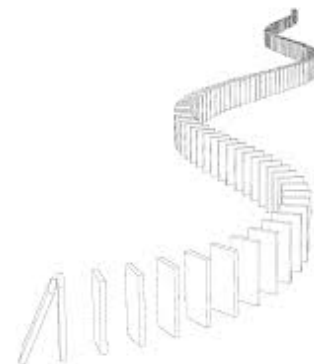




COMP 1002

Logic for Computer Scientists

Lecture 29



Admin stuff

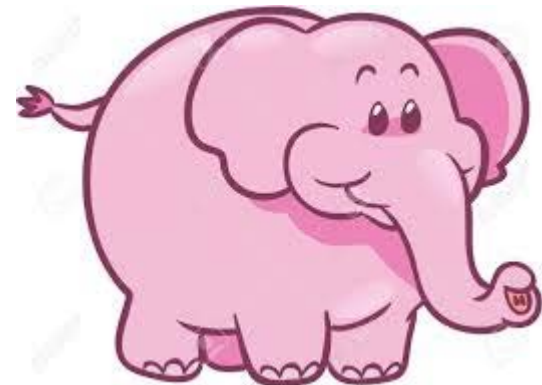
- Assignment 5 is posted.
 - Due April 2nd.
- Next week:
 - “Mini-lab” on Monday, April 3rd.
 - Review for the final exam on Tuesday, April 4th.
 - Practice final exam on Wednesday, April 5th
 - 9am-11am in CS-1019.





Probabilities and pink elephants

- What is the probability that walking down George street you'd see a pink elephant?
 - Your friend says: “It is $\frac{1}{2}$! You will either see the pink elephant, or not!”
 - Do you agree?





Probabilities and distributions

- What if outcomes are not equally likely?
 - Biased coins, etc.
- A function $Pr: S \rightarrow \mathbb{R}$ is a **probability distribution** on (a finite set) S if Pr satisfies the following:
 - For any outcome $s \in S$, $0 \leq Pr(s) \leq 1$
 - $\sum_{\{s \in S\}} Pr(s) = 1$
- **Uniform distribution:** for all $s \in S$, $Pr(s) = 1/|S|$
 - all outcomes are equally likely
 - Fair coin: $Pr(heads) = Pr(tails) = \frac{1}{2}$
- Biased coin: say heads twice as likely as tails.
 - $Pr(heads) + Pr(tails) = 1$. $Pr(heads) = 2 * Pr(tails)$
 - So $Pr(heads) = \frac{2}{3}$, $Pr(tails) = \frac{1}{3}$



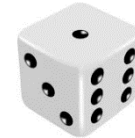
Probabilities of events

- Probability of an event A is a sum of probabilities of the outcomes in A :

- $\Pr(A) = \sum_{\{a \in A\}} \Pr(a)$

- Probability of A not occurring:

- $\Pr(\bar{A}) = 1 - \Pr(A)$



- Probability of the union of two events (either A or B happens) is

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

- By principle of inclusion-exclusion

- If A and B are disjoint, then $\Pr(A \cap B) = 0$, so $\Pr(A \cup B) = \Pr(A) + \Pr(B)$

- In general, if events $A_1 \dots A_n$ are pairwise disjoint

- that is, $\forall i, j$ if $i \neq j$ then $A_i \cap A_j = \emptyset$

- Then $\Pr(\bigcup_{i=1}^n A_i) = \Pr(A_1 \cup A_2 \cup \dots \cup A_n) = \sum_{i=1}^n \Pr(A_i)$

- That is, probability of that any of the events happens is the sum of their individual probabilities.



Probabilities of events

- Suppose a die is biased so that 3 appears twice as often as any other number (others equally likely).
 - Probability of 3: $2/7$. Probabilities of others: $1/7$
- What is the probability that an odd number appears?
 - Event: $A = \{1, 3, 5\}$
 - $\Pr(A) = 1/7 + 2/7 + 1/7 = 4/7$.
- What is a probability that either an odd number or a number divisible by 3 appears?
 - $A = \{1, 3, 5\}$. $B = \{3, 6\}$. $A \cap B = \{3\}$
 - $\Pr(A) = 4/7$. $\Pr(B) = 3/7$. $\Pr(A \cap B) = 2/7$
 - $\Pr(A \cup B) = \Pr(\{1, 3, 5, 6\})$
$$= \frac{4}{7} + \frac{3}{7} - \frac{2}{7} = \frac{1}{7} + \frac{2}{7} + \frac{1}{7} + \frac{1}{7} = \frac{5}{7}$$



Birthday paradox

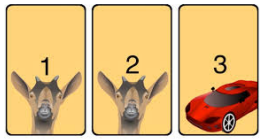
- How many people have to be in the room so that probability that two of them have the same birthday is at least $\frac{1}{2}$?





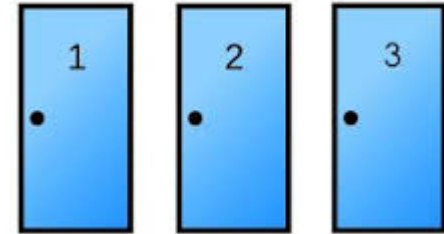
Birthday paradox

- How many people have to be in the room so that probability that two of them have the same birthday is at least $\frac{1}{2}$?
 - Considering all birthdays independent: no twins!
 - And considering all days equally likely
 - Otherwise probability would be higher.
 - Even counting leap years: 366 days.
- Product rule: number of combinations of birthdays of the first i people is $P(i, 366) = 366 * 365 * \dots * (366 - i + 1)$
 - Probability that the first i people all have different birthday is $\frac{P(i, 366)}{366^i} = \frac{365}{366} \frac{364}{366} \dots \frac{(366 - i + 1)}{366}$
 - So with probability $1 - \frac{P(i, 366)}{366^i}$ at least two out of first i people have birthday on the same day.
 - That comes up to about $i = 23$ people to reach $\frac{1}{2}$.

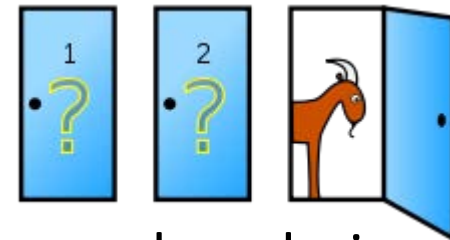


Puzzle: Monty Hall problem

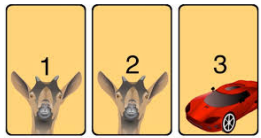
- Let's make a deal!
 - A player picks a door.
 - Behind one door is a car.
 - Behind two others are goats.



- A player chooses a door.
 - A host opens another door
 - Shows a goat behind it.
 - And asks the player if she wants to change her choice.



- Should she switch?
 - Originally, probability of picking the car is $1/3$
 - If she first picked a door with a car: ($1/3$ probability)
 - Then she would switch to a goat.
 - If she first picked a door with a goat ($2/3$ probability)
 - Then she would switch to a car.

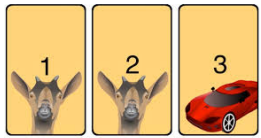


Conditional probabilities

- **Conditional probability** of an event A given event B, denoted $\Pr(A|B)$, is the probability of A if we know that B occurred.
 - Probability of car a behind door 2 if we chose door 1, and door 3 had a goat behind it.
- So it is probability of both A and B, given that we know B happened for sure:

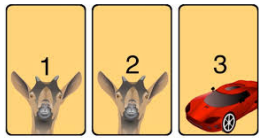
$$\Pr(A|B) = \frac{\Pr(A \cap B)}{\Pr(B)}$$

- Assume that $\Pr(B) > 0$: after all, B did happen.



Independent events

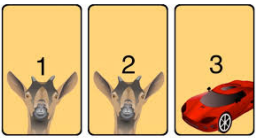
- If knowing B gives us no information about A and vice versa, then A and B are **independent** events:
 - Then $\Pr(A) = \Pr(A|B) = \frac{\Pr(A \cap B)}{\Pr(B)}$.
 - So A and B are independent iff $\Pr(A \cap B) = \Pr(A) \cdot \Pr(B)$.
 - In general, events $A_1 \dots A_n$ can be **pairwise independent** (that is, any two A_i, A_j) are independent, or (stronger condition) **mutually independent**: $\Pr(\cap_{i=1}^n A_i) = \prod_{i=1}^n \Pr(A_i)$
 - For example, different coin tosses/dice rolls are usually considered independent.



Bayes theorem

- Consider a medical test that
 - Has false positive rate of 3% (healthy labeled as sick)
 - Specificity 97%
 - Has false negative rate of 1% (sick labeled as healthy).
 - Sensitivity 99%
- What is the probability that a person has the disease given that the test came positive?
 - Let A: person tested positive, B: person is sick.
 $\Pr(B|A)$?
 - $\Pr(A|B) = 0.99$, $\Pr(\bar{A}|B) = 0.01\dots$
- Not enough information!





Bayes theorem

- **Bayes theorem** allows us to get $\Pr(B|A)$ from $\Pr(A|B)$, if we know probabilities of A and B:

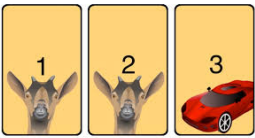
$$\Pr(B|A) = \frac{\Pr(A|B) \cdot \Pr(B)}{\Pr(A)} = \frac{\Pr(A|B) \cdot \Pr(B)}{\Pr(A|B) \Pr(B) + \Pr(A|\bar{B}) \Pr(\bar{B})}$$

- **Proof:**

$$- \Pr(A|B) = \frac{\Pr(A \cap B)}{\Pr(B)}. \quad \Pr(B|A) = \frac{\Pr(A \cap B)}{\Pr(A)}.$$

$$- \Pr(A \cap B) = \Pr(A|B) \Pr(B) = \Pr(B|A) \Pr(A)$$

$$- \text{So } \Pr(B|A) = \Pr(A|B) \Pr(B) / \Pr(A)$$



Bayes theorem

- Consider a medical test that
 - Has false positive rate of 3% (healthy labeled as sick).
 - Has false negative rate of 1% (sick labeled as healthy).
 - Tests for a disease that occurs in 5 in 1000 people.
- What is the probability that a person has the disease given that the test came positive?
 - Let A: person tested positive, B: person is sick. $\Pr(B|A)$?
 - $\Pr(A|B) = 0.99$, $\Pr(\bar{A}|B) = 0.01$. $\Pr(\bar{A}|\bar{B}) = 0.97$, $\Pr(A|\bar{B}) = 0.03$
 - $\Pr(B) = 0.005$.
 - $\Pr(A) = \Pr(A|B)\Pr(B) + \Pr(A|\bar{B})\Pr(\bar{B}) = 0.0348$
 - By Bayes theorem, $\Pr(B|A) = \frac{\Pr(A|B)\Pr(B)}{\Pr(A)} = 0.1422$
- So the probability that a person who tested positive has the disease is just 0.1422, that is, 14.22%.
 - By a similar argument, probability that a person who tested negative does not have a disease is whopping $0.99995 = 99.995\%$.



Hat-check problem



- Suppose n men came to an event, and checked in their hats at the door.
 - On the way out, in a hurry, they each picked up a random hat.
 - On average, how many men picked their own hat?

