



# Prove *Central Limit Theorem* using *Moment Generating Function*

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- MGF & k-th moment

- $\mu_k = E[X^k]$

- $M_X(t) = E(e^{tX}) \rightarrow$  MGF of r.v.  $X$


- If  $Z = X + Y$ , with  $X$  independent of  $Y$ , then
$$M_Z(t) = M_X(t) * M_Y(t)$$

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Theorem:  $X_1, X_2, \dots, X_n$  are i.i.d., with mean  $\mu$ , variance  $\sigma^2$ , we can have

$$\lim_{n \rightarrow \infty} \frac{X_1 + X_2 + \dots + X_n}{n} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

- Let  $S_n = X_1 + X_2 + \dots + X_n$ , so  $\bar{X} = \frac{S_n}{n}$
- $M_{S_n}(t) = E(e^{t(X_1 + X_2 + \dots + X_n)}) = [M_X(t)]^n$
- $M_{\bar{X}}(t) = E(e^{\frac{t(X_1 + X_2 + \dots + X_n)}{n}}) = E(e^{\frac{t}{n}(X_1 + X_2 + \dots + X_n)}) =$   
 $M_{S_n}\left(\frac{t}{n}\right) = \left[M_X\left(\frac{t}{n}\right)\right]^n$




- Standardize  $\bar{X}$  :  $Y = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} = \frac{\sqrt{n}}{\sigma} \bar{X} - \frac{\mu\sqrt{n}}{\sigma}$

- $$M_Y(t) = E(e^{tY}) = E\left(e^{t\frac{\sqrt{n}}{\sigma}\bar{X}} e^{-t\frac{\mu\sqrt{n}}{\sigma}}\right) = e^{-t\frac{\mu\sqrt{n}}{\sigma}} M_{\bar{X}}\left(\frac{\sqrt{n}}{\sigma} t\right) =$$

$$e^{-t\frac{\mu\sqrt{n}}{\sigma}} \left[M_X\left(\frac{\sqrt{n}}{\sigma n} t\right)\right]^n = e^{-t\frac{\mu\sqrt{n}}{\sigma}} \left[M_X\left(\frac{t}{\sigma\sqrt{n}}\right)\right]^n$$

- Now we just need to prove the **MGF of Y** has the form of a standard normal random variable as n goes to infinity!



$$M_Y(t) = e^{-t\frac{\mu\sqrt{n}}{\sigma}} \left[ M_X \left( \frac{t}{\sigma\sqrt{n}} \right) \right]^n$$

- $\ln M_Y(t) = \ln \left( e^{-t\frac{\mu\sqrt{n}}{\sigma}} \right) + \ln \left[ M_X \left( \frac{t}{\sigma\sqrt{n}} \right) \right]^n = -t\frac{\mu\sqrt{n}}{\sigma} + n \ln M_X \left( \frac{t}{\sigma\sqrt{n}} \right)$
- Let  $p = \frac{t}{\sqrt{n}\sigma}$ , then  $\sqrt{n} = \frac{t}{p\sigma}$ , *if  $n \rightarrow \infty, p \rightarrow 0$*
- $\ln M_Y(t) = -\frac{\mu t^2}{p\sigma^2} + \frac{t^2}{p^2\sigma^2} \ln M_X(p) = \frac{t^2}{\sigma^2} \left( -\frac{\mu}{p} + \frac{1}{p^2} \ln M_X(p) \right) =$   
 $\frac{t^2}{\sigma^2} \left( \frac{\ln M_X(p) - \mu p}{p^2} \right) \rightarrow$  *we need to use L'Hôpital's rule*

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*We let  $p \rightarrow 0$*

- $\lim_{p \rightarrow 0} \frac{t^2}{\sigma^2} \left( \frac{\ln M_X(p) - \mu p}{p^2} \right) = \lim_{p \rightarrow 0} \frac{t^2}{\sigma^2} \left( \frac{\frac{1}{M_X(p)} \frac{dM_X(p)}{dp} - \mu}{2p} \right) \rightarrow \textbf{L'Hôpital's rule again}$
- $= \lim_{p \rightarrow 0} \frac{t^2}{\sigma^2} \left( \frac{\frac{M'_X(p)}{M_X(p)} - \mu}{2p} \right) = \lim_{p \rightarrow 0} \frac{t^2}{\sigma^2} \left( \frac{\frac{M''_X(p) * M_X(p) - M'_X(p) * M'_X(p)}{(M_X(p))^2}}{2} \right)$
- $M_X(0) = 1, M'_X(0) = E[X] = \mu, M''_X(0) = E[X^2]$
- $= \frac{t^2}{\sigma^2} \frac{(E[X^2] - E^2[X])}{2} = \frac{t^2}{2}$





- $\ln M_Y(t) \rightarrow \frac{t^2}{2}$
- so  $M_Y(t) \rightarrow e^{\frac{t^2}{2}}$ , as  $n \rightarrow \infty$

➤ *Proof finished*



*Thanks for listening*