

Ex. $X \sim \text{Exp}(\lambda)$

$$f(x) = \lambda e^{-\lambda x} \text{ for } x > 0$$

$$\mu = EX = \frac{1}{\lambda} \text{ and } E[X^2] = \frac{2}{\lambda^2}$$

$$Eg(X) = \int g(x)f(x)dx$$

$$\text{Var}(X) = E[(X - \mu)^2] = \int_{\mathbb{R}} (x - \mu)^2 f(x) dx$$

↓

correct - but labor intensive

$$V_1 = E[X^2 - 2\mu X + \mu^2]$$

$$= E[X^2] - 2\mu EX + \mu^2$$

$$= \frac{2}{\lambda^2} - 2\left(\frac{1}{\lambda}\right)\left(\frac{1}{\lambda}\right) + \left(\frac{1}{\lambda}\right)^2$$

$$= \frac{2}{\lambda^2} - \frac{2}{\lambda^2} + \frac{1}{\lambda^2} = \boxed{\frac{1}{\lambda^2} = \text{Var}(X)}$$

$$E[aX + b] \quad (*)$$

$$= aEX + b$$

Theorem: Short-Cut Formula for Variance

$$\text{Var}(X) = E[X^2] - E[X]^2$$

= exp. of sq - sq. of exp.

Pf.

$$\text{Var}(X) = \mathbb{E}[(X - \mu)^2]$$

$$= \mathbb{E}[X^2 - 2\mu X + \mu^2]$$

$$= \mathbb{E}[X^2] - 2\mu \mathbb{E}X + \mu^2$$

$$= \mathbb{E}[X^2] - 2\mathbb{E}X \mathbb{E}X + (\mathbb{E}X)^2$$

$\mu = \mathbb{E}X$

$$= \mathbb{E}[X^2] - \mathbb{E}[X]^2$$

Revisit example

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

$$\mathbb{E}[X] = \frac{1}{\lambda} \text{ and } \mathbb{E}[X^2] = \frac{2}{\lambda^2}$$

$$\text{then } \text{Var}(X) = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \frac{1}{\lambda^2}$$

Ex. $X \sim \text{Bin}(n, p)$

$$\mathbb{E}[X] = np$$

$$\mathbb{E}[X^2] = np(np - p + 1)$$

$$\begin{aligned}
 \text{So } \text{Var}(X) &= E[X^2] - (EX)^2 \\
 &= np(np-p+1) - (np)^2 \\
 &= np(1-p)
 \end{aligned}$$

Standard deviation : $\text{sd}(X) = \sqrt{\text{Var}(X)}$

$$\text{sd}(X) = \sqrt{np(1-p)}$$

Theorem:

$$\text{Var}(aX+b) = a^2 \text{Var}(X)$$

two rules:

- ① multiply X by $a \rightarrow$ Variance gets multiplied by a^2
- ② ignore additive shifts b

pf.

$$\begin{aligned}
 \text{Var}(aX+b) &= E[(aX+b)^2] - (E[aX+b])^2 \\
 &= E[a^2X^2 + 2abX + b^2] - (aEX + b)^2
 \end{aligned}$$

$$= \mathbb{E}[a^2 X^2 + 2abX + b^2] - (a \mathbb{E}X + b)^2$$

$$= a^2 \mathbb{E}[X^2] + 2ab \mathbb{E}[X] + b^2 - (a^2 \mathbb{E}[X]^2 + 2ab \mathbb{E}[X] + b^2)$$

$$= a^2 (\mathbb{E}[X^2] - \mathbb{E}[X]^2)$$

$$= a^2 \text{Var}(X)$$

Defn: Moments of a RV

If r is a pos. integer we define the r^{th} moment of X to be

$$\mu_r = \mathbb{E}[X^r]$$

Ex.

$$\mu_1 = \mathbb{E}[X] = \mu, \mu_2 = \mathbb{E}[X^2], \mu_3 = \mathbb{E}[X^3], \dots$$

Defn: Moment Generating Function (MGF)

If X is a RV then the MGF of X is a function

$$M: \mathbb{R} \rightarrow \mathbb{R}$$

defined for $t \in \mathbb{R}$ as

$$M(t) = \mathbb{E}[e^{tX}].$$

$$M(t) = \mathbb{E}[e^{tx}]$$

For discrete

$$M(t) = \mathbb{E}[e^{tx}] = \sum_x e^{tx} f(x)$$

For continuous:

$$M(t) = \mathbb{E}[e^{tx}] = \int e^{tx} f(x) dx$$

Ex. $X \sim \text{Exp}(\lambda)$

$$f(x) = \lambda e^{-\lambda x} \text{ for } x > 0$$

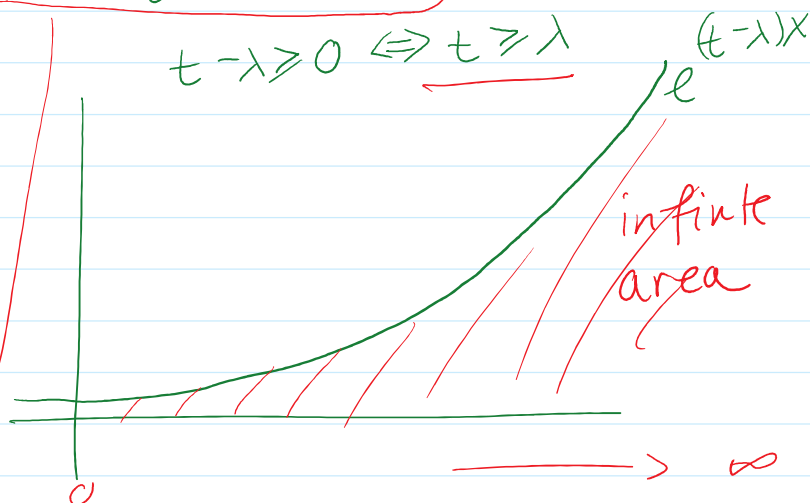
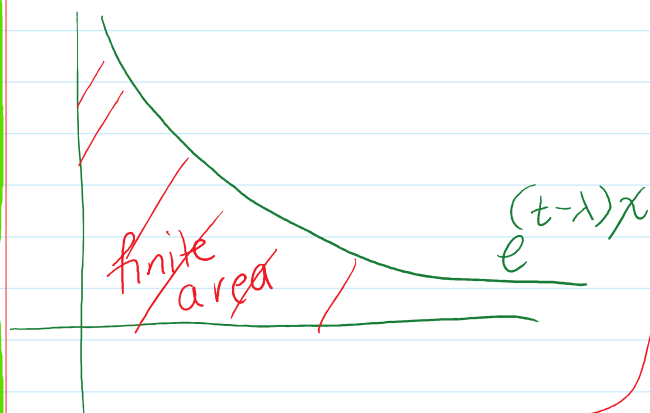
$$e^a e^b = e^{a+b}$$

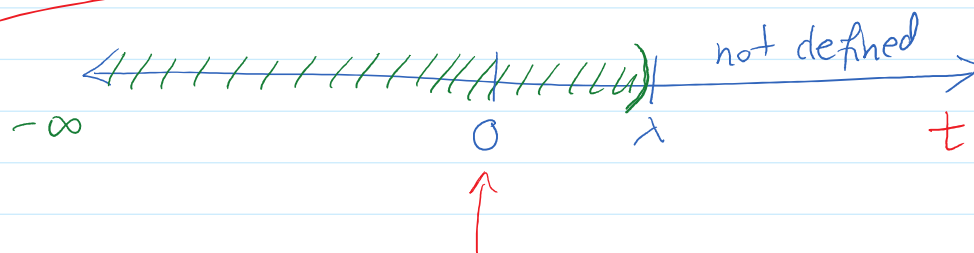
$$M(t) = \mathbb{E}[e^{tx}] = \int_{\mathbb{R}} e^{tx} f(x) dx = \int_0^{\infty} e^{tx} \lambda e^{-\lambda x} dx$$

$$= \lambda \int_0^{\infty} e^{(t-\lambda)x} dx$$

$$t - \lambda < 0 \Leftrightarrow t < \lambda$$

$$t - \lambda \geq 0 \Leftrightarrow t \geq \lambda$$





→ $t < \lambda$ then

$$M(t) = \lambda \int_0^{\infty} e^{(t-\lambda)x} dx = \lambda \frac{e^{(t-\lambda)x}}{t-\lambda} \Big|_0^{\infty}$$

$$= \frac{\lambda}{t-\lambda} (0 - 1) = \boxed{\frac{\lambda}{\lambda - t} \text{ for } t < \lambda}$$

Consider

$$\left. \frac{dM}{dt} \right|_{t=0} = \frac{\lambda}{(\lambda - t)^2} \Big|_{t=0} = \frac{\lambda}{\lambda^2} = \frac{1}{\lambda} = E[X]$$

$$\left. \frac{d^2 M}{dt^2} \right|_{t=0} = \frac{2\lambda}{(\lambda - t)^3} \Big|_{t=0} = \frac{2\lambda}{\lambda^3} = \frac{2}{\lambda^2} = E[X^2]$$

⋮

Theorem!

$d^r M /$

dt^r

$= E[X^r]$

$$\left. \frac{d^r M}{dt^r} \right|_{t=0} = M^{(r)}(0) = E[X^r] = \mu_r$$

pf. recall: $e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots$

$$e^{tx} = 1 + tx + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots$$

$$M(t) = E[e^{tx}] = 1 + t E[X] + \frac{t^2}{2!} E[X^2] + \frac{t^3}{3!} E[X^3] + \dots$$

$$\left. \frac{dM}{dt} \right|_{t=0} = E[X] + \frac{2t}{2!} E[X^2] + \frac{3t^2}{3!} E[X^3] + \dots$$

$$= E[X]$$

$$\left. \frac{d^2 M}{dt^2} \right|_{t=0} = E[X^2] + \frac{3 \cdot 2 \cdot t}{3!} E[X^3] + \dots$$

$$= E[X^2]$$

Ex. $X \sim \text{Bin}(n, p)$

$$E[X^2] = \sum_{x=0}^n x^2 \binom{n}{x} p^x (1-p)^{n-x}$$

$$x=0$$

Binomial theorem

$$(a+b)^n = \sum_{i=0}^n \binom{n}{i} a^i b^{n-i}$$

$$M(t) = \mathbb{E}[e^{tx}] = \sum_{x=0}^n \underbrace{e^{tx}}_{(e^t)^x} \binom{n}{x} \underbrace{p^x}_{p^x} (1-p)^{n-x}$$

$$= \sum_{x=0}^n \binom{n}{x} (pe^t)^x (1-p)^{n-x}$$

$$a = pe^t, \quad b = 1-p$$

$$= (a+b)^n = (pe^t + 1-p)^n = M(t)$$

$$\left. \frac{dM}{dt} \right|_{t=0} = \underbrace{n(pe^t + 1-p)^{n-1}}_{p+(1-p)=1} \underbrace{pe^t}_{\frac{1}{t}} = np = \mathbb{E}[X]$$

$$\left. \frac{d^2 M}{dt^2} \right|_{t=0} = n(n-1) \underbrace{(pe^t + 1-p)^{n-2}}_{1} \underbrace{pe^t}_{\frac{p}{t}} \underbrace{pe^t}_{\frac{p}{t}} + n \underbrace{(pe^t + 1-p)^{n-1}}_{1} \underbrace{pe^t}_{\frac{p}{t}}$$

$$= n(n-1)p^2 + np = \mathbb{E}[X^2]$$

Theorem: If $Y = aX + b$ then

$$M_Y(t) = e^{tb} M_X(at)$$

MGF of Y MGF of X

Prf.

$$\begin{aligned} M_Y(t) &= \mathbb{E}[e^{tY}] = \mathbb{E}[e^{t(ax+b)}] \\ &= \mathbb{E}[e^{(at)X} e^{tb}] \\ &= e^{tb} \underbrace{\mathbb{E}[e^{(at)X}]}_{M_X(at)} \end{aligned}$$