

Oct 30, 2017

Custom r.v's

Roulette in America

\$1 Bet on Black ^{Payout} is 1:1

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

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

$$= (\$1) \left(\frac{18}{38}\right) + (-\$1) \frac{20}{38} = \frac{1}{19} = -\$0.053$$

"the average will converge to -\\$0.053"

$\overline{X} \rightarrow M$
Lim of large #'s

$M \in \text{supp}[X]$
generally speaking.

$$\lim_{n \rightarrow \infty} T_n = \underbrace{X_1 + \dots + X_n}_{n(\bar{X})} = -\infty$$

\downarrow
 $\approx -\$0.053$

Bet on "Lucky 7" Payout 35:1

$$X \sim \begin{cases} \$35 \text{ (win)} & \text{wp } 1/38 \\ -\$1 & \text{wp } 37/38 \end{cases}$$

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

$$= 35 \cdot \frac{1}{38} + (-\$1) \frac{37}{38} = -\$0.053$$

Bet on "first dozen" 1-12 payout 2:1

$$X \sim \begin{cases} \$2 & \text{wp } 12/38 \\ -\$1 & \text{wp } 26/38 \end{cases}$$

$$E[X] = 2 \cdot \frac{12}{38} + (-1) \frac{26}{38} = -\$0.053$$

Roulette in Europe

\$1 Bet on Black Payout is 1:1

$$X \sim \begin{cases} \$1 & \text{wp } 18/37 \\ -\$1 & \text{wp } 19/37 \end{cases}$$

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

$$= (\$1) \frac{18}{37} + (-\$1) \frac{19}{37} = -\$0.027$$

Average

European roulette is "more ~~fair~~ fair" than American roulette.

"fair game" $E[X] = 0$.

$$P(\text{traffic}) = 0.3$$

if traffic, Uber takes 12 min

if no traffic, 7 min

Model the M car w.

$$W \sim \begin{cases} 12 \text{ min} & \text{wp } 0.3 \\ 7 \text{ min} & \text{wp } 0.7 \end{cases}$$

$$E[W] = 12(0.3) + 7(0.7) = 8.5 \text{ min}$$

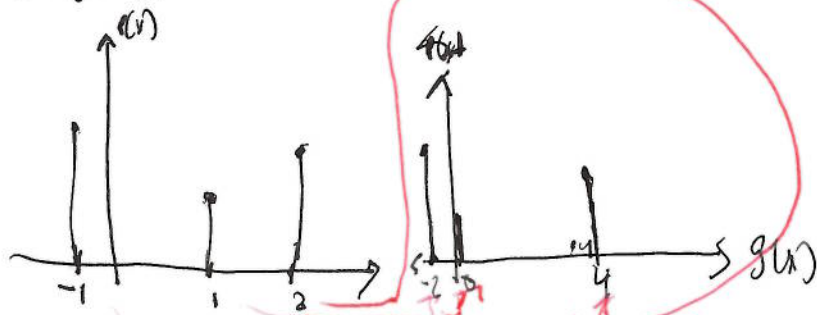
Uber charges \$.40/min Model B, the price paid for the time in taxi

$$B \sim \begin{cases} \$0.40 \times 12 = \$4.80 & \text{wp } 0.3 \\ \$0.40 \times 7 = \$2.80 & \text{wp } 0.7 \end{cases}$$

$$\begin{aligned} E[B] &= (4.80) 0.3 + (2.80) 0.7 = \$3.12 \\ &= \$0.40 \times 12 \times 0.3 + \$0.40 \times 7 \times 0.7 \\ &= \$0.40 (12 \times 0.3 + 7 \times 0.7) \\ &= \$0.40 E[W] \end{aligned}$$

$$Y = aX, a \in \mathbb{R} \quad E[Y] = E[aX] = E[g(X)] = \sum_{x \in \text{supp}(X)} a x p(x) = a \sum_{x \in \text{supp}(X)} x p(x) = a \cdot E[X]$$

$$E[g(X)]$$



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

Base Fare is \$3

Model T, the total price

$$T \sim \begin{cases} 3 + 4.80 = 7.80 & \text{wp } 0.3 \\ 3 + 2.80 = 5.80 & \text{wp } 0.7 \end{cases}$$

$$\begin{aligned} E[T] &= 7.80(0.3) + 5.80(0.7) = 6.12 \\ &= 3 + 4.80(0.3) + (3 + 2.80) \cdot 0.7 \\ &= 3 \cdot 0.3 + 4.80 \cdot 0.3 + 3 \cdot 0.7 + 2.80 \cdot 0.7 \\ &= 3(0.3 + 0.7) + 4.80 \cdot 0.3 + 2.80 \cdot 0.7 \\ &= 3 + E[B] \end{aligned}$$

$$Y = X + C \quad C \in \mathbb{R}$$

$$\begin{aligned} E[Y] &= E[X + C] = \sum_{x \in \text{Supp}(X)} (x + C) p(x) = \underbrace{\sum x p(x)}_{E[X]} + \sum C p(x) \\ &= E[X] + C \sum p(x) \\ &= E[X] + C \end{aligned}$$

Linear transformations
 $Y = aX + C, a, C \in \mathbb{R}$
 $E[Y] = aE[X] + C$

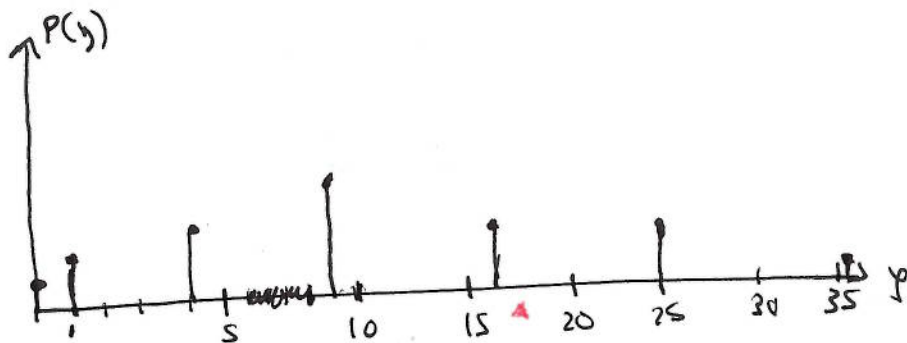
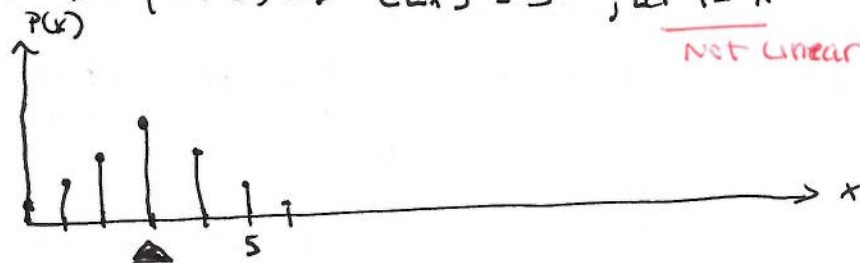
$$X \sim \text{Bin}(6, \frac{1}{2}) \Rightarrow E[X] = 3, \text{ let } Y = X^2$$

Not Linear

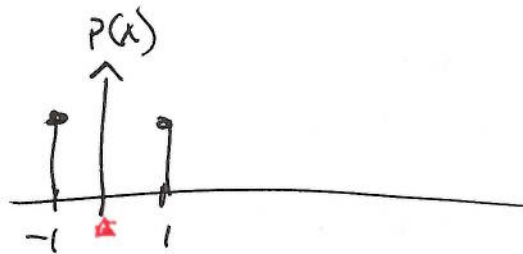
$$\text{Does } E[X] \neq (E[X])^2$$

$$\begin{aligned} E[Y] &= \sum_{x=0}^6 x^2 \binom{6}{x} p^x (1-p)^{6-x} \\ &= 17.5 \end{aligned}$$

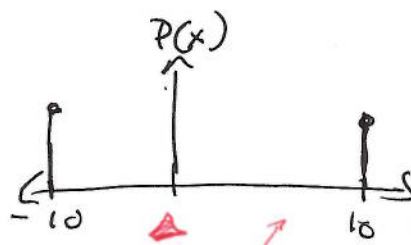
$$\Rightarrow E[g(x)] \neq g(E[X]) \text{ generally.}$$



$$X \sim \text{Ratenmacher} = \begin{cases} 1 & \text{wp } \frac{1}{2} \\ -1 & \text{wp } \frac{1}{2} \end{cases}$$



$$Y = 10X = \begin{cases} 10 & \text{wp } \frac{1}{2} \\ -10 & \text{wp } \frac{1}{2} \end{cases}$$



Y is more
a) spread out
b) deviant
c) variable

$$E[X] = 0$$

$$E[Y] = 10 E[X] = 0$$

$E[X]$ is a summary metric of r.v

$$E[X] \Rightarrow E[Y] \not\Rightarrow X \stackrel{d}{=} Y$$

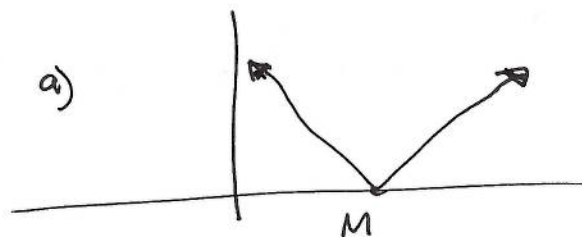
$$\int_{\mathbb{R}} g(x) dx = 3$$

$$\int_{\mathbb{R}} h(x) dx = 3$$

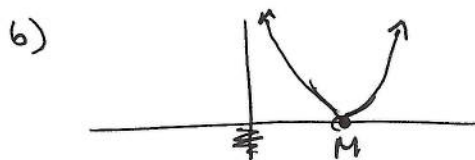
Theory of Error (Distance Functions)
measure error from x to M

$$a) e(x, M) = |x - M| \quad \text{L1 error metric}$$

you can minimize



$$b) e(x, M) = (x - M)^2 \quad \text{L2 error metric}$$



$$\text{Let } L = (X - M)^2$$

"Squared error loss"

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

$$\begin{aligned} \text{Var}[X] &= \sum_{x \in \text{supp}(X)} (x - M)^2 p(x) \\ &= (1 - 0)^2 \frac{1}{2} + (-1 - 0)^2 \frac{1}{2} \\ &= 1 \cdot \frac{1}{2} + 1 \cdot \frac{1}{2} = \boxed{1} \end{aligned}$$

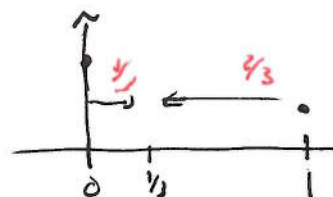
$$\begin{aligned} \text{Var}[Y] &= \sum_{y \in \text{supp}(Y)} (y - m_y)^2 p(y) \\ &= (10 - 0)^2 \frac{1}{2} + (-10 - 0)^2 \frac{1}{2} \\ &= 100 \cdot \frac{1}{2} + 100 \cdot \frac{1}{2} = \boxed{100} \end{aligned}$$

more spread out

$$X \sim \text{Bern}\left(\frac{1}{3}\right)$$

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

$$\begin{aligned} &= \sum_{x \in \text{supp}[X]} (x - \mu)^2 p(x) = \left(1 - \frac{1}{3}\right)^2 \frac{1}{3} + \left(0 - \frac{1}{3}\right)^2 \frac{2}{3} \\ &= \left(\frac{2}{3}\right)^2 \frac{1}{3} + \left(\frac{1}{3}\right)^2 \frac{2}{3} \\ &= \frac{4}{9} + \frac{2}{9} = \frac{2}{3} \end{aligned}$$



November 14

TEST

November 13

HW 5 Due

$$X \sim \text{Bern}(p)$$

Variance = "The average square distance from pivot"

$$\sigma^2 = \text{Var}[X] = \sum (x - \mu)^2 p(x)$$

$$= (1-p)^2 p + (0-p)^2 (1-p)$$

$$= (1 - 2p + p^2)p + p^2(1-p)$$

$$= p - 2p^2 + \cancel{p^3} + p^2 - \cancel{p^3} = p - p^2 = \boxed{p(1-p)}$$

$$\mu, \sigma^2$$

