

# A Brief Introduction to Probability

ECE/CS 541

# Foundations

- Notion of an “outcome of an experiment”
  - Experiment characterized by something tangible, e.g.
    - Flip a coin
      - Outcomes H, T
    - Flip a coin five times
      - Outcomes are sequences, e.g. HHHHH, HHHHT, HHHTH, and so on
    - Roll die 1 and die 2
      - Outcomes : faces shown, e.g. (die1 = 2, die2 = 4), (die1=1, die2=1), (die1=6, die2=1), ...
    - Run a system for 10,000 hours
      - Outcomes : Fails, does not fail

# Event Spaces

- Each outcome is from an *event space*  $\Omega$  of all possible outcomes
  - Experiment, flip a coin once,  $\Omega = \{H,T\}$
  - Experiment, flip a coin 5 times
    - $\Omega = \{ HHHHH, HHHHT, \dots, TTTHT, TTTTH, TTTTT \}$
  - Experiment, roll die1 and die2
    - $\Omega = \{(1,1),(1,2),(1,3), \dots, (6,4),(6,5),(6,6) \}$

# Event Spaces (cont'd)

Given an event space  $\Omega$  we *define* an event as a subset of  $\Omega$

Singleton event corresponds to one outcome, e.g.,

- coin face is 'H'
- Sequence of coin faces is 'HTHHT'
- die1 = 2, die2 = 4

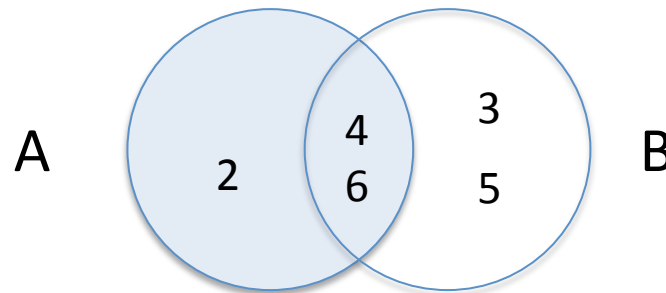
Some events refer to more interesting subsets, e.g.

- $\{(1,6),(2,5),(3,4),(4,3),(5,2),(6,1)\}$  is event that sum of die1 and die2 is 7
- Given an event and an outcome of an experiment, we can ask whether the outcome is in the event

# Algebra of Events

Ways in which events combine

- Union of A and B --- set-theoretic union, describes space of outcomes that are either in A, or B, or both
  - Example:  $A = \{ \text{die toss with even number} \}$ ,  $B = \{ \text{die toss with value at least 3} \}$
- Venn diagram view

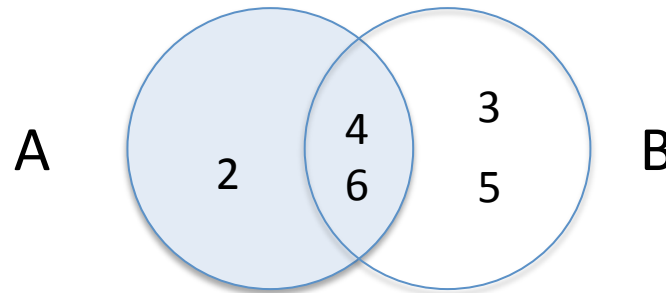


$$A \cup B = \{ 2, 3, 4, 5, 6 \}$$

# Algebra of Events

Ways in which events combine

- Intersection of A and B --- set-theoretic intersection, describes space of outcomes that are in A **and** B
- Example: A = { die toss with even number }, B = {die toss with value at least 3}
- Venn diagram view

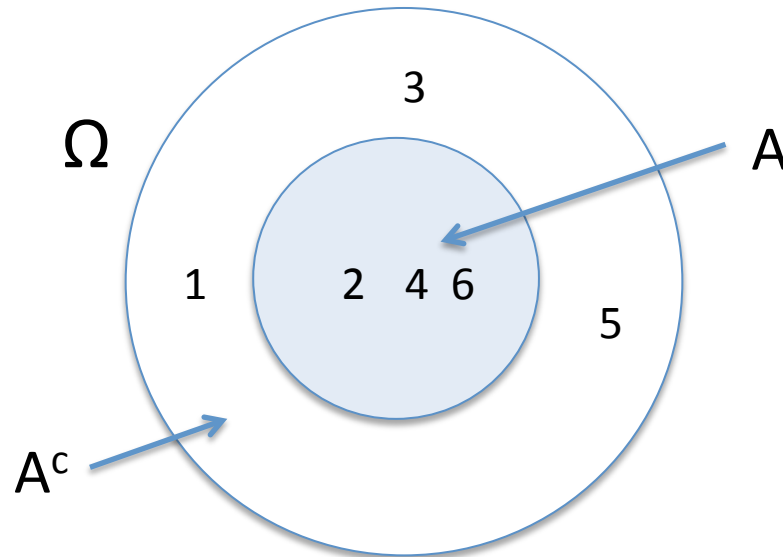


$$A \cap B = \{4, 6\}$$

# Algebra of Events

Ways in which events combine

- Complement of A --- set theoretic “entire event space excluding A”  $\Omega/A$  often denoted  $A^c$



# Boolean Algebra

Algebra of sets is Boolean algebra

- $\Omega$  plays the role of 1,  $\emptyset$  plays role of 0
- Set union plays the role of logical OR
- Set intersection plays the role of logical AND
- Set complement plays the role of NOT

$$\Omega \cap A = A, \Omega \cup A = \Omega, \emptyset \cap A = \emptyset, \emptyset \cup A = A$$

$$A \cup A^c = \Omega, \quad A \cap A^c = \emptyset$$

- Commutative laws  $A \cup B = B \cup A, A \cap B = B \cap A$

- Distributive laws

$$A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$$

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$$

- DeMorgan's Law

$$(A \cup B)^c = A^c \cap B^c, \quad (A \cap B)^c = A^c \cup B^c$$



# Now some “probability”

We aim to give weights to events

- N.B. For a given event  $A$  and outcome  $E$ , we can say that  $E$  is in  $A$  or not
  - Example:  $A = \{ \text{sum of two die is } 7 \}$ ,  $E = (\text{die1}=2, \text{die2}=6)$
- We have a notion then that if we repeated an experiment many times, some fraction of those outcomes would be in  $A$ , denote  $P(A)$ 
  - This the ‘frequency’ interpretation of probability

# Axioms of Probability

Function  $P: 2^\Omega \rightarrow [0,1]$  is a *probability measure* if

- If  $A_1, A_2, \dots, A_n$  are disjoint (i.e.,  $A_j \cap A_k = \emptyset$  for  $j \neq k$ ) then  $P(A_j \cup A_k) = P(A_j) + P(A_k)$
- $P(A) \geq 0$  for all events  $A \subseteq \Omega$
- $P(\Omega) = 1$

That's all!

# Consequences of Axioms

You can prove the following from these axioms

1.  $P(\emptyset) = 0$

2.  $P(A) + P(A^c) = 1$

# Consequences of Axioms

You can prove the following from these axioms

1.  $P(\emptyset) = 0$

$$A \cup \emptyset = \overset{\curvearrowright}{A}$$
$$P(A \cup \emptyset) = P(A) + P(\emptyset) = P(A)$$
$$\Rightarrow P(\emptyset) = 0$$

2.  $P(A) + P(A^c) = 1$

$$A \cup A^c = \Omega$$
$$P(A \cup A^c) = P(\Omega)$$
$$P(A) + P(A^c) = P(\Omega)$$
$$= 1$$

# Consequences of Axioms

You can prove the following from these axioms

$$3. P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

From point of view of Venn diagram think of  $P()$  as “area”

# Consequences of Axioms

You can prove the following from these axioms

$$3. P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

From point of view of Venn diagram think of  $P()$  as "area"



$$A = (A \cap B) \cup (A \cap B^c) \quad B = (A \cap B) \cup (B \cap A^c)$$

$$P(A) + P(B) - P(A \cap B)$$

$$P(A \cap B) + P(A \cap B^c) + P(A \cap B) + P(B \cap A^c) - P(A \cap B)$$

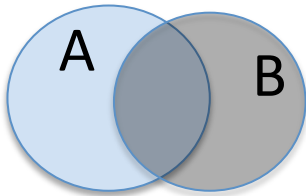
$$= P(A \cup B)$$

# Consequences of Axioms

You can prove the following from these axioms

$$3. P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

From point of view of Venn diagram think of  $P()$  as “area”

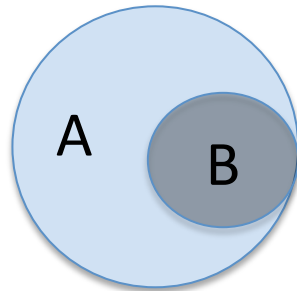


# Consequences of Axioms

You can prove the following from these axioms

4. If  $B \subset A$ , then  $P(B) \leq P(A)$ , furthermore

$$P(A/B) = P(A) - P(B)$$





# Consequences of Axioms

You can prove the following from these axioms

$$5. P(A_1 \cup A_2 \cup \dots \cup A_n) \leq P(A_1) + P(A_2) + \dots + P(A_n)$$

# Consequences of Axioms

You can prove the following from these axioms

$$5. P(A_1 \cup A_2 \cup \dots \cup A_n) \leq P(A_1) + P(A_2) + \dots + P(A_n)$$

Proof by Induction

$n=2$

$$P(A_1 \cup A_2) \leq P(A_1) + P(A_2)$$

$$P(A_1 \cup A_2) = P(A_1) + P(A_2) - P(A_1 \cap A_2)$$

Case proven for  $n=2$

Consider any  $n \geq 2$

Define  $B = A_2 \cup A_3 \cup \dots \cup A_n$

$$P(A_1 \cup A_2 \cup \dots \cup A_n) = P(A_1 \cup B) \quad \text{by induction hypothesis}$$

$$\leq P(A_1) + P(B)$$

$$P(B) = P(A_2 \cup \dots \cup A_n) \leq P(A_2) + \dots + P(A_n)$$

$$\leq P(A_1) + P(A_2) + \dots + P(A_n)$$

Proven

# Conditional Probability

Denote the probability of A given B by  $P(A | B)$

Frequency interpretation

- $n$  experiments,  $c(\omega, n)$  number whose outcome is  $\omega$
- $c(B, n)$  number whose outcomes are in B
- Frequency of  $\omega \in B$  relative to experiments with outcomes in B

$P(A|B)$  is the frequency of  $\omega \in A \cap B$  relative to the frequency of B

# Conditional Probability

Denote the probability of A given B by  $P(A | B)$

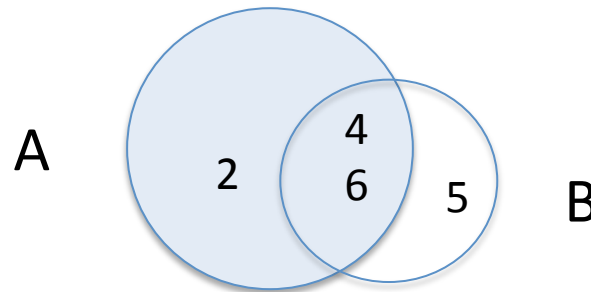
$$\begin{aligned} P(A|B) &= \sum_{\omega \in A \cap B} \lim_{n \rightarrow \infty} \frac{c(\omega, n)}{c(B, n)} \\ &= \frac{\sum_{\omega \in A \cap B} \lim_{n \rightarrow \infty} \frac{c(\omega, n)}{n}}{\lim_{n \rightarrow \infty} \frac{c(B, n)}{n}} \\ &= \frac{\lim_{n \rightarrow \infty} \frac{c(A \cap B, n)}{n}}{\lim_{n \rightarrow \infty} \frac{c(B, n)}{n}} = \frac{P(A \cap B)}{P(B)} \end{aligned}$$

# Conditional Probability

## Example

- A is die toss shows even number
  - But we know also the die toss is 4 or larger (event B)

## Venn diagram view



Frequency of  $\{2,4,6\}$  in space of  $\{4,5,6\}$  is same as frequency of  $\{4,6\}$  in  $\{4,5,6\}$  :  $( P(4)+P(6) ) / ( P(4) + P(5) + P(6) )$

In general

$$P(A | B) = P(A \cap B) / P(B)$$

# Independent Events

Events A and B are said to be independent if

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

Whether or not this is true depends on definitions of A and B

Example: 3 consecutive coin tosses

$$A = \{\text{at least 1 heads}\}, B = \{\text{exactly 1 heads}\}$$

$$A = \{HHH, HHT, HTH, THH, HTT, THT, TTH\}, P(A) = 7/8$$

$$B = \{HTT, THT, TTH\}, P(B) = 3/8$$

$$A \cap B = \{HTT, THT, TTH\}, P(A \cap B) = 3/8$$

$$P(A \cap B) = 3/8 \neq (7/8) * (3/8)$$

# Independent Events

Events A and B are said to be independent if

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

Whether or not this is true depends on definitions of A and B

Example: 3 consecutive coin tosses

A = {first two tosses have exactly one H}, B = {last toss is H}

$$A = \{ HTH, HTT, THH, THT \}, P(A) = 4/8 = 1/2$$

$$B = \{ HHH, HTH, THH, TTH \}, P(B) = 4/8 = 1/2$$

$$A \cap B = \{ HTH, THH \} = P(A \cap B) = 2/8 = 1/4$$

$$1/4 = P(A \cap B) = 1/2 * 1/2 = P(A) * P(B)$$

So A and B are independent

# Independent Events and Conditional Probability

An intuitive relationship:

if A is independent of B, then  $P(A)$  does not depend on B:

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



# Conditional Probability

Denote the probability of A given B by  $P(A | B)$

$P(A | B)$  is the frequency of  $\omega \in A \cap B$  relative to the frequency of B

$$\begin{aligned}P(A | B) &= \sum_{\omega \in A \cap B} \lim_{n \rightarrow \infty} c(\omega, n) / c(B, n) \\&= \lim_{n \rightarrow \infty} \sum_{\omega \in A \cap B} c(\omega, n) / c(B, n) \\&= \lim_{n \rightarrow \infty} \sum_{\omega \in A \cap B} (c(\omega, n) / n) / (c(B, n) / n) \\&= \left( \sum_{\omega \in A \cap B} \lim_{n \rightarrow \infty} c(\omega, n) / n \right) / \left( \lim_{n \rightarrow \infty} c(B, n) / n \right) \\&= \left( \lim_{n \rightarrow \infty} c(A \cap B, n) / n \right) / \left( \lim_{n \rightarrow \infty} c(B, n) / n \right) \\&= P(A \cap B) / P(B)\end{aligned}$$