

Last lecture

Random Variables (RV) (Ch 2.1)

- Probability Mass Function (pmf)
- Mean and Variance (Ch 2.2)

Agenda

Random Variables (RV) (Ch 2.1)

- Mean and Variance recap (Ch 2.2)

Conditional Probability (Ch 2.3)

- Motivation
- Examples
- Solver

Law of Total Probability (Ch 2.10)

Mean and Variance

Mean and Variance of linear RV functions

Always true:

By LOTUS:

$$\sum_i (ax_i + b) p(x_i) = a \sum_i x_i p(x_i) + b$$
$$\parallel$$

1. $E[aX + b] = a\mu_x + b = aE[X] + b$
2. $E[X + Y] = \mu_x + \mu_y$

By definition:

1. $Var(aX + b) = a^2 Var(X)$

2. $\sigma_{aX+b} = a\sigma_x$

$$Var(X) = E[(X - \mu_x)^2]$$
$$= E[X^2] - \mu_x^2$$

Detailed later - Requires the relationship between X, Y

1. $Var(X + Y) = E[(X + Y - \mu_x - \mu_y)^2]$

2. $E[(X + Y)^2]$

- All functions related to cross terms, e.g. $E[XY]$

Examples

$X \triangleq$ Roll a die

$A \in [2, 12]$

$A \triangleq$ "Rolling a die twice and sum the results"

$B \triangleq$ "Roll the same die once, multiply the result by 2"

$B \in \{2, 4, 6, 8, 10, 12\}$

• $p_A(k) = p_B(k)$? \times
possible outcomes are diff.

• $E[A] = E[B]$? \checkmark $\mu_A = E[X_1 + X_2] = \mu_{X_1} + \mu_{X_2}$

$\mu_B = E[2X_1] = 2 \mu_X = 2 \mu_X$

• $Var(A) = Var(B)$? \times indep., not in range now

$Var(A) = Var(X_1 + X_2) = Var(X_1) + Var(X_2)$
 $Var(B) = Var(2X_1) = 4 Var(X) = 2 Var(X)$

Standardized RV

$$\mu_Z = 0 \quad \text{Var}(Z) = 1$$
$$= \sigma_Z^2$$

For any RV X

- $\frac{X - \mu_X}{\sigma_X}$ is a **standardized** RV – Mean 0, variance 1

$$E[X - \mu_X] = \mu_X - \mu_X = 0$$

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

Conditional Probability (Ch 2.3)

Motivation

The probability of B happens given A happens

- $P(\text{pair of socks are same color})$ given $S_1 = B$
- $P(I \text{ win Texas Hold'em})$ given $X = \text{Ace} + \text{Ace}$
- $P(I \text{ pass 313})$ given I skip HW1...

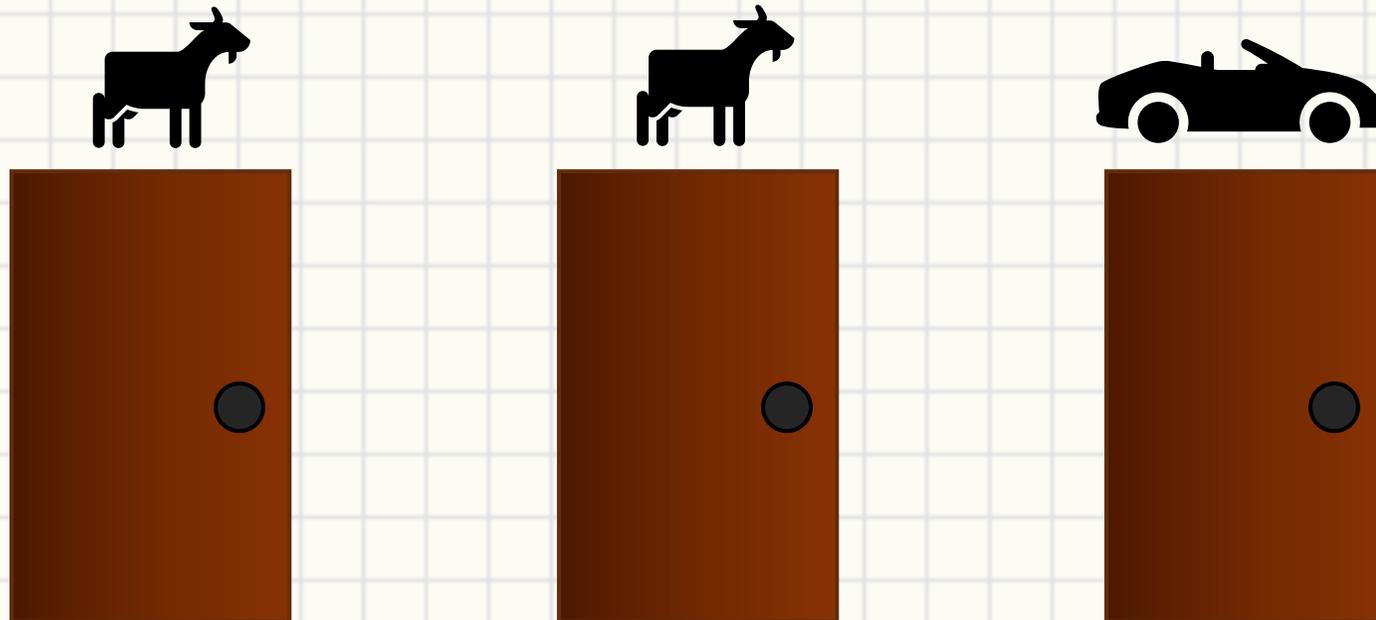
Why do we need conditional probability?

- Analyze the relationship between two events
- Find the optimal solution to make an event probable

Examples

3 doors (Monty Hall) problem

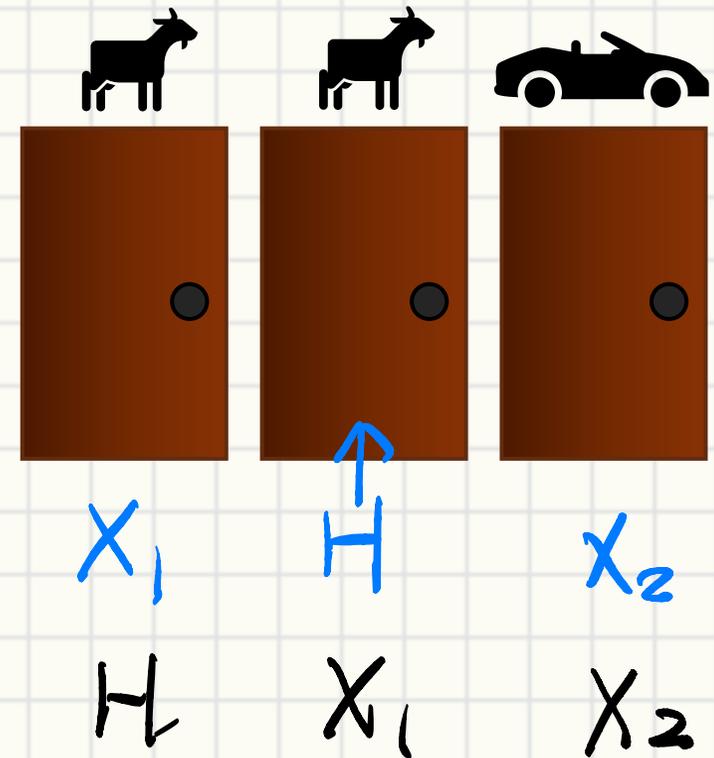
- 3 closed doors – 1 leads to a car, the others lead to goats
- After you choose one, the host will open a “Goat” door
- Should you change the door?



Examples

$P(X|Y) = P(X)$ conditioned on Y

$$X_i \in \{1, 2, 3\}$$



- Let X_i denotes i -th choice
- Never change (NC)
 - $P(W|NC) = P(W|X_2 = X_1)$
 $= P(X_1 = Car) = \frac{1}{3}$
- Change (C)
 - $P(W|C) = P(W|X_2 \neq X_1)$
 $\textcircled{=} P(X_1 = Goat)$
?
- What if there are 4 doors... 2 cars and 2 goats?

Conditional Probability

$$P(B|A) = \begin{cases} \frac{P(A \cap B)}{P(A)} & \text{if } P(A) > 0 \\ \text{Undefined} & \text{else} \end{cases}$$

Roll two dice, A = sum is 6; B = numbers are not equal

$P(B) = ?$ $P(B|A) = ?$ $P(B^c|A) = ?$

$E_A = \{ (1, 5)$

$(2, 4)$

$(3, 3)$

$(4, 2)$

$(5, 1) \}$

$P(B) = \frac{5}{6}$

$P(B^c|A) = \frac{\frac{1}{36}}{\frac{5}{36}} = \frac{1}{5}$

$P(B|A) = \frac{\frac{4}{36}}{\frac{5}{36}} = \frac{4}{5}$

$E_{A \cap B^c} = \{ (3, 3) \}$

Problem Solving

$P(A|B) \Rightarrow$ Is AB separable from B?
CARDS

YES

$P(\text{FULLHOUSE} \mid 3 \text{ OF A KIND})$
 $C_1 = C_2, \quad C_1, C_2, C_3$

$\hookrightarrow P(A|B)$ directly

NO. $P(\text{FULLHOUSE} \mid C_1, C_2 \text{ are diff suits})$

$\hookrightarrow \frac{P(AB)}{P(B)}$

Facts of conditional probability

- $P(B|A) \geq 0$
- $P(B|A) + P(B^c|A) = 1$
- $P(\Omega|A) = 1$
- $P(AB) = P(A|B)P(B)$
- $P(ABC) = P(A|BC)P(B|C)P(C)$

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

BC
= X

$$P(\underline{AX}) = P(A|X)P(X)$$

$$P(AB) = P(A|BC)P(BC)$$

$$P(B|C)P(C)$$

Unfair Games

Not textbook

Alice and Bob are playing a game

- They both play one of "O" or "X" simultaneously
- Alice earns the profits as follows
- If Alice wants to maximize her profits, $P_A(O) =$

	O	X
O	-1	2
X	2	-3

⇒ Nash Equilibrium $P_A(O) = P$

$$\begin{array}{l} \text{A O} \quad \text{A X} \\ \text{B O} \quad p(-1) \quad (1-p)(2) \\ \hline \text{B X} \quad p(2) \quad (1-p)(-3) \end{array}$$

$$\begin{aligned} E[A | B = "O"] \\ &= -p + 2(1-p) \\ &= 2 - 3p \end{aligned}$$

$$E[A|B = \text{"X"}] = 2p + (1-p)(-3)$$

$$= 5p - 3$$

$$2 - 3p = 5p - 3$$

$$\Rightarrow p^* = \frac{5}{8}$$

Nash Eq.

$$\min(2 - 3p, 5p - 3)$$

Law of total probability

$$2 - 3p^* = \frac{3}{8}$$

$$= 5p^* - 3$$