Fall 2018 STSCI 5080 Discussion 8 (10/19)

Normal approximation to Poisson

Theorem 1 (Continuity theorem for mgfs). Let X_n and X have mgfs ψ_n and ψ , respectively. If $\psi_n(\theta) \to \psi(\theta)$ for any θ in an open interval containing the origin, then $X_n \stackrel{d}{\to} X$.

Theorem 2. Let $X_n \sim Po(\lambda_n)$ and $\lambda_n \to \infty$ as $n \to \infty$. Then

$$\frac{X_n - \lambda_n}{\sqrt{\lambda_n}} \stackrel{d}{\to} N(0, 1).$$

Proof. Recall that $E(X_n) = \lambda_n$ and $Var(X_n) = \lambda_n$. The mgf of X_n is

$$\psi_{X_n}(\theta) = e^{\lambda_n(e^{\theta} - 1)}.$$

Hence the mgf of $Y_n = (X_n - \lambda_n)/\sqrt{\lambda_n}$ is

$$\psi_{Y_n}(\theta) = E(e^{\theta Y_n}) = E(e^{\theta(X_n - \lambda_n)/\sqrt{\lambda_n}}) = e^{-\theta\sqrt{\lambda_n}} E(e^{\theta X_n/\sqrt{\lambda_n}})$$
$$= e^{-\theta\sqrt{\lambda_n}} \psi_{X_n}(\theta/\sqrt{\lambda_n})$$
$$= e^{-\theta/\sqrt{\lambda_n}} e^{\lambda_n} (e^{\theta/\sqrt{\lambda_n}} - 1).$$

By Taylor's theorem, we can expand e^x as

$$e^x = 1 + x + \frac{x^2}{2} + x^2 R(x),$$

where $\lim_{x\to 0} R(x) = 0$. Hence

$$\lambda_n(e^{\theta/\sqrt{\lambda_n}}-1) = \theta\sqrt{\lambda_n} + \frac{\theta^2}{2} + \theta^2 R(\theta/\sqrt{\lambda_n}),$$

so that

$$e^{-\theta/\sqrt{\lambda_n}}e^{\lambda_n(e^{\theta/\sqrt{\lambda_n}}-1)} = e^{-\theta^2/2} + \theta^2 R(\theta/\sqrt{\lambda_n}).$$

Since $\lim_{n\to\infty} R(\theta/\sqrt{\lambda_n}) = 0$, we conclude that

$$\lim_{n \to \infty} \psi_{Y_n}(\theta) = e^{-\theta^2/2},$$

which is the mgf of N(0,1). By the continuity theorem, we have $Y_n \stackrel{d}{\to} N(0,1)$.

Problems

- 1. Show that if $E(|X_n|) \to 0$, then $X_n \stackrel{P}{\to} 0$.
- 2. Let X_n be such that

$$P(X_n = n) = \frac{1}{n}$$
 and $P(X_n = 0) = 1 - \frac{1}{n}$.

Show that $E(X_n) = 1$ but $X_n \stackrel{P}{\to} 0$. So convergence in probability does not imply moment convergence.

- 3. (Rice 5.4.1) Let X_1, X_2, \ldots be a sequence of independent random variables with $E(X_i) = \mu$ and $Var(X_i) = \sigma_i^2$. Show that if $n^{-2} \sum_{i=1}^n \sigma_i^2 \to 0$, then $\overline{X}_n \stackrel{P}{\to} \mu$.
- 4. (Rice 5.4.3) Suppose that the number of insurance claims, N, filed in a year is Poisson distributed with E(N) = 10,000. Use the normal approximation to the Poisson to approximate P(N > 10,200).
- 5. (Rice 5.4.5) Using mgfs, show that as $n \to \infty$, $p \to 0$, and $np \to \lambda$, the binomial distribution with parameters n and p tends to the Poisson distribution.
- 6. Let $X_1, \ldots, X_n \sim U[0,1]$ i.i.d. Show that $nX_{(1)} \stackrel{d}{\to} Ex(1)$ by directly evaluating the cdf of $X_{(1)}$.
- 7. (Rice 5.4.10) A six-sided die is rolled 100 times. Using the normal approximation, find the probability that the face showing a six turns up between 15 and 20 times. Find the probability that the sum of the face values of the 100 trials is less than 300.
- 8. (Rice 5.4.16) Suppose that X_1, \ldots, X_{20} are independent random variables with density function

$$f(x) = 2$$
 if $0 \le x \le 1$.

Let $S = X_1 + \cdots + X_{20}$. Use the central limit theorem to approximate $P(S \le 10)$.

Solutions

1. By Markov's inequality,

$$P(|X_n| > \varepsilon) \le \frac{E(|X_n|)}{\varepsilon} \to 0$$

as $n \to \infty$. This implies that $X_n \stackrel{P}{\to} 0$.

2. We have

$$E(X_n) = 0 \cdot \left(1 - \frac{1}{n}\right) + n \cdot \frac{1}{n} = 1.$$

In addition, for any $\varepsilon > 0$,

$$P(|X_n| > \varepsilon) = P(X_n = n) = \frac{1}{n}$$

for sufficiently large n. Hence

$$\lim_{n \to \infty} P(|X_n| > \varepsilon) = 0,$$

which implies that $X_n \stackrel{P}{\to} 0$.

3. (Rice 5.4.1) By Chebyshev's inequality, we have

$$P(|\overline{X}_n - \mu| > \varepsilon) \le \frac{\operatorname{Var}(\overline{X}_n)}{\varepsilon^2} = \frac{n^{-2} \sum_{i=1}^n \sigma_i^2}{\varepsilon^2}.$$

Since $n^{-2} \sum_{i=1}^{n} \sigma_i^2 \to 0$ by assumption, we have $\overline{X}_n \stackrel{P}{\to} \mu$.

4. (**Rice 5.4.3**) $N \sim Po(\lambda)$ with $\lambda = 10000$. So

$$P(N > 10200) = P\left(\frac{N - 10000}{100} > \frac{10200 - 10000}{100}\right) \approx 1 - \Phi(2) \approx 0.023.$$

5. (Rice 5.4.5) The mgf of $X_n \sim Bin(n, p)$ is

$$\psi_n(\theta) = \{1 + p(e^{\theta} - 1)\}^n.$$

Let $\lambda_n = np$ so that $p = \lambda_n/n$. Then

$$\psi_n(\theta) = \left\{1 + \frac{\lambda_n}{n}(e^{\theta} - 1)\right\}^n.$$

Since $\lambda_n \to \lambda$, the RHS $\to e^{\lambda(e^{\theta}-1)}$, which is the mgf of $Po(\lambda)$. By the continuity theorem, we have $X_n \stackrel{d}{\to} Po(\lambda)$.

6. The cdf of $X_{(1)}$ is

$$P(X_{(1)} \le x) = 1 - (1 - x)^n$$

for $0 \le x \le 1$. Hence, for $x \ge 0$,

$$P(nX_{(1)} \le x) = P(X_{(1)} \le x/n) = 1 - (1 - x/n)^n \to 1 - e^{-x}.$$

If x < 0, then $P(nX_{(1)} \le x) = 0$, so that we have $nX_{(1)} \stackrel{d}{\to} Ex(1)$.

7. (**Rice 5.4.10**) Let X be the number that the face shows a six; then $X \sim Bin(100, 1/6)$. We want to find P(15 < X < 20). We have E(X) = 100/6 = 50/3 and $Var(X) = 100 \cdot 5/36 = 125/9$. Hence,

$$\begin{split} P(15 < X < 20) &= P\left(\frac{15 - 50/3}{125/9} < \frac{X - 50/3}{125/9} < \frac{20 - 50/3}{125/9}\right) \\ &\approx \Phi(0.24) - \Phi(-0.12) = \Phi(0.24) + \Phi(0.12) - 1 \approx 0.182. \end{split}$$

Next, let Y_i denote the face value of the *i*-th trial, and we want to find $P(S \leq 300)$ where $S = \sum_{i=1}^{100} Y_i$. The mean and variance of Y_i are

$$E(Y_i) = \sum_{j=1}^{6} \frac{j}{6} = \frac{7}{2}$$
 and $Var(Y_i) = E(Y_i^2) - \{E(Y_i)\}^2 = \sum_{j=1}^{6} \frac{j^2}{6} - \frac{49}{4} = \frac{91}{6} - \frac{49}{4} = \frac{35}{12}$.

Hence,

$$P(S \le 300) = P\left(\frac{S - 350}{\sqrt{875/3}} \le \frac{300 - 350}{\sqrt{875/3}}\right) \approx \Phi(-2.93) \approx 0.002.$$

8. (Rice 5.4.16) The mean and variance of X_i are

$$E(X_i) = \int_0^1 2x^2 dx = \frac{2}{3}$$
 and $Var(X_i) = E(X_i^2) - \{E(X_i)\}^2 = \int_0^1 2x^3 dx - \frac{4}{9} = \frac{1}{2} - \frac{4}{9} = \frac{1}{18}$.

So,

$$P(S \le 10) = P\left(\frac{S - 40/3}{\sqrt{10/9}} \le \frac{10 - 40/3}{\sqrt{10/9}}\right) \approx \Phi(-3.16) \approx 0.001.$$