

Conditional Probability

July, 20, 2020

- 1) The Monty Hall Problem
- 2) Conditional Probability
- 3) Bayes' Rule
- 4) Total Probability Rule.

Last time:

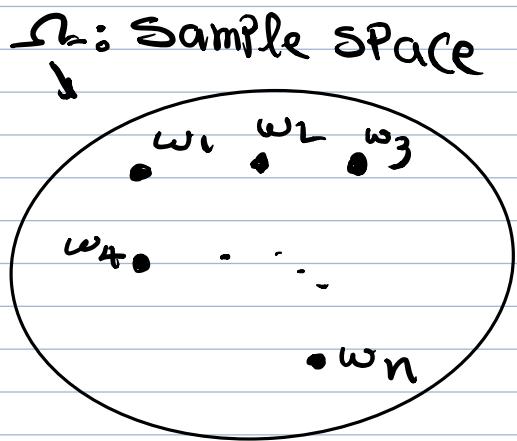
w_i : Sample Point (Outcome)
 $i=1, \dots, n$

$$0 \leq \Pr[w_i] \leq 1$$

$$\sum_{i=1}^n \Pr[w_i] = 1$$

$$\text{Events: } E \Rightarrow \Pr[E] = \sum_{w_i \in E} \Pr[w_i]$$

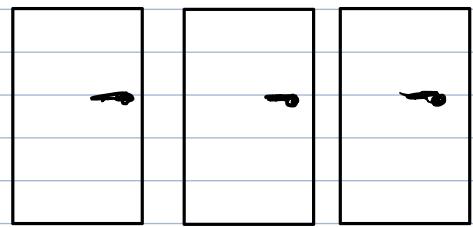
$$\text{The complement of event } E: \bar{E} \Rightarrow \Pr[\bar{E}] = 1 - \Pr[E]$$



I) The Monty Hall Problem:

1 2 3

- There are three doors.
- A car is behind one of the doors!
- Goats behind the other two doors.



-
1. The contestant picks a door (but does not open it)
 2. Then, Hall's assistant opens one of the other two doors, revealing a goat.
 3. The contestant is then given the option of sticking with their current choice or switching to the other unopened door.
 4. He/she wins the car if and only if their chosen door is the correct one.

Question: Does the contestant have a better chance of winning if he/she switches doors?

Assume the prize is equally likely to be behind any of the three doors.

sample space

c: car

g: goat

$$\Omega = \{ cgg, gcg, gg \}$$

1	2	3	Pr
w ₁ : c	g	g	$\frac{1}{3}$
w ₂ : g	c	g	$\frac{1}{3}$
w ₃ : g	g	c	$\frac{1}{3}$

WLOG we assume w₁ happens

$$\Omega' = \{ 1, 2, 3 \}$$

1	2	3
c	g	g

$$Pr[1] = Pr[2] = Pr[3] = \frac{1}{3}$$

E₁: winning by switching

E ₁ :	chosen		switch
	1	2, 3	
	1	2, 3	3, 2
	2	3	1
	3	2	1

$$Pr[E_1] = Pr[1 \rightarrow 2] + Pr[1 \rightarrow 3] + Pr[2 \rightarrow 1]$$

$$+ Pr[3 \rightarrow 1] = 0 + 0 + \frac{1}{3} + \frac{1}{3} = \frac{2}{3}$$

$$Pr[1 \rightarrow 2] = \frac{1}{3} \times 0 = 0$$

$$Pr[E_1] = \frac{2}{3}$$

$$\Pr[1 \rightarrow 1] = \frac{1}{3} \times 1 = 0$$

$$\Pr[2 \rightarrow 1] = \frac{1}{3} \times 1 = \frac{1}{3}$$

$$\Pr[3 \rightarrow 1] = \frac{1}{3} \times 1 = \frac{1}{3}$$

E_2 : winning by not switching

		chosen opened	Stick to
$E_2:$	1	{ 2, 3 }	1
	2	{ 3 }	2
	3	{ 2 }	3

$$\Pr[E_2] = \Pr[1 \rightarrow 1] + \Pr[2 \rightarrow 2] + \Pr[3 \rightarrow 3]$$

$$= \frac{1}{3} + 0 + 0 = \frac{1}{3} \Rightarrow \boxed{\Pr[E_2] = \frac{1}{3}}$$

$$\Pr[1 \rightarrow 1] = \frac{1}{3} \times 1 = \frac{1}{3}$$

$$\Pr[2 \rightarrow 2] = \frac{1}{3} \times 0 = 0$$

$$\Pr[3 \rightarrow 3] = \frac{1}{3} \times 0 = 0$$

we have a better chance of winning if
we use switching strategy.

2) Conditional Probability:

Examples: Two coin flips. First flip is heads.

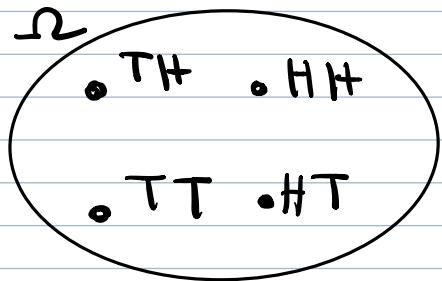
Probability of two heads?

$\Omega = \{\underline{H}\underline{H}, \underline{H}\underline{T}, \underline{T}\underline{H}, \underline{T}\underline{T}\}$, uniform probability space.

Event A = first flip is heads: $A = \{\underline{H}\underline{H}, \underline{H}\underline{T}\}$

New sample space

Event B = two heads = $\{\underline{H}\underline{H}\}$



The Probability of two heads if the first flip is heads.

The Probability of B given A is $Pr[B|A] = \frac{1}{2}$

Example: Two coin flips. At least one of the flips is heads. Probability of two heads?

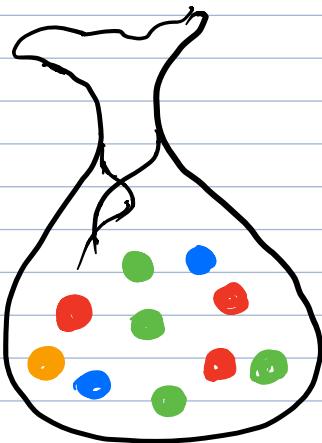
Event A = $\{\underline{H}\underline{T}, \underline{T}\underline{H}, \underline{H}\underline{H}\}$

Event B = $\{\underline{H}\underline{H}\}$

\Rightarrow The Probability of B given A

$$\Rightarrow Pr[B|A] = \frac{1}{3}$$

A non-uniform example



Experiment



$$\Omega = \{\text{Red, Blue, Green, Orange}\}$$

$$\Pr[\text{red} | \text{red or green}] = \frac{3}{7}$$

\downarrow \downarrow
B A

$$A = \{3 \text{ reds, 4 greens}\}$$

$$B = \{3 \text{ reds}\}$$

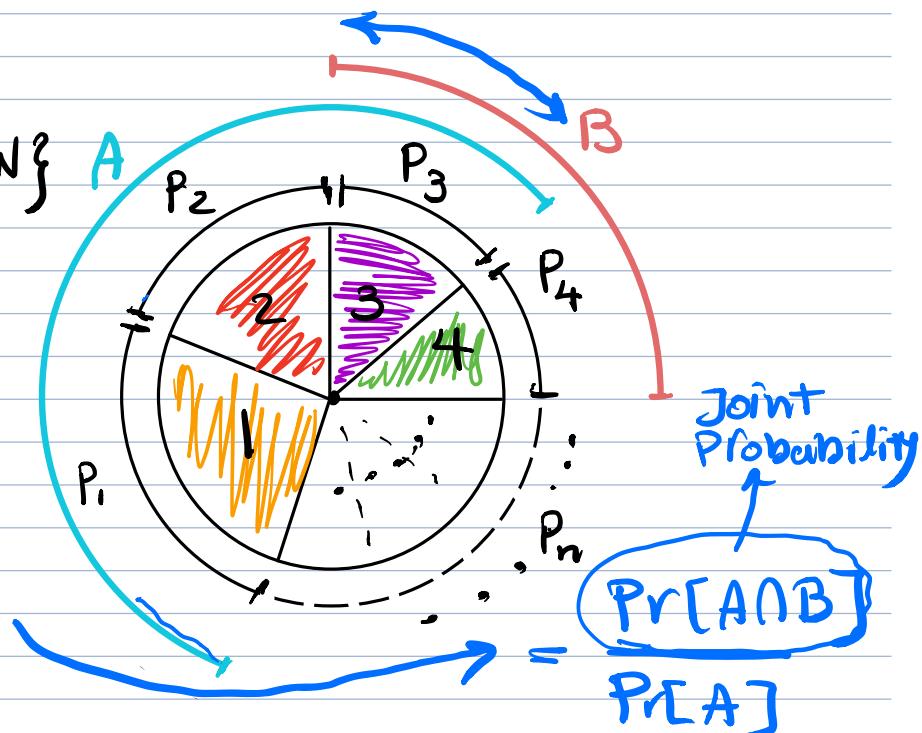
Another example:

Consider $\Omega = \{1, 2, \dots, N\}$
with $\Pr[n] = P_n$

$$A = \{1, 2, 3\}$$

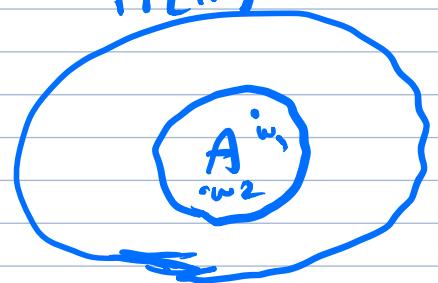
$$B = \{3, 4\}$$

$$\Pr[B|A] = \frac{P_3}{P_1 + P_2 + P_3}$$



Assume sample point $\omega \in A$, then

$$\Pr[\omega|A] = ? = \frac{\Pr[\omega \cap A]}{\Pr[A]} = \frac{\Pr[\omega]}{\Pr[A]} \text{ since } \Pr[\omega] = 1$$



Then

$$\Pr[B|A] = ? = \frac{\sum_{\omega \in B} \Pr[\omega \cap A]}{\Pr[A]} = \frac{\Pr[B \cap A]}{\Pr[A]}$$

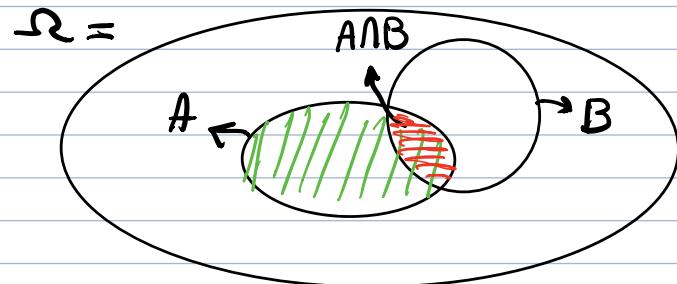
Definition: The conditional Probability of \underline{B}

given A is

$$\Pr[B|A] = \frac{\Pr[B \cap A]}{\Pr[A]}$$

A given B

$$\Pr[A|B] = \frac{\Pr[A \cap B]}{\Pr[B]}$$



one more example:

SUPPOSE I toss 3 balls into 3 bins, one at the time

$A = "1^{\text{st}} \text{ bin is empty}" ; B = "2^{\text{nd}} \text{ bin is empty}"$

what is $\Pr[A|\underline{B}]$?

$$\Omega = \underbrace{\{1, 2, 3\}}_3^3 \rightarrow |\Omega| = 3^3 = 27$$

$$A = \{2, 3\}^3, B = \{1, 3\}^3 \quad \left. \begin{array}{l} |A| = 2^3 = 8 \\ |B| = 2^3 = 8 \end{array} \right\} \quad \left. \begin{array}{l} A \cap B = \{3\}^3 \\ |A \cap B| = 1^3 = 1 \end{array} \right\}$$

0 0 0

1 2 3

$$\Pr[A|B] = \frac{\Pr[A \cap B]}{\Pr[B]} = \frac{\frac{1}{27}}{\frac{8}{27}} = \frac{1}{8}.$$

$$\Pr[B] = \frac{|B|}{|S|} = \frac{8}{27}, \quad \Pr[A \cap B] = \frac{|A \cap B|}{|S|} = \frac{1}{27}$$

Bayesian Inference:

- A way update knowledge after making an observation
- we may have an estimate of probability of a given event A. \rightarrow Prior knowledge
- After event B occurs we can update this estimate to $\Pr[A|B]$
- In this interpretation

$\Pr[A]$: Prior Probability
$\Pr[A B]$: Posterior Probability

Example:

There is a new test for a certain disease.

1) when the test applied to an affect person, the test comes up positive 90% of cases. and negative in 10%. \rightarrow ("False negative")

2. when applied to a healthy person the test $\rightarrow \bar{A}$

comes up negative 80% of cases and Positive
20%. "False Positive"

- Suppose that only 5% of the population has this disease. (Prior) $\rightarrow A$

Q. When a random person is tested positive, what is the probability that the person has the disease? $\rightarrow \underline{\text{Pr}[A|B]}$

Let's define events:

A: affected

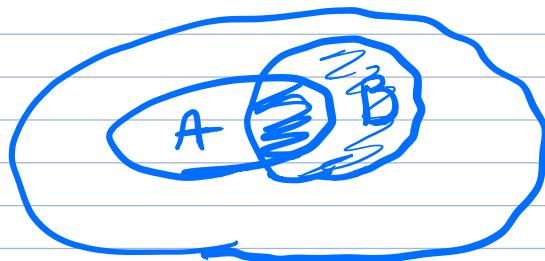
B: Test is Positive.

$$\text{Pr}[A] = 0.05, \quad \underline{\text{Pr}[B|A]} = 0.9, \quad \text{Pr}[B|\bar{A}] = 0.2$$

$$\text{Pr}[A|B] = \frac{\text{Pr}[A \cap B]}{\text{Pr}[B]} = \frac{\text{Pr}[B|A] \text{Pr}[A]}{\text{Pr}[B]}$$

$$\text{Pr}[B|A] = \frac{\text{Pr}[A \cap B]}{\text{Pr}[A]} \Rightarrow \text{Pr}[A \cap B] = \underline{\text{Pr}[B|A]} \underline{\text{Pr}[A]}$$

$$\text{Pr}[B] = \text{Pr}[A \cap B] + \text{Pr}[\bar{A} \cap B]$$



$$\Pr[A \cap B] = \Pr[B|A] \Pr[A]$$

$$\Pr[\bar{A} \cap B] = \Pr[B|\bar{A}] \Pr[\bar{A}] = \Pr[B|\bar{A}](1 - \Pr[A])$$

$$\Pr[A|B] = \frac{\Pr[B|A] \Pr[A]}{\Pr[B|A] \Pr[A] + \Pr[B|\bar{A}](1 - \Pr[A])}$$

3) Bayes' Rule: $\Pr[A|B] = \frac{\Pr(B|A) \Pr(A)}{\Pr(B)}$

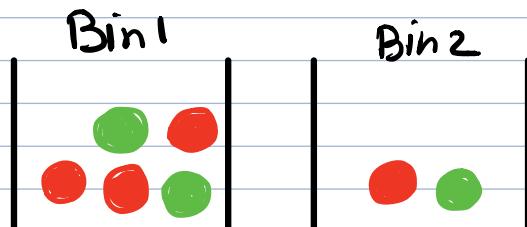
$$\Pr[A|B] = \frac{\Pr[A \cap B]}{\Pr[B]}$$

$$\Pr[B|A] = \frac{\Pr[A \cap B]}{\Pr[A]}$$

4) Total Probability Rule:

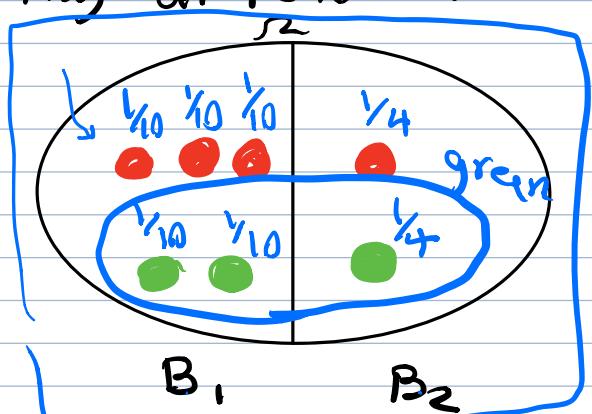
Imagine two bins containing some number of red and green.

- One bin is chosen with equal probability $\Rightarrow 50\%$.



- Then a ball is drawn uniformly at random.

- What is the probability that we picked Bin 1 given that a green ball was drawn?



$$\Pr[B_1 | \text{green}] = \frac{\Pr[\text{green} | B_1] \Pr[B_1]}{\Pr[\text{green}]} = \frac{\frac{1}{5}}{\frac{4}{9}} = \frac{9}{20}$$

$$\begin{aligned}\Pr[\text{green}] &= \Pr[\text{green} \cap B_1] + \Pr[\text{green} \cap B_2] \\ &= \Pr[\text{green} | B_1] \Pr[B_1] + \Pr[\text{green} | B_2] \Pr[B_2] \\ &= \frac{2}{5} \times \frac{1}{2} + \frac{1}{2} \times \frac{1}{2} = \frac{9}{20}\end{aligned}$$

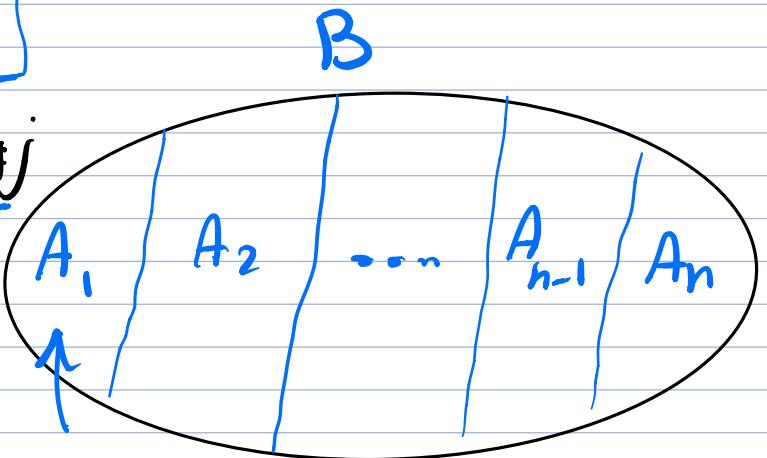
Definition: Event B is Partitioned into n events A_1, \dots, A_n if

$$1. B = \underline{A_1 \cup A_2 \cup \dots \cup A_n}$$

$$2. A_i \cap A_j = \emptyset \text{ for all } i \neq j$$

(i.e. A_1, \dots, A_n are

mutually exclusive



Then the total probability is

$$\Pr[B] = \sum_{i=1}^n \Pr[B \cap A_i] = \sum_{i=1}^n \Pr[B | A_i] \Pr[A_i]$$

So for the bayes rule

$$\Pr[A_i | B] = \frac{\Pr[B | A_i] \Pr[A_i]}{\Pr[B]} = \frac{\Pr[B | A_i] \Pr[A_i]}{\sum_{j=1}^n \Pr[B | A_j] \Pr[A_j]}$$