

# Class - 2022/09/29

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Conditional Probability

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Example

## Conditional Probability

- Independence

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Recall: Two events  $A, B$  are mutually exclusive if  $A \cap B = \emptyset \rightarrow P(AB) = 0$  (They cannot occur together).

Two events  $A, B$  are independent if  $P(A|B) = P(A)$  and  $P(B|A) = P(B)$

That is, they have no effect on each other

The formula can also be written as:

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

## • Proposition

- Assume  $A$  &  $B$  are independent, show  $A$  &  $B^c$  are also independent.

## – Proof

- $A = AB \cup AB^c$  (They are Disjoint = Mutually Exclusive)
- $P(A) = P(AB \cup AB^c) = P(AB) + P(AB^c)$
- $P(A) = P(A)P(B) + P(AB^c)$
- $P(AB^c) = P(A) - P(A)P(B) = P(A)(1 - P(B)) = P(A)P(B^c)$
- $P(A)P(B^c) = P(AB^c)$
- Therefore,  $A$  &  $B^c$  are also independent.

## • Example

Suppose that chance Lannie earns an A in the course is 40% and the chance that Lannie gets an A in Analysis is 70%. If these events are independent, find the probability that Lannie:

- Gets A in both  $\Rightarrow P(AP) = 0.28$
- Gets A in neither  $\Rightarrow P(A'P') = 0.18$
- Gets exactly one A  $\Rightarrow P(AP') + P(A'P) = 0.42 + 0.12 = 0.54$

	P	P'	Total
A	$P(AP) = 0.7 \cdot 0.4 = 0.28$	$P(AP') = 0.7 \cdot 0.6 = 0.42$	0.7
A'	$P(A'P) = 0.3 \cdot 0.4 = 0.12$	$P(A'P') = 0.3 \cdot 0.6 = 0.18$	0.3
Total	0.4	0.6	1

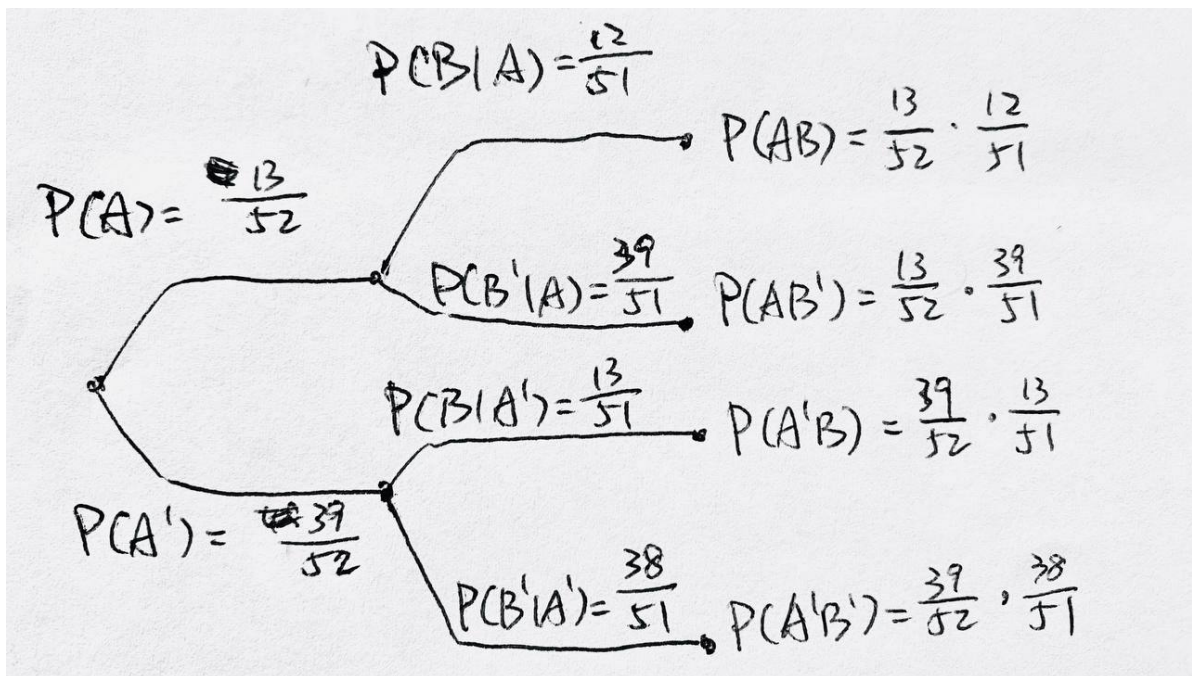
Two cards are dealt sequentially without replacement.

Suppose  $A$  is the event the first card is SPADE.

Suppose  $B$  is the event the second card is SPADE.

Find  $P(B|A)$  and  $P(A|B)$

Are these events independent?



$$P(B|A) = \frac{12}{51} \neq \frac{1}{4} = P(B)$$

$$P(B) = \frac{13}{52} \cdot \frac{12}{51} + \frac{39}{52} \cdot \frac{13}{51} = \frac{1}{4}$$

$A$  &  $B$  are **NOT** independent!

## Chapter 4: Random Variables

Recall: Experiment  $\rightarrow$  Sample Space  $\rightarrow$  (new) Real number

- Ex: flip 2 coins  $\rightarrow S = \{HH, HT, TH, TT\} \rightarrow$  See example below

### • Definition

A random variable is a **function** that assigns a real number to each element in the sample space.

- We often use capital letters  $X, Y, Z$  to represent  $R.V.$
- $X : S \rightarrow \mathbb{R}, X(s) = r$
- Often express  $X$  as its possible outcomes

### – Example:

- Flip 2 coins.
- Let  $X$  be the  $R.V.$  that counts the number of heads.
  - $X = \{0, 1, 2\}$
  - $X : S \rightarrow \mathbb{R}$
  - $X(HH) = 2$
  - $X(HT) = 1$
  - $X(TH) = 1$
  - $X(TT) = 0$
- Let  $Y$  be the  $RV$  that subtracts number of heads from number of tails
  - $Y = \{2, 0, -2\}$

- Let  $Z$  be the  $RV$  that multiplies number of heads by 3 and adds to number of tails squared
  - $Z = \{6, 4\}$
  - $Z(HH) = 6$
  - $Z(HT) = 4$
  - $Z(TH) = 4$
  - $Z(TT) = 4$
- A moral is a number  $RV$  defined on a given sample space, and it's not unique
- However:** Once a  $RV$  is defined, then the probability distribution function(PDF) is unique
  - Examples for  $f_X, f_Y, f_Z$

	$X$	$f_X$	$Y$	$f_Y$	$Z$	$f_Z$
	0	1/4	2	1/4	6	1/4
	1	1/2	0	1/2	4	3/4
	2	1/3	-2	1/4		
Total		1		1		1

## • Discrete Random Variables (DRV)

A Discrete Random Variable on a Sample Space  $S$ , is a Random Variable that takes on a finite number of values, or countably infinite number of values.

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- Countable:
  - finite
  - countably infinite set:  $\mathbb{N}, \mathbb{Z}, 2\mathbb{Z}, \mathbb{Q}, \dots$  (all with equal sizes)
  - think about it like gaps
- Uncountable:
  - $\mathbb{R}, (0, 1), \dots$  or worse
  - no gaps

Therefore, we have

Experiment  $\rightarrow S \rightarrow$  (Define)  $X \rightarrow f_X$

### – Example:

- Flip a coin until a heads appears.  $S = \{H, TH, TTH, TTTH, \dots\}$
- Define  $X$  be the  $RV$  that counts the number of trials need to to get the first head.
  - $X = \{1, 2, 3, \dots\}$
  - $X$  is DRV

## • Probability Mass Function (PMF)

Our **probability distribution function** for a **DRV** is called a probability mass function (pmf)

- EX: The pmf for a DRV,  $X$
- $p : X \rightarrow [0, 1]$ 
  - $0 \leq p(x) \leq 1$  for all  $x \in X$
  - $\sum_{x \in X} p(X) = 1$
  - $p(a) = P(X = a)$  (assigns probabilities to each  $x$  in  $X$ )

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For the continuous mass function, the properties would change slightly

- same
- change the sum to integral
- not  $P(X = a)$

### - Example

- A pmf for a DRV,  $I$ , is given by:
  - $p : I \rightarrow [0, 1], p(i) = \frac{c\lambda^i}{i!}, i = 0, 1, 2, \dots (I = \{0, 1, 2, \dots\})$
- Find the constant  $c$ .

$$\sum_{i \in I} p(i) = 1 = \sum_{i=0}^{\infty} \frac{c\lambda^i}{i!} = c \sum_{i=0}^{\infty} \frac{\lambda^i}{i!}$$

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Recall :  $e^x = \sum_{i=0}^{\infty} \frac{x^i}{i!}$

$$\begin{aligned} \sum_{i=0}^{\infty} \frac{c\lambda^i}{i!} &= c \sum_{i=0}^{\infty} \frac{\lambda^i}{i!} = c \cdot e^{\lambda} = 1 \\ c &= \frac{1}{e^{\lambda}} = e^{-\lambda} \end{aligned}$$

- pmf:  $p(i) = \frac{e^{-\lambda}\lambda^i}{i!}$

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PMF for an important named DRV

- $p(0) = \frac{e^{-\lambda}\lambda^0}{0!} = e^{-\lambda}$
- $P(I \geq 2) = 1 - P(I \leq 1) = 1 - p(0) - p(1) = 1 - e^{-\lambda} - e^{-\lambda}\lambda$