

HW #6

4. $f(x) = p_1 f_1(x) + p_2 f_2(x)$, $p_1 + p_2 = 1$

$$f_1(x) = \begin{cases} 2x, & 0 \leq x \leq 1 \\ 0, & \text{o/w} \end{cases}, \quad f_2(x) = \begin{cases} 2(1-x), & 0 \leq x \leq 1 \\ 0, & \text{o/w} \end{cases}$$

4a. Data: $\{x_1, x_2, x_3, x_4, x_5\} = \{0.1, 0.2, 0.3, 0.4, 0.7\}$. Write likelihood fn.

$$\begin{aligned} L(p_1, p_2 | x) &= \prod_{i=1}^5 f(x_i) \\ &= \prod_{i=1}^5 [p_1 f_1(x_i) + p_2 f_2(x_i)] \\ &= [p_1 f_1(0.1) + p_2 f_2(0.1)][p_1 f_1(0.2) + p_2 f_2(0.2)] \\ &\quad [p_1 f_1(0.3) + p_2 f_2(0.3)][p_1 f_1(0.4) + p_2 f_2(0.4)] \\ &\quad [p_1 f_1(0.7) + p_2 f_2(0.7)] \end{aligned}$$

$$\Rightarrow L(p_1, p_2 | x) = [0.2p_1 + 1.8p_2][0.4p_1 + 1.6p_2][0.6p_1 + 1.4p_2] \\ \quad [0.8p_1 + 1.2p_2][1.4p_1 + 0.6p_2]$$

Since $p_2 = 1 - p_1$, we can further simplify:

$$\begin{aligned} L(p_1 | x) &= [0.2p_1 + 1.8(1-p_1)][0.4p_1 + 1.6(1-p_1)][0.6p_1 + 1.4(1-p_1)] \\ &\quad [0.8p_1 + 1.2(1-p_1)][1.4p_1 + 0.6(1-p_1)] \end{aligned}$$

$$\Rightarrow L(p_1 | x) = [1.8 - 1.6p_1][1.6 - 1.2p_1][1.4 - 0.8p_1] \\ \quad [1.2 - 0.4p_1][0.6 + 0.8p_1]$$

4b. Data $\{x_1, x_2, x_3, x_4, x_5\} = \{0.1, 0.2, 0.3, 0.4, 0.9\}$. Write likelihood fn.

$$\begin{aligned} L(p_1, p_2 | x) &= \prod_{i=1}^5 f(x_i) \\ &= \prod_{i=1}^5 [p_1 f_1(x_i) + p_2 f_2(x_i)] \\ &= [p_1 f_1(0.1) + p_2 f_2(0.1)][p_1 f_1(0.2) + p_2 f_2(0.2)] \\ &\quad [p_1 f_1(0.3) + p_2 f_2(0.3)][p_1 f_1(0.4) + p_2 f_2(0.4)] \\ &\quad [p_1 f_1(0.9) + p_2 f_2(0.9)] \end{aligned}$$

$$\Rightarrow L(p_1, p_2 | x) = [0.2p_1 + 1.8p_2][0.4p_1 + 1.6p_2][0.6p_1 + 1.4p_2] \\ \quad [0.8p_1 + 1.2p_2][1.8p_1 + 0.2p_2]$$

• Since $p_2 = 1 - p_1$,

$$L(p_1 | x) = [0.2p_1 + 1.8(1-p_1)][0.4p_1 + 1.6(1-p_1)] \\ [0.6p_1 + 1.4(1-p_1)][0.8p_1 + 1.2(1-p_1)] \\ [1.8p_1 + 0.2(1-p_1)]$$

$$L(p_1 | x) = [1.8 - 1.6p_1][1.6 - 1.2p_1][1.4 - 0.8p_1] \\ [1.2 - 0.4p_1][0.2 + 1.6p_1]$$

4C. Data $\{x_1, x_2, x_3, x_4, x_5\} = \{0.1, 0.2, 0.3, 0.6, 0.9\}$. Write likelihood fcn.

$$\begin{aligned} L(p_1, p_2 | x) &= \prod_{i=1}^5 f(x_i) \\ &= \prod_{i=1}^5 [p_1 f_1(x_i) + p_2 f_2(x_i)] \\ &= [p_1 f_1(0.1) + p_2 f_2(0.1)][p_1 f_1(0.2) + p_2 f_2(0.2)] \\ &\quad [p_1 f_1(0.3) + p_2 f_2(0.3)][p_1 f_1(0.6) + p_2 f_2(0.6)] \\ &\quad [p_1 f_1(0.9) + p_2 f_2(0.9)] \end{aligned}$$

$$\Rightarrow L(p_1, p_2 | x) = [0.2p_1 + 1.8p_2][0.4p_1 + 1.6p_2][0.6p_1 + 1.4p_2] \\ [1.2p_1 + 0.8p_2][1.8p_1 + 0.2p_2]$$

• Since $p_2 = 1 - p_1$,

$$L(p_1 | x) = [0.2p_1 + 1.8(1-p_1)][0.4p_1 + 1.6(1-p_1)] \\ [0.6p_1 + 1.4(1-p_1)][1.2p_1 + 0.8(1-p_1)] \\ [1.8p_1 + 0.2(1-p_1)]$$

$$\Rightarrow L(p_1 | x) = [1.8 - 1.6p_1][1.6 - 1.2p_1][1.4 - 0.8p_1] \\ [0.8 + 0.4p_1][0.2 + 1.6p_1]$$

6. $X_1, X_2, X_3, X_4 \sim N_3(\mu, \Sigma)$:

$$X = [x_{jk}] = \begin{bmatrix} 3 & 6 & 0 \\ 4 & 4 & 3 \\ ? & 8 & 3 \\ 5 & ? & ? \end{bmatrix} \quad \begin{aligned} X_1^T &= \text{obs 1} \\ X_2^T &= \text{obs 2} \\ X_3^T &= \text{obs 3} \\ X_4^T &= \text{obs 4} \end{aligned}$$

Use EM algo. to estimate μ + Σ .

6a. Impute values to obtain \tilde{X} , data matrix wr missing values filled by initial estimates.

$$\cdot \tilde{X}_{31} = \frac{3+4+5}{3} = \frac{12}{3} = 4$$

$$\cdot \tilde{X}_{42} = \frac{6+4+8}{3} = \frac{18}{3} = 6$$

$$\cdot \tilde{X}_{43} = \frac{0+3+3}{3} = \frac{6}{3} = 2$$

So,

$$\tilde{X} = \begin{bmatrix} 3 & 6 & 0 \\ 4 & 4 & 3 \\ 4 & 8 & 3 \\ 5 & 6 & 2 \end{bmatrix}$$

6b. Find $\tilde{\mu}$, $\tilde{\Sigma}$ estimated from \tilde{X} .

$$\cdot \tilde{\mu} = \begin{bmatrix} (3+4+4+5)/4 \\ (6+4+8+6)/4 \\ (0+3+3+2)/4 \end{bmatrix} = \begin{bmatrix} \frac{16}{4} \\ \frac{24}{4} \\ \frac{8}{4} \end{bmatrix} = \begin{bmatrix} 4 \\ 6 \\ 2 \end{bmatrix} = \tilde{\mu}$$

$$\cdot \tilde{\Sigma} = \begin{bmatrix} \tilde{\sigma}_1^2 & \tilde{\sigma}_{12} & \tilde{\sigma}_{13} \\ \tilde{\sigma}_{12} & \tilde{\sigma}_2^2 & \tilde{\sigma}_{23} \\ \tilde{\sigma}_{13} & \tilde{\sigma}_{23} & \tilde{\sigma}_3^2 \end{bmatrix}$$

$$\cdot \tilde{\sigma}_1^2 = \frac{1}{4} [(3-4)^2 + (4-4)^2 + (4-4)^2 + (5-4)^2] = \frac{1}{4}[2] = \frac{1}{2}$$

$$\cdot \tilde{\sigma}_2^2 = \frac{1}{4} [(6-6)^2 + (4-6)^2 + (8-6)^2 + (6-6)^2] = \frac{1}{4}[4+4] = 2$$

$$\cdot \tilde{\sigma}_3^2 = \frac{1}{4} [(0-2)^2 + (3-2)^2 + (3-2)^2 + (2-2)^2] = \frac{1}{4}[4+1+1] = \frac{3}{2}$$

$$\cdot \tilde{\sigma}_{12} = \frac{1}{4} [(3-4)(6-6) + (4-4)(4-6) + (4-4)(8-6) + (5-4)(6-6)] = \frac{1}{4}[0] = 0$$

$$\cdot \tilde{\sigma}_{13} = \frac{1}{4} [(3-4)(0-2) + (4-4)(3-2) + (4-4)(3-2) + (5-4)(2-2)] = \frac{1}{4}[2] = \frac{1}{2}$$

$$\begin{aligned}\tilde{\sigma}_{23} &= \frac{1}{4} [(6-6)(0-2) + (4-6)(3-2) + (8-6)(3-2) + (6-6)(2-2)] \\ &= \frac{1}{4} [0 - 2 + 2 + 0] \\ &= 0\end{aligned}$$

So,

$$\tilde{\Sigma} = \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 2 & 0 \\ \frac{1}{2} & 0 & \frac{3}{2} \end{bmatrix}$$

6c. Iterate to find the first revised estimates:

6c(i). Find revised estimate of x_{31} , using estimated $\tilde{\mu}, \tilde{\Sigma}$ from Part (b). What are updated estimates of $\tilde{\mu}, \tilde{\Sigma}$?

$$\begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} \approx N_3 \left(\begin{bmatrix} 4 \\ 6 \\ 2 \end{bmatrix}, \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 2 & 0 \\ \frac{1}{2} & 0 & \frac{3}{2} \end{bmatrix} \right)$$

We partition as

$$\begin{aligned}X_{(1)} &\rightarrow \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \approx N_2 \left(\begin{bmatrix} 4 \\ 6 \end{bmatrix}, \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 2 & 0 \end{bmatrix} \right) \\ X_{(23)} &\rightarrow \begin{bmatrix} X_1 \\ X_3 \end{bmatrix} \approx N_2 \left(\begin{bmatrix} 4 \\ 2 \end{bmatrix}, \begin{bmatrix} \frac{1}{2} & 0 & \frac{3}{2} \end{bmatrix} \right)\end{aligned}$$

So,

$$\begin{aligned}X_{(1)} | X_{(23)} = X_{(23)} &\approx N(\tilde{\mu}_{(1)} + \tilde{\Sigma}_{(1)(23)} \tilde{\Sigma}_{(23)(23)}^{-1} (X_{(23)} - \tilde{\mu}_{(23)}), \tilde{\Sigma}_{(1)(1)} - \tilde{\Sigma}_{(1)(23)} \tilde{\Sigma}_{(23)(23)}^{-1} \tilde{\Sigma}_{(23)(1)}) \\ &= N\left(4 + [0 \ \frac{1}{2}] \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{2}{3} \end{bmatrix} \left(\begin{bmatrix} x_2 \\ x_3 \end{bmatrix} - \begin{bmatrix} 6 \\ 2 \end{bmatrix}\right), \frac{1}{2} - [0 \ \frac{1}{2}] \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix}\right) \\ &= N\left(4 + [0 \ \frac{1}{3}] \begin{bmatrix} x_2 - 6 \\ x_3 - 2 \end{bmatrix}, \frac{1}{2} - [0 \ \frac{1}{2}] \begin{bmatrix} 0 \\ \frac{1}{3} \end{bmatrix}\right) \\ &= N\left(4 + \frac{1}{3}(x_3 - 2), \frac{1}{2} - \frac{1}{6}\right) \\ &= N\left(4 + \frac{1}{3}x_3 - \frac{2}{3}, \frac{1}{3}\right) \\ &= N\left(\frac{10}{3} + \frac{1}{3}x_3, \frac{1}{3}\right)\end{aligned}$$

• So, $\tilde{X}_{31} = \mathbb{E}[\tilde{X}_{(1)} | X_{(23)} = (8, 3)]$
 $= \frac{10}{3} + \frac{1}{3}(3)$

$$\tilde{X}_{31} = \frac{13}{3} \approx 4.33$$

• Now,

$$\tilde{\mathbf{X}} = \begin{bmatrix} 3 & 6 & 0 \\ 4 & 4 & 3 \\ \frac{13}{3} & 8 & 3 \\ 5 & 6 & 2 \end{bmatrix}$$

• So

$$\tilde{\mu} = \begin{bmatrix} (3+4+\frac{13}{3}+5)/4 \\ (6+4+8+6)/4 \\ (0+3+3+2)/4 \end{bmatrix} = \begin{bmatrix} \frac{49}{3}/4 \\ 24/4 \\ 8/4 \end{bmatrix} = \begin{bmatrix} \frac{49}{12} \\ 6 \\ 2 \end{bmatrix} = \tilde{\boldsymbol{\mu}}$$

• $\tilde{\sigma}_2^2$, $\tilde{\sigma}_3^2$, & $\tilde{\sigma}_{23}$ are unchanged, since we have only updated \tilde{X}_{31} .

$$\begin{aligned} \tilde{\sigma}_1^2 &= \frac{1}{4} \left[(3 - \frac{49}{12})^2 + (4 - \frac{49}{12})^2 + (\frac{13}{3} - \frac{49}{12})^2 + (5 - \frac{49}{12})^2 \right] \\ &= \frac{1}{4} \left[\frac{169}{144} + \frac{1}{144} + \frac{1}{16} + \frac{121}{144} \right] \\ &= \frac{1}{4} \left[\frac{25}{12} \right] \\ &= \frac{25}{48} \end{aligned}$$

$$\begin{aligned} \tilde{\sigma}_{12} &= \frac{1}{4} \left[(3 - \frac{49}{12})(6 - 6) + (4 - \frac{49}{12})(4 - 6) + (\frac{13}{3} - \frac{49}{12})(8 - 6) + (5 - \frac{49}{12})(6 - 6) \right] \\ &= \frac{1}{4} \left[0 + (-\frac{1}{12})(-2) + (\frac{1}{4})(2) + 0 \right] \\ &= \frac{1}{4} \left[\frac{1}{6} + \frac{1}{2} \right] \\ &= \frac{2}{12} \\ &= \frac{1}{6} \end{aligned}$$

$$\begin{aligned} \tilde{\sigma}_{13} &= \frac{1}{4} \left[(3 - \frac{49}{12})(0 - 2) + (4 - \frac{49}{12})(3 - 2) + (\frac{13}{3} - \frac{49}{12})(3 - 2) + (5 - \frac{49}{12})(2 - 2) \right] \\ &= \frac{1}{4} \left[(-\frac{13}{12})(-2) + (-\frac{1}{12})(1) + (\frac{1}{4})(1) + 0 \right] \\ &= \frac{1}{4} \left[\frac{13}{6} - \frac{1}{12} + \frac{1}{4} \right] \\ &= \frac{1}{4} \left[\frac{7}{3} \right] \\ &= \frac{7}{12} \end{aligned}$$

• So,

$$\tilde{\Sigma} = \begin{bmatrix} \frac{25}{48} & \frac{1}{6} & \frac{7}{12} \\ \frac{1}{6} & 2 & 0 \\ \frac{7}{12} & 0 & \frac{3}{2} \end{bmatrix}$$

6(c)(ii) Find revised estimates of X_{42} , X_{43} , using estimated $\tilde{\mu} + \tilde{\Sigma}$ from Step (c)(i). Write out updated $\tilde{\mu}$, $\tilde{\Sigma}$ after first iteration.

$$\bullet \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} \approx N_3 \left(\begin{bmatrix} \frac{49}{12} \\ 6 \\ 2 \end{bmatrix}, \begin{bmatrix} \frac{25}{78} & \frac{1}{6} & \frac{7}{12} \\ \frac{1}{6} & 2 & 0 \\ \frac{7}{12} & 0 & \frac{3}{2} \end{bmatrix} \right)$$

• We partition as

$$\begin{aligned} X_{(1)} &\rightarrow \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \approx N_2 \left(\begin{bmatrix} \frac{49}{12} \\ 6 \end{bmatrix}, \begin{bmatrix} \frac{25}{48} & \frac{1}{6} \\ \frac{1}{6} & 2 \end{bmatrix} \right) \\ X_{(23)} &\rightarrow \begin{bmatrix} X_1 \\ X_3 \end{bmatrix} \approx N_2 \left(\begin{bmatrix} \frac{49}{12} \\ 2 \end{bmatrix}, \begin{bmatrix} \frac{7}{12} & 0 \\ 0 & \frac{3}{2} \end{bmatrix} \right) \\ &\quad \begin{array}{ccc} \downarrow \tilde{\mu}_{(1)} & \downarrow \tilde{\Sigma}_{(1)(1)} & \sqrt{\tilde{\Sigma}_{(1)(23)}} \\ \tilde{\mu}_{(23)} & \tilde{\Sigma}_{(23)(1)} & \tilde{\Sigma}_{(23)(23)} \end{array} \end{aligned}$$

• So,

$$\begin{aligned} X_{(23)} | X_{(1)} = x_{(1)} &\approx N(\tilde{\mu}_{(23)} + \tilde{\Sigma}_{(23)(1)} \tilde{\Sigma}_{(1)(1)}^{-1} (x_{(1)} - \tilde{\mu}_{(1)}), \tilde{\Sigma}_{(23)(23)} - \tilde{\Sigma}_{(23)(1)} \tilde{\Sigma}_{(1)(1)}^{-1} \tilde{\Sigma}_{(1)(23)}) \\ &= N \left(\begin{bmatrix} 6 \\ 2 \end{bmatrix} + \begin{bmatrix} \frac{1}{6} \\ \frac{7}{12} \end{bmatrix} \cdot \frac{48}{25} (x_{(1)} - \frac{49}{12}), \begin{bmatrix} 2 & 0 \\ 0 & \frac{3}{2} \end{bmatrix} - \begin{bmatrix} \frac{1}{6} \\ \frac{7}{12} \end{bmatrix} \frac{48}{25} \begin{bmatrix} \frac{1}{6} & \frac{7}{12} \end{bmatrix} \right) \\ &= N \left(\begin{bmatrix} 6 \\ 2 \end{bmatrix} + \begin{bmatrix} \frac{8}{25} \\ \frac{28}{25} \end{bmatrix} (x_{(1)} - \frac{49}{12}), \begin{bmatrix} 2 & 0 \\ 0 & \frac{3}{2} \end{bmatrix} - \frac{48}{25} \begin{bmatrix} \frac{1}{36} & \frac{7}{72} \\ \frac{7}{72} & \frac{49}{144} \end{bmatrix} \right) \\ &= N \left(\begin{bmatrix} 6 \\ 2 \end{bmatrix} + \begin{bmatrix} \frac{8}{25} x_{(1)} - \frac{98}{75} \\ \frac{28}{25} x_{(1)} - \frac{343}{75} \end{bmatrix}, \begin{bmatrix} 2 & 0 \\ 0 & \frac{3}{2} \end{bmatrix} - \begin{bmatrix} \frac{4}{75} & \frac{14}{75} \\ \frac{14}{75} & \frac{49}{75} \end{bmatrix} \right) \\ &= N \left(\begin{bmatrix} \frac{352}{75} + \frac{8}{25} x_{(1)} \\ -\frac{193}{75} + \frac{28}{25} x_{(1)} \end{bmatrix}, \begin{bmatrix} \frac{146}{75} & -\frac{14}{75} \\ -\frac{14}{75} & \frac{127}{150} \end{bmatrix} \right) \end{aligned}$$

$$\text{So, } \begin{bmatrix} \tilde{X}_{42} \\ \tilde{X}_{43} \end{bmatrix} = \mathbb{E} \left[\begin{bmatrix} \tilde{X}_{(23)} | X_{(1)} = 5 \end{bmatrix} \right]$$

$$\begin{aligned} &= \begin{bmatrix} \frac{352}{75} + \frac{8}{25}(5) \\ -\frac{193}{75} + \frac{28}{25}(5) \end{bmatrix} \\ &= \begin{bmatrix} \frac{352}{75} + \frac{8}{5} \\ -\frac{193}{75} + \frac{28}{5} \end{bmatrix} \end{aligned}$$

$$= \begin{bmatrix} \frac{472}{75} \\ \frac{227}{75} \end{bmatrix}$$

$$\Rightarrow \tilde{X}_{42} = \frac{472}{75} \approx 6.293$$

$$\tilde{X}_{43} = \frac{227}{75} \approx 3.027$$

Now,

$$\tilde{X} = \begin{bmatrix} 3 & 6 & 0 \\ 4 & 4 & 3 \\ \frac{13}{3} & 8 & 3 \\ 5 & \frac{472}{75} & \frac{227}{75} \end{bmatrix}$$

So

$$\tilde{\mu} = \begin{bmatrix} (3+4+\frac{13}{3}+5)/4 \\ (6+4+8+\frac{472}{75})/4 \\ (0+3+3+\frac{227}{75})/4 \end{bmatrix} = \begin{bmatrix} \frac{49}{3}/4 \\ \frac{1822}{75}/4 \\ \frac{677}{75}/4 \end{bmatrix} = \begin{bmatrix} \frac{49}{12} \\ \frac{911}{150} \\ \frac{677}{300} \end{bmatrix} \approx \begin{bmatrix} 4.083 \\ 6.073 \\ 2.257 \end{bmatrix} = \tilde{\mu}$$

$\tilde{\sigma}_1^2$ is unchanged.

$$\begin{aligned} \tilde{\sigma}_2^2 &= \frac{1}{4} [(6 - \frac{91}{150})^2 + (4 - \frac{91}{150})^2 + (8 - \frac{91}{150})^2 + (\frac{472}{75} - \frac{91}{150})^2] \\ &= \frac{1}{4} [\frac{121}{22500} + \frac{96721}{22500} + \frac{83521}{22500} + \frac{121}{2500}] \\ &= \frac{1}{4} [\frac{15121}{1875}] \end{aligned}$$

$$= \frac{15121}{7500} \approx 2.016$$

$$\begin{aligned} \tilde{\sigma}_3^2 &= \frac{1}{4} [(0 - \frac{677}{300})^2 + (3 - \frac{677}{300})^2 + (3 - \frac{677}{300})^2 + (\frac{227}{75} - \frac{677}{300})^2] \\ &= \frac{1}{4} [\frac{458329}{90000} + \frac{49729}{90000} + \frac{49729}{90000} + \frac{5929}{10000}] \\ &= \frac{1}{4} [\frac{50929}{7500}] \end{aligned}$$

$$= \frac{50929}{30000} \approx 1.698$$

$$\begin{aligned} \tilde{\sigma}_{12}^2 &= \frac{1}{4} [(3 - \frac{49}{12})(6 - \frac{91}{150}) + (4 - \frac{49}{12})(4 - \frac{91}{150}) + (\frac{13}{3} - \frac{49}{12})(8 - \frac{91}{150}) + (5 - \frac{49}{12})(\frac{472}{75} - \frac{91}{150})] \\ &= \frac{1}{4} [(-\frac{13}{12})(-\frac{11}{150}) + (-\frac{1}{12})(-\frac{311}{150}) + (\frac{1}{4})(\frac{289}{150}) + (\frac{11}{12})(\frac{11}{50})] \\ &= \frac{1}{4} [\frac{143}{1800} + \frac{311}{1800} + \frac{289}{600} + \frac{121}{600}] \\ &= \frac{1}{4} [\frac{421}{450}] \end{aligned}$$

$$= \frac{421}{1800} \approx 0.234$$

$$\begin{aligned} \tilde{\sigma}_{13}^2 &= \frac{1}{4} [(3 - \frac{49}{12})(0 - \frac{677}{300}) + (4 - \frac{49}{12})(3 - \frac{677}{300}) + (\frac{13}{3} - \frac{49}{12})(3 - \frac{677}{300}) + (5 - \frac{49}{12})(\frac{227}{75} - \frac{677}{300})] \\ &= \frac{1}{4} [(-\frac{13}{12})(-\frac{677}{300}) + (-\frac{1}{12})(\frac{223}{300}) + (\frac{1}{4})(\frac{223}{300}) + (\frac{11}{12})(\frac{77}{100})] \\ &= \frac{1}{4} [\frac{8801}{3600} - \frac{223}{3600} + \frac{223}{1200} + \frac{847}{1200}] \\ &= \frac{1}{4} [\frac{2947}{900}] \end{aligned}$$

$$= \frac{2947}{3600} \approx 0.819$$

$$\begin{aligned}
 \widetilde{\sigma}_{23} &= \frac{1}{4} \left[(6 - \frac{911}{150})(0 - \frac{677}{300}) + (4 - \frac{911}{150})(3 - \frac{677}{300}) + (8 - \frac{911}{150})(3 - \frac{677}{300}) \right. \\
 &\quad \left. + (\frac{472}{75} - \frac{911}{150})(\frac{227}{75} - \frac{677}{300}) \right] \\
 &= \frac{1}{4} \left[(-\frac{11}{150})(-\frac{677}{300}) + (-\frac{311}{150})(\frac{223}{300}) + (\frac{289}{150})(\frac{223}{300}) + (\frac{11}{50})(\frac{77}{100}) \right] \\
 &= \frac{1}{4} \left[\frac{7447}{45000} - \frac{69353}{45000} + \frac{64447}{45000} + \frac{847}{5000} \right] \\
 &= \frac{1}{4} \left[\frac{847}{3750} \right] \\
 &= \frac{847}{15000} \approx 0.056
 \end{aligned}$$

Thus,

$$\widetilde{\Sigma} = \begin{bmatrix} \frac{25}{48} & \frac{421}{1800} & \frac{2947}{3600} \\ \frac{421}{1800} & \frac{15121}{7500} & \frac{847}{15000} \\ \frac{2947}{3600} & \frac{847}{15000} & \frac{50929}{30000} \end{bmatrix} \approx \begin{bmatrix} 0.521 & 0.234 & 0.819 \\ 0.234 & 2.016 & 0.056 \\ 0.819 & 0.056 & 1.698 \end{bmatrix}$$

7. Dissimilarity measures d_1, d_2 defined in space S are global-order equivalent if

$$d_1(x, y) \leq d_1(w, z) \Leftrightarrow d_2(x, y) \leq d_2(w, z) \quad \forall x, y, w, z \in S$$

7a Let $d(a, b) = \sqrt{(a-b)^T(a-b)}$ be Euclidean dist b/w $a, b \in \mathbb{R}^P$,
 $D(a, b) = (d(a, b))^2$.

Show $d_1 = d$, $d_2 = D$ are global-order equivalent.

• (\Rightarrow) Let $x, y, w, z \in \mathbb{R}^P$ such that $d(x, y) \leq d(w, z)$.

Then

$$\begin{aligned} & \sqrt{(x-y)^T(x-y)} \leq \sqrt{(w-z)^T(w-z)} \\ & \Rightarrow \sqrt{(x-y)^T(x-y)} \sqrt{(x-y)^T(x-y)} \leq \sqrt{(w-z)^T(w-z)} \sqrt{(x-y)^T(x-y)} \quad \text{since } \sqrt{(x-y)^T(x-y)} \geq 0 \\ & \Rightarrow (x-y)^T(x-y) \leq (w-z)^T(w-z) \sqrt{(w-z)^T(w-z)} \quad \text{since } \sqrt{(x-y)^T(x-y)} \leq \sqrt{(w-z)^T(w-z)} \\ & \Rightarrow D(x, y) \leq (w-z)^T(w-z) \\ & \qquad \qquad \qquad = D(w, z) \end{aligned}$$

Thus, $d(x, y) \leq d(w, z) \Rightarrow D(x, y) \leq D(w, z)$.

• (\Leftarrow) Let $x, y, w, z \in \mathbb{R}^P$ s.t. $D(x, y) \leq D(w, z)$.

Then

$$\begin{aligned} & (x-y)^T(x-y) \leq (w-z)^T(w-z) \\ & \Rightarrow \sqrt{(x-y)^T(x-y)} \leq \sqrt{(w-z)^T(w-z)} \quad \text{since the square root is} \\ & \qquad \qquad \qquad \text{a monotone function on } [0, \infty), \\ & \qquad \qquad \qquad \text{and } D(x, y), D(w, z) \geq 0 \end{aligned}$$

$$\Rightarrow d(x, y) \leq d(w, z)$$

Thus, $D(x, y) \leq D(w, z) \Rightarrow d(x, y) \leq d(w, z)$

So, d & D are global-order equivalent. \square

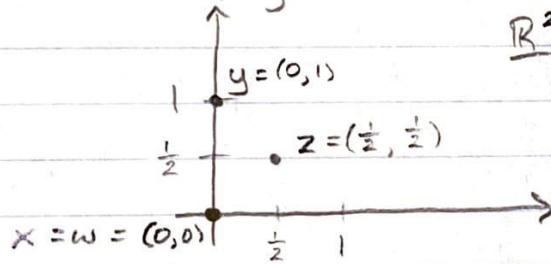
- 7b. Show by counterexample that goe does not hold
 for $d_1 = \text{Euclidean dist.}$ & $d_2 = \text{Canberra dist.}$

$$d_2(x, y) = \begin{cases} \sum_{i=1}^p \frac{|x_i - y_i|}{|x_i| + |y_i|}, & x_i - y_i \neq 0 \\ 0, & x_i - y_i = 0 \end{cases}$$

for $x, y \in \mathbb{R}^p$.

Consider points in \mathbb{R}^2 (i.e., $p=2$).

Let $x = (0, 0)$, $y = (0, 1)$, $w = (0, 0)$, $z = (\frac{1}{2}, \frac{1}{2})$:



$$\cdot d_1(x, y) = \sqrt{(1-0)^2 + (0-0)^2} = \sqrt{1+0} = 1$$

$$\cdot d_1(w, z) = \sqrt{(\frac{1}{2}-0)^2 + (\frac{1}{2}-0)^2} = \sqrt{\frac{1}{4} + \frac{1}{4}} = \sqrt{\frac{1}{2}} = \frac{1}{\sqrt{2}} \approx 0.707$$

$$\cdot \text{So, } d_1(x, y) = 1 > \frac{1}{\sqrt{2}} = d_1(w, z)$$

$$\cdot d_2(x, y) = 0 + \frac{|0-1|}{|0|+|1|} = 0 + \frac{1}{0+1} = 1$$

$$\cdot d_2(w, z) = \frac{|\frac{1}{2}-0|}{|\frac{1}{2}|+|0|} + \frac{|\frac{1}{2}-0|}{|\frac{1}{2}|+|0|} = \frac{\frac{1}{2}}{\frac{1}{2}} + \frac{\frac{1}{2}}{\frac{1}{2}} = 1+1=2$$

$$\cdot \text{So, } d_2(x, y) = 1 < 2 = d_2(w, z)$$

Thus, $\exists x, y, w, z \in \mathbb{R}^2$ s.t. $d_1(x, y) > d_1(w, z)$ but $d_2(x, y) < d_2(w, z)$. Thus, $d_1 \circ d_2$ are not global-order equivalent. \square