

# Probability Review

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CSC 343-643

WAKE FOREST  
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Department of Computer Science

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## Basic Definitions

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- **Random experiment** - Observing the outcome of a chance event
- **Elementary outcomes** - All possible results of the experiment
- **Sample space** - Set of all the elementary outcomes
- Consider tossing a coin
  - Random experiment consists of recording the outcome
  - Elementary outcomes are heads and tails
  - Sample space is {heads, tails}

*What about rolling a single die?*

## Probability

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- We want to assign a numerical weight to each outcome
  - This **probability** measures the likelihood of it occurring
  - Denoted as  $p\{x_i\}$ , which is the probability of  $x_i$
- Consider rolling a fair die and recording the side facing up
  - Since the die is fair, all sides have an equal probability
  - $p\{1\} = p\{2\} = p\{3\} = p\{4\} = p\{5\} = p\{6\} = \frac{1}{6}$
- Of course a *loaded* die would have one side appear more often
  - $p\{1\} = \frac{1}{4}$
  - $p\{2\} = p\{3\} = p\{4\} = p\{5\} = p\{6\} = \frac{3}{20}$

*What is the assumption for the remaining sides?*

## Characteristic Properties of Probability

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- Probabilities are never negative
  - $p\{x_i\} \geq 0$
  - Probability of zero means the event cannot happen
- Probabilities are never greater than 1
  - $p\{x_i\} \leq 1$
  - Probability of 1 means the event is certain to happen
- The total probability of all possible outcomes is one

$$\sum_{i=1}^n p\{x_i\} = p\{x_1\} + \dots + p\{x_n\} = 1$$

*Did the previous examples adhere to these rules?*

## Events

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- An **event** is a set of elementary outcomes
  - The probability of an event is the sum of the probabilities of the elementary outcomes
- Consider rolling a pair of dice

Event	Elementary Outcomes	Probability
A = dice add to three	$\{(1, 2), (2, 1)\}$	$p\{A\} = \frac{2}{36}$
B = dice add to six	$\{(1, 5), (2, 4), (3, 3), (4, 2), (5, 1)\}$	$p\{B\} = \frac{5}{36}$
C = first die shows 1	$\{(1, 1), (1, 2), (1, 3), (1, 4), (1, 5), (1, 6)\}$	$p\{C\} = \frac{6}{36}$
D = second die shows 1	$\{(1, 1), (2, 1), (3, 1), (4, 1), (5, 1), (6, 1)\}$	$p\{D\} = \frac{6}{36}$

*Where did the denominator of 36 come from?*

## Combining Events

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- Given two events, we can combine them to create a new event
  - **E and F**, the event E and the event F both occur
  - **E or F**, the event E or the event F (or both) occur
  - **not E**, the event E does not occur
- Remember events can be represented as sets
  - Combining events is really just set manipulation

## Addition Rule

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- For any events E and F,

$$p\{E \text{ or } F\} = p\{E\} + p\{F\} - p\{E \text{ and } F\}$$

- Adding *double counts* the elementary outcomes shared by E and F, so we must subtract the extra amount

- *What is the probability of the first or second die being one?*

$$p\{C \text{ or } D\} = p\{C\} + p\{D\} - p\{C \text{ and } D\} = \frac{6}{36} + \frac{6}{36} - \frac{1}{36} = \frac{11}{36}$$

- If E and F are mutually exclusive, then the two events will not *double count*, as a result

$$p\{E \text{ or } F\} = p\{E\} + p\{F\}$$

*What is a dice example of this?*

## Multiplication Rule

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- For any events E and F,

$$p\{E \text{ and } F\} = p\{E|F\} \cdot p\{F\}$$

- Where  $p\{E|F\}$  is the conditional probability of E will occur given that F has already occurred

- If E and F are independent, then the occurrence of one has no influence on the other

$$p\{E \text{ and } F\} = p\{E\} \cdot p\{F\}$$

- *What is the probability of snake eyes?*

$$p\{\text{both die are } 1\} = p\{\text{first die is } 1\} \cdot p\{\text{second die is } 1\} =$$

$$\frac{1}{6} \cdot \frac{1}{6} = \frac{1}{36}$$

## Subtraction Rule

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- For any event E,

$$p\{\text{not } E\} = 1 - p\{E\}$$

- *What is the probability of not rolling snake eyes?*

$$\begin{aligned} p\{\text{both die are not } 1\} &= 1 - p\{\text{both die are } 1\} = \\ &1 - \frac{1}{36} = \frac{35}{36} \end{aligned}$$

## Random Variables

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- A random variable,  $X$ , is the numerical outcome of an experiment
- For example  $X$  could represent
  - The sum of the faces of a pair of dice
  - Number of heads when two coins are tossed
  - Number of bits transmitted in error in a frame
- We want to observe the probabilities of the outcomes
  - The probability of the RV  $X$  having the value  $x$

$$Pr\{X = x\} = p\{x\}$$

## RV Examples

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- Consider tossing two coins, let  $X$  = number of heads

$x$	0	1	2
$Pr\{X = x\}$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$

*How did we get these probability values?*

- Consider rolling a pair of dice, let  $X$  = sum of the dice

$x$	2	3	4	5	6	7	8	9	10	11	12
$Pr\{X = x\}$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

## Mean of a RV

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- The mean of a random variable  $X$  is

$$\sum_{\forall x} x_i \cdot p\{x_i\}$$

– Sum of the possible values, each weighted by its probability

- Is also called the **expected value** of  $X$ ,  $e[X]$
- The expected number of heads when tossing two coins is

$$e[X] = \sum_{i=0}^2 x_i \cdot p\{x_i\} = 0 \cdot \frac{1}{4} + 1 \cdot \frac{1}{2} + 2 \cdot \frac{1}{4} = 1$$

## Bernoulli Trial

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- An experiment (trial) that has the following properties
  1. Result of a trial is either a *success* or a *failure*
  2. The probability  $p$  of success is the same in every trial
  3. Independence, one trial has no influence on later outcomes
- Often used in network modeling

*What are some other example experiments that can be considered a Bernoulli trial?*

## Binomial RV

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- $X$  is the number of successes in  $n$  repeated Bernoulli trials with probability  $p$
- For example, number of heads in two flips of a coin, ( $n = 2, p = \frac{1}{2}$ )

$k = \text{number of successes}$	0	1	2
$Pr\{X = k\}$	$\frac{1}{4}$	$\frac{1}{2}$	$\frac{1}{4}$

*How did we get the probability values?*

- The equation is

$$Pr\{X = k\} = \binom{n}{k} p^k (1-p)^{n-k}$$

- Where

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

- Is called  $n$  choose  $k$ , which counts the possible ways of getting  $k$  successes in  $n$  trials

### Example problem

*Three bits are grouped together to form a frame and transmitted over a channel. Assuming the probability of a single bit error is 0.25 and errors are independent...*

- The elementary outcomes are
  - Let b represent the original bit, e represent an error

Bit Pattern	Probability
[bbb]	$\frac{3}{4} \frac{3}{4} \frac{3}{4} = 0.421875$
[ebb]	$\frac{1}{4} \frac{3}{4} \frac{3}{4} = 0.140625$
[beb]	$\frac{3}{4} \frac{1}{4} \frac{3}{4} = 0.140625$
[bbe]	$\frac{3}{4} \frac{3}{4} \frac{1}{4} = 0.140625$
[eeb]	$\frac{1}{4} \frac{1}{4} \frac{3}{4} = 0.046875$
[bee]	$\frac{3}{4} \frac{1}{4} \frac{1}{4} = 0.046875$
[ebe]	$\frac{1}{4} \frac{3}{4} \frac{1}{4} = 0.046875$
[eee]	$\frac{1}{4} \frac{1}{4} \frac{1}{4} = 0.015625$

- What is the probability that a frame has at least one error?

- Could use *brute force*

$$p\{\text{at least one error}\} = p\{[ebb]\} + p\{[beb]\} + p\{[bbe]\} + p\{[eeb]\} + p\{[bee]\} + p\{[ebe]\} + p\{[eee]\} = 0.578125$$

- Could use Binomial RV

$$p\{\text{at least one error}\} = p\{1 \text{ error}\} + p\{2 \text{ errors}\} + p\{3 \text{ errors}\} = \binom{3}{1} \left(\frac{1}{4}\right)^1 \left(\frac{3}{4}\right)^2 + \binom{3}{2} \left(\frac{1}{4}\right)^2 \left(\frac{3}{4}\right)^1 + \binom{3}{3} \left(\frac{1}{4}\right)^3 \left(\frac{3}{4}\right)^0 = 0.578125$$

- Could use *not*

$$p\{\text{at least one error}\} = 1 - p\{\text{no error}\} = 1 - 0.421875 = 0.578125$$

- Assume the third bit is a parity bit, what is the probability of an undetected error? Parity cannot detect an even number of errors

- Brute force

$$p\{2 \text{ errors}\} = p\{[eeb]\} + p\{[bee]\} + p\{[ebe]\} = 0.140625$$

- Binomial RV

$$\binom{3}{2} \left(\frac{1}{4}\right)^2 \left(\frac{3}{4}\right)^1 = 0.140625$$

- *What is the average number of transmissions required to receive the frame with no errors?*
  - Let  $g = p\{\text{error in frame}\}$ ,  $1 - g = p\{\text{no error in frame}\}$ , and  $i$  be a RV that represents the number of transmissions
  - If  $i$  transmissions are required, then we would have  $(i - 1)$  failures followed by 1 success
  - Use the expected RV formula

$$\begin{aligned} e[i] &= \sum_{i=0}^{\infty} i \cdot p\{i \text{ transmissions}\} = \sum_{i=0}^{\infty} i \cdot g^{i-1} \cdot (1 - g) \\ &= \frac{1}{1 - g} = \frac{1}{1 - 0.578125} = 2.37 \end{aligned}$$