

2) (a) $P(A|C=0) = \frac{0.28}{0.4} = 0.7$

$P(B|C=0) = \frac{0.08}{0.32} = \frac{0.25}{0.4} = 0.2$

$P(A, B|C=0) = \frac{0.056}{0.4} = 0.14$

| | | | | |
|-------|-------|-------|------|------|
| 0.420 | 0.096 | 0.024 | 0.27 | 0.03 |
| 0.580 | 0.224 | 0.056 | 0.27 | 0.03 |
| | 0.320 | 0.080 | 0.54 | 0.06 |

$P(A=0|C=0) = 1-0.7=0.3$

$P(B=0|C=0) = 1-0.2=0.8$

$P(A=1, B=0|C=0) = 0.224/0.4 = 0.56$

$P(A=0, B=1|C=0) = 0.06$

$P(A=0, B=0|C=0) = 0.24$

(b) $P(A|C=1) = \frac{0.27+0.03}{0.6} = 0.5$

$P(B|C=1) = \frac{0.06}{0.60} = 0.1$

$P(A, B|C=1) = \frac{0.07}{0.60} = 0.05$

$P(A=0|C=1) = 0.5$

$P(B=0|C=1) = 0.9$

$P(A=1, B=0|C=1) = 0.27/0.60 = 0.45$

$P(A=0, B=1|C=1) = 0.03/0.60 = 0.05$

$P(A=0, B=0|C=1) = 0.45$

(c) Yes, mutually independent b/c

i.e. $P(A \cap B) = P(A) \cdot P(B)$

[yes, conditionally indep.] ✓

$P(A|C=0) \cdot P(B|C=0) = P(A, B|C=0)$

$C=0$ $0.7 \cdot 0.2 = 0.14$

$C=1$ $0.5 \cdot 0.1 = 0.05$

for all cases.

(d) $P(A) = 0.580$ $A=1$

$P(B) = 0.14$ $B=1$

$P(A, B) = 0.086$ $A=B=1$

$P(A=0) = 0.42$

$P(B=0) = 0.86$

$P(A=0, B=0) = 0.366$

$P(A=1, B=0) = 0.494$

$P(A=0, B=1) = 0.054$

(e) $P(A) \cdot P(B) = P(A, B)$

$0.58 \cdot 0.14 \neq 0.086$

NOT INDEPENDENT

does not hold for all combos of A, B

(3) (a) $P(G) = P(G=0) = 2/8 = 1/4$ $P(G=1) = 3/4$

$P(X|G) \rightarrow P(\text{not } G) =$

| | | |
|--------------------|--------------------|--------------------|
| $P(B=0 G=0) = 0$ | $P(A=0 G=0) = 1$ | $P(C=0 G=0) = 1/4$ |
| $P(B=1 G=0) = 1$ | $P(A=0 G=1) = 1/6$ | $C=0 G=1 = 1/4$ |
| $P(B=0 G=1) = 0.5$ | $P(A=1 G=0) = 0$ | $C=1 G=0 = 1/4$ |
| $P(B=1 G=1) = 0.5$ | $P(A=1 G=1) = 5/6$ | $C=1 G=1 = 1/2$ |

$P(B=0|G=0) = 0$
 $P(B=1|G=0) = 1$
 $P(B=0|G=1) = 1/3$
 $P(B=1|G=1) = 2/3$

counting in graph

(b) $P(G=0) \cdot P(B=0|G=0) \cdot P(A=1|G=0) \cdot P(C=0|G=0) \cdot P(A=1|G=0)$

$= 1/4 \cdot 0 \cdot 1 \cdot 1/4 \cdot 0 = 0$

$P(G=1) \cdot P(B=0|G=1) \cdot P(A=1|G=1) \cdot P(C=0|G=1) \cdot P(A=1|G=1)$

$= 3/4 \cdot 1/2 \cdot 1/3 \cdot 1/2 \cdot 5/6 = \frac{5}{96}$ $G_1=1 \geq G_2=0$

$G=1$ for sample 9

$P(G=0) \cdot P(B=1|G=0) \cdot P(A=1|G=0) \cdot P(C=1|G=0) \cdot P(A=1|G=0)$

$= 1 \cdot 1 \cdot 1/2 \cdot 0 \cdot 1/4 = 0$

$P(G=1) \cdot P(B=1|G=1) \cdot P(A=1|G=1) \cdot P(C=1|G=1) \cdot P(A=1|G=1)$

$= \frac{3}{4} \cdot \frac{1}{2} \cdot \frac{1}{3} \cdot \frac{1}{2} \cdot \frac{5}{6} = \frac{5}{96}$

$G_1=1 \geq G_2=0$

sample 10: $G=1$

(c) $P(B=0|G=0) = 1/4$ $P(A=0|G=0) = 3/4$ $P(B=0|G=0) = 1/4$

$P(B=1|G=0) = 3/4$ $P(A=1|G=0) = 1/4$ $P(B=1|G=0) = 3/4$

$P(B=0|G=1) = 1/8$ $P(A=0|G=1) = 2/3$ $P(B=0|G=1) = 1/8$

$P(B=1|G=1) = 7/8$ $P(A=1|G=1) = 1/3$ $P(B=1|G=1) = 3/8$

$P(C=0|G=0) = 1/2$

$P(C=1|G=0) = 1/2$

$P(C=0|G=1) = 1/2$

$P(C=1|G=1) = 1/2$

$$\begin{aligned}
 (d) \quad & P(G=0) \cdot P(O=0|G=0) \cdot P(B=1|G=0) \cdot P(C=0|G=0) \cdot P(A=1|G=0) \\
 &= \frac{2}{8} \cdot \frac{1}{4} \cdot \frac{3}{4} \cdot \frac{1}{2} \cdot \frac{1}{4} = 0.00586
 \end{aligned}$$

$$\begin{aligned}
 & P(G=1) \cdot P(O=0|G=1) \cdot P(B=1|G=1) \cdot P(C=0|G=1) \cdot P(A=1|G=1) \\
 &= \frac{1}{2} \cdot \frac{1}{4} \cdot \frac{1}{2} \cdot \frac{3}{4} \cdot \frac{6}{8} = 0.0527
 \end{aligned}$$

for sample 9, $G_1 > G_0 \Rightarrow \boxed{G=1}$

$$\begin{aligned}
 & P(G=0) \cdot P(O=1|G=0) \cdot P(B=1|G=0) \cdot P(C=1|G=0) \cdot P(A=1|G=0) \\
 &= \frac{2}{8} \cdot \frac{3}{4} \cdot \frac{3}{4} \cdot \frac{1}{4} \cdot \frac{1}{4} = 0.0176
 \end{aligned}$$

$$\begin{aligned}
 & P(G=1) \cdot P(O=1|G=1) \cdot P(B=1|G=1) \cdot P(C=1|G=1) \cdot P(A=1|G=1) \\
 &= \frac{1}{2} \cdot \frac{3}{8} \cdot \frac{1}{2} \cdot \frac{3}{4} \cdot \frac{6}{8} = 0.0527
 \end{aligned}$$

For sample 10: $G_1 > G_0 \Rightarrow \boxed{G=1}$

$$(4) \quad P(x^1, x^2, \dots, y^1, y^2, \dots, y^m) = \prod_{k=1}^m P(x^k, y^k)$$

$$= \prod_{k=1}^m P(y^{(k)}) \cdot P(x^{(k)} | y^{(k)}) = \prod_{k=1}^m P(y^{(k)}) \cdot P$$

=

Problem 4

$$(a) \prod_{i=1}^m P(x^{(i)}, y^{(i)}) = \prod_{i=1}^m P(y^{(i)}) \cdot P(x^{(i)} | y^{(i)})$$

$$= \prod_{i=1}^m \left[\theta_0^{1(y^{(i)}=0)} (1-\theta_0)^{1(y^{(i)}=1)} \prod_{j=1}^n \left(\prod_{k=1}^S \theta_{j,k,y=0}^{1[x_j^{(i)}=k, y^{(i)}=0]} \cdot \left(1 - \sum_{k=1}^{S-1} \theta_{j,k,y=0}\right)^{1[x_j^{(i)}=S, y^{(i)}=0]} \right. \right.$$

$$(b) \log \cdot (a) = \sum_{i=1}^m [y^{(i)}=1] \log(1-\theta_0) + [y^{(i)}=0] \log \theta_0$$

$$+ \sum_{j=1}^n \sum_{k=1}^S 1[x_j^{(i)}=k, y^{(i)}=0] \log \theta_{j,k,y=0}$$

$$+ 1[x_j^{(i)}=S, y^{(i)}=0] \log \left(1 - \sum_{k=1}^{S-1} \theta_{j,k,y=0}\right)$$

$$+ 1[x_j^{(i)}=k, y^{(i)}=1] \log \theta_{j,k,y=1}$$

$$+ 1[x_j^{(i)}=S, y^{(i)}=1] \log \left(1 - \sum_{k=1}^{S-1} \theta_{j,k,y=1}\right)$$

$$\frac{\partial}{\partial \theta_0} = \sum_{i=1}^m [y^{(i)}=1] \log(1-\theta_0) + [y^{(i)}=0] \log \theta_0$$

$$N_{k,0} = 1[x_j^{(i)}=k, y^{(i)}=0] \quad N_{k,1} = 1[x_j^{(i)}=k, y^{(i)}=1] \quad N_{S,0} = 1[x_j^{(i)}=S, y^{(i)}=0]$$

$$N_{S,1} = 1[x_j^{(i)}=S, y^{(i)}=1]$$

$$\frac{-N_1}{1-\theta_0} + \frac{N_0}{\theta_0} = 0 \quad \text{where } N_1 = \sum_{i=1}^m 1[y^{(i)}=1] \quad N_0 = \sum_{i=1}^m 1[y^{(i)}=0]$$

$$\hat{\theta}_0 = \frac{N_0}{(N_0 + N_1)}$$

$$\sum_{j=1}^n$$

$$\sum_{k=1}^S$$

disappears % w.r.t k

$$\frac{\partial \log \cdot (a)}{\partial \theta_{j,k,y=0}} = \sum_{i=1}^m \sum_{j=1}^n \frac{N_{k,0}}{\theta_{j,k,y=0}} + \frac{-N_{S,0}}{1 - \sum_{k=1}^{S-1} \theta_{j,k,y=0}} = 0$$

$$\Rightarrow \frac{\sum_{i=1}^m N_{k,0}}{\theta_{j,k,y=0}} = \frac{\sum_{i=1}^m N_{S,0}}{\theta_{j,S,y=0}}$$

$$\hat{\theta}_{j,k,y=0} = \theta_{j,S,y=0} \frac{\sum_{i=1}^m N_{k,0}}{\sum_{i=1}^m N_{S,0}}$$

next

$$\hat{\theta}_{0,k|y=1} = \theta_{0,s|y=1} = \frac{\sum_{i=1}^m N_{k,i}}{\sum_{i=1}^m N_{s,i}} \quad \text{by symmetry}$$

$\hat{\theta}_0$ the proportion of all trials whose coin flip landed $y=0$

$\hat{\theta}_{0,k|y=0}$ the proportion to the return of k trials whose ^{flip} y is
~~not the max val~~ over the k trials whose ~~flip~~ y is 0 is
 s (the max) all with ~~flip~~ $y=0$

$\hat{\theta}_{s,k|y=1}$ the same thing except don't flip $y=1$

5-

$$f_{xy}(x,y) = \frac{\exp \left\{ \frac{-1}{2(\sigma_1 \sigma_2 \sqrt{1-\rho_{xy}})} \left[\left(\frac{x-m_1}{\sigma_1} \right)^2 - 2\rho_{xy} \left(\frac{x-m_1}{\sigma_1} \right) \left(\frac{y-m_2}{\sigma_2} \right) + \left(\frac{y-m_2}{\sigma_2} \right)^2 \right] \right\}}{2\pi\sigma_1\sigma_2\sqrt{1-\rho_{xy}^2}}$$

$$\Sigma = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \rightarrow |\Sigma| = ad-bc$$

$$ad-bc = \sigma_1^2 \sigma_2^2 - \rho_{xy}^2 \sigma_1^2 \sigma_2^2$$

$$\Sigma = \begin{bmatrix} \sigma_1^2 & \rho_{xy}\sigma_1\sigma_2 \\ \rho_{xy}\sigma_1\sigma_2 & \sigma_2^2 \end{bmatrix} \quad \begin{matrix} ad = \uparrow \\ bc = \downarrow \end{matrix}$$

$$\Sigma^{-1} = \frac{1}{\sigma_1^2 \sigma_2^2 - \rho_{xy}^2 \sigma_1^2 \sigma_2^2} \begin{bmatrix} \sigma_2^2 & -\rho_{xy}\sigma_1\sigma_2 \\ -\rho_{xy}\sigma_1\sigma_2 & \sigma_1^2 \end{bmatrix}$$

$$z = \begin{bmatrix} x \\ y \end{bmatrix} \quad \mu = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$$

$$(z-\mu)^T \Sigma^{-1} (z-\mu)$$

$$= \begin{bmatrix} x-c_1 \\ y-c_2 \end{bmatrix}^T \begin{bmatrix} x-c_1 \\ y-c_2 \end{bmatrix}$$

$$= \frac{1}{(\sigma_1^2 \sigma_2^2 - \rho_{xy}^2 \sigma_1^2 \sigma_2^2)} \left[\left(\frac{x-c_1}{\sigma_1} \right)^2 - 2\rho_{xy} \frac{(x-c_1)(y-c_2)}{\sigma_1 \sigma_2} + \left(\frac{y-c_2}{\sigma_2} \right)^2 \right]$$

$$\mu = \begin{bmatrix} m_1 \\ m_2 \end{bmatrix}$$

$$c) \quad \overline{\text{cov}[X_i, X_j]} =$$

X_i, X_j are independent

$$E(X_i, X_j) = \underbrace{E[X_i] \cdot E[X_j]} \leftarrow \int \int x_i x_j p(x_i, x_j)$$

$$= \int x_i p_{x_i} \int x_j p_{x_j}$$

$$\text{cov}[X_i, X_j] = \cancel{E[X_i X_j]} - \cancel{E[X_i]} \cancel{E[X_j]}$$

$$= 0$$

$$\begin{bmatrix} \sigma_1^2 & 0 & 0 & 0 & 0 \\ & \sigma_2^2 & 0 & 0 & 0 \\ & & \ddots & & \\ & & & 0 & \\ & & & & \sigma_n^2 \end{bmatrix}$$