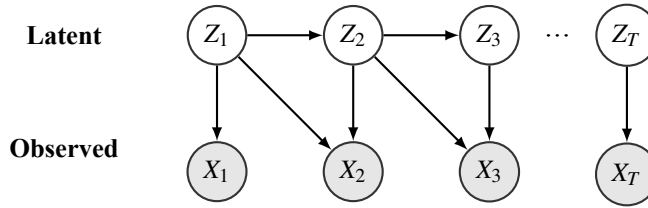


# Notes

September 28

We tend to find the Forward-backward algorithm for the new model.

Suppose we have T observations and T hidden states. Let  $X_i$  denote the ith observation where  $X_i \in \{v_1, v_2, \dots, v_k\}$  ( or just 1,  $\dots$  k for convinience). Let  $Z_i$  denote the ith hidden states where  $Z_i \in \{o_1, o_2, \dots, o_m\}$  ( or just 1,  $\dots$  m for convinience). The graphical model is below.



Generally we need to know  $P(X_{1:T})$  for future computation. From property of HMM we know

$$P(X_{1:T}, Z_{1:T}) = P(Z_1) P(X_1|Z_1) \prod_{t=2}^T P(Z_t|Z_{t-1}) P(X_t|Z_{t-1}, Z_t)$$

Thus we repeat same operation,

$$\begin{aligned} P(X_{1:T}) &= \sum_{Z_{1:T}} P(X_{1:T}, Z_{1:T}) \\ &= \sum_{Z_{1:T}} \underbrace{P(Z_1)P(X_1|Z_1)}_{S_1(Z_1)} \prod_{t=2}^T P(Z_t|Z_{t-1}) P(X_t|Z_{t-1}, Z_t) \\ &= \sum_{Z_{2:T}} \left( \sum_{Z_1} S_1 P(Z_2|Z_1) P(X_2|Z_1, Z_2) \right) \prod_{i=3}^T P(Z_i|Z_{i-1}) P(X_i|Z_{i-1}, Z_i) \\ &\quad \underbrace{\hspace{10em}}_{S_2(Z_2)} \\ &\dots \\ &= \sum_{Z_{j+1:T}} \left( \sum_{Z_j} S_j P(Z_{j+1}|Z_j) P(X_{j+1}|Z_j, Z_{j+1}) \right) \prod_{t=j+2}^T P(Z_t|Z_{t-1}) P(X_t|Z_{t-1}, Z_t) \\ &\quad \underbrace{\hspace{10em}}_{S_{j+1}(Z_{j+1})} \end{aligned}$$

By same kind of trick, we also have

$$\begin{aligned}
P(X_{1:T}) &= \sum_{Z_{1:T}} P(X_{1:T}, Z_{1:T}) \\
&= \sum_{Z_{1:T-1}} \underbrace{\sum_{Z_T} P(Z_T|Z_{T-1})P(X_T|Z_{T-1}, Z_T)P(Z_1)P(X_1|Z_1)}_{R_{T-1}(Z_{T-1})} \prod_{t=2}^{T-1} P(Z_t|Z_{t-1})P(X_t|Z_{t-1}, Z_t) \\
&= \sum_{Z_{1:T-1}} \underbrace{\left( \sum_{Z_{T-1}} R_{T-1}P(Z_{T-1}|Z_{T-2})P(X_{T-1}|Z_{T-2}, Z_{T-1}) \right) P(Z_1)P(X_1|Z_1)}_{R_{T-2}(Z_{T-2})} \prod_{i=2}^{T-2} P(Z_i|Z_{i-1})P(X_i|Z_{i-1}, Z_i) \\
&\dots \\
&= \sum_{Z_{1:j}} \underbrace{\left( \sum_{Z_j} R_jP(Z_j|Z_{j-1})P(X_j|Z_{j-1}, Z_j) \right) P(Z_1)P(X_1|Z_1)}_{R_{j-1}(Z_{j-1})} \prod_{t=2}^{j-1} P(Z_t|Z_{t-1})P(X_t|Z_{t-1}, Z_t)
\end{aligned}$$

Thus we decomposed  $P(X_{1:T})$  as function of  $P(Z_{j+1}|Z_j)$ ,  $P(X_{j+1}|Z_j, Z_{j+1})$  (or  $P(X_1|Z_1)$ ) and  $P(Z_1)$  in two ways. Then we set the parameter vector  $\theta$  as

- $\pi = (\pi_1, \dots, \pi_m)$ , where  $\pi_i = P(Z_1 = i)$
- $\phi_0 = (b_{in})_{m \times k}$ , where  $b_{in} = P(X_1 = n|Z_1 = i)$
- $\phi = (b_{ijn})_{m \times m \times k}$ , where  $b_{ijn} = P(X_t = n|Z_{t-1} = i, Z_t = j)$
- $T = (t_{ij})_{m \times m}$ , where  $t_{ij} = P(Z_{t+1} = j|Z_t = i)$
- $\theta = (\pi, \phi_0, \phi, T)$

And the two notations, S and R, have specific meanings. ( If having priors  $\theta$ , we may condition on it).

$$\begin{aligned}
S_t(Z_t) &= P(X_1, \dots, X_t, Z_t | \theta) \\
R_t(Z_t) &= P(X_T, \dots, X_{t+1} | Z_t, \theta)
\end{aligned}$$

They are easy to see by checking the first term and induction relations:

$$\begin{aligned}
S_t(Z_t) &= \sum_{Z_{t-1}} S_{t-1}P(Z_t|Z_{t-1})P(X_t|Z_{t-1}, Z_t) \\
R_t(Z_t) &= \sum_{Z_{t+1}} R_{t+1}P(Z_{t+1}|Z_t)P(X_{t+1}|Z_t, Z_{t+1})
\end{aligned}$$