

Homework 6

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1 PCA

1.1

(a) For any i , we have:

$$\begin{aligned}\frac{\partial J}{\partial p_{i2}} &= 2e_{i2}^T(x_i - p_{i1}e_1 - p_{i2}e_2) \\ &= 2(e_{i2}^T x_i - p_{i1}e_{i2}^T e_1 - p_{i2}e_{i2}^T e_2) \\ &= 2(e_{i2}^T - p_{i2}) = 0 \\ p_{i2} &= e_2^T x_i\end{aligned}$$

(b)

$$\begin{aligned}\frac{\partial \tilde{J}}{\partial e_2} &= -2Se_2 + 2\lambda_2 e_2 + \lambda_{12}e_1 = 0 \\ \tilde{e}_2 &= \frac{1}{2}(S - I\lambda)^{-1}\lambda_{12}e_1 \\ \frac{\partial \tilde{J}}{\partial \lambda_2} &= e_2^T e_2 - 1 = 0 \\ e_2^T e_2 &= 1 \\ \frac{\partial \tilde{J}}{\partial \lambda_{12}} &= e_2^T e_1 - 0 = 0 \\ e_2^T e_1 &= 0\end{aligned}$$

so e_2 is the second largest eigenvalues of S

1.2

(a)

eigenvalue : 1626.52644399, 7.09745924, 28.98609676
corresponding eigenvector:

$$\begin{aligned}&(0.21793758, 0.41449518, 0.88357057)^T \\ &(0.94428286, -0.31834854, -0.0835709)^T \\ &(-0.24664366, -0.85255378, 0.46078081)^T\end{aligned}$$

(b)

eigenvector $(0.94428286, -0.31834854, -0.0835709)^T$ with eigenvalue 7.09745924, because it has minimum eigenvalue, which means it has lowest information (minimum variance)

(c)

eigenvector $(0.21793758, 0.41449518, 0.88357057)^T$ with eigenvalue 1626.52644399 contains the most of information(maximum variance)

2 HMM

(a)

$$\begin{aligned}
\alpha_1(1) &= b_{1A}\pi_1 = 0.28 \\
\alpha_1(2) &= b_{2A}\pi_2 = 0.06 \\
\alpha_2(1) &= b_{1G}(\alpha_1(1)a_{11} + \alpha_1(2)a_{21}) = 0.0992 \\
\alpha_2(2) &= b_{2G}(\alpha_1(1)a_{12} + \alpha_1(2)a_{22}) = 0.0184 \\
\alpha_3(1) &= b_{1C}(\alpha_2(1)a_{11} + \alpha_2(2)a_{21}) = 0.008672 \\
\alpha_3(2) &= b_{2C}(\alpha_2(1)a_{11} + \alpha_2(2)a_{22}) = 0.009264 \\
\alpha_4(1) &= b_{1G}(\alpha_3(1)a_{11} + \alpha_3(2)a_{21}) = 0.00425728 \\
\alpha_4(2) &= b_{2G}(\alpha_3(1)a_{12} + \alpha_3(2)a_{22}) = 0.00145856 \\
\alpha_5(1) &= b_{1T}(\alpha_4(1)a_{11} + \alpha_4(2)a_{21}) = 0.0003989248 \\
\alpha_5(2) &= b_{2T}(\alpha_4(1)a_{12} + \alpha_4(2)a_{22}) = 0.0005179776 \\
\alpha_6(1) &= b_{1A}(\alpha_5(1)a_{11} + \alpha_5(2)a_{21}) = 0.000210532352 \\
\alpha_6(2) &= b_{2A}(\alpha_5(1)a_{12} + \alpha_5(2)a_{22}) = 0.00007811 \\
P(O; \Theta) &= \alpha_6(1) + \alpha_6(2) = 0.000288642352
\end{aligned}$$

(b)

$$\begin{aligned}
\beta_6(1) &= 1 \\
\beta_6(2) &= 1 \\
\beta_5(1) &= (\beta_{1A}a_{11}\beta_6(1) + \beta_{2A}a_{12}\beta_6(2)) = 0.36 \\
\beta_5(2) &= (\beta_{1A}a_{21}\beta_6(1) + \beta_{2A}a_{22}\beta_6(2)) = 0.28 \\
\beta_4(1) &= (\beta_{1T}a_{11}\beta_5(1) + \beta_{2T}a_{12}\beta_5(2)) = 0.0456 \\
\beta_4(2) &= (\beta_{1T}a_{21}\beta_5(1) + \beta_{2T}a_{22}\beta_5(2)) = 0.0648 \\
\beta_3(1) &= (\beta_{1G}a_{11}\beta_4(1) + \beta_{2G}a_{12}\beta_4(2)) = 0.017184 \\
\beta_3(2) &= (\beta_{1G}a_{21}\beta_4(1) + \beta_{2G}a_{22}\beta_4(2)) = 0.015072 \\
\beta_2(1) &= (\beta_{1C}a_{11}\beta_3(1) + \beta_{2C}a_{12}\beta_3(2)) = 0.00227904 \\
\beta_2(2) &= (\beta_{1C}a_{21}\beta_3(1) + \beta_{2C}a_{22}\beta_3(2)) = 0.00340032 \\
\beta_1(1) &= (\beta_{1G}a_{11}\beta_2(1) + \beta_{2G}a_{12}\beta_2(2)) = 0.0008643056 \\
\beta_1(2) &= (\beta_{1G}a_{21}\beta_2(1) + \beta_{2G}a_{22}\beta_2(2)) = 0.0007726848 \\
P(X_6 = S_1|O; \theta) &= \frac{\alpha_6(1)\beta_6(1)}{\alpha_6(1)\beta_6(1) + \alpha_6(2)\beta_6(2)} = 0.72938 \\
P(X_6 = S_2|O; \theta) &= 1 - P(X_6 = S_1|O; \theta) = 0.27062
\end{aligned}$$

(c)

$$P(X_4 = S_1|O; \theta) = \frac{\alpha_4(1)\beta_4(1)}{\alpha_4(1)\beta_4(1) + \alpha_4(2)\beta_4(2)} = 0.672559$$

$$P(X_4 = S_2|O; \theta) = 1 - P(X_4 = S_1|O; \theta) = 0.327441$$

(d)

$$P(X_1 = S_1|O; \theta) = \frac{\alpha_1(1)\beta_1(1)}{\alpha_1(1)\beta_1(1) + \alpha_1(2)\beta_1(2)} = 0.83938$$

$$P(X_1 = S_2|O; \theta) = 1 - P(X_1 = S_1|O; \theta) = 0.16062$$

max is S_1

$$P(X_2 = S_1|O; \theta) = \frac{\alpha_2(1)\beta_2(1)}{\alpha_2(1)\beta_2(1) + \alpha_2(2)\beta_2(2)} = 0.78324$$

$$P(X_2 = S_2|O; \theta) = 1 - P(X_2 = S_1|O; \theta) = 0.21676$$

max is S_1

$$P(X_3 = S_1|O; \theta) = \frac{\alpha_3(1)\beta_3(1)}{\alpha_3(1)\beta_3(1) + \alpha_3(2)\beta_3(2)} = 0.51627$$

$$P(X_3 = S_2|O; \theta) = 1 - P(X_3 = S_1|O; \theta) = 0.48373$$

max is S_1

$$P(X_5 = S_1|O; \theta) = \frac{\alpha_5(1)\beta_5(1)}{\alpha_5(1)\beta_5(1) + \alpha_5(2)\beta_5(2)} = 0.49753$$

$$P(X_5 = S_2|O; \theta) = 1 - P(X_5 = S_1|O; \theta) = 0.50247$$

max is S_2

Most likely explanation is $S_1S_1S_1S_1S_2S_1$

(e)

A:

$$P(O_7 = A|O; \theta) = P(X_6 = S_1|O; \theta)b_{1A} + P(X_6 = S_2|O; \theta)b_{2A} = 0.345876$$

C:

$$P(O_7 = C|O; \theta) = P(X_6 = S_1|O; \theta)b_{1C} + P(X_6 = S_2|O; \theta)b_{2C} = 0.154124$$

G:

$$P(O_7 = G|O; \theta) = P(X_6 = S_1|O; \theta)b_{1G} + P(X_6 = S_2|O; \theta)b_{2G} = 0.345876$$

T:

$$P(O_7 = T|O; \theta) = P(X_6 = S_1|O; \theta)b_{1T} + P(X_6 = S_2|O; \theta)b_{2T} = 0.154124$$

A,G is most likely after $o_{1:6}$