

Q1

(1)

$$+ x_1^2 + x_1 x_2 + \frac{x_2^2}{2} = \frac{1}{2(1-\rho^2)} \left(\frac{(x_1 - \mu_1)^2}{\sigma_1^2} - \frac{2\rho(x_1 - \mu_1)(x_2 - \mu_2)}{\sigma_1 \sigma_2} + \frac{(x_2 - \mu_2)^2}{\sigma_2^2} \right)$$

expand both sides.

$$\text{RHS} = \frac{x_1^2}{2(1-\rho^2)\sigma_1^2} + x_1 \left(\frac{-2\rho\mu_2}{2(1-\rho^2)\sigma_1^2} + \frac{2\rho\mu_1}{2(1-\rho^2)\sigma_1\sigma_2} \right)$$

$$\frac{x_2^2}{2(1-\rho^2)\sigma_2^2} + x_2 \left(\frac{-2\rho\mu_1}{2(1-\rho^2)\sigma_2^2} + \frac{2\rho\mu_2}{2(1-\rho^2)\sigma_1\sigma_2} \right)$$

$$+ \frac{-2\rho x_1 x_2}{2(1-\rho^2)\sigma_1\sigma_2} + \left(\frac{\mu_1^2}{\sigma_1^2} + \frac{\mu_2^2}{\sigma_2^2} - \frac{2\rho\mu_1\mu_2}{\sigma_1\sigma_2} \right) \frac{1}{2(1-\rho^2)}$$

$$\frac{x_1^2}{2(1-\rho^2)\sigma_1^2} = x_1^2$$

$$\frac{x_2^2}{2(1-\rho^2)\sigma_2^2} = \frac{x_2^2}{2}$$

$$\frac{-2\rho}{2(1-\rho^2)\sigma_1\sigma_2} x_1 x_2 = x_1 x_2$$

$$\left(\frac{\mu_2 \rho}{\sigma_1 \sigma_2} - \frac{\mu_1}{\sigma_1^2} \right) \frac{1}{(1-\rho^2)} = 0$$

$$\left(\frac{\rho \mu_1}{\sigma_1 \sigma_2} - \frac{\mu_2}{\sigma_2^2} \right) \frac{1}{(1-\rho^2)} = 0$$

$$\left(\frac{\mu_1^2}{\sigma_1^2} - \frac{2\rho\mu_1\mu_2}{\sigma_1\sigma_2} + \frac{\mu_2^2}{\sigma_2^2} \right) \frac{1}{2(1-\rho^2)} = 0$$

$\mu_1 = \mu_2 = 0$ solves eqns on right!

$$\textcircled{A} \quad \sigma_1^2 = \frac{1}{2(1-\rho^2)}$$

$$\textcircled{B} \quad \sigma_2^2 = \frac{1}{(1-\rho^2)}$$

 $\left(\frac{1}{2}\right)$

$$\textcircled{C} \quad \sigma_1 \sigma_2 = \frac{-\rho}{(1-\rho^2)}$$

$$\textcircled{C} = \frac{-\rho}{(1-\rho^2)} = \frac{1}{2(1-\rho^2)} \frac{1}{\sqrt{(1-\rho^2)}} = \frac{1}{(1-\rho^2)} \frac{1}{\sqrt{2}}$$

$$\leadsto \rho = -\frac{1}{\sqrt{2}}$$

$$\leadsto \sigma_1^2 = \frac{1}{2(1-\frac{1}{2})} = 1$$

$$\sigma_2^2 = \frac{1}{(1-\frac{1}{2})} = 2$$

$$\sigma_1 = 1, \sigma_2 = \sqrt{2}, \rho = -\frac{1}{\sqrt{2}}$$

 $\left(\frac{1}{2}\right)$

b) i) Marginal distribution.

$$\phi(x_1) = \frac{1}{\sigma_1 \sqrt{2\pi}} \exp\left(-\frac{(x_1 - \mu_1)^2}{2\sigma_1^2}\right)$$

$$= \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x_1^2}{2}\right)$$

 $\textcircled{1}$

b ii) Correlation $E[x_1, x_2] = \rho = -\frac{1}{\sqrt{2}}$

(1.3)

(1)

c) Note that

$$x_1^2 + x_1 x_2 + \frac{x_2^2}{2} = (x_1 \ x_2) \begin{pmatrix} a & b \\ b & c \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= ax_1^2 + 2bx_1x_2 + cx_2^2$$

$$a=1 \quad b=\frac{1}{2} \quad c=\frac{1}{2}$$

Hence,

$$\frac{1}{2} \Sigma^{-2} = \begin{pmatrix} 1 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

(1/2)

$$\Sigma^{-2} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}^{-1} = \frac{1}{1} \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix} = \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix}$$

Also note $\mu = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

(1/2)

Also note $\Sigma^{-2} = \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix} = \begin{pmatrix} 1 & -\frac{1}{\sqrt{2}}\sqrt{2} \\ -\frac{1}{\sqrt{2}}\sqrt{2} & (\sqrt{2})^2 \end{pmatrix} = \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix}$

$$\det(\Sigma^{-2}) = 2 - 1 = 1$$

$$\phi(x_1, x_2) = \frac{1}{2\pi} \exp\left(-\frac{1}{2} \mathbf{x}^T \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix} \mathbf{x}\right)$$

(1)

Q2

2-1

a) i) Sample space is all sets of possible rolls.

$$S = \{(1,1), (1,2), \dots, (1,6), (2,1), \dots, (2,6), \dots, (6,1), \dots, (6,5), (6,6)\}$$

ii) The prob. dist. $\phi: S \rightarrow \mathbb{R}$.

$$\phi_S(i, j) = \phi(i) \phi(j) \quad i, j = 1, \dots, 6, \text{ where } \phi(i), \phi(j) \text{ are given by the table.}$$

b) i) $\{A \geq 4\} = \{(i, j) \in S \mid j \in \{4, 5, 6\}, i \in \{1, 2, 3, \dots, 6\}\}$

$$P(A \geq 4) = \sum_{j=4}^6 \sum_{i=1}^6 \phi_S(i, j)$$

$$= \sum_{i=1}^6 \sum_{j=4}^6 \phi(i) \phi(j)$$

$$= \sum_{i=1}^6 \phi(i) \underbrace{\sum_{j=4}^6 \phi(j)}_1$$

$$= 0.15 + 0.15 + 0.3 = 0.6 \quad \textcircled{1}$$

ii) $\{B \geq 4\} = \{(i, j) \in S \mid i \in \{1, \dots, 6\}, j \in \{4, \dots, 6\}\}$

$$P(B \geq 4) = \sum_{j=4}^6 \phi(j)$$

$$= 0.6$$

①

$$c) i) P_r(A \geq 4 \cup B \geq 4) = P_r(A \geq 4) + P_r(B \geq 4)$$

$$= P_r(A \geq 4 \cap B \leq 4) + P_r(A \geq 4 \cap B \geq 4) + P_r(A \leq 4 \cap B \geq 4)$$

$$P_r(A \geq 4 \cap B \leq 4) = (0.6) \cdot (0.4) = P_r(A \geq 4) (1 - P(B \geq 4))$$

$$P_r(A \geq 4 \cap B \geq 4) = (0.6)(0.6)$$

$$P_r(A \leq 4 \cap B \geq 4) = (0.4)(0.6)$$

$$P_r(A \geq 4 \cup B \geq 4) = 0.6(0.4 + 0.6 + 0.4)$$

$$= (0.6)(1.4)$$

$$= 0.84$$

②

∴ (A) Divide event into mutually disjoint events + use sum of probabilities law.

(B) Use independence of dice rolls for $P_r(A \cap B) = P_r(A)P_r(B)$

(C) Compute $P_r(B \leq 4) = 1 - P_r(B \geq 4)$. ③

$$\begin{aligned} \text{Cii)} \quad \Pr(A \geq 4 \cap B \geq 4) \\ = \Pr(A \geq 4) \Pr(B \geq 4) = (0.6)^2 = 0.36 \end{aligned} \quad \left(\frac{1}{2}\right)$$

Independence of ~~events~~ die rolls ensures product of probabilities. $\left(\frac{1}{2}\right)$

$$\begin{aligned} \text{Ciii)} \quad \Pr(A + B) &= \Pr(A=2, B=6) &= (0.15)(0.3) \\ &+ \Pr(A=3, B=5) &+ (0.15)(0.15) \\ &+ \Pr(A=4, B=4) &(0.15)(0.15) \\ &+ \Pr(A=5, B=3) &(0.15)(0.15) \\ &+ \Pr(A=6, B=2) &(0.3)(0.15) \end{aligned}$$

$$\begin{aligned} &= (0.15)(0.3 + 3(0.15) + 0.3) \\ &= 0.1575 \end{aligned} \quad \left(\frac{1}{2}\right)$$

→ Disjoint events + sum over probabilities $\left(\frac{1}{2}\right)$

$$\begin{aligned} \text{Civ)} \quad \Pr(A+B=8 | A \geq 4) &= \frac{\Pr(A+B=8, A \geq 4)}{\Pr(A \geq 4)} \\ &= \frac{(0.15)(0.15) + (0.15)(0.15) + (0.3)(0.15)}{0.6} \\ &= 0.15 \end{aligned} \quad \left(\frac{1}{2}\right)$$

→ Rule of conditional probability + then compute by sum of disjoint events. $\left(\frac{1}{2}\right)$

$$d i) \quad E[X] = \sum_{x=2}^{12} \phi(x) x$$

$$= 8 \quad \swarrow \text{take from table. } \textcircled{1}$$

$$d ii) \quad E[(x-\bar{x})^2] = \sum_{x=2}^{12} \phi(x) (x-8)^2$$

~~$$\begin{aligned}
 &= 2(0.01)(36) \\
 &+ 0.05(25) \\
 &+ 0.075(16) \\
 &+ 0.075(9) \\
 &+ 0.15(4) \\
 &+ 0.1575(1)
 \end{aligned}$$~~

①

$$= 6$$

Basic approach is list ~~variance~~ $(x-\bar{x})^2$ terms under table of ϕ_x + then do the calculator thing.

(2.5)

e)

i) Use a z-test because the variance of null hypothesis is known $\sigma_{ab}^2 = 5.833$. (1)

$$\begin{aligned} \text{ii) } z &= \frac{\bar{x} - \mu_{ab}}{\sigma_{ab} / \sqrt{n}} = \frac{7.7 - 7}{\sqrt{5.833} / \sqrt{50}} \\ &= ~~1.0137~~ 2.05 \end{aligned} \quad (1)$$

$$\text{iii) } P = ~~0.0207~~ = 0.020 \quad \text{p-value for z-score} \quad (1)$$

iv) The significance level $\alpha = 0.05$.

The alt. hypoth is two sided hence (1)

We compare $P = 0.0207 \leq 0.025 = \frac{1}{2}\alpha$.

The P value is ~~not~~ less than the significance level & hence ~~the test does not~~ support the claim.

Q3

3.1

a i) This is justified by marginalization of the joint probabilities

$$\sum_{c \in \{H, D\}} \phi(T_i | c | R=1) = \phi(T | R=1) \quad \textcircled{1}$$

a ii) Bayes law

$$\phi(T_i | c | R=1) = \frac{\phi(R=1 | T_i, c) \phi(T_i, c)}{\phi(R=1)} \quad \textcircled{1}$$

a iii) Conditional probabilities

$$\phi(T_i, c) = \phi(T | c) \phi(c) \quad \textcircled{1}$$

$$b) \quad \phi(T=H | R=1) \phi(R=1) = \sum_{c \in \{H, D\}} \phi(R=1 | T_i, c) \phi(T | c) \phi(c)$$

$$= (0.05)(0.4)(0.6) + (0.2)(0.7)(0.4)$$

$$= 0.068 \quad \textcircled{1/2}$$

3.2

$$\begin{aligned}
 \text{b i)} \quad \phi(T=S | R=1) \phi(R=1) &= \sum_{C=\{W, S\}} \phi(R=1 | T=C) \phi(T=C) \phi(C) \\
 &= (0.2)(0.6)(0.6) \\
 &\quad + (0.05)(0.3)(0.4) \\
 &= 0.078 \quad \text{①}
 \end{aligned}$$

$$\text{b ii)} \quad P_r(T=H | R=1) = \frac{0.068}{0.068 + 0.078} = 0.466 \quad \text{①}$$

b iii) I would recommend $(T=W)$ wet weather tires. ①

$$\begin{aligned}
 \text{c)} \quad \phi(T=H | R=\{1, 2, 3\}) &= \sum_{C=\{W, D\}} \frac{\phi(R=\{1, 2, 3\} | T=C) \phi(T=C) \phi(C)}{\phi(R=\{1, 2, 3\})} \\
 &= (0.05 + 0.05 + 0.1)(0.4)(0.6) \\
 &\quad + (0.2 + 0.2 + 0.2)(0.7)(0.4) \quad \cancel{\phi(R=\{1, 2, 3\})} \\
 &= 0.216 \quad \cancel{\phi(R=\{1, 2, 3\})} \quad \text{①}
 \end{aligned}$$

$$P(T=S | R=\{1,2,3\}) = \frac{1}{P(R=\{1,2,3\})} \sum_{c \in \{0,1\}} P(R=\{1,2,3\} | T=c) P(T=c) \quad (303)$$

$$= \frac{1}{P(R=\{1,2,3\})} \left((0.2+0.2+0.2)(0.6)(0.6) + (0.05+0.05+0.1)(0.3)(0.4) \right)$$

$$= \frac{1}{P(R=\{1,2,3\})} (0.24) \quad \left(\frac{1}{2}\right)$$

$$P(T=H | R=\{1,2,3\}) = \frac{0.216}{0.216 + 0.24} = 0.47 \quad \left(\frac{1}{2}\right)$$

For $R=\{1,2,3\}$ choose Hard trees $\left(\frac{1}{2}\right)$
 $T=H$

G4)

a) i)

$$\mu_p = \int_0^{\infty} x \lambda e^{-\lambda x} dx \quad (1)$$

ii)

$$\frac{d}{dx} (x^2 e^{-\lambda x}) = 2x e^{-\lambda x} - \lambda x^2 e^{-\lambda x}$$

$$\left(\frac{1}{2}\right) \int_0^{\infty} \frac{d}{dx} (x e^{-\lambda x}) = \int_0^{\infty} e^{-\lambda x} - \int_0^{\infty} \lambda x e^{-\lambda x} dx$$

$$\Rightarrow \left[x e^{-\lambda x} \right]_{x=0}^{x=\infty} = \left[-\frac{e^{-\lambda x}}{\lambda} \right]_0^{\infty} - \mu_p$$

$$\Rightarrow 0 - 0 = 0 + \frac{1}{\lambda} - \mu_p$$

$$\Rightarrow \boxed{\mu_p = \frac{1}{\lambda}}$$

$$b) \quad \bar{\Phi}_\lambda(x) = \int_0^x \lambda e^{-\lambda z} dz \quad \left(\frac{1}{2}\right)$$

$$= \left[-e^{-\lambda z} \right]_0^x$$

$$= -e^{-\lambda x} + 1$$

$$= 1 - e^{-\lambda x} \quad \left(\frac{1}{2}\right) \text{ QED.}$$

c) i) Sampling dist

$$\phi_{\hat{\mu}}(x) = \phi_\lambda(x) = \lambda e^{-\lambda x} \quad \left(\frac{1}{2}\right)$$

ii) ~~Assume~~

A single sample drawn from dist. ϕ_λ

has distribution

$$\hat{\mu} = x_1 \sim \phi_\lambda$$

$\left(\frac{1}{2}\right)$

$$d) \quad x_i = 10$$

50% conf. lower bound $\Rightarrow P = 0.25$

upper bound $P = 0.75$

$$\hat{\mu} = 10$$

$$\hat{\phi} = \phi_{1/\hat{\mu}} = \phi_{1/10}$$

$$\left(\frac{1}{2}\right)$$

lower bound x_L

$$F(x_L) = 0.25$$

$$\Rightarrow 1 - e^{-x_L/10} = 0.25$$

$$-10 \ln(0.75) = x_L$$

$$x_L = 2.88$$

$$\left(\frac{1}{2}\right)$$

for the idea.

upper bound:

$$F(x_u) = 0.75$$

$$1 - e^{-x_u/10} = 0.75$$

$$x_u = -10 \ln(0.25)$$

$$= 13.86$$

$$\text{Conf Int}_{50\%} = [2.88, 13.86]$$

QS

5.1

a i)

$$\phi(x_k=i | y_k=j) = \frac{\phi(y_k=j | x_k=i) \phi(x_k=i)}{\phi(y_k=j)}$$

$\frac{1}{2}$

$$= \frac{\phi(y_k=j | x_k=i) \cdot \frac{1}{2}}{\sum_s \phi(y_k=j | x_k=s) \underbrace{\phi(x_k=s)}_{\frac{1}{2}}}$$

$$= \frac{\phi(y_k=j | x_k=i)}{\sum_s \phi(y_k=j | x_k=s)}$$

$$\sum_s \phi(y_k=0 | x_k=s) = 0.6 + 0.4 = 1$$

$\frac{1}{2}$

$$\sum_s \phi(y_k=1 | x_k=s) = 0.4 + 0.6 = 1$$

$$\leadsto \boxed{\phi(x_k=i | y_k=j) = \phi(y_k=j | x_k=i)}$$

$$\text{ii) } \phi(x_k=0 | y_k=0) = 0.6 \quad \left(\frac{1}{2}\right)$$

$$\phi(x_k=1 | y_k=0) = 0.4 \quad \left(\frac{1}{2}\right)$$

$$\phi(x_k=0 | y_k=1) = 0.4 \quad \left(\frac{1}{2}\right)$$

$$\phi(x_k=1 | y_k=1) = 0.6 \quad \left(\frac{1}{2}\right)$$

$$a-iii) \quad \Pr(y_k \neq x_k) = \frac{1}{2} \left(\Pr(y_k=0 | x_k=1) + \Pr(y_k=1 | x_k=0) \right)$$

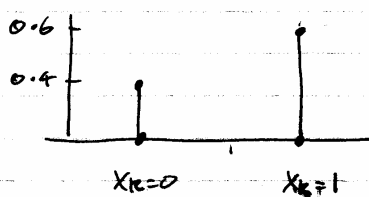
$$= \frac{1}{2} \left(\Pr(x_k=1) + \Pr(x_k=0) \right)$$

$$= \frac{1}{2} (0.4 + 0.4)$$

$$= 0.4$$

b)

i)



$$\text{or } \Pr(X_k=0) = 0.4$$

$$\Pr(X_k=1) = 0.6$$

$$ii) \quad E[X_k] = (0.4) \cdot 0 + (0.6) \cdot 1 = 0.6$$

$$iii) \quad E[T] = 1000 E[X_k] = 600$$

Each instance X_k is an indep event.

5-3

c) i) Prob of transmitting a false character

$$= \Pr(x_1=0, x_2=1, x_3=1) + \Pr(x_1=1, x_2=0, x_3=1) + \Pr(x_1=1, x_2=1, x_3=0) \\ + \Pr(x_1=0, x_2=0, x_3=0)$$

$$= 0.6 \times (0.4)^2 + 0.6 \times 0.4^2 + 0.6 \times (0.4)^2 \\ + (0.4)^3$$

$$= 0.496$$

(1)

Call this event 'F' $\rightarrow \Pr(F) = 0.352$

Prob of trans a valid character is (event V)

$$\Pr(V) = \cancel{0.504} 0.504$$

cii) Answer

for formulae

$$\Pr(2 \text{ correct bits}) = \Pr(x_1=1, x_2=1 | V)$$

(1/2)

$$= \frac{\Pr(x_1=1, x_2=1, V)}{\Pr(V)}$$

$$= \frac{\Pr(x_1=1, x_2=1, x_3=1)}{\Pr(V)} \quad \leftarrow \text{by independence}$$

$$= \frac{(0.6)^3}{0.504} = \cancel{0.429} 0.429 \quad \left(\frac{1}{2} \right)$$

$$\begin{aligned}
 \Pr(1 \text{ bit} | V) &= \Pr(x_1=1, x_2=0 | V) + \Pr(x_1=0, x_2=1 | V) \quad \textcircled{5.4} \\
 &= \frac{\Pr(x_1=1, x_2=0, x_3=0) + \Pr(x_1=0, x_2=1, x_3=0)}{\Pr(V)} \\
 &= \frac{(0.4)^2(0.6) + (0.4)^2(0.6)}{0.6} \\
 &= 0.381 \quad \textcircled{\frac{1}{2}}
 \end{aligned}$$

$$\begin{aligned}
 \Pr(0 \text{ bits} | V) &= \Pr(x_1=0, x_2=0 | V) \\
 &= \frac{\Pr(x_1=0, x_2=0, x_3=1)}{\Pr(V)} \\
 &= \frac{(0.4)^2(0.6)}{0.6} \quad \textcircled{\frac{1}{2}} \\
 &= 0.190
 \end{aligned}$$

$$\begin{aligned}
 \text{c) iii) } E[X] &= \left(2\Pr(2 \text{ correct bits}) + 1\Pr(1 \text{ correct}) + 0\Pr(0 \text{ correct}) \right) \frac{1}{2} \\
 &= \left(2(0.429) + (0.381) + 0(0.190) \right) \frac{1}{2} \\
 &= 1.024 \frac{1}{2} = 0.62 \quad \textcircled{\frac{1}{2}}
 \end{aligned}$$

$$\text{d) i) } T' \leq E[X] \cdot \frac{2}{3} \cdot 1000 \leq 413 \quad \textcircled{\frac{1}{2}} \quad \boxed{T' < T}$$

ii) Although less errors are made the overhead in transmission with reduce the data rate. In fact T' is smaller than $E[X] \cdot \frac{2}{3} \cdot 1000$ since some symbols are retransmitted! $\textcircled{\frac{1}{2}}$