

Lecture 4: Introduction to Probability

• 4.1 Introduction

- **Probability** is a numerical measure of the likelihood that an event will occur
- Probability values are always assigned on a scale from 0 to 1
 - A probability near 0 indicates an event is very unlikely to occur
 - A probability near 1 indicates an event is almost certain to occur
- **Why Should We Care:** we cannot predict events with absolute certainty. Instead, we make decisions about the data we have collected using probability – our inferences are stated in probabilistic terms
 - inference is made using **inductive** reasoning, that is, from a limited number of observations to general rules, and the uncertainty is expressed in probabilistic terms
- **Probability** is the study of **randomness** and **uncertainty**
- In the early days, probability was associated with *games of chance* (gambling)...The chance of winning ...
- Some simple **games**:
 - Game 1: A fair die is rolled. If the result is 2, 3, or 4, you win \$1; if it is 5, you win \$2; but if it is 1 or 6, you lose \$3. Should you play this game?
 - Game 2: Before two fair coins are tossed, you are given a choice of the following payoffs:
 - Payoff 1: Win \$1 for each head. Lose \$3 for getting two tails.
 - Payoff 2: Win \$1 if the coins are different. Win \$2 if both coins turn up tails. Lose \$3 if both coins turn up head.
 - Payoff 3: Win \$3 for getting two heads. Win \$1 for getting one of each. Lose \$4 for getting two tails.
 Which payoff (if any) should you choose?
 - Game 3: Game 3: A certain lottery has a jackpot of \$10M and the chance of winning is one in 500 million. Should you play this game?
- Modern **applications**:
 - Epidemiology
 - Weather forecast
 - Business applications: (Finance, Insurance, Marketing)

• 4.2 Experiments, Sample Spaces, and Events

- A **simple experiment** is the process of obtaining observations or measurements. More formally, an experiment is some well-defined act or process that leads to a single well-defined outcome

- Examples of simple experiments:

- tossing a coin
- picking a card from a deck
- measuring temperature from patients

- The **sample space** for an experiment is the set of all possible experimental outcomes

- A **sample point**, or an **elementary event**, is any member of the sample space

- Any set of elementary events is an **event**, or an **event class**. (e.g., H,T; spades, clubs)

- Example: The experiment is to toss a coin 3 times

- Sample space $S = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\}$

- Examples of event include

$$A = \{HHH, HHT, HTH, THH\} = \{\text{at least two heads}\}$$

$$B = \{HTT, THT, TTH\} = \{\text{exactly two tails}\}$$

• 4.3 Probabilities

- **Definition:** Given a sample space S and the family of events in S , a probability function associates with each event $A \subseteq S$, $p(A)$, the probability of event A , such that the following axioms are true:

- $p(A) \geq 0$, $A \subseteq S$

- $p(S) = 1$

- If there exists some *countable* set of events, $\{A_1, A_2, \dots, A_N\}$, and if these events are all mutually exclusive,

$$p\left(\bigcup_{i=1}^N A_i\right) = \sum_{i=1}^N p(A_i)$$

[The probability of the union of the A_N mutually exclusive events is equal to the sum of their separate probabilities]

• 4.4 Probability Rules

- There are some **basic probability relationships** that can be used to compute the probability of an event without knowledge of all the sample point probabilities

\emptyset is the empty set; that is, a set that contains no events

$A \subset S$ A is a subset of S , or is fully contained in S

$A \cup B$ A union B , the set of all elements that are either in A or B or both
($A \cup B = S$)

$A \cap B$ A intersection B , the set of all elements that are in both A and B
(if A and B are mutually exclusive, $A \cap B = \emptyset$)

A^c A complement; that is, the event consisting of all sample points that are not in A (if A and B are mutually exclusive, $A^c = B$ and $B = A^c$, $\sim A$, or \overline{A})

$A \cup A^c = S$ A union its complement is S

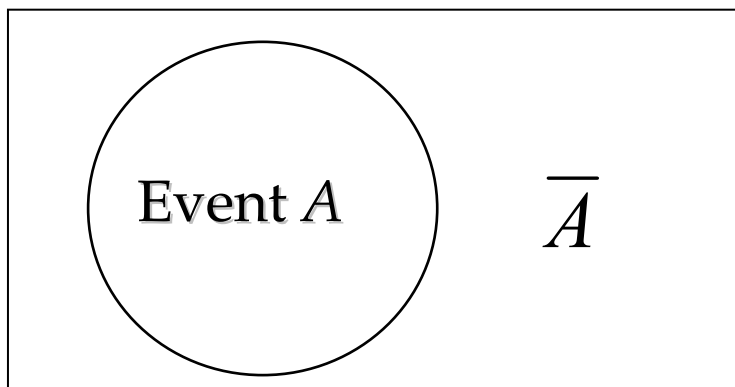
- **Rule 1: The complement rule of probability**

- The **complement** of an event A is defined to be the event consisting of all sample points that are not in A

- $p(\sim A) = 1 - p(A)$

- The complement of A is denoted by $\sim A$ or A^c or \overline{A}

- The **Venn diagram** below illustrates the concept of a complement



- **Rule 2: Probability Range**

$$0 \leq p(A) \leq 1$$

- **Rule 3: Rule of the impossible event**

$$p(\emptyset) = 0, \text{ for any } \mathcal{S}$$

- **Rule 4: The “or” rule of probability**

- The probability that event A or B (both in \mathcal{S}) will occur is

$$p(A \cup B) = p(A) + p(B) - p(A \cap B)$$

where $P(A \cap B)$ is the probability that both events, A and B, will occur

- A and B are **mutually exclusive events** if both cannot occur together in the same experiment

$$p(A \cap B) = 0$$

- The **Additional Rule for Mutually Exclusive Events:**

If A, B, C, ... are mutually exclusive events, then

$$p(A \cup B \cup C \cup \dots) = p(A) + p(B) + p(C) + \dots$$

- **The Multiplication Rule of Probability**

- The probability that both A and B will occur when an experiment is performed is given by

$$p(A \cap B) = p(A) \cdot p(B|A),$$

where $p(B|A)$ is the probability of B if A has occurred, and is called the **conditional probability** of B given A

- Events A and B are **independent** if the probability of each event is not affected by whether or not the other event has occurred

- **The Multiplication Rule for Independent Events:**

- If A, B, C, ... are independent events, then

$$p(ABC \dots) = p(A) \cdot p(B) \cdot p(C) \cdot \dots$$

- **Test for Independent Events:**

- A and B are **independent** if either $p(A|B) = p(A)$ or $p(AB) = p(A) \cdot p(B)$. Otherwise, the events are dependent

- **The Condition Rule of Probability**

- The conditional probability of B , given that A has occurred, is given by

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

Provided that $A \neq 0$

• **4.5 Assigning Probabilities**

- **Relative Frequency Method** - Assigning probabilities based on experimentation or historical data

- The probability of event E is denoted by $P(E)$ and is the proportion of the time that E can be expected to occur **in the long run**

- If we let A be the event

A : a female birth in the United States

then the probability of E is estimated to be

$$P(E) \approx \frac{1,983,000}{4,065,000} = .49$$

- **Bernoulli's Theorem**

- If the probability of occurrence of the event X is $p(X)$ and if N trials are made, independently and under exactly the same conditions, the probability that the relative frequency of occurrence of X differs from $p(X)$ by any amount, however small, approaches zero as the number of trial grows indefinitely large

$$\lim_{N \rightarrow \infty} F(X) - p(X) = 0,$$

where $F(X)$ is the relative frequency of X to N

- **Classical/Theoretical Method** - Assigning probabilities based on the assumption of equally likely outcomes

- Suppose a sample space has S equally likely points, and A is an event consisting of a points. Then

$$p(A) = \frac{a}{s}$$

- The sample space of the experiment is

(a,a)	(a,b)	(a,c)
(b,a)	(b,b)	(b,c)
(c,a)	(c,b)	(c,c)

The sample space has $s = 9$ equally likely outcomes

- The event of interest is
 A : both answers are correct

Event A contains $a=1$ point. Therefore, by the classical method of assigning probabilities, the probability that the student will answer both questions correctly is

$$P(A) = \frac{a}{s} = \frac{1}{9}$$

- **Subjective Method** - Assigning probabilities based on the assignor's judgment

- When economic conditions and a company's circumstances change rapidly it might be inappropriate to assign probabilities based solely on historical data

- We can use any data available as well as our experience and intuition, but ultimately a probability value should express our **degree of belief** that the experimental outcome will occur

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

• 4.6 Probability Distributions

- A **probability distribution** is any statement of a function associating each of a set of mutually exclusive and exhaustive events with its probability
- The sum of the probabilities of all classes must be 1.00

Probability Distribution

Height (inches)	p
78-82	.002
73-77	.021
68-72	.136
63-67	.682
58-62	.136
53-57	.021
48-52	.002
	1.000

- Given a frequency distribution, each interval has associated a probability, as

$$p(A_I) = f(A_I) / f(A_N) ,$$

or the frequency of a particular interval divided by the total frequency

- Given a theoretical probability distribution and N observations made independently and at random with replacement, there is a **theoretical frequency distribution**, in which for any event class A

$$F(A) \text{ from } N = Np(A),$$

where $F(A)$ is the theoretical frequency of event A out of N observations

Theoretical Frequency Distribution

Height (inches)	f
78-82	2
73-77	21
68-72	136
63-67	682
58-62	136
53-57	21
48-52	2
	$1,000 = N$

• 4.7 Random Variables

- Let X be a function that associates a real number with each and every elementary event in some sample space \mathcal{S} . Then X is called a **random variable** on the sample space \mathcal{S}

- A random variable X represents values that are associated with elementary events, so that particular values of X occur when the appropriate elementary events occur, however the association might be

- Random variables can be specified in three ways: listing all possible numerical events and their associated probability, graphing this relationship, or expressing a rule of the probability for each value

- Typical notation is to use capital letters, such as X, Y, Z to denote random variables and lowercase letters, such as x, y, z to denote particular values of the random variable

- A variable X is said to be a **discrete random variable** if it can assume only a particular finite or a countable infinite set of values. Given discrete random variables, probability calculations are often simple

- Example: two dices, and $X = \text{die}_1 + \text{die}_2$

Values of X for all elementary event in \mathcal{S}

6	7	8	9	10	11	12
5	6	7	8	9	10	11
4	5	6	7	8	9	10
3	4	5	6	7	8	9
2	3	4	5	6	7	8
1	2	3	4	5	6	7
	1	2	3	4	5	6

- From the definition of X we can now estimate the probability distribution of all values x of X , as

Probability Distribution for X

X	p
12	1/36
11	2/36
10	3/36
9	4/36
8	5/36
7	6/36
6	5/36
5	4/36
4	3/36
3	2/36
2	1/36
	36/36

- And from this distribution of X , one can compute other probability questions

- Example 1: $p(3 \leq X \leq 5)$?

$$\begin{aligned} p(3 \leq X \leq 5) &= p(3 \cup 4 \cup 5) = p(3) + p(4) + p(5) \\ &= p(X=3) + p(X=4) + p(X=5) \\ &= 2/36 + 3/36 + 4/36 = 9/36 = 1/4 = .25 \end{aligned}$$

- Example 2: $p(X < 5)$?

$$\begin{aligned} p(X < 5) &= p(X=2) + p(X=3) + p(X=4) \\ &= 1/36 + 2/36 + 3/36 = 6/36 = 1/6 \end{aligned}$$

• 4.8 Function Rules for Discrete Random Variables

- Sometimes the most suitable way to specify the distribution of a random variable is by its rule or function

- Example 1: Let X be a random variable that can take one of six values 1, 2, ..., 6. If all values have exactly the same probability of occurrence, the function rule for X can be expressed as

$$p(x) = \begin{cases} 1/6 & (\text{if } x=1, 2, \dots, 6) \\ 0 & (\text{otherwise}) \end{cases}$$

- Example 2: Let X be a random variable that can take one of six values 1, 2, ..., 6. Let the function rule for X be expressed as

$$p(x) = \begin{cases} x/12 & (\text{if } x=1, 2, 3) \\ (7-x)/12 & (\text{if } x=4, 5, 6) \\ 0 & (\text{otherwise}) \end{cases}$$

What are the probabilities to values of X ?

$$p(X=1) = 1/12; \quad p(X=2) = 2/12; \quad p(X=3) = 3/12;$$

$$p(X=4) = (7-4)/12 = 3/12;$$

$$p(X=5) = (7-5)/12 = 2/12;$$

$$p(X=6) = (7-6)/12 = 1/12;$$

• 4.9 Continuous Random Variables

- A random variable is **continuous** when it can be represented in terms of arbitrarily small class intervals of size ΔX , with the probability of any interval corresponding exactly to the area cut off by the interval under a smooth curve
- As ΔX approaches 0, the p associated with any class interval also approaches 0, because the corresponding area under the curve is being reduced
- Hence, the occurrence of **any exact value** of X is said to have zero probability
- And probabilities are not discussed for a particular value of a continuous random variable X , but for intervals of X in a continuous distribution
- Similarly, for a particular value a , probability is described as the **probability density** of X at value a , expressed as

$$f(a) = \text{probability density of } X \text{ at } a$$

- The probability density of X at a is the **rate of change** in the p of an interval with lower limit a , for very small changes in the size of the interval
- This rate of change depends on:
 - the function rule assigning probabilities to intervals
 - the particular region of X values of interest
- For discrete random variables, probability at value x is equivalent to density at value x ; this is not true for continuous random variables
- The probability of an interval for a continuous random variable depends on the weighted sum of the probability densities associated with all the values in the interval, as

$$p(a \leq X \leq b) = \int_a^b f(x)dx.$$

or the area cut off by the interval under the curve for the probability densities, when X is a continuous random variable and the total area = 1

- The distribution of a random variable can also be described using the **cumulative distribution function**. This function represents the relation between the possible values a of a random variable X and the probability that the value of X is less than or equal to a
 - The cumulative probability $F(a) = p(X \leq a)$
 - Similarly, $F(b) - F(a) = p(a \leq X \leq b)$

- Graphic representation of continuous distributions (probability densities, areas, and probabilities in a continuous distribution).

