

Conditional Probability

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

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

Why are A and B not same?

Conditional Probability

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Why are A and B not same?
Ground Truth (Actual)

		No Covid	HAS COVID
Prediction/Test	No Covid		
	HAS covid		

Conditional Probability

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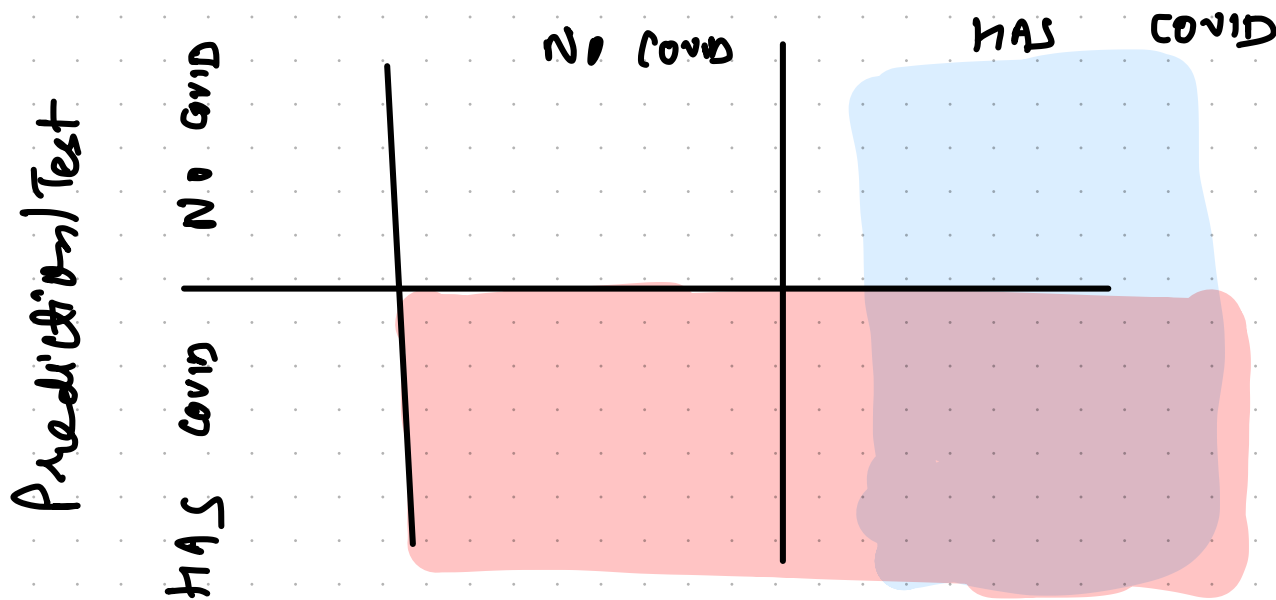
		No Covid	HAS COVID
Prediction/Test	No Covid		
	HAS covid		

Conditional Probability

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Why are A and B not same?
Ground Truth (Actual)



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

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

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

$$A^c \cap B^c = \{ 5, 11, 12 \}$$

$$A^c \cap B = \{ 6, 9, 10 \}$$

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

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

Confusion Matrix

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

Conditional Probability

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

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I really

TP, TN high

FN, FP low

Confusion Matrix

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

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

$$P[A] =$$

$$P[B] =$$

$$P[A \cap B]$$

$$P[A|B] =$$

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

Conditional Probability

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

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

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

Conditional Probability

$$A = \{ \text{greater than 3} \}$$

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

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

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

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

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

Conditional Probability

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

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

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

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

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

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

4 sided die

X : first roll ; Y : second roll

B : $\min(X, Y) = 2$; $M = \max(X, Y) = 3$

$P[M|B] = ?$

4 sided die

x : first roll ; y : second roll

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

$P[M|B] = ?$

4				
3				
2				
1				
	1	2	3	4

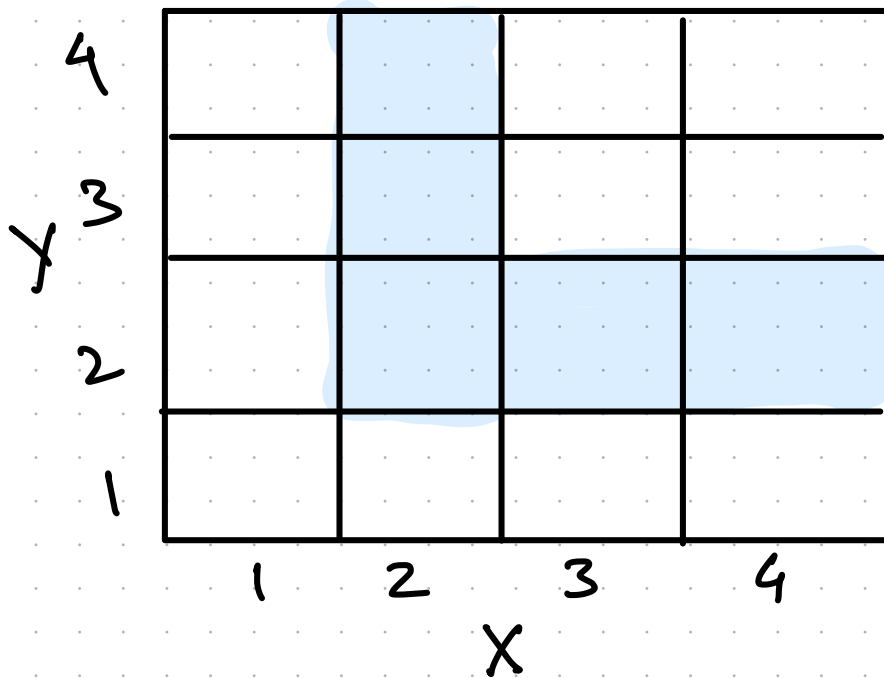
x

4 sided die

x : first roll ; y : second roll

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

$P[M|B] = ?$



Marking B

examples

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

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

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

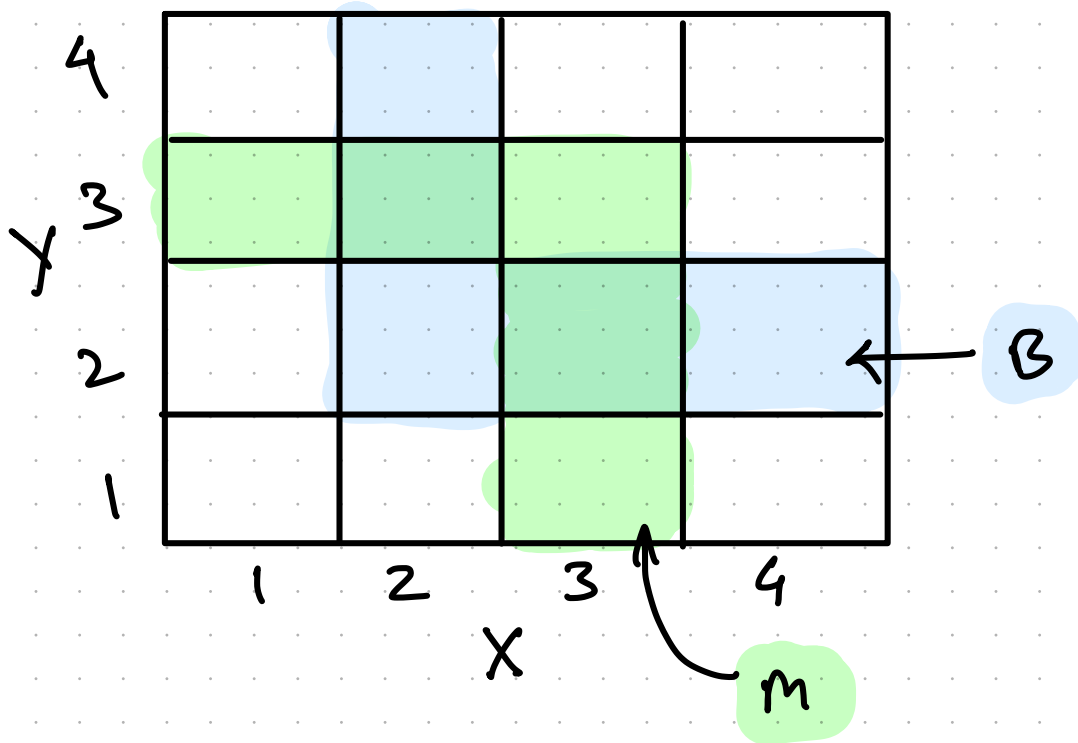
$$\min(4, 3) = 3 \neq 2$$

4 sided die

x : first roll ; y : second roll

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

$P[M|B] = ?$

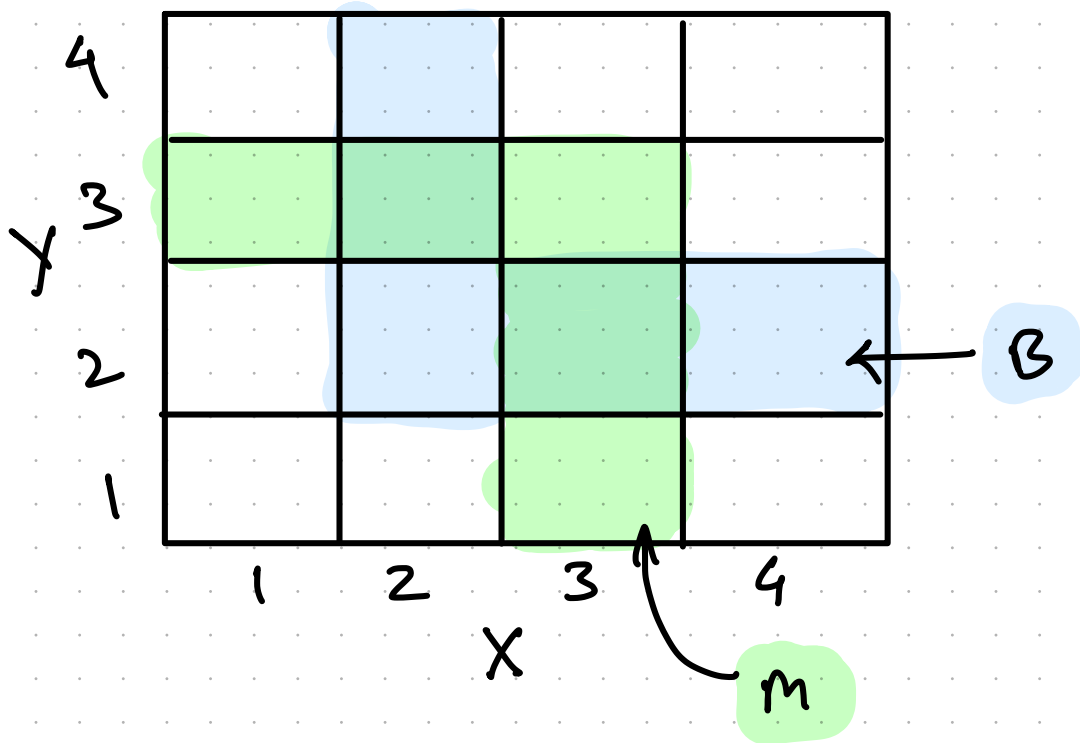


4 sided die

x : first roll ; y : second roll

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

$P[M \cap B] = ?$



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

$$n(B) = 5$$

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

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 $P[B] > 0$; $P[A|B]$ 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: $P(\Omega | B) = 1$

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

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

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

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

3) Disjoint A and C

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

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

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

* 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.

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

Independence

Two events A and B are independent if:

Knowing one occurred gives no extra information about the 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:

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

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

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

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

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

If 1st coin is H; 2nd still has 1/2 prob. of H

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

For 2 coin toss:

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

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

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

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

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

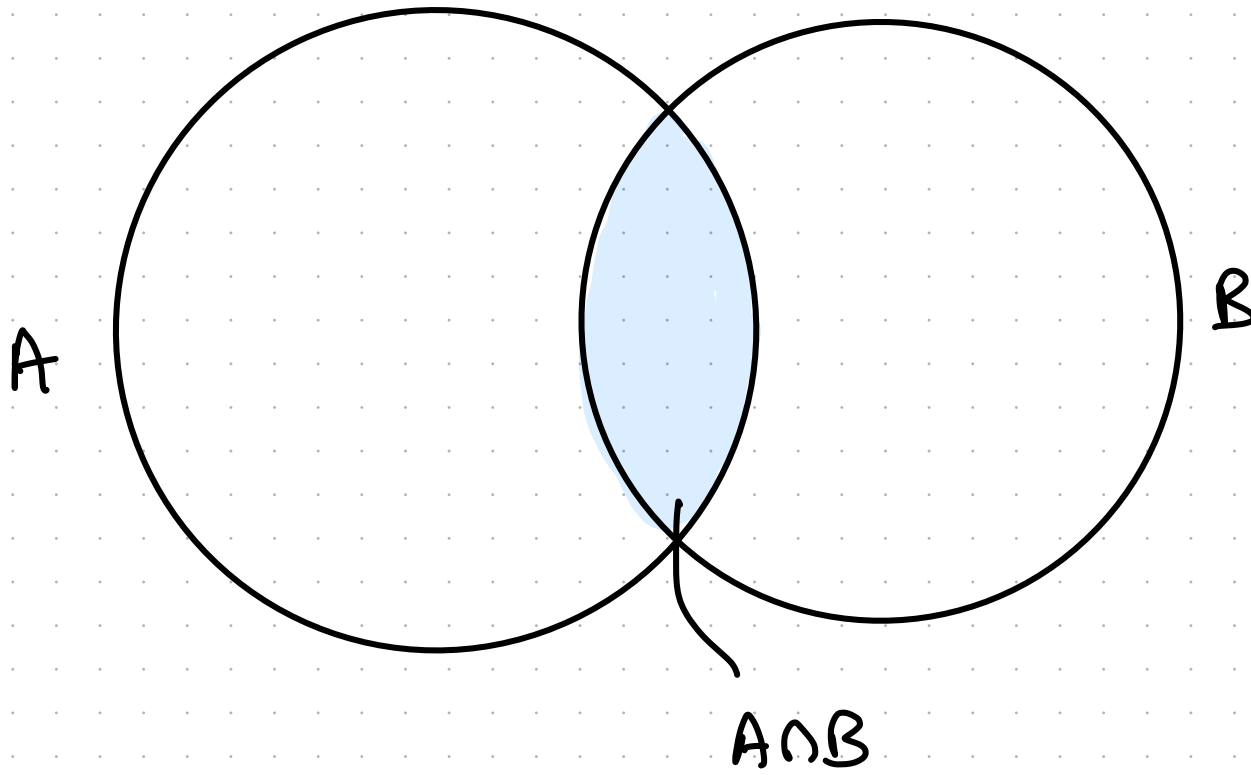
$$P(A|B) = 1/2$$

$$= \frac{\quad}{\quad}$$

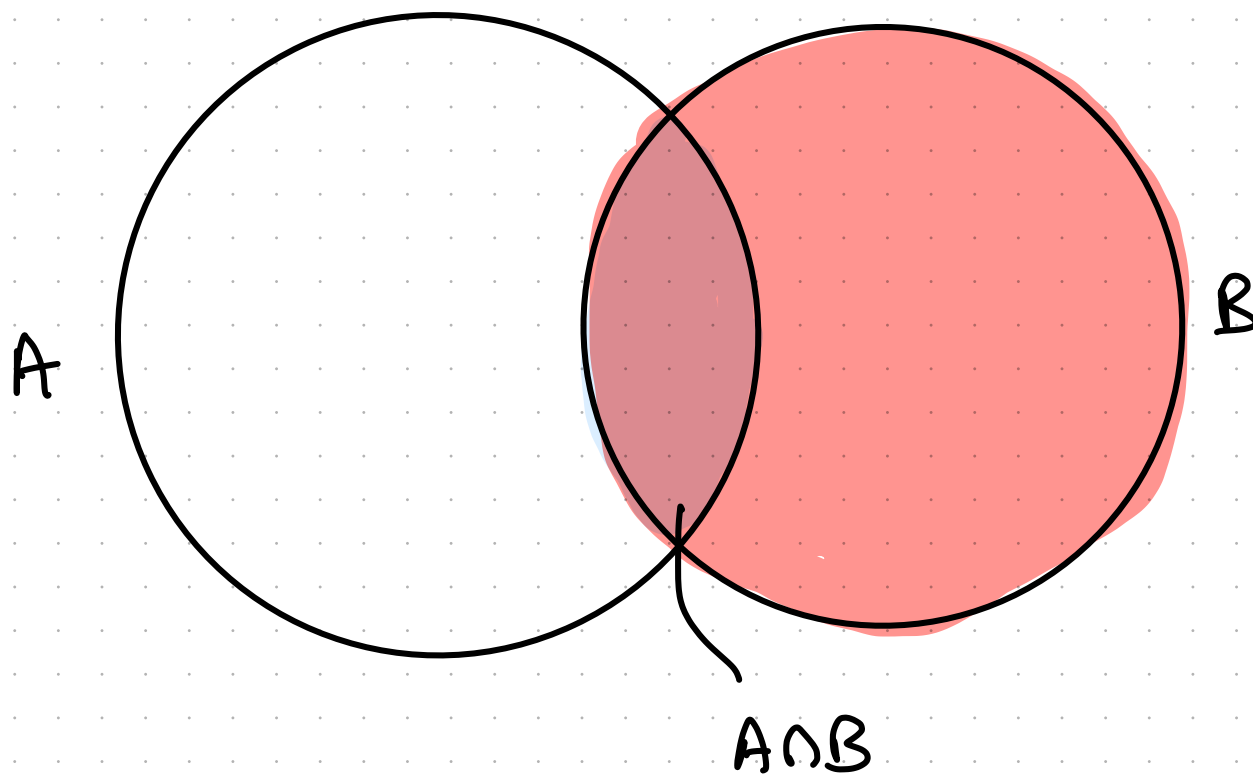
Disjoint vs Independent

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

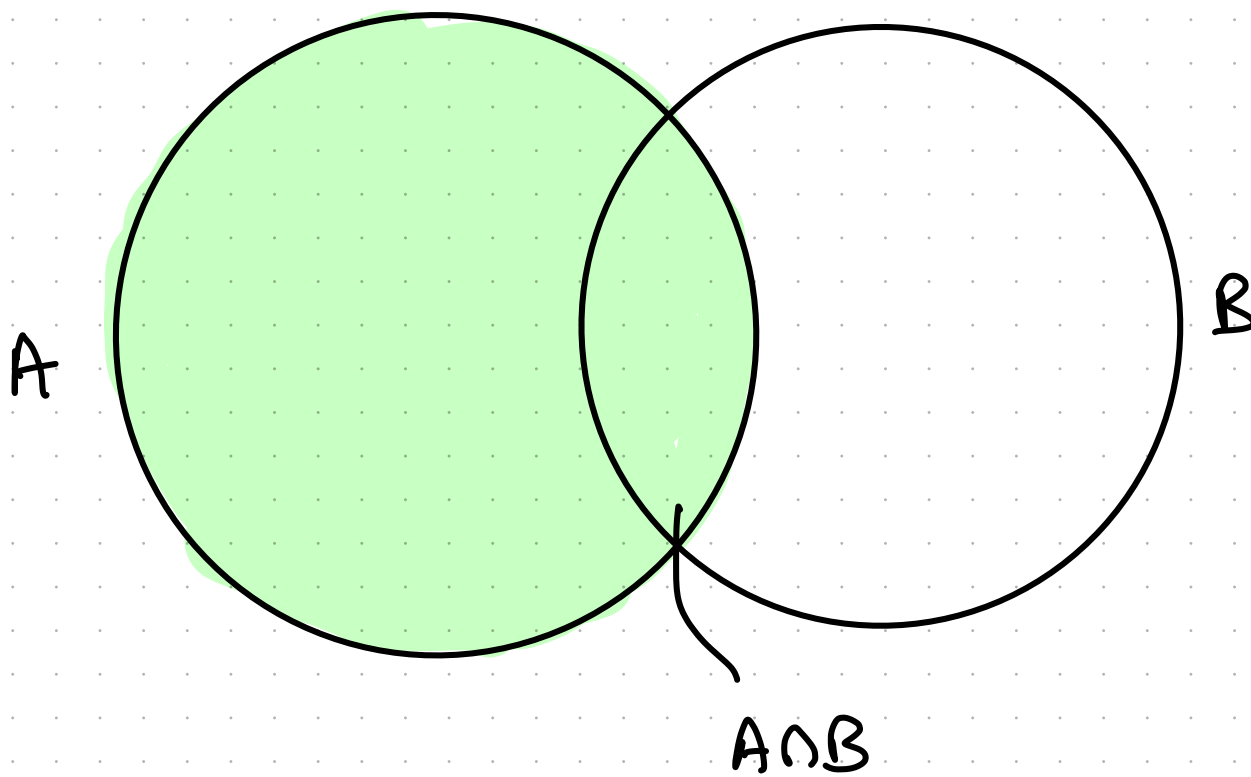
Independent $P(A \cap B) = P(A) \cdot P(B)$



$$P(A|B) =$$

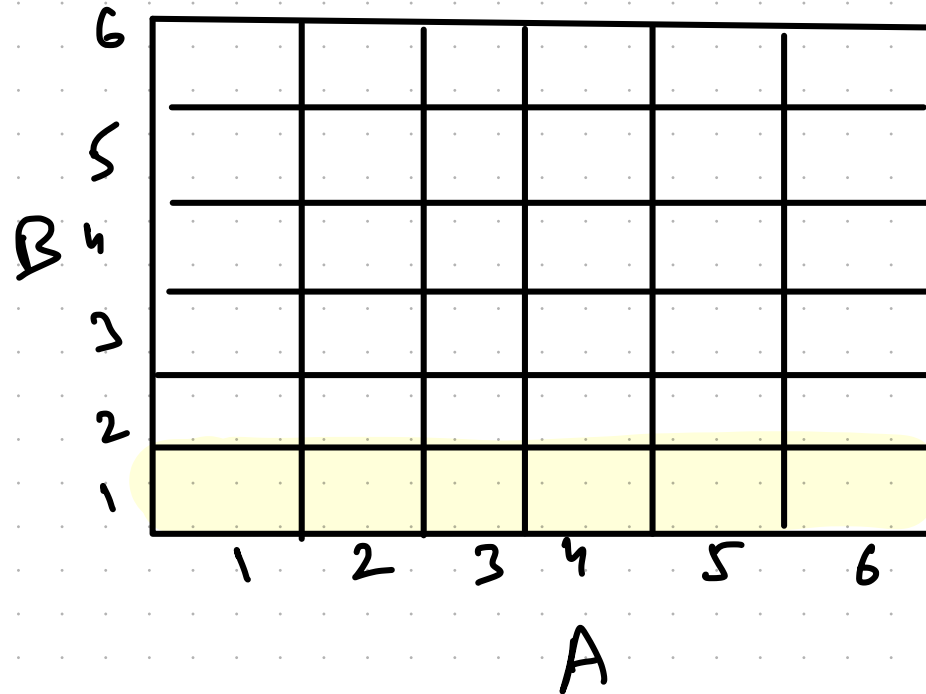


$$P(A|B) = \frac{\text{light blue bar}}{\text{red bar}}$$



$$P(A|B) = \frac{\text{blue bar}}{\text{red bar}} = \frac{P(A)}{P(\Omega)} = \frac{\text{green bar}}{1}$$

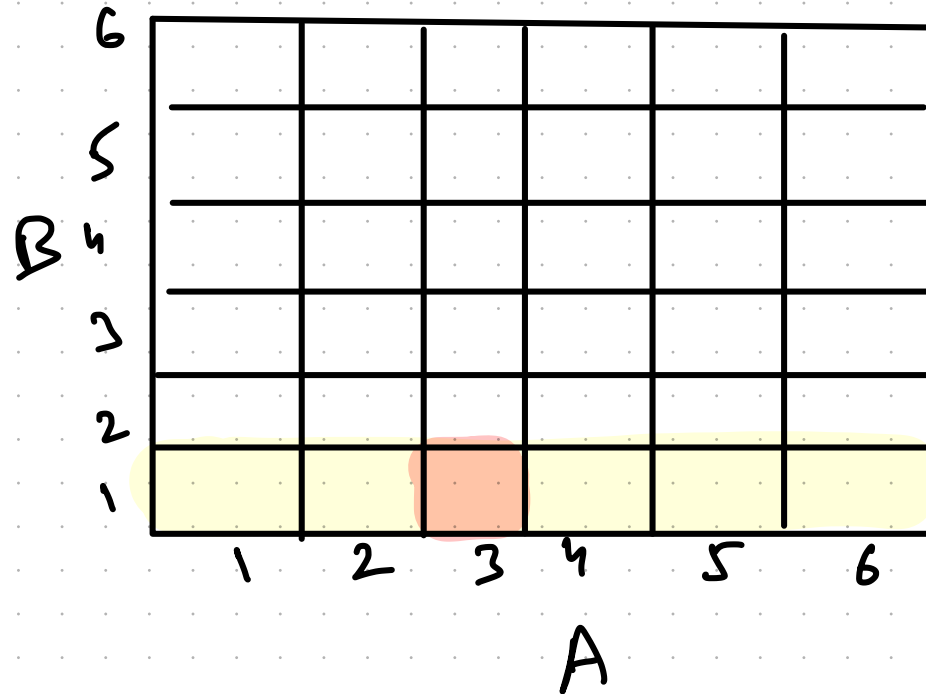
I) $A = \{1^{\text{st}} \text{ dice is } 3\}$; $B = \{2^{\text{nd}} \text{ dice is } 4\}$



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

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

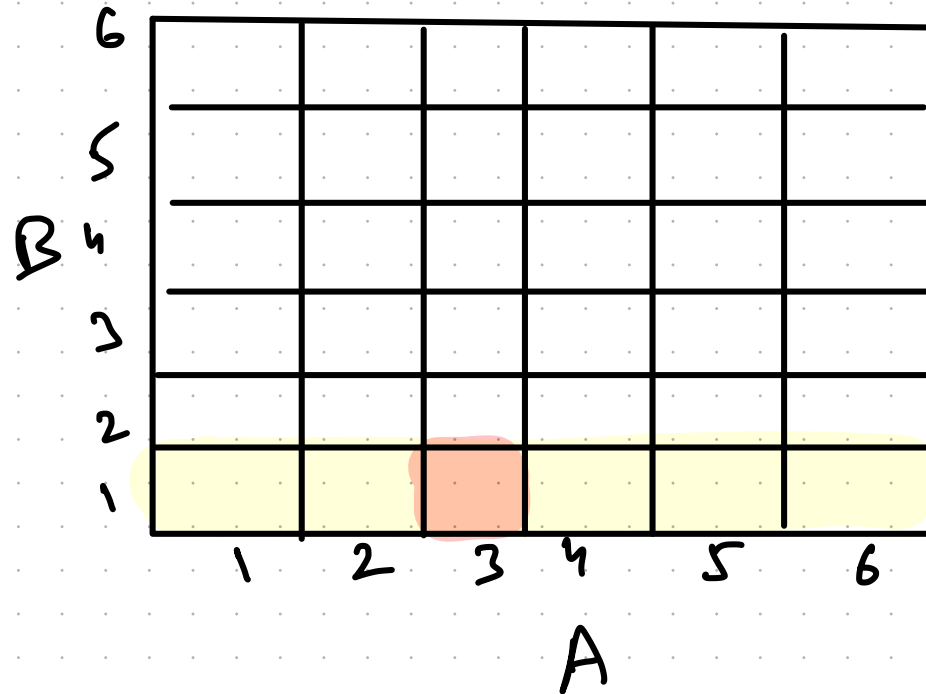
I) $A = \{1^{\text{st}} \text{ dice is } 3\}$; $B = \{2^{\text{nd}} \text{ dice is } 4\}$



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

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

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



$$P(A) = \frac{1}{6} = \frac{\text{Yellow}}{\text{Yellow + Orange}}$$

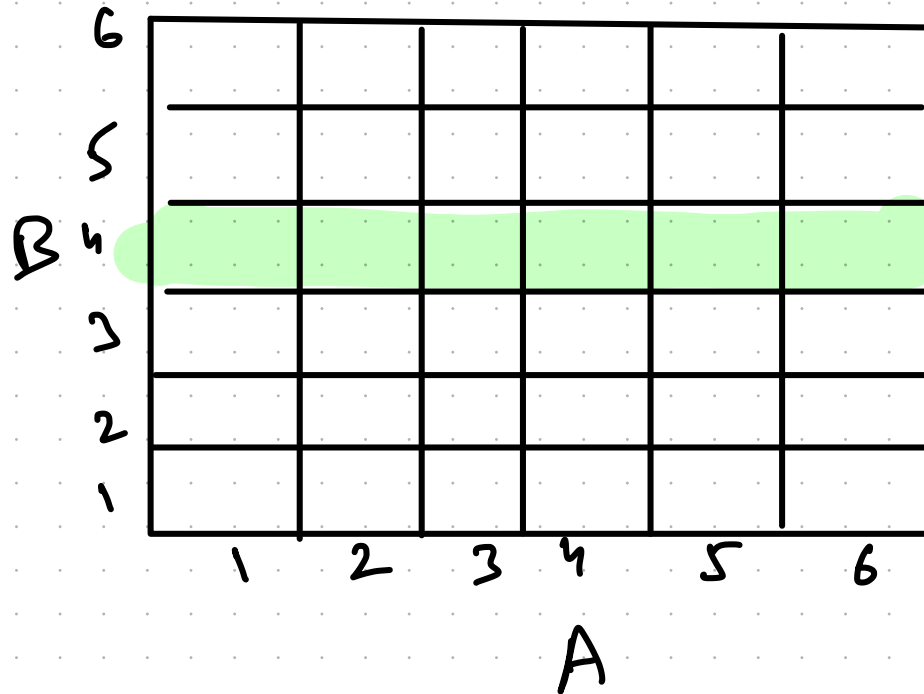
$$P(B) = \frac{1}{6}$$

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

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

$\therefore A$ & B are independent.

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



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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

$$\underline{\text{V)}} \quad A = \{ \text{max is } 2 \} \quad ; \quad B = \{ \text{min is } 2 \}$$

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

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

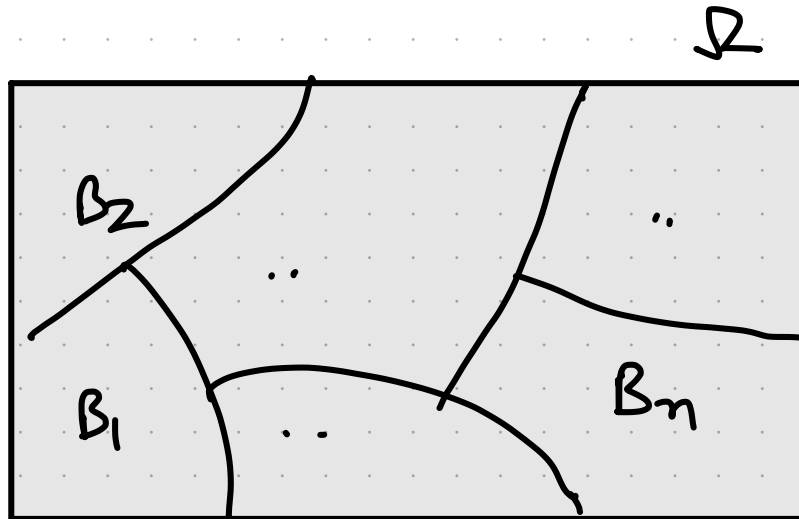
$$A \cap B = \{ (2, 2) \}$$

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

i. i. d.

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

Total Probability Law

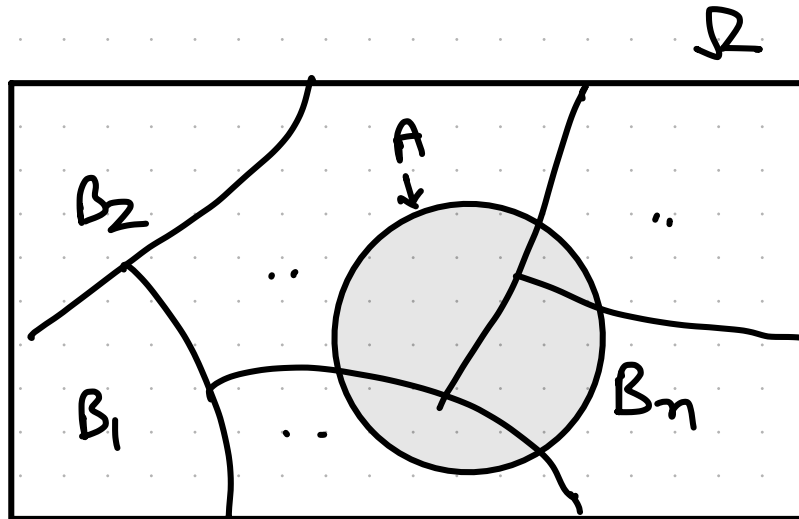


$\{B_1, \dots, B_n\}$ is
partition on Ω

(i) $B_i \cap B_j = \emptyset$ if $i \neq j$

(ii) $\bigcup_{i=1}^n B_i = \Omega$

Total Probability Law

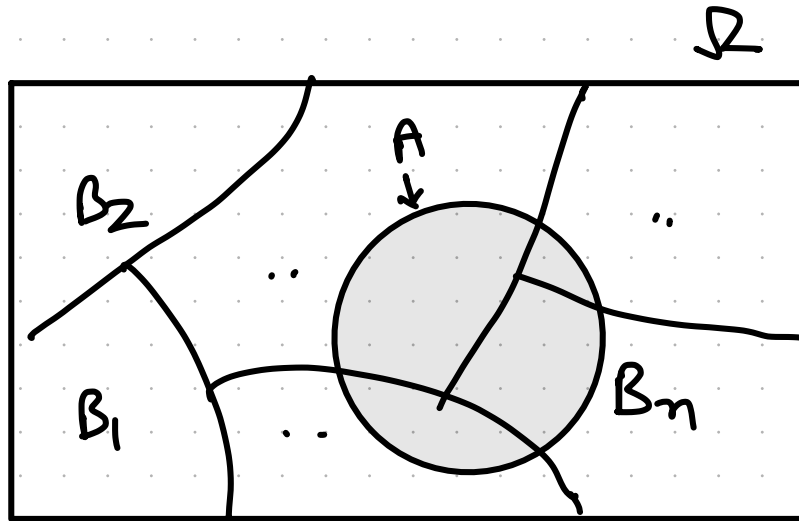


$\{B_1, \dots, B_n\}$ is
partition on Ω

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

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

Total Probability Law



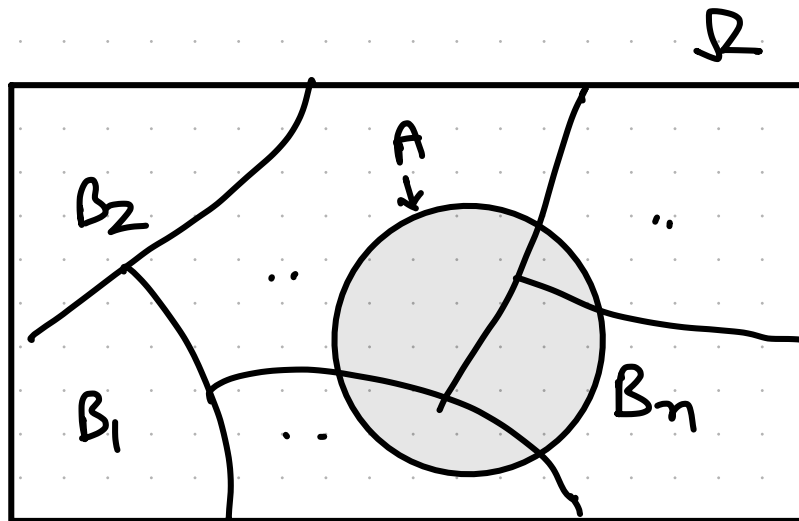
$\{B_1, \dots, B_n\}$ is
partition on Ω

- (i) $B_i \cap B_j = \emptyset$ if $i \neq j$
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$$P(A) = \sum_{i=1}^n P(A|B_i) \cdot P(B_i)$$

Prove

Total Probability Law



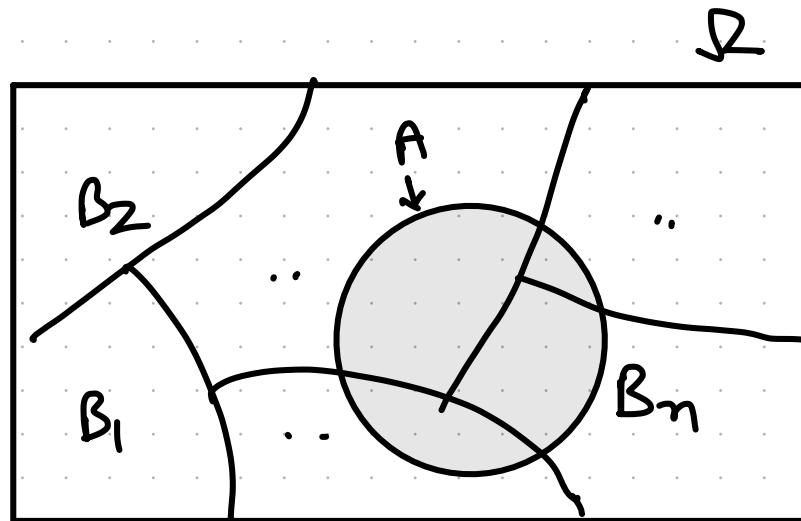
$\{B_1, \dots, B_n\}$ is
partition on Ω

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

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

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

Total Probability Law



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

- (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 $P(\text{Test}^+)$

8) 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 $P(\text{Test}^+)$

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

8) 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 $P(\text{Test}^+)$

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

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

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

$= 0.045 + 0.0475 = 0.0925$

Q) Flight delays

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

Sun Cloud Rain

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

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

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$$\begin{aligned} P(D) &= P(D|S) \cdot P(S) + P(D|R) \cdot P(R) + P(D|C) \cdot P(C) \\ &= .5 * .1 + .6 * .2 + .3 * .3 \\ &= .26 \end{aligned}$$

Q) COVID example

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

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

$P(\text{COVID})$ = 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

$$P(\text{Test}^+) = 0.09725$$

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Probability that Random person has COVID = 0.05

$$P(\text{Test}^+) = 0.09725$$

$$P(\text{COVID} | \text{Test}^+) = \frac{0.90 \times 0.05}{0.09725} = 0.486$$

Bayes Theorem

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

Sensitivity = TRUE POSITIVE RATE = $P(\text{Test}^+ | \text{COVID}) = 0.90$

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$$P(\text{COVID} | \text{Test}^+) = \frac{0.90 \times 0.05}{0.09725} = 0.486$$

(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.

- Define

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_2 or D_3 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_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

$$\Rightarrow P(H_2 | C_2) = 0 \quad ; \quad P(H_3 | C_2) = 1$$

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

$$\Rightarrow P(H_2 | C_3) = 0 ; P(H_3 | C_3) = 1$$

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

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

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

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

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

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

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

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

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

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

∴ Presented w/ new data,
switching to Door 2 is better

