

* Probability \Rightarrow quantify uncertainty

* Same Representation for different problems

random variable X

X	0	1
$P(X)$	$\frac{1}{2}$	$\frac{1}{2}$

X	1	2	3	4	5	6
$P(X)$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$

* Look at an average value of the random variable

look at the probabilities to decide how much weight to put on every value of x

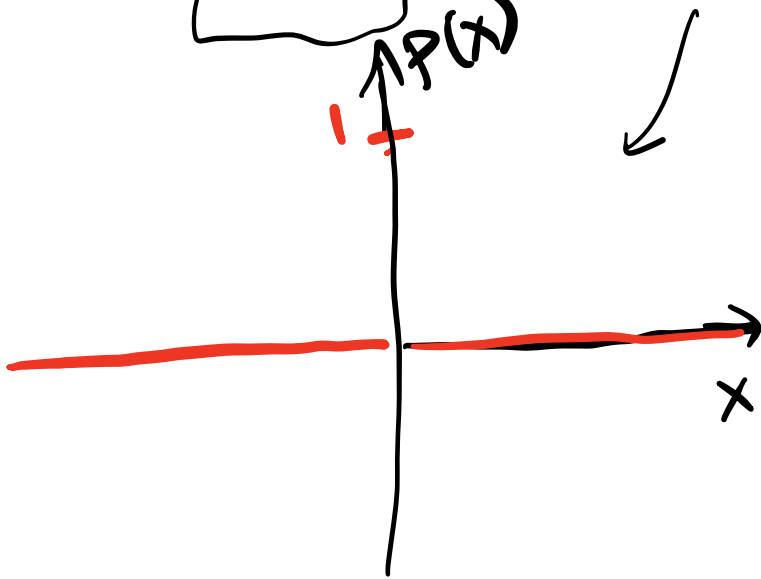
$$0 \cdot \frac{1}{2} + 1 \cdot \frac{1}{2} = \frac{1}{2}$$

$$1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + 3 \cdot \frac{1}{6} + 4 \cdot \frac{1}{6} + 5 \cdot \frac{1}{6} + 6 \cdot \frac{1}{6} = 3.5$$

$$E(X) = \sum_a a p(X=a) = \text{average value of } X \\ = \text{expected value of } X$$

X	2	3	4	5	6	7	8	9	10	11	12
$p(X)$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

$$E(X) = 7$$



$$E(X) = 0$$



$$E(X) = 0$$

\Rightarrow the two distributions don't have the same spread and $E(X)$ is not able to capture such information

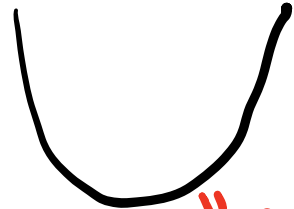
\Rightarrow Need another metric

50	60	60	50	→	55
			80	→	55
30	80	30			

$$(50 - 55) + (60 - 55) + (60 - 55) + (50 - 55) = 0$$

negative
not good

square ✓
absolute value



smoother



sharp

$$(50 - 55)^2 + (60 - 55)^2 + (60 - 55)^2 + (50 - 55)^2 = 100$$

$$(30 - 55)^2 + (80 - 55)^2 + (30 - 55)^2 + (80 - 55)^2 = 2500$$

variance

standard deviation

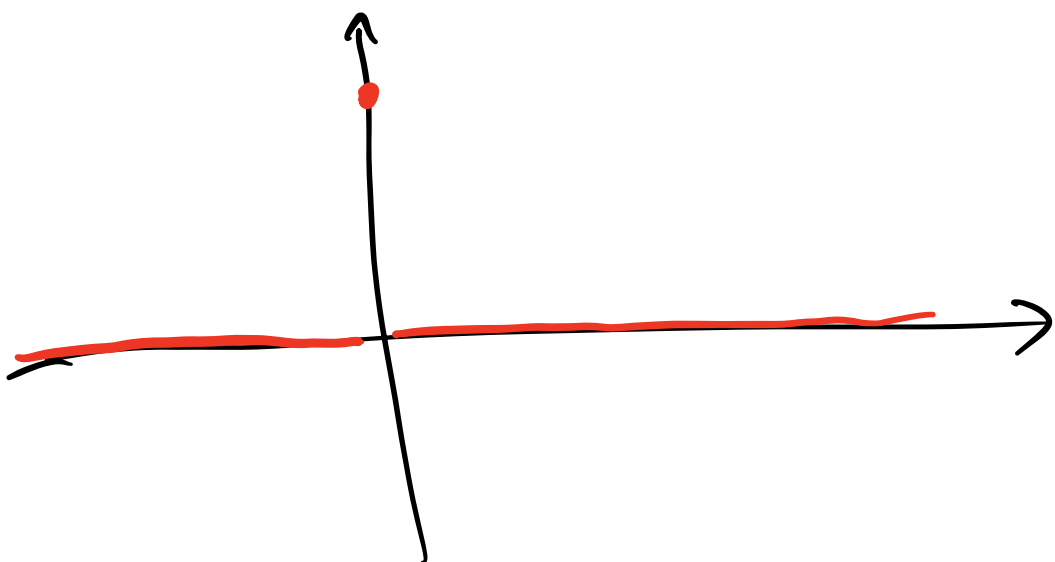
Let's take a square root

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

$$\sigma(X) = \sqrt{\text{Var}(X)}$$

give me intuition
about the
spread of the
square values

Standard deviation



$$E(X) = 0$$

X	0	1
$P(X)$	$\frac{1}{2}$	$\frac{1}{2}$

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

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

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

* 1 4 6 9 $E(X) = 5$

* 2 8 12 18 $E(2X) = 10 = 2E(X)$

$$E(aX) = aE(X)$$

more probability (trust me, will convince you later)

* $E(X_1 + X_2) = E(X_1) + E(X_2)$ (pinkie promise)
two random variables X_1 and X_2

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

(think of it as $a^2 - 2a + 1 = (a-1)^2$)

coin example

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

X	0	1
$P(X)$	$\frac{1}{2}$	$\frac{1}{2}$

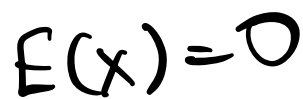
X^2	0	1
$P(X)$	$\frac{1}{2}$	$\frac{1}{2}$

$$E(X^2) = 0^2 \cdot \frac{1}{2} + 1^2 \cdot \frac{1}{2} = \frac{1}{2}$$

$$\text{Var}(X) = \frac{1}{2} - \left(\frac{1}{2}\right)^2 = \frac{1}{4}$$

$$\sigma(X) = \sqrt{\text{Var} X} = \frac{1}{2}$$

Var makes sense



$$p(x) = \begin{cases} 1 & \text{if } x=0 \\ 0 & \text{if } x \neq 0 \end{cases}$$

$$E(X^2) = 0^2 \cdot 1 = 0$$

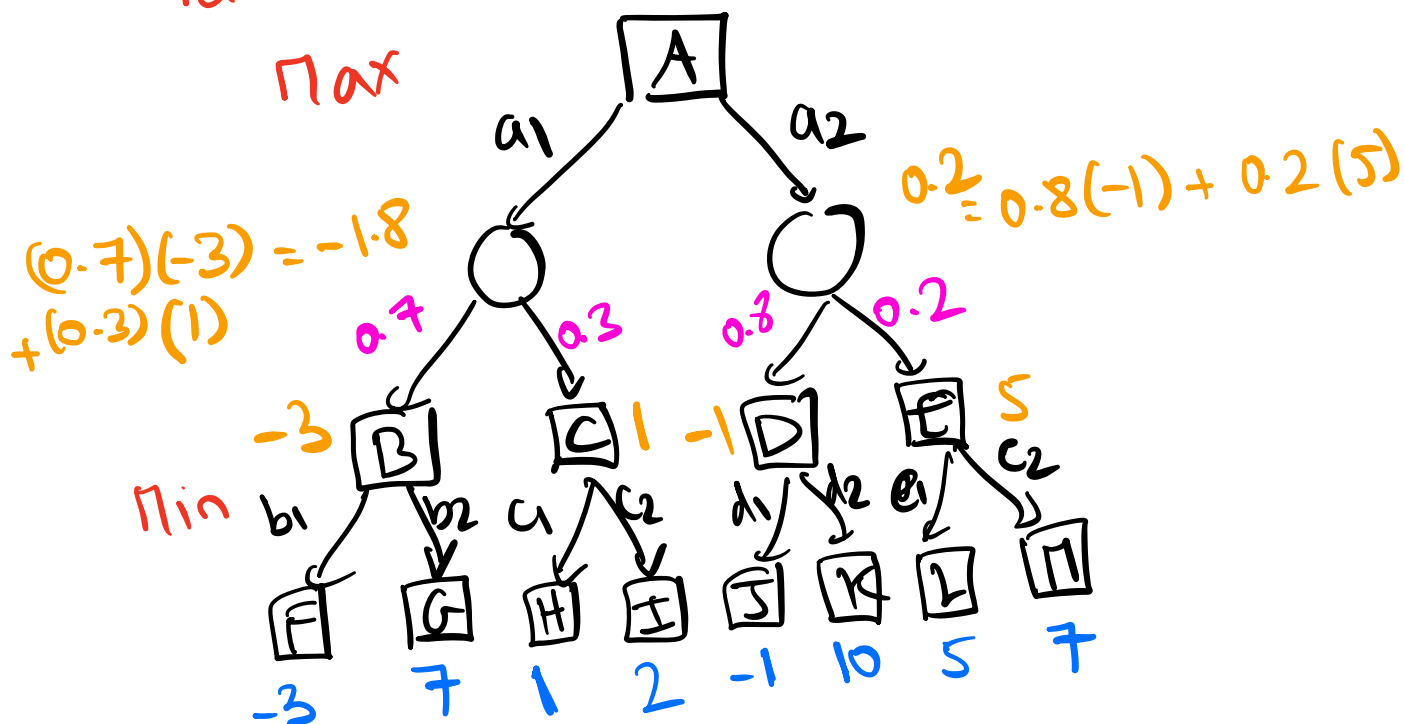
$$\text{Var}(X) = E(X^2) - (E(X))^2 = 0 - 0^2 = 0.$$

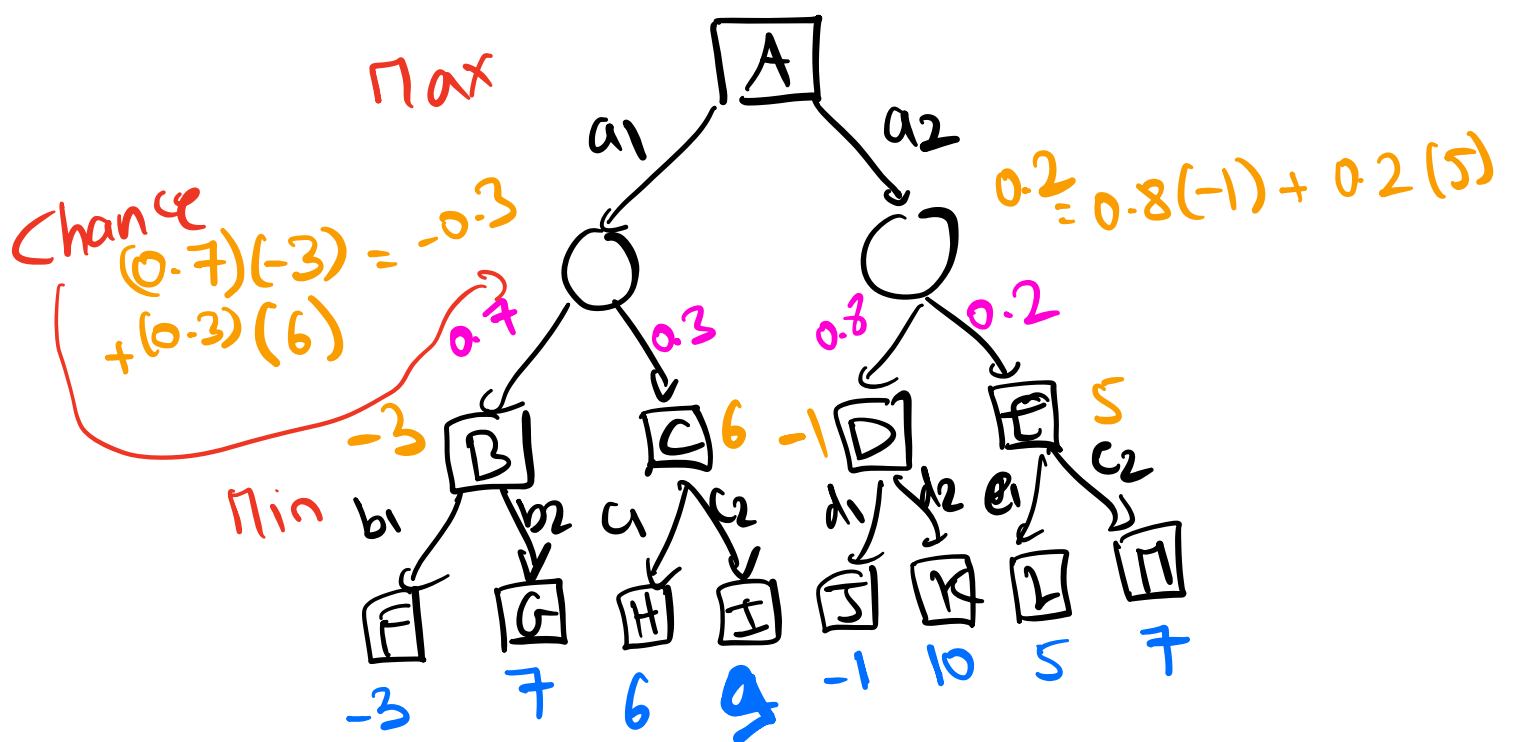
Games with randomness

(Stochastic games)
"random"

random

Proof





$$\text{Expectiminimax}(s) = \begin{cases} \text{Utility}(s) & \text{if } s \text{ is a terminal state} \\ \max_{a \in \text{actions}} \text{Expectiminimax}(s') & \text{if Player = Max} \\ \min_{a \in \text{actions}} \text{Expectiminimax}(s') & \text{if Player = Min} \end{cases}$$

$s' \in \text{succ}(s)$

$$\left[\sum_{s'} P(s \rightarrow s') \cdot \text{Expectiminimax}(s') \right] \text{ if Player = Chance}$$

$$P(B) \text{Expectiminimax}(B) + P(C) \parallel (C)$$

s'_1
 s'_2

(more for coding chance is not really a player)

$$\sum_{s'} P(s \rightarrow s') \text{Expectiminimax}(s')$$

Algorithm (similar to what we did last time we will need function max, min and chance which call each other.)

Exercise

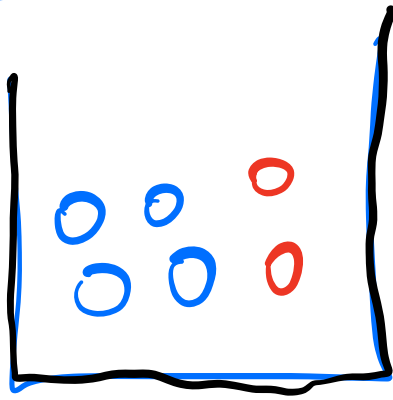
X_1, X_2 random variables

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(X_1 = a | X_2 = b) = \frac{P(X_1 = a \text{ and } X_2 = b)}{P(X_2 = b)}$$

$$P(X_2 = b | X_1 = a) = \frac{P(X_2 = b \text{ and } X_1 = a)}{P(X_1 = a)}$$

$$P(X_1=a \text{ and } X_2=b) = P(X_1=a) P(X_2=b|X_1=a)$$



$$P(X_1 = \text{blue})$$

$$P(X_2 = \text{blue})$$

← depends on what I got for X_1

$$P(X_1=a \text{ and } X_2=b \text{ and } X_3=c)$$

$$P(\overbrace{X_1=a}^B, \overbrace{X_2=b}^A, X_3=c) = P(X_1=a) \cdot P(X_2=b|X_1=a) \cdot P(X_3=c|X_1=a, X_2=b)$$

joint distribution

$$P(A \cap B) = P(X_3=c|X_1=a, X_2=b) \cdot P(X_1=a, X_2=b)$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \Rightarrow P(A \cap B) = P(A|B) \cdot P(B)$$

$$P(X_1=a, X_2=b, X_3=c) = P(X_3=c|X_1=a, X_2=b) \underbrace{P(\overbrace{X_1=a}^B, \overbrace{X_2=b}^A)}_{P(A \cap B) = P(A|B) \cdot P(B)}$$

$$= P(X_3=c | X_1=a, X_2=b) \cdot P(X_2=b | X_1=a) \cdot P(X_1=a)$$