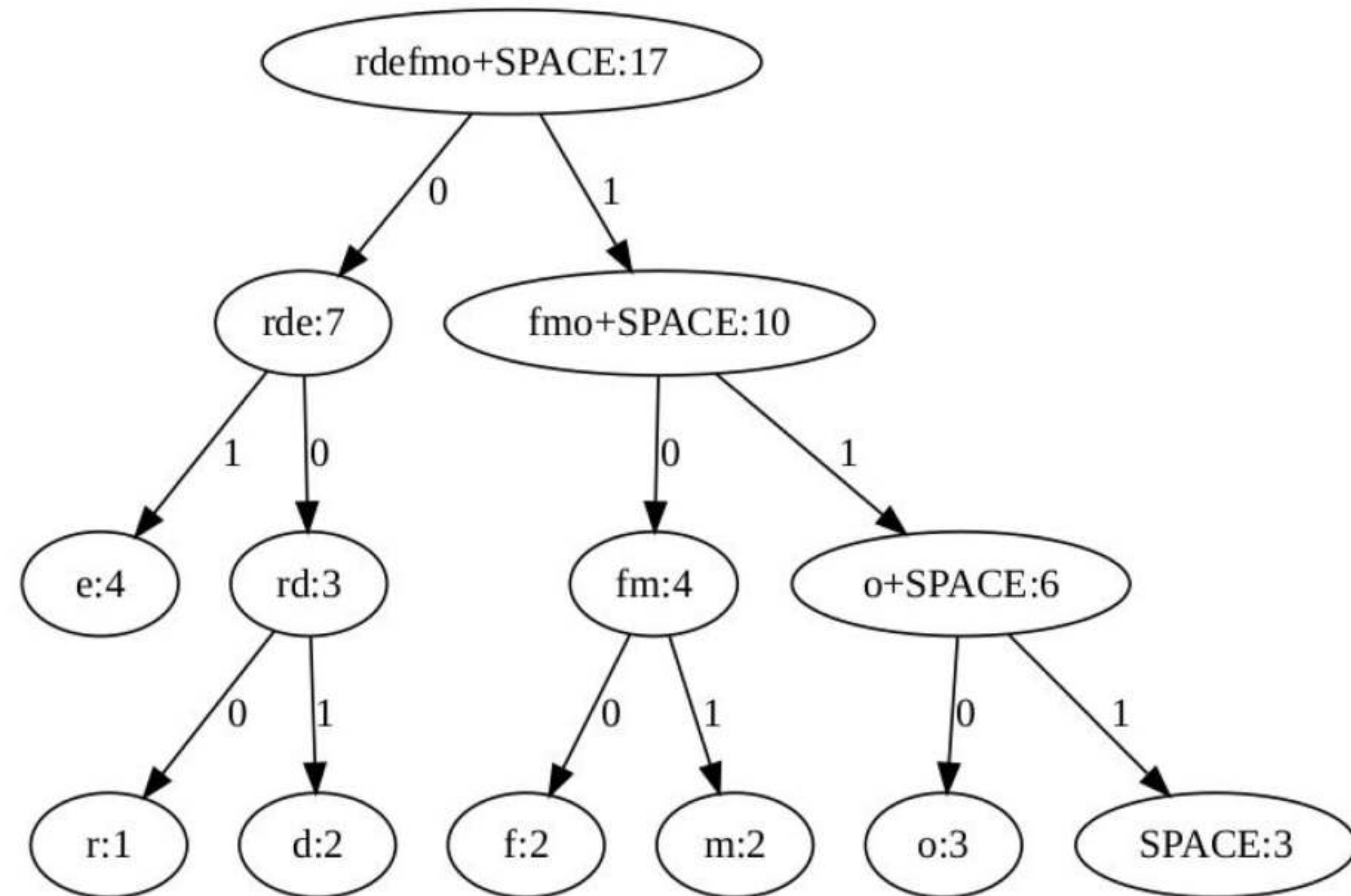
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One can often make inferences about a growing network from its current state alone. For example, it is generally possible to determine how a network changed over time or pick among plausible mechanisms explaining its growth. In practice, however, the extent to which such problems can be solved is limited by existing techniques, which are often inexact, inefficient, or both. In this Letter, we derive exact and efficient inference methods for growing trees and demonstrate them in a series of applications: network interpolation, history reconstruction, model fitting, and model selection.

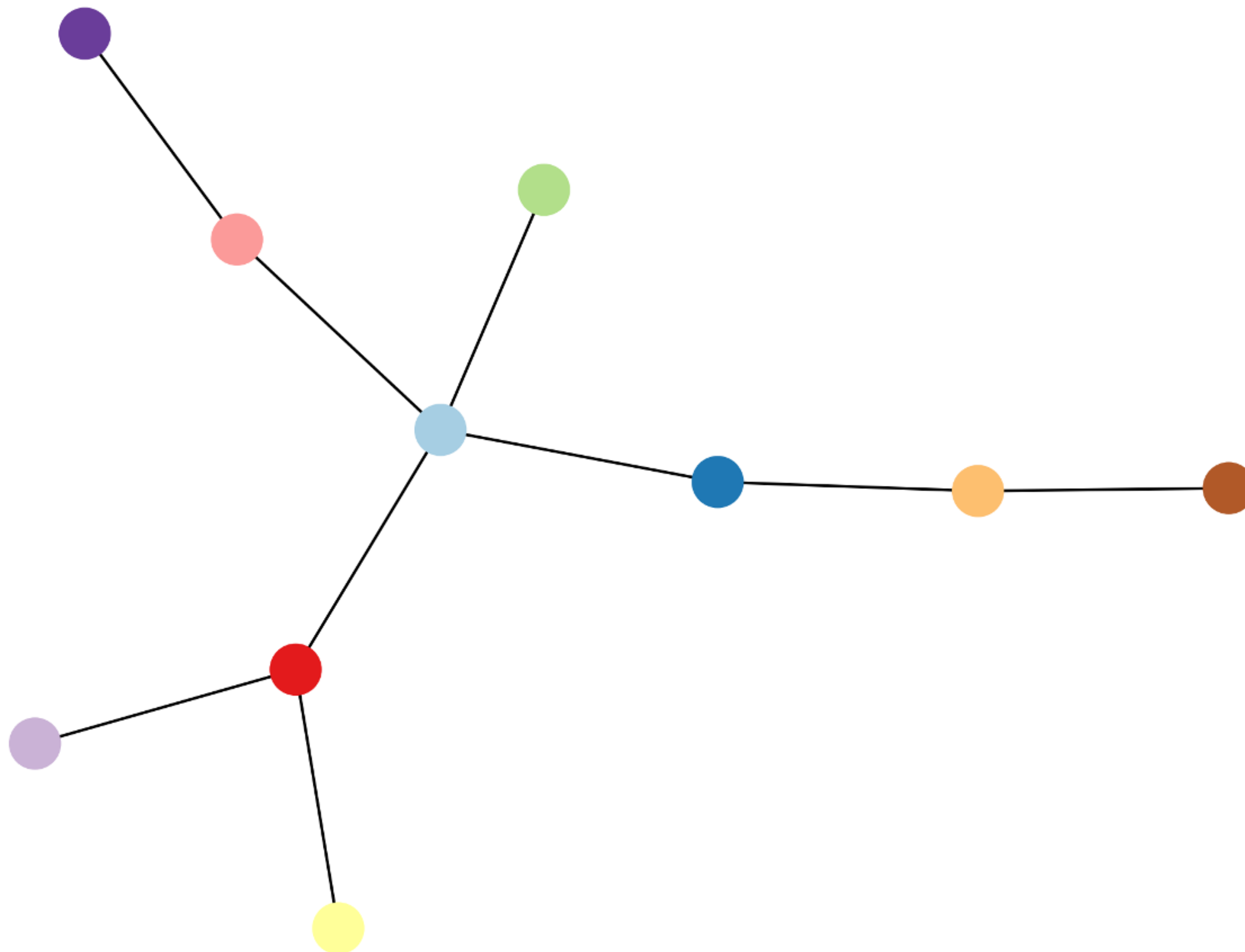
Growing trees

- Trees: simple undirected, acyclic graph. If there are n nodes, then there are $n - 1$ edges
- Growing: each node is added one at a time, such that the “history” of a growing tree can be represented as a sequence of nodes
- Examples: Academic trees, social networks

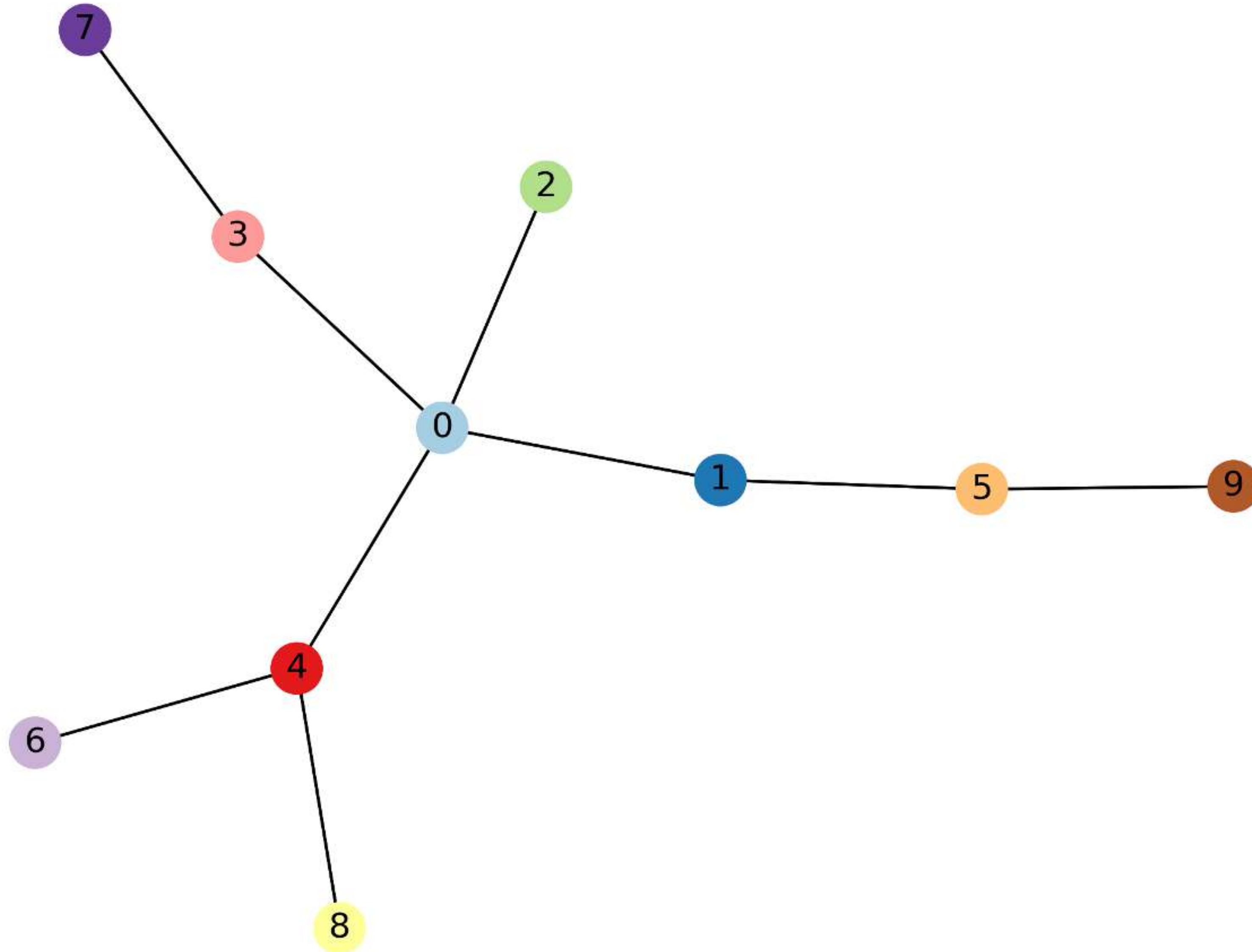


A tree (not a growing one), from Chad's talk on Huffman Encoding

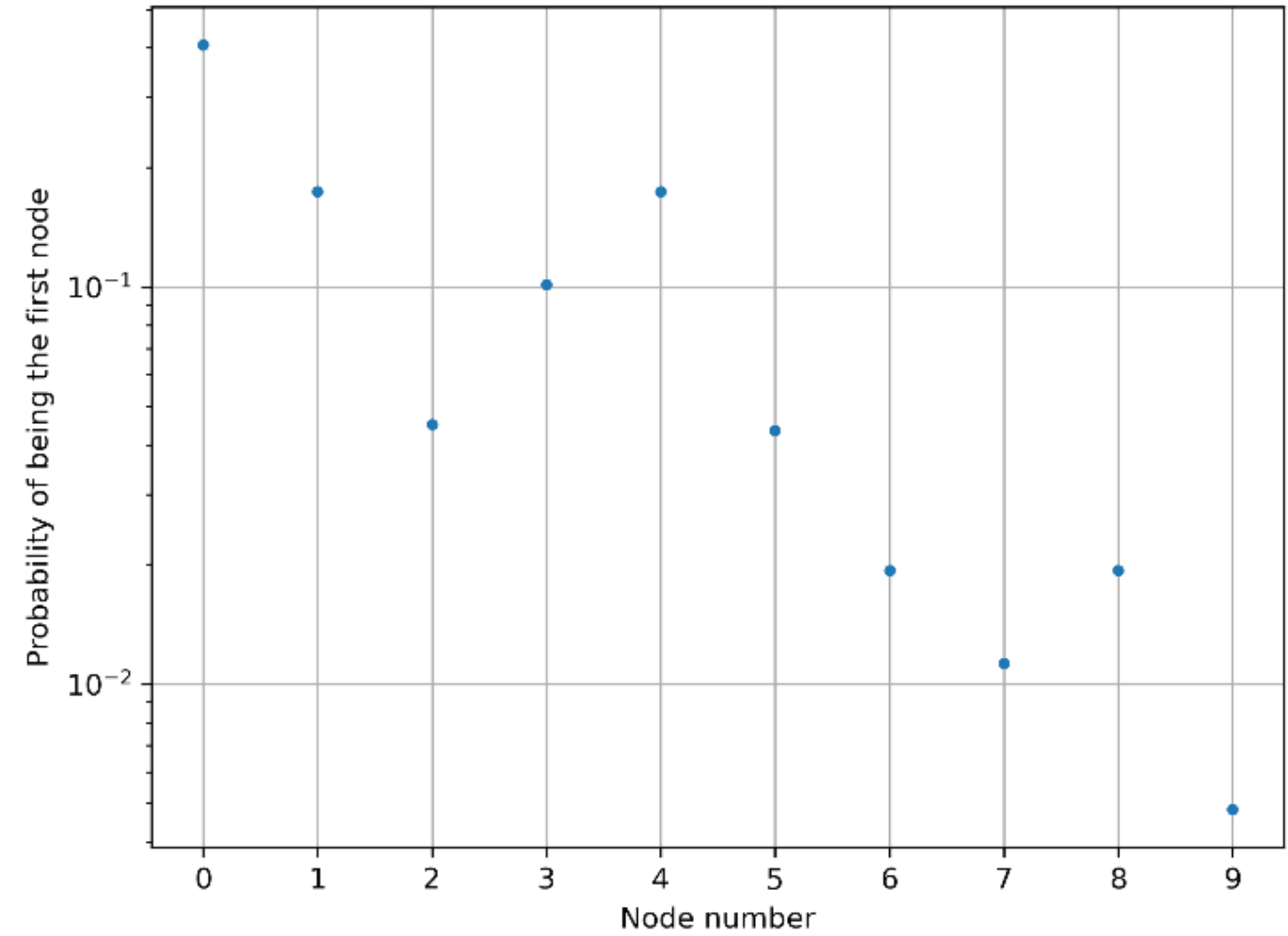
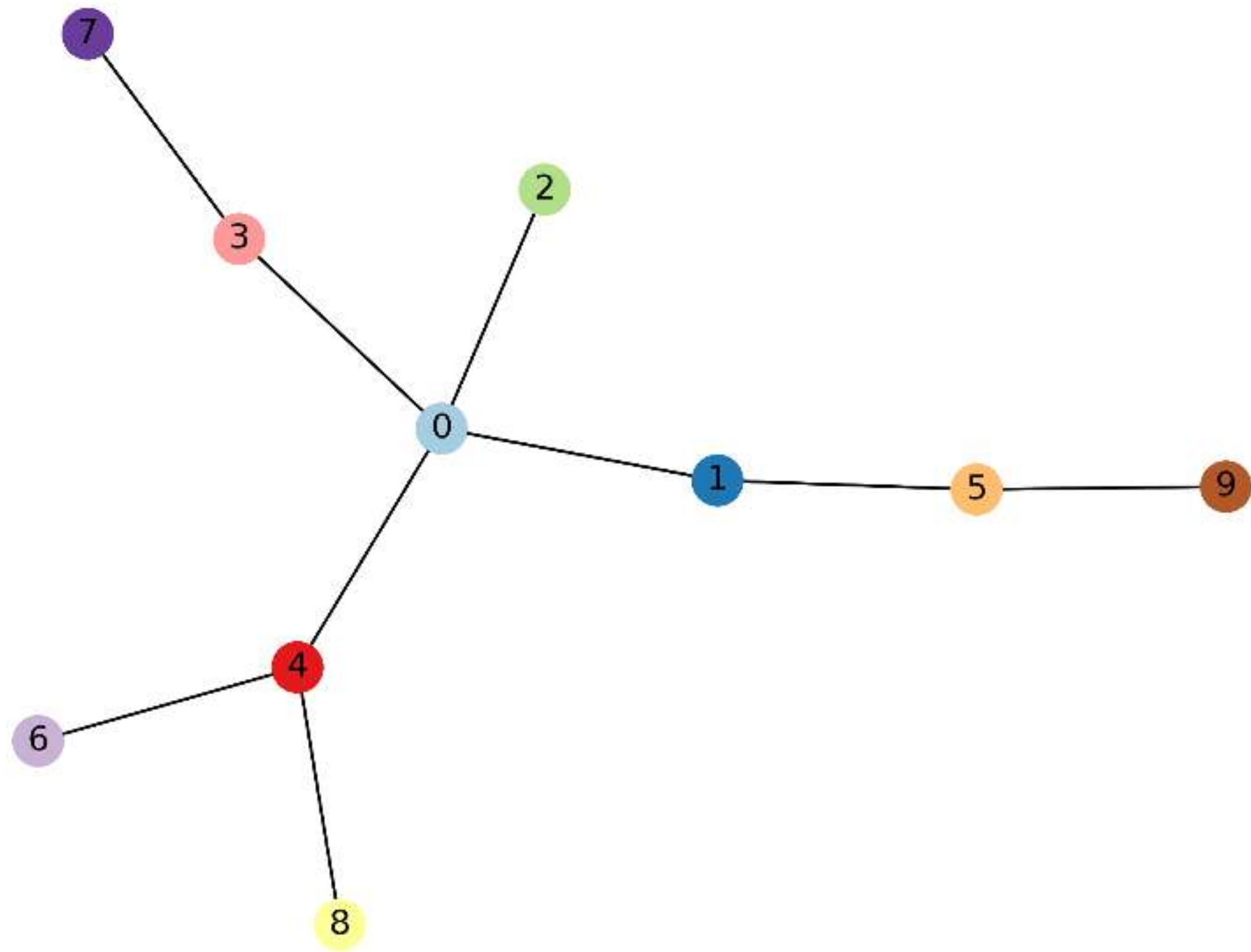
Which node came first?



Which node came first?



Which node came first?



Histories

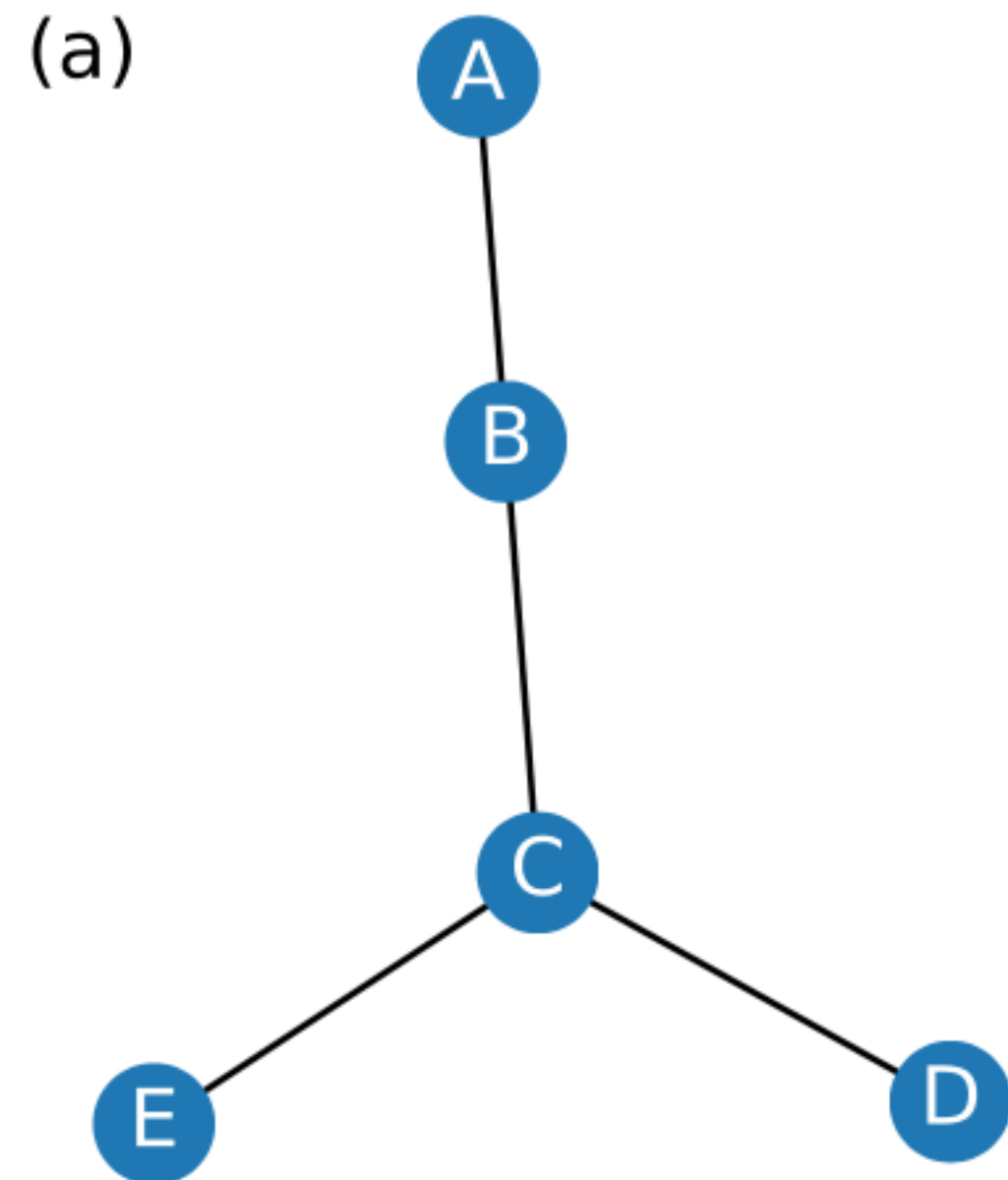


TABLE I. Enumeration of all consistent histories for the network of Fig. 1.

<i>ABCDE</i>	<i>ABCED</i>	<i>BACDE</i>	<i>BACED</i>
<i>BCADE</i>	<i>BCAED</i>	<i>BCDAE</i>	<i>BCDEA</i>
<i>BCEAD</i>	<i>BCEDA</i>	<i>CBADE</i>	<i>CBAED</i>
<i>CBDAE</i>	<i>CBDEA</i>	<i>CBEAD</i>	<i>CBEDA</i>
<i>CDBAE</i>	<i>CDBEA</i>	<i>CDEBA</i>	<i>CEBAD</i>
<i>CEBDA</i>	<i>CEDBA</i>	<i>DCBAE</i>	<i>DCBEA</i>
<i>DCEBA</i>	<i>ECBAD</i>	<i>ECBDA</i>	<i>ECDBA</i>

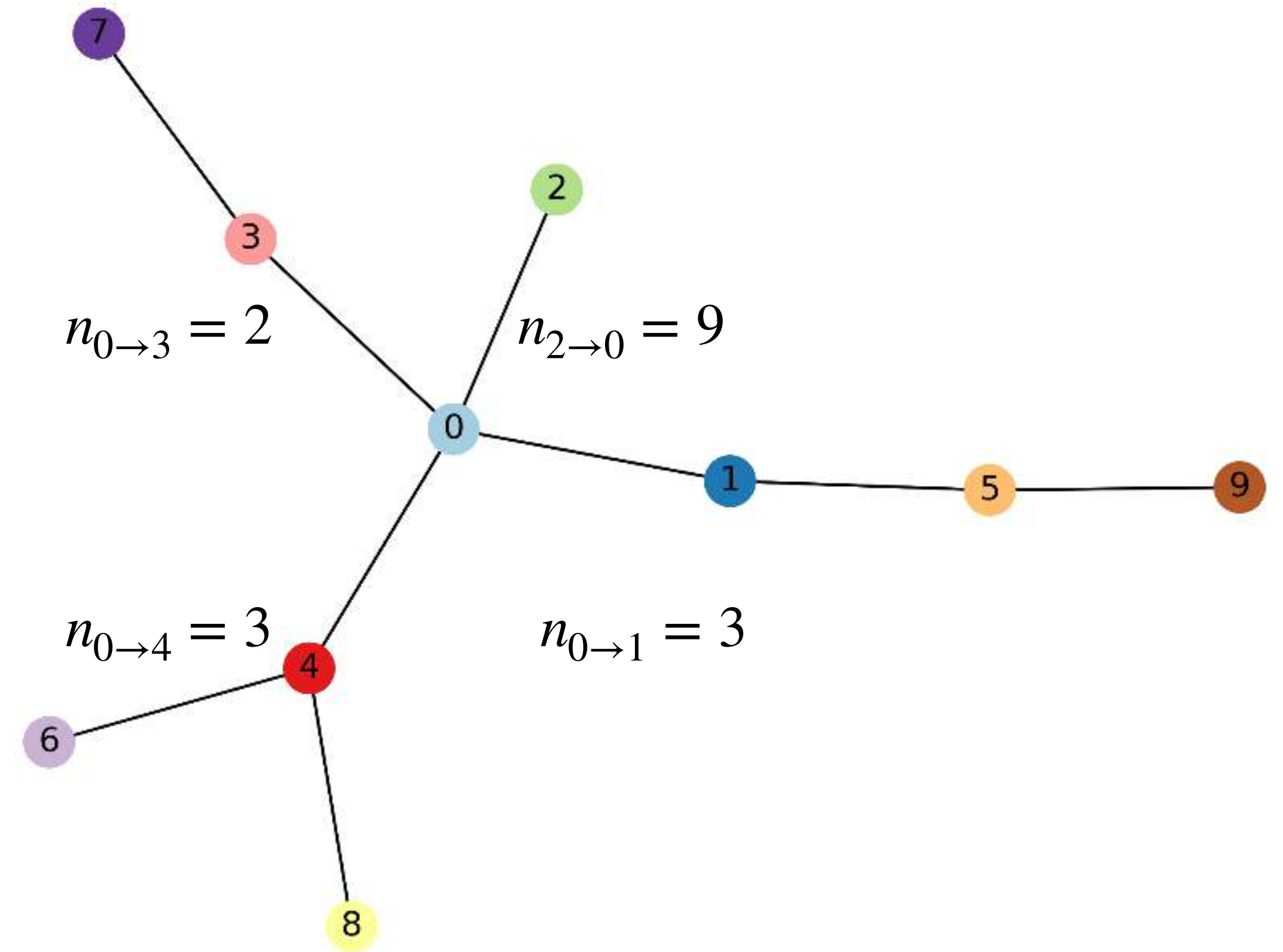
**Of the 120 possible sequences of nodes, only 28 are possible histories.
For example, the sequence EABCD is not possible**

Counting descendants

Let $n_{i \rightarrow j}$ be the number of nodes in the branch containing node j when edge (i, j) is removed. Then

$$n_{i \rightarrow j} = 1 + \sum_{k \in N_j \setminus i} n_{j \rightarrow k}$$

where the sum is over all of j 's neighbors except for i

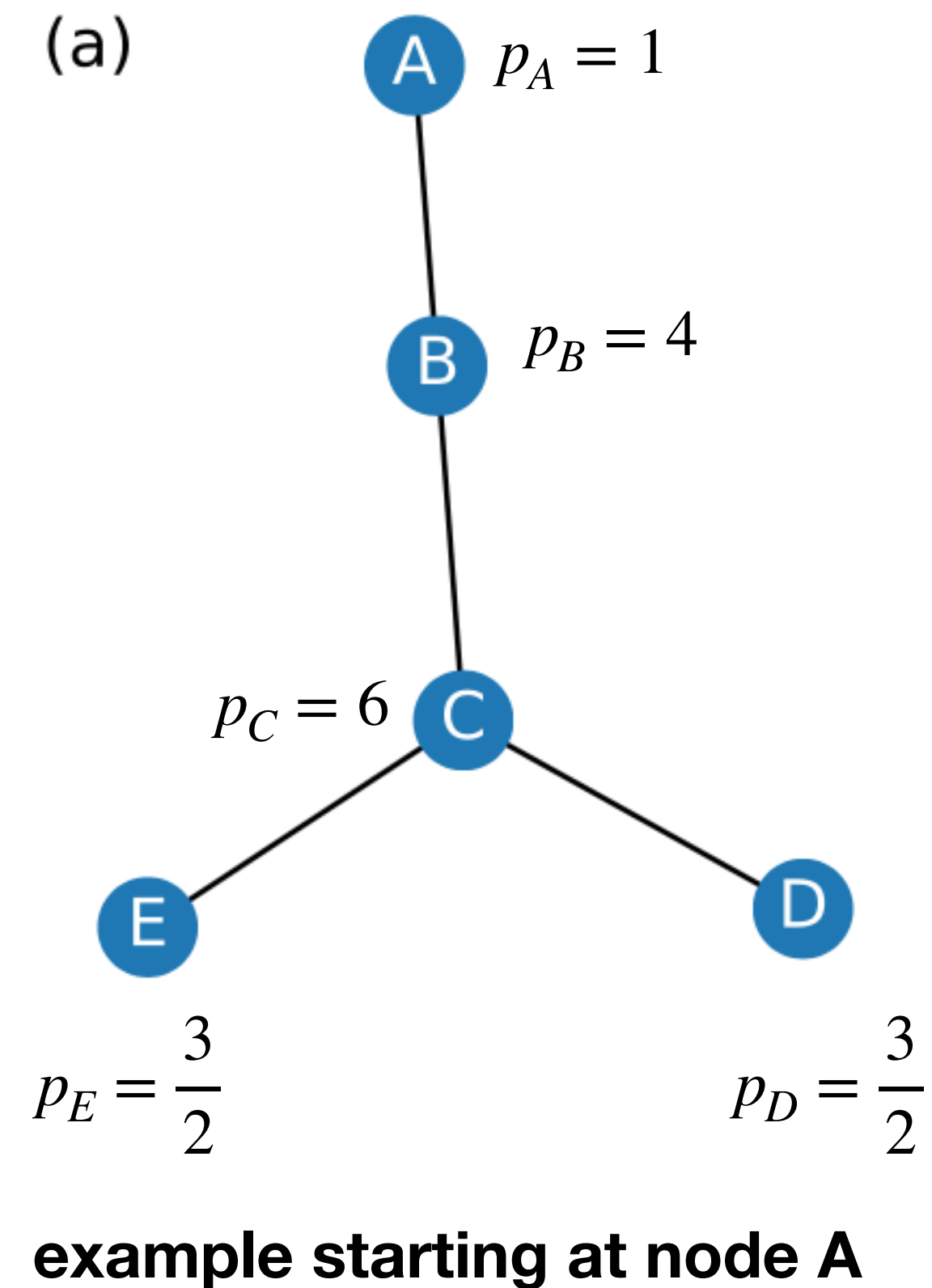


Probability of being the root

Let p_i be the probability that node i comes first. Set $p_0 = 1$ (arbitrary starting point), then

$$p_j = p_i \left(\frac{n_{i \rightarrow j}}{n - n_{i \rightarrow j}} \right)$$

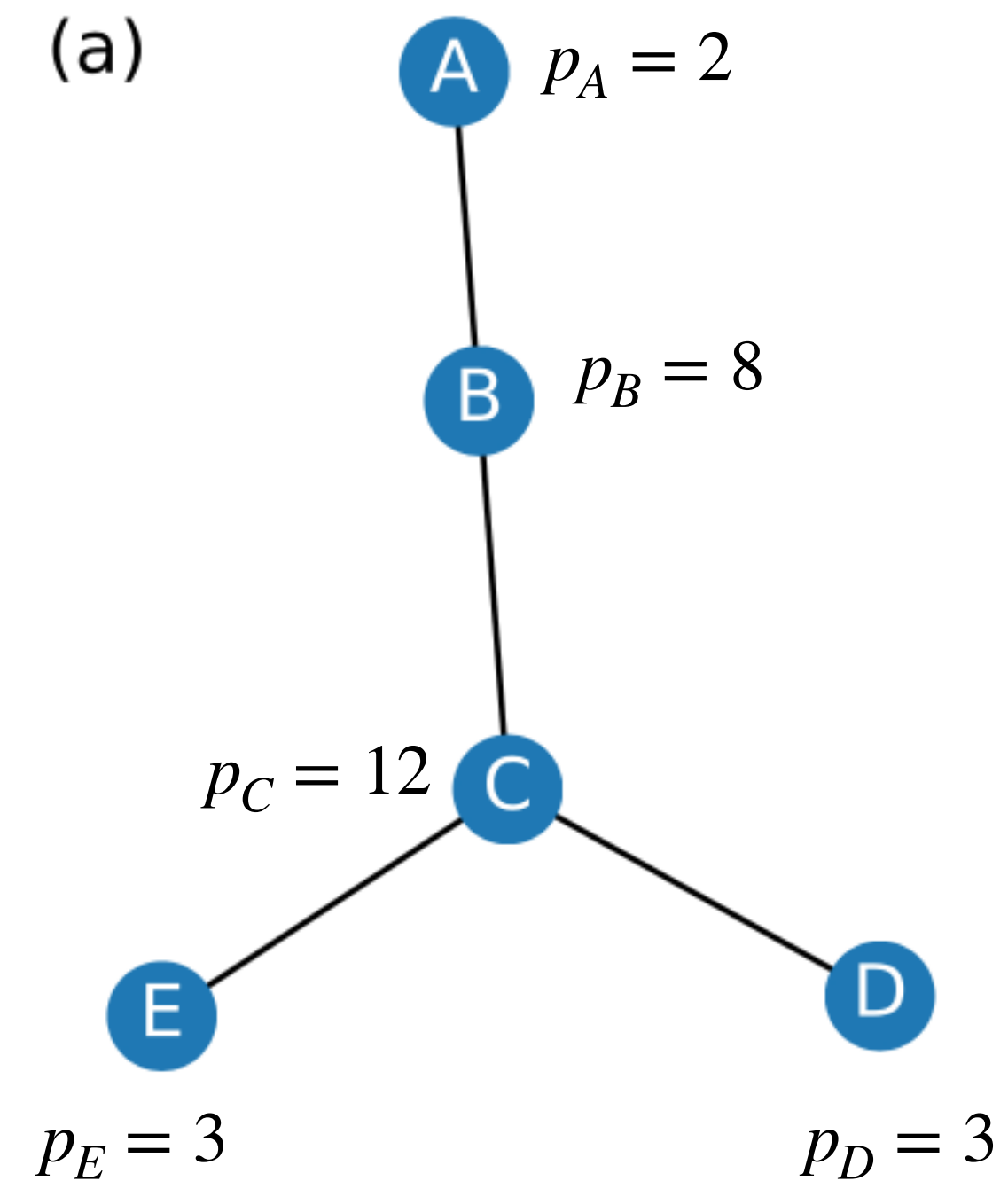
will give the unnormalized probabilities (will be proven momentarily)



Probability of being the root

TABLE I. Enumeration of all consistent histories for the network of Fig. 1.

<i>ABCDE</i>	<i>ABCED</i>	<i>BACDE</i>	<i>BACED</i>
<i>BCADE</i>	<i>BCAED</i>	<i>BCDAE</i>	<i>BCDEA</i>
<i>BCEAD</i>	<i>BCEDA</i>	<i>CBADE</i>	<i>CBAED</i>
<i>CBDAE</i>	<i>CBDEA</i>	<i>CBEAD</i>	<i>CBEDA</i>
<i>CDBAE</i>	<i>CDBEA</i>	<i>CDEBA</i>	<i>CEBAD</i>
<i>CEBDA</i>	<i>CEDBA</i>	<i>DCBAE</i>	<i>DCBEA</i>
<i>DCEBA</i>	<i>ECBAD</i>	<i>ECBDA</i>	<i>ECDBA</i>



**multiplying all the previous results by 2,
we see this is the right answer!**

Counting histories

$$\text{i.e. } p_i = \frac{h_i}{\sum_j h_j}$$

Let h_i be the number of histories in which i comes first, and $h_{i \rightarrow j}$ be the number of histories of the subtree rooted at j when edge (i, j) is removed. Then

$$h_i = (n - 1)! \prod_{j \in N_i} \frac{h_{i \rightarrow j}}{n_{i \rightarrow j}!}$$

NOT $\prod_{j \in N_i} h_{i \rightarrow j}$!!!!!!!

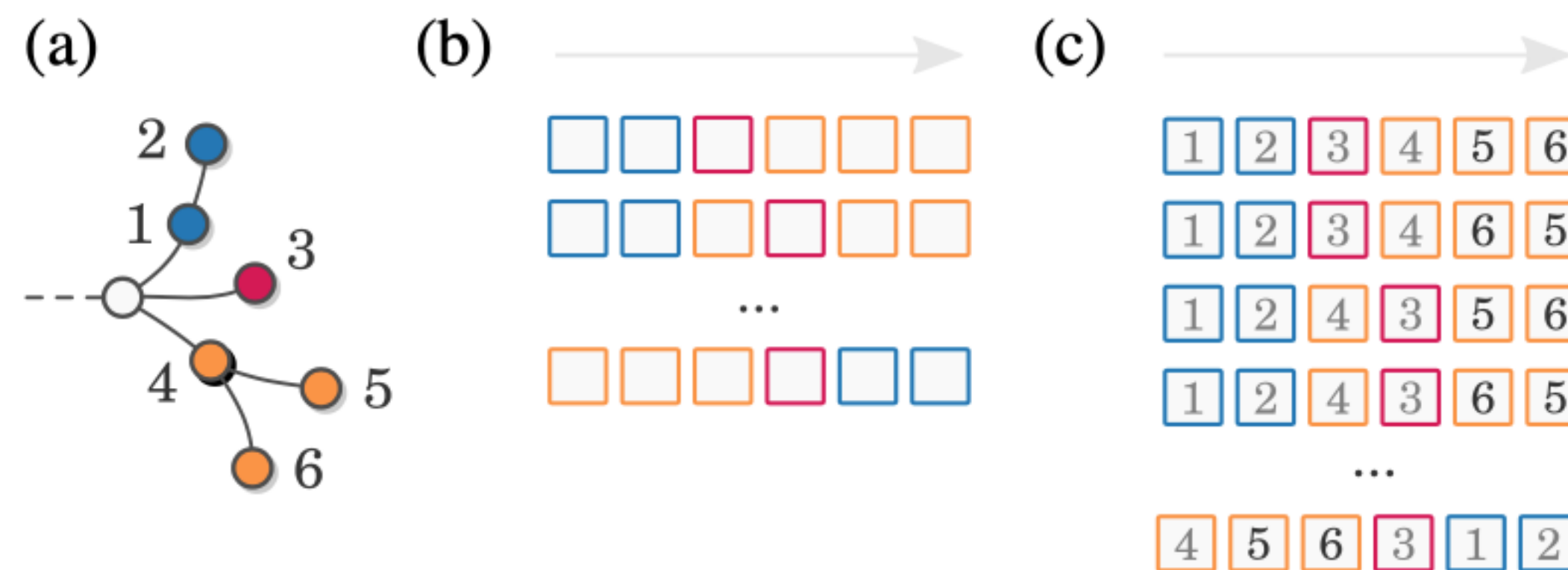


FIG. 1. Interlacing of histories. (a) We count the number of histories for the tree rooted on the focal node, shown in white. (b) There are $(6!/1!2!3!) = 60$ ways to interlace the nodes of the three subtrees, and (c) $1 \times 1 \times 2 = 2$ distinct orderings of these nodes for each of the interlacings. The total number of possible histories for the subtree rooted on the white node is therefore equal to $2 \times 60 = 120$.

Proof of $p_j = p_i(\dots)$

Note that the same game played in previous slide works for subtrees:

$$h_i = (n - 1)! \prod_{j \in N_i} \frac{h_{i \rightarrow j}}{n_{i \rightarrow j}!}$$

$$\implies h_{i \rightarrow j} = (n_{i \rightarrow j} - 1)! \prod_{k \in N_j | i} \frac{h_{j \rightarrow k}}{n_{j \rightarrow k}!}$$

$$\implies h_{i \rightarrow j} \frac{h_{j \rightarrow i}}{n_{j \rightarrow i}!} \frac{(n - 1)!}{(n_{i \rightarrow j} - 1)!} = h_j$$

Proof of $p_j = p_i(\dots)$

$$h_j = h_{i \rightarrow j} \frac{h_{j \rightarrow i} (n-1)!}{n_{j \rightarrow i}! (n_{i \rightarrow j} - 1)!}$$

$$\implies \frac{p_j}{p_i} = \frac{h_j}{h_i} = \frac{n_{i \rightarrow j}! (n_{j \rightarrow i} - 1)!}{n_{j \rightarrow i}! (n_{i \rightarrow j} - 1)!} = \frac{n_{i \rightarrow j}}{n_{j \rightarrow i}} = \frac{n_{i \rightarrow j}}{n - n_{i \rightarrow j}}$$

noting that $n_{i \rightarrow j} + n_{j \rightarrow i} = n$. This is the relation we used to calculate all of the root probabilities

Probability of arriving at t

Let $p_i(t)$ be the probability that node i arrived at time t , and $g_{i \rightarrow j}(t)$ be the number of histories in which node j arrived *before* time t in the subtree containing j when edge (i, j) is removed. Then

$$p_i(t) = \frac{1}{Z} \sum_{j \in N_i} \underbrace{g_{i \rightarrow j}(t)}_{\text{red}} \underbrace{h_{j \rightarrow i}}_{\text{blue}} \binom{n - t - 1}{\underbrace{n_{i \rightarrow j} - t}_{\text{purple}}}$$

Normalization, number of histories in which j comes before t excluding any histories containing i and its descendants, number of histories of the subtree rooted at i , number of subsequences of both subtrees consistent with previous two occurrences

Probability of arriving at t

Let $h_{i,k \rightarrow j}$ be the number of histories rooted at j when the edges (i, j) and (k, j) are removed:

$$h_{i,k \rightarrow j} = (n_{k \rightarrow j} - 1 - n_{j \rightarrow 1})! \prod_{l \in N_j | i, k} \frac{h_{j \rightarrow l}}{n_{j \rightarrow l}!}$$

similar to previous expression for counting histories but minus another branch

$$\implies g_{i \rightarrow j}(t+1) - g_{i \rightarrow j}(t) = \sum_{j \in N_j | i} g_{j \rightarrow k}(t) h_{i, k \rightarrow j} \binom{n-t-1}{n_{j \rightarrow k}-t}$$

Demo

Summary

- Presented a means of counting histories and calculating probabilities for growing trees
- $p_i(t)$ calculation is $\mathcal{O}(n^2)$, whereas the number of possible (but not necessarily consistent) sequences of nodes is $n!$
- Probabilities can be used to do model selection on growing trees

Network	n	Known time line	Reconstructed
Phylogenetic tree	4120	$[-0.41, -0.20]$	$[-0.52, -0.39]$
Twitter reply tree	748	$[0.93, 1.03]$	$[0.89, 1.00]$
Erdős collaborators	6927	...	$[1.18, 1.21]$

generated 100 networks of 2048 nodes from each model. Bayes factor [33] correctly identified the model in every case with an average strength of evidence of $\langle |\log K| \rangle = 25.25$.

We end with a demonstration using empirical data (see Table I). When applied to real trees without temporal metadata our method finds that the coauthorship network centered on Erdős [34] is plausibly grown by a super-preferential attachment mechanism [27], a network of retweets on Twitter [35] is explained by a regular preferential attachment mechanism, and the phylogenetic tree of West Nile virus [36] certainly did not grow by this mechanism. The credible intervals found when using only the static network overlap with the one we find when using available temporal metadata.