

MSO201A

Hints to Solution: Quiz-IV

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1. If X and Y are independent, then for a discrete random variable X taking values in $\{x_1, x_2, \dots, x_n\}$ with positive probability, we have

$$P[X + Y = \alpha] = \sum_{x=x_1}^{x_n} P[X = x]P[Y = \alpha - x] \text{ (due to independence).}$$

So, if Y is a continuous random variable, then $P[X + Y = \alpha] = 0$. If Y is a discrete random variable, then $\sum_{x=x_1}^{x_n} P[X = x]P[Y = \alpha - x] < 1$.

Similarly, it can be argued when X is continuous. Under independence, both X and Y are degenerate random variables.

If independence is removed, then take $Y = \alpha - X$, hence both can be non-degenerate random variables.

If X is degenerate at c and Y is non-degenerate random variable, then $P[X + Y = \alpha] = P[Y = \alpha - c] < 1$.

2. Note that $G(Y) \geq c \iff F(c) \leq Y$, hence $P[G(Y) \geq \log_e 5] = \exp(-\log_e 5) = 1/5$.
3. If $h(\cdot)$ and $\phi(\cdot)$ are non-negative, non-decreasing and continuous, then $\phi(h(\cdot))$ is also non-decreasing and continuous. Hence all functions given are cumulative distribution functions.
- 4.

$$P[\text{unit is faulty} | \text{unit is passed by detector}] = \frac{P[\text{unit is faulty}]P[\text{unit passed} | \text{unit is faulty}]}{P[\text{unit is passed}]}$$
$$\implies P[\text{unit is faulty} | \text{unit is passed by detector}] = \frac{0.05 \times (1 - 0.5)}{(1 - 0.05 \times 0.5)} = \frac{1}{39}$$

Given $Y = 1000$, the number of passed products given they are faulty

$\sim \text{Bin}\left(40000 - 1000, \frac{1}{39}\right)$. So,

$$E(X|Y = 1000) = 1000 + 39000 \times \frac{1}{39} = 2000.$$

5. Consider $F(x) = \begin{cases} 0 & x \leq 1 \\ 1 - \frac{1}{kx} & e^{k-1} < x \leq e^k \text{ for } k = 1, 2, \dots \end{cases}$

Now, $x(1 - F(x))$ converges to 0 as $x \rightarrow \infty$, but $E(X) = \int_0^\infty (1 - F(x))dx = \infty$.

$$P[A \cup B \cup C] = 1 - P[A^c \cap B^c \cap C^c] = 1 - P[A^c | B^c \cap C^c] P[B^c \cap C^c]$$

Let $X \sim U(0, 1)$, and define $Y = \begin{cases} X & , \text{ if } X \text{ is irrational} \\ 0 & , \text{ else} \end{cases}$
then correlation between Y and X is 1.

Let X and Y be independent and have probability density function (pdf) as follows:

$$f(x) = \begin{cases} \frac{1}{x^2} & x \geq 1 \\ 0 & \text{otherwise.} \end{cases}$$

Then, $\min\{X, Y\}$ has the following pdf:

$$g(x) = 2[1 - F(x)]^{2-1} f(x),$$

and hence $\min\{X, Y\}$ has finite expectation.

Consider (X, Y, Z) follow a multivariate normal with following covariance matrix:

$$\Sigma = \begin{bmatrix} 1 & 0.5 & -0.25 \\ 0.5 & 1 & 0.5 \\ -0.25 & 0.5 & 1 \end{bmatrix}.$$

Then, Y is positively correlated with X and Z , but X and Z are negatively correlated.

6. Observe that $X \sim \text{Bernoulli}(\mu)$.

7. A random variable X with mgf $M_X(t) = (1 - \theta t)^{-k}$ with $t < 1/\theta$ is *Gamma*(k, θ). So, $\mathbb{E}(X) = k\theta$ and $\mathbb{E}(X^2) = k\theta^2 + k^2\theta^2$.

8. Let $S_n = X_1 + X_2 + \dots + X_n$, where X_1, X_2, \dots, X_n are i.i.d. *Poisson*(1) random variables. Using CLT, we get $\frac{1}{\sqrt{n}}(S_n - n) \xrightarrow{d} N(0, 1)$ as $n \rightarrow \infty$. Now,

$$e^{-n} \sum_{k=0}^n \frac{n^k}{k!} = P[S_n \leq n] = P\left[\frac{1}{\sqrt{n}}(S_n - n) \leq 0\right] \xrightarrow{n \rightarrow \infty} \Phi(0) = 0.5.$$

9. Observe that $X, -X \sim N(0, 1)$ and

$$\begin{aligned}
 P(Y \leq y) &= P(Y \leq y | W = 1)P(W = 1) + P(Y \leq y | W = -1)P(W = -1) \\
 &= 0.5P(X \leq y | W = 1) + 0.5P(-X \leq y | W = -1) \\
 &= 0.5P(X \leq y) + 0.5P(-X \leq y) \text{ as } X \perp\!\!\!\perp W \\
 &= 0.5P(X \leq y) + 0.5P(X \leq y) \\
 &= P(X \leq y).
 \end{aligned}$$

So, $Y \sim N(0, 1)$, $\mathbb{E}(Y) = 0$ and $Var(Y) = 1$.

Further, note that

$$Z = \begin{cases} 2X, & \text{with probability } 0.5 \\ 0, & \text{with probability } 0.5. \end{cases}$$

So, we get

$$\begin{aligned}
 P(Z \leq 0) &= 0.5P(2X \leq 0) + 0.5\mathbb{I}(0 \in (-\infty, 0]) \\
 &= 0.25 + 0.5 = 0.75,
 \end{aligned}$$

$$\mathbb{E}(Z) = 0 \text{ and } Var(Z) = 0.5Var(2X) = 2.$$

10. Easy.

11. Observe the following facts:

- If $X \sim Poisson(\lambda_1)$, $Y \sim Poisson(\lambda_2)$ and $X \perp\!\!\!\perp Y$, then $X + Y \sim Poisson(\lambda_1 + \lambda_2)$.
- If $X \sim Poisson(\lambda_1)$, $Y \sim Poisson(\lambda_2)$, then $X - Y \not\sim Poisson$, because $X - Y$ may be negative.
- If $X \sim Poisson(\lambda_1)$, $Y \sim Poisson(\lambda_2)$ and $X \perp\!\!\!\perp Y$, then $X | X+Y = t \sim Binomial(t, \frac{\lambda_1}{\lambda_1 + \lambda_2})$.
- If $X \sim Poisson(\lambda_1)$, then $2X \not\sim Poisson$ as $2X$ can not be an odd integer.

12. Observe following facts:

- If $\rho = 0$:
 $X \perp\!\!\!\perp Y$.
 $\frac{X}{Y} \sim t_1 \equiv Cauchy$.
 $X + Y \perp\!\!\!\perp X - Y$ and $X + Y, X - Y \sim N(0, 2)$.
 $X^2, Y^2 \sim \chi^2$ & $X^2 \perp\!\!\!\perp Y^2 \Rightarrow X^2 + Y^2 \perp\!\!\!\perp \frac{X^2}{X^2 + Y^2}$.
- If $\rho \neq 0$:
 $Cov(X, Y - \rho X) = Cov(X, Y) - \rho Cov(X, X) = \rho - \rho = 0$.
 $Cov(X, Y - 2\rho X) = Cov(X, Y) - 2\rho Cov(X, X) = \rho - 2\rho = -\rho$.