$$\underbrace{\xi_{X}}_{X} \times \sim \operatorname{Exp}(\lambda)$$

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

$$\mu = E_{X} = \frac{1}{\lambda} \quad ; \quad E_{X}^{2} = \frac{2}{\lambda^{2}}$$

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

$$= E[(X - \mu)^{2}] = \int (x - \mu)^{2} f(x) dx$$

$$\lim_{X \to \infty} \operatorname{Correct}_{X} \quad \text{but probably time intensive}$$

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

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

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

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

$$= \frac{2}{\lambda^2} - \frac{2}{\lambda^2} + \frac{1}{\lambda^2} = \sqrt{2} = \sqrt{2} = \sqrt{2}$$

Theorem: Short-Cut Formula For Variance $Var(X) = E[X^2] - E[X]^2$

 $F: Vav(X) = E[(X-\mu)^{2}]$ $= E[X^{2} - 2\mu X + \mu^{2}] \qquad \mu = EX$ $= E[X^{2}] - 2\mu E[X] + \mu^{2}$ $= E[X^{2}] - 2 E[X]^{2} + E[X]^{2}$ $E[X^{2}] - E[X]^{2}$

Ex. Revisit prev.

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

$$= \frac{2}{\lambda^2} - \frac{1}{\lambda^2} = \frac{1}{\lambda^2}$$

Meoreur:

$$Var(a \times + b) = a^2 Var(x)$$

- multiply X by a ~> var is multiplied by a2
- 2) ignore additive shifts

$$\underline{pf} \cdot Var(aX+b) = \mathbb{E}[(aX+b)^2] - (\mathbb{E}[aX+b])^2$$

$$\frac{\text{pf.}}{\text{Var}(a \times +b)} = \mathbb{E}[(a \times +b)^{2}] - (\mathbb{E}[a \times +b])^{2}$$

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

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

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

$$= a^{2} Var(x)$$

$$\begin{array}{l} \{\chi, \chi \sim \text{Bin}(n, p) \\ f(\chi) = {n \choose \chi} p^{\chi} (1-p)^{n-\chi} \\ E[\chi] = np \\ \text{turns out that } E[\chi^2] = np(np-p+1) \\ \text{So } \text{Var}(\chi) = E[\chi^2] - (E\chi)^2 \\ = np(np-p+1) - (np)^2 \\ = np(1-p) \\ \\ \underline{Sfandard deviation:} \quad Sd(\chi) = \sqrt{\text{Var}(\chi)} \end{array}$$

So
$$Sd(X) = \sqrt{np(1-p)}$$

Defn: Moments of a RV

If r is a pos. integer we define the

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

 $\underline{\mathcal{E}}_{X}$. $\underline{\mathcal{M}}_{1} = \mathbb{E}_{X}$, $\underline{\mathcal{M}}_{2} = \mathbb{E}_{X}^{2}$, $\underline{\mathcal{M}}_{3} = \mathbb{E}_{X}^{3}$

Defu: Moment Generating Function (MGF)

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

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

defined for tER

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

For discrete:

$$M(t) = Ee^{tX} = \sum_{x} e^{tx} f(x)$$

for continuos:

$$M(t) = Ee^{tx} = \int e^{tx} f(x) dx$$
 $Ex. \quad X \sim Exp(X)$

$$f(x) = \lambda e^{-\lambda x} \quad for \quad x \neq 0$$
 $M(t) = E[e^{tx}] = \int e^{tx} f(x) dx = \int e^{tx} \lambda e^{-\lambda x} dx$
 $\lambda \int e^{(t-\lambda)x} dx$
 $t = \lambda$
 $t = \lambda$

$$\begin{array}{c}
 \text{nbhd aroud} \\
 \text{Zero} \\
 \end{array}$$

$$\begin{array}{c}
 \text{M(t)} = \frac{\lambda}{\lambda - t} \\
 \text{M(t)}
\end{array}$$

Recall:
$$EX = \frac{1}{\lambda}$$
; $EX^2 = \frac{\lambda}{\lambda^2}$

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

Theorem:

If X is a RV w/ MGF M then

$$\frac{dM}{dt'} = M(0) = \mathbb{E}[X^{r}] = \mu_{r}$$

Pt. recall:
$$e^{\chi} = 1 + \chi + \frac{\chi^2}{z!} + \frac{\chi^3}{3!} + \frac{\chi^4}{4!} + \cdots$$

$$e^{\pm \chi} = 1 + \pm \chi + \pm \frac{\chi^2}{2!} + \pm \frac{\chi^3}{3!} + \frac{\chi^4}{4!} + \cdots$$

$$e^{2x} = 1 + tx + tx^{2} + tx^{3}x^{3} + \cdots$$

$$e^{\pm x} = 1 + \pm x + \pm \frac{2}{2!} + \pm \frac{3}{3!} + \cdots$$

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

$$\frac{\text{CIM}}{\text{at}} = \frac{2t}{2!} + \frac{3t^2}{2!} + \frac{3t^2}{3!} + \cdots$$

$$\frac{d^{2}M}{dt^{2}} = \mathbb{E}[X^{2}] + \frac{3 \cdot 2t}{3!} \mathbb{E}[X^{3}] + \cdots$$

$$\mathbb{E}_{X}$$
, $X \sim \mathbb{B}$ in (n,p) ,

$$\mathbb{E}\left[\chi^{2}\right] = \sum_{\chi=0}^{n} \chi^{2}(\chi) p^{\chi} (1-p)^{n-\chi}$$

Instead use MGF

Recall: binomial theorem

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

$$M(t) = \mathbb{E}[e^{tx}] = \sum_{x=0}^{n} e^{tx} f(x)$$

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

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

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

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

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

$$= np = \mathbb{E}x$$

$$\frac{dM}{dt} = n (n-1)(pe^{t} + 1-p)^{n-1} pe^{t} = n$$

$$= n (n-1)p^{2} + np$$

$$M_{\chi}(t) = e^{tb} M_{\chi}(at)$$

ats a b

Theorem: If X and Y are RVs and $M_X(t) = M_Y(t)$ for t in some neighborhood of zero

then $\chi \stackrel{d}{=} \%$.