SUTD 2021 50.007 Homework 4

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Hidden Markov Model

Question 1

The transition probabilities are estimated as:

$$a_{u,v} = \frac{\text{Count}(u,v)}{\text{Count}(u)}$$

$u \backslash v$	X	Y	Z	STOP
START	2/4	0	2/4	0
X	0	2/5	2/5	1/5
Y	1/5	0	1/5	3/5
Z	2/5	3/5	0	0

The emission probabilities are estimated as:

$$b_u(o) = \frac{\text{Count}(u \to o)}{\text{Count}(u)}$$

$u \setminus o$	a	b	c
X	2/5	3/5	0
Y	1/5	0	4/5
Z	1/5	3/5	1/5

Question 2

Base case:

$$\pi(0, \text{START}) = 1$$
, otherwise $\pi(0, v) = 0$ if $v \neq \text{START}$

Moving forward recursively:

$$k = 1$$

$$\pi(1, X) = \pi(0, \text{START}) \times a_{\text{START}, X} \times b_X(\mathbf{b}) = 2/4 \times 3/5 = 3/10$$

 $\pi(1, Y) = \pi(0, \text{START}) \times a_{\text{START}, Y} \times b_Y(\mathbf{b}) = 0$
 $\pi(1, Z) = \pi(0, \text{START}) \times a_{\text{START}, Z} \times b_Z(\mathbf{b}) = 2/4 \times 3/5 = 3/10$

$$k = 2$$

$$\begin{split} \pi(2,X) &= \max_{u \in \mathcal{T}} \{\pi(1,u) \times a_{u,X} \times b_X(\mathbf{c})\} \\ &= \max\{3/10 \times 0 \times 0, 0, 3/10 \times 2/5 \times 0\} \\ &= 0 \\ \pi(2,Y) &= \max_{u \in \mathcal{T}} \{\pi(1,u) \times a_{u,Y} \times b_Y(\mathbf{c})\} \\ &= \max\{3/10 \times 2/5 \times 4/5, 0, 3/10 \times 3/5 \times 4/5\} \\ &= 18/125 \\ \pi(2,Z) &= \max_{u \in \mathcal{T}} \{\pi(1,u) \times a_{u,Z} \times b_Z(\mathbf{c})\} \\ &= \max\{3/10 \times 2/5 \times 1/5, 0, 3/10 \times 0 \times 1/5\} \\ &= 3/125 \end{split}$$

k = 3

$$\pi(3, \text{STOP}) = \max_{u \in \mathcal{T}} \{ \pi(2, u) \times a_{u, \text{STOP}} \}$$

$$= \max\{0 \times 1/5, 18/125 \times 3/5, 3/125 \times 0 \}$$

$$= 54/625$$

Backtracking:

$$y_2^* = \arg\max_{v \in \mathcal{T}} \{\pi(2, v) \times a_{v, \text{STOP}}\} = Y$$
$$y_1^* = \arg\max_{v \in \mathcal{T}} \{\pi(1, v) \times a_{v, Y}\} = Z$$

Therefore, the optimal sequence is: Z, Y.

Question 3

The set of possible states is $\{START, X, Y, Z, STOP\}$. In an HMM, for a sequence of length n, we have

$$P(x_1, ..., x_{i-1}, y_i = u, x_i, ..., x_n)$$

= $P(x_1, ..., x_{i-1}, y_i = u) \times P(x_i, ..., x_n | y_i = u)$
= $\alpha_u(i)\beta_u(i)$,

where:

$$\alpha_u(i) = P(x_1, \dots, x_{i-1}, y_i = u),$$

 $\beta_u(i) = P(x_i, \dots, x_n | y_i = u)$

As shown in class:

$$P(s_i = u | o_1 = \mathbf{b}, o_2 = \mathbf{c}) = \frac{\alpha_u(i)\beta_u(i)}{\sum_v \alpha_v(j)\beta_v(j)},$$

with any choice of $j \in \{1, 2\}$.

Hence, to calculate the marginal distribution probability term, we will use the Forward-Backward Algorithm.

Forward

• Base case:

$$\alpha_X(1) = a_{\text{START},X} = 2/4$$

$$\alpha_Y(1) = a_{\text{START},Y} = 0$$

$$\alpha_Z(1) = a_{\text{START},Z} = 2/4$$

• Moving forward recursively:

$$\alpha_X(2) = \sum_{v} \alpha_v(1) \times a_{v,X} \times b_v(\mathbf{b})$$

$$= (2/4 \times 0 \times 3/5) + 0 + (2/4 \times 2/5 \times 3/5) = 21/50$$

$$\alpha_Y(2) = \sum_{v} \alpha_v(1) \times a_{v,Y} \times b_v(\mathbf{b})$$

$$= (2/4 \times 2/5 \times 3/5) + 0 + (2/4 \times 3/5 \times 3/5) = 3/10$$

$$\alpha_Z(2) = \sum_{v} \alpha_v(1) \times a_{v,Z} \times b_v(\mathbf{b})$$

$$= (2/4 \times 2/5 \times 3/5) + 0 + (2/4 \times 0 \times 3/5) = 3/25$$

Backward

• Base case:

$$\beta_X(2) = a_{X,\text{STOP}} \times b_X(\mathbf{c}) = 1/5 \times 0 = 0$$

 $\beta_Y(2) = a_{Y,\text{STOP}} \times b_Y(\mathbf{c}) = 3/5 \times 4/5 = 12/25$
 $\beta_Z(2) = a_{Z,\text{STOP}} \times b_Z(\mathbf{c}) = 0 \times 1/5 = 0$

• Moving backward recursively:

$$\beta_X(1) = \sum_{v} a_{X,v} \times b_X(\mathbf{b}) \times \beta_v(2)$$

$$= 0 + (2/5 \times 3/5 \times 12/25) + 0 = 72/625$$

$$\beta_Y(1) = \sum_{v} a_{Y,v} \times b_Y(\mathbf{b}) \times \beta_v(2)$$

$$= 0 + (0 \times 0 \times 12/25) + 0 = 0$$

$$\beta_Z(1) = \sum_{v} a_{Z,v} \times b_Z(\mathbf{b}) \times \beta_v(2)$$

$$= 0 + (3/5 \times 3/5 \times 12/25) + 0 = 108/625$$

Thus:

$$\sum_{v} \alpha_v(j)\beta_v(j) = (2/4 \times 72/625) + (2/4 \times 108/625) = (3/10 \times 12/25) = 18/125$$

Hence, the marginal distribution table would be:

$u \setminus i$	1	2
X	0.0576/0.144 = 2/5	0
Y	0	0.144/0.144 = 1
Z	0.0864/0.144 = 3/5	0

Taking the maximum marginal probability value for each column i from the table:

$$s_1^* = \arg \max_{s_1} P(s_1|o_1 = \mathbf{b}, o_2 = \mathbf{c}) = \arg \max_{s_1} P(s_1, o_1 = \mathbf{b}, o_2 = \mathbf{c}) = Z$$

 $s_2^* = \arg \max_{s_2} P(s_2|o_1 = \mathbf{b}, o_2 = \mathbf{c}) = \arg \max_{s_2} P(s_2, o_1 = \mathbf{b}, o_2 = \mathbf{c}) = Y$

Therefore, we would also consistently obtain and arrive at the optimal sequence of: Z, Y.

Question 4

Assume we have a set of possible states $\{0, 1, \dots, N-1, N\}$, where 0 = START and N = STOP.

$$P(x_1, ..., x_{i-1}, y_1, ..., y_{i-1}, z_i = u, x_i, ..., x_n, y_i, ..., y_n; \theta)$$

$$= P(x_1, ..., x_{i-1}, y_1, ..., y_{i-1}, z_i = u; \theta) \times P(x_i, ..., x_n, y_i, ..., y_n | z_i = u; \theta)$$

$$= \alpha_u(i)\beta_u(i),$$

where

$$\alpha_u(i) = P(x_1, \dots, x_{i-1}, y_1, \dots, y_{i-1}, z_i = u; \theta),$$

 $\beta_u(i) = P(x_i, \dots, x_n, y_i, \dots, y_n | z_i = u; \theta)$

Forward

• Base case:

$$\alpha_u(1) = a_{\text{START},u}, \ \forall \ u \in \{1, \dots, N-1\}$$

• Moving forward recursively, for i = 1, ..., n - 1:

$$\alpha_u(i+1) = \sum_v \alpha_v(i) \times a_{v,u} \times b_v(x_i) \times c_{x_i}(y_i),$$

where

$$b_v(x) = P(x|v),$$

$$c_x(y) = P(y|x)$$

Backward

• Base case:

$$\beta_u(n) = a_{u,\text{STOP}} \times b_u(x_n) \times c_{x_n}(y_n), \ \forall \ u \in \{1, \dots, N-1\}$$

• Moving backward recursively, for $i = n - 1, \dots, 1$:

$$\beta_u(i) = \sum_v a_{u,v} \times b_u(x_i) \times c_{x_i}(y_i) \times \beta_v(i+1)$$

At each time step/position, there are N forward (α) and N backward (β) terms to compute. Hence, to compute each term, there are O(N) operations. Thus, at each time step/position, there are $O(N^2)$ operations. Therefore, assuming that the length of the sentence is n, which is the number of different time steps/positions, the total time complexity for this algorithm is $O(nN^2)$.