

Certificate Program in Machine Learning & Artificial Intelligence: Batch-3

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Session Outline

- ▶ Probability
 - ▶ Why Probability?
 - ▶ What is Probability? - Definition and Meaning
 - ▶ Properties of Probability
 - ▶ Conditional Probability
 - ▶ Bayes' Theorem

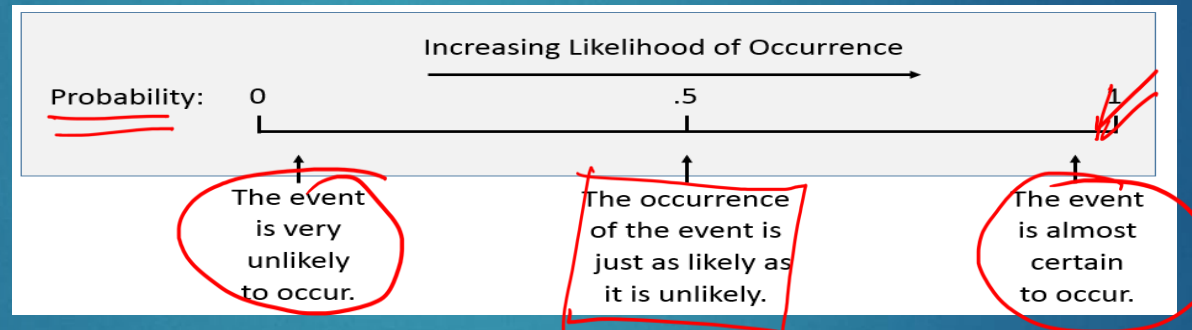
Business Use Cases of Probability

- ▶ Managers often base their decisions on an analysis of uncertainties such as the following
- ▶ What are the *chances* that the sales will decrease if we increase prices?
- ▶ What is the *likelihood* a new assembly method will increase productivity?
- ▶ What are the *odds* that a new investment will be profitable?

Probability & Random Experiments

► What is probability?

- Numerical measure of the likelihood that an event will occur.



Probability: A Few Terms

$\checkmark \{H, T\}$ $\checkmark \{1, 2, 3, 4, 5, 6\}$

- ▶ Experiment, Sample space, Single-step and multi-step experiment

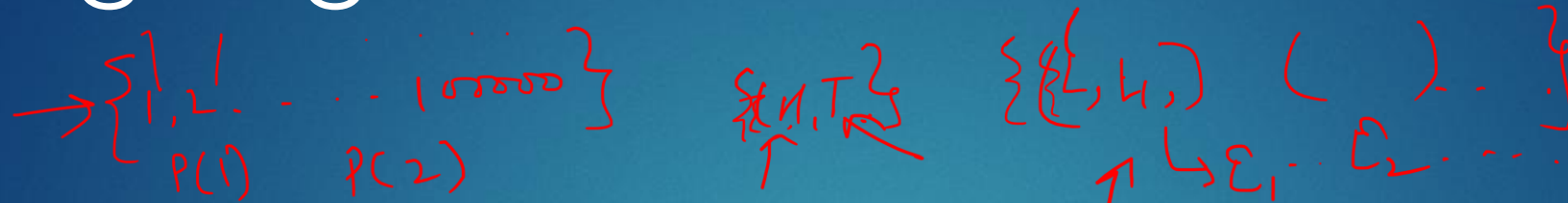
▶ Combinations and Permutations

- ▶ Number of experimental outcomes when the experiment involves selecting n objects from a set of N objects – Combinations

- ▶ Number of experimental outcomes when the experiment involves selecting n objects from a set of N objects and the order of selection is important – Permutations

$\{H, 1\}$ $\{H, 2\}$... $\{H, 6\}$
 $\{T, 1\}$ $\{T, 2\}$... $\{T, 6\}$

Assigning Probabilities



- ▶ The probability assigned to each experimental outcome must be between 0 and 1, inclusively.

$$\underline{0 \leq P(E_i) \leq 1 \text{ for all } i}$$

where E_i is the i^{th} experimental outcome and $\underline{P(E_i)}$ is its probability

- ▶ The sum of the probabilities for all experimental outcomes must equal 1.

$$\underline{P(E_1) + P(E_2) + \dots + P(E_n) = 1}$$

where n is the number of experimental outcomes

Assigning Probabilities

- ▶ Classical Method

- ▶ Assigning probabilities based on the assumption of equally likely outcomes; E.g., Rolling a die

- ▶ Relative Frequency Method

- ▶ Assigning probabilities based on experimental or historical data; E.g., No. of deliveries on a particular day in a week

- ▶ Subjective Method

- ▶ Assigning probability based on judgment.

The best probability estimates often are obtained by combining the estimates from the classical or relative frequency approach with the subjective estimate

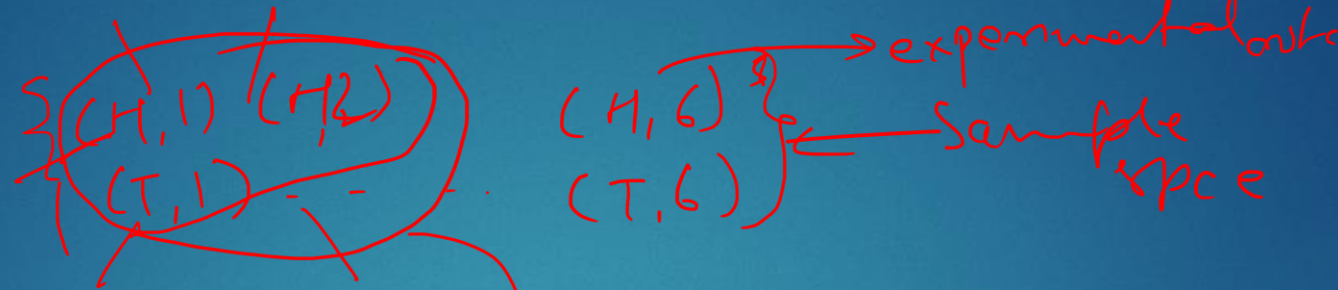
Events & Their Probabilities

▶ Event

- ▶ A collection of sample points
- ▶ Two upstream suppliers' case – together being able to deliver in less than 10 days

▶ Probability of an event

- ▶ Sum of the probabilities of the sample points in the event



Some Probability Relationships

- ▶ Complement of an Event, A^c

- ▶ Event consisting of all sample points that are not in A
- ▶ Computing Probability using the complement



- ▶ Addition Law – Union of Two Events

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

- ▶ Mutually Exclusive events

- ▶ If the events have no sample points in common
- ▶ $P(A \text{ intersection } B) = 0$

Conditional Probability

$\{(H,1) \dots (H,6)$
 $\{ (T,1) \dots (T,6) \}$

A - custom
B → demo

- ABP
- ▶ The probability of an event given that another event has occurred is called a conditional probability.

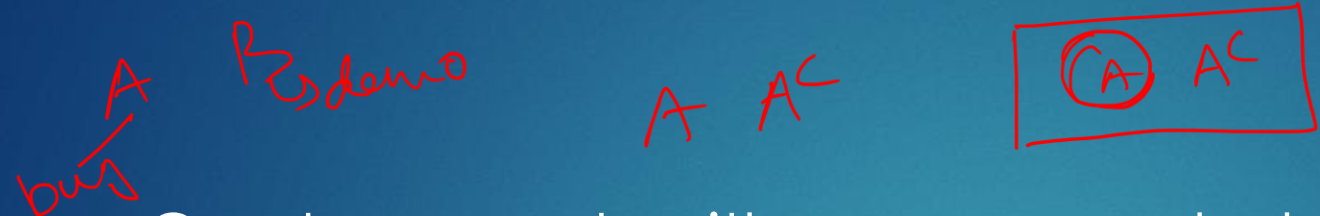
- ▶ The conditional probability of A given B has already occurred denoted by $P(A|B)$.

- ▶ A conditional probability is computed as follows:

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

- ▶ Examples? Business Use Cases?

Independent events



- ▶ Can two events with non-zero probability be both mutually exclusive and independent? →
- ▶ Mutually exclusive events are always dependent events

A → SA ✓
B → SB ✓
▶ Non-mutually exclusive events may or may not be independent



▶ Independent events:

$$P(A|B) = P(A) \text{ or } P(B|A) = P(B)$$

dependent

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

buy → email

Questions

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

▶ Question 1

▶ $P(A) = 0.30$; $p(B) = 0.40$; A and B are mutually independent;

▶ $p(A \text{ Intersection } B) = ?$

▶ $P(A|B) = ?$

not

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

▶ Question 2

▶ $P(A) = 0.30$; $p(B) = 0.40$; A and B are mutually exclusive;

▶ $p(A \text{ Intersection } B) = ?$

▶ $P(A|B) = ?$

$$P(A) = 0.30$$

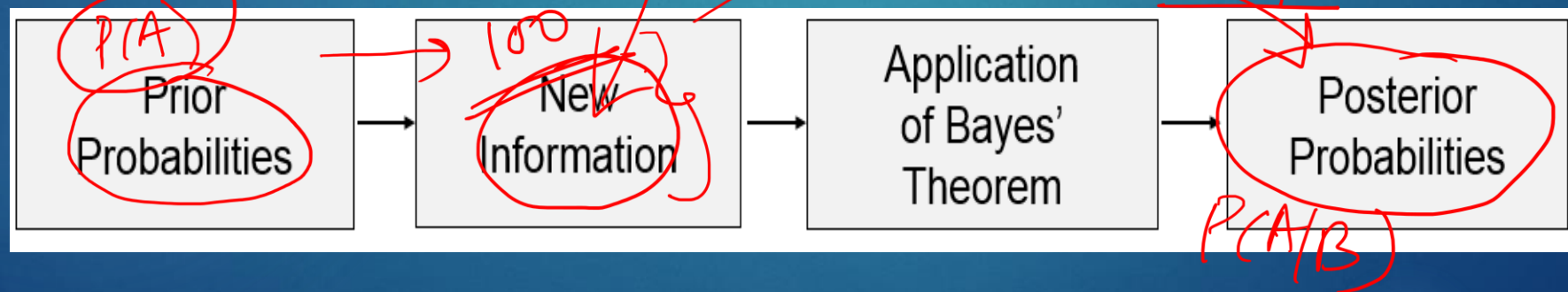
demanded → *bought*
A & B are dependent

$$P(A) = 0.30$$
$$P(B) = 0.40$$

$$\frac{P(A \cap B)}{P(B)} = \frac{0.3}{0.4} = 0.75$$

Bayes' Theorem

- ▶ Probability Analysis begins with 'prior probabilities'
- ▶ Then, additional information is obtained – from a sample, product test, report
- ▶ Using this information, 'posterior probabilities' are calculated – using Bayes' Theorem – A means to revise prior probabilities



Bayes' Theorem: Applications

- ▶ Use case
 - ▶ Impact of marketing approaches



$P(A_i|B)$
 $P(\text{---})$

posterior

$$P(A_i|B) = \frac{P(A_i)P(B|A_i)}{P(A_1)P(B|A_1) + P(A_2)P(B|A_2) + \dots + P(A_n)P(B|A_n)}$$

$A_1, A_2, A_3, \dots, A_n$

- ▶ Bayes' theorem is applicable when the events for which we want to compute posterior probabilities are mutually exclusive and collectively exhaustive

Bayes' Theorem: Applications

$$P(A_1|B) = ?$$

$$P(A_1|G) = \frac{P(G|A_1) \cdot P(A_1)}{P(G|A_1) \cdot P(A_1) + P(G|A_2) \cdot P(A_2)}$$

▶ Question 1

▶ Two suppliers (A1 & A2)

▶ 65% supplies from A1 & 35% from A2

▶ $P(G|A_1) = 0.98$ & $p(G|A_2) = 0.95$

$$\rightarrow P(B|A_2) = 0.05$$

$$P(A_1) = 0.65 ; P(A_2) = 0.35$$

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▶ Question is: A bad part is detected, what is the probability that it came from A1 and what is the probability that it came from A2?

▶ Solution

$$P(B|A_1) = 0.02$$

▶ If a part is chosen at random, $p(A_1) = 0.65$; $p(A_2) = 0.35$

▶ $P(A_1|B) = p(A_1) \cdot p(B|A_1) / (p(A_1) \cdot p(B|A_1) + p(A_2) \cdot p(B|A_2))$

$$P(A_1|B) = \frac{P(B|A_1) \cdot P(A_1)}{P(B|A_1) \cdot P(A_1) + P(B|A_2) \cdot P(A_2)}$$

$$\hookrightarrow \frac{P(B|A_2) \cdot P(A_2)}{P(B|A_1) \cdot P(A_1) + P(B|A_2) \cdot P(A_2)}$$

A Supply Chain Problem

- ▶ **Scenario: Identifying Defective Products from Multiple Manufacturing Plants**
- ▶ A company, *TechWear*, produces wearable fitness trackers at three different manufacturing plants: **Plant A**, **Plant B**, and **Plant C**. Recently, there has been an uptick in defective products, and *TechWear* wants to identify the most likely source of these defects.
- ▶ The company has the following information:
 - 50% of the fitness trackers are produced at **Plant A**. ✓
 - 30% of the fitness trackers are produced at **Plant B**. ✓
 - 20% of the fitness trackers are produced at **Plant C**. ✓
 - The probability of a defective fitness tracker from **Plant A** is 2%.
 - The probability of a defective fitness tracker from **Plant B** is 4%.
 - The probability of a defective fitness tracker from **Plant C** is 6%.
- ▶ A fitness tracker has been identified as defective, and the company wants to determine the probability that it was produced at **Plant C**.

A Customer Service Problem

► Scenario: Customer Service Issue in a Telecom Company

► A telecom company, *ConnectCo*, provides customer support through three different channels: **Phone Support**, **Live Chat**, and **Email Support**. Recently, the company has been receiving customer complaints about poor service experiences, and it wants to identify which channel is most likely responsible for the complaints.

► The company has the following data:

- 50% of customer interactions happen via **Phone Support**.
- 30% of customer interactions happen via **Live Chat**.
- 20% of customer interactions happen via **Email Support**.
- The probability of a complaint arising from **Phone Support** is 4%.
- The probability of a complaint arising from **Live Chat** is 6%.
- The probability of a complaint arising from **Email Support** is 10%.

► A customer complaint has been received, and the company wants to determine the probability that this complaint came from **Email Support**.

A_1 $P(A) = 0.5$
 A_2 0.3
 A_3 0.2

$P(A_3|C) = \frac{0.2 \times 0.1}{0.2 \times 0.1 + 0.3 \times 0.06 + 0.5 \times 0.04}$
 $P(A_1|C) =$
 $P(A_2|C) =$