Ma 551 — Advanced Probability

Solutions for Problem Set #3 due November 17, 2009

Prof. Sawyer — Washington University

Six problems. See the Math 551 Web site for the statement of problems.

1. For each $\epsilon > 0$, there exists $K = K_{\epsilon}$ such that $|h(y)| \ge 2C/\epsilon$ for $|y| \ge K$. Then $P(|X_n| \ge K) \le P(|h(X_n)| \ge 2C/\epsilon) \le (\epsilon/(2C))E(|h(X_n)|) < \epsilon$ for all $n \ge 1$, or equivalently $P(X_n \in [-K, K]) \ge 1 - \epsilon$ for all n.

2. Let $H_n(y) = P(X_n + h_n Y_n \leq y)$. For each $\epsilon > 0$, there exists $K = K_{\epsilon}$ such that $P(|Y_n| \geq K) < \epsilon$. Chose n_0 such that $|h_n| < \epsilon/K$ for $n \geq n_0$. Then, if $n \geq n_0$, $P(|h_n Y_n| > \epsilon) = P(|Y_n| > \epsilon/h_n) \leq P(|Y_n| > K) < \epsilon$. This implies

$$H_n(y) = P(X_n \le y - h_n Y_n) \le P(X_n \le y + \epsilon) + \epsilon$$

$$\ge P(X_n \le y - \epsilon) - \epsilon$$

(**Remark**: This implies $\rho(H_n, F_n) \to 0$ for the Lévy metric $\rho(F, G)$ defined in Problem 4. Since $\rho(F_n, F) \to 0$ and ρ satisfies the triangle inequality, $\rho(H_n, F) \to 0$ and we are done. The rest of the proof is for those who have not yet done Problem 4.)

Assume that y and $y \pm \epsilon$ are points of continuity of F(y). Then $P(X_n \leq y \pm \epsilon) \rightarrow F(y \pm \epsilon)$ and

$$F(y-\epsilon)-\epsilon \leq \liminf_{n\to\infty} H_n(y) \leq \limsup_{n\to\infty} H_n(y) \leq F(y+\epsilon)+\epsilon$$

Since y is a point of continuity of F(y), and since the above holds for a sequence $\epsilon = \epsilon_m \to 0$, it follows that $H_n(y) \to F(y)$.

3. Let $\phi(\theta) = E(e^{i\theta X_1}) = \int_{-\infty}^{\infty} e^{i\theta y} F_X(dy)$. Then

$$E\left(e^{i\theta Y}\right) = \sum_{n=0}^{\infty} E\left(e^{i\theta Y}I_{[M=n]}\right) = \sum_{n=0}^{\infty} E\left(\prod_{k=1}^{n} e^{i\theta X_{k}}I_{[M=n]}\right)$$

Since the $\{M, X_1, \dots\}$ are independent,

$$E\left(e^{i\theta Y}\right) = \sum_{n=0}^{\infty} \phi(\theta)^n P(M=n) = e^{-\mu} \sum_{n=0}^{\infty} \mu^n \phi(\theta)^n / n! = \exp\left(\mu(\phi(\theta) - 1)\right)$$

which implies the desired result.

4. (a) (i) $\rho(F,F) = 0$, and, since F(x) and G(x) are right continuous, $\rho(F,G) \neq 0$ if $F \neq G$. (ii) Note $F(x - \epsilon) - \epsilon \leq G(x)$ and $G(x) \leq F(x + \epsilon) + \epsilon$ for all x implies $G(x - \epsilon) - \epsilon \leq F(x)$ and $F(x) \leq G(x + \epsilon) + \epsilon$ for all x. Thus $\rho(F,G) = \rho(G,F)$. (iii) Suppose that $\rho(F,G) < \epsilon_1$ and $\rho(G,H) < \epsilon_2$. Then, for all x,

$$H(x - \epsilon_1 - \epsilon_2) - \epsilon_1 - \epsilon_2 \leq G(x - \epsilon_1) - \epsilon_1 \leq F(x)$$

with similar upper inequalities. Thus $\rho(F, H) < \epsilon_1 + \epsilon_2$ and ρ satisfies the triangle inequality.

(b) If $\rho(F_n, F) \to 0$, then

$$F(y-\epsilon) - \epsilon \leq F_n(y) \leq F(y+\epsilon) + \epsilon$$
 (1)

for $n \ge n_0(\epsilon)$ and all y. Thus $F_n(y) \to F(y)$ at all points of continuity y of F(y). Now assume $F_n \to F$ in distribution. If $\rho(F_n, F)$ does not converge to zero, there exists a subsequence (while we also call $\{F_n\}$) and a value $\epsilon > 0$ such that $\rho(F_n, F) \ge 2\epsilon > 0$, and thus real values y_n such that

$$F(y_n - \epsilon) - \epsilon \ge F_n(y_n) \text{ or } F_n(y_n) \ge F(y_n + \epsilon) + \epsilon$$
 (2)

with, choosing a further subsequence if necessary, one of the two inequalities in (2) holding for all n. Since F_n, F are distribution functions, there cannot exist a subsequence $y_{n_k} \to \infty$ or $y_{n_k} \to -\infty$, so that y_n are bounded. Hence there exists a further subsequence (which we also call F_n, y_n) such that $y_n \to y$ for $y \in R$. Recall that if $F_n \to F$ and $y_n \to y$, then

$$F(y-) \leq \liminf_{n \to \infty} F_n(y_n) \leq \limsup_{n \to \infty} F_n(y_n) = F(y)$$

Then by (2)

$$F(y-\epsilon) - \epsilon \ge F(y-)$$
 or $F(y) \ge F((y+\epsilon)-) + \epsilon$

for some $\epsilon > 0$, either of which provides a contradiction.

5. (a) $E(|X_k|^{2-\delta}) = 2 \int_1^\infty y^{2-\delta} y^{-3} dy < \infty$ for $\delta > 0$ but not for $\delta \le 0$, etc. (b) Set $S_n = (X_1 + X_2 + \ldots + X_n)/a_n$ for $a_n = c\sqrt{n \log n}$ for some constant c > 0. Then $E(e^{i\theta S_n}) = \phi(\theta/a_n)^n$ for $\phi(\theta) = E(e^{i\theta X_1})$. Since the X_k are symmetrically distributed, $\phi(\theta) = \phi(-\theta)$ and, for $\theta > 0$,

$$1 - \phi(\theta) = \int_{|y| \ge 1} \left(1 - e^{i\theta y}\right) \frac{dy}{|y|^3} = 2 \int_1^\infty \left(1 - \cos(\theta y)\right) \frac{dy}{y^3}$$
$$= 2\theta^2 \int_\theta^\infty (1 - \cos y) \frac{dy}{y^3}$$

By L'Hopital's rule (or otherwise), $\lim_{\theta \to 0} \int_{\theta}^{\infty} (1 - \cos y) y^{-3} dy / \log(1/\theta) = 1/2$, so that

$$\lim_{\theta \to 0} \frac{1 - \phi(\theta)}{\theta^2 \log(1/\theta)} = 1$$

In particular, for fixed $\theta > 0$,

$$1 - \phi(\theta/a_n) \sim \frac{\theta^2}{a_n^2} \log\left(\frac{a_n}{\theta}\right) = \frac{\theta^2}{c^2 n \log n} \log\left(\frac{c\sqrt{n\log n}}{\theta}\right)$$
$$= \frac{\theta^2}{2c^2 n \log n} \left(\log n + \log\log n + 2\log(c/\theta)\right) \sim \frac{\theta^2}{2c^2 n \log n}$$

as $n \to \infty$. Thus

$$\log \phi(\theta/a_n) = \log (1 - (1 - \phi(\theta/a_n))) = -(1 - \phi(\theta/a_n)) + O(1/n^2)$$

and

$$\log\left(E\left(e^{i\theta S_n}\right)\right) = n\log\phi(\theta/a_n) = -n\left(1-\phi(\theta/a_n)\right) + O(1/n) \rightarrow -\frac{\theta^2}{2c^2}$$

Thus $E(e^{i\theta S_n}) \to \exp(-\theta^2/2)$ for all θ for c = 1, which implies that S_n converges in distribution to a standard normal distribution.

6. (a) Here $S_n = (X_1 + X_2 + ... + X_n)/\sqrt{n}$ where

$$\phi_k(\theta) = E\left(e^{i\theta X_k}\right) = \frac{e^{i\theta\sqrt{k}}}{2k} + \left(1 - \frac{1}{k}\right) + \frac{e^{-i\theta\sqrt{k}}}{2k} = 1 - \frac{1 - \cos(\theta\sqrt{k})}{k} \quad (1)$$

Then $E(e^{i\theta S_n}) = \prod_{k=1}^n \phi_k(\theta/\sqrt{n})$. We give two proofs of part (a), the first using Theorem 28.3 in the text, which was proven in class, and the second arguing directly from $\Phi_n(\theta) = E(e^{i\theta S_n})$.

Proof I. Thus $S_n = \sum_{k=1}^n X_{nk}$ for $X_{nk} = X_k/\sqrt{n}$. Since $E(X_{nk}) = 0$, $E(X_{nk}^2) = E(X_k^2)/n = 1/n$, and $E(S_n^2) = 1$, the S_n are the row sums of a triangular array in the sense of page 272 of the text. Thus by Theorem 28.3

$$\lim_{n \to \infty} \left(E\left(e^{i\theta S_n}\right) - \exp\left(\sum_{k=1}^n \left(\phi_k\left(\frac{\theta}{\sqrt{n}}\right) - 1\right)\right) \right) = 0$$
(2)

for all θ . By (1)

$$\sum_{k=1}^{n} \left(\phi_k \left(\frac{\theta}{\sqrt{n}} \right) - 1 \right) = -\sum_{k=1}^{n} \frac{1 - \cos(\theta \sqrt{k/n})}{k} = -\frac{1}{n} \sum_{k=1}^{n} \left(\frac{1 - \cos(\theta \sqrt{k/n})}{k/n} \right)$$

Since $1 - \cos(y) \le (1/2)y^2$ for all y, the term within the last set of parentheses in the display is uniformly bounded for fixed θ and $1 \le k \le n$. Hence for all θ

$$\lim_{n \to \infty} \sum_{k=1}^{n} \left(\phi_k \left(\frac{\theta}{\sqrt{n}} \right) - 1 \right) = -\int_0^1 \frac{1 - \cos(\theta \sqrt{y})}{y} \, dy = -2 \int_0^1 \frac{1 - \cos(\theta y)}{y} \, dy$$
ence by (2)

Hence by (2)

$$\lim_{n \to \infty} E\left(e^{i\theta S_n}\right) = \psi(\theta) = \exp\left(-2\int_0^1 \frac{1 - \cos(\theta y)}{y} \, dy\right)$$
(3)
$$= \exp\left(-2\int_0^\theta \frac{1 - \cos(y)}{y} \, dy\right)$$

Proof II. Since $1 - \cos(y) \le (1/2)y^2$ for all y, it follows from (1) that

$$1 - \phi_k\left(\frac{\theta}{\sqrt{n}}\right) = \frac{1}{n}\left(\frac{1 - \cos(\theta\sqrt{k/n})}{k/n}\right) = O(1/n)$$

uniformly for $1 \le k \le n$ and

$$\log \phi_k\left(\frac{\theta}{\sqrt{n}}\right) = \log \left(1 - \left(1 - \phi_k\left(\frac{\theta}{\sqrt{n}}\right)\right)\right) = -\left(1 - \phi_k\left(\frac{\theta}{\sqrt{n}}\right)\right) + O\left(\frac{1}{n^2}\right)$$

where the error term is uniform for $1 \le k \le n$. Hence the logarithms below exist

where the error term is uniform for $1 \leq \kappa \leq n$. Hence the logarithms below exist and

$$\log\left(E\left(e^{i\theta S_n}\right)\right) = \sum_{k=1}^n \log\phi_k\left(\frac{\theta}{\sqrt{n}}\right) = -\sum_{k=1}^n \left(1 - \phi_k\left(\frac{\theta}{\sqrt{n}}\right) + O\left(\frac{1}{n^2}\right)\right)$$
$$= \left(-\frac{1}{n}\sum_{k=1}^n \left(\frac{1 - \cos(\theta\sqrt{k/n})}{k/n}\right)\right) + O\left(\frac{1}{n}\right)$$
$$\to -\int_0^1 \frac{1 - \cos(\theta\sqrt{y})}{y} \, dy = -2\int_0^1 \frac{1 - \cos(\theta y)}{y} \, dy$$

and (3) follows as above.

(b) The function $\psi(\theta)$ is continuous in θ and $\lim_{\theta \to 0} \psi(\theta) = 1$. Thus, by the Lévy continuity theorem, there exists a distribution function F(y) such that $F_n(y) = P(S_n \leq y) \to F(y)$ at all continuity points of F(y) and $\psi(\theta) = \int e^{i\theta y} F(dy)$.

If F(y) were normal, then $\psi(\theta) = \psi_N(\theta) = \exp(i\mu\theta - (1/2)\sigma^2\theta^2)$ for some choice of real constants μ and σ . However

$$\frac{d}{d\theta}\log\psi(\theta) = -2\frac{1-\cos\theta}{\theta} \quad \text{and} \quad \frac{d}{d\theta}\log\psi_N(\theta) = i\mu - \sigma^2\theta$$

Note that $(d/d\theta) \log \psi(\theta) = 0$ if and only if $\theta = 2n\pi$ for some $n \neq 0$, which is not true for $(d/d\theta) \log \psi_N(\theta)$ for any values of μ and σ . Thus F(y) cannot be a normal distribution.