

11/8/17

$$\text{Var}(X) = E((X - \mu)^2) = \sum_{x \in \text{supp}(X)} (x - \mu)^2 p(x)$$

$$\text{Var}(aX + c) = a^2 \sigma^2 \Rightarrow \text{SE}(aX + c) = |a| \sigma$$

$$E(X_1 + X_2) = E(X_1) + E(X_2)$$

$$\begin{aligned} \text{Var}(X_1 + X_2) &= E(((X_1 + X_2) - (\mu_1 + \mu_2))^2) \\ &= E(X_1^2 + X_2^2 + \mu_1^2 + \mu_2^2 + 2X_1X_2 - 2X_1\mu_1 - 2X_1\mu_2 - 2X_2\mu_1 - 2X_2\mu_2 + 2\mu_1\mu_2) \\ &= E(X_1^2) + E(X_2^2) + \mu_1^2 + \mu_2^2 + 2E(X_1X_2) - 2\mu_1E(X_1) - 2\mu_2E(X_1) \\ &\quad - 2\mu_1E(X_2) - 2\mu_2E(X_2) + 2\mu_1\mu_2 \\ &= \sigma_1^2 + \sigma_2^2 + 2(E(X_1X_2) - \mu_1\mu_2) \\ &\quad \text{Covariance}(X_1, X_2) \end{aligned}$$

Assume  $X_1, X_2$  independent

$$P(X_1, X_2) = P(X_1)P(X_2)$$

$$\begin{aligned} E(X_1X_2) &= \sum_{x_1 \in \text{supp}(X_1)} \sum_{x_2 \in \text{supp}(X_2)} x_1x_2 P(X_1, X_2) = \sum_{x_1} \sum_{x_2} x_1x_2 P(X_1)P(X_2) \\ &= \sum_{x_1} x_1 P(X_1) \sum_{x_2} x_2 P(X_2) \end{aligned}$$

$$\text{Var}(\sum X_i) = \sum \text{Var}(X_i) = n\sigma^2$$

$$\text{Var}(\bar{X}) = \text{Var}(\frac{1}{n} \sum X_i) = \frac{1}{n^2} \text{Var}(\sum X_i) = \frac{1}{n^2} \sum \text{Var}(X_i) = \frac{\sigma^2}{n}$$

$$\text{SE}(\bar{X}) = \frac{\sigma}{\sqrt{n}} \Rightarrow \bar{X} \rightarrow \mu \text{ (Law of Large Numbers)}$$

 $X \sim \text{Binomial}(n, p)$ 

$$\text{Var}(X) = \sum_{x=0}^n (x - np)^2 \binom{n}{x} p^x (1-p)^{n-x} = \dots = np(1-p)$$

$$\text{Var}(X) = \text{Var}(\sum X_i) = n\sigma^2 = np(1-p)$$

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



$$X \sim \text{Geometric}(p) = (1-p)^{x-1} p$$

$$\text{Supp}(X) = \mathbb{N}$$

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

$$E(X^2) = \sum_{x=1}^{\infty} x^2 (1-p)^{x-1} p = \sum_{y=0}^{\infty} (y+1)^2 (1-p)^y p = \sum_{y=0}^{\infty} y^2 (1-p)^y p$$

$$+ \sum_{y=0}^{\infty} 2y(1-p)^y + p \sum_{y=0}^{\infty} (1-p)^y$$

$$= (1-p) \sum_{y=1}^{\infty} y^2 (1-p)^{y-1} p + 2(1-p) \sum_{y=1}^{\infty} y(1-p)^{y-1} p + 1$$

$$\Rightarrow E(X^2) = (1-p) E(X^2) + \frac{2(1-p)}{p} + 1$$

$$\Rightarrow E(X^2) = (1-p) E(X^2) + \frac{2-p}{p}$$

$$E(X^2)(1-(1-p)) = \frac{2-p}{p}$$

$$E(X^2) = \frac{2-p}{p^2}$$

$$\text{Var}(X) = E(X^2) - \frac{1}{p^2} = \frac{2-p}{p^2} - \frac{1}{p^2} = \frac{1-p}{p^2}$$

$$P(X=7) = (1-p)^6 p$$

$$P(X=17) = (1-p)^{16} p$$

$$P(X=17 | X > 10) = P(X=7) = \frac{P(X=17 \cap X > 10)}{P(X > 10)} = \frac{P(X=17)}{P(X > 10)} = \frac{(1-p)^{16} p}{1 - F(10)}$$

$$P(X=a) = P(X=a+b | X > b) = \frac{P(X=a+b)}{P(X > b)} = \frac{(1-p)^{a+b-1} p}{(1-p)^b} = (1-p)^{a-1} p = P(X=a)$$

Memoryless property

$$T \sim \text{Geometric}(p) = (1-p)^{t-1} p$$

Every sec we do  $n$  iid Bern( $p$ )'s

IF  $n=10$ ,  $\text{Supp}(t) = \{0.1, 0.2, \dots\}$

For arbitrary  $n$ ,  $p(t) = (1-p)^{nt-1} p$ ,  $\text{Supp}(t) = \{\frac{1}{n}, \frac{2}{n}, \dots\}$

$$\text{Let } \lambda = np \Rightarrow p = \frac{\lambda}{n}$$

$$P(t) = (1 - \frac{\lambda}{n})^{nt-1} \cdot \frac{\lambda}{n}$$

$$\lim_{n \rightarrow \infty} p(t) = \lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^{nt-1} \cdot \lim_{n \rightarrow \infty} \frac{\lambda}{n} = 0 \quad \forall t$$

Limiting CDF

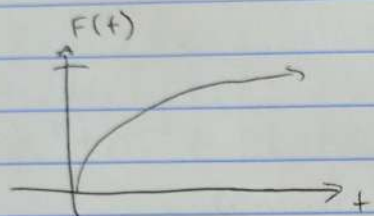
$$\lim_{n \rightarrow \infty} 1 - \left(1 - \frac{\lambda}{n}\right)^{nt} = 1 - \left(\lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^n\right)^t = 1 - e^{-\lambda t}$$

$$\lim_{n \rightarrow \infty} \left(1 + \frac{1}{n}\right)^n = e$$

$$\lim_{n \rightarrow \infty} \left(1 + \frac{a}{n}\right)^n = \left(\lim_{c \rightarrow \infty} \left(1 + \frac{1}{c}\right)^c\right)^a = e^a$$

$$\text{Let } c = \frac{n}{a} \Rightarrow n = ac$$

$$n \rightarrow \infty \Rightarrow c \rightarrow \infty$$



$$F(0) = 0$$

$$\lim_{t \rightarrow \infty} F(t) = 1$$

$$F'(t) = \lambda e^{-\lambda t} = \frac{\lambda}{e^{\lambda t}} > 0$$

$\Rightarrow F(t)$  is monotonically increasing