

Week 7

* the moments do not capture all info abt the dist. in the unbounded support scenario

Q: Why do we care?

A: we're in need to approximate RVs

Then (convergence of MGF's leads to convergence of CDF).
 Suppose X_1, X_2, \dots is a sequence of random variables,
 each w/ mgf $M_{X_i}(t)$.
 Suppose $\lim_{i \rightarrow \infty} M_{X_i}(t) \rightarrow M_X(t)$
 for all t in a neighborhood of 0, and $M_X(t)$ is a mgf. (or)
 Then $X_i \rightarrow X$ in distribution
 where X is a RV w/ mgf $M_X(t)$
 i.e. $F_{X_i}(x) \rightarrow F_X(x)$ at all points x where cdf F_X is continuous

Proof (idea). Beyond scope of class

\rightarrow cdf $\exists h > 0$ s.t. $\forall t \in (-h, h), M_{X_i}(t) = M_X(t)$

def: $X_i \rightarrow X$ in distribution ($X_i \xrightarrow{d} X$)

if $F_{X_i}(x) \rightarrow F_X(x)$ at all continuity points of F_X

- implicit in above def.

Ex (Poisson Approx): Let $X \sim \text{Binomial}(n, p)$. We know $\begin{cases} E(X) = np \\ V(X) = np(1-p) \end{cases}$

As n gets large, X "behaves" like a Poisson RV.

$Y \sim \text{Poisson}(\lambda)$ if

$$f_Y(y) = e^{-\lambda} \frac{\lambda^y}{y!}, \quad y = 0, 1, \dots$$

$$E(Y) = V(Y) = \lambda$$

$$M_Y(t) = e^{\lambda(e^t - 1)}$$

We will establish that

$$P(X \leq x) \approx P(Y \leq x) \text{ as } n \rightarrow \infty \quad \forall x \in \mathbb{N}$$

More precisely, we also need $p \rightarrow 0$

s.t. np constant $\approx \lambda$.

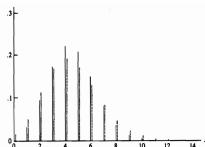


Figure 2.3.3. Poisson (dotted line) approximation to the binomial (solid line), $n = 15$, $p = .3$

Suppose $\begin{cases} np = \lambda = \text{const} \\ n \rightarrow \infty, p \rightarrow 0 \end{cases}$

$$M_X(t) = (pe^t + 1 - p)^n$$

Since $X_n = y_1 + \dots + y_n$ where $y_i \sim \text{Bernoulli}(p)$ s.t. $\sum_{i=1}^n P(y_i = 1) = p$

$$\Rightarrow M_{X_n}(t) = E(e^{tX_n}) = E(e^{t(y_1 + \dots + y_n)}) = E(e^{ty_1}) \dots E(e^{ty_n})$$

$$\text{and } y_1, \dots, y_n \text{ are independent} \\ = \prod_{i=1}^n E(e^{ty_i}) = (pe^t + 1 - p)^n$$

$$\text{So, } M_{X_n}(t) = (pe^t + 1 - p)^n$$

$$= (1 + p(e^t - 1))^n$$

$$= (1 + \frac{1}{n}(e^t - 1)\lambda)^n, \quad p = \frac{1}{n}\lambda$$

$$= (1 + \frac{1}{n}(e^t - 1)\lambda)^n$$

$$= (1 + \frac{1}{n}(e^t - 1)\lambda)^n \xrightarrow{\text{Stirling?}} e^{(e^t - 1)\lambda}$$

$$= (1 + \frac{1}{n}(e^t - 1)\lambda)^{\frac{1}{(e^t - 1)\lambda} \cdot (e^t - 1)\lambda} \xrightarrow{\text{Stirling?}} e^{(e^t - 1)\lambda}, \quad e = \lim_{x \rightarrow \infty} (1 + \frac{1}{x})^x$$

$$\Rightarrow e^{(e^t - 1)\lambda} \text{ as } n \rightarrow \infty$$

Hence, $M_X(t) \rightarrow M_Y(t)$ $\forall t$

By the convergence theorem of mgf.

If $X_n \sim \text{Binomial}(n, p)$, $\lambda = np$. Then

$$F_{X_n}(x) \rightarrow F_Y(x) \text{ as } n \rightarrow \infty \quad \forall x \text{ where } F_Y \text{ is continuous.}$$

For Poisson, F_Y is a step function w/ continuity at $y \in \mathbb{N}$.

For $y \in \mathbb{N}$, $F_Y(y) = F_Y(y + \frac{1}{2})$ continuous at $y + \frac{1}{2}$.

So,

$$P(X_n \leq x) \rightarrow P(Y \leq x) \text{ as } n \rightarrow \infty \quad \forall x \in \mathbb{N} \text{ (in fact, } \forall x \in \mathbb{R}).$$

Remarks: Still holds if $x_n \sim \text{Binomial}(n, p_n)$ and $n \rightarrow \infty, p_n \rightarrow 0$ s.t. $np_n \rightarrow \lambda$.

Thm: $\forall a, b \in \mathbb{R}$,

$$M_{a+b}(t) = e^{bt} M_a(at)$$

Proof (Sketch): $M_{a+b} = E[e^{t(aX+b)}]$

$$= E[e^{atX} e^{bt}]$$

$$= e^{bt} E[e^{atX}]$$

$$= e^{bt} M_a(at) \quad \square$$

2.4: TOOLS

Interchange Integral and Differentiation

Thm (Leibniz's Rule): If

- $f(x, y), a(y), b(y)$ are differentiable wrt y

- $\frac{\partial f}{\partial y}(x, y)$ is continuous on $x \in (x_0, x_1)$ can be weakened

Then for $a(y), b(y) \in (x_0, x_1)$:

$$\begin{aligned} \frac{d}{dy} \int_{a(y)}^{b(y)} f(x, y) dx &= f(b(y), y) \frac{db}{dy} - f(a(y), y) \frac{da}{dy} \\ &+ \int_{a(y)}^{b(y)} \frac{\partial f}{\partial y}(x, y) dx \end{aligned}$$

Corollary: If $a(y) = a, b(y) = b$, then

$$\frac{d}{dy} \int_a^b f(x, y) dx = \int_a^b \frac{\partial f}{\partial y}(x, y) dx$$

To strengthen the above result for infinite domain of the integral, we need another tool:

Interchanging Limit and Integral

Thm: Suppose

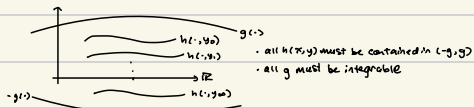
- $h(x, y)$ is continuous at $y=y_0$ for each fixed x
- there is a function $g(x)$ s.t.

$$\begin{aligned} &|h(x, y)| \leq g(x) \quad \forall x, y \\ &\text{evaluate condition} \int_{-\infty}^{\infty} g(x) dx < \infty \end{aligned}$$

$$\text{Then, } \lim_{y \rightarrow y_0} \int_{-\infty}^{\infty} h(x, y) dx = \int_{-\infty}^{\infty} \lim_{y \rightarrow y_0} h(x, y) dx$$

Remarks: g is called the dominating or envelope function of f

- known as (a version/consequence of) Lebesgue's Dominated Convergence Thm (DCVT)
- think of h as a collection of functions in x if $x, y \in \mathbb{R}$



Interchange Differentiation and Integral

Thm: Suppose

- for each $x, f(x, y)$ is differentiable wrt y , at $y \in (y_0)$
- for each $y \in (y_0)$, there is a function $g(x, y)$ and $\delta_0 > 0$ s.t.

$$\begin{aligned} &\int_{-\infty}^{\infty} \frac{\partial f}{\partial y}(x, y) dy \leq g(x, y) \quad \forall y \in (y_0 - \delta_0, y_0 + \delta_0) \\ &\int_{-\infty}^{\infty} g(x, y) dx < \infty \quad \text{for each } y \in (y_0) \end{aligned}$$

$$\text{Then } \frac{d}{dy} \int_{-\infty}^{\infty} f(x, y) dx = \int_{-\infty}^{\infty} \frac{\partial f}{\partial y}(x, y) dx$$

holds for each $y \in (y_0)$

Remarks: Leibniz's rule holds under this weaker envelope condition for $f(x, y)$.

- if we need only to differentiate at $y = y_0$, then it is sufficient that the envelope condition be satisfied for a neighborhood of y_0 i.e. $y \in (y_0 - \delta_0, y_0 + \delta_0)$ for some $\delta_0 > 0$, and only a dominating function $g(x, y_0)$ (for fixed $y = y_0$) is required
- this thm is a direct consequence of Lebesgue's dominated convergence thm (DCVT)

$$\begin{aligned} \text{Proof (Leibniz's Rule): } (HS) &= \lim_{h \rightarrow 0} \frac{1}{h} \left(\int_{a(y_0+h)}^{b(y_0+h)} f(x, y_0+h) dx - \int_{a(y_0)}^{b(y_0)} f(x, y_0) dx \right) \\ &= \lim_{h \rightarrow 0} \frac{1}{h} \left(\int_{a(y_0)}^{b(y_0)} [f(x, y_0+h) - f(x, y_0)] dx + \int_{b(y_0)}^{b(y_0+h)} f(x, y_0+h) dx + \int_{a(y_0+h)}^{a(y_0)} f(x, y_0+h) dx \right) \end{aligned}$$

$$\text{where } A = \lim_{h \rightarrow 0} \frac{1}{h} \int_{a(y_0)}^{b(y_0)} [f(x, y_0+h) - f(x, y_0)] dx$$

$$= \lim_{h \rightarrow 0} \int_{a(y_0)}^{b(y_0)} \frac{1}{h} [f(x, y_0+h) - f(x, y_0)] dx$$

$$= \int_{a(y_0)}^{b(y_0)} \lim_{h \rightarrow 0} \frac{1}{h} [f(x, y_0+h) - f(x, y_0)] dx \quad \text{next theorem}$$

$$B = \lim_{h \rightarrow 0} \frac{1}{h} \int_{b(y_0)}^{b(y_0+h)} f(x, y_0+h) dx$$

$$= \lim_{h \rightarrow 0} \frac{1}{h} \frac{a(y_0+h) - a(y_0)}{h} = \frac{1}{a'(y_0)} \int_{a(y_0)}^{a(y_0+h)} f(x, y_0+h) dx$$

$$= - \frac{d}{dy} \int_{a(y)}^{b(y)} f(x, y) dy$$

Likewise,

$$C = \frac{d}{dy} \int_{a(y)}^{b(y)} f(x, y) dy \quad \square$$

