

# GB13604 - Maths for Computer Science

## Lecture 8 – Probability, Part I

Claus Aranha

caranha@cs.tsukuba.ac.jp

College of Information Science

2018-11-21

Last updated December 5, 2018

This course is based on Mathematics for Computer Science, Spring 2015, by Albert Meyer and Adam Chlipala, Massachusetts Institute of Technology OpenCourseWare.



## Last Topic: Probability

Probability is absolutely essential to the Engineering, Sciences, and Social studies. And it is also very important to understand in daily life as well.

- Probability as the study of gambling (Lottos, Casinos, Gatchas).
- Probability for Extrapolating information about society (Chances of Death and Sickness, Insurance, Average behavior of populations)
- Probability for the analysis of Noisy Data and Noisy Processes. (Stochastic Algorithms, Experiments, Resilience Engineering)

# Unit Goals and Outline

- Discrete Probability
- Conditional Probability
- Independence and Causality
- Random Variables and Density Functions
- Expectation
- Deviation
- Sampling and Confidence
- Random Walks

# Poker Example: Probability of Two Jacks



What is the probability of getting **exactly two aces** in a poker hand?

This can be seen as a counting problem:

## Poker Example: Probability of Two Jacks



What is the probability of getting **exactly two aces** in a poker hand?

This can be seen as a counting problem:

- All outcomes:  $\binom{52}{5}$  sets of 5 cards

## Poker Example: Probability of Two Jacks



What is the probability of getting **exactly two aces** in a poker hand?

This can be seen as a counting problem:

- All outcomes:  $\binom{52}{5}$  sets of 5 cards
- Desired outcomes:  $\binom{4}{2} \binom{52-4}{5-2}$  two ace hands.

# Poker Example: Probability of Two Jacks



What is the probability of getting **exactly two aces** in a poker hand?

This can be seen as a counting problem:

- All outcomes:  $\binom{52}{5}$  sets of 5 cards
- Desired outcomes:  $\binom{4}{2} \binom{52-4}{5-2}$  two ace hands.
- Probability: Desired outcome/All outcomes =  $\frac{\binom{4}{2} \binom{48}{3}}{\binom{52}{5}}$
- (About 0.04)

# Probability as a counting problem

**Basic Idea:** “What fraction of the time do I get what I want?”

$$\Pr(\text{event}) = \frac{\text{Outcomes we want}}{\text{All possible outcomes}}$$



# Probability as a counting problem: nomenclature

- A set of experimental **outcomes**
- A subset of outcomes is an **event**
- Probability of an event:

$$\Pr(\text{event}) = \frac{\# \text{ outcomes in the event}}{\text{total } \# \text{ of outcomes}}$$

Applies to a lot of cases (but not all of them)

# The Monty Hall Problem

- 1970's American TV show “Let's make a Deal”  
(Hosted by Monty Hall)



- One door has a good prize (a car?) and two doors have a bad prize (a goat?)

# Rules of the Monty Hall Problem

- Goats behind two doors, car behind one door;
- Contestant **chooses a door**;
- Monty **reveals a door** with a goat behind it;
- Contestant choose to **stick or switch** doors;

Mary Savant published a column in a science magazine about the game that sparked a debate on two positions:

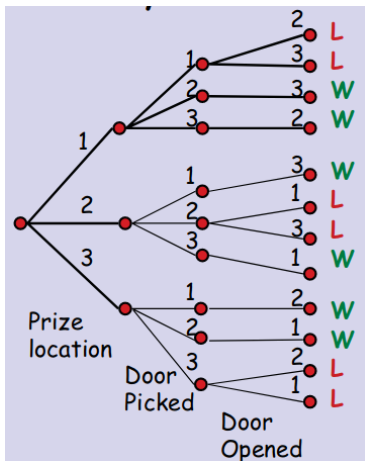
- ① **stick and switch** are equally good
- ② **switch** is much better than **stick**

# Analysing Monty Hall

- What are the outcomes? What is the event?
- We will use a **probability tree** to analyse the game step by step and define these sets.

# Analysis: Switch Strategy

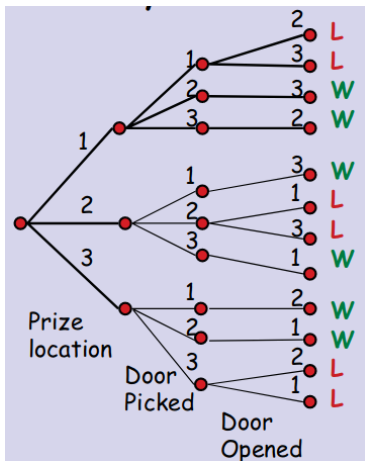
Switch: Pick a door, Reveal Goat, Switch Door



- For each step, we break down the possible outcomes.
- Each branch of the tree is labeled W/L
- **Winning outcomes: 6**
- **Losing outcomes: 6**

# Analysis: Switch Strategy

Switch: Pick a door, Reveal Goat, Switch Door



- For each step, we break down the possible outcomes.
- Each branch of the tree is labeled W/L

- **Winning outcomes: 6**
- **Losing outcomes: 6**

- Since the # of winning outcomes and losing outcomes is the same, **both strategies are the same.**  
(This is a **Bad Conclusion**)

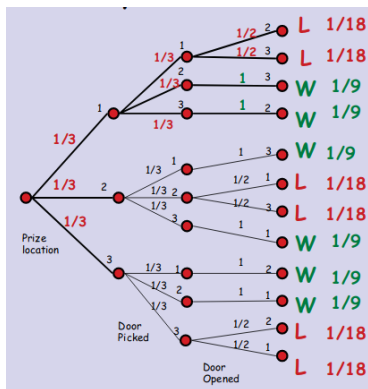
# Analysis: Bad Conclusion

“Since the number of winning outcomes (6) and losing outcomes (6) is the same, then stick strategy and switch strategy are the same”

**Another way:** “After door opening, one goat and one prize are left. The probability of being at the goat door or prize door is the same”

**Problem:** Outcomes do not have the same probability

# Outcomes with variable probability



- We can assign probabilities for each of the events in the probability tree.
- By counting the probabilities of the branches, we can figure out the probability of the leaves.
- Although there are six wins, and six losses, if we compare the total probabilities:
  - Total Win probability:  $6/9$
  - Total Loss probability:  $3/9$



# Conclusion

Switch is better than Stick



## 4-part method for calculating probabilities:

- 1 Identify the outcomes (tree helps)
- 2 Identify the event (winning)
- 3 Assign the outcome probabilities
- 4 Compute the probability of the event (add it up)

Avoid intuition in Probabilities!

# Probability Spaces

# Probability Spaces

- **Sample Spaces:** a **countable** set  $S$  where the elements are outcomes;
- **Probability Function:**  $\Pr : S \rightarrow \{0, 1\}$  so that:

$$\sum_{\omega \in S} \Pr(\omega) = 1$$

- The probability function defines the probability of each outcome in the Sample Space  $S$ .

# Probability Spaces and the Tree Model

The Tree Model presented in the previous example serves to help transforming the problem description into the probability space.

- **Outcomes:** Leaves of the tree
- **Outcome Probabilities:** Calculated from Branch probabilities

# Probability Spaces

- **Sample Spaces:** a **countable** set  $S$  where the elements are outcomes;
- **Probability Function:**  $\Pr : S \rightarrow \{0, 1\}$  so that:

$$\sum_{\omega \in S} \Pr(\omega) = 1$$

- **Event:** A subset  $E \subseteq S$

$$\Pr(E) = \sum_{\omega \in E} \Pr(\omega)$$

- **Corolary:** The sum rule

# The Sum Rule

For **pairwise disjoint** events  $A_0, A_1, \dots$ ,  
 $\Pr(A_0 \cup A_1 \cup \dots) = \Pr(A_0) + \Pr(A_1) + \dots$

$$\Pr(\cup_{i \in \mathbb{N}} A_i) = \sum_{i \in \mathbb{N}} \Pr(A_i)$$

**Discrete** spaces = **Countable** spaces  
Allows the use of **Sums** instead of **Integrals**

# Generalized Probability Rules

- **Difference Rule:**  $\Pr(A - B) = \Pr(A) - \Pr(A \cap B)$

Note how similar these rules are to the **Rules of set size**



# Generalized Probability Rules

- **Difference Rule:**  $\Pr(A - B) = \Pr(A) - \Pr(A \cap B)$
- **Inclusion-Exclusion:**  $\Pr(A \cup B) = \Pr(A) + \Pr(B) - \Pr(A \cap B)$

Note how similar these rules are to the **Rules of set size**

# Generalized Probability Rules

- **Difference Rule:**  $\Pr(A - B) = \Pr(A) - \Pr(A \cap B)$
- **Inclusion-Exclusion:**  $\Pr(A \cup B) = \Pr(A) + \Pr(B) - \Pr(A \cap B)$
- **Union Bound:**  $\Pr(A \cup B) \leq \Pr(A) + \Pr(B)$

Note how similar these rules are to the **Rules of set size**

# Generalized Probability Rules

- **Difference Rule:**  $\Pr(A - B) = \Pr(A) - \Pr(A \cap B)$
- **Inclusion-Exclusion:**  $\Pr(A \cup B) = \Pr(A) + \Pr(B) - \Pr(A \cap B)$
- **Union Bound:**  $\Pr(A \cup B) \leq \Pr(A) + \Pr(B)$
- **Boole's Inequality:**  $\Pr(\cup_{i \in \mathbb{N}} A_i) \leq \sum_{i \in \mathbb{N}} \Pr(A_i)$

Note how similar these rules are to the **Rules of set size**

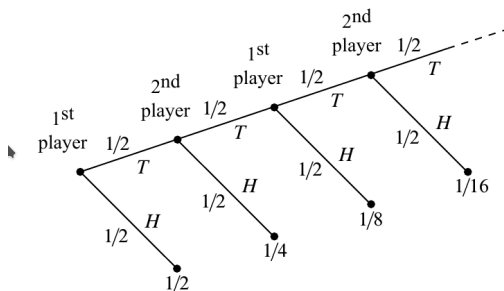
# QUIZ: A Coin Game

**GAME:** Two players flip the same, fair coin.

- What is the probability that the first player wins?
- What is the probability that no player wins?

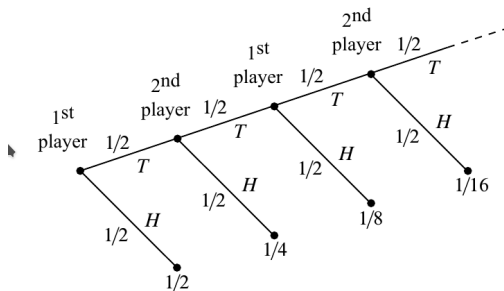
# QUIZ: A Coin Game

**GAME:** Two players flip the same, fair coin.



# QUIZ: A Coin Game

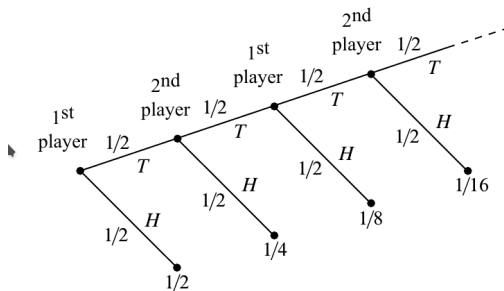
**GAME:** Two players flip the same, fair coin.



- $\Pr(\text{First Player Wins}) = \frac{1}{2} + \frac{1}{8} + \frac{1}{32} + \dots$

# QUIZ: A Coin Game

**GAME:** Two players flip the same, fair coin.



- $\Pr(\text{First Player Wins}) = \frac{1}{2} + \frac{1}{8} + \frac{1}{32} + \dots$
- $\Pr(\text{First Player Wins}) = \frac{1}{2} \sum_{n=0}^{\infty} \left(\frac{1}{4}\right)^n = \frac{1}{2} \cdot \frac{1}{1-1/4} = \frac{2}{3}$

# Strange Dice

- Dice A: {2,6,7}
- Dice B: {1,5,9}
- Dice C: {3,4,8}
- Game 1: You roll one dice, I roll another. Which dice wins?
- Game 2: You roll one dice 2 times and add. I roll one dice 2 times and add. Which dice wins?



# Conditional Probability: Definitions

Probability that one event occurs, given that another event occurred.

- What is the probability that a person will have a health problem, given their health history?
- What is the probability that a stock will rise, given past price?
- What is the probability that a server will overload, given the number of requests?

# Conditional Probability: A fair dice example

Probability of rolling a 1 in a D6.

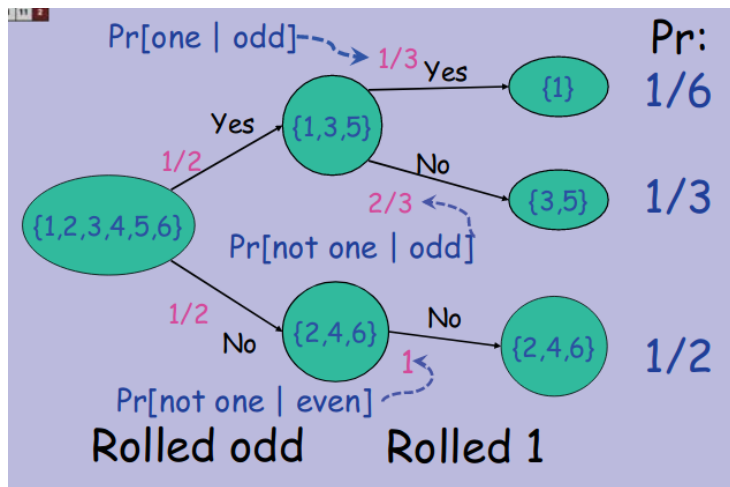
$$\Pr(\text{Roll } 1) = \frac{|\{1\}|}{|\{1, 2, 3, 4, 5, 6\}|} = \frac{1}{6}$$

“Knowledge” changes probabilities

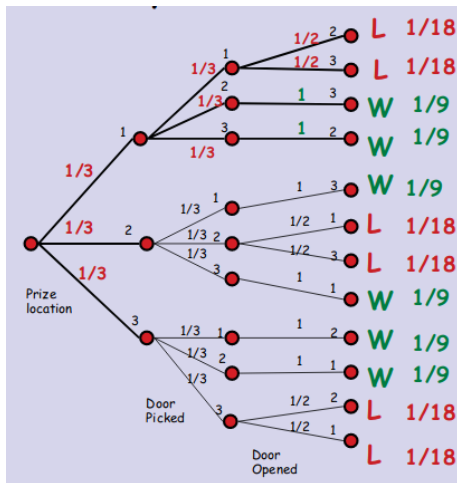
$\Pr(\text{Roll } 1, \text{ knowing that the result was odd})$ :

$$\frac{|\{1\}|}{|\{1, 3, 5\}|} = \frac{1}{3}$$

# Fair dice and the Tree Model



# Conditional Probability and the Tree Model

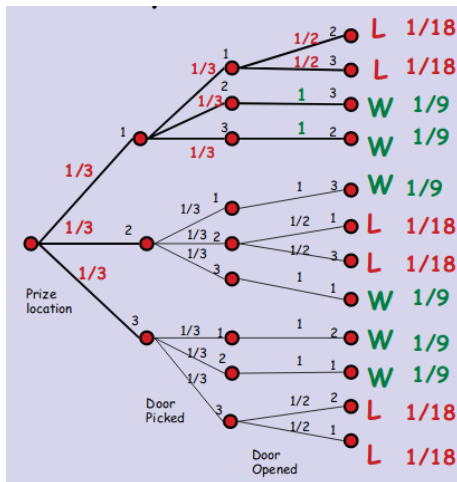


The tree's edges are the conditional probabilities!

- $\Pr(\text{pick 1} | \text{prize 1}) = 1/3$
- $\Pr(\text{pick 2} | \text{prize 3}) = 1/3$
- $\Pr(\text{open 3} | \text{prize 1 \& pick 1}) = 1/2$

We can compose conditional probs!

# Conditional Probability and the Tree Model



The tree's edges are the conditional probabilities!

- $\Pr(\text{pick 1} | \text{prize 1}) = 1/3$
- $\Pr(\text{pick 2} | \text{prize 3}) = 1/3$
- $\Pr(\text{open 3} | \text{prize 1 \& pick 1}) = 1/2$

We can compose conditional probs!

The tree model was using conditional probabilities implicitly!

# Probability: Product Rules

What is the probability that **both** A & B happen?

$$\Pr(A \cap B) = \Pr(A) \cdot \Pr(B|A) \quad (1)$$

This implies a **definition of conditional Prob:**

$$\Pr(B|A) ::= \frac{\Pr(A \cap B)}{\Pr(A)} \quad (2)$$

# Conditional Probability as a Probability Space

Another way to think about conditional probability is that, when we define a probability conditional to  $A$ , we create a new probability space where all events not following  $A$  have a probability of 0.

Let  $\Pr_A$  be the probability function conditional to  $A$ , then:

$$\Pr_A(\omega) = 0, \text{ if } \omega \notin A$$

$$\Pr_A(\omega) = \frac{\Pr(\omega)}{\Pr(A)}, \text{ otherwise.}$$

# Conditional Probability: Quiz

Imagine that I roll **two** dice, and the sum is a **four**. What is the probability that **one of the dice is a 3**?



# Conditional Probability: Quiz

Imagine that I roll **two** dice, and the sum is a **four**. What is the probability that **one of the dice is a 3**?

- Total Events:  $(1,1), (1,2), (1,3), \dots, (6,5), (6,6)$

# Conditional Probability: Quiz

Imagine that I roll **two** dice, and the sum is a **four**. What is the probability that **one of the dice is a 3**?

- Total Events:  $(1,1), (1,2), (1,3), \dots, (6,5), (6,6)$
- Events when sum = 4:  $(1,3), (2,2), (3,1)$

# Conditional Probability: Quiz

Imagine that I roll **two** dice, and the sum is a **four**. What is the probability that **one of the dice is a 3**?

- Total Events:  $(1,1), (1,2), (1,3), \dots, (6,5), (6,6)$
- Events when sum = 4:  $(1,3), (2,2), (3,1)$
- Events when dice = 3 & sum = 4:  $(1,3), (3,1)$

## Conditional Probability: Quiz

Imagine that I roll **two** dice, and the sum is a **four**. What is the probability that **one of the dice is a 3**?

- Total Events:  $(1,1), (1,2), (1,3), \dots, (6,5), (6,6)$
- Events when sum = 4:  $(1,3), (2,2), (3,1)$
- Events when dice = 3 & sum = 4:  $(1,3), (3,1)$
- Probability:  $2/3$

# A 99% accurate TB testing

**Imagine** a great looking test for Tuberculosis:

- If you **have TB**, the test is always **correct**
- If you **don't have TB**, the test is correct **99% of the time**

# A 99% accurate TB testing

**Imagine** a great looking test for Tuberculosis:

- If you **have TB**, the test is always **correct**
- If you **don't have TB**, the test is correct **99% of the time**
- The doctor does this test **and says you have TB!**
- The test is 99% accurate!
- Recently, many antibiotics don't worry anymore!

# A 99% accurate TB testing

**Imagine** a great looking test for Tuberculosis:

- If you **have TB**, the test is always **correct**
- If you **don't have TB**, the test is correct **99% of the time**
- The doctor does this test **and says you have TB!**
- The test is 99% accurate!
- Recently, many antibiotics don't worry anymore!
- Should you worry?

# 99% diagnostic test and conditional probability

What is the probability that you have TB, if the %99 accuracy test says Yes?

$$\Pr(\text{TB} \mid \text{test positive}) = ?$$

What do we know about this test?



# 99% diagnostic test and conditional probability

What is the probability that you have TB, if the %99 accuracy test says Yes?

$$\Pr(\text{TB} \mid \text{test positive}) = ?$$

What do we know about this test?

- $\Pr(\text{Positive} \mid \text{TB}) = 1$
- $\Pr(\text{Positive} \mid \text{Not TB}) = 1/100$

## 99% diagnostic test and conditional probability

What is the probability that you have TB, if the %99 accuracy test says Yes?

$$\Pr(\text{TB} \mid \text{test positive}) = ?$$

What do we know about this test?

- $\Pr(\text{Positive} \mid \text{TB}) = 1$
- $\Pr(\text{Positive} \mid \text{Not TB}) = 1/100$
- False Positive Rate = 1%

# Do you have TB or Not?

- Definition of **Conditional Probability**:
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{Positive and TB})}{\Pr(\text{Positive})}$

# Do you have TB or Not?

- Definition of **Conditional Probability**:
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{Positive and TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{Positive} \mid \text{TB}) \cdot \Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{1 \cdot \Pr(\text{TB})}{\Pr(\text{Positive})} = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})}$

# Do you have TB or Not?

- Definition of **Conditional Probability**:
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{Positive and TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{Positive} \mid \text{TB}) \cdot \Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{1 \cdot \Pr(\text{TB})}{\Pr(\text{Positive})} = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{Positive}) = ?$
- By the **Total Probability formula**:
- $\Pr(\text{Positive}) = \Pr(\text{Positive} \mid \text{TB}) \cdot \Pr(\text{TB}) + \Pr(\text{Positive} \mid \text{Not TB}) \cdot \Pr(\text{Not TB})$

# Do you have TB or Not?

- Definition of **Conditional Probability**:
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{Positive and TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{Positive} \mid \text{TB}) \cdot \Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{1 \cdot \Pr(\text{TB})}{\Pr(\text{Positive})} = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{Positive}) = ?$
- By the **Total Probability formula**:
- $\Pr(\text{Positive}) = \Pr(\text{Positive} \mid \text{TB}) \cdot \Pr(\text{TB}) + \Pr(\text{Positive} \mid \text{Not TB}) \cdot \Pr(\text{Not TB})$
- $\Pr(\text{Positive}) = 1 \cdot \Pr(\text{TB}) + \frac{1}{100} \cdot \Pr(\text{Not TB})$

# Do you have TB or Not?

- Conditional Probability:
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})}$

# Do you have TB or Not?

- Conditional Probability:
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{Positive}) = 1 \cdot \Pr(\text{TB}) + \frac{1}{100} \cdot (1 - \Pr(\text{TB})) = 99/100 \Pr(\text{TB}) + 1/100$



# Do you have TB or Not?

- **Conditional Probability:**
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{Positive}) = 1 \cdot \Pr(\text{TB}) + \frac{1}{100} \cdot (1 - \Pr(\text{TB})) = 99/100 \Pr(\text{TB}) + 1/100$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})} = 100 \cdot \frac{100 \Pr(\text{TB})}{99\Pr(\text{TB})+1}$
- Now we need to know “ $\Pr(\text{TB})$ ”. In the United States, it is about 11.000 per 100 million people, or 0.001.

# Do you have TB or Not?

- **Conditional Probability:**
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})}$
- $\Pr(\text{Positive}) = 1 \cdot \Pr(\text{TB}) + \frac{1}{100} \cdot (1 - \Pr(\text{TB})) = 99/100 \Pr(\text{TB}) + 1/100$
- $\Pr(\text{TB} \mid \text{Positive}) = \frac{\Pr(\text{TB})}{\Pr(\text{Positive})} = 100 \cdot \frac{100 \Pr(\text{TB})}{99\Pr(\text{TB})+1}$
- Now we need to know “ $\Pr(\text{TB})$ ”. In the United States, it is about 11.000 per 100 million people, or 0.001.
- $\Pr(\text{TB} \mid \text{positive}) = \frac{100/10000}{99/10000+1} = 1/100$

# Conditional Probability and Tricky Base Rates

- Even if you have the **positive diagnostic**, your chance of having TB is 1%.
- This is because the **TB Rate (0.01%)** is much lower than the **False Positive Rate (1%)**
- So the 99% accurate test was not so accurate here...
- This is a good example of **Bayes' Rule**

$$\Pr(B|A) = \frac{\Pr(A|B) \cdot \Pr(B)}{\Pr(A)}$$

# Probability in Computer Sciences: The Birthday Paradox

- In a class with 95 people, what is the probability that **two people have a birthday in the same day**?

# Probability in Computer Sciences: The Birthday Paradox

- In a class with 95 people, what is the probability that **two people have a birthday in the same day**?
- $n = 95, d = 365$ .
- Event space is sequences of  $n$  numbers with no repetitions.

# Probability in Computer Sciences: The Birthday Paradox

- In a class with 95 people, what is the probability that **two people have a birthday in the same day**?
- $n = 95, d = 365$ .
- Event space is sequences of  $n$  numbers with no repetitions.
- **The Birthday Principle**: When  $n = \sqrt{2d}$ , the probability of two people having the same birthday is about 0.632
- This same principle is applied to **hashing collisions**.

## Class Summary

- We can see probability problems as counting or combinatorial problems.

$$\text{Probability} = \text{Events of Interest} / \text{Total Events}$$

- We can use the **Tree Model** to make problems easier to understand.
- Conditional Probability, and Bayes' Rule

# Image Credits

- Poker hand photo by “Poker Photos”
- Let’s Make a deal photo by “[www.letsmakeadeal.com](http://www.letsmakeadeal.com)”
- Switch and Stick photo by Wikipedia