

MTH211A: Theory of Statistics

Problem set 1

1. Let the random variable X have pdf

$$f(x) = \sqrt{\frac{2}{\pi}} \exp\{-x^2/2\}, \quad x > 0.$$

- (a) Find $E(X)$ and $\text{var}(X)$.
(b) Find an appropriate transformation $Y = g(X)$ and α, β , so that $Y \sim \text{Gamma}(\alpha, \beta)$.
2. Let X is distributed as $\text{Gamma}(\alpha, \beta)$ distribution, $\alpha, \beta > 0$. Then show that the r -th order population moment

$$E(X^r) = \beta^{-r} \frac{\Gamma(r + \alpha)}{\Gamma(\alpha)}, \quad r > -\alpha.$$

3. Let the bivariate random variable (X, Y) has a joint pdf

$$f_{X,Y}(x, y) = \begin{cases} C(x + 2y) & \text{if } 0 < y < 1, 0 < x < 2, \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Find the marginal distribution of Y .
(b) Find the conditional distribution of Y given $X = 1$.
(c) Compare the expectations of the above two distributions of Y .
(d) Find the covariance between X and Y .
(e) Find the distribution of $Z = 9/(2Y + 1)^2$.
(f) What is $P(X > Y)$?
4. Let $X \sim \text{normal}(0, 1)$. Define $Y = -X\mathbb{I}(|X| \leq 1) + X\mathbb{I}(|X| > 1)$.
- (a) Find the distribution of Y (Hint: Start by finding the CDF of Y).
(b) Prove or disprove: The distribution of (X, Y) is bivariate normal.
(Hint: Argue that the distribution of $X + Y$ is not continuous.)
5. Let $X \sim \text{normal}(0, 1)$. Define $Y = \text{sign}(X)$ and $Z = |X|$. Here $\text{sign}(\cdot)$ is a $\mathbb{R} \rightarrow \{0, 1\}$ function such that $\text{sign}(a) = 1$ if $a \geq 0$, and $\text{sign}(a) = -1$ otherwise.
- (a) Find the marginal distributions of Y and Z .
(b) Find the joint CDF of (Y, Z) . Hence or otherwise prove that Y and Z are independently distributed.
6. Suppose the distribution of Y , conditional on $X = x_0$ is $\text{normal}(x_0, x_0^2)$, and the marginal distribution of X is $\text{uniform}(0, 1)$. Show that $Z = Y/X$ and X are independently distributed. Find the distribution of Y/X .

7. (a) Let (X, Y) is jointly distributed as $N_2(\mu_x, \mu_y, \sigma_x^2, \sigma_y^2, \rho)$. Suppose (X, Y) are uncorrelated in the sense that $\text{cov}(X, Y) = E[\{X - E(X)\}\{Y - E(Y)\}] = 0$. Then show that X and Y are independently distributed.
- (b) Let $\mathbf{X} = (X_1, \dots, X_k)^\top$ is distributed as a k -variate normal distribution with parameters $\boldsymbol{\mu}$ and $\Sigma = \text{Diag}(\sigma_1^2, \dots, \sigma_k^2)$, $\sigma_j > 0$ for all $j = 1, \dots, k$. Show that X_1, \dots, X_k are mutually independent.
8. Let $X_i \stackrel{\text{iid}}{\sim} \text{normal}(\mu, \sigma^2)$. Define $U = (a_1 X_1 + \dots + a_n X_n)$ and $V = (b_1 X_1 + \dots + b_n X_n)$.
- (a) Show that (U, V) jointly follow a bi-variate normal distribution. Identify the parameters of the distribution. (Hint: For any $\mathbf{c} \in \mathbb{R}^2$, show that $\mathbf{c}^\top \begin{bmatrix} U \\ V \end{bmatrix}$ has an univariate normal distribution.)
- (b) Find conditions on \mathbf{a} and \mathbf{b} such that U and V are independently distributed. (Hint: Apply problem 7.)
9. Suppose $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{normal}(\mu_x, \sigma^2)$, $Y_1, \dots, Y_m \stackrel{\text{iid}}{\sim} \text{normal}(\mu_y, \sigma^2)$, and all the random variables $\{X_1, \dots, X_n, Y_1, \dots, Y_m\}$ are mutually independent. Then find the distribution of $T := S_X^{*2}/S_Y^{*2}$, where S_X^{*2} and S_Y^{*2} are the unbiased sample variances of X and Y , respectively.
10. Let X_1, \dots, X_n be iid random variables with continuous CDF F_X , and suppose $E(X_1) = \mu$. Define the random variables Y_1, \dots, Y_n as follows:
- $$Y_i = \begin{cases} 1 & \text{if } X_i > \mu \\ 0 & \text{otherwise.} \end{cases}$$
- (a) Find $E(Y_1)$.
- (b) Find the distribution of $\sum_{i=1}^n Y_i$.
11. Let $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{normal}(\mu, \sigma^2)$, and S_n^2 be the sample variance. Find a function of S_n^2 , say $g(S_n^2)$, which satisfies $E[g(S_n^2)] = \sigma$. (Hint: You may use problem 2.)
12. Let $\mathbf{X}_1, \dots, \mathbf{X}_n$ be a random sample from a k -variate distribution F , where $\mathbf{X}_i = (Y_{1,i}, \dots, Y_{k,i})^\top$. Let the expectation $E(\mathbf{X}_1) = \boldsymbol{\mu}$ and the variance covariance matrix of F , $\Sigma = E\{(\mathbf{X}_1 - \boldsymbol{\mu})(\mathbf{X}_1 - \boldsymbol{\mu})^\top\}$, have all finite components, and $S_n = n^{-1} \sum_{i=1}^n (\mathbf{X}_i - \bar{\mathbf{X}}_n)(\mathbf{X}_i - \bar{\mathbf{X}}_n)^\top$ be the sample variance-covariance matrix, where $\bar{\mathbf{X}}_n = n^{-1} \sum_{i=1}^n \mathbf{X}_i$. Show that
- (a) $\bar{\mathbf{X}}_n = (\bar{Y}_1, \dots, \bar{Y}_k)^\top$
- (b) For all $j, l = 1, \dots, k$, the (j, l) -th component of Σ is $\sigma_{j,l} = E[\{Y_j - E(Y_j)\}\{Y_l - E(Y_l)\}]$. Also, for all $j, l = 1, \dots, k$, the (j, l) -th component of S_n is $S_{j,l} = n^{-1} \sum_{i=1}^n (Y_{j,i} - \bar{Y}_j)(Y_{l,i} - \bar{Y}_l)$.
- (c) $E(\bar{\mathbf{X}}_n) = \boldsymbol{\mu}$ and $\text{var}(\bar{\mathbf{X}}_n) = E\{(\bar{\mathbf{X}}_n - \boldsymbol{\mu})(\bar{\mathbf{X}}_n - \boldsymbol{\mu})^\top\} = n^{-1}\Sigma$.
- (d) $E(nS_n) = (n-1)\Sigma$.
13. Let X_1, \dots, X_n be iid with pdf f_X and CDF F_X , and let $r < s$. Then find the conditional distribution of $X_{(r)}$ given $X_{(s)} = y$ in terms of f_X and F_X . In particular, if X_1, \dots, X_n is a random sample from **uniform**(0, 1) distribution, then can you identify the conditional distribution of $X_{(r)}$ given $X_{(s)} = y$?
14. Let X_1, \dots, X_n be iid with pdf f_X and CDF F_X . Find the CDF of r -th order statistics $X_{(r)}$. Hence derive the pdf of $X_{(r)}$.
15. Let Y have a **Cauchy**(0, 1) distribution.
- (a) Find the CDF of Y .
- (b) Hence provide a method of simulating random samples from **Cauchy**(0, 1) distribution, starting from **uniform**(0, 1) random variables.