

Lesson 15 Best 201 10/29/15

1

Talk about $E(X)$ what about $Y = g(X)$ some

I take Uber. Van Wyck: 7 mi, Jewel St: 12 mi
(no traffic) (if traffic)

$$P(\text{traffic}) = 0.7$$

$$W \sim \begin{cases} 7 \text{ min} & \text{up } 0.3 \\ 12 \text{ min} & \text{up } 0.7 \end{cases} \quad E(W) = 7 \cdot 0.3 + 12 \cdot 0.7 = 7.8 \text{ min}$$

14 percent

Uber charges \$0.40/min

$$B = \$0.40/\text{min} \cdot W = g(W)$$

$E(B)$? We need to figure out $E(W)$
if $Y = g(X)$ what is $E(Y)$? Assume Y is discrete random

$$\begin{aligned} E(Y) &:= \int_{\Omega} Y(\omega) dP(\omega) = \int_{\Omega} g(X(\omega)) dP(\omega) = \int_{\{\omega: X(\omega)=x_1\}} g(X(\omega)) dP(\omega) + \int_{\{\omega: X(\omega)=x_2\}} g(X(\omega)) dP(\omega) + \dots \\ &= g(x_1) \int dP(\omega) + g(x_2) \int dP(\omega) + \dots \end{aligned}$$

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

$$E(B) = E[0.70W] = \sum_{w \in \text{supp}(W)} 0.70w p(w) = 2.7 p(7) + 7.8 p(12) \\ = 2.7 \cdot 0.7 + 7.8 \cdot 0.3 \\ = \$3.10 \text{ i.e. } 310 \text{ cents}$$

let $g(x) = aX$ s.t. $a \in \mathbb{R}$ Note

$$E(aX) = \sum_{x \in \text{supp}(X)} ax p(x) = a \sum x p(x) = a E(X)$$

What is $P(B=b)$ i.e. the PMF of B ? Can't say this... sorry!

There is also a \$3 base fee so...

$T = \$3 + B$ What is $E(T)$?
 $\nwarrow g(w)$

$$E(T) = E(\$3 + B) = \sum_{w \in \text{supp}(W)} (\$3 + 0.70w) p(w) = 5.40 p(7) + 7.8 p(12) \\ = 5.4 \cdot 0.7 + 7.8 \cdot 0.3 \\ = \$6.12$$

let $g(x) = x + c$

$$E(g(x)) = E(x+c) = \sum (x+c) p(x) = \sum x p(x) + \sum c p(x) \\ = E(x) + c$$

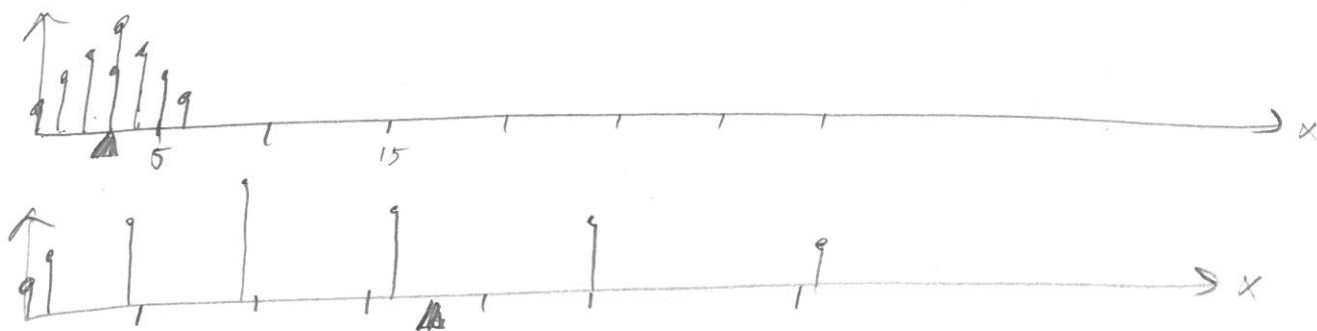
$$g(x) = aX + c \Rightarrow E(g(x)) = a E(X) + c$$

Let $X \sim \text{Bin}(6, \frac{1}{2})$

$$Y = X^2 = g(X)$$

$E(Y)$?

3



$0 \rightarrow 0$

$1 \rightarrow 1$

$2 \rightarrow 4$

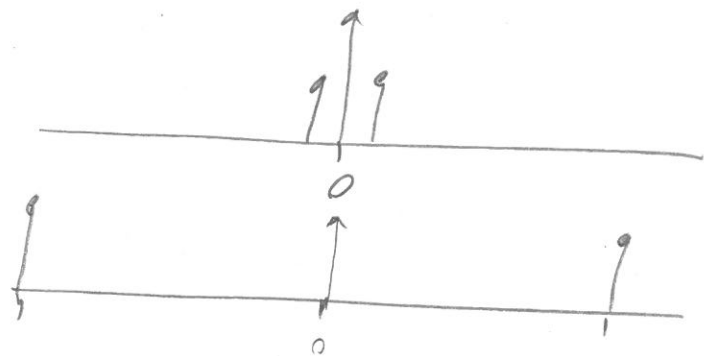
$6 \rightarrow 36$

$$Eg(X) = \sum_{x=0}^6 x^2 \binom{6}{x} \frac{1}{2^6} = 17.5 \text{ no easy way!}$$

$E(X) = 3$, $E(X^2) = 17.5$ any pattern? No...

only if $g(x)$ is linear...

$X \sim \text{Rademacher}$, $Y = 10X \sim \begin{cases} -10 & \text{w.p. } \frac{1}{2} \\ 10 & \text{w.p. } \frac{1}{2} \end{cases}$



$$E(X) = 0$$

$$E(Y) = E(10X) = 10E(X) = 10 \cdot 0 = 0$$

$$\text{obviously } E(X) = E(Y) \neq X = Y$$

Y is more "dispersed" than X around its pivot, 0.

Can we capture this concept of dispersion?

Need to define dispersed about the prior

penalty function

$$d(x, \mu) = x - \mu \quad \text{distance}$$

But we don't care if it's \oplus or \ominus so...

$$d(x, \mu) = |x - \mu| \quad \text{but abs value hard to work with } d'(x, \mu) ? \text{ piecewise...}$$

$$d(x, \mu) = (x - \mu)^2 \quad \text{"squared error loss" very natural... function always, the higher the fee.}$$

Any penalty function is arbitrary!

$$\text{Let } L := (X - \mu)^2 = g(X) \quad \text{a r.v. representing loss}$$

$E[L]$ is how far away from the prior in sqd distance on avg.

for Rademacher

$$L = (X - \mu)^2 = X^2 \sim \{1 \text{ w.p. } 1\} \Rightarrow E[L] = 1$$

for $10X$ (Rademacher $\times 10$)

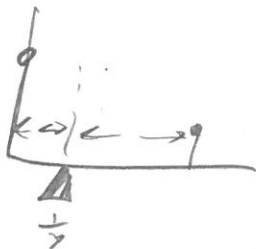
more dispersed!

$$L = (10X - \mu)^2 = (10X)^2 = 100X^2 \sim \{100 \text{ w.p. } 1\} \Rightarrow E[L] = 100$$

$$\text{ET } \sigma^2 = \text{Var}(X) :=$$

$$E[L] := E[(X - \mu)^2] \quad \text{and call it "variance" of r.v. } X$$

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



$\frac{1}{3}$ of the time you're $\frac{2}{3}$ away or $\frac{4}{9}$ sqd dist

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$$L \sim \begin{cases} \frac{4}{9} \text{ w.p. } \frac{2}{3} \\ \frac{1}{9} \text{ w.p. } \frac{1}{3} \end{cases} \quad \text{Var}[X] = \frac{1}{9} \cdot \frac{2}{3} + \frac{4}{9} \cdot \frac{1}{3} = \frac{7}{27} \approx 0.259$$

$$\text{Var}(X) = \sum_{x=0}^1 (x-\mu)^2 p(x) = (0-\frac{1}{3})^2 p(0) + (1-\frac{1}{3})^2 p(1) \\ = \frac{1}{9} \cdot \frac{2}{3} + \frac{4}{9} \cdot \frac{1}{3} = .251$$

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

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

Roulette: Bet on lucky #7

$$X \sim \begin{cases} \$35 \text{ up } \frac{1}{38} \\ -\$1 \text{ up } \frac{37}{38} \end{cases} \quad \mu = -0.053 \quad (\text{from 1st class})$$

$$\text{Var}(X) = E[(X-\mu)^2] = (35 - -0.053)^2 p(35) + (-1 - -0.053)^2 p(-1) \\ = 33.207 \text{ \2$

In arg... your winnings will be $33.207 \text{ \2 any from the center as measured by sqd distance.

$\text{\2 ? What is the interpretation?

$\text{SE}(X) :=$

Let $\sigma := \sqrt{\sigma^2} = \sqrt{\text{Var}(X)}$ the "standard error" or "standard deviation"

$$\text{Here } \text{SE}(X) = \$5.76$$

same unit as $\text{Exp}(X)$

Interpretation? None.

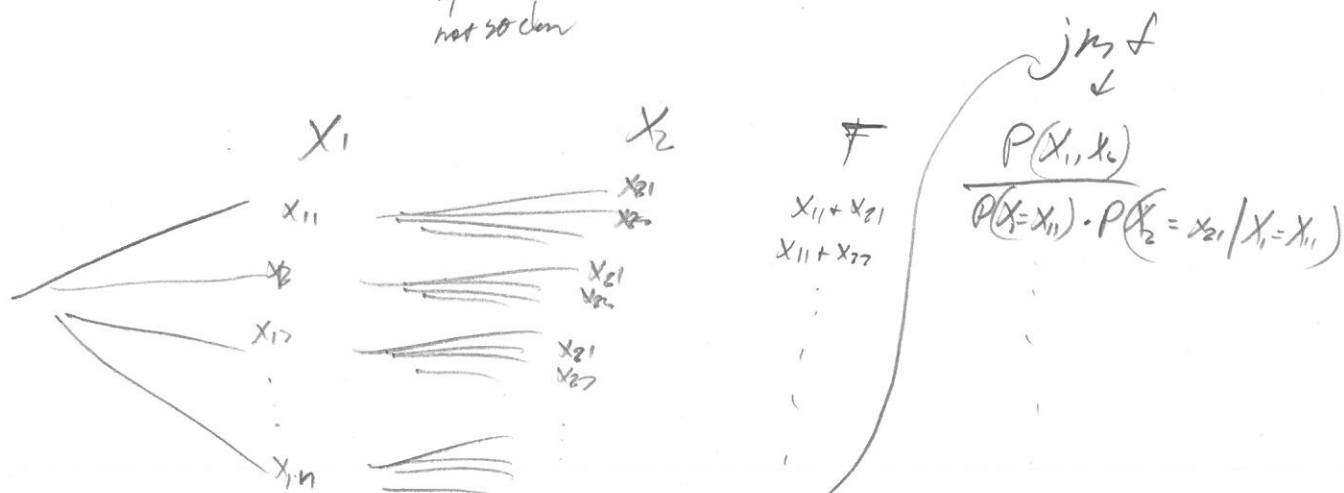
Synonym for variance

Try to measure spread... soon will see that it's useful...

Recall $T = X_1 + X_2$

What is $E(T)$? $= \sum_{t \in \text{supp}(T)} t p(t)$

not so clear



$T = g(X_1, X_2)$

$E(T) = E[g(X_1, X_2)] = \sum_{\langle x_1, x_2 \rangle \in \text{supp}(X_1) \times \text{supp}(X_2)} g(x_1, x_2) p(x_1, x_2)$

The proof is a generalization of the previous proof using a double integral

$\sum_{x_1 \in \text{supp}(X_1)} \sum_{x_2 \in \text{supp}(X_2)} g(x_1, x_2) p(x_1, x_2)$

Imagine if X_1, X_2 independent...

PMF for X_1
PMF for X_2

$p(x_1, x_2) = p_1(x_1) p_2(x_2)$

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

$= \underbrace{\sum_{x_1} x_1 p(x_1)}_{E(X_1)} \underbrace{\sum_{x_2} p(x_2)}_1 + \sum_{x_1} p(x_1) \underbrace{\sum_{x_2} x_2 p(x_2)}_{E(X_2)}$

$E(X_1 + X_2) = E(X_1) + E(X_2)$ if X_1, X_2 independent