

Ma 551 — Advanced Probability

Problem Set #3 — Due November 17, 2009

Prof. Sawyer — Washington University

Problems in text are from Patrick Billingsley, *Probability and Measure*, 3rd edn, John Wiley & Sons, 1995. Here i.r.v. means independent random variables and i.i.d. means independent and identically distributed (random variables).

Six problems on two pages.

1. Let X_1, X_2, \dots be r.v.s with $F_n(y) = P(X_n \leq y)$. Assume $\sup_n E(h(X_n)) = C < \infty$ where $h(y) \geq 0$ and $\lim_{y \rightarrow \pm\infty} h(y) = \infty$. Prove that the family $\{F_n(y)\}$ is tight.

2. Let X_n, Y_n be random variables such that

(a) $P(X_n \leq y) \rightarrow F(y)$ at all points of continuity of the d.f. $F(y)$

(b) The family $\{G_n(y)\}$ for $G_n(y) = P(Y_n \leq y)$ is tight.

Assume $h_n \rightarrow 0$ as $n \rightarrow \infty$. Then show

$$P(X_n + h_n Y_n \leq y) \rightarrow F(y)$$

at all points of continuity y of $F(y)$.

3. Let $M, X_1, X_2, \dots, X_n, \dots$ be i.r.v.s such that $P(X_k \leq y) = F_X(y)$ for all y and k . (That is, the X_k are i.i.d.) Assume that M has the Poisson distribution $P(M = n) = e^{-\mu} \mu^n / n!$ for $n = 0, 1, 2, \dots$ and let $Y = \sum_{j=1}^M X_j$. (That is, Y is the sum of a random number of i.i.d. with distribution $F_X(y)$.) Prove that

$$E(e^{i\theta Y}) = \exp \left[\mu \int (e^{i\theta y} - 1) F_X(dy) \right]$$

4. (Problem 14.5, p198) Define

$$\rho(F, G) = \inf \{ \epsilon > 0 : \text{for all } x, F(x - \epsilon) - \epsilon \leq G(x) \leq F(x + \epsilon) + \epsilon \}$$

for d.f.s F, G . Prove that

(a) ρ is a metric on d.f.s

(b) For d.f.s F_n, F , $F_n(y) \rightarrow F(y)$ weakly if and only if $\rho(F_n, F) \rightarrow 0$.

(For distributions on R , $\rho(F, G)$ is called the *Lévy distance* between F and G .)

The analog of ρ for distributions on a Banach space is called the *Skorokhod metric*. Among other things, this shows that the topology of weak convergence of d.f.s on R is a metric topology.)

5. (Like Problem 27.11, p368) Let $X_1, X_2, \dots, X_n, \dots$ be i.i.d. with density $f(x) = 1/|x|^3$ for $|x| \geq 1$ and $f(x) = 0$ for $|x| < 1$. Prove that

- (a) $E(|X_k|^{2-\delta}) < \infty$ for all $\delta > 0$, $E(X_k) = 0$, but $E(X_k^2) = \infty$.
- (b) Find a constant $c > 0$ such that for all real y

$$\lim_{n \rightarrow \infty} P\left(\frac{X_1 + X_2 + \dots + X_n}{c\sqrt{n \log n}} \leq y\right) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^y e^{-(1/2)x^2} dx$$

(Hint: Let $\varphi_n(\theta) = E(e^{i\theta S_n})$ for $S_n = (X_1 + \dots + X_n)/a_n$ for appropriate a_n . Since $\varphi(\theta) = E(e^{i\theta X_k})$ is real and $|1 - \varphi(\theta)| < 1/2$ for $|\theta| < \delta$, you can take logarithms with a clear conscience. This gives an example of a central limit theorem for summands of infinite variance but with a larger denominator.)

6. Let X_k be i.r.v. with

$$\begin{aligned} P(X_k = +\sqrt{k}) &= 1/(2k) \\ P(X_k = 0) &= 1 - (1/k) \\ P(X_k = -\sqrt{k}) &= 1/(2k) \end{aligned}$$

for $k = 1, 2, \dots$. Note that $E(X_k) = 0$ and $E(X_k^2) = 1$ for all k , but that X_k are not i.i.d. Let $S_n = (X_1 + \dots + X_n)/\sqrt{n}$. Prove that

- (a) For all θ ,

$$\begin{aligned} \varphi_n(\theta) = E(e^{i\theta S_n}) &\rightarrow \exp\left(2 \int_0^1 \frac{\cos(\theta y) - 1}{y} dy\right) \\ &= \exp\left(\int_{-1}^1 (e^{i\theta y} - 1) \frac{dy}{|y|}\right) \end{aligned}$$

- (b) For all points of continuity y of some d.f. $F(y)$

$$P\left(\frac{X_1 + \dots + X_n}{\sqrt{n}} \leq y\right) \rightarrow F(y)$$

but $F(y)$ is not normal. (In particular, $F(y)$ is not the d.f. of $\mu + \sigma Z$ for any constants μ, σ where $Z \approx N(0, 1)$.)

Note the similarity between the characteristic function above and the characteristic function in Problem 3, although here the analog of $F_X(dy)$ is not normalizable.