

Lecture 35.

Recall that a sequence of RV's $\{X_n\}$ with distribution functions $F_n(t)$ "converges in distribution" to a RV X with c.d.f. $F(t)$ iff

$$\lim_{n \rightarrow \infty} F_n(t) = F(t) \quad \forall t \text{ where } F(t) \text{ is continuous.}$$

Lemma: Suppose X_n has moment generating function $M_{X_n}(t)$. If $M_{X_n}(t) \rightarrow M_X(t) \quad \forall t$, then X_n converges in distribution to X .

Theorem (Central Limit Theorem)

Let $X_1, X_2, \dots$ be a sequence of independent identically distributed RV's with mean μ and variance σ^2 . Then

$$Z_n = \frac{X_1 + \dots + X_n - n\mu}{\sigma\sqrt{n}}$$

converges in distribution to $Z = N(0, 1)$.

Proof: We first consider the situation where the X_i are identically distributed with mean $\mu = 0$ and variance $\sigma^2 = 1$.

- We assume that the moment generating function exists and is finite, $M(t) = E[e^{tx_i}] \quad \forall i$.

- Observe that the moment generating function of $X_i/\sqrt{n}$ is $E[e^{tX_i/\sqrt{n}}] = M\left(\frac{t}{\sqrt{n}}\right)$.

- By independence, the moment generating function of $Z_n = \sum_{i=1}^n \frac{X_i}{\sqrt{n}}$ is $\left[M\left(\frac{t}{\sqrt{n}}\right)\right]^n$

- consider the function

$$L(t) = \log(M(t))$$

this is called the cumulant generating function. $L^{(n)}(0)$ is the n th cumulant.

- Observe that:

$$L(0) = 0$$

$$L'(0) = \frac{m'(0)}{m(0)} = \frac{\mu}{1} = 0.$$

$$L''(0) = \frac{m(0)m''(0) - (m'(0))^2}{m(0)^2} = \frac{\text{Var}(X_1)}{1} = \sigma^2 = 1.$$

Goal: $\left[M\left(\frac{t}{\sqrt{n}}\right)\right]^n \rightarrow e^{t^2/2} = E[e^{tZ}]$ as $n \rightarrow \infty$.

Equiv: $nL\left(\frac{t}{\sqrt{n}}\right) \rightarrow t^2/2$ as $n \rightarrow \infty$.

Now, we use some calculus:

$$\lim_{n \rightarrow \infty} nL\left(\frac{t}{\sqrt{n}}\right) = \lim_{n \rightarrow \infty} \frac{L\left(\frac{t}{\sqrt{n}}\right)}{n^{-1}}$$

$$= \lim_{n \rightarrow \infty} \frac{L'(t/\sqrt{n}) n^{-3/2} t (-1/2)}{-n^{-2}} \quad (\text{L'Hôpital})$$

$$= \lim_{n \rightarrow \infty} \frac{L'(t/\sqrt{n}) t}{2 n^{-1/2}}$$

$$= \lim_{n \rightarrow \infty} \frac{L''(t/\sqrt{n}) \cancel{(-1/2)} \cancel{n^{-3/2}} t^2}{2 \cancel{(-1/2)} \cancel{n^{-3/2}}} \quad (\text{L'Hôpital})$$

$$= \lim_{n \rightarrow \infty} L''(t/\sqrt{n}) \frac{t^2}{2}$$

$$= L''(0) \frac{t^2}{2} = \frac{t^2}{2},$$

as required.

For the case where σ^2, μ are arbitrary,
 apply the above case to $X_i^* := \frac{(X_i - \mu)}{\sigma}$.

Review session begins.

Ex: Let X be the uniform RV on $(2, 3)$
Let Y be uniform on $(4, 6)$.

Assume X and Y are independent. Find the density function of $X + Y$.

Solution: Then density function of X and Y are

$$f_X(x) = \begin{cases} \frac{1}{3-2} = 1 & , \quad 2 < x < 3 \\ 0 & \text{o/w.} \end{cases}$$

$$f_Y(y) = \begin{cases} \frac{1}{6-4} = \frac{1}{2} & 4 < y < 6 \\ 0 & \text{o/w.} \end{cases}$$

The density function of $X + Y$ is the convolution of f_X and f_Y :

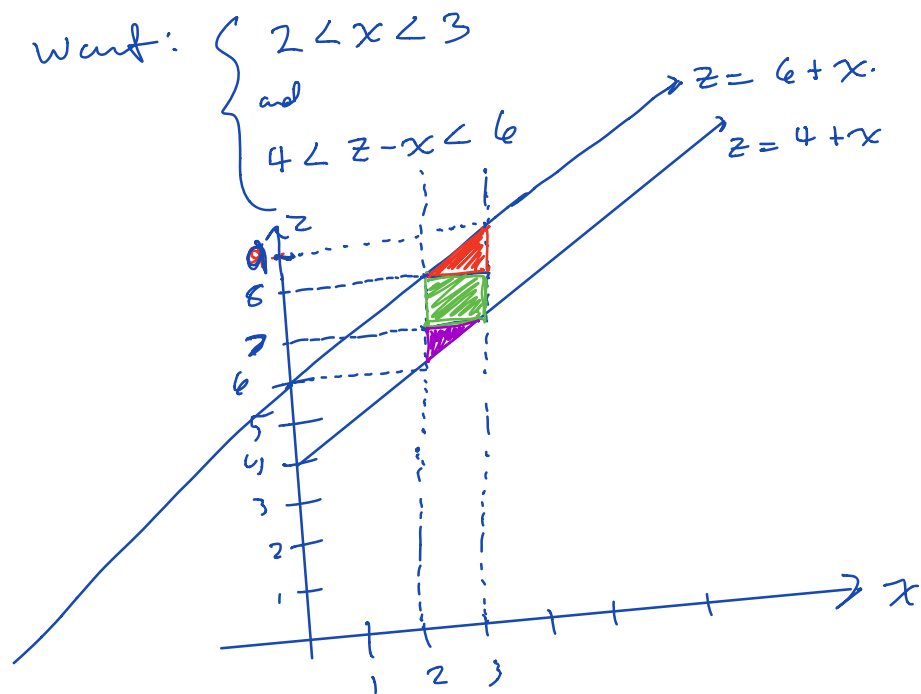
$$f_{X+Y}(z) = (f_X * f_Y)(z) = \int_{-\infty}^{\infty} f_X(z-y) f_Y(y) dy = \int_{-\infty}^{\infty} f_Y(z-x) f_X(x) dx.$$

$$= \int_2^3 f_Y(z-x) 1 dx$$

$$f_Y(z-x) = \begin{cases} 1/2 & \text{if } 4 < z-x < 6 \\ 0 & \text{o/w.} \end{cases}$$

these ending the same.

so the integral
is non-zero iff
 $2 < x < 3$
and $4 < z-x < 6$.



Three regions:

when $6 < z \leq 7$, we have $2 < x < z-4$.

when $7 < z \leq 8$, we have $2 < x < 3$.

when $8 \leq z < 9$ we have $z-6 < x < 3$

$$\text{So } f_{x+y}(z) = \begin{cases} \int_2^{z-4} \frac{1}{2} dx & 6 < z \leq 7 \\ \int_2^3 \frac{1}{2} dx & 7 < z \leq 8 \\ \int_{z-6}^3 \frac{1}{2} dx & 8 \leq z < 9 \end{cases}$$

$$= \begin{cases} \frac{z-6}{2} & 6 < z \leq 7 \\ \frac{1}{2} & 7 < z < 8 \\ \frac{9-z}{2} & 8 \leq z < 9 \\ 0 & \text{otherwise} \end{cases}$$

