Name	Jerry
Student Number	

STA 302s13 Quiz 10 A

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$$\begin{split} E(g(X)) &= \int_{-\infty}^{\infty} g(x) f_X(x) dx, \quad \text{or } E(g(X)) = \sum_x g(x) p_X(x) \\ Var(X) &= E[(X - \mu_X)^2] \\ & \text{Cov}(X, Y) = E[(X - \mu_X)(Y - \mu_Y)] \\ & \text{with } E(X) = M_X \quad \text{and} \quad E(Y) = M_Y \end{split}$$

1. (5 points) In this question, X and Y are random variables, while a and b are constants. Circle the letter corresponding to the correct statement, and prove it. You may well-known properties of expected value without proof.

(a)
$$Cov(X + a, Y + b) = Var(X) + Var(Y) + 2abCov(X, Y)$$

(b) $Cov(X + a, Y + b) = Cov(X, Y)$
(c) $Cov(X + a, Y + b) = 0$
 $E(X + a) = M_{x} + a$ $\sum E(Y + b) = M_{y} + b$
 $Cov(X + a, Y + b) = E\left\{(X + a - (\mu_{x} + a))(Y + b - (\mu_{y} + b))\right\}$
 $= E\left\{(X + a - M_{x} - A)(Y + b - M_{y} - b)\right\}$
 $= E\left\{(X + a - M_{x} - A)(Y + b - M_{y} - b)\right\}$
 $= E\left\{(X - M_{x})(Y - M_{y})\right\}$
 $= C_{OV}(X, Y)$

2. (5 points) Let Y_1, \ldots, Y_n be a random sample from a normal distribution with mean μ and variance σ^2 , so that $T = \frac{\sqrt{n}(\overline{Y}-\mu)}{S} \sim t(n-1)$. This is something you may use without proof. Derive a $(1-\alpha)100\%$ confidence interval for μ . "Derive" means show all the high school algebra. Use the symbol $t_{\alpha/2}$ for the number satisfying $Pr(T > t_{\alpha/2}) = \alpha/2$.

1-2 = P(-ta/2 T < ta/2) = P(-tarz < Un (P-M) < tarz) = P(-taz = < F- M < taz =) = P(- 7-ta/2 = <- u<- F+ta/2 =) $= P(Y + t_{a_1} = M > Y - t_{a_2} = T_{a_2})$ = P(Y-to, E So the confidence interval is (Y-ta/2 In, Y+ ta/2 In), 07 y + tan in

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$$\begin{split} E(g(X)) &= \int_{-\infty}^{\infty} g(x) \, f_X(x) \, dx, \quad \text{ or } E(g(X)) = \sum_x g(x) \, p_X(x) \\ Var(X) &= E[(X - \mu_X)^2] \qquad \qquad Cov(X,Y) = E[(X - \mu_X)(Y - \mu_Y)] \end{split}$$

1. (5 points) Let X and Y be random variables with $E(X) = \mu_x$ and $E(Y) = \mu_y$. Calculate Var(X + Y) in terms of variances and covariances. Show your work. You may use well-known properties of expected value without proof.

$$E(X+Y) = M_{x} + M_{y}, so$$

$$Van(X+Y) = E\left\{\left(X+Y-(\mu_{x}+\mu_{y})\right)^{2}\right\}$$

$$= E\left\{\left(X-\mu_{x} + Y-\mu_{y}\right)^{2}\right\}$$

$$= E\left\{\left(X-\mu_{x}\right)^{2} + 2\left(X-\mu_{x}\right)\left(Y-\mu_{y}\right) + \left(Y-\mu_{y}\right)^{2}\right\}$$

$$= E\left\{\left(X-\mu_{x}\right)^{2} + 2E\left\{(X-\mu_{x})\left(Y-\mu_{y}\right)\right\} + E\left\{\left(Y-\mu_{y}\right)^{2}\right\}$$

$$= E\left\{\left(X-\mu_{x}\right)^{2} + 2E\left\{(X-\mu_{x})\left(Y-\mu_{y}\right)\right\} + E\left\{\left(Y-\mu_{y}\right)^{2}\right\}$$

$$= Van(X) + 2Cov(X,Y) + Van(Y)$$

2. (5 points) Recall that an inverse of the square matrix \mathbf{A} (denoted \mathbf{A}^{-1}) is defined by two properties: $\mathbf{A}^{-1}\mathbf{A} = \mathbf{I}$ and $\mathbf{A}\mathbf{A}^{-1} = \mathbf{I}$. Prove that inverses are unique, as follows. Let \mathbf{B} and \mathbf{C} both be inverses of \mathbf{A} . Show that $\mathbf{B} = \mathbf{C}$. This is fast if you do it the matrix \mathbf{A} and \mathbf{A} .



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