

**1.** Let  $X$  and  $Y$  be integer-valued random variables with joint pmf  $p_{X,Y}(k, m)$ , and let  $Z = X + Y$ .

- (a) Show that for every integer  $n$  we have

$$\mathbb{P}(\{Z \leq n\}) = \sum_{k=-\infty}^{\infty} \sum_{m=-\infty}^{n-k} p_{X,Y}(k, m)$$

and therefore that

$$p_Z(n) = \sum_{k=-\infty}^{\infty} p_{X,Y}(k, n-k).$$

- (b) Conclude that when  $X$  and  $Y$  are independent we have

$$p_Z(n) = \sum_{k=-\infty}^{\infty} p_X(k)p_Y(n-k).$$

**2.** Recall that the correlation coefficient of two random variables  $X$  and  $Y$  defined on the same probability space is

$$\rho = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y},$$

where  $\sigma_X$  and  $\sigma_Y$  are the respective standard deviations of  $X$  and  $Y$ . In class I asserted that  $|\rho| \leq 1$ , with equality if and only if  $X = cY$  for some nonzero constant  $c$ . Let's prove it.

- (a) For any real number  $c$ , we know that

$$\text{Var}(X - cY) \geq 0$$

with equality if and only if  $X = cY$ . Show that this is the same as

$$\text{Var}(X) + c^2 \text{Var}(Y) - 2c\text{Cov}(X, Y) \geq 0$$

with equality if and only if  $X = cY$ .

- (b) The result of part (a) holds in particular for

$$c^* = \frac{\text{Cov}(X, Y)}{\text{Var}(Y)}.$$

From this fact conclude that  $|\rho| \leq 1$  with equality if and only if  $X = c^*Y$ .

- (c) Show from scratch that if  $X = cY$ , then necessarily  $c = c^*$ , with  $c^*$  defined as in (b).

**3.**  $P$  is a random variable distributed uniformly on  $[0, 1]$ . You have a coin whose probability of coming up heads is  $P$ . You toss the coin  $n$  times and set  $X_k = 1$  if the  $k$ th toss comes up heads and  $X_k = 0$  if the  $k$ th toss comes up tails, where we assume that the tosses — and hence the  $X_k$  — are conditionally independent given the value of  $P$ . Let  $Y = \sum_{k=1}^n X_k$ .

- (a) Find  $\mathbb{E}(Y)$  and  $\text{Var}(Y)$ .
- (b) Find  $\text{Cov}(P, Y)$ .
- (c) Find the linear minimum mean-square estimator (LMMSE) of  $P$  given  $Y$ . That is, find constants  $a$  and  $b$  that minimize

$$\mathbb{E}((P - b - aY)^2).$$

**4.** Let  $X$  be uniform on  $[0, 1]$  and  $N$  be geometric with parameter  $p \in (0, 1)$ . Assume that  $X$  and  $N$  are independent.

- (a) Find  $\mathbb{E}(X^N | N = n)$ .
- (b) Find  $\mathbb{E}(X^N)$ . You might want to use the identity

$$\sum_{k=1}^{\infty} \frac{a^k}{k} = -\ln(1-a) \text{ for all } a \in (0, 1).$$

**5.** Consider a random sinusoidal signal of the form

$$X(t) = A \cos(\omega_o t + \Theta),$$

where  $\omega_o$  is fixed and known, while  $A$  and  $\Theta$  are independent random variables with  $\Theta$  a continuous random variable uniformly distributed on  $[0, 2\pi]$ . Find the correlation coefficient  $\rho$  between  $X(t_0)$  and  $X(t_1)$ , where  $t_0$  and  $t_1$  are specified sampling times. Note that the random variable  $A$  can be continuous or discrete and that the final answer turns out not to depend on how  $A$  is distributed.

**6.** Cornell has  $N$  living alumni, where  $N$  is a discrete random variable with pmf

$$p_N(n) = p^{n-1}(1-p) \text{ for all } n > 0,$$

where  $p \in (0, 1)$  is given. Every year Cornell holds a fundraiser that each living alum attends with probability  $q \in (0, 1)$  independent of all other alumni. An alum attending the fundraiser donates an amount of money  $X$  that has exponential pdf

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & \text{when } x \geq 0 \\ 0 & \text{otherwise.} \end{cases}$$

The different alums' donations are independent of each other, of the random variable  $N$ , and of whether each alum attends the fundraiser or not.

- (a) Find the expected value and variance of the number of alumni attending the fundraiser.
- (b) Find the expected value and variance of the total amount of money raised at the fundraiser.

Rami Bellumbi

ECE3100 HW11

NOT TURNED IN

①  $X, Y$  integer valued random variables with joint pmf  $P_{X,Y}(k,m)$  and let

$$Z = X + Y$$

(a) Show  $\forall n \in \mathbb{Z}$  we have

$$P(Z \leq n) = \sum_{k=-\infty}^{+\infty} \sum_{m=-\infty}^{n-k} P_{X,Y}(k,m)$$

and therefore that

$$P_Z(n) = \sum_{k=-\infty}^{+\infty} P_{X,Y}(k, n-k)$$

$P(Z \leq n)$  = sum of  $P_{X,Y}(k,m)$  over all pairs  $(k,m)$  such that  $k+m \leq n$ . This is shown in double sum. For  $k$  anything,  $m \leq n-k$ .

$$\begin{aligned} P_Z(n) &= P(Z \leq n) - P(Z \leq n-1) \\ &= \sum_{k=-\infty}^{+\infty} \sum_{m=-\infty}^{n-k} P_{X,Y}(k,m) - \sum_{k=-\infty}^{+\infty} \sum_{m=-\infty}^{n-1-k} P_{X,Y}(k,n) \\ &= \sum_{k=-\infty}^{+\infty} \left[ \sum_{m=-\infty}^{n-k} P_{X,Y}(k,m) - \sum_{m=-\infty}^{n-1-k} P_{X,Y}(k,n) \right] \\ &= \sum_{k=-\infty}^{+\infty} \left[ P_{X,Y}(k, n-k) + \sum_{m=-\infty}^{n-k-1} (P_{X,Y}(k,m) - P_{X,Y}(k,n)) \right] \\ &= \sum_{k=-\infty}^{+\infty} P_{X,Y}(k, n-k) \end{aligned}$$

(b)  $X, Y$  independent  $\iff P_{X,Y}(k, n-k) = p_X(k)p_Y(n-k)$

thus

$$\sum_{k=-\infty}^{+\infty} p_{X,Y}(k, n-k) = \sum_{k=-\infty}^{+\infty} p_X(k)p_Y(n-k)$$

②

$$P = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} \quad \left. \begin{array}{l} \text{Correlation} \\ \text{coefficient} \end{array} \right\}$$

(a)  $\forall c \in \mathbb{R}$ , know

$$\text{Var}(X - cY) \geq 0$$

with equality iff  $X = cY$ .

Show this is the same as

$$\text{Var}(x) + c^2 \text{Var}(Y) - 2c \text{Cov}(X, Y) \geq 0$$

with equality iff  $X = cY$

$$\text{Var}(X - cY) = \mathbb{E}[(X - cY)^2] - (\mathbb{E}[X - cY])^2 \geq 0$$

$$= \mathbb{E}[X^2] - \mathbb{E}[2XcY] + \mathbb{E}[c^2Y^2] - (\mathbb{E}[X] - \mathbb{E}[cY])^2$$

$$= \mathbb{E}(X^2) - (\mathbb{E}(X))^2 + \mathbb{E}(c^2Y^2) - (\mathbb{E}(cY))^2 - [2c(\mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y))]$$

$$= \text{Var}(X) + c^2 \text{Var}(Y) - 2c \text{Cov}(X, Y) \geq 0$$

if  $X = cY$

$$c^2 \text{Var}(Y) + c^2 \text{Var}(Y) - 2c^2 \text{Var}(Y) = 0$$

(b)

$$c^* = \frac{\text{Cov}(X, Y)}{\text{Var}(Y)}$$

Conclude  $|p| \leq 1$  with equality iff  $X = c^* Y$

$$\text{Var}(X) + (c^*)^2 \text{Var}(Y) - 2c^* \text{Cov}(X, Y) \geq 0$$

$$\text{Var}(X) - \frac{(\text{Cov}(X, Y))^2}{\text{Var}(Y)} \geq 0$$

$$1 \geq (p(X, Y))^2$$

$$|p(X, Y)| \leq 1$$

(c) Let  $X, Y$  be two random variables such that  $X = cY$ ;  $c \in \mathbb{R}$ .

Thus

$$X - \mathbb{E}(X) = c(Y - \mathbb{E}(Y))$$

and

$$(X - \mathbb{E}(X))(Y - \mathbb{E}(Y)) = c(Y - \mathbb{E}(Y))^2$$

$$\mathbb{E}[(X - \mathbb{E}(X))(Y - \mathbb{E}(Y))] = \mathbb{E}[c(Y - \mathbb{E}(Y))^2]$$

$$\text{Cov}(X, Y) = c \text{Var}(Y)$$

$$c = \frac{\text{Cov}(X, Y)}{\text{Var}(Y)}$$

### ③ $P \sim \text{Uniform}[0,1]$ .

Coin whose probability of being heads is  $P$ .

Toss coin  $n$  times; set  $X_k = 1$  if  $k^{\text{th}}$  toss comes up heads  
else 0 -  $1 \leq k \leq n$ .

- Thus  $X_k$  conditionally independent given value  $P$ .

Let

$$Y = \sum_{k=1}^n X_k$$

(a)  $\text{IE}(Y)$ ,  $\text{Var}(Y)$ ?

Given  $P = p$ ,  $Y$  is the sum of  $n$  Bernoulli  $p$  random variables.

So

$$\text{IE}(Y|P=p) = np \rightarrow \text{IE}(Y|P) = nP$$

Thus

$$\begin{aligned} \text{IE}(Y) &= \int_{-\infty}^{+\infty} \text{IE}(Y|P=p) f_p(p) dp \\ &= \int_0^1 np dp = \left. \frac{1}{2} n p^2 \right|_0^1 = \frac{n}{2} \end{aligned}$$

Similarly,

$$\text{Var}(Y|P=p) = np(1-p) \rightarrow \text{Var}(Y|P) = nP(1-P)$$

$$\begin{aligned} \text{Var}(Y) &= \text{IE}(\text{Var}(Y|P)) + \text{Var}(\text{IE}(Y|P)) \\ &= \text{IE}(nP(1-P)) + \text{Var}(nP) \\ &= \frac{n}{6} + \frac{n^2}{12} = \frac{n(n+2)}{12} \end{aligned}$$

b)

$$\begin{aligned}
 \text{Cov}(P, Y) &= \mathbb{E}[(P - \mathbb{E}(P))(Y - \mathbb{E}(Y))] \\
 &= \mathbb{E}\left[(P - \frac{1}{2})(Y - \frac{n}{2})\right] \\
 &= \mathbb{E}\left[\mathbb{E}\left[(P - \frac{1}{2})(Y - \frac{n}{2})\right] \mid P = p\right] \\
 &= \mathbb{E}\left[\left(p - \frac{1}{2}\right)\mathbb{E}\left[Y \mid P = p\right] - \frac{n}{2}\right] \\
 &= \mathbb{E}\left[\left(p - \frac{1}{2}\right)\left(\mathbb{E}[Y \mid P = p] - \frac{n}{2}\right)\right] \\
 &= \mathbb{E}\left[\left(p - \frac{1}{2}\right)\left(np - \frac{n^2}{2}\right)\right] \quad p \rightarrow P \\
 \rightarrow \mathbb{E}\left[\left(P - \frac{1}{2}\right)n\left(P - \frac{1}{2}\right)\right] \\
 &= n\mathbb{E}\left[\left(P - \frac{1}{2}\right)^2\right] \\
 &= n\text{Var}(P) \\
 &= n/12
 \end{aligned}$$

(c) Linear Minimum Mean Square Estimator of  $P$  given  $Y$ ?

i.e.;  $a, b$  that minimize

$$\mathbb{E}[(P - b - aY)^2]$$

$$\begin{aligned}
 \hat{a} &= \frac{\text{Cov}(P, Y)}{\text{Var}(Y)}, & \hat{b} &= \mathbb{E}(P) - \hat{a}\mathbb{E}(Y) \\
 &= \frac{n/12}{n(n+2)/12} = \boxed{\frac{1}{n+2}} & &= \frac{1}{2} - \frac{n}{2(n+2)} = \boxed{\frac{1}{n+2}}
 \end{aligned}$$

④  $X \sim \text{Uniform}[0,1]$

$N \sim \text{Geometric}(p)$ ;  $p \in (0,1)$   $f_{X|N}(x|n) = f_X(x) \neq x, n$   
 $X, N$  independent. by independence

$$\begin{aligned}
 (a) \mathbb{E}[X^n | N=n] &= \int_{-\infty}^{+\infty} x^n f_{X|N}(x|n) dx \\
 &= \int_{-\infty}^{+\infty} x^n f_X(x) dx \\
 &= \int_0^1 x^n dx = \frac{1}{n+1}
 \end{aligned}$$

$$(b) \mathbb{E}(X^n) = \sum_{n=-\infty}^{+\infty} \mathbb{E}(X^n | N=n) p_N(n)$$

$$\sum_{n=1}^{\infty} \frac{1}{n+1} p(1-p)^{n-1}$$

$$m = n+1$$

$$= \sum_{m=2}^{\infty} \frac{p(1-p)^{m-2}}{m}$$

$$= \frac{p}{(1-p)^2} \sum_{m=2}^{\infty} \frac{(1-p)^m}{m}$$

$$= -\frac{p}{(1-p)^2} (\ln(p) + (1-p))$$

(5)

$$X(t) = A \cos(\omega_0 t + \Theta)$$

$\omega_0$  fixed;  $\Theta \sim \text{Uniform}[0, 2\pi]$

$A$  a random variable

Given  $t_0, t_1$ , find  $\rho$  between  $X(t_0)$  and  $X(t_1)$ .

$$\rho = \frac{\text{Cov}(X(t_0), X(t_1))}{\sqrt{\text{Var}(X(t_0))} \sqrt{\text{Var}(X(t_1))}}$$

$$\begin{aligned}\text{Var}(X(t_0)) &= \mathbb{E}[(X(t_0))^2] + (\mathbb{E}[X(t_0)])^2 \\ &= \mathbb{E}[A^2 \cos^2(\omega_0 t_0 + \Theta)] \\ &= \mathbb{E}(A^2) \mathbb{E}[\cos^2(\omega_0 t_0 + \Theta)] \\ &= \mathbb{E}(A^2) \left( \frac{1}{2} \int_0^{2\pi} \cos^2(\omega_0 t_0 + \theta) d\theta \right) \\ &= \mathbb{E}(A^2) \left( \frac{1}{2} \right) = \boxed{\frac{1}{2} \mathbb{E}(A^2)}\end{aligned}$$

$\rightarrow 0 \leftarrow \frac{1}{2\pi} \int_0^{2\pi} A \cos(\omega_0 t_0 + \theta) d\theta = 0$

$\leftarrow A, \Theta \text{ independent}$

Thus,

$$\text{Var}(X(t_1)) = \frac{1}{2} \mathbb{E}(A^2) \text{ as well.}$$

$$\begin{aligned}\text{Cov}(X(t_0), X(t_1)) &= \mathbb{E}[X(t_0)X(t_1)] - \mathbb{E}[X(t_0)] \mathbb{E}[X(t_1)] \\ &= \mathbb{E}[A^2 \cos(\omega_0(t_0-t_1) + \Theta)] \\ &= \frac{1}{2} \mathbb{E}(A^2) \cos(\omega_0(t_0-t_1))\end{aligned}$$

$\rightarrow 0$

Thus

$$\rho = \frac{\frac{1}{2} \mathbb{E}(A^2) \cos(\omega_0(t_0-t_1))}{\sqrt{\frac{1}{2} \mathbb{E}(A^2)} \sqrt{\frac{1}{2} \mathbb{E}(A^2)}} = \boxed{\cos(\omega_0(t_0-t_1))}$$

⑥ Cornell has  $N$  living alumni,  $N$  a discrete random variable with pmf

$$P_N(n) = p^{n-1} (1-p) \quad \forall n > 0 ; \quad p \in (0,1) \text{ given}$$

Every year Cornell holds a fundraiser that each living alum attends w/ probability  $q \in (0,1)$  independent of other alumni.

An alum attending the fundraiser donates an amount of money  $X$  that has exponential pdf

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & , x \geq 0 \\ 0 & , \text{ else} \end{cases}$$

Different alum donations independent of each other, of  $N$ , and of whether each alum attends the fundraiser or not.

(a) Let  $Q$  = number of alumni attending fundraiser.

$\mathbb{E}(Q)$ ?  $\text{Var}(Q)$ ?

Note  $\Pr(\{Q=k\} \mid \{N=n\}) = \binom{n}{k} q^k (1-p)^{n-k}$  ← Binomial pmf!

$$\mathbb{E}(Q) = \sum_{n=1}^{\infty} \underbrace{\mathbb{E}(\{Q=k\} \mid \{N=n\})}_{nq} \underbrace{\Pr(\{N=n\})}_{P_N(n)} \quad \text{Law of Total Expectation}$$

$$= q \sum_{n=1}^{\infty} n P_N(n) = q \mathbb{E}[N] = \frac{q}{(1-p)}$$

OR

$$\mathbb{E}(Q) = \mathbb{E}(\mathbb{E}(Q|N)) = \mathbb{E}(qN) = q \mathbb{E}[N] = \frac{q}{1-p} \quad \leftarrow \text{Law of Iterated Expectation}$$

$$\text{Var}(Q|N=n) = nq(1-q)$$

$$\text{Var}(Q|N) = Nq(1-q)$$

$$\begin{aligned}\text{Var}(Q) &= \text{IE}(\text{Var}(Q|N)) + \text{Var}(\text{IE}(Q|N)) \\ &= \text{IE}(Nq(1-q)) + \text{Var}(Nq) \\ &= q(1-q)\text{IE}[N] + q^2 \text{Var}(N) \\ &= \boxed{\frac{q(1-q)}{1-p} + q^2 \frac{p}{(1-p)^2}}\end{aligned}$$

(b) Given  $Q=k$

Let

$$Z = \text{Total amount fundraised} = X_1 + X_2 + \dots + X_k$$

where  $X_m$  is the donation of attendee  $m$  for  $1 \leq m \leq k$ .

$X_m$  are independent exponential random variable w/ rate parameter  $\lambda$ .

So,

$$\text{IE}(Z|Q=k) = \sum_{m=1}^k \text{IE}(X_m|Q=k) \stackrel{X_m \text{ independent of } Q}{=} \sum_{m=1}^k \text{IE}(X_m) = \frac{k}{\lambda}$$

$$\text{Thus } \text{IE}(Z|Q) = \frac{Q}{\lambda}$$

Law of iterated expectation

$$\text{IE}(Z) = \text{IE}(\text{IE}(Z|Q)) = \text{IE}\left(\frac{Q}{\lambda}\right) = \frac{1}{\lambda} \text{IE}(Q) = \frac{q}{(1-p)\lambda}$$

$$\text{Var}(Z|Q=k) = \sum_{m=1}^k \text{Var}(X_m|Q=k) = \sum_{m=1}^k \text{Var}(X_m) = \frac{k}{\lambda^2}$$

Thus

$$\text{Var}(Z|Q) = Q/\lambda^2$$

$$\text{Var}(Z) = \mathbb{E}(\text{Var}(Z|Q)) + \text{Var}(\mathbb{E}(Z|Q))$$

$$= \mathbb{E}(Q/\lambda^2) + \text{Var}(Q/\lambda)$$

$$= \frac{1}{\lambda^2} \mathbb{E}(Q) + \frac{1}{\lambda^2} \text{Var}(Q)$$

$$= \frac{q}{\lambda^2(1-p)} + \frac{1}{\lambda^2} \left[ \frac{q(1-q)}{1-p} + q^2 \frac{p}{(1-p)^2} \right]$$