



Probability and Stochastic Processes

Lecture 02: Sample Space, Algebra, σ -Algebra, Measure

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Probability Theory – Humble Beginnings

- Bernoulli (1713) and de Moivre (1718) gave the first definition of probability:

$$\text{probability of an event} = \frac{\text{\# favourable outcomes}}{\text{total number of outcomes}}.$$

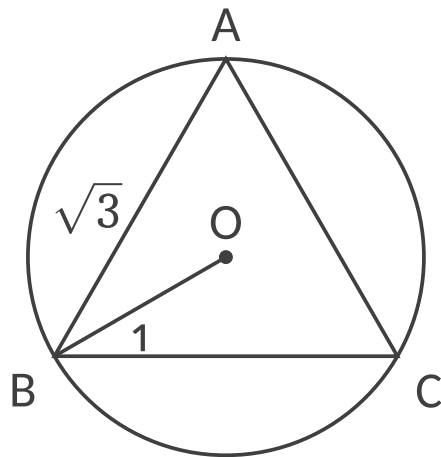
- Cournot (1843):
“An event with very small probability is morally impossible; an event with very high probability is morally certain.”
- French mathematicians of the day were satisfied with the “frequentist” approach to probability, but not the German and English mathematicians of the day
- Frequentist approach could not satisfactorily explain certain paradoxes

Why Axiomatic Theory? Bertrand's Paradox

Take a circle with unit radius and inscribe an equilateral triangle in it. Draw a random chord. What is the probability that the length of the “random chord” is greater than $\sqrt{3}$?

Bertrand's perfectly valid arguments:

- Mid-point of chord should lie inside incircle of radius $1/2$



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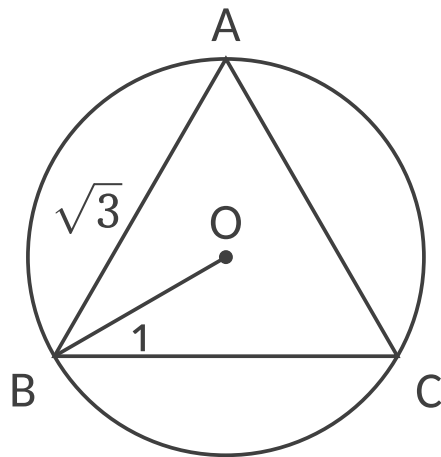
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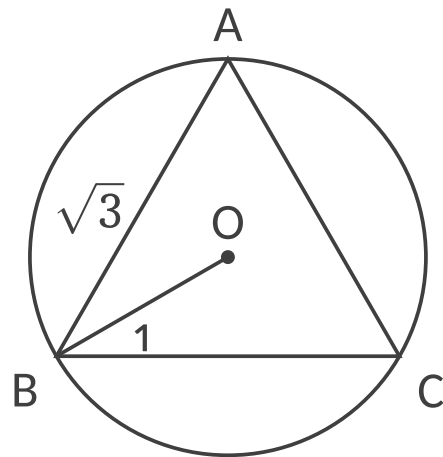
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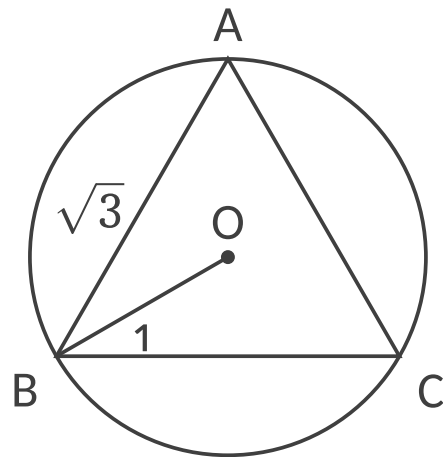


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- Mid-point of chord should be between O and projection of O onto side BC

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Borel to the Rescue

- Contributions to Measure Theory by Borel (1894) provided a shift in perspective
- Countable unions played a key role in Borel's theory
- Kolmogorov's genius was in applying Borel's theory to formalise the axioms of probability, laying the foundation stone for modern probability theory
- For more details on the history of probability, see [[Shafer and Vovk, 2018](#)] and [[Kolmogorov, 2004](#)]

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We begin with two universally accepted entities:

- Random experiment
- Outcome (denoted by ω) – **source of randomness**

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- If our interest is in the velocity with which the coin lands on ground, then $\Omega = [0, \infty) = \mathbb{R}_+$
- If our interest is in the number of times coin flips in air, then $\Omega = \mathbb{N}$



Example: Toss a coin n times, for some $n < \infty$.
Interest: faces that show up

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Event

Informal Definition (Event)

Informally,^a an event is a subset of outcomes “of interest” to us.

^aWe shall give a more formal definition of an event later.

Example: Toss a coin 3 times; interest is in the faces that show up

Sample space $\Omega =$

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Event A of interest: at least 2 heads show up

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$$E = \{HHH, THH, HTH, HHT\} \subset \Omega$$

$$\{H, T\}^N \quad \{H, T\}^\infty$$

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$$\Omega = \{H, T\}^3 = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\}$$

Event A of interest: at least 2 heads show up

$$A = \{HHH, THH, HTH, HHT\}$$

$$\{HHH\}$$

Note

If an outcome $\omega \in A$ occurs, we say that the event A occurs.

Definition (Algebra)

Let Ω be a sample space.

A collection \mathcal{A} of subsets of Ω is called an **algebra** if it satisfies the following properties:

1. $\Omega \in \mathcal{A}$. $\nearrow \Omega \setminus A$
2. $A \in \mathcal{A} \implies A^c \in \mathcal{A}$ (closure under **complements**).
3. $A, B \in \mathcal{A} \implies A \cup B \in \mathcal{A}$.

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Property 3 above implies, by mathematical induction, that

$$A_1, A_2, \dots, A_n \in \mathcal{A} \implies \bigcup_{i=1}^n A_i \in \mathcal{A} \quad \text{for all } n \in \mathbb{N} \quad (\text{closure under **finite unions**}).$$

Exercise

Show that an algebra is closed under finite intersections.

Closure under finite intersections:

To show: Given $n \in \mathbb{N}$ and $A_1, \dots, A_n \in \mathcal{A}$,

$$\bigcap_{i=1}^n A_i \in \mathcal{A}$$

$$A_1 \cap A_2 \cap \dots \cap A_n \quad \leftarrow$$

$$A_i \in \mathcal{A} \Rightarrow A_i^c \in \mathcal{A} \quad \forall i \in \{1, \dots, n\} \quad (\text{prop. 2})$$

$$\Rightarrow \left(\bigcup_{i=1}^n A_i^c \right) \in \mathcal{A} \quad (\text{finite union closure})$$
$$= B$$

$$\Rightarrow \left(\bigcup_{i=1}^n A_i^c \right)^c \in \mathcal{A}$$

$$\text{But note that } \left(\bigcup_{i=1}^n A_i^c \right)^c = \bigcap_{i=1}^n A_i \quad (\text{De Morgan's law})$$

σ -Algebra – Motivation

Motivating example: toss a coin until first head shows up

$$\Omega = \{H, TH, TTH, TTTH, \dots\}$$

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Event of interest (A) = # of tosses is even

$$A = \{TH, TTH, \dots\}$$

$$\{H, TH\} = \{H\} \cup \{TH\}$$

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Remark

Consider an algebra \mathcal{A} of subsets of Ω . Then, $A \notin \mathcal{A}$.

Definition (σ -Algebra)

Let Ω be a sample space.

A collection \mathcal{F} of subsets of Ω is called a σ -algebra if it satisfies the following properties:

- $\Omega \in \mathcal{F}$.
- $A \in \mathcal{F} \implies A^c \in \mathcal{F}$ (closed under complements).
- $A_1, A_2, \dots \in \mathcal{F} \implies \bigcup_{i \in \mathbb{N}} A_i \in \mathcal{F}$ (closure under countably infinite unions).

$$A_1 \cup A_2 \cup A_3 \cup A_4 \cup \dots$$

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- Every σ -algebra is also an algebra, but the converse is not true (see homework)

$$\bigcup_{i=1}^{\infty} A_i = \bigcup_{i \in \mathbb{N}} A_i, \text{ with } A_i = \emptyset \quad \forall i \geq n+1$$

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- An event $A \in \mathcal{F}$ is also referred to as an **\mathcal{F} -measurable set**
- ✓ Every σ -algebra is also an algebra, but the converse is not true (see homework)
- The pair (Ω, \mathcal{F}) is called a **measurable space**

2

One can readily verify the following properties of σ -algebras:

- Every σ -algebra is closed under countable intersections
- Let \mathcal{I} be an **arbitrary** index set, and let $\{\mathcal{F}_i : i \in \mathcal{I}\}$ be a collection of σ -algebras of subsets of a given sample space Ω . Then, $\bigcap_{i \in \mathcal{I}} \mathcal{F}_i$ is a σ -algebra of subsets of Ω .

σ -Algebra – Examples

Fix a sample space Ω .

- The most trivial σ -algebras $\{\emptyset, \Omega\}$ and 2^Ω
- For any $A \subset \Omega$, $\{\emptyset, \Omega, A, A^c\}$ is a σ -algebra
- Given a collection \mathcal{C} of subsets of Ω , the smallest σ -algebra containing \mathcal{C} is denoted by $\sigma(\mathcal{C})$.



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σ -Algebra – Examples

$$\{\emptyset\} \subseteq \mathcal{A}$$

$$\{\emptyset, \Omega, \{1\}, \{2, 3, 4, 5, 6\}\}$$

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Example: $\Omega = \{1, 2, \dots, 6\}$, $\mathcal{C} = \{\{1, 2\}, \{1, 3\}\}$

$$\{1\} \in \mathcal{A}$$

$$\{\{1\}\} \subseteq \mathcal{A}$$

$$\mathcal{A} = \{\Omega, \emptyset, \{1, 2\}, \{1, 3\}, \{3, 4, 5, 6\}, \{2, 4, 5, 6\}, \{1, 2, 3\}, \{4, 5, 6\}, \\ \{1, 3, 4, 5, 6\}, \{2\}, \{1, 2, 4, 5, 6\}, \{3\}, \{2, 3\}, \{1, 4, 5, 6\}, \\ \{2, 3, 4, 5, 6\}, \{1\}\}$$

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$$\sigma(\mathcal{C}) = \left\{ \emptyset, \Omega, \{1, 2\}, \{1, 3\}, \{1\}, \{2, 3, 4, 5, 6\}, \{1, 2, 3\}, \{4, 5, 6\}, \right. \\ \left. \{3, 4, 5, 6\}, \{2, 4, 5, 6\}, \{2\}, \{3\}, \{1, 3, 4, 5, 6\}, \{1, 2, 4, 5, 6\}, \{2, 3\}, \{1, 4, 5, 6\} \right\}$$

Measure

Fix a measurable space (Ω, \mathcal{F}) .

Definition (Measure)

A function $\mu : \mathcal{F} \rightarrow [0, +\infty]$ is called a **measure** on (Ω, \mathcal{F}) if it satisfies the following properties:

1. $\mu(\emptyset) = 0$.
2. If A_1, A_2, \dots is a **countable** collection of **disjoint** sets, with $A_i \in \mathcal{F}$ for each $i \in \mathbb{N}$, then

$$\mu \left(\bigcup_{i \in \mathbb{N}} A_i \right) = \sum_{i \in \mathbb{N}} \mu(A_i).$$

Property 2 above is called the property of **countable additivity**.

The triplet $(\Omega, \mathcal{F}, \mu)$ is called a **measure space**.

Measure

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

$$\mathcal{F} = \{ \emptyset, \Omega, \{1\}, \{2, 3\} \}$$

\downarrow
0

\downarrow
1

\downarrow
0

\downarrow
1

$$\mathbb{P} : \mathcal{F} \rightarrow [0, 1]$$

- When $\mu(\Omega) < +\infty$, the measure μ is called a **finite measure**
- When $\mu(\Omega) = +\infty$, the measure μ is called an **infinite measure**
- When $\mu(\Omega) = 1$, the measure μ is called a **probability measure**, and denoted by \mathbb{P} .

$$[0, 1] \rightarrow \{0, 1\}^{\mathbb{N}}$$

$$f : [0, 1] \rightarrow \mathbb{R}$$

Probability Measure

$$\Omega = \{0,1\}^{\mathbb{N}} \rightarrow \text{uncountable}$$
$$\mathcal{F}$$

Fix a measurable space (Ω, \mathcal{F}) .

Definition (Probability Measure)

A function $\mathbb{P} : \mathcal{F} \rightarrow [0, 1]$ is called a **probability measure** if the following properties are satisfied:

1. $\mathbb{P}(\emptyset) = 0$.
2. $\mathbb{P}(\Omega) = 1$.
3. If A_1, A_2, \dots is a **countable** collection of **disjoint** sets, with $A_i \in \mathcal{F}$ for each $i \in \mathbb{N}$, then

$$\mathbb{P} \left(\bigcup_{i \in \mathbb{N}} A_i \right) = \sum_{i \in \mathbb{N}} \mathbb{P}(A_i).$$

References



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