Same setup as in Felsenstein.

Outline

Input: T_{σ} , \mathcal{D}

Output: All marginal distributions $P(V_u|\mathcal{D},T)$ for $u \in V \setminus L(T)$

This requires two passes of "message passing" (see N. Friedman Probabilistic graphical models: Principles and techniques, D Koller - 2009 - MIT Press: Cambridge, MA, USA).

Message

The messages and the DP algorithm to compute them is closely related to Felsenstein.

Consider any tree edge $B \rightarrow A$ between inner nodes A and B.

Let $\mathcal{D}_{|u}$ for any $u \in V \setminus L(T)$ denote the data restricted to leaves below u. Let $\mathcal{D}_{|-u}$ for any $u \in V \setminus L(T)$ denote the data restricted to all other leaves not below u.

Upward message:

$$egin{aligned} \delta \uparrow_{A,B} &= \sum_a P(A=a|B) \prod_{C \in child(A)} \delta \uparrow_{C,A=a} \ &= \sum_a P(A=a|B) lpha(A,a) \ &= P(D_{|A}|B) \end{aligned}$$

with $\delta \uparrow_{A,B=b} = P(D_{|A}|B=b)$. If $P_{B,A}$ is the transition matrix along edge $B \to A$ then $\delta \uparrow_{A,B} = P_{B,A}\alpha(A,\cdot)$. Initial case (where A is leaf): $\delta \uparrow_{A,B} = P_{B,A}\mathcal{D}_A$.

Downward message:

$$egin{aligned} \delta \downarrow_{B,A} &= \left(\prod_{C \in child(B) \setminus \{A\}} \delta \uparrow_{C,B=b}
ight) \sum_{p} P(B|parent(B) = p) \delta \downarrow_{parent(B) = p,B} \ &= rac{lpha(B, \cdot)}{\delta \uparrow_{A,B}} \sum_{p} P(B|parent(B) = p) \delta \downarrow_{parent(B) = p,B} \ &= rac{lpha(B, \cdot)}{\delta \uparrow_{A,B}} P_{parent(B),B}^T \delta \downarrow_{parent(B),B} \ &= P(D_{-A}, B) \end{aligned}$$

with $\delta \downarrow_{B=b,A} = P(D_{-A}, B=b)$. Initial case (where B is root): $\delta \downarrow_{B,A} = P(B) \frac{\alpha(B,\cdot)}{\delta_{A,B}}$ with α computed by Felsenstein.

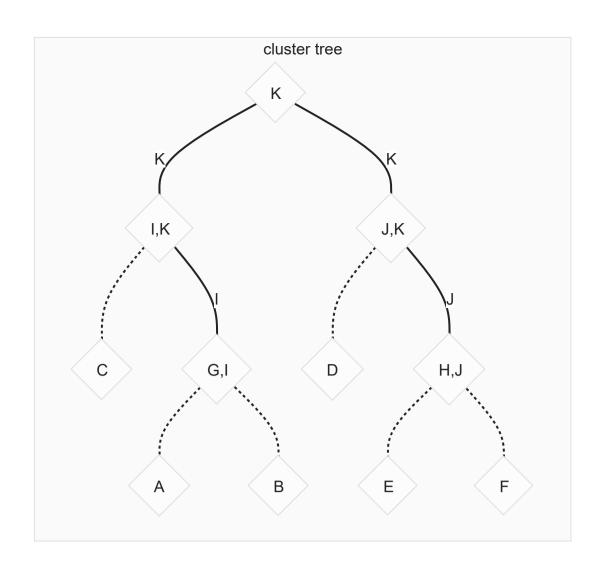
Beliefs:

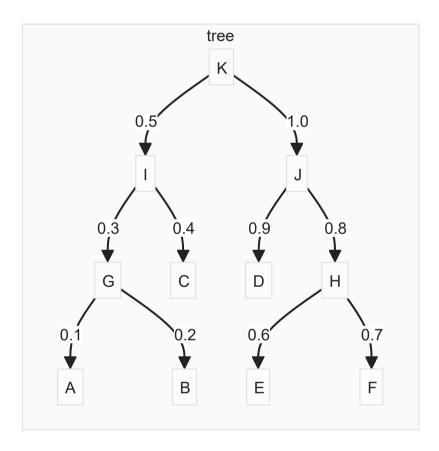
$$egin{aligned} eta_X &= \left(\prod_{C \in child(X)} \delta \uparrow_{C,X}
ight) \sum_p P(X|parent(X) = p) \delta \downarrow_{parent(X) = p,X} \ &= lpha(X,\cdot) \sum_p P(X|parent(X) = p) \delta \downarrow_{parent(X) = p,X} \ &= lpha(X,\cdot) P_{parent(X),X}^T \delta \downarrow_{parent(X),X} \ &= P(\mathcal{D},X) \end{aligned}$$

for $X \neq root$ and $\beta_{root} = \alpha(root, \cdot)\pi$.

Example

Distribution P(A, B, C, D, E, F, G, H, I, J, K) = P(K)P(I|K)P(J|K)P(C|I)P(G|I)P(A|G)P(B|G)P(D|J)P(H|J)P(E|H)P(F|H). We have evidence $e = \{A, B, C, D\}$ so they disappear in the <u>cluster tree</u>.





 \mathcal{D}

	Α	В	С	D	E	F
1	1	0	0	0	1	1
2	0	1	0	0	0	0
3	0	0	1	0	0	0
4	0	0	0	1	0	0

Model: Jukes-Cantor

$$P_{i,i}^ au = rac{1}{4} + rac{3}{4} \mathrm{exp}(-rac{4}{3} au)$$

$$P_{i,j}^{ au} = rac{1}{4} - rac{1}{4} ext{exp}(-rac{4}{3} au)$$

Upward pass:

 $\delta\uparrow_{A,G}=(0.90638,0.031207,0.031207,0.031207)^T$

 $\delta\uparrow_{B,G}=(0.058518,0.824446,0.058518,0.058518)^T$

 $\delta\uparrow_{C,I}=(0.103338,0.103338,0.689985,0.103338)^T$

 $\delta\uparrow_{D,J}=(0.174701,0.174701,0.174701,0.475896)^T$

 $\delta\uparrow_{E,H} = (0.586997, 0.137668, 0.137668, 0.137668)^T$

```
\begin{split} \delta \uparrow_{F,H} &= (0.544931, 0.15169, 0.15169, 0.15169)^T \\ \alpha(G,\cdot) &= (0.05303954, 0.02572849, 0.00182617, 0.00182617)^T \\ \alpha(H,\cdot) &= (0.31987286, 0.02088286, 0.02088286, 0.02088286)^T \\ \delta \uparrow_{G,I} &= (0.04234655, 0.02403941, 0.0080172, 0.0080172)^T \\ \delta \uparrow_{H,J} &= (0.17280427, 0.06990572, 0.06990572, 0.06990572)^T \\ \alpha(I,\cdot) &= (0.00437601, 0.00248418, 0.00553175, 0.00082848)^T \\ \alpha(J,\cdot) &= (0.03018908, 0.0122126, 0.0122126, 0.03326785)^T \\ \delta \uparrow_{I,K} &= (0.00385493, 0.00288363, 0.0044483, 0.00203356)^T \\ \delta \uparrow_{J,K} &= (0.02413692, 0.01939837, 0.01939837, 0.02494847)^T \\ \alpha(K,\cdot) &= (9.30461370e - 05, 5.59377217e - 05, 8.62897693e - 05, 5.07342107e - 05)^T \end{split}
```

Downward pass and Beliefs:

 $\delta \downarrow_{KI} = (0.00603423, 0.00484959, 0.00484959, 0.00623712)^T$

```
\begin{split} \delta\downarrow_{K,J} &= (0.00096373, 0.00072091, 0.00111207, 0.00050839)^T \\ \delta\downarrow_{I,G} &= (0.00059633, 0.00053348, 0.00356204, 0.0006071)^T \\ \delta\downarrow_{J,H} &= (0.00015068, 0.0001395, 0.00015751, 0.00035334)^T \\ \beta_G &= (4.43660557e - 05, 2.04372134e - 05, 5.15791894e - 06, 1.54072264e - 06)^T \\ \beta_H &= (5.85991776e - 05, 3.74529033e - 06, 3.87472667e - 06, 5.28214031e - 06)^T \\ \beta_I &= (2.52526318e - 05, 1.28245382e - 05, 2.85575679e - 05, 4.86720826e - 06)^T \\ \beta_J &= (2.60383153e - 05, 9.75177438e - 06, 1.10109992e - 05, 2.47007606e - 05)^T \\ \beta_K &= (2.32615342e - 05, 1.39844304e - 05, 2.15724423e - 05, 1.26835527e - 05)^T \end{split}
```

Marginals:

```
P(G|\mathcal{D}) = (0.62048769, 0.28582751, 0.0721368, 0.02154799)^{T}
P(H|\mathcal{D}) = (0.81955362, 0.0523807, 0.05419097, 0.07387471)^{T}
P(I|\mathcal{D}) = (0.35317405, 0.17935929, 0.39939567, 0.06807099)^{T}
P(J|\mathcal{D}) = (0.36416282, 0.13638493, 0.15399601, 0.34545625)^{T}
P(K|\mathcal{D}) = (0.32532723, 0.19558108, 0.30170421, 0.17738748)^{T}
```