

Lecture

## **Foundations of Artificial Intelligence**

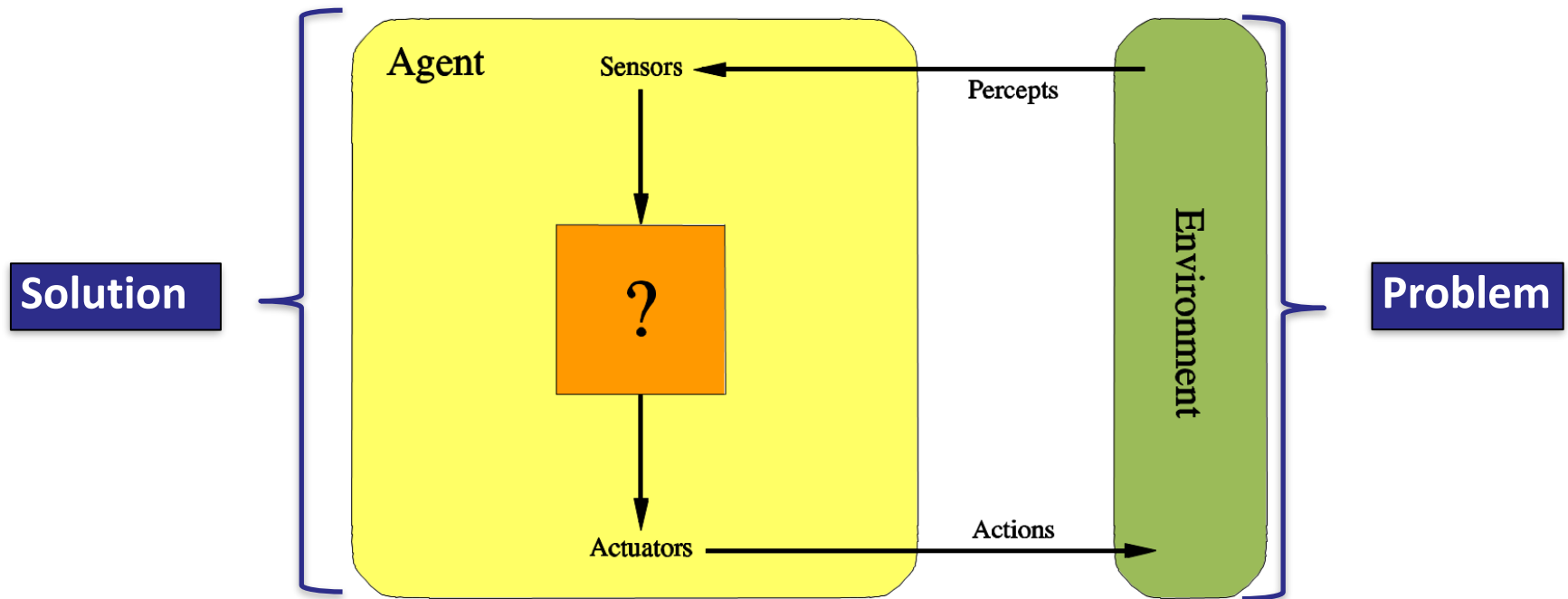
### **Part 7 – Uncertainty**

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# Recall



# Recall ...

- Two major optimization problems
  - **Optimal path finding:** start and goal states are clearly defined
  - **Optimal goal state finding:** start state is given but the goal state is unknown. Instead we **have a set of constraints that should be satisfied**

# Recall ...

## Environment variables

- **Observability:** fully vs partial
- **Certainty:** deterministic vs strategic vs stochastic
- **Temporal succession:** episodic vs sequential
- **Continuity:** static vs dynamic
- **Scale:** discrete vs continuous
- **Population:** single-agent vs multi-agent

# Observability

## Fully observable

- the complete state of the environment can be observed (relevant parts)
- no need to keep track of internal states

## Partially observable

- parts of the environment cannot be observed



# Certainty

## Deterministic

- next environment state is determined only by current state and executed action

## Strategic

- only the opponents' actions cannot be foreseen

## Stochastic

- next environment state is uncertain (e.g., throwing dices).



# Any other open questions?



# Motivation

In previous lectures, we developed agent functions for environments that are **fully observable and deterministic**

- states true or false (maybe unknown) and
- Given state and an action, next state is always predictable

Unfortunately, the real world is not like that

- the whole truth about the world is not accessible
- the next state in many environments is not predictable **certainly**

→ agents must deal with uncertainty in many environments



# Certainty

## Deterministic

- next environment state is determined only by current state and executed action

## Strategic

- only the opponents' actions cannot be foreseen

## Stochastic

- next environment state is **uncertain** (e.g., throwing dices).



# In this lecture you learn about ...

- **Uncertainty**
  - What is it?
  - Why is it happening?
  - How to deal with it?
- **Probability**
  - Prior probability
  - Posterior probability
  - Joint probability
  - Bayes' rule

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**Today**

# Uncertainty

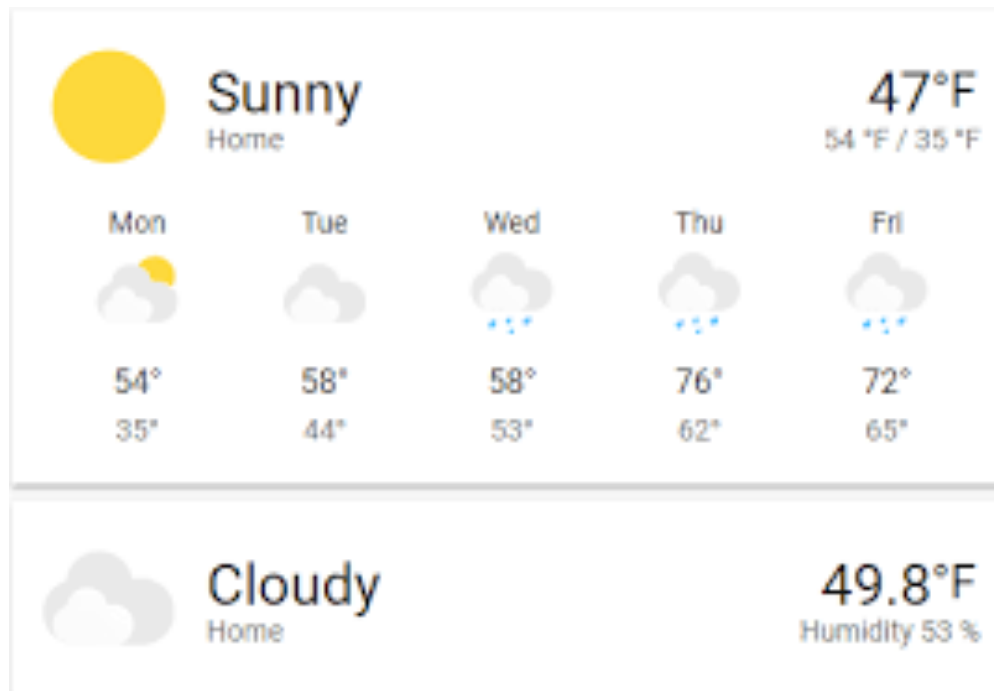
# Uncertainty

- The action in the backgammon game is to throw the dices
- The state of the board depends on the output of the action
- Although we know all possibilities for the output of the action
- We cannot be certain, which possible numbers show up on dices



# Uncertainty

- Weather forecasting might be certain for today
- It's uncertain for tomorrow and next days



# Uncertainty

- Many different actions for getting to the airport:
- Action  $A_t = \text{leave for the airport } t \text{ minutes before departure}$
- typical questions:
  - Will action  $A_t$  get me to the airport in time?
  - Which action is the best choice?

- Refers to situations involving imperfect or unknown **information**
- Happens because of the lack of knowledge
- Uncertainty arises in **partially observable** or **stochastic** environments, as well as due to **ignorance**, **indolence**, or both.

# Problems with uncertainty

Risks involved in plan “ $A_{90}$  will get me to the airport in time”

- **partial observability** (road state, other drivers' plans, etc.)
- **noisy sensors** (traffic reports may be wrong)
- **uncertainty in action outcomes** (flat tire, accident, etc.)
- **immense complexity** of modeling and predicting traffic

## An **ideal rational plan**

$A_{90}$  will get me to the airport in time as long as my car doesn't break down, I don't run out of gas, no accident, the bridge doesn't fall down, ...

- impossible to model **all things that can go wrong**

## A **more cautious plan**

- $A_{1440}$  will get me to the airport in time
- will (almost) certainly succeed, but clearly suboptimal  
→ we have to pay for a night in a hotel



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**Today**

# Dealing with Uncertainty

# Probabilities

Probabilities are one way of handling uncertainty

- e.g.  $A_{90}$  will get me to the airport in time with probability 0.5

The probability summarizes effects that are due to

- **Theoretical ignorance:** our theory might be incomplete
  - e.g., we cannot completely model the weather
- **Practical ignorance:** even if we know all theoretical odds, we might not be aware of the actual current situation.
  - e.g. “Today starts a big festival just next to the airport.”

# Probabilities and beliefs

Probabilities are related to one's beliefs

- A probability  $p$  attached to a statement means that I believe that the statement will be true in  $p * 100\%$  of the cases
  - Example: I assume a  $0.1$  probability of traffic jam on the A40  
→ there might be jam (10% of cases), but usually there is none
- It does not mean that it is true with  $p\%$
- Example: the traffic on the A40 is jammed with a degree of 10%

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  - Example: the traffic on the A40 is jammed with a degree of 10%

Probability theory is about **degree of belief**.

- other techniques (e.g., fuzzy logic) deal with **degree of truth**

# Probability and knowledge

Probability of an event depends on how much knowledge about other events

$$P(A_{25} \text{ at airport in time} \mid \text{no reported accidents}) = 0.06$$

- in 6% of the days I get to the airport in 25 minutes if no accidents are reported

$$P(A_{25} \text{ at airport in time} \mid \text{no reported accidents, 5 a.m.}) = 0.15$$

- chances are higher at 5 in the morning

# Making decisions under uncertainty

Suppose I believe the following

- $P(A_{25} \text{ at airport on time} \mid \text{current knowledge}) = 0.04$
- $P(A_{90} \text{ at airport on time} \mid \text{current knowledge}) = 0.70$
- $P(A_{120} \text{ at airport on time} \mid \text{current knowledge}) = 0.95$
- $P(A_{1440} \text{ at airport on time} \mid \text{current knowledge}) = 0.9999$

Which action should I choose?

We make a decision based on our **preferences**.

- How bad would it be to miss the flight?
- How bad would it be to wait for an hour at the airport?

Utility theory is used to represent and infer preferences.

**Decision theory = probability theory + utility theory**

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**Today**

# Probability

# Terminology

- **Experiment (or trial):** is any procedure that produces a definite outcome that cannot be predicted with certainty. An experiment is said to be *random* if it has more than **one possible** outcome,
- **Outcome: a possible** result of an experiment
- **Sample space (S):** the set of all possible outcomes of that experiment.
- **Event:** a property that we are studying in an experiment
- **Event space:** A subset from the sample space with a common property.
- **Probability Theory:** We are interested in estimating the probability of occurring an event = observing an outcome that belongs to the event space



# Flipping a coin

- **Experiment (or trial):** flipping a coin (<https://g.co/kgs/dmoetH>)
- **Outcome:** one possible output is “H”
- **Sample space:**  $S = \{\text{“H”}, \text{“T”}\}$
- **Event (A):** How likely is to see a head?
- **Event space:**  $A = \{\text{“H”}\}$
- **Event (B):** How likely is to see a tail?
- **Event space:**  $B = \{\text{“T”}\}$

# Rolling a dice

- **Experiment (or trail):** rolling a dice (<https://www.calculator.net/dice-roller.html>)
- **Outcome:** one possible output is “1”
- **Sample space:**  $S = \{“1”, “2”, “3”, “4”, “5”, “6”\}$
- **Event A:** How likely is it to see an even number?
- **Event space:**  $A = \{“2”, “4”, “6”\}$
- **Event B:** How likely is it to see a number greater than 3?
- **Event space:**  $B = \{“4”, “5”, “6”\}$

# Identifying gender and job of a person

- **Experiment (or trial):** randomly select a person and ask about her job and/or gender
- **Outcome:** one possible output is “Female/Teacher”
- **Sample space for job:**  $S1 = \{\text{Teacher, Healer}\}$
- **Sample Space for gender:**  $S2 = \{\text{Male, Female}\}$
- **Event A:** How likely is it to see a female worker?
- Event space  $A1 = \{\text{Female}\}$
- **Event B:** How likely is it to see a teacher?
- Event space  $B = \{\text{Teacher}\}$

# Probability of an event

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Given a sample space for an experiment, we are interested in the probability of a specific event

There are different ways to estimate the probability of an event

- **Count-based**
- Neural models
- ...

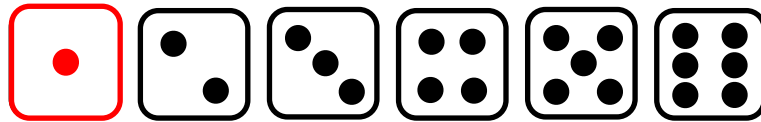
# Prior probability

**Prior probability:** probability of an event in a sample space without any additional knowledge

$$P(A) = \frac{|\text{Event Space}|}{|\text{Sample Space}|}$$

# Example

- Probability of rolling “1” with a single 6-sided fair dice?



$$P\left(\frac{\text{“good“ events}}{\text{all events}}\right) = \frac{1}{6}$$

# Prior probability

**Prior probability:** probability of an event without any additional knowledge

- $p(\text{female}) = 18.3 \text{ million} / 44 \text{ million} = 0.42$
- $p(\text{teacher}) = 673225 / 44 \text{ million} = 0.015$
- $p(\text{healer}) = 45000 / 44 \text{ million} = 0.001$

	total	male	female
working population (our sample)	44 million	25.7 million	18.3 million
teacher	673 225		
holistic healer	45000*		

# Boy-or-girl problem

**Experiment:** Mr. Jones has two children.

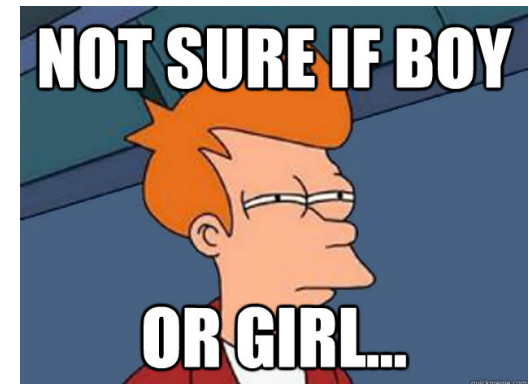
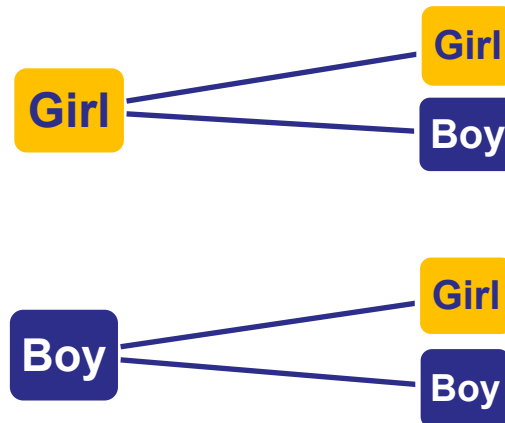
**Outcome:** {B, G}

**Sample Space:** {BG, BB, GB, GG}

**Event Space:** {GG}

$$p = \frac{\text{event space}}{\text{sample space}} = \frac{1}{4}$$

Older Child      Younger Child





# Boy-or-girl problem

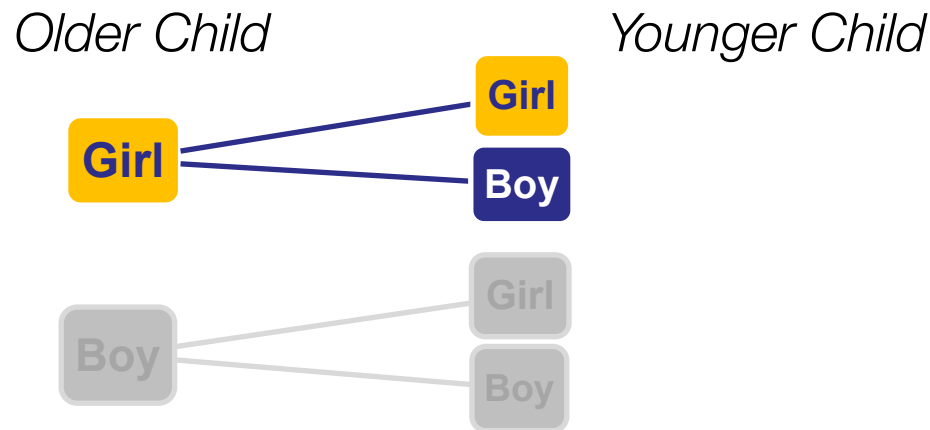
**Experiment:** Mr. Jones has two children. *The older child is a girl.*

**Event:** What is the probability that both children are girls?

**Sample Space:** {GB, GG}

**True Events:** {GG}

$$p = \frac{\text{event space}}{\text{sample space}} = \frac{1}{2}$$



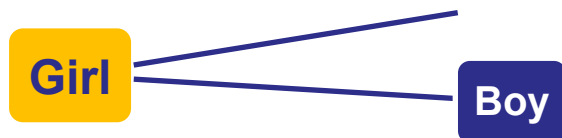
# Boy-or-girl problem

**Experiment:** Mr. Jones has two children. At least one of them is a girl.

**Event:** What is the probability that both children are girls?

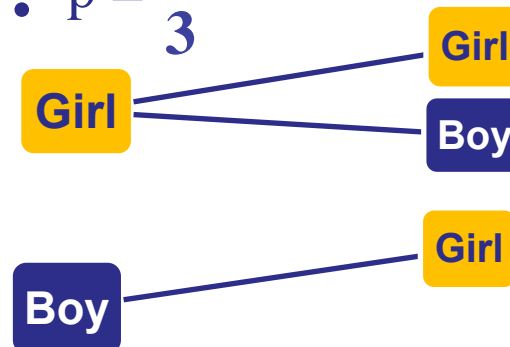
First way of reasoning

- Sample Space: {GB, GG}
- Event Space: {GG}
- $p = \frac{1}{2}$



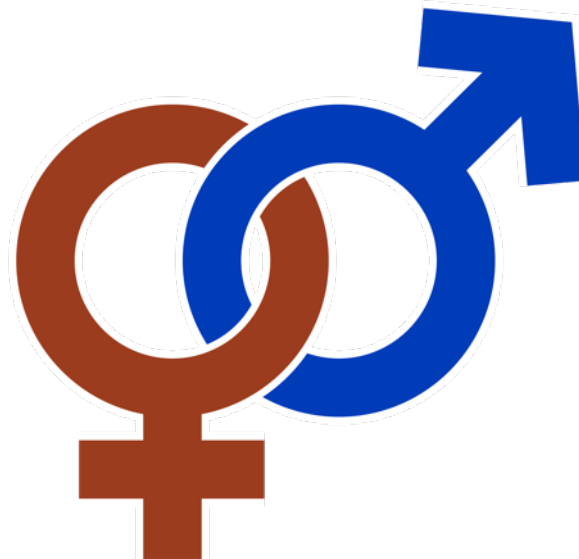
Second way of reasoning

- Sample Space: {GB, GG, BG}
- Event Space: {GG}
- $p = \frac{1}{3}$



# Paradox

- If two ways of reasoning lead to different answers for the same question, it is often called a paradox.
- It has been shown that the answers for the boy-or-girl paradox vary when the wording of the question is modified slightly.



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**Today**

# Posterior Probability

# Posterior probability

How likely is an event after incorporating knowledge that  $B$  is true.

- $P(\text{teacher} \mid \text{femaleWorker})$ : how likely is it that Sarah is a teacher?
- $P(\text{tumor} \mid \text{headache})$ : I have a headache. Do I have a tumor?
- $P(\text{pregnant} \mid \text{noPeriod})$ : A girl's period is three days overdue. Is she pregnant?
- $P(\text{passExam} \mid \text{noSubmissions})$ : I haven't done any exercises. Will I pass?

# Posterior probability

**Conditional probability: probability of an event A given that another event B has occurred.**

**$P(A | B)$**

- *“the conditional probability of A given B”*
- *“the probability of A under the condition B”*

**$P(\text{teacher} | \text{femaleWorker})$ :** if we take a female worker from our sample, how likely is it that she is a teacher?

**$P(\text{teacher}, \text{femaleWorker})$ :** if we take a worker from our sample, how likely is it that the worker is female and a teacher.

# Posterior probability

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

$$P(\text{femaleworker} | \text{teacher}) = \frac{P(\text{femaleworker}, \text{teacher})}{P(\text{teacher})}$$

$$P(\text{femaleworker} | \text{teacher}) = \frac{\frac{491454}{44\text{million}} = 0.011}{0.015} = 0.73$$

	total	male	female
working population (our sample)	44 million	25.7 million	18.3 million
teacher	673 225	181 771	491 454
holistic healer	45000*	N/A	N/A

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**Today**

# Joint Probability



# Joint probability

**Joint probability:** probability of two events occurring together

- $P(\text{teacher}, \text{femaleworker})$

	total	male	female
working population (our sample)	44 million	25.7 million	18.3 million
teacher	673 225		
holistic healer	45000*		

# Joint probability

- Multiplication rule for joint probabilities of dependent variables:

$$P(A, B) = P(A) * P(B|A)$$

$$\begin{aligned} P(\text{femaleWorker}, \text{teacher}) &= P(\text{femaleWorker}) * P(\text{teacher}|\text{femaleWorker}) \\ &= 0.42 * \frac{491454}{18.3 \text{ million}} = 0.011 \end{aligned}$$

	total	male	female
working population (our sample)	44 million	25.7 million	18.3 million
teacher	673 225	181 771	491 454
holistic healer	45000*	N/A	N/A

# Joint probability for independent events

$$P(A, B) = P(A) \times P(B)$$

$$P(\text{teacher, femaleWorker}) = P(\text{teacher}) * P(\text{femaleWorker})$$

- $P(\text{teacher, femaleWorker}) = 0.015 * 0.42 = 0.0063$

→ this would lead to an estimate of 277 200 female teachers

**Not independent!**

	total	male	female
working population (our sample)	44 million	25.7 million	18.3 million
teacher	673 225	181 771	491 454
holistic healer	45000*	N/A	N/A

# Bayes' rule

- Estimating the probability of an event, based on prior knowledge of conditions that might be related to the event

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

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**Today**

# Bayes' Rule

# Conditional probability

$$\bullet \quad P(\text{teacher} \mid \text{femaleWorker}) = \frac{P(\text{femaleWorker} \mid \text{teacher}) * P(\text{teacher})}{P(\text{femaleWorker})}$$

$$\bullet \quad P(\text{teacher} \mid \text{femaleWorker}) = \frac{\frac{491454}{673225} * 0.015}{0.42} = 0.026$$

$$P(A \mid B) = \frac{P(B \mid A) * P(A)}{P(B)}$$

	total	male	female
working population (our sample)	44 million	25.7 million	18.3 million
teacher	673 225	181 771	491 454
holistic healer	45000*	N/A	N/A

# Back to girls and boys

Sample Space: {BG, BB, GB, GG}

- $P(GG) = \frac{1}{4} = 0.25$

- $P(BB) = \frac{1}{4} = 0.25$

- $P(B) = \frac{3}{4} = 0.75$

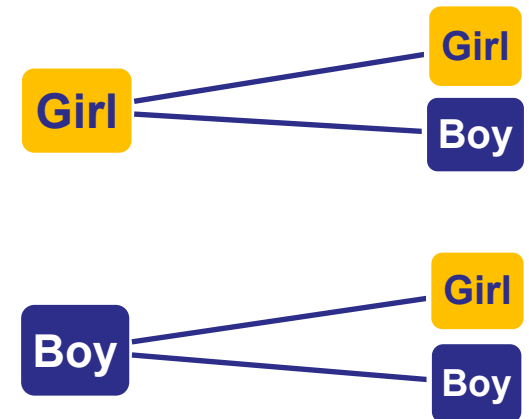
- $P(G) = \frac{3}{4} = 0.75$

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

- $P(GG|G) = \frac{P(G|GG) * P(GG)}{P(G)}$

- $P(GG|G) = \frac{1 * 0.25}{0.75} = \frac{1}{3}$

Given one child is a girl, how likely is it that the other child is a girl, too?



**Knowledge  
reduces  
uncertainty**



# Monty Hall Problem

Scenario: You're on a game show, and you are supposed to choose one of three doors: Behind one door is a car; behind the other two are goats. You pick a door, say number 1, and the host, who knows what's behind each door, opens another door, say number 2, which reveals a goat.

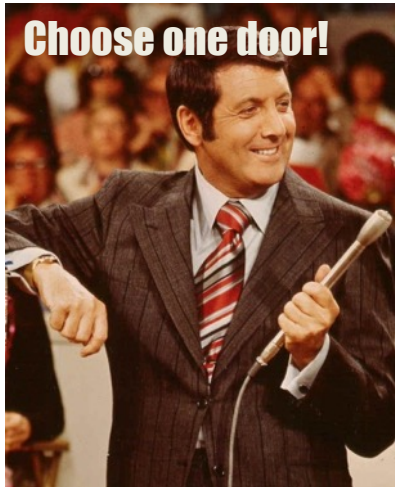
He then asks you: *"Do you want to switch your choice to door No. 3?"*  
What do you do??



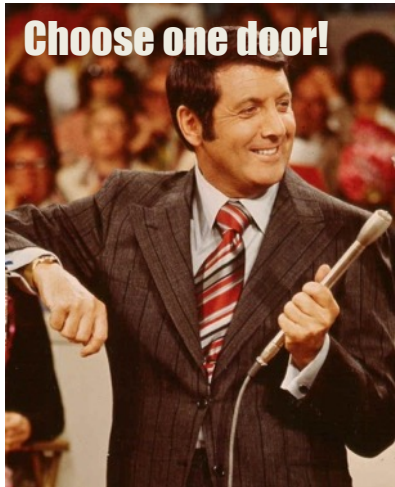
# Monty Hall Problem



# Monty Hall Problem



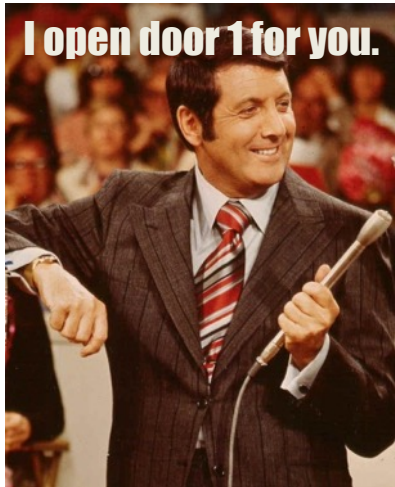
# Monty Hall Problem



$$P(\text{car}) = \frac{1}{3} \quad P(\text{car}) = \frac{1}{3} \quad P(\text{car}) = \frac{1}{3}$$



# Monty Hall Problem



# What would you do?



stay? or switch?



# Monty Hall Solution

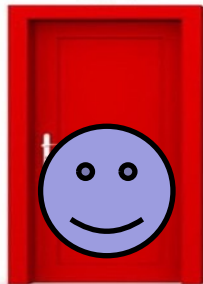
Switching is always better than sticking to your primary selected door.

Why?

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# Monty Hall Solution

$$P(\text{car}) = \frac{1}{3}$$



$$P(\text{car}) = \frac{2}{3}$$

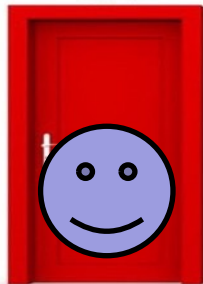




# Monty Hall Solution

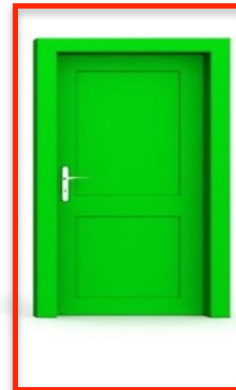
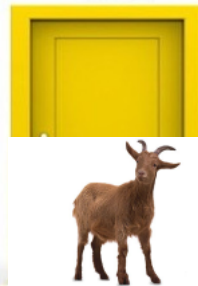
$$P(\text{car}) = \frac{1}{3}$$

$$P(\text{goat}) = \frac{2}{3}$$



# Monty Hall Solution

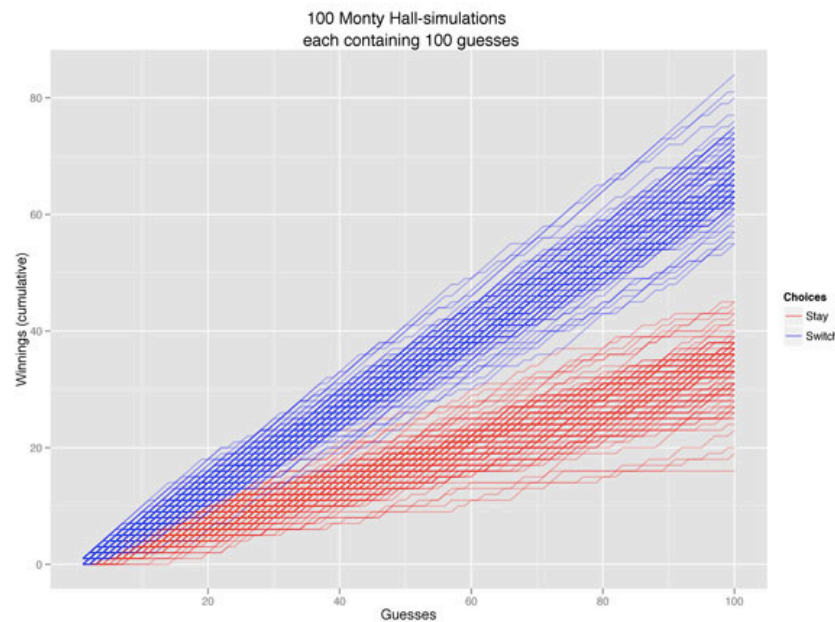
$$P(\text{car}) = \frac{1}{3} \quad P(\text{car}) = \frac{0}{3} \quad P(\text{car}) = \frac{2}{3}$$



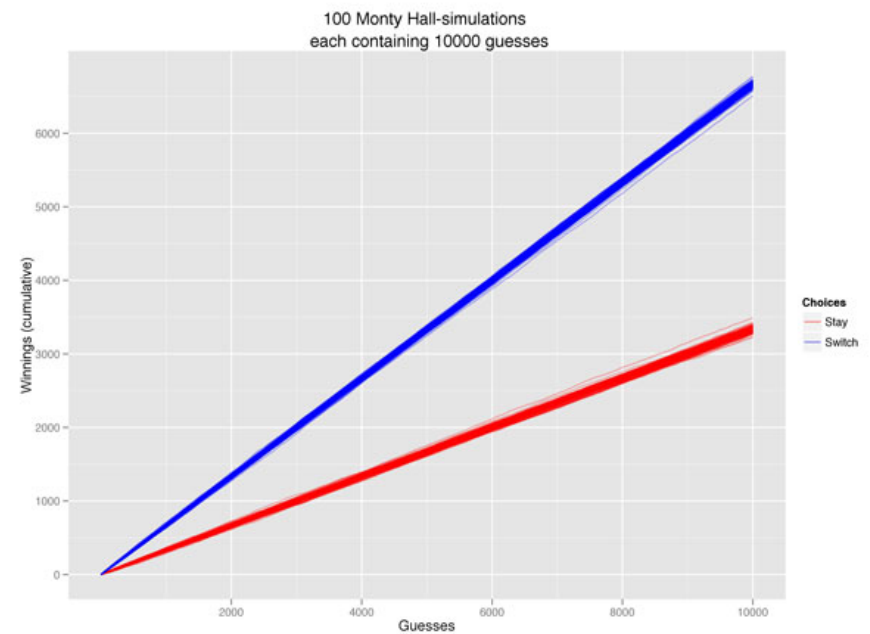
# Monty Hall Solution

[Simulation](#)

# Simulation



<http://rpsychologist.com/monty-hall-simulation>



# Summary

- **Uncertainty**
  - What is it?
  - Why is it happening?
  - How to deal with it?
- **Probability**
  - Prior probability
  - Posterior probability
  - Joint probability
  - Bayes' rule

# Readings

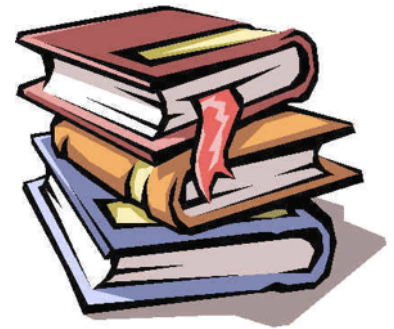
- Christopher D. Manning & Hinrich Schütze (1999): *Foundations of Statistical Natural Language Processing*

## Mandatory

- Chapter 2: *Mathematical Foundations*,
  - 2.1 *Elementary Probability Theory*, p.40-50.

## Optional

- Rest of chapter 2.



# Next lecture

- How do we use probabilities to let an agent act in an uncertain environment?  
→ Machine learning basics



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**Today**

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**Thank You**