

I. (i) Transmission parameter  $a_{uv}$  and Emission parameter  $b_u(v)$  are the parameters associated with the HMM.

(ii) Transmission parameter  $a_{uv}$

<del>u\ v</del>	X	Y	Z	STOP
START	0.11	0	0.6	0
X	0	0.5	0.333	0.167
Y	0.167	0	0.167	0.667
Z	0.5	0.5	0	0

Emission parameter  $b_u(v)$

<del>u\ v</del>	a	b	c
X	0.167	0.5	0.333
Y	0.333	0	0.667
Z	0.167	0.333	0.5

$$a_{\text{START}, X} = \frac{\text{Count}(\text{START}; X)}{\text{Count}(\text{START})} = \frac{2}{5} = 0.4$$

$$b_x(a) = \frac{\text{Count}(X \rightarrow a)}{\text{Count}(X)} = \frac{1}{6} = 0.167$$

$$a_{\text{START}, Y} = \frac{\text{Count}(\text{START}; Y)}{\text{Count}(\text{START})} = \frac{0}{5} = 0$$

$$b_x(b) = \frac{\text{Count}(X \rightarrow b)}{\text{Count}(X)} = \frac{3}{6} = 0.5$$

$$a_{\text{START}, Z} = \frac{\text{Count}(\text{START}; Z)}{\text{Count}(\text{START})} = \frac{3}{5} = 0.6$$

$$b_x(c) = \frac{\text{Count}(X \rightarrow c)}{\text{Count}(X)} = \frac{2}{6} = 0.333$$

$$a_{\text{START}, \text{STOP}} = \frac{\text{Count}(\text{START}; \text{STOP})}{\text{Count}(\text{START})} = \frac{0}{5} = 0$$

$$b_y(a) = \frac{\text{Count}(Y \rightarrow a)}{\text{Count}(Y)} = \frac{2}{6} = 0.333$$

$$a_{X, X} = \frac{\text{Count}(X; X)}{\text{Count}(X)} = \frac{0}{6} = 0$$

$$b_y(b) = \frac{\text{Count}(Y \rightarrow b)}{\text{Count}(Y)} = \frac{0}{6} = 0$$

$$a_{X, Y} = \frac{\text{Count}(X; Y)}{\text{Count}(X)} = \frac{3}{6} = 0.5$$

$$b_y(c) = \frac{\text{Count}(Y \rightarrow c)}{\text{Count}(Y)} = \frac{4}{6} = 0.667$$

$$a_{X, Z} = \frac{\text{Count}(X; Z)}{\text{Count}(X)} = \frac{2}{6} = 0.333$$

$$b_z(a) = \frac{\text{Count}(Z \rightarrow a)}{\text{Count}(Z)} = \frac{1}{6} = 0.167$$

$$a_{X, \text{STOP}} = \frac{\text{Count}(X; \text{STOP})}{\text{Count}(X)} = \frac{1}{6} = 0.167$$

$$b_z(b) = \frac{\text{Count}(Z \rightarrow b)}{\text{Count}(Z)} = \frac{2}{6} = 0.333$$

$$a_{Y, X} = \frac{\text{Count}(Y; X)}{\text{Count}(Y)} = \frac{1}{6} = 0.167$$

$$b_z(c) = \frac{\text{Count}(Z \rightarrow c)}{\text{Count}(Z)} = \frac{3}{6} = 0.5$$

$$a_{Y, Y} = \frac{\text{Count}(Y; Y)}{\text{Count}(Y)} = \frac{0}{6} = 0$$

$$a_{Y, Z} = \frac{\text{Count}(Y; Z)}{\text{Count}(Y)} = \frac{1}{6} = 0.167$$

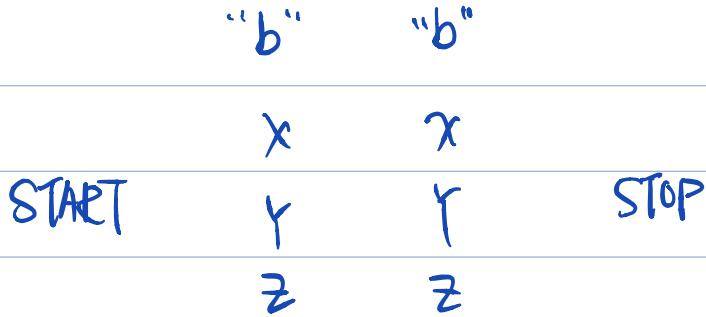
$$a_{Y, \text{STOP}} = \frac{\text{Count}(Y; \text{STOP})}{\text{Count}(Y)} = \frac{4}{6} = 0.667$$

$$a_{Z, X} = \frac{\text{Count}(Z; X)}{\text{Count}(Z)} = \frac{3}{6} = 0.5$$

$$a_{Z, Y} = \frac{\text{Count}(Z; Y)}{\text{Count}(Z)} = \frac{3}{6} = 0.5$$

$$a_{Z, Z} = \frac{\text{Count}(Z; Z)}{\text{Count}(Z)} = \frac{0}{6} = 0, a_{Z, \text{STOP}} = \frac{\text{Count}(Z; \text{STOP})}{\text{Count}(Z)} = \frac{0}{6} = 0$$

2.



$$\pi(0, \text{START}) = 1$$

$$\pi(1, x) = \max \{ \pi(0, \text{START}) \times 0.4 \times 0.5 \} = 0.2$$

$$\pi(1, y) = \max \{ \pi(0, \text{START}) \times 0 \times 0.5 \} = 0$$

$$\pi(1, z) = \max \{ \pi(0, \text{START}) \times 0.6 \times 0.333 \} = 0.1998$$

$$\pi(2, x) = \max \{ \pi(1, x) \times b_x("b") \times a_{x,x}, \pi(1, y) \times b_y("b") \times a_{y,x}, \pi(1, z) \times b_z("b") \times a_{z,x} \}$$

$$= \max \{ 0.2 \times 0.5 \times 0, 0, 0.1998 \times 0.5 \times 0.5 \} = 0.1$$

$$\pi(2, y) = \max \{ \pi(1, x) \times b_y("b") \times a_{x,y}, \pi(1, y) \times b_y("b") \times a_{y,y}, \pi(1, z) \times b_y("b") \times a_{z,y} \} = 0$$

$$\pi(2, z) = \max \{ \pi(1, x) \times b_z("b") \times a_{x,z}, \pi(1, y) \times b_z("b") \times a_{y,z}, \pi(1, z) \times b_z("b") \times a_{z,z} \}$$

$$= \max \{ 0.2 \times 0.333 \times 0.333, 0, 0.1998 \times 0.333 \times 0 \} = 0.0222$$

$$\therefore y_2^* = \arg \max \{ \pi(2, x) \times a_{x,\text{stop}}, \pi(2, y) \times a_{y,\text{stop}}, \pi(2, z) \times a_{z,\text{stop}} \}$$

$$= \arg \max \{ 0.1 \times 0.167, 0, 0.0222 \times 0 \} = x$$

$$y_1^* = \arg \max \{ \pi(1, x) \times a_{x,x}, \pi(1, y) \times a_{y,x}, \pi(1, z) \times a_{z,x} \}$$

$$= \arg \max \{ 0.2 \times 0, 0, 0.1998 \times 0.5 \} = z$$

$\Rightarrow$  The most probable state sequence = (z / "b", x / "b")

$$3. \quad y^* = \arg \max_{y_1, \dots, y_n} p(x_1, \dots, x_n, y_1, \dots, y_{i-1}, y_{i+1}, \dots, y_n \mid y_i \neq v)$$

Base case:

$$\pi(0, v) = \begin{cases} 1 & \text{if } v = \text{START} \\ 0 & \text{otherwise} \end{cases}$$

Then:

For  $j=0, \dots, i-2$  for each  $u \in T$ ,

$$\pi(j+1, u) = \max \{ \pi(j, v) \times b_u(x_{j+1}) \times a_{v, u} \}$$

For  $j=i-1$ , for each  $u \in T$  and  $u \neq v$ ,

$$\begin{aligned} \pi(j+1, u) &= \max \{ \pi(j, v) \times b_u(x_{j+1}) \times a_{v, u} \} \\ &= \max \{ \pi(i-1, v) \times b_u(x_i) \times a_{v, u} \} \end{aligned}$$

For  $j=i, \dots, n-1$ , for each  $u \in T$ ,

$$\pi(j+1, u) = \max \{ \pi(j, v) \times b_u(x_{j+1}) \times a_{v, u} \}$$

Final step:

$$\pi(n+1, \text{stop}) = \max \{ \pi(n, v) \times a_{v, \text{stop}} \}$$

Followed by backward tracking:

$$y_n^* = \arg \max_u \{ \pi(n, u) \cdot a_{u, \text{stop}} \}$$

For  $j=n-1, \dots, 0$

$$y_j^* = \arg \max_u \{ \pi(j, u) \cdot a_{u, y_{j+1}^*} \}$$

#### 4. ① Estimate the most probable state sequence

- Initialization:  $\pi(0, u) = \begin{cases} 1 & \text{if } u = \text{START} \\ 0 & \text{otherwise} \end{cases}$

- then  $\pi(j+1, u) = \max_u \{\pi(j, v) \times b_u((x_{j+1}, y_{j+1})) \times a_{v, u}\}$   
for  $j=0, \dots, n-1$  for each  $u \in T$ .

- Final step:  $\pi(n+1, \text{stop}) = \max_v \{\pi(n, v) \times a_{v, \text{stop}}\}$

- Now every node stores a score of the highest scoring path from START to that node  $(j, u)$ .

- Backtrack the score:

$$y_n^* = \arg \max_u \{\pi(n, u) \cdot a_{u, \text{stop}}\}$$

$$y_j^* = \arg \max_u \{\pi(j, u) \cdot a_{u, y_{j+1}}\} \quad \text{for } j=n-1, \dots, 0$$

$$\textcircled{2} \quad P(z_j=u | x, y; \theta) = P(x_1, x_2, \dots, x_n, y_1, \dots, y_n, z_j=u; \theta)$$

$$= P(x_1, \dots, x_{j-1}, y_1, \dots, y_{j-1}, z_j=u; \theta) \cdot P(x_j, \dots, x_n, y_j, \dots, y_n | x_1, \dots, x_{j-1}, y_1, \dots, y_{j-1}, z_j=u; \theta)$$

$$= P(x_1, \dots, x_{j-1}, y_1, \dots, y_{j-1}, z_j=u; \theta) \cdot P(x_j, \dots, x_n, y_j, \dots, y_n | z_j=u; \theta)$$

$$= P((x_1, y_1), \dots, (x_{j-1}, y_{j-1}), z_j=u; \theta) \cdot P((x_j, y_j), \dots, (x_n, y_n) | z_j=u; \theta)$$

let the forward probability  $\alpha_u(y_j) = P((x_1, y_1), \dots, (x_{j-1}, y_{j-1}), z_j=u; \theta)$

let the backward probability  $\beta_u(y_j) = P((x_j, y_j), \dots, (x_n, y_n) | z_j=u; \theta)$

∴ Forward Algorithm:

- Base case:  $\alpha_u(\text{start}) = d_{\text{START}, u}$

- Recursive case:  $\alpha_u(y_{j+1}) = \sum_v \alpha_v(y_j) d_{v,u} b_v((x_j, y_j))$

∴ Backward Algorithm:

- Base case:  $\beta_u(n) = a_{u, \text{stop}} b_u((x_n, y_n))$

- Recursive case:  $\beta_u(y_j) = \sum_v a_{v,u} b_u((x_j, y_j)) \beta_v(y_{j+1})$

③ Time complexity =  $O(nT^2)$  for both forward and backward algorithm.

For each node in one layer, it has to visit all nodes in next layer.

$$\Rightarrow |T^2|$$

There are  $n$  layers in total .

$$\Rightarrow O(nT^2)$$

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