

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 σ_X and σ_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 ω_o is fixed and known, while A and Θ are independent random variables with Θ a continuous random variable uniformly distributed on $[0, 2\pi]$. Find the correlation coefficient ρ 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.

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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^{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)

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

ω_0 fixed; $\Theta \sim \text{Uniform}[0, 2\pi]$

A a random variable

Given t_0, t_1 , find ρ 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 λ .

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

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

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

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

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

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