

lecture Math 241 15

Bet on #7 \$1

$$X_7 \sim \begin{cases} \$3.5 & \text{wp } 1/38 \\ \$-1 & \text{wp } 37/38 \end{cases} \Rightarrow M = -\$0.053$$

$$M = E(X) = \sum_{x \in \text{supp}(X)} x p(x)$$

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

$$= (3.5 - 0.053)^2 \frac{1}{38} + (-1 - -0.053)^2 \frac{37}{38}$$

$$= 33.207 \2$

$$= (1 - -0.053)^2 \left(\frac{18}{38}\right) + (-1 - -0.053)^2 \left(\frac{20}{38}\right)$$

$$= 0.997 \2$

$$\lim_{n \rightarrow \infty} \frac{1}{n} = 0$$

$$\lim_{n \rightarrow \infty} \frac{1}{e^n} = 0$$

Bet on Black \$1

$$X \sim \begin{cases} \$1 & \text{wp } 18/38 \\ \$-1 & \text{wp } 20/38 \end{cases} \Rightarrow M = -\$0.053$$

$$\bar{X}_7 \rightarrow M$$

$$\bar{X}_8 \rightarrow M$$

Law of Large #s

$$\bar{X}_7 \rightarrow M \text{ slower than } \bar{X}_8 \rightarrow M$$

$\2 is not a comprehensible unit so

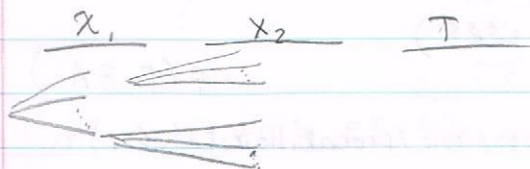
$$\sigma = \sqrt{33.207 \$^2} = \$5.79$$

$$\sigma = \sqrt{0.997 \$^2} = \$1.08$$

Standard Deviation or Standard error

$$\sigma = E[X] = \text{sp}[X] := \sqrt{\text{var}[X]}$$

$$T_2 = X_1 + X_2 \quad E[T_2] = \sum_{t \in \text{supp}(T_2)} t p(t) \text{ impractical}$$



* Don't need to know the following proof

$$E[g(x)] = \sum_{x \in \text{supp}(x)} g(x) p(x)$$

$$E[g(x_1, x_2)] = \sum_{x_1 \in \text{supp}(x_1)} \sum_{x_2 \in \text{supp}(x_2)} g(x_1, x_2) \overset{\text{joint mass function}}{p(x_1, x_2)}$$

$$E[x_1 + x_2] = \sum_{x_1} \sum_{x_2} (x_1 + x_2) p(x_1, x_2) = \sum_{x_1} \sum_{x_2} x_1 p(x_1, x_2) + \sum_{x_2} \sum_{x_1} x_2 p(x_1, x_2)$$

$$= \sum_{x_1} x_1 \sum_{x_2} p(x_1, x_2) + \sum_{x_2} x_2 \sum_{x_1} p(x_1, x_2)$$

x_1, x_2 independent

$$\Rightarrow p(x_1, x_2) = p(x_1) p(x_2)$$

$$E[x_1 + x_2] = \sum_{x_1} x_1 \sum_{x_2} p(x_1) p(x_2) + \sum_{x_2} x_2 \sum_{x_1} p(x_1) p(x_2)$$

$$= \underbrace{\sum_{x_1} x_1 p(x_1)}_{E[x_1]} \underbrace{\sum_{x_2} p(x_2)}_1 + \sum_{x_2} x_2 p(x_2) \underbrace{\sum_{x_1} p(x_1)}_1$$

$$= E[x_1] + E[x_2]$$

$$\text{supp } p(x_1) = \{1, 7, 9\}$$

$$\text{supp } p(x_2) = \{5, 23, 88\}$$

		x_1			
		1	7	9	
	5	$1/15$	$1/3$	$2/15$	$4/30$
x_2	23	$1/30$	$1/10$	$1/30$	$5/30$
	88	$1/30$	$1/5$	$1/15$	$9/30$

$p(x_1, x_2)$

$$P(B) = P(B \& A_1)$$

$$+ P(B \& A_2)$$

$$+ P(B \& A_3)$$

$$+ P(B \& A_4)$$

why?

All possible probabilities of (x_1)

$$\sum_{x_1} \sum_{x_2} p(x_1, x_2) = 1$$

$$P(x_1=1) = \begin{cases} 1 & \text{wp } 4/30 \\ 7 & \text{wp } 14/30 \\ 9 & \text{wp } 7/30 \end{cases}$$

x_1, x_2 independent? No

$$p(x_1=1, x_2=5) = p(x_1=1) p(x_2=5)$$

$$\frac{1}{15} \neq \frac{4}{30} \cdot \frac{16}{30}$$

$$\sum_{x_1} \text{marging out } x_1 \quad p(x_1, x_2) = p(x_2)$$

$$\int_{\mathbb{R}} f(x) dx = 1 \quad \int_{\mathbb{R}} f(x, y) dy = g(x)$$

$$\rightarrow \sum_x x_1 p(x_1) + \sum_{x_2} x_2 p(x_2) = E[x_1] + E[x_2]$$

under 'identically distributed assumption'

$$E[T_n] = \sum_{i=1}^n E(x_i)$$

$$E[\bar{x}_n] = E\left[\frac{1}{n} T_n\right] = \frac{1}{n} E[T_n] = \frac{1}{n} nm = \boxed{\mu} \quad *$$

$$X \sim \text{Hyper}(n, h, N)$$

$$E[X] = \sum_{x \in \text{supp}(X)} x \frac{\binom{n}{x} \binom{N-n}{n-x}}{\binom{N}{n}}$$

$$X = x_1 + x_2 + \dots + x_n$$

$$x_1 \sim \text{Bern}\left(\frac{n}{N}\right)$$

$$x_2 \sim \text{Bern}\left(\frac{n}{N}\right)$$

$$x_n \sim \text{Bern}\left(\frac{n}{N}\right)$$

$$x_2 | x_1 = 0 \sim \text{Bern}\left(\frac{n-1}{N-1}\right)$$

$$E[X] = n \frac{n}{N}$$

dependent -
you get info,
things change

identically
distributed

but
NOT IDEP

$$\begin{aligned} \text{Var}[X] &= E[(X - \mu)^2] \\ &= E[X^2 - 2\mu X + \mu^2] \\ &= E[X^2] + E[-2\mu X] + E[\mu^2] \\ &= E[X^2] - 2\mu \underbrace{E[X]}_{\mu} + \mu^2 \end{aligned}$$

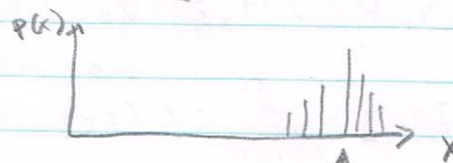
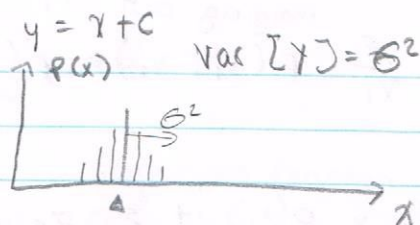
$$\boxed{\begin{aligned} \sigma^2 &= \text{Var}[X] = E[X^2] - \mu^2 \\ E[X^2] &= \sigma^2 + \mu^2 \end{aligned}}$$

$$\sum_x \mu^2 p(x) = \mu^2 \sum_x p(x) = \mu^2 \cdot 1 = \mu^2$$

$$Y = aX + c, a, c \in \mathbb{R}$$

$$E[Y] = aE[X] + c$$

$$\text{Var}[Y] =$$

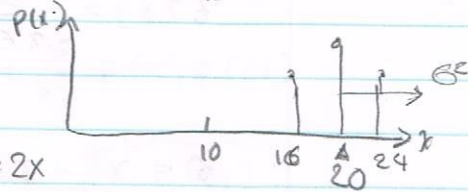
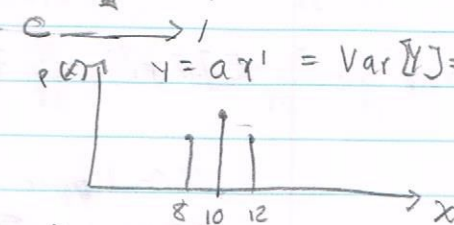


$$\text{Var}[X+c] = E[(X+c) - \underbrace{E[X+c]}_{m+c}]^2$$

$$= E[(X+c) - (m+c)]^2$$

$$= E[(X-m)^2]$$

$$= \sigma^2$$



$$y = 2x$$

$$\text{Var}[Y] = 4\sigma^2$$

$$\text{Var}(aX) = E\left((aX - \underbrace{E[aX]}_{am})^2\right)$$

$$= E[a^2(X-m)^2]$$

$$= E[a^2(X-m)^2] = a^2 E[(X-m)^2] = a^2 \sigma^2$$

$$Y = aX + c, a, c \in \mathbb{R}$$

$$E[Y] = aE[X] + c$$

$$\text{Var}[Y] = a^2 \sigma^2$$

$$S.E.[Y] = |a| \sigma$$