

Use Theorem D.1.1(c) as our usual definition of \Rightarrow in \mathbb{R}

$$\text{i.e. } \mu_n = \mathcal{L}(X_n), \quad \mu = \mathcal{L}(X)$$

$$\mu_n \Rightarrow \mu \text{ iff } \mu_n((-\infty, x]) \rightarrow \mu((-\infty, x]) \text{ as } n \rightarrow \infty$$

$$\text{In all } x \rightarrow \mu(\{x\}) = 0$$

$$\mu_n((-\infty, x]) = F_n(x), \quad F_n = \text{cdf of } \mathcal{L}(X_n)$$

$$F = \text{cdf of } \mu = \mathcal{L}(X)$$

$$\mu(\{x\}) = 0 \text{ iff } F \text{ is continuous at } x$$

Special case:

X_n, X are \mathbb{N}_0 valued, i.e. non-negative integer valued

$$F_n(x) = P(X_n \leq x)$$

$$= \sum_{j=0}^{\infty} P(X_n = j) \quad \leftarrow \text{is a finite sum} \quad (\text{sum of a finite number of terms})$$

$$\therefore F(x) = \sum_{j=0}^{\infty} P(X=j)$$

at all x for which F is continuous

i.e. all x except for those that are point masses of $\mathcal{L}(X)$

$$\text{i.e. } x + j, \quad j \in \mathbb{N}_0$$

$$a_j = P(X=j)$$

$$a_{j,n} = P(X_n=j)$$

$$\therefore F(x) = \sum_{j=0}^{\infty} a_j$$

$$F_n(x) = \sum_{j=0}^{\infty} a_{j,n} \quad P(X_n=j) \rightarrow P(X=j)$$

$$\therefore F_n \Rightarrow F \text{ iff } \sum_{j=0}^{\infty} a_{j,n} \rightarrow \sum_{j=0}^{\infty} a_j$$

Eg: $X_n \sim \text{Binomial}(n, p_n) \rightarrow \lambda > 0$ as $n \rightarrow \infty$

To show X_n converges in distribution to Poisson, λ

$$\begin{aligned} \text{Show: } P(X_n=j) &= \binom{n}{j} p_n^j (1-p_n)^{n-j} & \binom{n}{j} &= \frac{n!}{j!(n-j)!} \\ &\rightarrow \lambda^j e^{-\lambda} & \text{as } n \rightarrow \infty & \end{aligned}$$

Eg ②: $X_n \geq 0, X \geq 0$ w.p. 1 (a.s.) X_n iid F

$$Y_n = \min(X_1, \dots, X_n)$$

$$G_n = \text{cdf of } Y_n$$

$$\text{For } x > 0 \quad P(Y_n \leq x) = 1 - P(Y_n > x)$$

$$\begin{cases} G_n(x) = P(Y_n \leq x) = 1 - (1 - F(x))^n \\ = 1 - (1 - F(x))^n \end{cases}$$

$$\text{If } F(x) = \int_0^x f(u) du \approx x f(0) \quad (\text{if } f(0) = \lim_{x \rightarrow 0^+} f(x))$$

$$\text{Aside: small } x, \text{ say } \frac{w}{n} \quad \frac{F(x)}{x} \rightarrow \lambda > 0 \text{ as } x \rightarrow 0^+$$

$$G_n\left(\frac{w}{n}\right) = 1 - \left(1 - \frac{w}{n}\right)^n \rightarrow 1 - e^{-w}$$

$$P(Y_n \leq \frac{w}{n})$$

$$= P(n \cdot Y_n \leq w)$$

$$= H_n(w)$$

$$H_n = \mathcal{L}(n Y_n)$$

$$\begin{aligned} F(x) &\approx \text{small} \\ &= \lambda x^2 \end{aligned}$$

$$H_n(w) = \frac{F(x)}{x^n} = \frac{1 - (1 - F(x))^n}{x^n} = e^{-\lambda x}$$

If $\frac{F(x)}{x} \rightarrow \lambda > 0$

$$\frac{G_n(w)}{G_n(\frac{w}{n})} = \frac{1 - (1 - F(x))^n}{1 - (1 - \lambda(\frac{w}{n}))^n}$$

$$H_n(w) = P(\sqrt{n}Y_n \leq w) = 1 - (1 - \frac{\lambda w}{n})^n \rightarrow 1 - e^{-\lambda w}$$

$H_n = \text{dist. of } \sqrt{n}Y_n$

Cannot work directly using this method
to study Central Limit Theorems.

$$X_n \sim \text{Binom}(n, \theta) \quad \theta \text{ fixed}$$

$$Z_n = \frac{(X_n - n\theta)}{\sqrt{n\theta(1-\theta)}}$$

$$\theta = \frac{1}{2}$$

$$Z_n = \frac{(X_n - \frac{n}{2})}{\sqrt{\frac{n}{4}}}$$

$$F_n(x) = P(Z_n \leq x)$$

$$= P(X_n \leq \frac{n}{2} + \frac{1}{2}\sqrt{n} \cdot x)$$

$$= \sum_{j=0}^{\lfloor \frac{n}{2} + \frac{1}{2}\sqrt{n} \cdot x \rfloor} \binom{n}{j} \left(\frac{1}{2}\right)^j \left(\frac{1}{2}\right)^{n-j}$$

$$= \sum_{j=0}^{\lfloor \frac{n}{2} + \frac{1}{2}\sqrt{n} \cdot x \rfloor} \binom{n}{j} \cdot \frac{1}{2^n}$$

$$\text{April 7} \quad 12:30 - 3:30 \text{ EST (EDT?)}$$

$$\leq 9:30 - 12:30 \text{ PST (PDT?)}$$