

# Conditional Probability

A = {Covid Test + }

B = {Has Covid}

Why are A and B not same?

# Conditional Probability

$$A = \{\text{Covid Test +}\}$$

$$B = \{\text{Has Covid}\}$$

Why are A and B not same?

Ground Truth (Actual)

Predictional Test

		No Covid	Has COVID
Has Covid	No Grid		
	Has Grid		

# Conditional Probability

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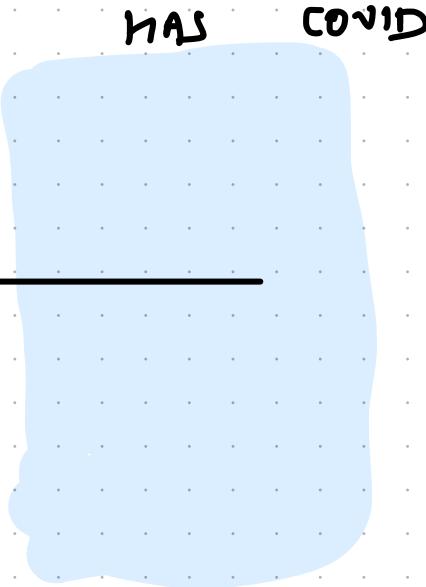
Why are A and B not same?

Ground Truth (Actual)

Prediction / Test

Has Covid No Grid

No Covid



# Conditional Probability

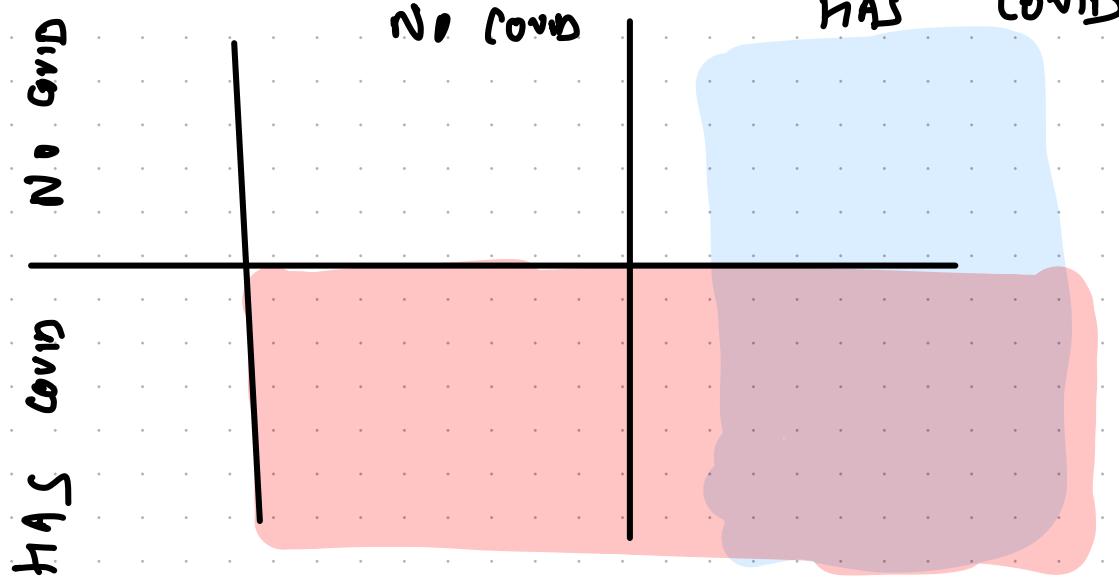
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Why are A and B not same?

Ground Truth (Actual)

Prediction / Test



# Conditional Probability

$$A = \{\text{Covid Test +}\}$$

$$B = \{\text{Has Covid}\}$$

$$A = \{P_1, P_2, P_3, P_4, P_7, P_8\}$$

$$B = \{P_1, P_2, P_6, P_7, P_8, P_9, P_{10}\}$$

$$U = \{P_1, \dots, P_{12}\}$$

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Confusion Matrix

		GT	
		No COVID	Covid
Pred.	No COVID	$A^c \cap B^c$	$A^c \cap B$
	Covid	$A \cap B^c$	$A \cap B$

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$$U = \{P_1, \dots, P_{12}\}$$

$$A^c = \{P_5, P_6, P_9, P_{10}, P_{11}, P_{12}\}$$

$$B^c = \{P_3, P_4, P_5, P_{11}, P_{12}\}$$

## Confusion Matrix

		GT	
		No COVID	Covid
Pred.	No COVID	$A^c \cap B^c$	$A^c \cap B$
	Covid	$A \cap B^c$	$A \cap B = \{1, 2, 7, 8\}$

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## Confusion Matrix

		GT	
		No COVID	Covid
Pred.	No COVID	$A^c \cap B^c = \{5, 4, 12\}$	$A^c \cap B = \{6, 9, 10\}$
	Covid	$A \cap B^c = \{3, 4\}$	$A \cap B = \{1, 2, 7, 8\}$

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## Confusion Matrix

		GT	
		NO COVID	COVID
Pred.	NO COVID	3	3
	COVID	2	4

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## Confusion Matrix

		G <sub>T</sub>	
		No COVID	COVID
Pred.	No COVID	3 (TN)	3 (FN)
	COVID	2 (FP)	4 (TP)

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I deadly  
 TP, TN high  
 FN, FP low

## Confusion Matrix

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		No COVID	COVID
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	COVID	2 (FP)	4 (TP)

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## Confusion Matrix

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		No COVID	COVID
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	COVID	$A \cap B^c$	$A \cap B$

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$$B = \{P_1, P_2, P_6, P_7, P_8, P_9, P_{10}\}$$

$$U = \{P_1, \dots, P_{12}\}$$

$$P[A] =$$

$$P[B] =$$

$$P[A \cap B]$$

$$P[A|B] =$$

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$$U = \{P_1, \dots, P_{12}\}$$

$$P[A] = 6/12 = 0.5$$

$$P[B] = 7/12$$

$$P[A \cap B] = 4/12 = 1/3$$

$$P[A|B] = 4/7 ; P[B|A] = 4/6 = 2/3$$

		Confusion Matrix	
		GT	
Pred.	NO COVID	3 (TN)	3 (FN)
	COVID	2 (FP)	4 (TP)

# Conditional Probability

$$A = \{ \text{Get 3} \}$$

$$B = \{ \text{odd numbers} \} = \{ 1, 3, 5 \}$$

$$P[A \cap B] = ? ; P[B|A] = ?$$

# Conditional Probability

$$A = \{ \text{Get 3} \}$$

$$B = \{ \text{odd numbers} \} = \{ 1, 3, 5 \}$$

$$P[A \cap B] = ? ; P[B|A] = ?$$

$$A \cap B = A = \{ 3 \}$$

$$P[A \cap B] = \frac{n(\{3\})}{n(\{1, 3, 5\})} = \frac{1}{3}$$

$$= \frac{P(\{3\})}{P(\{1, 3, 5\})} = \frac{\frac{1}{6}}{\frac{3}{6}} = \frac{1}{3}$$

# Conditional Probability

$$A = \{ \text{Get 3} \}$$

$$B = \{ \text{odd numbers} \} = \{ 1, 3, 5 \}$$

$$P[A \cap B] = ? ; P[B|A] = ?$$

$$A \cap B = A = \{ 3 \}$$

$$P[B|A] = \frac{n(\{3\})}{n(\{3\})} = 1$$

$$= \frac{P(\{3\})}{P(\{3\})} = \frac{\frac{1}{6}}{\frac{1}{6}} = 1$$

4 sided dice

$x$ : first roll ;  $y$ : second roll

$B$ :  $\min(x, y) = 2$  ;  $M = \max(x, y) = 3$

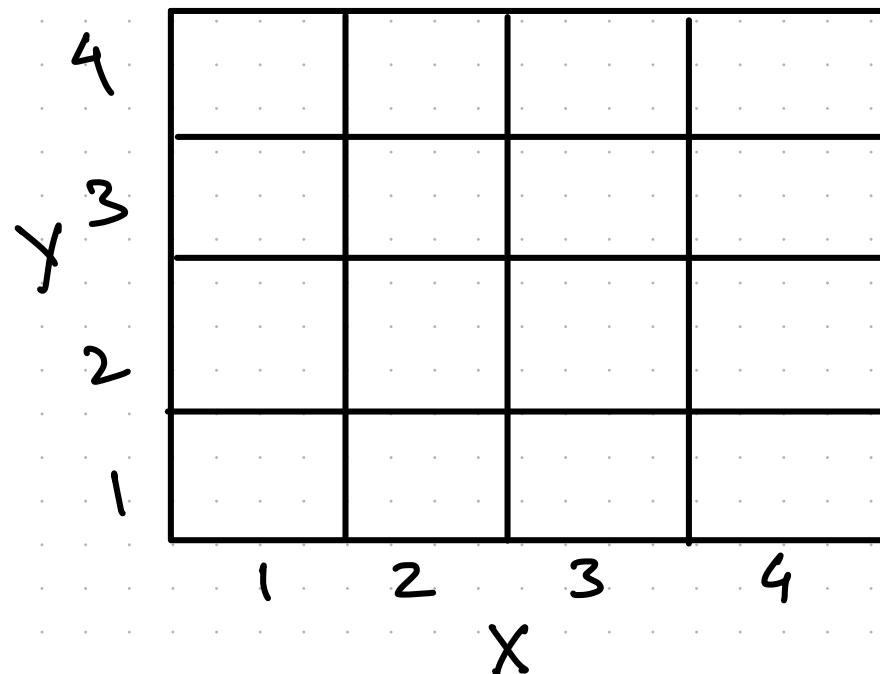
$P[M|B] = ?$

4 sided dice

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$B$ :  $\min(x, y) = 2$  ;  $M = \max(x, y) = 3$

$$P[M|B] = ?$$

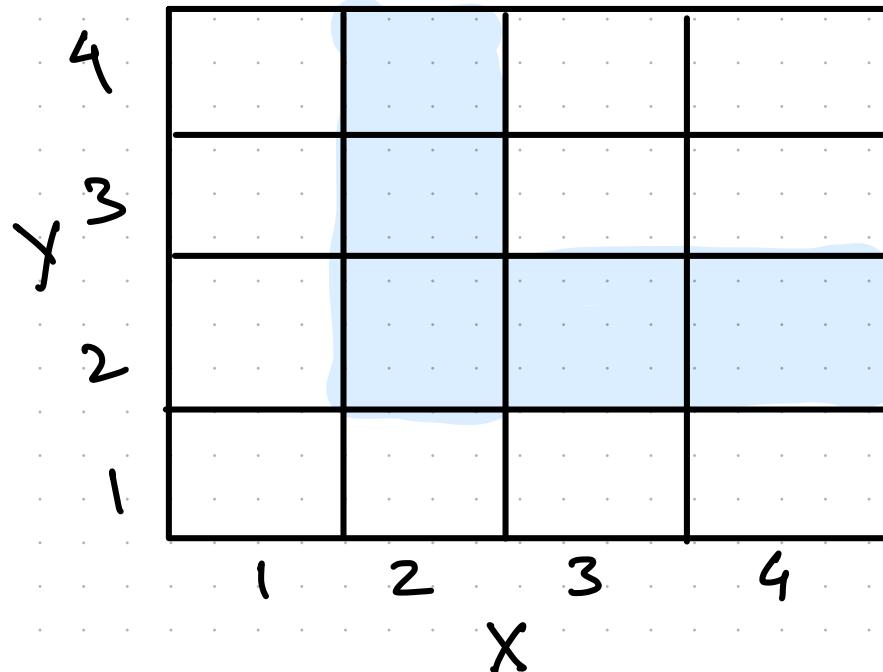


4 sided dice

$x$ : first roll;  $y$ : second roll

$B$ :  $\min(x, y) = 2$ ;  $M = \max(x, y) = 3$

$$P[M|B] = ?$$



Marking B

examples

$$\min(1,1) = 1 \neq 2$$

$$\min(2,1) = 2 = 2$$

$$\min(2,4) = 2 = 2$$

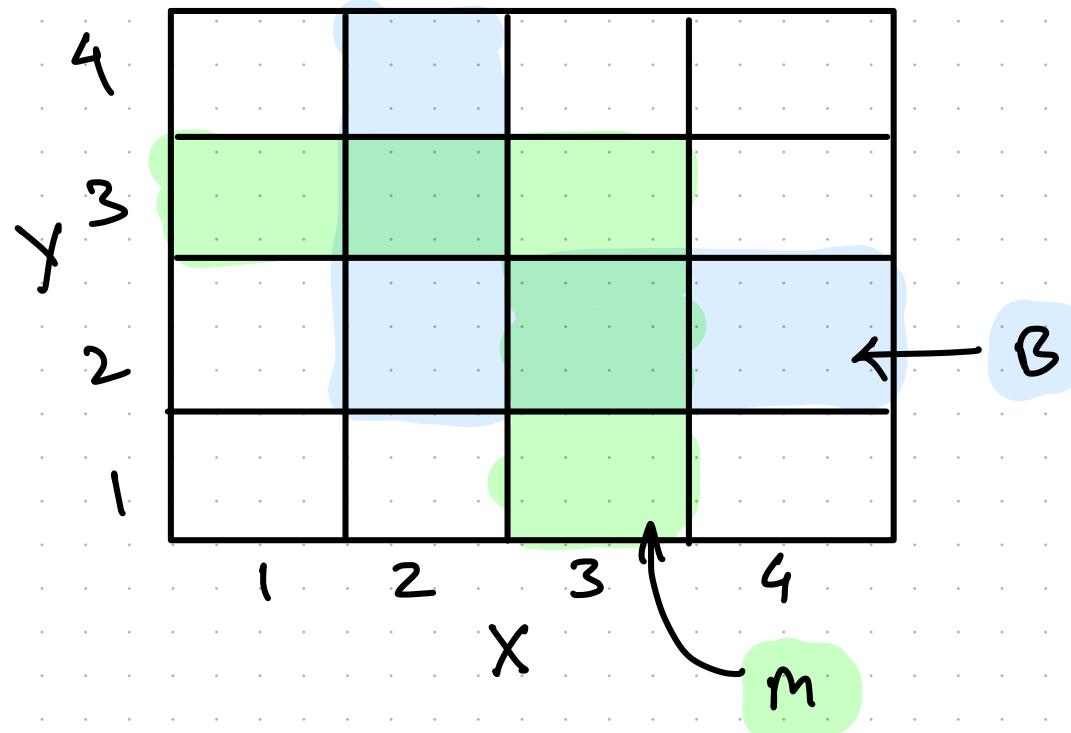
$$\min(4,3) = 3 \neq 2$$

4 sided dice

$x$ : first roll ;  $y$ : second roll

$B$ :  $\min(x, y) = 2$  ;  $M = \max(x, y) = 3$

$$P[M|B] = ?$$

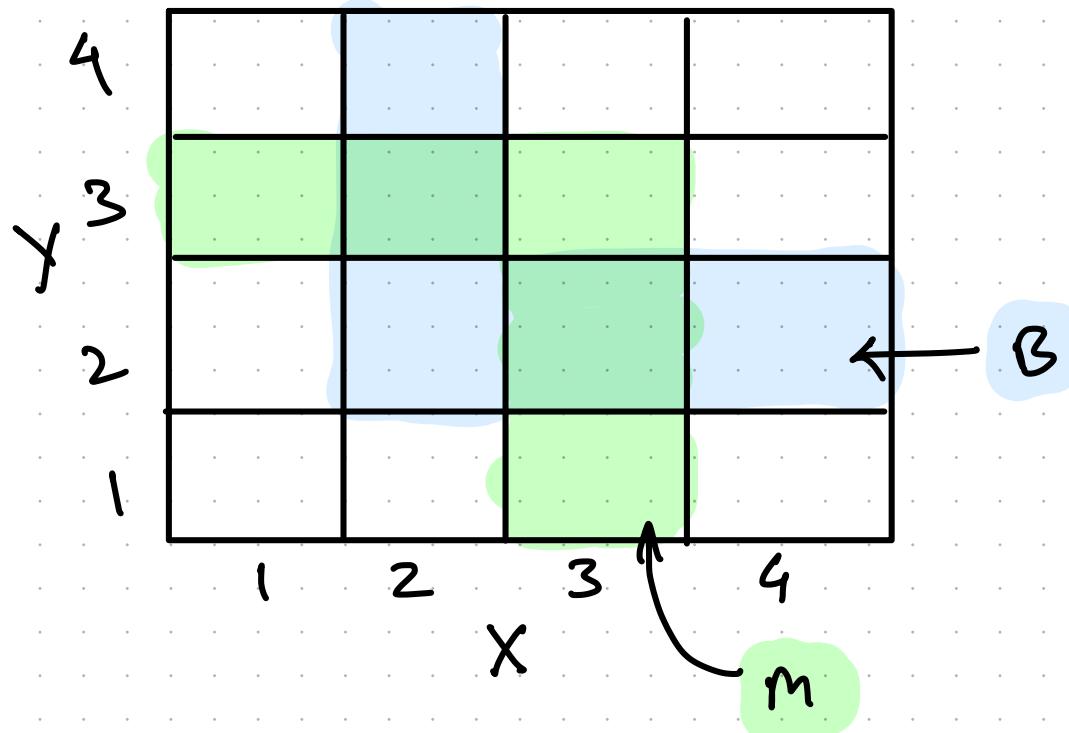


4 sided dice

$x$ : first roll ;  $y$ : second roll

$B$ :  $\min(x, y) = 2$  ;  $M = \max(x, y) = 3$

$$P[M|B] = ?$$



$$n(M \cap B) = 2$$

$$n(B) = 5$$

$$\therefore P[M|B] = \frac{2}{5}$$

Let  $P[B] > 0$ ;  $P[A|B]$  satisfies 3 Axioms of Probability

1) Non-negativity:  $P[A|B] = \frac{P[A \cap B]}{P[B]} > 0$   
 $\therefore P[A|B] > 0$

2) Normalization:  $P(\Omega|B) = 1$

Let  $P[B] > 0$ ;  $P[A \setminus B]$  satisfies 3 Axioms of Probability

1) Non-negativity:  $P[A \setminus B] = \frac{P[A \cap \bar{B}]}{P[\bar{B}]} > 0$   
 $\therefore P[A \setminus B] > 0$

2) Normalization:  $P(\Omega \setminus B) = 1$   
 $= \frac{P(\Omega \cap \bar{B})}{P(\bar{B})} = \frac{P(\bar{B})}{P(B)} = 1$

3) Disjoint A and C

$$P[A \cup C | B] = P[A|B] + P[C|B]$$

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$$P[A \cup C | B] = P[A|B] + P[C|B]$$

LHS

$$P[A \cup C | B] = \frac{P[(A \cup C) \cap B]}{P[B]}$$

$$= \frac{P[(A \cap B) \cup (C \cap B)]}{P[B]}$$

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$$P[A \cup C | B] = P[A|B] + P[C|B]$$

LHS

$$P[A \cup C | B] = \frac{P[(A \cup C) \cap B]}{P[B]}$$

$$= \frac{P[(A \cap B) \cup (C \cap B)]}{P[B]}$$

$\because A$  and  $C$  are disjoint;

$(A \cap B)$  and  $(C \cap B)$  are disjoint.

$$\therefore P[A \cup C | B] = \frac{P[A \cap B] + P[C \cap B]}{P[B]} = P[A|B] + P[C|B]$$

## Independence

- \* Two Coins toss
- \* Event B = {1<sup>st</sup> coin is H}
- \* Event A = {2<sup>nd</sup> coin is H}
- \* Does knowing "B" change anything in "A".

## Independence

Two events A and B are independent if:

Knowing one occurred gives no extra information about other.

$$P(A|B) = P(A)$$

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Two events A and B are independent if:

Knowing one occurred gives no extra information about other.

$$P(A|B) = P(A)$$

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

$$\Rightarrow P(A \cap B) = P(A) \cdot P(B)$$

For 2 coin toss:

$$B = \{2^{\text{nd}} \text{ coin is H}\}$$

$$P(B) = 1/2$$

$$A = \{1^{\text{st}} \text{ coin is H}\}$$

$$P(A) = 1/2$$

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

For 2 coin toss:

$$B = \{1^{\text{st}} \text{ coin is } H\}$$

$$P(B) = \frac{1}{2}$$

$$A = \{2^{\text{nd}} \text{ coin is } H\}$$

$$P(A) = \frac{1}{2}$$

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

If 1<sup>st</sup> coin is H; 2<sup>nd</sup> still has  $\frac{1}{2}$  prob. of H

$$\therefore P(A|B) = \frac{1}{2}$$

$$\begin{aligned} &= \frac{P(A \cap B)}{P(B)} = \frac{n(\{H,H\})}{n(\{(H,H), (H,T), (T,H), (T,T)\})} \\ &= \frac{1}{2} \end{aligned}$$

For 2 coin toss:

$$B = \{1^{\text{st}} \text{ coin is } H\}$$

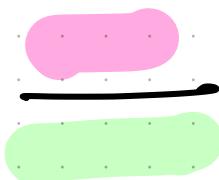
$$P(B) = \frac{1}{2}$$

$$A = \{2^{\text{nd}} \text{ coin is } H\}$$

$$P(A) = \frac{1}{2}$$

$$\Omega = \{HH, HT, TH, TT\}$$

$$P(A|B) = \frac{1}{2}$$

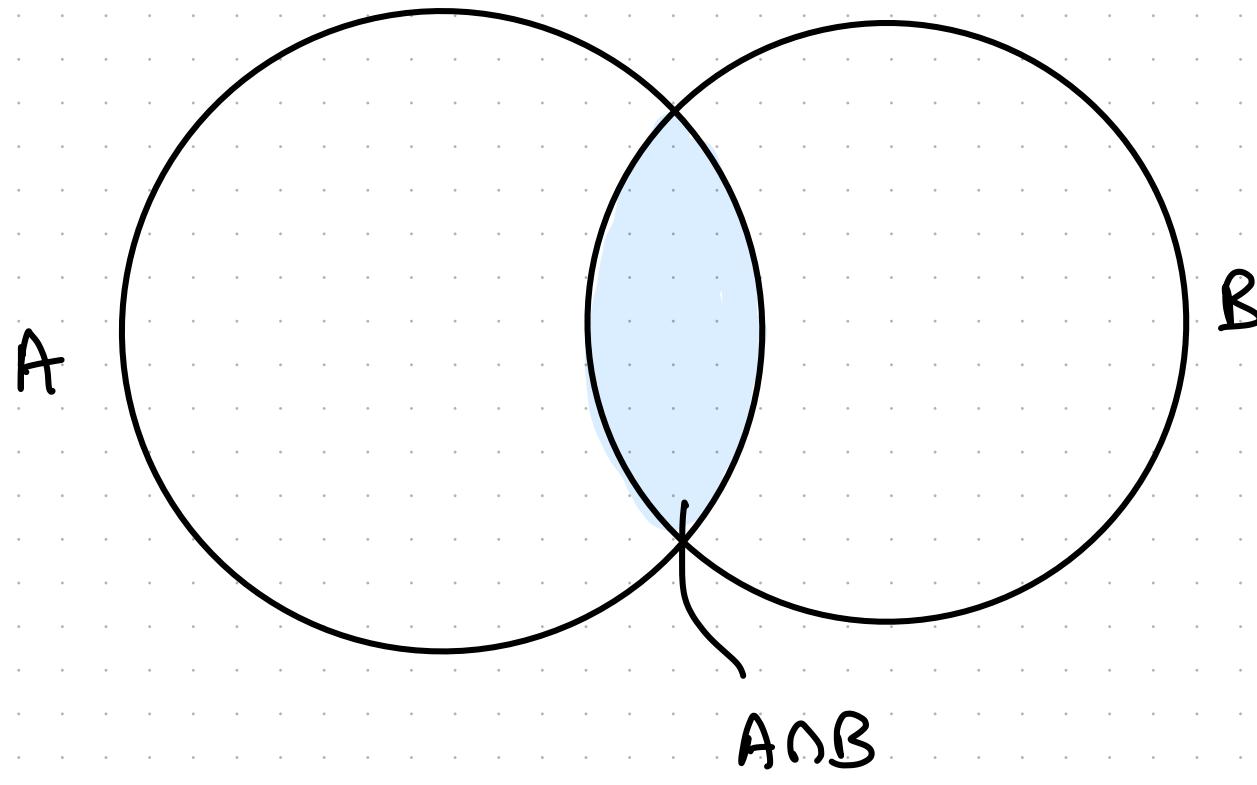


Disjoint vs Independent:

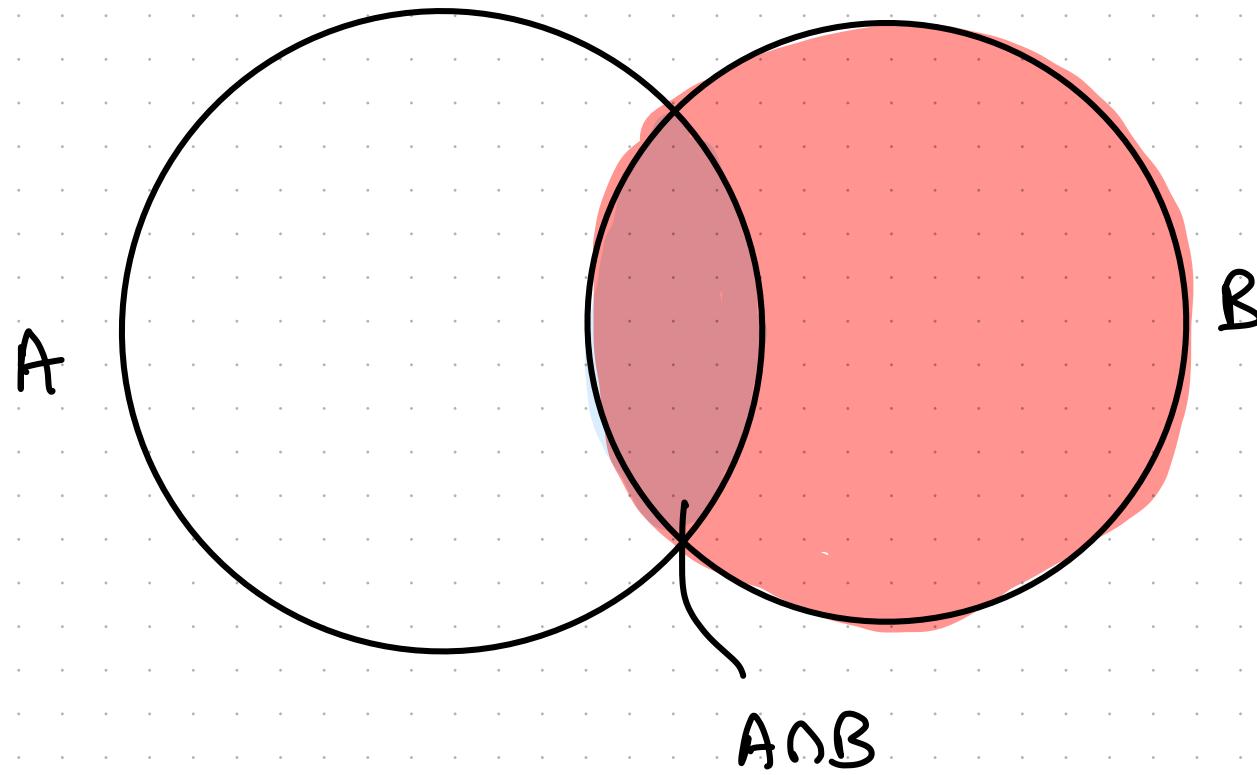
Disjoint  $A \cap B = \emptyset$

$$\Rightarrow P(A \cap B) = 0$$

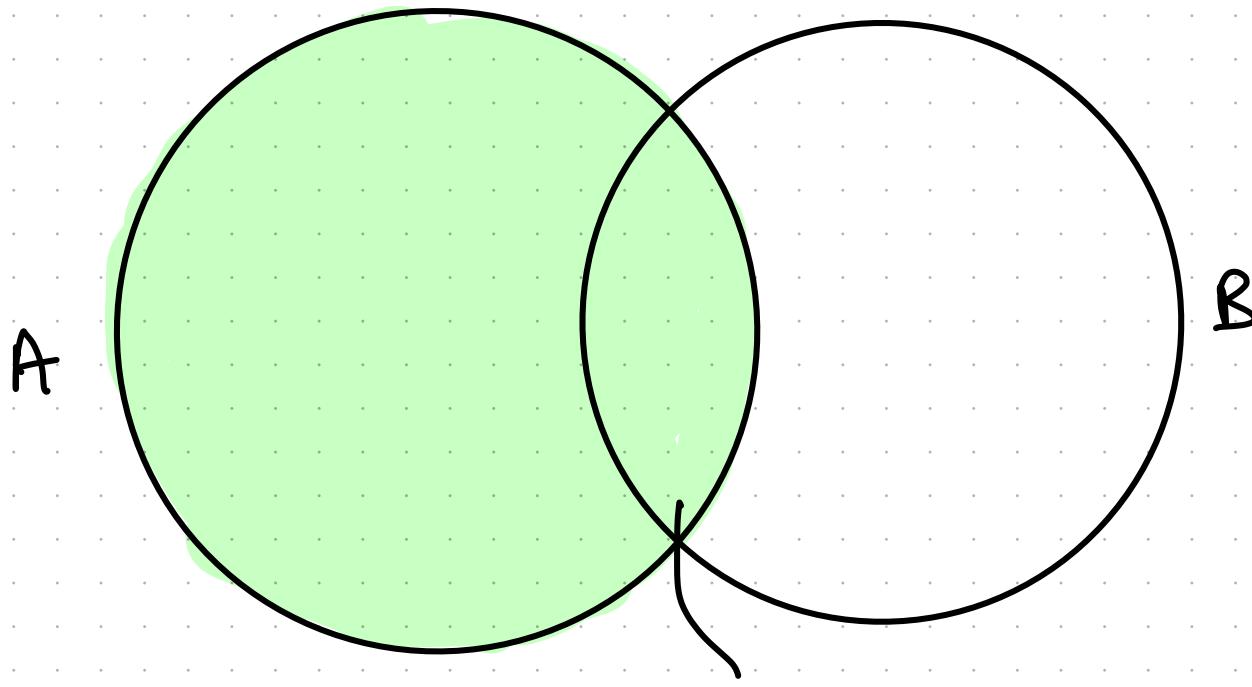
Independent  $P(A \cap B) = P(A) \cdot P(B)$



$$P(A|B) =$$



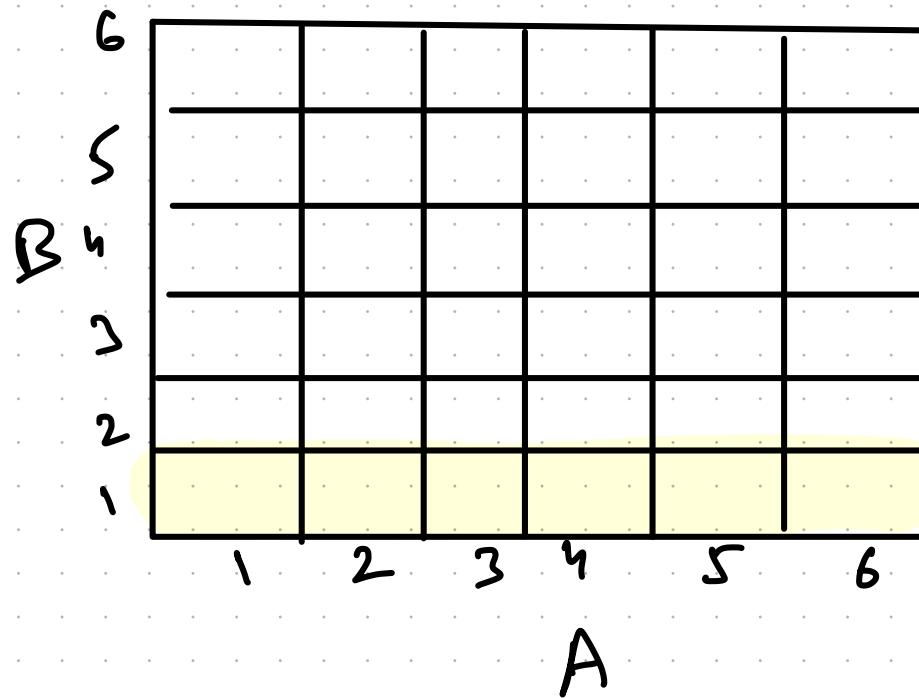
$$P(A|B) = \frac{\underline{\text{Red Area}}}{\underline{\text{Total Area}}}$$



$A \cap B$

$$P(A|B) = \frac{\text{_____}}{\text{_____}} = \frac{P(A)}{P(B)} = \frac{\text{_____}}{1}$$

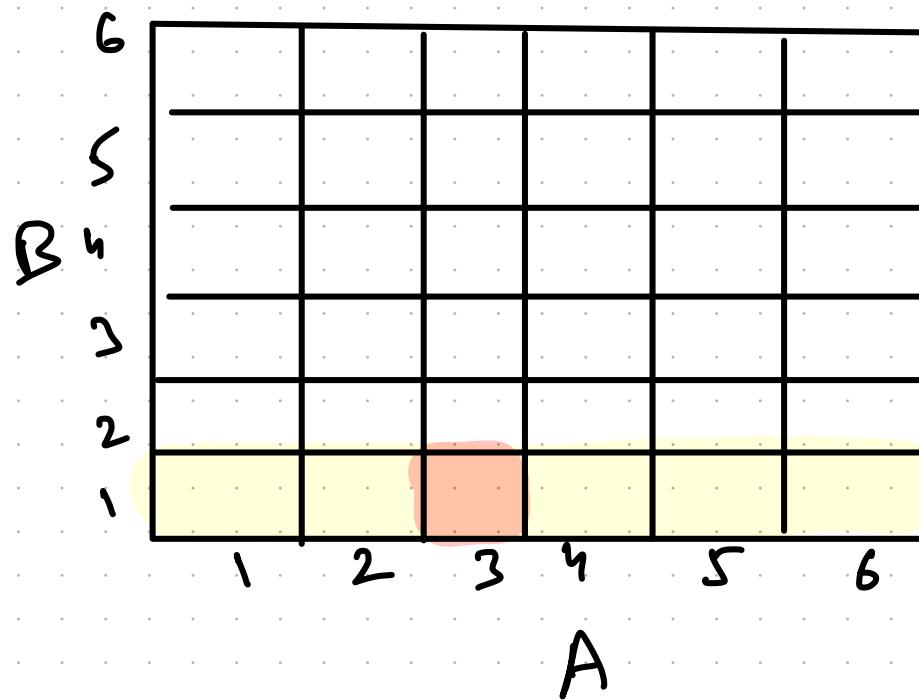
I)  $A = \{1^{\text{st}} \text{ dice is } 3\}$  ;  $B = \{2^{\text{nd}} \text{ dice is } 4\}$



$$P(A) = \frac{1}{6} = \underline{\hspace{2cm}}$$

$$P(B) = \frac{1}{6}$$

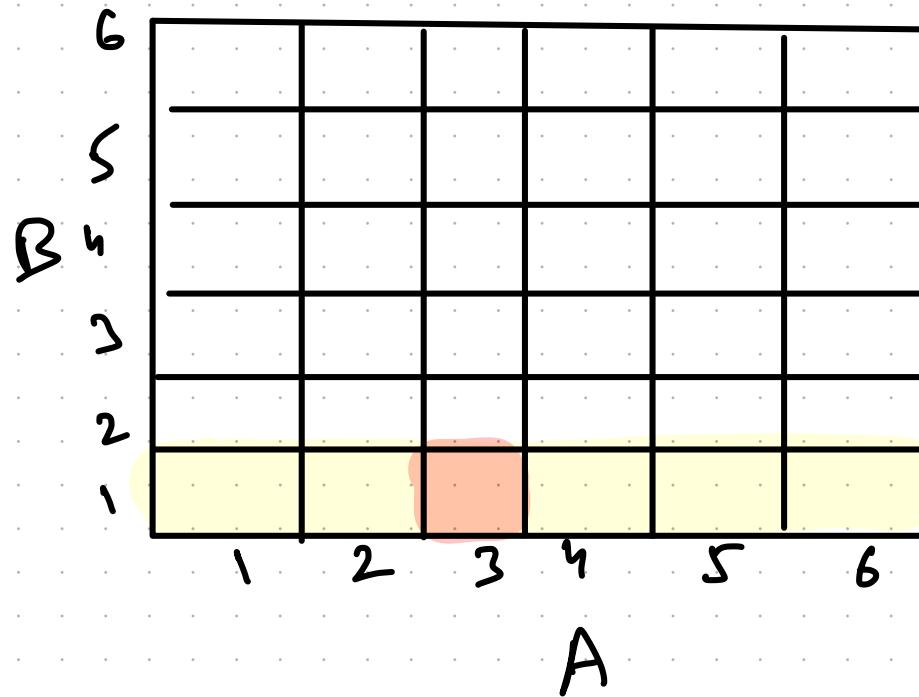
I)  $A = \{1^{\text{st}} \text{ dice is } 3\}$  ;  $B = \{2^{\text{nd}} \text{ dice is } 4\}$



$$P(A) = \frac{1}{6} = \underline{\quad}$$

$$P(B) = \frac{1}{6}$$

I)  $A = \{1^{\text{st}} \text{ dice is } 3\}$ ;  $B = \{2^{\text{nd}} \text{ dice is } 4\}$



$$P(A) = \frac{1}{6} = \underline{\quad}$$

$$P(A \cap B) = \frac{1}{36}$$

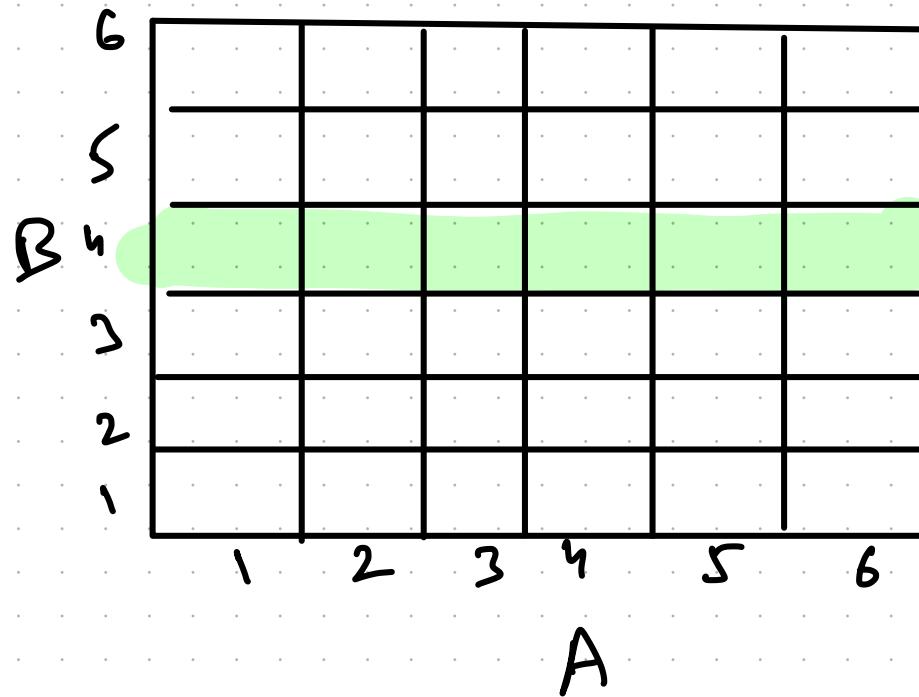
$$P(B) = \frac{1}{6}$$

$$P(A) \cdot P(B) = \frac{1}{36}$$

$$= P(A \cap B)$$

$\therefore A \text{ & } B \text{ are independent.}$

I)  $A = \{1^{\text{st}} \text{ dice is } 3\}$ ;  $B = \{2^{\text{nd}} \text{ dice is } 4\}$



$$P(A|B) = \frac{1}{6} = P(A)$$

II)  $A = \{ \text{1st dice is } 1 \}$

$B = \{ \text{sum of first \& second is } 7 \}$

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$B = \{ \text{sum of first \& second is } 7 \}$

$$P(A) = \frac{1}{6}$$

$B = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$

$$P(B) = \frac{6}{36} = \frac{1}{6}$$

II)  $A = \{ \text{1st dice is } 1 \}$

$B = \{ \text{sum of first \& second is } 7 \}$

$$P(A) = \frac{1}{6}$$

$B = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$

$$P(B) = \frac{6}{36} = \frac{1}{6}$$

$$A \cap B = \{(1,6)\} ; P(A \cap B) = \frac{1}{36}$$

$$P(A) \cdot P(B) = \frac{1}{36} \quad \therefore A \text{ \& } B \text{ are independent}$$

II)  $A = \{ \text{1st dice is } 1 \}$

$B = \{ \text{sum of first \& second is } 7 \}$

$$P(A) = \frac{1}{6}$$

$B = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}$

$$P(B) = \frac{6}{36} = \frac{1}{6}$$

$$P(A|B) = \frac{n(\{(1,6)\})}{n(B)} = \frac{1}{6} = P(A)$$

II)  $A = \{1^{\text{st}} \text{ dice is } 1\}$

$B = \{\text{sum is } 8\}$

$$P(A) = \frac{1}{6}$$

$B = \{(2,6), (3,5), (4,4), (5,3), (6,2)\}$

If  $B$  occurs, can  $A$  occur?

$$P(A|B) = 0$$

$$\neq P(A)$$

$\therefore A \& B$  are dependent

IV)

$$A = \{ \text{1st dice is } 2 \}$$

$$B = \{ \text{sum is } 8 \}$$

$$B = \{ (2, 6), (3, 5), (4, 4), (5, 3), (6, 2) \}$$

$$P(B) = \frac{5}{36} ; P(A) = \frac{1}{6}$$

$$P(A|B) = \frac{n(\{(2, 6)\})}{n(B)} = \frac{1}{5} \neq P(A)$$

$\therefore A \& B$  are dependent

IV)  $A = \{ \text{max is } 2 \} ; B = \{ \text{min is } 2 \}$

$$A = \{(1,2), (2,1), (2,2)\}$$

$$B = \{(2,2), (2,3), (2,4), (2,5), (2,6), (3,2), (4,2), (5,2), (6,2)\}$$

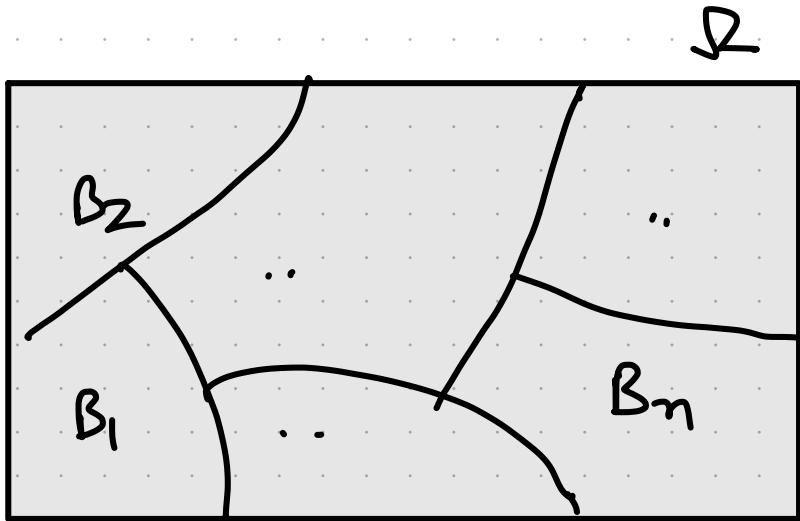
$$A \cap B = \{(2,2)\}$$

$$P(A) = \frac{3}{36} = \frac{1}{12}; P(A|B) = \frac{n(A \cap B)}{n(B)} = \frac{1}{9} \neq P(A)$$

i. i. d.

Sample #	Colour	Radius	Condition
1	0.9	0.7	
2	0.6	0.	
3			
4			
.			
.			

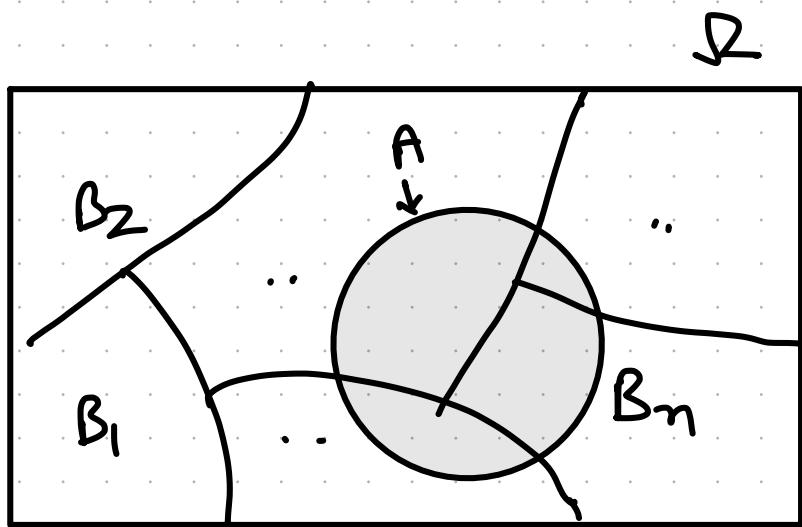
## Total Probability Law



$\{B_1, \dots, B_n\}$  is  
partition on  $\Omega$

- i,  $B_i \cap B_j = \emptyset$  if  $i \neq j$
- ii,  $\bigcup_{i=1}^n B_i = \Omega$

## Total Probability Law

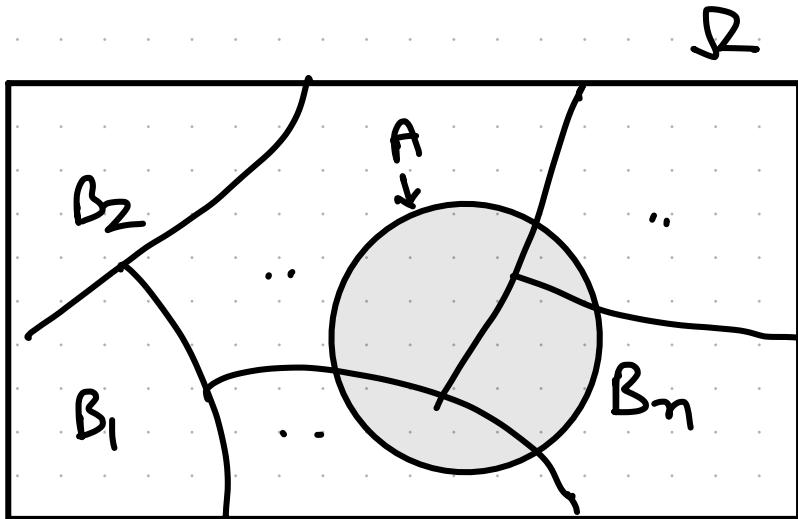


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$$P(A) = \sum_{i=1}^n P(A|B_i) \cdot P(B_i)$$

## Total Probability Law



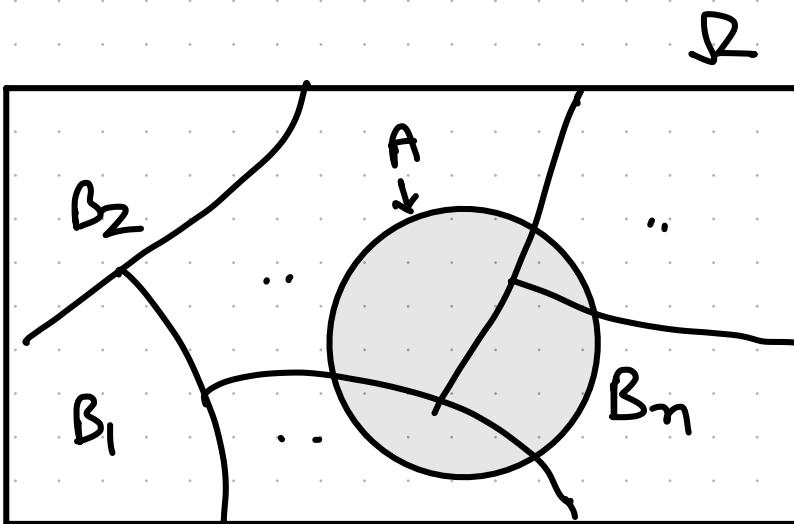
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Prove

## Total Probability Law



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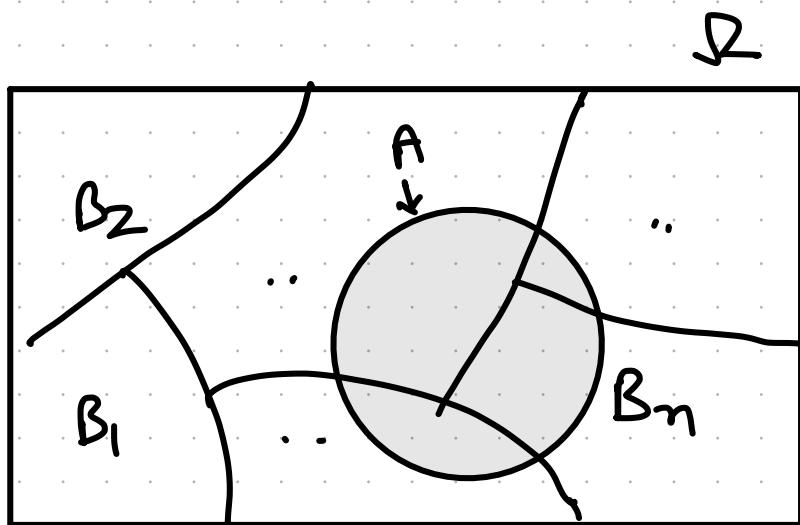
- i,  $B_i \cap B_j = \emptyset$  if  $i \neq j$
- ii,  $\bigcup_{i=1}^n B_i = \Omega$

$$P(A) = \sum_{i=1}^n P(A|B_i) \cdot P(B_i)$$

Prove

$$\begin{aligned} RHS &= \sum_{i=1}^n P(A \cap B_i) &= P[(A \cap B_1) \cup (A \cap B_2) \cup \dots \cup (A \cap B_n)] \\ &&= P[A] \end{aligned}$$

## Total Probability Law



$\{B_1, \dots, B_n\}$  is  
partition on  $\Omega$

- i,  $B_i \cap B_j = \emptyset$  if  $i \neq j$
- ii,  $\bigcup_{i=1}^n B_i = \Omega$

We can also write

$$P(A) = P(A|B) \cdot P(B) + P(A|B^c) \cdot P(B^c)$$

## Q) COVID example

Sensitivity = True Positive Rate =  $P(\text{Test}^+ | \text{covid}) = 0.90$

Specificity = True negative Rate =  $P(\text{Test}^- | \neg \text{covid}) = 0.95$

Probability that Random person has COVID = 0.05

What is  $P(\text{Test}^+)$

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$$P(\text{Test}^+) = P(\text{Test}^+ | \text{covid}) \cdot P(\text{covid}) + P(\text{Test}^+ | \neg \text{covid}) \cdot P(\neg \text{covid})$$

$$= 0.9 \cdot 0.05 + (1 - 0.95) \cdot 0.95$$

$$= 0.045 + 0.0475 = 0.0925$$

Q) Flight delays

$$P(S) = 0.5 ; P(C) = 0.3 ; P(R) = 0.2$$

Sun                      Cloud                      Rain

$$P(D|S) = 0.1 ; P(D|C) = 0.3 ; P(D|R) = 0.6$$

$$P(D) = ? \quad (\text{Any random day flight delay})$$

Q) Flight delays

$$P(S) = 0.5 ; P(C) = 0.3 ; P(R) = 0.2$$

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$$P(D|S) = 0.1 ; P(D|C) = 0.3 ; P(D|R) = 0.6$$

$P(D) = ?$  (Any random day flight delay)

$$\begin{aligned} P(D) &= P(D|S) \cdot P(S) + P(D|R) \cdot P(R) + P(D|C) \cdot P(C) \\ &= 0.5 \cdot 0.1 + 0.6 \cdot 0.2 + 0.3 \cdot 0.3 \\ &= 0.26 \end{aligned}$$

## Q) COVID example

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Probability that Random person has COVID = 0.05

What is  $P(\text{Test}^+)$

$P(\text{covid}) = 0.05$ ;  $P(\neg \text{covid}) = 0.95$

$$P(\text{Test}^+) = 0.0925$$

$$P(\text{covid} | \text{Test}^+) = ?$$

## Bayes Theorem

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

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## Bayes Theorem

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

$$P(\text{COVID} | \text{Test}^+) = \frac{P(\text{Test}^+ | \text{COVID}) \cdot P(\text{COVID})}{P(\text{Test}^+)}$$

$P(\text{COVID})$  = Prior probability of COVID w/o Test

$P(\text{COVID} | \text{Test}^+)$  = Posterior probability after Test  
(updated belief)

## Bayes Theorem

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

$$P(\text{COVID} | \text{Test}^+) = \frac{P(\text{Test}^+ | \text{COVID}) \cdot P(\text{COVID})}{P(\text{Test}^+)}$$

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Probability that Random person has COVID = 0.05

$$P(\text{Test}^+) = 0.9725$$

## Bayes Theorem

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

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Probability that Random person has COVID = 0.05

$$P(\text{Test}^+) = 0.9725$$

$$P(\text{COVID} | \text{Test}^+) = \frac{0.90 \times 0.05}{0.9725} = 0.486$$

## Bayes Theorem

$$P(\text{COVID} | \text{Test}^+) = \frac{P(\text{Test}^+ | \text{COVID}) \cdot P(\text{COVID})}{P(\text{Test}^+)}$$

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(App)

## Monty Hall Problem

3 doors

- one door has a car behind it
- goat behind other two
- you initially choose a door
- host (Monty Hall) opens one of remaining door  
that has a goat
- Question
  - Stay with your door
  - Switch

## Monty Hall Problem

- Say you chose Door 1.

- Define

$C_1$  : Car behind Door 1

$C_2$  : " .. Door 2

$C_3$  : " .. Door 3

## Monty Hall Problem

- Say you chose Door 1.

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$C_1$  : Car behind Door 1

$C_2$  : " .. Door 2

$C_3$  : " .. Door 3

$$P(C_1) = P(C_2) = P(C_3) = \frac{1}{3}$$

## Monty Hall Problem

- Say you chose Door 1.

- Define

$C_1$  : Car behind Door 1

$C_2$  : " .. Door 2

$C_3$  : " .. Door 3

- Hall opens

- Door 1  $\rightarrow H_1$

2  $\rightarrow H_2$

3  $\rightarrow H_3$

- Hall never opens door with car

## Monty Hall Problem

- you picked Door 1
- Case I
  - Car behind Door 1
  - Hall chooses D<sub>2</sub> or D<sub>3</sub> with equal prob
  - $\Rightarrow P(H_2 | C_1) = P(H_3 | C_1) = 0.5$

## Monty Hall Problem

- you picked Door 1
- Case I
  - Car behind Door 1
  - Hall chooses D<sub>2</sub> or D<sub>3</sub> with equal prob
  - $P(H_2 | C_1) = P(H_3 | C_1) = 0.5$
- Case II
  - Car behind Door 2
  - Hall has to open Door 3
  - $\Rightarrow P(H_2 | C_2) = 0 ; P(H_3 | C_2) = 1$

## Monty Hall Problem

- you picked Door 1
- Case I
  - Car behind Door 1
  - Hall chooses  $D_2$  or  $D_3$  with equal prob
  - $\Rightarrow P(H_2 | C_1) = P(H_3 | C_1) = 0.5$
- Case II
  - Car behind Door 2
  - Hall has to open Door 3
  - $\Rightarrow P(H_2 | C_2) = 0 ; P(H_3 | C_2) = 1$
- Case III
  - Car behind Door 3
  - $P(H_2 | C_3) = 1 ; P(H_3 | C_3) = 0$

## Monty Hall Problem

- you picked Door 1
- let's assume Hall opened Door 3

## Monty Hall Problem

- you picked Door 1
- let's assume Hall opened Door 3
- Question

$$P(C_2 | H_3) = ? \quad (\text{switch})$$

## Monty Hall Problem

- you picked Door 1
- let's assume Hall opened Door 3
- Question

$$P(C_2 | H_3) = ? \quad (\text{switch})$$

$$P(C_1 | H_3) = ? \quad (\text{stay})$$

$$P(C_1) = \frac{1}{3} \quad (\text{no data on door})$$

## Monty Hall Problem

$$P(C_2 | H_3) = ? \quad (\text{switch})$$

$$P(C_2 | H_3) = \frac{P(H_3 | C_2) \cdot P(C_2)}{P(H_3)}$$

Now  $P(H_3) =$

## Monty Hall Problem

$$P(C_2 | H_3) = ? \quad (\text{switch})$$

$$P(C_2 | H_3) = \frac{P(H_3 | C_2) \cdot P(C_2)}{P(H_3)}$$

$$\text{Now } P(H_3) = P(H_3 | C_1) \cdot P(C_1) + P(H_3 | C_2) \cdot P(C_2) \\ + P(H_3 | C_3) \cdot P(C_3)$$

$$= 0.5 \times \frac{1}{3} + 1 \times \frac{1}{3} + 0 \times \frac{1}{3}$$

$$= \frac{1}{2} \times \frac{1}{3} + 1 \times \frac{1}{3} = \frac{3}{2} \times \frac{1}{3} = \frac{1}{2}$$

## Monty Hall Problem

$$P(C_2 | H_3) = ? \quad (\text{switch})$$

$$P(C_2 | H_3) = \frac{P(H_3 | C_2) \cdot P(C_2)}{P(H_3)}$$

$$P(H_3) = \frac{1}{2}$$

$$P(C_2 | H_3) = \frac{1 \times \frac{1}{3}}{\frac{1}{2}} = \frac{2}{3} = 66.7\%$$

# Monty Hall Problem

$$P(C_2 | H_3) = ? \quad (\text{switch})$$

$$P(C_2 | H_3) = \frac{P(H_3 | C_2) \cdot P(C_2)}{P(H_3)}$$

$$P(H_3) = \frac{1}{2}$$

$$P(C_2 | H_3) = \frac{1 \times \frac{1}{3}}{\frac{1}{2}} = \frac{2}{3} = 66.7\%$$

$$P(C_1 | H_3) = \frac{1}{3} = 33.3\%$$

## Monty Hall Problem

$$P(C_1) = 33.3\%$$

$$P(C_2 | H_3) = 66.7\%$$

$$P(C_1 | H_3) = 33.3\%$$

$\therefore$  Presented w/ new data,  
switching to Door 2 is better

