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# Homework 5

CSCI 567: Machine Learning

Seyed Mohammad Asghari Pari

USCID: 5788788474

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## Hidden Markov Models

### Solution to Question 1.1:

We want to calculate  $P(\mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta)$  where  $\mathbf{O}_{1:6} = [AGCGTA]$ . To this end, we use the formula

$$P(\mathbf{X}_{1:6} = \mathbf{O}_{1:6}) = \sum_{j=1}^2 \alpha_6(j), \quad (1)$$

where

$$\alpha_1(j) = \pi_j P(o_1 | Z_1 = s_j), \quad (2)$$

$$\alpha_t(j) = P(o_t | Z_t = s_j) \sum_{i=1}^2 a_{ij} \alpha_{t-1}(i). \quad (3)$$

Note that since we have only two states  $s_1, s_2$ , we can simplify (2) and (3) for  $j = 1, 2$  as follows:

$$\alpha_1(1) = \pi_1 P(o_1 | Z_1 = s_1) = 0.7 \times P(o_1 | Z_1 = s_1), \quad (4)$$

$$\alpha_1(2) = \pi_2 P(o_1 | Z_1 = s_2) = 0.3 \times P(o_1 | Z_1 = s_2), \quad (5)$$

$$\alpha_t(1) = P(o_t | Z_t = s_1) \sum_{i=1}^2 a_{i1} \alpha_{t-1}(i) = P(o_t | Z_t = s_1) [0.8 \alpha_{t-1}(1) + 0.4 \alpha_{t-1}(2)], \quad (6)$$

$$\alpha_t(2) = P(o_t | Z_t = s_2) \sum_{i=1}^2 a_{i2} \alpha_{t-1}(i) = P(o_t | Z_t = s_2) [0.2 \alpha_{t-1}(1) + 0.6 \alpha_{t-1}(2)]. \quad (7)$$

Now, we use (4)-(7) to calculate  $\alpha_t(j)$ ,  $t = 1, \dots, 6$ ,  $j = 1, 2$ .

- $t = 1$ : We have  $o_1 = A$ . Hence, from (4) and (5) we have

$$\alpha_1(1) = 0.7 P(X_1 = A | Z_1 = s_1) = 0.7 \times 0.4 = 0.28, \quad (8)$$

$$\alpha_1(2) = 0.3 P(X_1 = A | Z_1 = s_2) = 0.3 \times 0.2 = 0.06. \quad (9)$$

- $t = 2$ : We have  $o_2 = G$ . Hence, from (6) and (7) we have

$$\begin{aligned} \alpha_2(1) &= P(X_2 = G | Z_2 = s_1) [0.8 \alpha_1(1) + 0.4 \alpha_1(2)] \\ &= 0.4 [0.8 \times 0.28 + 0.4 \times 0.06] = 0.0992, \end{aligned} \quad (10)$$

$$\begin{aligned} \alpha_2(2) &= P(X_2 = G | Z_2 = s_2) [0.2 \alpha_1(1) + 0.6 \alpha_1(2)] \\ &= 0.2 [0.2 \times 0.28 + 0.6 \times 0.06] = 0.0184. \end{aligned} \quad (11)$$

- $t = 3$ : We have  $o_3 = C$ . Hence, from (6) and (7) we have

$$\begin{aligned}\alpha_3(1) &= P(X_3 = C|Z_3 = s_1)[0.8\alpha_2(1) + 0.4\alpha_2(2)] \\ &= 0.1[0.8 \times 0.0992 + 0.4 \times 0.0184] = 0.0087,\end{aligned}\quad (12)$$

$$\begin{aligned}\alpha_3(2) &= P(X_3 = C|Z_3 = s_2)[0.2\alpha_2(1) + 0.6\alpha_2(2)] \\ &= 0.3[0.2 \times 0.0992 + 0.6 \times 0.0184] = 0.0093.\end{aligned}\quad (13)$$

- $t = 4$ : We have  $o_4 = G$ . Hence, from (6) and (7) we have

$$\begin{aligned}\alpha_4(1) &= P(X_4 = G|Z_4 = s_1)[0.8\alpha_3(1) + 0.4\alpha_3(2)] \\ &= 0.4[0.8 \times 0.0087 + 0.4 \times 0.0093] = 0.0043,\end{aligned}\quad (14)$$

$$\begin{aligned}\alpha_4(2) &= P(X_4 = G|Z_4 = s_2)[0.2\alpha_3(1) + 0.6\alpha_3(2)] \\ &= 0.2[0.2 \times 0.0087 + 0.6 \times 0.0093] = 0.0015.\end{aligned}\quad (15)$$

- $t = 5$ : We have  $o_5 = T$ . Hence, from (6) and (7) we have

$$\begin{aligned}\alpha_5(1) &= P(X_5 = T|Z_5 = s_1)[0.8\alpha_4(1) + 0.4\alpha_4(2)] \\ &= 0.1[0.8 \times 0.0043 + 0.4 \times 0.0015] = 3.98 \times 10^{-4},\end{aligned}\quad (16)$$

$$\begin{aligned}\alpha_5(2) &= P(X_5 = T|Z_5 = s_2)[0.2\alpha_4(1) + 0.6\alpha_4(2)] \\ &= 0.3[0.2 \times 0.0043 + 0.6 \times 0.0015] = 5.17 \times 10^{-4}.\end{aligned}\quad (17)$$

- $t = 6$ : We have  $o_6 = A$ . Hence, from (6) and (7) we have

$$\begin{aligned}\alpha_6(1) &= P(X_6 = A|Z_6 = s_1)[0.8\alpha_5(1) + 0.4\alpha_5(2)] \\ &= 0.4[0.8 \times 3.98 \times 10^{-4} + 0.4 \times 5.17 \times 10^{-4}] = 2.1 \times 10^{-4},\end{aligned}\quad (18)$$

$$\begin{aligned}\alpha_6(2) &= P(X_6 = A|Z_6 = s_2)[0.2\alpha_5(1) + 0.6\alpha_5(2)] \\ &= 0.2[0.2 \times 3.98 \times 10^{-4} + 0.6 \times 5.17 \times 10^{-4}] = 7.79 \times 10^{-5}.\end{aligned}\quad (19)$$

Now, we use (1) to calculate  $P(\mathbf{X}_{1:6} = \mathbf{O}_{1:6})$  as follows

$$P(\mathbf{X}_{1:6} = \mathbf{O}_{1:6}) = \alpha_6(1) + \alpha_6(2) = 2.88 \times 10^{-4}.\quad (20)$$

### Solution to Question 1.2:

First note that

$$\begin{aligned}\operatorname{argmax}_{\mathbf{z}_{1:6}} P(\mathbf{Z}_{1:6} = \mathbf{z}_{1:6} | \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta) &= \operatorname{argmax}_{\mathbf{z}_{1:6}} \frac{P(\mathbf{Z}_{1:6} = \mathbf{z}_{1:6}, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta)}{P(\mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta)} \\ &= \operatorname{argmax}_{\mathbf{z}_{1:6}} P(\mathbf{Z}_{1:6} = \mathbf{z}_{1:6}, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta),\end{aligned}\quad (21)$$

where the last equality is correct because the denominator does not depend on  $\mathbf{z}_{1:6}$ . Therefore, we need to find most likely path and to this end, we use the following formula,

$$\delta_t(j) = \max_{i=1,2} \delta_{t-1}(i) a_{ij} P(X_t = o_t | Z_t = s_j). \quad (22)$$

Note that since we have only two states  $s_1, s_2$ , we can simplify (2) and (3) for  $j = 1, 2$  as follows:

$$\delta_1(1) = \pi_1 P(o_1 | Z_1 = s_1) = 0.7 \times P(o_1 | Z_1 = s_1), \quad (23)$$

$$\delta_1(2) = \pi_2 P(o_1 | Z_1 = s_2) = 0.3 \times P(o_1 | Z_1 = s_2), \quad (24)$$

$$\begin{aligned}\delta_t(1) &= \max\{\delta_{t-1}(1) a_{11} P(X_t = o_t | Z_t = s_1), \delta_{t-1}(2) a_{21} P(X_t = o_t | Z_t = s_1)\} \\ &= \max\{\underbrace{0.8\delta_{t-1}(1) P(X_t = o_t | Z_t = s_1)}_{\mathbb{T}_t(1,1)}, \underbrace{0.4\delta_{t-1}(2) P(X_t = o_t | Z_t = s_1)}_{\mathbb{T}_t(2,1)}\},\end{aligned}\quad (25)$$

$$\begin{aligned}\delta_t(2) &= \max\{\delta_{t-1}(1) a_{12} P(X_t = o_t | Z_t = s_2), \delta_{t-1}(2) a_{22} P(X_t = o_t | Z_t = s_2)\} \\ &= \max\{\underbrace{0.2\delta_{t-1}(1) P(X_t = o_t | Z_t = s_2)}_{\mathbb{T}_t(1,2)}, \underbrace{0.6\delta_{t-1}(2) P(X_t = o_t | Z_t = s_2)}_{\mathbb{T}_t(2,2)}\}.\end{aligned}\quad (26)$$

Now, we use (25)-(26) to calculate  $\delta_t(j)$ ,  $t = 1, \dots, 6$ ,  $j = 1, 2$ .

- $t = 1$ : We have  $o_1 = A$ . Hence, from (23)-(24) we have

$$\delta_1(1) = 0.7P(X_1 = A|Z_1 = s_1) = 0.7 \times 0.4 = 0.28, \quad (27)$$

$$\delta_1(2) = 0.3P(X_1 = A|Z_1 = s_2) = 0.3 \times 0.2 = 0.06. \quad (28)$$

- $t = 2$ : We have  $o_2 = G$ . Hence, from (25)-(26) we have

$$\begin{aligned} \mathbb{T}_2(1, 1) &= 0.8\delta_1(1)P(X_2 = G|Z_2 = s_1) = 0.8 \times 0.28 \times 0.4 = 0.0896 \\ \mathbb{T}_2(2, 1) &= 0.4\delta_1(2)P(X_2 = G|Z_2 = s_1) = 0.4 \times 0.06 \times 0.2 = 0.0048 \\ \delta_2(1) &= \max\{\mathbb{T}_2(1, 1), \mathbb{T}_2(2, 1)\} = 0.0896 \end{aligned} \quad (29)$$

$$\begin{aligned} \mathbb{T}_2(1, 2) &= 0.2\delta_1(1)P(X_2 = G|Z_2 = s_2) = 0.2 \times 0.28 \times 0.4 = 0.0224 \\ \mathbb{T}_2(2, 2) &= 0.6\delta_1(2)P(X_2 = G|Z_2 = s_2) = 0.6 \times 0.06 \times 0.2 = 0.0072 \\ \delta_2(2) &= \max\{\mathbb{T}_2(1, 2), \mathbb{T}_2(2, 2)\} = 0.0224. \end{aligned} \quad (30)$$

Since  $\mathbb{T}_2(1, 1) > \mathbb{T}_2(2, 1)$ , if the most probable state at  $t = 2$  is  $s_1$ , then the most probable state at time  $t = 1$  is also  $s_1$ . Further, since  $\mathbb{T}_2(1, 2) > \mathbb{T}_2(2, 2)$ , if the most probable state at  $t = 2$  is  $s_2$ , then the most probable state at time  $t = 1$  is  $s_1$ .

- $t = 3$ : We have  $o_3 = C$ . Hence, from (25)-(26) we have

$$\begin{aligned} \mathbb{T}_3(1, 1) &= 0.8\delta_2(1)P(X_3 = C|Z_3 = s_1) = 0.8 \times 0.0896 \times 0.1 = 0.0072 \\ \mathbb{T}_3(2, 1) &= 0.4\delta_2(2)P(X_3 = C|Z_3 = s_1) = 0.4 \times 0.0224 \times 0.3 = 0.0027 \\ \delta_3(1) &= \max\{\mathbb{T}_3(1, 1), \mathbb{T}_3(2, 1)\} = 0.0072 \end{aligned} \quad (31)$$

$$\begin{aligned} \mathbb{T}_3(1, 2) &= 0.2\delta_2(1)P(X_3 = C|Z_3 = s_2) = 0.2 \times 0.0896 \times 0.1 = 0.0012 \\ \mathbb{T}_3(2, 2) &= 0.6\delta_2(2)P(X_3 = C|Z_3 = s_2) = 0.6 \times 0.0224 \times 0.3 = 0.0040 \\ \delta_3(2) &= \max\{\mathbb{T}_3(1, 2), \mathbb{T}_3(2, 2)\} = 0.0040. \end{aligned} \quad (32)$$

Since  $\mathbb{T}_3(1, 1) > \mathbb{T}_3(2, 1)$ , if the most probable state at  $t = 3$  is  $s_1$ , then the most probable state at time  $t = 2$  is also  $s_1$ . Further, since  $\mathbb{T}_3(1, 2) < \mathbb{T}_3(2, 2)$ , if the most probable state at  $t = 3$  is  $s_2$ , then the most probable state at time  $t = 2$  is also  $s_2$ .

- $t = 4$ : We have  $o_4 = G$ . Hence, from (25)-(26) we have

$$\begin{aligned} \mathbb{T}_4(1, 1) &= 0.8\delta_3(1)P(X_4 = G|Z_4 = s_1) = 0.8 \times 0.0072 \times 0.4 = 0.0023 \\ \mathbb{T}_4(2, 1) &= 0.4\delta_3(2)P(X_4 = G|Z_4 = s_1) = 0.4 \times 0.0040 \times 0.2 = 3.22 \times 10^{-4} \\ \delta_4(1) &= \max\{\mathbb{T}_4(1, 1), \mathbb{T}_4(2, 1)\} = 0.0023 \end{aligned} \quad (33)$$

$$\begin{aligned} \mathbb{T}_4(1, 2) &= 0.2\delta_3(1)P(X_4 = G|Z_4 = s_2) = 0.2 \times 0.0072 \times 0.4 = 5.73 \times 10^{-4} \\ \mathbb{T}_4(2, 2) &= 0.6\delta_3(2)P(X_4 = G|Z_4 = s_2) = 0.6 \times 0.0040 \times 0.2 = 4.83 \times 10^{-4} \\ \delta_4(2) &= \max\{\mathbb{T}_4(1, 2), \mathbb{T}_4(2, 2)\} = 5.73 \times 10^{-4}. \end{aligned} \quad (34)$$

Since  $\mathbb{T}_4(1, 1) > \mathbb{T}_4(2, 1)$ , if the most probable state at  $t = 4$  is  $s_1$ , then the most probable state at time  $t = 3$  is also  $s_1$ . Further, since  $\mathbb{T}_4(1, 2) > \mathbb{T}_4(2, 2)$ , if the most probable state at  $t = 4$  is  $s_2$ , then the most probable state at time  $t = 3$  is  $s_1$ .

- $t = 5$ : We have  $o_5 = T$ . Hence, from (25)-(26) we have

$$\begin{aligned} \mathbb{T}_5(1, 1) &= 0.8\delta_4(1)P(X_5 = T|Z_5 = s_1) = 0.8 \times 0.0023 \times 0.1 = 1.83 \times 10^{-4} \\ \mathbb{T}_5(2, 1) &= 0.4\delta_4(2)P(X_5 = T|Z_5 = s_1) = 0.4 \times 5.73 \times 10^{-4} \times 0.3 = 6.88 \times 10^{-5} \\ \delta_5(1) &= \max\{\mathbb{T}_5(1, 1), \mathbb{T}_5(2, 1)\} = 1.83 \times 10^{-4} \end{aligned} \quad (35)$$

$$\begin{aligned} \mathbb{T}_5(1, 2) &= 0.2\delta_4(1)P(X_5 = T|Z_5 = s_2) = 0.2 \times 0.0023 \times 0.1 = 4.58 \times 10^{-5} \\ \mathbb{T}_5(2, 2) &= 0.6\delta_4(2)P(X_5 = T|Z_5 = s_2) = 0.6 \times 5.73 \times 10^{-4} \times 0.3 = 1.03 \times 10^{-4} \\ \delta_5(2) &= \max\{\mathbb{T}_5(1, 2), \mathbb{T}_5(2, 2)\} = 1.03 \times 10^{-4}. \end{aligned} \quad (36)$$

Since  $\mathbb{T}_5(1, 1) > \mathbb{T}_5(2, 1)$ , if the most probable state at  $t = 5$  is  $s_1$ , then the most probable state at time  $t = 4$  is also  $s_1$ . Further, since  $\mathbb{T}_5(1, 2) < \mathbb{T}_5(2, 2)$ , if the most probable state at  $t = 5$  is  $s_2$ , then the most probable state at time  $t = 4$  is also  $s_2$ .

- $t = 6$ : We have  $o_6 = A$ . Hence, from (25)-(26) we have

$$\begin{aligned}
\mathbb{T}_6(1, 1) &= 0.8\delta_5(1)P(X_6 = A|Z_6 = s_1) = 0.8 \times 1.83 \times 10^{-4} \times 0.4 = 5.87 \times 10^{-5} \\
\mathbb{T}_6(2, 1) &= 0.4\delta_5(2)P(X_6 = A|Z_6 = s_1) = 0.4 \times 1.03 \times 10^{-4} \times 0.2 = 8.25 \times 10^{-6} \\
\delta_6(1) &= \max\{\mathbb{T}_6(1, 1), \mathbb{T}_6(2, 1)\} = 5.87 \times 10^{-5} \\
\mathbb{T}_6(1, 2) &= 0.2\delta_5(1)P(X_6 = A|Z_6 = s_2) = 0.2 \times 1.83 \times 10^{-4} \times 0.4 = 1.46 \times 10^{-5} \\
\mathbb{T}_6(2, 2) &= 0.6\delta_5(2)P(X_6 = A|Z_6 = s_2) = 0.6 \times 1.03 \times 10^{-4} \times 0.2 = 1.23 \times 10^{-5} \\
\delta_6(2) &= \max\{\mathbb{T}_6(1, 2), \mathbb{T}_6(2, 2)\} = 1.46 \times 10^{-5}.
\end{aligned} \tag{37}$$

Since  $\mathbb{T}_6(1, 1) > \mathbb{T}_6(2, 1)$ , if the most probable state at  $t = 6$  is  $s_1$ , then the most probable state at time  $t = 5$  is also  $s_1$ . Further, since  $\mathbb{T}_6(1, 2) > \mathbb{T}_6(2, 2)$ , if the most probable state at  $t = 6$  is  $s_2$ , then the most probable state at time  $t = 5$  is  $s_1$ .

Now, since  $\delta_6(1) > \delta_6(2)$ , the most probable state at  $t = 6$  is  $s_1$ . If we trace this path back, we can see that the most probable state at time  $t = 5$  is also  $s_1$ . Continuing this procedure, we find the most probable path as  $\mathbf{z}_{1:6}^* = [s_1, s_1, s_1, s_1, s_1, s_1]$ .

### Solution to Question 1.3:

We want to find  $x$  maximizing  $x^* = \operatorname{argmax}_x P(X_7 = x | \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta)$  where  $\mathbf{O}_{1:6} = [AGCGTA]$ . To this end, note that

$$\begin{aligned}
\operatorname{argmax}_x P(X_7 = x | \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta) &= \operatorname{argmax}_x \frac{P(X_7 = x, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta)}{P(\mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta)} \\
&= \operatorname{argmax}_x P(X_7 = x, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta),
\end{aligned} \tag{39}$$

where the last equality is correct because the denominator does not depend on  $x$ .

Since  $x$  can have one of four values  $A, C, G, T$ , we calculate  $P(X_7 = x, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta)$  for all these values and choose the one maximizing that term. Note that we have already calculated  $\alpha_6(1), \alpha_6(2)$  and we can use them to find  $\alpha_7(1), \alpha_7(2)$ .

- $x = A$ . In this case, we have,

$$\begin{aligned}
\alpha_7(1) &= P(X_7 = A | Z_7 = s_1)[0.8\alpha_6(1) + 0.4\alpha_6(2)] \\
&= 0.4[0.8 \times 2.1 \times 10^{-4} + 0.4 \times 7.79 \times 10^{-5}] = 7.96 \times 10^{-5},
\end{aligned} \tag{40}$$

$$\begin{aligned}
\alpha_7(2) &= P(X_7 = A | Z_7 = s_2)[0.2\alpha_6(1) + 0.6\alpha_6(2)] \\
&= 0.2[0.2 \times 2.1 \times 10^{-4} + 0.6 \times 7.79 \times 10^{-5}] = 1.77 \times 10^{-5}.
\end{aligned} \tag{41}$$

Hence, we have

$$P(X_7 = A, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta) = \alpha_7(1) + \alpha_7(2) = 9.73 \times 10^{-5}. \tag{42}$$

- $x = C$ . In this case, we have,

$$\begin{aligned}
\alpha_7(1) &= P(X_7 = C | Z_7 = s_1)[0.8\alpha_6(1) + 0.4\alpha_6(2)] \\
&= 0.1[0.8 \times 2.1 \times 10^{-4} + 0.4 \times 7.79 \times 10^{-5}] = 1.99 \times 10^{-5},
\end{aligned} \tag{43}$$

$$\begin{aligned}
\alpha_7(2) &= P(X_7 = C | Z_7 = s_2)[0.2\alpha_6(1) + 0.6\alpha_6(2)] \\
&= 0.3[0.2 \times 2.1 \times 10^{-4} + 0.6 \times 7.79 \times 10^{-5}] = 2.66 \times 10^{-5}.
\end{aligned} \tag{44}$$

Hence, we have

$$P(X_7 = C, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta) = \alpha_7(1) + \alpha_7(2) = 4.66 \times 10^{-5}. \tag{45}$$

- $x = G$ . In this case, we have,

$$\begin{aligned}
\alpha_7(1) &= P(X_7 = G | Z_7 = s_1)[0.8\alpha_6(1) + 0.4\alpha_6(2)] \\
&= 0.4[0.8 \times 2.1 \times 10^{-4} + 0.4 \times 7.79 \times 10^{-5}] = 7.96 \times 10^{-5},
\end{aligned} \tag{46}$$

$$\begin{aligned}
\alpha_7(2) &= P(X_7 = G | Z_7 = s_2)[0.2\alpha_6(1) + 0.6\alpha_6(2)] \\
&= 0.2[0.2 \times 2.1 \times 10^{-4} + 0.6 \times 7.79 \times 10^{-5}] = 1.77 \times 10^{-5}.
\end{aligned} \tag{47}$$

Hence, we have

$$P(X_7 = G, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta) = \alpha_7(1) + \alpha_7(2) = 9.73 \times 10^{-5}. \tag{48}$$

- $x = T$ . In this case, we have,

$$\begin{aligned}\alpha_7(1) &= P(X_7 = T | Z_7 = s_1)[0.8\alpha_6(1) + 0.4\alpha_6(2)] \\ &= 0.1[0.8 \times 2.1 \times 10^{-4} + 0.4 \times 7.79 \times 10^{-5}] = 1.99 \times 10^{-5},\end{aligned}\quad (49)$$

$$\begin{aligned}\alpha_7(2) &= P(X_7 = T | Z_7 = s_2)[0.2\alpha_6(1) + 0.6\alpha_6(2)] \\ &= 0.3[0.2 \times 2.1 \times 10^{-4} + 0.6 \times 7.79 \times 10^{-5}] = 2.66 \times 10^{-5}.\end{aligned}\quad (50)$$

Hence, we have

$$P(X_7 = T, \mathbf{X}_{1:6} = \mathbf{O}_{1:6}, \Theta) = \alpha_7(1) + \alpha_7(2) = 4.66 \times 10^{-5}. \quad (51)$$

As can be seen, the maximizers are  $x^* = A, G$ .