STAT 709 Second Exam 9:55am-10:45am, Oct 26, 2016

Please show all your work for full credits.

- 1. Let $Z, Y, X_1, X_2, ...$ be random variables. Suppose that $X_n \to_d Z$ as $n \to \infty$.
 - (a) (3 points) Suppose that Y is independent of X_n for any n. Show that the sequence of c.d.f. of $X_n + Y$ converges weakly to a c.d.f., and identify that c.d.f.
 - (b) (2 points) Suppose that $E|X_n| \to E|Z| < \infty$ and $E|Y| < \infty$. Show that the sequence $\{|X_n + Y|, n = 1, 2, ...\}$ is uniformly integrable.
- 2. Let $X_1, X_2, ...$ be a sequence of independent random variables. Suppose that for each i, X_i has Lebesgue p.d.f.

$$f_i(x) = \begin{cases} \frac{1}{\Gamma(i^s)} x^{i^s - 1} e^{-x} & x > 0\\ 0 & x \le 0 \end{cases}$$

i = 1, 2, ..., where s is a constant, -1 < s < 1.

(a) (2 points) Show that

$$\frac{1}{n} \sum_{i=1}^{n} (X_i - i^s) \to_{a.s.} 0.$$

(b) (4 points) Show that

$$\sum_{i=1}^{n} (X_i - i^s) / \left(\sum_{i=1}^{n} i^s\right)^{1/2} \to_d N(0, 1).$$

(Hint:
$$|X_i - i^s|^3 \le (X_i - i^s)^2 (X_i + i^s)$$
 and $n^{-(t+1)} \sum_{i=1}^n i^t \to (t+1)^{-1}$ for $t > -1$.)

- (c) (2 points) Let $Y_n = \sum_{i=1}^n X_i / \sum_{i=1}^n i^s$. Show that $n^{(s+1)/2} \log Y_n$ converges in distribution to a random variable Z and identify the distribution of Z.
- 3. Let $X_1, ..., X_n$ $(n \ge 2)$ be i.i.d. random variables with Lebesgue p.d.f.

$$f_{\alpha,\beta}(x) = \begin{cases} \beta \alpha^{\beta} x^{-(\beta+1)} & x > \alpha \\ 0 & x \le \alpha \end{cases}$$

where $\alpha > 0$ and $\beta > 0$ are parameters.

- (a) (2 points) When both α and β are unknown, show that the transformed variables $Y_i = \log X_i, i = 1, ..., n$, has a distribution in a location-scale family, and identify the location and scale parameters.
- (b) (3 points) When β is known and α is unknown, show that $X_{(1)} = \min(X_1, ..., X_n)$ is minimal sufficient for α .
- (c) (2 points) When both α and β are unknown, obtain a 2 dimensional minimal sufficient statistic for (α, β) , and show the proof.

Solution:

- 1. (a) Using the characteristic functions, we can show that $X_n + Y \to_d Z + \tilde{Y}$, where \tilde{Y} has the same distribution as Y but is independent of Z. Note that Y is not necessarily independent of Z.
 - (b) From Theorem 1.8, $E|X_n| \to E|Z| < \infty$ implies that $\{|X_n|, n = 1, 2, ...\}$ is uniformly integrable. Since $|X_n + Y| \le |X_n| + |Y|$ and |Y| is integrable, $\{|X_n + Y|, n = 1, 2, ...\}$ is uniformly integrable.
- 2. (a) Apply Theorem 1.14(b). $EX_i = i^s$. Consider random variable $X_i EX_i = X_i i^s$. $E(X_i i^s)^2 = i^s$. If s < 1,

$$\sum_{i=1}^{\infty} \frac{i^s}{i^2} < \infty$$

(b) Apply Liapounov's condition. $EX_i = i^s$, $EX_i^2 = i^s(i^s+1)$, $EX_i^3 = i^s(i^s+1)(i^s+2)$, $E(X_i - i^s)^2 = i^s$, $\sigma_n^2 = \sum_{i=1}^n i^s$, Then

$$E|X_{i} - i^{s}|^{3} \leq E(X_{i} - i^{s})^{2}(X_{i} + i^{s})$$

$$= E(X_{i}^{3} - 2i^{s}X_{i}^{2} + i^{2s}X_{i}) + i^{s}E(X_{i} - i^{s})^{2}$$

$$= i^{s}(i^{s} + 1)(i^{s} + 2) - 2(i^{s})^{2}(i^{s} + 1) + i^{3s} + i^{2s}$$

$$= 2(i^{2s} + i^{s})$$

Then

$$\frac{1}{\sigma_n^{3/2}} \sum_{i=1}^n E|X_i - i^s|^3 \le \frac{2\sum_{i=1}^n (i^s)^2 + i^s}{(\sum_{i=1}^n i^s)^{3/2}} \approx 2\left(\frac{n^{2s+1}}{2s+1} + \frac{n^{s+1}}{s+1}\right) / \left(\frac{n^{s+1}}{s+1}\right)^{3/2} \to 0$$

if
$$-1 < s < 1$$
.

(c) Apply the delta method to

$$\left(\sum_{i=1}^{n} i^{s}\right)^{1/2} \left(Y_{n} - 1\right) = \left(\sum_{i=1}^{n} i^{s}\right)^{1/2} \left(\frac{\sum_{i=1}^{n} X_{i}}{\sum_{i=1}^{n} i^{s}} - 1\right) \to_{d} N(0, 1)$$

and
$$g(y) = \log y$$
, $g(1) = 0$, $g'(y) = 1/y$, $g'(1) = 1$. Then $Z \sim N(0, s + 1)$.

- 3. (a) Y_i has the two parameter exponential distribution, location parameter $\log \alpha$ and scale parameter $1/\beta$.
 - (b) Apply Theorem 2.3(iii). For $x = (x_1, ..., x_n)$ and $y = (y_1, ..., y_n)$,

$$\frac{f_{\alpha,\beta}(x)}{f_{\alpha,\beta}(y)} = \frac{\prod_i y_i^{\beta+1}}{\prod_i x_i^{\beta+1}} \frac{I_{(\alpha,\infty)}(x_{(1)})}{I_{(\alpha,\infty)}(y_{(1)})}$$

The first factor does not depend on α . The 2nd factor does not depend on α implies that $x_{(1)} = y_{(1)}$.

(c) The minimal sufficient statistic is $(\prod_i X_i, X_{(1)})$. The proof is similar to that of (b).