

Statistics 164 Homework 6

- If X has characteristic function φ , find the characteristic function of $Y = aX + b$.
- Show that if X and $-X$ have the same distributions then the characteristic function of X is real-valued.
- Find the characteristic functions of the following random variables.
 - $X \sim \text{Poisson}(\lambda)$.
 - $X \sim U[0, 1]$
 - $X \sim DE(1)$
- Use the characteristic function of the double exponential and the Inversion Theorem to find the characteristic function of the standard Cauchy distribution. Now let X_1, \dots, X_n be i.i.d. Cauchy random variables, and define $Z = n^{-1}(X_1 + \dots + X_n)$. Show that Z has the same distribution as X_1 . What does this tell you about the law of large numbers?
- (B & D) If $X \sim N(\theta, 1)$, the random variable $Y = X^2$ is said to have a noncentral χ^2 distribution with 1 degree of freedom and noncentrality parameter θ^2 .
 - Show that Y has density $f_Y(y) = \frac{1}{2\sqrt{2\pi y}} e^{-\frac{1}{2}(y+\theta^2)}(e^{\theta\sqrt{y}} + e^{-\theta\sqrt{y}})$, $y > 0$.
 - Show that we can write $f_Y(y) = \sum_{i=0}^{\infty} P(R = i) f_{2i+1}(y)$ where $R \sim \text{Pois}(\frac{1}{2}\theta^2)$ and f_m is the χ_m^2 density. Give a probabilistic interpretation of this formula. Hint: Use the Taylor expansion for $e^{\theta\sqrt{y}}$ and $e^{-\theta\sqrt{y}}$ in powers of \sqrt{y} .
- (B & D) Let S_n have a χ_n^2 distribution.
 - Show that if n is large, $\sqrt{S_n} - \sqrt{n}$ has approximately a $N(0, \frac{1}{2})$ distribution. This is known as Fisher's approximation. Hint: Use the delta method.
 - Using (a), deduce that $P[S_n < x] \approx \Phi(\sqrt{2x} - \sqrt{2n})$.
- (B&D) Let X_1, \dots, X_n be a sample from a population with mean μ and variance $\sigma^2 < \infty$. Suppose h has a second derivative $h^{(2)}$ continuous at μ and that $h^{(1)}(\mu) = 0$.
 - Show that $\sqrt{n}[h(\bar{X}) - h(\mu)] \rightarrow 0$ while $n[h(\bar{X}) - h(\mu)]$ is asymptotically distributed as $\frac{1}{2}h^{(2)}(\mu)\sigma^2 V$ where $V \sim \chi_1^2$.
 - Use part (a) to show that $n[\bar{X}(1 - \bar{X}) - \mu(1 - \mu)] \Rightarrow -\sigma^2 V$ with $V \sim \chi_1^2$ when $\mu = \frac{1}{2}$.

8. Let $\mathbf{X} = (X_1, \dots, X_k)$ be a random vector such that $EX_i^2 < \infty, i = 1, \dots, k$. The variance-covariance matrix of \mathbf{X} is defined by $\text{Var}(\mathbf{X}) = E[(\mathbf{X} - E\mathbf{X})(\mathbf{X} - E\mathbf{X})^T]$. Show that

(a) $\text{Var}(\mathbf{X}) = E(\mathbf{X}\mathbf{X}^T) - (E\mathbf{X})(E\mathbf{X})^T$.

(b) If A is an $m \times k$ constant matrix, then $\text{Var}(A\mathbf{X}) = A \text{Var}(\mathbf{X})A^T$.

(c) If \mathbf{X} and \mathbf{Y} are $k \times 1$ independent random vectors, then $\text{Var}(\mathbf{X} + \mathbf{Y}) = \text{Var}(\mathbf{X}) + \text{Var}(\mathbf{Y})$.

9. Let \mathbf{X} ($k \times 1$) and \mathbf{Y} ($l \times 1$) be random vectors, and define the $(k \times l)$ covariance matrix $\text{Cov}(\mathbf{X}, \mathbf{Y}) = E[(\mathbf{X} - E\mathbf{X})(\mathbf{Y} - E\mathbf{Y})^T]$. Show that

(a) If \mathbf{X} and \mathbf{Y} are independent, $\text{Cov}(\mathbf{X}, \mathbf{Y}) = \mathbf{0}$.

(b) $\text{Cov}(A\mathbf{X} + \mathbf{a}, B\mathbf{Y} + \mathbf{b}) = A \text{Cov}(\mathbf{X}, \mathbf{Y})B^T$.

(c) If $k = l$ then $\text{Var}(\mathbf{X} + \mathbf{Y}) = \text{Var}(\mathbf{X}) + \text{Cov}(\mathbf{X}, \mathbf{Y}) + \text{Cov}(\mathbf{Y}, \mathbf{X}) + \text{Var}(\mathbf{Y})$.

10. (MKB) Let U and V be independent $\mathcal{N}(0, 1)$ random variables. Define $Y = V$ and let

$$X = \begin{cases} U & \text{if } UV \geq 0 \\ -U & \text{if } UV < 0 \end{cases}$$

a. Show that X and Y each have a standard normal distribution, but that (X, Y) is not bivariate normal.

b. Show that X^2 and Y^2 are independent.

11. (MKB) Let (X, Y) be a random vector having a bivariate Pareto distribution with density $f(x, y) = c(x + y - 1)^{-p-2}$ with $p > 0$.

a. Show that $c = p(p + 1)$.

b. If $p > 1$, show that $EX = EY = p(p - 1)$.

c. For $p > 2$, find the covariance matrix of (X, Y) .