

# Probability: Axioms, Properties, Interpretations

Engineering Statistics  
Section 2.2

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# Chains of Unions, Chains of Intersections

Just as a **chain of sums** can be written compactly...

$$\sum_{k=1}^n a_k = a_1 + a_2 + \cdots + a_{n-1} + a_n$$

$$\sum_{k=1}^{\infty} a_k = a_1 + a_2 + a_3 + \cdots$$

...a **chain of unions** can similarly be compactly written...

$$\bigcup_{k=1}^n A_k = A_1 \cup A_2 \cup \cdots \cup A_{n-1} \cup A_n$$

$$\bigcup_{k=1}^{\infty} A_k = A_1 \cup A_2 \cup A_3 \cup \cdots$$

...as can a **chain of intersections**...

$$\bigcap_{k=1}^n A_k = A_1 \cap A_2 \cap \cdots \cap A_{n-1} \cap A_n$$

$$\bigcap_{k=1}^{\infty} A_k = A_1 \cap A_2 \cap A_3 \cap \cdots$$

...as can a **chain of products**.

$$\prod_{k=1}^n a_k = a_1 a_2 \cdots a_{n-1} a_n$$

$$\prod_{k=1}^{\infty} a_k = a_1 a_2 a_3 \cdots$$

# Probability Axioms

The **probability** of an event  $E$ , denoted  $\mathbb{P}(E)$ , is a real number. It turns out only 3 **axioms** are needed to ensure probability behaves properly:

## Proposition

*(Probability Axioms)*

Let  $E \subseteq \Omega$  be an event from sample space  $\Omega$  of an experiment.

Let  $E_1, E_2, E_3, \dots$  be an infinite collection of pairwise disjoint events. Then:

- (A1)  $\mathbb{P}(E) \geq 0$  *Probability of an Event is Non-Negative*
- (A2)  $\mathbb{P}(\Omega) = 1$  *Probability of Sample Space is always One*
- (A3)  $\mathbb{P}\left(\bigcup_{k=1}^{\infty} E_k\right) = \sum_{k=1}^{\infty} \mathbb{P}(E_k)$  *Probability of Infinite Disjoint Union is a Sum*

# Basic Properties of Probability

The 3 probability axioms can be used to build more useful properties:

## Proposition

*(Basic Probability Properties)*

*Let  $E \subseteq \Omega$  be an event from sample space  $\Omega$  of an experiment.*

*Let  $E_1, E_2, E_3, \dots$  be an infinite collection of pairwise disjoint events. Then:*

(P1)  $\mathbb{P}(\emptyset) = 0$  *Probability of Empty Set is always Zero*

(P2)  $\mathbb{P}\left(\bigcup_{k=1}^n E_k\right) = \sum_{k=1}^n \mathbb{P}(E_k)$  *Probability of Finite Disjoint Union is a Sum*

# Basic Properties of Probability

The 3 probability axioms can be used to build more useful properties:

## Proposition

*(Basic Probability Properties)*

Let  $E \subseteq \Omega$  be an event from sample space  $\Omega$  of an experiment.

Let  $E_1, E_2, \dots, E_n$  be a finite collection of pairwise disjoint events. Then:

(P1)  $\mathbb{P}(\emptyset) = 0$  Probability of Empty Set is always Zero

(P2)  $\mathbb{P}\left(\bigcup_{k=1}^n E_k\right) = \sum_{k=1}^n \mathbb{P}(E_k)$  Probability of Finite Disjoint Union is a Sum

PROOF: (P1) Let  $E_1 = E_2 = E_3 = \dots = \emptyset$ . Then  $\bigcup_{k=1}^{\infty} E_k = \emptyset$

Moreover, the events are pairwise disjoint since  $E_i \cap E_j = \emptyset \cap \emptyset \stackrel{SS}{=} \emptyset$  for  $i \neq j$

$\therefore \mathbb{P}\left(\bigcup_{k=1}^{\infty} E_k\right) = \mathbb{P}(\emptyset) \xrightarrow{A3} \sum_{k=1}^{\infty} \mathbb{P}(E_k) = \mathbb{P}(\emptyset) \implies \sum_{k=1}^{\infty} \mathbb{P}(\emptyset) = \mathbb{P}(\emptyset)$

$\implies \mathbb{P}(\emptyset) = 0$  (as it's the only way for LHS & RHS to equal each other)  $\square$

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PROOF: (P2) Let  $E_{n+1} = E_{n+2} = E_{n+3} = \dots = \emptyset$

Then  $\bigcup_{k=1}^{\infty} E_k = \left(\bigcup_{k=1}^n E_k\right) \cup \left(\bigcup_{k=n+1}^{\infty} \emptyset\right) \stackrel{\text{S8}}{=} \left(\bigcup_{k=1}^n E_k\right) \cup \emptyset \stackrel{\text{S8}}{=} \bigcup_{k=1}^n E_k$

Finally, apply (A3) and then (P1)  $\square$

## Proposition

*(More Probability Properties)*

*Let  $E \subseteq \Omega$  be an event from sample space  $\Omega$  of an experiment. Then:*

*(P3)  $\mathbb{P}(E^c) = 1 - \mathbb{P}(E)$     Probability of Complement*

*(P4)  $\mathbb{P}(E) \leq 1$     Probability is never greater than One*

## Proposition

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(P4)  $\mathbb{P}(E) \leq 1$  *Probability is never greater than One*

PROOF:

(P3) Note that  $E \cup E^c = \Omega$  and  $E \cap E^c = \emptyset \implies E, E^c$  are disjoint events.

Then  $\mathbb{P}(E \cup E^c) = \mathbb{P}(\Omega) \xrightarrow{P2} \mathbb{P}(E) + \mathbb{P}(E^c) = \mathbb{P}(\Omega) \xrightarrow{A2} \mathbb{P}(E) + \mathbb{P}(E^c) = 1$

$\therefore \mathbb{P}(E^c) = 1 - \mathbb{P}(E)$   $\square$



## Proposition

*(More Probability Properties)*

Let  $E \subseteq \Omega$  be an event from sample space  $\Omega$  of an experiment. Then:

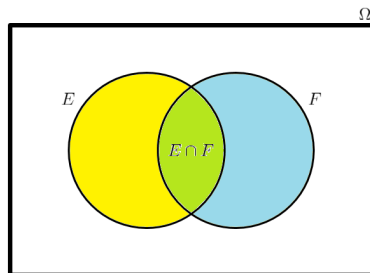
$$(P3) \quad \mathbb{P}(E^c) = 1 - \mathbb{P}(E) \quad \text{Probability of Complement}$$

$$(P4) \quad \mathbb{P}(E) \leq 1 \quad \text{Probability is never greater than One}$$

PROOF:

$$(P4) \quad \text{By (P3), } \mathbb{P}(E) + \mathbb{P}(E^c) = 1 \stackrel{A1}{\implies} \mathbb{P}(E) \leq 1 \quad \square$$

# Probability of a Union of Two Events



Not Mutually Exclusive

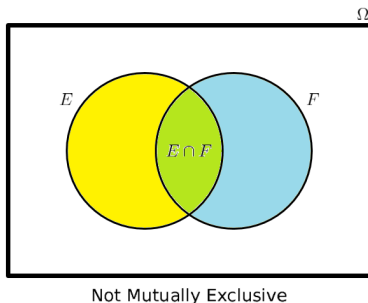
## Proposition

*(Probability of a Union of Two Events)*

Let  $E, F \subseteq \Omega$  be two events from sample space  $\Omega$  of an experiment. Then:

$$(P5) \quad \mathbb{P}(E \cup F) = \mathbb{P}(E) + \mathbb{P}(F) - \mathbb{P}(E \cap F) \quad \text{Probability of Union}$$

# Probability of a Union of Two Events



## Proposition

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Let  $E, F \subseteq \Omega$  be two events from sample space  $\Omega$  of an experiment. Then:

$$(P5) \quad \mathbb{P}(E \cup F) = \mathbb{P}(E) + \mathbb{P}(F) - \mathbb{P}(E \cap F) \quad \text{Probability of Union}$$

PROOF: The Venn Diagram indicates that  $\mathbb{P}(E) + \mathbb{P}(F)$  counts  $\mathbb{P}(E \cap F)$  twice!  
See the textbook for lower-level proof.  $\square$

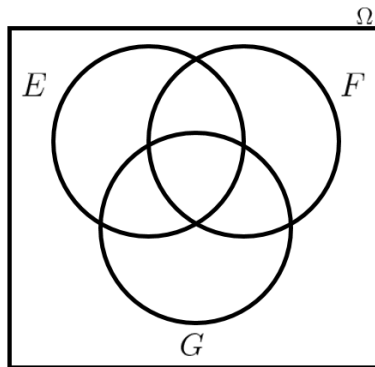
# Principle of Inclusion-Exclusion

## Proposition

*(Principle of Inclusion-Exclusion)*

*Let  $E, F, G \subseteq \Omega$  be three events from sample space  $\Omega$  of an experiment.  
Then:*

$$\mathbb{P}(E \cup F \cup G) = \mathbb{P}(E) + \mathbb{P}(F) + \mathbb{P}(G) - \mathbb{P}(E \cap F) - \mathbb{P}(E \cap G) - \mathbb{P}(F \cap G) + \mathbb{P}(E \cap F \cap G)$$



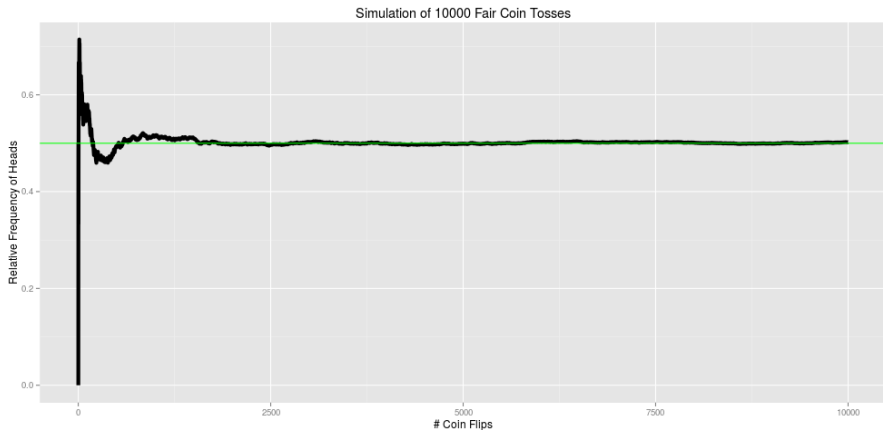
# Deep Interpretation of Probability

The axioms & properties do not give a complete interpretation of probability!!  
The most intuitive interpretation is to treat probability as a relative frequency:



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The most intuitive interpretation is to treat probability as a relative frequency:



# Interpretation of Probability

## Definition

(Interpretation of Probability)

The **probability of an outcome** in a sample space is the likelihood of the outcome, which is a number between 0 and 1 inclusive.

The **probability of an event**  $E$ , denoted  $\mathbb{P}(E)$ , is the sum of the probabilities of the outcomes that comprise  $E$ .

Interpretation of Probability:

	$\mathbb{P}(E) = 0$	$\implies$	Event $E$ is impossible
0	$< \mathbb{P}(E) < 0.50$	$\implies$	Event $E$ is not likely to occur
	$\mathbb{P}(E) = 0.50$	$\implies$	Event $E$ has 50-50 chance of occurring
0.50	$< \mathbb{P}(E) < 1$	$\implies$	Event $E$ is likely to occur
	$\mathbb{P}(E) = 1$	$\implies$	Event $E$ is certain to occur

## Definition

(Interpretation of Probability)

The **probability of an outcome** in a sample space is the likelihood of the outcome, which is a number between 0 and 1 inclusive.

The **probability of an event**  $E$ , denoted  $\mathbb{P}(E)$ , is the sum of the probabilities of the outcomes that comprise  $E$ .

Examples of Probability:

- "There's a 30% chance of snow tomorrow."  $[\mathbb{P}(\text{Snow tomorrow}) = 0.30]$
- "25% of adults get seven hours of sleep."  $[\mathbb{P}(7 \text{ hrs of sleep}) = 0.25]$
- "All dogs play fetch."  $[\mathbb{P}(\text{Playing fetch}) = 1]$
- "There's a 1 in 1000 chance of winning."  $[\mathbb{P}(\text{Winning}) = \frac{1}{1000}]$
- "None of my cats catch mice."  $[\mathbb{P}(\text{Catch mice}) = 0]$



# Measure of a Set (Definition)

## Definition

(Measure of a Set)

The **measure** of a **countable set** is  $|E| := (\# \text{ of elements in } E)$

The **measure** of a **1D set** is  $|E| := (\text{Length of curve } E)$

The **measure** of a **2D set** is  $|E| := (\text{Area of region } E)$

The **measure** of a **3D set** is  $|E| := (\text{Volume of solid } E)$

The **measure** of the **empty set** is defined to be zero:  $|\emptyset| := 0$

# Measure of a Countable Set (Example)

## Definition

(Measure of a Set)

The **measure** of a **countable set** is  $|E| := (\# \text{ of elements in } E)$

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The **measure** of the **empty set** is defined to be zero:  $|\emptyset| := 0$

Example: Let  $S = \{\text{Heads, Tails}\}$ .

Then,  $|S| = (\# \text{ of elements of } S) = \boxed{2}$

# Measure of a 1D Set (Example)

## Definition

(Measure of a Set)

The **measure** of a **countable set** is  $|E| := (\# \text{ of elements in } E)$

The **measure** of a **1D set** is  $|E| := (\text{Length of curve } E)$

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The **measure** of a **3D set** is  $|E| := (\text{Volume of solid } E)$

The **measure** of the **empty set** is defined to be zero:  $|\emptyset| := 0$

Example: Let  $\ell$  be a line segment with length 13.

Then,  $|\ell| = (\text{Length of } \ell) = \boxed{13}$

# Measure of a 2D Set (Example)

## Definition

(Measure of a Set)

The **measure** of a **countable set** is  $|E| := (\# \text{ of elements in } E)$

The **measure** of a **1D set** is  $|E| := (\text{Length of curve } E)$

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The **measure** of a **3D set** is  $|E| := (\text{Volume of solid } E)$

The **measure** of the **empty set** is defined to be zero:  $|\emptyset| := 0$

Example: Let  $R$  be a rectangle with length 2 and width 3.

Then,  $|R| = (\text{Area of } R) = (\text{Length}) \times (\text{Width}) = 2 \times 3 = \boxed{6}$

# Measure of a 3D Set (Example)

## Definition

(Measure of a Set)

The **measure** of a **countable set** is  $|E| := (\# \text{ of elements in } E)$

The **measure** of a **1D set** is  $|E| := (\text{Length of curve } E)$

The **measure** of a **2D set** is  $|E| := (\text{Area of region } E)$

The **measure** of a **3D set** is  $|E| := (\text{Volume of solid } E)$

The **measure** of the **empty set** is defined to be zero:  $|\emptyset| := 0$

Example: Let  $C$  be a cube of length 4.

Then,  $|C| = (\text{Volume of } C) = (\text{Length})^3 = 4^3 = \boxed{64}$

# BE CAREFUL WITH THE VERTICAL BARS!!!

The vertical bars  $| \cdot |$  are heavily overloaded in mathematics:

<b>MATHEMATICAL OBJECT</b>	<b>TYPICAL EXPRESSION</b>	<b>MEANING OF VERTICAL BARS</b>
Scalar	$ a  =  -3 $	Absolute Value of $a$
Vector	$\ \mathbf{v}\  = \left\  \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\ $	Norm of $\mathbf{v}$
Matrix	$ A  = \begin{vmatrix} 1 & 2 \\ 3 & 4 \end{vmatrix}$	Determinant of $A$
Set	$ E  =  \{1, 2, 3, 4\} $	Measure of $E$

For this course, vectors and matrices will not be considered, so just be mindful of the difference between the absolute value of a scalar & measure of a set.

# Probability: Equally Likely Outcomes

Very often, all the outcomes of an experiment are **equally likely to occur**:

## Definition

(Probability of an Event)

Let  $\Omega$  be the sample space of an experiment with **equally likely outcomes**. Let  $E$  be an event of the experiment.

Then the **probability** of event  $E$  occurring is defined as:  $\mathbb{P}(E) = \frac{|E|}{|\Omega|}$

i.e. The probability is the proportion of outcomes that comprise the event.

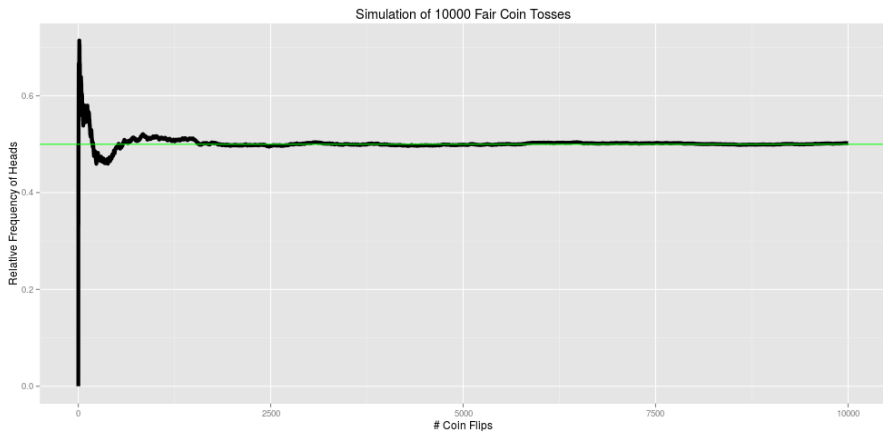
## REMARKS:

Often it's impractical to list every outcome of a sample space  $\Omega$  or event  $E$ . When computing probability, only the **measures** of  $E$  &  $\Omega$  are needed.

Fair coins & fair dice always result in equally likely outcomes.

# Probability: Equally Likely Outcomes (Example)

Consider the experiment of flipping a coin once with (H)eads and (T)ails. Let event  $E \equiv$  "Heads". Then, sample space  $\Omega = \{H, T\}$ .



$$\text{Then: } \mathbb{P}(\text{Heads}) \equiv \mathbb{P}(E) = \frac{|E|}{|\Omega|} = \frac{|\{H\}|}{|\{H, T\}|} = \frac{1}{2} = 0.5$$



## Textbook Logistics for Section 2.2

- Difference(s) in Terminology:

<b>TEXTBOOK TERMINOLOGY</b>	<b>SLIDES/OUTLINE TERMINOLOGY</b>
Null Event $\emptyset$	Empty Set $\emptyset$
Number of Outcomes in $E$	Measure of $E$

- Difference(s) in Notation:

<b>CONCEPT</b>	<b>TEXTBOOK NOTATION</b>	<b>SLIDES/OUTLINE NOTATION</b>
Sample Space	$\mathcal{S}$	$\Omega$
Complement of Event	$A'$	$A^c$
Probability of Event	$P(A)$	$\mathbb{P}(A)$
Measure of Event	$N(A)$	$ A $

Fin.