

Probability and Statistics

UG2, Core course, IIIT,H

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- ① Motivation for Probability
- ② Classical Probability
- ③ Digress: Game Theory
- ④ Random Walks

- ⑤ Conditional Probability, Bayes Theorem
- ⑥ Define Conditional Probability and Chain Rule
- ⑦ The Monty Hall Problem
- ⑧ Independence
- ⑨ Conditional Independence

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1 Why do we need Probability?

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What is probability?

Probability is a **measure** that we use to measure how likely an event or outcome is. For example, today there is 10% chance of rain.

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How likely is that one player or team will win the game?



1 Motivation for Probability...

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How likely is that one will win the lottery?



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How likely is that one will win the lottery?



How likely it is to rain or snow?



1 Study of Probability started with Games of Chance...

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1 Game of Chance, Probability, and War

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Figure: illustration of great war

- Mahabharat, happened around 900 BCE (Proof dates to 400BCE)

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Figure: illustration of great war

- Mahabharat, happened around 900 BCE (Proof dates to 400BCE)
- Struggle between two major group of cousins: Adventure, Drama, Suspense, Thriller, etc
- ... and a game of chance!

1 What Does Game of Chance has to do with Mahabharat in India?

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- It is a game of Pasha, similar to modern ludo
- In ludo, there is one dice with 6 sides
- There are two cuboidal dices with number of dots on them
- The move is determined by a randomly throwing the two dices

1 Game of Chance and Early History of Probability

| 8



Figure: Cardino, Italy

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- Early theory of probability arose from games of chance played in Europe

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- Early theory of probability arose from games of chance played in Europe
- On the left is Cardano, an Italian mathematician, who studied game of chance
- He gambled for about 25 years!
- His work on probability were published in famous 15 page “Liber de Ludo Aleae”

Figure: Cardano, Italy

1 Game of Chance and Early History of Probability...

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- 
- A black and white engraving of Girolamo Cardano, showing him from the chest up in profile, facing left. He has a prominent mustache and is wearing a dark, buttoned-up coat over a white collared shirt.
- Cardano's contributions have led some to consider Cardano as the real father of probability.



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- Oystein Ore's biography of Cardano, titled *Cardano: The Gambling Scholar*:

... I have gained the conviction that this pioneer work on probability is so extensive and in certain questions so successful that it would seem much more just to date the beginnings of probability theory from Cardano's treatise rather than the customary reckoning from Pascal's discussions with his gambling friend de Méré and the ensuing correspondence with Fermat, all of which took place at least a century after Cardano began composing his *De Ludo Aleae*.

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Trial

When we repeat a random experiment several times, we call each one of them a **trial**.

1 Examples of Experiments and Samples Spaces

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Sample space: $S = \{0, 1, 2, 3, \dots\}$.

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So there is one general rule, namely, that we should consider the whole circuit, and the number of those casts which represents in how many ways the favorable result can occur, and compare that number to the rest of the circuit, and according to that proportion should the mutual wagers be laid so that one may contend on equal terms.

*If favorable result occurs
= $\frac{\# \text{ no. of possibilities}}{\# \text{ of outcomes}}$*

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2 Definition of Classical Probability...



Figure: Pascal, France



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 - Engineering: Tidal Dynamics
 - Mathematics: Laplace equation, Laplace Transform
 - Statistics: Bayesian Interpretation
 - Physics: Existence of black holes, Gravitational collapse, stability of solar, Speed of sound, Surface tension, etc

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2 Definition of Classical Probability...the way it evolved

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equally likely

La théorie des hasards consiste à réduire tous les événemens du même genre, à un certain nombre de cas également possibles, c'est-à-dire, tels que nous soyons également indécis sur leur existence; et à déterminer le nombre des cas favorables à l'événement dont on cherche la probabilité. Le rapport de ce nombre à celui de tous les cas possibles, est la mesure de cette probabilité qui n'est ainsi qu'une fraction dont le numérateur est le nombre des cas favorables, et dont le dénominateur est le nombre de tous les cas possibles.

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Figure: Book: Théorie analytique des probabilités, Laplace, 1812

Translation of last line: The probability of an event is a **fraction** whose **numerator** is the number of **favourable cases** and whose **denominator** is **total number of cases**. In the first line, he also mentions that all the events are **equally possible** (or equally likely).

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- Is $p_i = 0$ possible?

2 Finite Probability Space with Equally Likely Outcomes

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- If A is any event with cardinality $|A| = M$, then

$$P(A) = \frac{|A|}{|S|} = \frac{M}{N}$$

2 Probability and Axioms of Probability

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Probability Axioms...

 **Axiom 1:** For any event A , $P(A) \geq 0$

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Probability Axioms...

- **Axiom 1:** For any event A , $P(A) \geq 0$
- **Axiom 2:** Probability of the sample space S is $P(S) = 1$
- **Axiom 3:** If A_1, A_2, A_3, \dots are disjoint events, then
$$P(A_1 \cup A_2 \cup A_3 \dots) = P(A_1) + P(A_2) + P(A_3) + \dots$$

2 Quiz

Quiz

Prove the following:

- For any event A , $P(A^c) = 1 - P(A)$

$$S = A \cup A^c$$

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Axiom-2:

$$P(S) = P(A \cup A^c) \xrightarrow{\text{Axiom-2}} 1$$

A^c are disjoint,

Axiom-3:

Since A and A^c are disjoint,

$$P(A \cup A^c) = P(A) + P(A^c)$$

From \star $P(A) + P(A^c) = 1$

$$\Rightarrow P(A^c) = 1 - P(A)$$

2 Quiz

Quiz

Prove the following:

- For any event A , $P(A^c) = 1 - P(A)$
- The probability of the empty set is zero, i.e., $P(\emptyset) = 0$

Take $A_1 = S$, $A_2 = \emptyset$ | 19
 Here $S \cap \emptyset = \emptyset \Rightarrow$
 S and \emptyset are mutually
 disjoint. Axiom-3
 $\Rightarrow P(S \cup \emptyset) = P(S) + P(\emptyset)$

Axiom-1: $P(S) = 1$

From above

$$\begin{aligned} 1 &= P(S) = P(S \cup \emptyset) = P(S) + P(\emptyset) \\ \Rightarrow P(\emptyset) &= 1 - P(S) \stackrel{\text{Axiom-1}}{=} 1 - 1 \\ &= 0 \end{aligned}$$

2 Quiz

Quiz

Prove the following:

- For any event A , $P(A^c) = 1 - P(A)$
- The probability of the empty set is zero, i.e., $P(\emptyset) = 0$
- For any event A , $P(A) \leq 1$

$$P(A^c) > 0, \text{ because } A^c \text{ is some event.}$$

From before,

$$\begin{aligned} P(A^c) &= 1 - P(A) > 0 \\ \Rightarrow P(A) &\leq 1 \end{aligned}$$

2 Quiz

Vanshi

$$A \cup X = B$$

$$X = B \setminus A$$

Here A & X are mutually disjoint. (Axiom-3)

Hence $P(A \cup X) = P(A) + P(X) = P(B)$

Here $P(A) > 0$, $P(B) > 0$, $P(X) > 0$
because A, B, X are events. (Axiom)

Since $P(X) > 0$

$$\Rightarrow \boxed{P(A) \leq P(B)}$$

Quiz

Prove the following:

- For any event A , $P(A^c) = 1 - P(A)$
- The probability of the empty set is zero, i.e., $P(\emptyset) = 0$
- For any event A , $P(A) \leq 1$ ←
- If $A \subseteq B$, then $P(A) \leq P(B)$

2 Quiz

① Are $A \setminus B$ & $A \cap B$ mutually exclusive? yes |¹⁹

Quiz

Prove the following:

- For any event A , $P(A^c) = 1 - P(A)$
- The probability of the empty set is zero, i.e., $P(\emptyset) = 0$
- For any event A , $P(A) \leq 1$
- If $A \subseteq B$, then $P(A) \leq P(B)$
- $P(A - B) = P(A) - P(A \cap B)$

Axiom - 3 Since $A \setminus B$ and $A \cap B$ are mutually exclusive

$$\Rightarrow P((A \setminus B) \cup (A \cap B)) = P(A \setminus B) + P(A \cap B) = P(A) - P(A \cap B)$$

$$\Rightarrow P(A \setminus B) = P(A) - P(A \cap B) \quad \text{=====}$$

2 Calculate Probabilities...

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Coin Toss example

A coin is considered fair if the likelihood of getting heads or tails is same. Calculate the probability of obtaining head, when the coin is tossed once.

$$A = \{H\} \leftarrow \text{favourable event.}$$

$$S = \{H, \bar{T}\} \leftarrow \text{Sample space}$$

$$P(A) = \frac{1}{2} = 50\%.$$

Coin Toss example

A coin is considered fair if the likelihood of getting heads or tails is same. Calculate the probability of obtaining head, when the coin is tossed once.

Answer

There are two favourable cases, the set of all possible cases is $\{H, T\}$. The number of head possible in one toss of the coin is 1. Hence, the probability of obtaining head is the fraction: $\frac{1}{2}$

2 Quiz

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Equally likely.

favourable event

Quiz

We roll a fair dice. What is the probability of the event $E = \{1, 6\}$?



$$P(E) = \frac{2}{6} = \frac{1}{3}$$

$$= 33.33\dots$$

$$\approx 33\%$$

$$E = \{1, 6\}$$

$$S = \{1, 2, 3, 4, 5, 6\}$$

Coin Toss example

A coin is considered **fair** if the likelihood of getting heads or tails is **same**.

2 Calculate Probabilities...

equally likely

H H
H T
T H
T T

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Coin Toss example

A coin is considered fair if the likelihood of getting heads or tails is same. Calculate the probability of obtaining at least one head, when two fair coins are tossed simultaneously.

Random Experiment.

Sample Space, $S = \{(H, H), (H, T), (T, H), (T, T)\}$

Favourable event : $A = \{(H, H), (H, T), (T, H)\}$

$$P(A) = \frac{3}{4} = 75\%$$

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A coin is considered **fair** if the likelihood of getting heads or tails is **same**. Calculate the probability of obtaining at least one head, when two fair coins are tossed simultaneously.

Answer

- Total number possibilities are
 $\{H, H\}, \{H, T\}, \{T, T\}, \{T, H\}$

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Answer

- Total number possibilities are
 $\{H, H\}, \{H, T\}, \{T, T\}, \{T, H\}$
- Favourable cases are $\{H, H\}, \{H, T\}, \{T, H\}$
- Hence the probability of atleast one head is = $\frac{3}{4}$

2 Calculate Probabilities...

$$P(A^c) = 1 - P(A)$$

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Coin Toss example

Ten ~~fair~~ coins are tossed simultaneously. What is the probability of getting atleast one head?

Define:

A = event of getting atleast one head.
 A^c = event of getting no head., i.e.;
 getting $\{(T, T, T, T, T, T, T, T, T, T)\}$

We know

$$P(A^c) = 1 - P(A)$$

$$\Rightarrow P(A) = 1 - P(A^c)$$

$$= 1 - \frac{1}{2^{10}}$$

Sample Space: S

$$\text{& } |S| = 2^{10}$$

$$P(A^c) = \frac{1}{2^{10}} = \frac{1}{1024}$$

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Ten fair coins are tossed simultaneously. What is the probability of getting atleast one head?

Answer

- Total number possibilities are 2^{10}

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Ten fair coins are tossed simultaneously. What is the probability of getting atleast one head?

Answer

- Total number possibilities are 2^{10}
- Only case with no head: $\{T, T, T, T, T, T, T, T, T, T\}$. Favourable cases are: $2^{10} - 1$

Coin Toss example

Ten fair coins are tossed simultaneously. What is the probability of getting atleast one head?

Answer

- Total number possibilities are 2^{10}
- Only case with no head: $\{T, T, T, T, T, T, T, T, T, T\}$. Favourable cases are: $2^{10} - 1$
- Hence the probability of atleast one head is $= \frac{2^{10} - 1}{2^{10}}$