

I take uber

Van Wget. = 7 min (no traffic)

Jewel Ave: 12 min (traffic)

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

Law of $\bar{X} \rightarrow \mu$
large #'s

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

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

average time in the car
8.5 if I take many, many
Ubers.. Meaning of 8.5

average - sample
Mean - Expectation

Expectation - Long Run average

Uber charges \$0.40 min

$$\underline{B} = \underbrace{\$0.40/\text{min}}_{g(w)} \times W$$

$$E[B] = E[g(w)] = \sum_{w \in \text{supp}(w)} g(w) \overbrace{P(W=w)}^{P(W=w)}$$

0.4×7 0.4×12

$$E[B] = g(7)p(7) + g(12)p(12) \\ = \$2.80 \times 0.7 + \$4.80 \times 0.3 = \$3.40$$

If you take many many Uber the
in the long run the average that
you spend in the bill will be \$3.40

$$Y := g(X)$$

$$E[Y] := \int_{\Omega} Y(\omega) dP(\omega)$$

X is discrete $\Rightarrow \text{supp}[X] = \{X_1, X_2, \dots\}$

$$\Omega = \int_{\Omega} g(X(\omega)) dP(\omega) = \int_{\{\omega: X(\omega)=X_1\}} g(X(\omega)) dP(\omega) + \int_{\{\omega: X(\omega)=X_2\}} g(X(\omega)) dP(\omega) + \dots$$

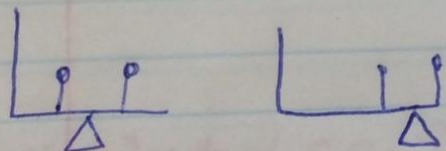
$$= g(X_1) \int_{\{\omega: X(\omega)=X_1\}} dP(\omega) + g(X_2) \int_{\{\omega: X(\omega)=X_2\}} dP(\omega) + \dots$$

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

$$E[B] = \sum_{b \in \text{supp}[B]} b p(b) \quad \leftarrow P(B=b)$$

* $Y = y(X) = aX, a \in \mathbb{R}$

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



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

* There is a \$3 base fee

$$T = \underbrace{\$3 + B}_{h(B)} = \$3 + \underbrace{\$0.40/m \cdot W}_{g(W)}$$

$$E[T] = g(7) p(7) + g(12) p(12)$$

$$= (3 + 0.4 \times 7) \cdot p(7) + (3 + 0.4 \times 12) p(12) = \$6.40$$

↑
0.7

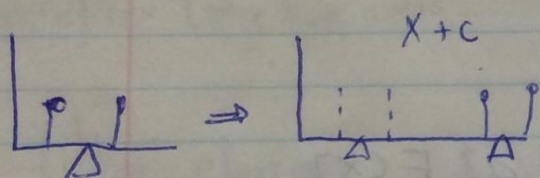
↑
0.3

over a long-term period
of time, that will be
the bill

• $Y = X + c$, $c \in \mathbb{R}$

$$E[\underbrace{X+c}_{g(x)}] = \sum g(x) p(x) = \sum (x+c) p(x) = \underbrace{\sum x p(x)}_{E[X]} + \underbrace{\sum c p(x)}_{c \sum p(x) = 1}$$

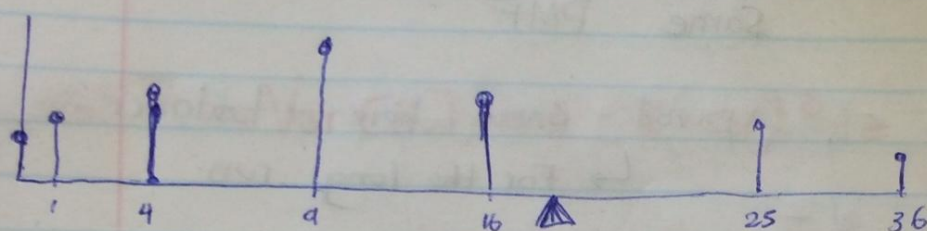
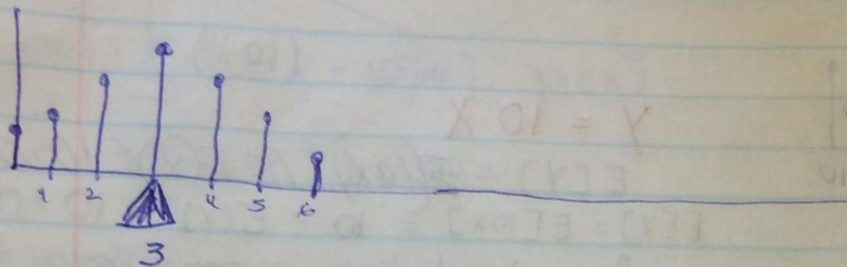
$$E[X] + c$$



Linear transformations

$$\boxed{E[aX + c] = a E[X] + c}$$

$$X \sim \text{Bin}\left(6, \frac{1}{2}\right) \quad Y = X^2 = g(X)$$



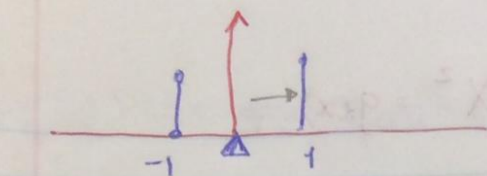
$$\text{Supp}[g(X)] = \{g(x) : x \in \text{Supp}[X]\}$$

$$E[X^2] = \sum_{x \in \text{Supp}[X]} x^2 p(x)$$

$$= \sum_{x=0}^6 x^2 \binom{6}{x} \frac{1}{2^6} = \frac{1}{2^6} \left(\binom{6}{1} + 4 \binom{6}{2} + 9 \binom{6}{3} + 16 \binom{6}{4} + 25 \binom{6}{5} + 36 \right)$$

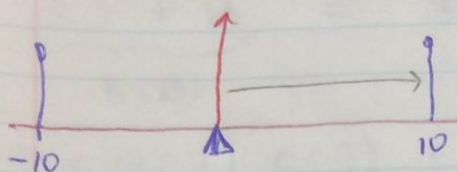
$$= 17.5$$

$$E[X^2] \neq (E[X])^2$$



$X \sim \text{Rademacher}$

$$E[X] = 0$$



$$Y = 10X$$

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

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

* equal in distribution means exact same PMF

$$E[X] = E[Y]$$

$$\Rightarrow X \neq Y$$

Why are they different?

→ Different distance from the pivot

pivot - know where it balances

↳ For the long run

* Split out → Support

Dispersion

Far from the pivot

Distance

Error

Far from where μ want to be

cost

Loss

Variance

Diviance

$$e(X, \mu) = X - \mu$$

$$L(X, \mu) = |X - \mu|$$

$$e(X, \mu) := (X - \mu)^2 \quad \text{right loss function}$$

quadratic loss,

square error loss

on average what's out square error loss

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

$$= \sum_{x \in \{-1, 1\}} x^2 \frac{1}{2} = 1$$

How far ± 1 m away on average

$$= \sum (10x - 10 \cancel{0})^2 p(x)$$

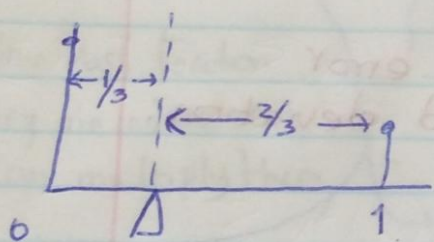
$$= 100 \sum x^2 p(x)$$

$$= 100$$

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

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

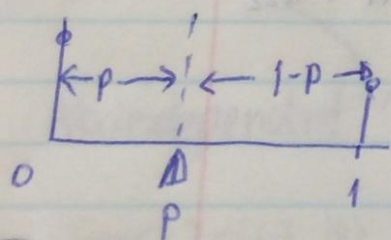
Variance of X



$$\begin{aligned} \sigma^2 &= \left(0 - \frac{1}{3}\right)^2 \frac{2}{3} + \left(1 - \frac{1}{3}\right)^2 \frac{1}{3} \\ &= \frac{1}{9} \cdot \frac{2}{3} + \frac{4}{9} \cdot \frac{1}{3} \end{aligned}$$

$$= .259$$

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



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

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

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

$$= \underline{p(1-p)}$$

Variance of any Bernoulli

- Bet on lucky # 7

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

$$E[X] = -\$0.053$$

$$\begin{aligned} \text{Var}[X] &= E[X - \mu]^2 \\ &= (\$35 - -\$0.053)^2 \frac{1}{38} + (-\$1 - -\$0.053)^2 \frac{37}{38} \\ &= 33.21 \$^2 \quad \text{— Distance away} \end{aligned}$$

$$SE[X] = \sqrt{\text{Var}[X]} = \$5.76$$

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

Standard error
Standard deviation.

- $T = \underbrace{X_1 + X_2}_{g(X_1, X_2)}$

$$E[T] = \sum_{t \in \text{supp}(T)} t p(t)$$

	X_1	X_2	T	$P(X_1, X_2)$
	X_{11}	X_{21}	$X_{11} + X_{21}$	$P(X_{11})P(X_{21} X_{11})$
	X_{12}	X_{22}	$X_{11} + X_{22}$	
	X_{13}			

$$E[g(X_1, X_2)] = \sum_{\substack{\langle X_1, X_2 \rangle \in \\ \text{Supp}[X_1] \times \\ \text{Supp}[X_2]}} g(X_1, X_2) P(X_1, X_2)$$

$$= \sum_{X_1 \in \text{Supp}[X_1]} \sum_{X_2 \in \text{Supp}[X_2]} g(X_1, X_2) P(X_1, X_2)$$

$$E[X_1 + X_2] = \sum_{X_1} \sum_{X_2} (X_1 + X_2) P(X_1, X_2)$$

$$= \sum_{X_1} \sum_{X_2} X_1 P(X_1, X_2) + \sum_{X_1} \sum_{X_2} X_2 P(X_1, X_2)$$

under no independence will be stuck here

If X_1, X_2 independent $\Rightarrow P(X_1, X_2) = P(X_1)P(X_2)$

Join Mass function
If they are independent
I can multiply them

$$= \sum_{X_1} \sum_{X_2} X_1 P(X_1) P(X_2) + \sum_{X_1} \sum_{X_2} X_2 P(X_1) P(X_2)$$

Independence

$$= \underbrace{\sum_{X_1} X_1 P(X_1)}_{E[X_1]} \underbrace{\sum_{X_2} P(X_2)}_1 + \underbrace{\sum_{X_1} P(X_1)}_1 \underbrace{\sum_{X_2} X_2 P(X_2)}_{E[X_2]}$$

No independent $\sim \sum_{X_1} X_1 \sum_{X_2} P(X_1, X_2) + \sum_{X_2} X_2 \sum_{X_1} P(X_1, X_2)$

$$\text{Supp}[X_1] = \{1, 7, 19\}$$

$$\text{Supp}[X_2] = \{5, 23, 88\}$$

	1	7	19
5	1/15	1/3	2/15
23	1/30	1/10	1/30
88	1/30	1/5	1/15