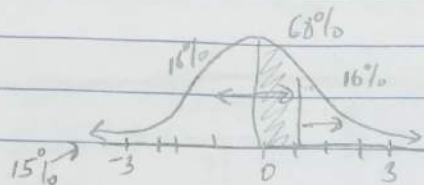


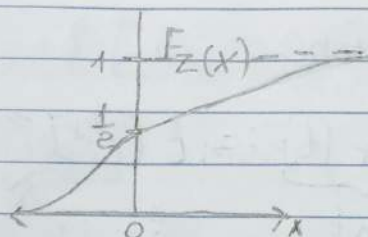
$$Z \sim N(0,1) := \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$$

$$Z = +1$$

$$Z = -1.5$$



$$X \sim N(\mu, \sigma^2) := \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$



$$X \sim N(70'', 3''^2)$$

$$P(\text{some US}^{\text{male}} \text{ is taller than } 73'') = P(X > 73'') = P\left(\frac{X-70}{3''} > \frac{73-70}{3''}\right)$$

← standard normal

$$= P(Z > 1)$$

Parameter space

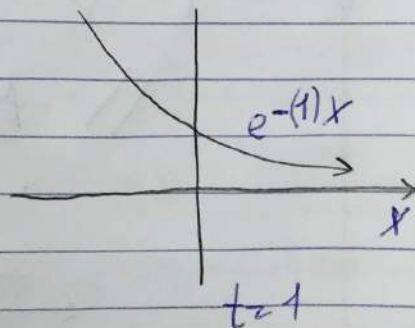
$$\mu \in \mathbb{R}$$

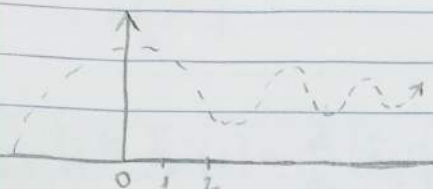
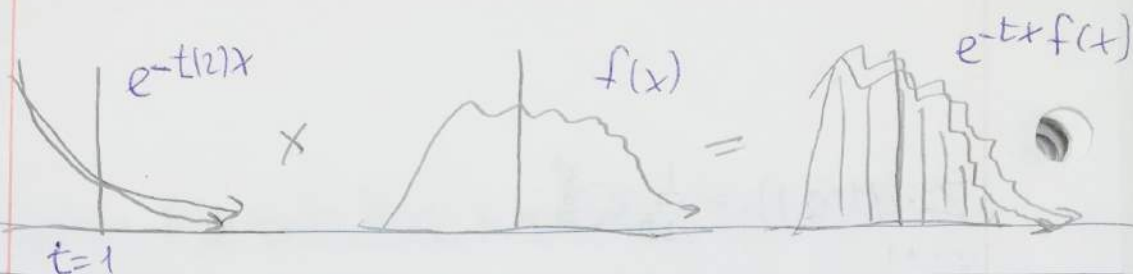
$$\sigma^2 \in (0, \infty)$$

$$L(t) := \int_{\mathbb{R}} e^{-tx} f(x) dx$$

\mathbb{R}

Biland
laplace
transform





$L(t)$ and $f(x)$ are 1:1

Define

$$L(-t) = \underbrace{M_X(t)}_{\text{moment generating function (mgt)}} := E[e^{tx}] = \int_{\mathbb{R}} e^{tx} f(x) dx$$

continuous

$$\Rightarrow \sum_{x \in \text{supp}(X)} e^{tx} p(x)$$

discrete

$$X \stackrel{d}{=} Y$$

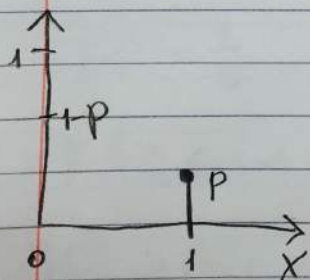
continuous case

discrete

$$f_X(x) = f_Y(y)$$

$$M_X(t) = M_Y(t)$$

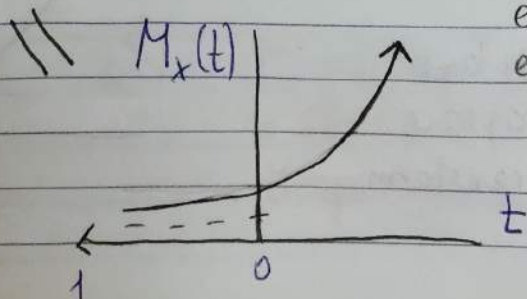
$$p_X(x) = p_Y(y)$$



$$M_X(t) = E[e^{tx}] = \sum_{x=0}^{\infty} e^{tx} p^x (1-p)^{1-x} =$$

$$= 1-p + e^t p$$

$$e^{t(0)} p^0 (1-p)^{1-0} + e^{t(1)} p^1 (1-p)^{1-1}$$



geometric

$$X \sim \text{Binom}(n, p) \quad E[X], E[X^2], E[X^3]$$

not possible

$$f(x) \approx f(c) + \frac{f'(c)}{1!} (x-c) + \frac{f''(c)}{2!} (x-c)^2 + \dots = \sum_{i=0}^{\infty} \frac{f^{(i)}(c)}{i!} (x-c)^i$$

$$\approx f(0) + \frac{f'(0)}{1!} x + \frac{f''(0)}{2!} x^2$$

$$e^x \approx 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!}$$

$$\bullet L(-t) = M_X(t) := E[e^{tx}] = E\left[1 + tx + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots\right]$$

$$= 1 + tE[X] + \frac{t^2}{2!} E[X^2] + \frac{t^3}{3!} E[X^3] + \dots$$

$$M'_X(t) = E[X] + tE[X^2] + \frac{t^2}{2!} E[X^3] + \frac{t^3}{3!} E[X^4] + \dots$$

$$m'_X(0) = E[X]$$

$$m''_X(t) = E[X^2] + tE[X^3] + \frac{t^2}{2!} E[X^4] + \dots$$

$$\textcircled{\text{I}} \quad m_X^k(0) = E[X^k]$$

$$T = X_1 + X_2 \quad \text{if } X_1, X_2 \text{ independent}$$

$$M_T(t) = M_{X_1+X_2}(t) = E[e^{t(X_1+X_2)}] = E[e^{tX_1} e^{tX_2}] =$$

$$= E[e^{tX_1}] E[e^{tX_2}] = M_{X_1}(t) \cdot M_{X_2}(t)$$

II

$$M_T(t) = \prod_{i=1}^n m_{X_i}(t) = (M_X(t))^n$$

↑ by independence

↑ if identically distributed

III

$$Y = aX + c$$

$$M_Y(t) = E[e^{t(aX+c)}] = E[e^{taX} e^c] = e^c E[e^{taX}] = e^c M_X(t') = e^c M_X(at)$$

- $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Bern}(p)$

$$M_T(t) = (M_X(t))^n = (1-p + e^{tp})^n$$

- $T \sim \text{Binom}(n, p)$

$$E(e^{tT}) = \sum_{x=0}^n \binom{n}{x} p^x (1-p)^{n-x} (e^t)^x = \sum_{x=0}^n \binom{n}{x} (e^t p)^x (1-p)^{n-x} = (1-p + e^t p)^n$$

(Binomial is iid Bernoullis)

- $X \sim \text{Exp}(\lambda)$

$$M_X(t) = E(e^{tX}) = \int_0^\infty e^{tx} \lambda e^{-\lambda x} dx = \lambda \int_0^\infty e^{(t-\lambda)x} dx = \frac{\lambda}{t-\lambda} [e^{(t-\lambda)x}]_0^\infty = \frac{\lambda}{t-\lambda} \left(\lim_{x \rightarrow \infty} e^{(t-\lambda)x} - 1 \right) = \frac{\lambda}{\lambda-t} \quad t < \lambda$$

* $Y = aX$

$$M_Y(t) = M_X(at) = \frac{\lambda}{\lambda - at} \left(\frac{1/a}{1/a} \right) = \frac{\lambda}{\lambda - t} = \frac{\lambda'}{\lambda' - t} \Rightarrow Y \sim \text{Exp}(\lambda') = \text{Exp}\left(\frac{\lambda}{a}\right)$$

$$(x-t)^2 = x^2 - 2xt + t^2$$

$$e^{-\frac{1}{2}(x-t)^2 - \frac{t^2}{2}} = e^{-\frac{1}{2}(x-t)^2} \cdot e^{-\frac{t^2}{2}}$$

$$e^{-\frac{1}{2}(x^2 - 2tx)}$$

$$Z \sim N(0,1) \quad \text{PDF for } Z$$

$$f(t) = \int_{\mathbb{R}} e^{tx} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{tx - \frac{x^2}{2}} dx = e^{\frac{t^2}{2}} \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-t)^2}{2}} dx$$

density
support 0 = 1

$$= \boxed{e^{\frac{t^2}{2}}}$$

$f(x)$ for $N(t, t^2)$

$$E(Z) = 0$$

$$M'_Z(0) = \left[t e^{\frac{t^2}{2}} \right]_0 = 0$$

$$\text{Var}[Z] = 1$$

$$E[Z^2] = 1$$

$$M''_Z(0) = \left[t^2 e^{\frac{t^2}{2}} + e^{\frac{t^2}{2}} \right]_0 = 1$$

• $X_1, \dots, X_n \stackrel{\text{iid}}{\sim}$ some distribution with $M, \sigma^2 < \infty$

$$\bar{X} = \frac{X_1 + \dots + X_n}{n}, E(\bar{X}) = M, SE[\bar{X}] = \frac{\sigma}{\sqrt{n}}$$

Goal: find distr. of C_n if n is large

$$C_n := \frac{\bar{X} - M}{\frac{\sigma}{\sqrt{n}}} = \frac{\frac{X_1 + \dots + X_n}{n} - M}{\frac{\sigma}{\sqrt{n}}} = \frac{(X_1 - M) + \dots + (X_n - M)}{\sigma \sqrt{n}}$$

$$\frac{\sigma}{\sqrt{n}} = \frac{1}{\sqrt{n}} \left(\frac{(X_1 - M)}{\sigma} + \dots + \frac{(X_n - M)}{\sigma} \right)$$

$$E(C_n) = 0$$

$$E(C_n) = 1$$

$$\frac{1}{\sqrt{n}} (Z_1 + \dots + Z_n) = \frac{Z_1}{\sqrt{n}} + \dots + \frac{Z_n}{\sqrt{n}}$$