

# Part IV: Probability Theory

L13: Introduction to probability

L14: Conditional probability and Bayes' Theorem

L15: Expectation and variance

# L13: Introduction to Probability

- Reading: Rosen 7.1, 7.2

# The Hatcheck Problem Revisited

- Total number of permutations:  $n!$
- Number of derangements:

$$D_n = n! \left[ 1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \cdots (-1)^n \frac{1}{n!} \right]$$

- Probability of that a random permutation is a derangement:

$$p = \frac{D_n}{n!} = 1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \cdots (-1)^n \frac{1}{n!}$$

# Definitions

- An **experiment** is a procedure that yields one of a given set of possible outcomes.
- The **sample space** of the experiment is the set of all possible outcomes.
- An **event** is a subset of the sample space.
- If  $S$  is a finite sample space of equally likely outcomes, and  $E$  is an event, i.e., a subset of  $S$ , then the **probability** of  $E$  is

$$p(E) = \frac{|E|}{|S|}$$

# Probability = Counting

- Example 1

A bag contains 4 blue balls and 5 red balls. What is the probability that a ball chosen from the bag is blue?

- Solution

$4/9$

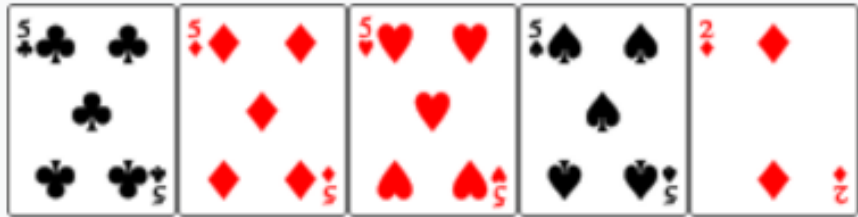

- Example 2

What is the probability that when two dice are rolled, the sum of the numbers on the two dice is 7?

- Solution

By the product rule there are  $6^2 = 36$  possible outcomes. Six of these sum to 7. Hence, the probability of obtaining a 7 is  $6/36 = 1/6$ .

# Poker

Four of a kind	
Full house	

- Probability of four of a kind

Total number of hands of five cards:  $C(52,5)$

Number of ways to get four of a kind:  $13 \times 48$

Probability =  $\frac{13 \times 48}{C(52,5)} \approx 0.00024$
- Probability of full house

Total number of hands of five cards:  $C(52,5)$

Number of ways to get full house:  $P(13,2)C(4,3)C(4,2)$

Probability =  $\frac{P(13,2)C(4,3)C(4,2)}{C(52,5)} \approx 0.0014$

# Mark Six (六合彩)

- Player chooses 6 numbers from 49
- 6 numbers are drawn + 1 extra

Prize	Criteria	Probability
1st Division	All 6 drawn numbers	$\frac{1}{\binom{49}{6}} = \frac{1}{13,983,816}$
2nd Division	5 out of 6 drawn numbers, plus the extra number	$\frac{\binom{6}{5}}{\binom{49}{6}} = \frac{1}{2,330,636}$
3rd Division	5 out of 6 drawn numbers	$\frac{\binom{6}{5} \binom{42}{1}}{\binom{49}{6}} \approx \frac{1}{55,491.33}$



# Sampling

- What is the probability that the numbers 11, 4, 17, 39, and 23 are drawn in that order from a bin containing 50 balls labeled with the numbers 1, 2, ..., 50 if
  - (a) the ball selected is not returned to the bin before the next ball is selected (**sampling without replacement**), and
  - (b) the ball selected is returned to the bin before the next ball is selected (**sampling with replacement**)?
- Solution
  - Number of ways the event happens: 1
  - Total number of ways to draw numbers:
    - (a)  $P(50, 5)$ , probability is  $1/P(50, 5)$
    - (b)  $50^5$ , probability is  $1/50^5$



# Complement of Event

- Theorem

Let  $E$  be an event in a finite sample space  $S$ . The probability of the event  $\bar{E}$ , the complementary event of  $E$ , is given by  $p(\bar{E}) = 1 - p(E)$ .

- Proof

Using the fact that  $|\bar{E}| = |S| - |E|$ ,

$$p(\bar{E}) = \frac{|S| - |E|}{|S|} = 1 - \frac{|E|}{|S|} = 1 - p(E).$$

# Complement of Event

- **Example:**

A sequence of 10 bits is chosen randomly. What is the probability that at least one of these bits is 0?

- **Solution:**

$E$ : the event that at least one of the 10 bits is 0.

$\bar{E}$ : the event that all of the bits are 1s.

The size of the sample space  $S$  is  $2^{10}$ . Hence,

$$p(E) = 1 - p(\bar{E}) = 1 - \frac{|\bar{E}|}{|S|} = 1 - \frac{1}{2^{10}} = 1 - \frac{1}{1024} = \frac{1023}{1024}.$$

# Union of Events

- **Theorem**

Let  $E_1$  and  $E_2$  be events in the sample space  $S$ . Then

$$p(E_1 \cup E_2) = p(E_1) + p(E_2) - p(E_1 \cap E_2)$$

- **Proof:**

Given the inclusion-exclusion formula  $|A \cup B| = |A| + |B| - |A \cap B|$ , it follows that

$$\begin{aligned} p(E_1 \cup E_2) &= \frac{|E_1 \cup E_2|}{|S|} = \frac{|E_1| + |E_2| - |E_1 \cap E_2|}{|S|} \\ &= \frac{|E_1|}{|S|} + \frac{|E_2|}{|S|} - \frac{|E_1 \cap E_2|}{|S|} \\ &= p(E_1) + p(E_2) - p(E_1 \cap E_2). \end{aligned}$$

# Union of Events

- **Example:**

What is the probability that a positive integer selected at random from the set of positive integers not exceeding 100 is divisible by either 2 or 5?

- **Solution:**

$E_1$ : the event that the integer is divisible by 2;

$E_2$ : the event that it is divisible 5;

$E_1 \cup E_2$ : The event that the integer is divisible by 2 **or** 5;

$E_1 \cap E_2$ : The event that it is divisible by 2 **and** 5.

It follows that:

$$\begin{aligned} p(E_1 \cup E_2) &= p(E_1) + p(E_2) - p(E_1 \cap E_2) \\ &= 50/100 + 20/100 - 10/100 = 3/5. \end{aligned}$$

# Inclusion-Exclusion Principle for Probability

- **Theorem**

$$p\left(\bigcup_{i=1}^n E_i\right) = \sum_{1 \leq i \leq n} p(E_i) - \sum_{1 \leq i < j \leq n} p(E_i \cap E_j) + \sum_{1 \leq i < j < k \leq n} p(E_i \cap E_j \cap E_k) \\ - \dots + (-1)^{n+1} p(E_1 \cap E_2 \cap \dots \cap E_n).$$

# Probability Distribution

- **Definition**

Let  $S$  be the sample space of an experiment with a finite number of outcomes. A **probability distribution** on  $S$  is characterized by a **probability mass function (pmf)**  $p: S \rightarrow \mathbf{R}$  such that

(a)  $0 \leq p(s) \leq 1$  for each  $s \in S$ ,

(b)  $\sum_{s \in S} p(s) = 1$ ,

where  $p(s)$  is the probability of an outcome  $s$ .

# Uniform Distribution

- **Definition**

Suppose that  $S$  is a set with  $n$  elements. The **uniform distribution** assigns the probability  $1/n$  to each element of  $S$ .

- **Example:**

For a fair dice with 6 sides, we have  $p(x) = \frac{1}{6}$  for all  $x$ .

What is the probability that an odd number appears when we roll this dice?

# Non-Uniform Distribution

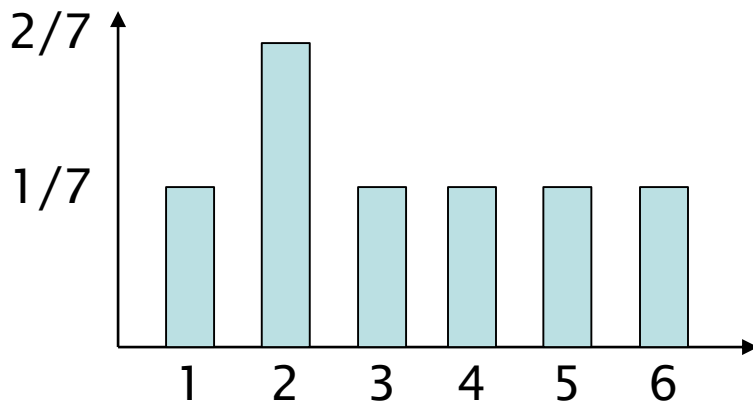
- **Example:**

Suppose that a dice is biased so that

$$(1) \ p(2) = \frac{2}{7},$$

$$(1) \ p(x) = \frac{1}{7} \text{ for all other } x.$$

What is the probability that an odd number appears when we roll this dice?



Histogram of the pmf



# Probability of an Event

- **Definition:**

The probability of the event  $E$  is the sum of the probabilities of the outcomes in  $E$ .

$$p(E) = \sum_{s \in S} p(s)$$

- Uniform distributions: probability = count
- General distributions: probability = sum

# Complement and Union of Events

- Complement:  $p(\overline{E}) = 1 - p(E)$
- Union:  $p(E_1 \cup E_2) = p(E_1) + p(E_2) - p(E_1 \cap E_2)$
- Inclusion-exclusion principle for probability

$$p\left(\bigcup_{i=1}^n E_i\right) = \sum_{1 \leq i \leq n} p(E_i) - \sum_{1 \leq i < j \leq n} p(E_i \cap E_j) + \sum_{1 \leq i < j < k \leq n} p(E_i \cap E_j \cap E_k) \\ - \dots + (-1)^{n+1} p(E_1 \cap E_2 \cap \dots \cap E_n).$$

- Disjoint union:  
If  $E_1, E_2, \dots$  is a sequence of pairwise disjoint events in a sample space  $S$ , then

$$p\left(\bigcup_i E_i\right) = \sum_i p(E_i)$$

# Independence

- **Example**

Roll two dice. What's the probability that they both turn up heads?

- **Definition**

Two events  $E$  and  $F$  are independent if and only if  $p(E \cap F) = p(E)p(F)$ .

- How to check independence?
  - Two unrelated events are independent
  - Use definition

# Example

- **Example**

In a randomly generated bit string of length 4:

$E$ : it begins with a 1

$F$ : it contains an even number of 1s.

Are  $E$  and  $F$  independent?

- **Solution**

- $p(E) = 1/2$ .
- $p(F) = 1/2$ .
- $p(E \cap F) = |\{1111, 1100, 1010, 1001\}|/16 = 1/4$ .
- So  $E$  and  $F$  are independent.

# Example

- **Example:**

Suppose a family have 3 children, each of which has equal probability to be a boy or a girl. Are these two events independent?

- $E$ : The family has children of both sexes
- $F$ : The family has at most one boy

- **Solution:**

- $$p(E) = \frac{|\{BBG, BGB, BGG, GBB, GBG, GGB\}|}{8} = \frac{3}{4}$$

- $$p(F) = \frac{|\{BGG, GBG, GGB, GGG\}|}{8} = \frac{1}{2}$$

- $$p(E \cap F) = \frac{|\{BGG, GBG, GGB\}|}{8} = \frac{3}{8}$$

- So they are independent

# Pairwise and Mutual Independence

- **Definition:**

The events  $E_1, E_2, \dots, E_n$  are **pairwise independent** if and only if  $p(E_i \cap E_j) = p(E_i) p(E_j)$  for all pairs  $i$  and  $j$  with  $1 \leq i < j \leq n$ .

- **Definition:**

The events are **mutually independent** if

$$p(E_{i_1} \cap E_{i_2} \cap \dots \cap E_{i_m}) = p(E_{i_1})p(E_{i_2}) \dots p(E_{i_m})$$

whenever  $i_j, j = 1, 2, \dots, m$ , are integers with

$$1 \leq i_1 < i_2 < \dots < i_m \leq n \text{ and } m \geq 2.$$

- **Note**

- Mutual independence  $\Rightarrow$  pairwise independence,
- but the reverse is not necessarily true.

# Random Variables

- **Definition**

A **random variable** is a function from the sample space of an experiment to the set of real numbers, i.e., a random variable assigns a real number to each possible outcome.

- Note: This function is not random!

- **Example**

Suppose that a coin is flipped three times. Let  $X(t)$  be the random variable that equals the number of heads that appear when  $t$  is the outcome. Then  $X(t)$  is the function where:

$$X(HHH) = 3, X(TTT) = 0,$$

$$X(HHT) = X(HTH) = X(THH) = 2,$$

$$X(TTH) = X(THT) = X(HTT) = 1.$$

# Distribution of a Random Variable

- **Definition**

The **distribution** of a random variable  $X$  on a sample space  $S$  is the set of pairs  $(r, p(X = r))$  for all  $r \in X(S)$ , where  $p(X = r)$  is the probability that  $X$  takes the value  $r$ , i.e.,  $p(X = r) = \sum_{s \in S, X(s)=r} p(s)$

- **Example**

Continuing with the previous example, we have

$$p(X = 3) = 1/8$$

$$p(X = 2) = 3/8$$

$$p(X = 1) = 3/8$$

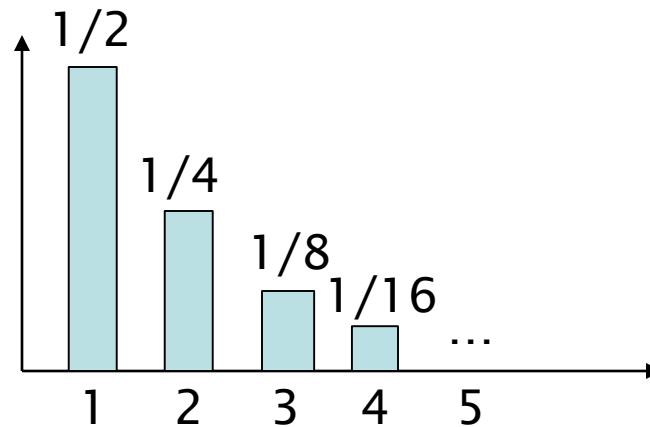
$$p(X = 0) = 1/8$$



# Infinite Sample Space

- **Example**

Experiment: Flip a fair coin until it turns up heads.  
Let random variable  $X$  = the number of flips.

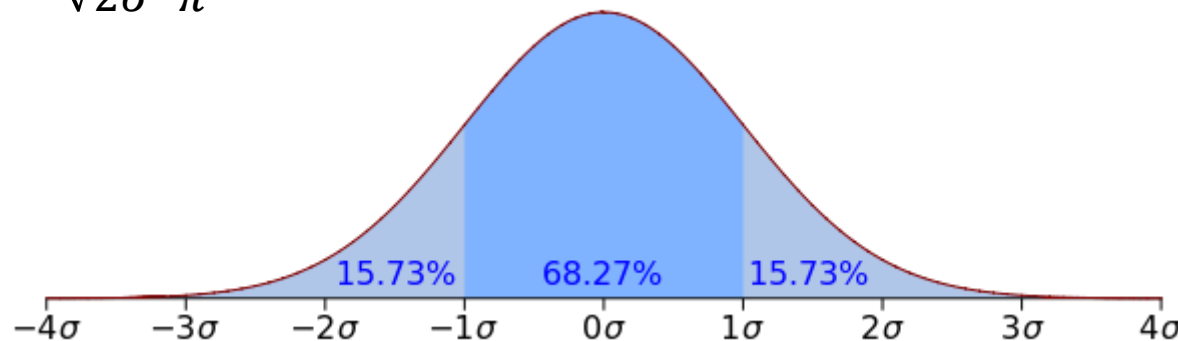


- Verify  $\sum_{x=1}^{\infty} p(X = x) = \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \dots = 1$
- The probability of the event that we get 3 or more flips =  $p(X = 3) + p(X = 4) + \dots = \frac{1}{8} + \frac{1}{16} + \dots = \frac{1}{4}$ .

# Continuous Probability Distribution

- Defined over a domain of real numbers
- Characterized by a **probability density function (pdf)**
- Example: The pdf of a normal (Gaussian) distribution

is  $f(x) = \frac{1}{\sqrt{2\sigma^2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$ .



- Remark:  $f(x)$  does **not** represent a probability!
- The probability of an interval:  $\Pr[a \leq X \leq b] = \int_a^b f(x)dx$

# Bernoulli Trials

- **Definition:**

Each performance of an experiment with two possible outcomes is called a **Bernoulli trial**. The two possible outcomes are generally referred to as **success** and **failure**.

- **Example:**

A coin is biased so that the probability of heads is  $2/3$ . What is the probability that exactly four heads occur when the coin is flipped seven times?

- **Solution:**

$$C(7,4) \left(\frac{2}{3}\right)^4 \left(\frac{1}{3}\right)^3$$

# Binomial Distribution

- **Theorem**

Let  $X$  be the number of successes in  $n$  independent Bernoulli trials, with probability of success  $p$  and probability of failure  $q = 1 - p$ . Then

$$p(X = k) = b(k: n, p) = C(n, k)p^k q^{n-k}$$

- **Definition:**

$X$  follows the **binomial distribution** with parameters  $n$  and  $p$ , and  $b(k: n, p)$  is the pmf.

- Check (from the binomial theorem)

$$\sum_{k=0}^n \binom{n}{k} p^k q^{n-k} = (p + q)^n = 1$$

# Binomial Distribution

- **Question**

What is  $\Pr(10 \leq X \leq 20)$  ?

- **Solution**

$$\sum_{k=10}^{20} \binom{n}{k} p^k q^{n-k},$$

which, unfortunately, has no closed form.

- For large  $n$ , and  $p$  not too close to 0 or 1, the normal distribution approximates binomial distribution well.
  - A special case of the central limit theorem.

