

# IS 7033: Artificial Intelligence and Machine Learning

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# Agenda

- Course Logistics
- Course Overview
- What is AI?

# Purpose of this class

- To teach you the basics of artificial intelligence
- To excite you about **AI (beyond Deep Learning)**

# Course Objectives

- At the end of this course, you should
  - Be able to tackle real-world tasks with the appropriate techniques
  - Be more proficient at math and programming

# Prerequisite

- Programming
- Discrete math, mathematical rigor
- Probability

# Coursework

- Homework (60%)
  - Introduction
  - Graph
  - Probabilistic Graph Models
  - Bayesian networks
  - Markov Hidden Models
  - Reinforcement
  - Logic
- Project (40%)

Milestones: proposal, progress report, poster session, and final report

# The Honor Code

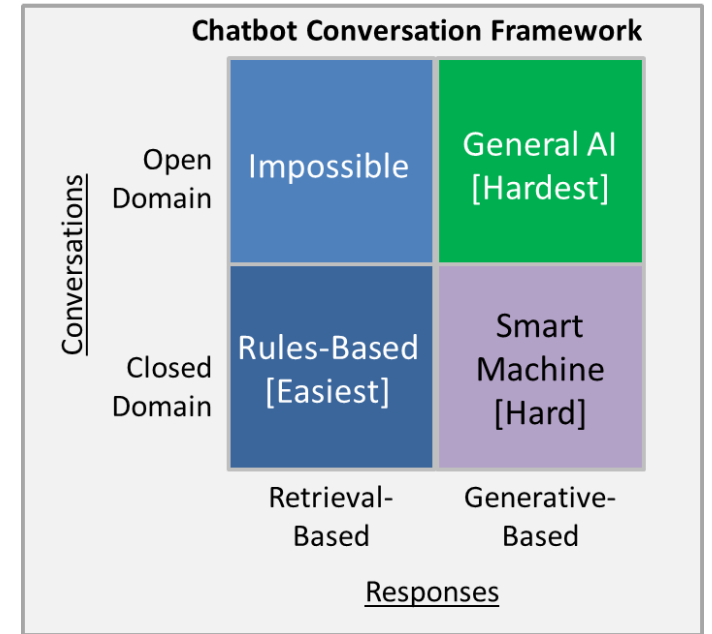
- Do collaborate and discuss together, but write up and code independently
- Do not look at anyone else's writeup or code
- Do not show anyone else your writeup or code or post it online (e.g. GitHub)

# Project – Chatbot

- Open domain – user can ask anything. There isn't necessarily have a well-defined goal or intention.

The infinite number of topics and the fact that a certain amount of world knowledge is required to create reasonable responses makes this a hard problem. The chatbot [mitsuku](#) is the example for this.

- Closed domain – you are solving a particular business problem (ex: pizza bot, Banking, Medical bot, ...)





# Projects – Computer Vision

- What is AI?
- Course overview?
- Course logistics?
- Optimization

AI

## Number of AI papers on Scopus by subcategory (1998–2017)

Source: Elsevier

Growth of annual  
Source: ScopusGrowth in papers (relative to 1996)  
9x  
7x  
5x  
3x  
1x

Number of papers

60,000

40,000

20,000

0

2000

2005

2010

2015

Machine Learning and  
Probabilistic Reasoning

Neural Networks

Computer Vision

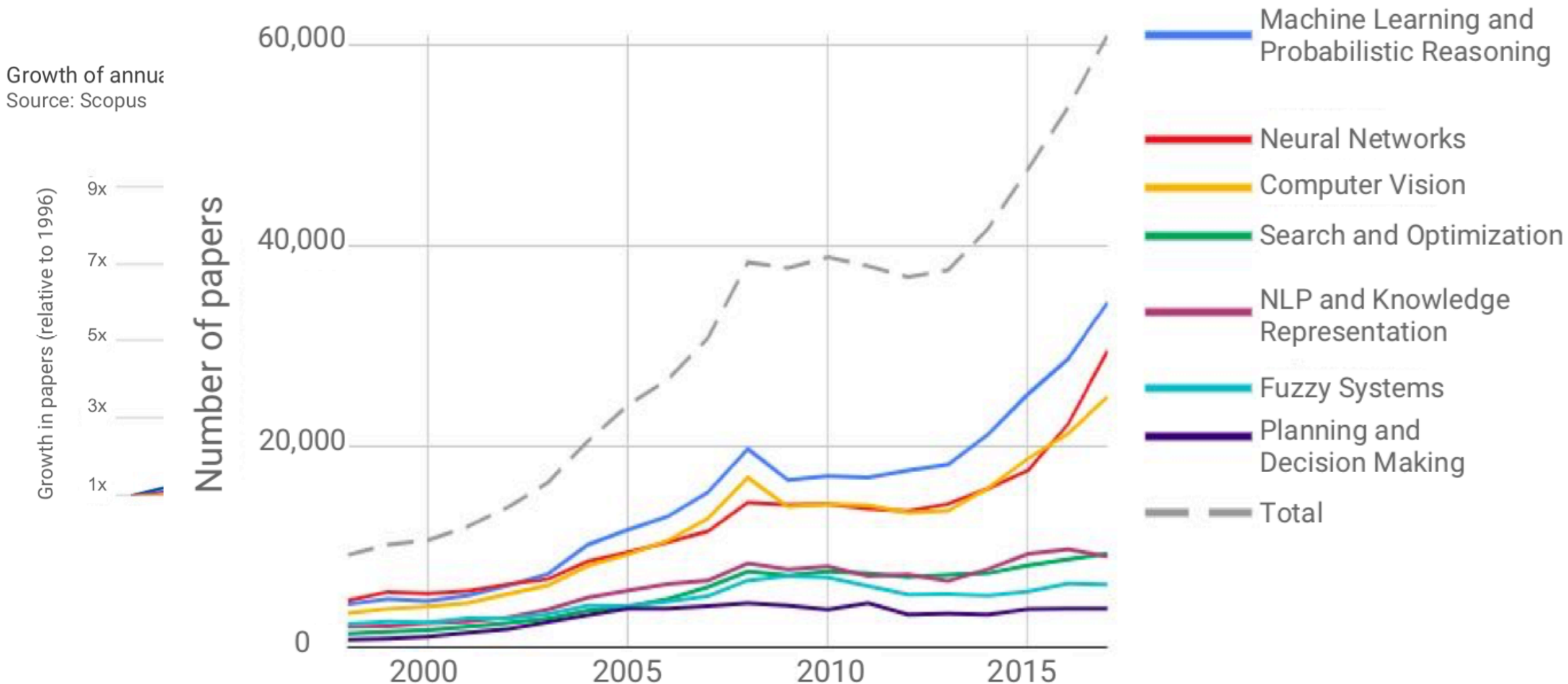
Search and Optimization

NLP and Knowledge  
Representation

Fuzzy Systems

Planning and  
Decision Making

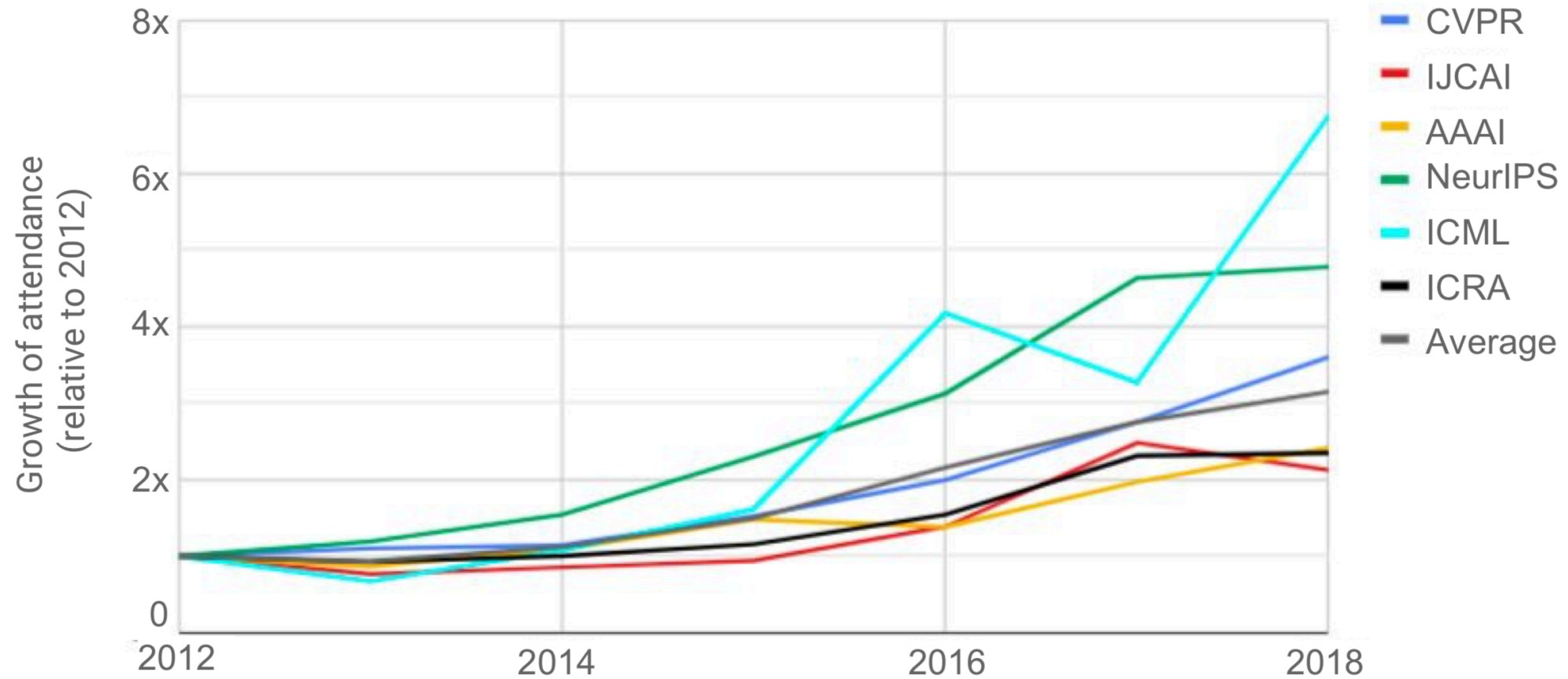
Total



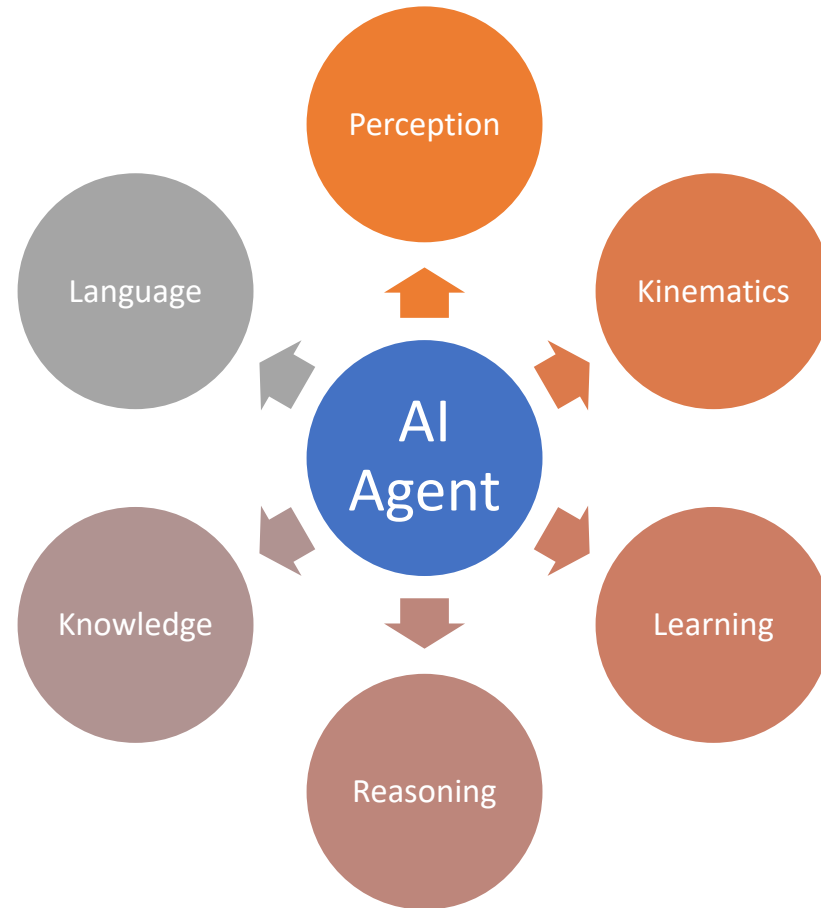
# AI index

Growth of large conference attendance (2012–2018)

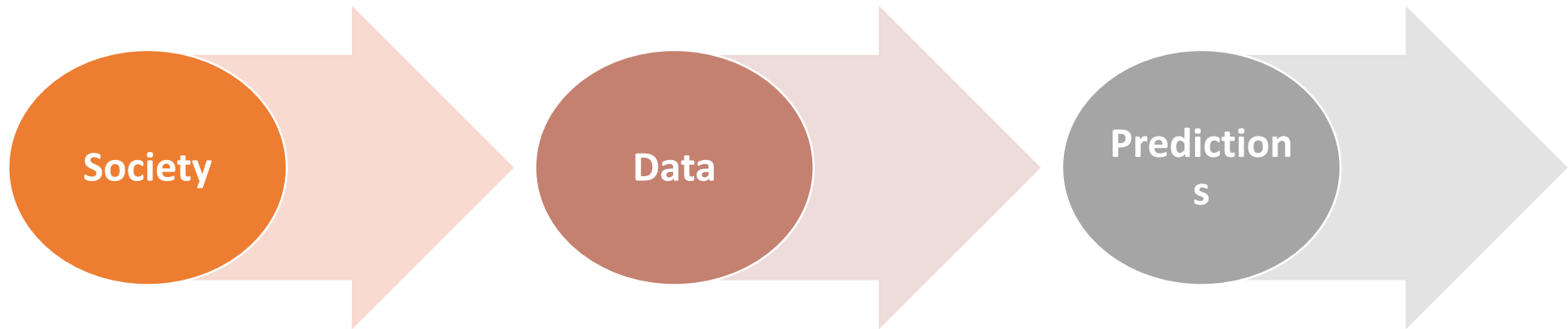
Source: Conference provided data



# An intelligent agent



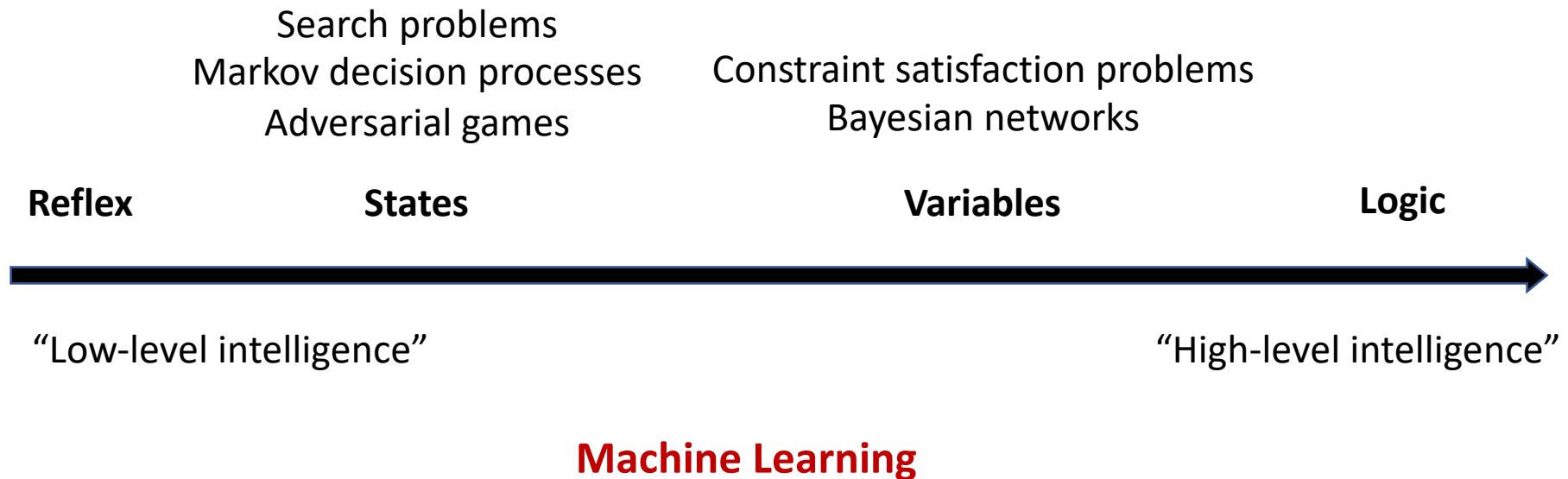
- AI Security
- Bias in machine learning
- Fairness



# Two views of AI

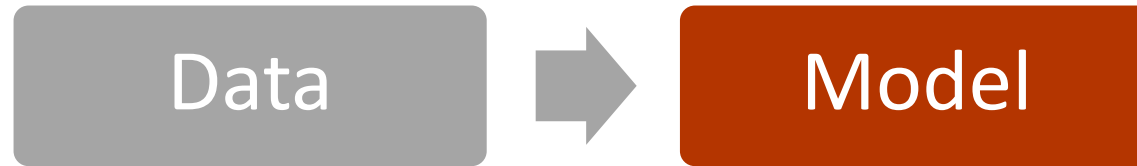
- **AI Agents**: how can we re-create intelligence? Building agents with human-level intelligence.
- **AI Tools**: how can we benefit society? Developing tools that can benefit society.

# Course Plan





# Data Driven Learning



- The main driver of recent successes in artificial intelligence (AI)
- Move from “code” to data to manage the information complexity
- Requires a leap of faith: generalization

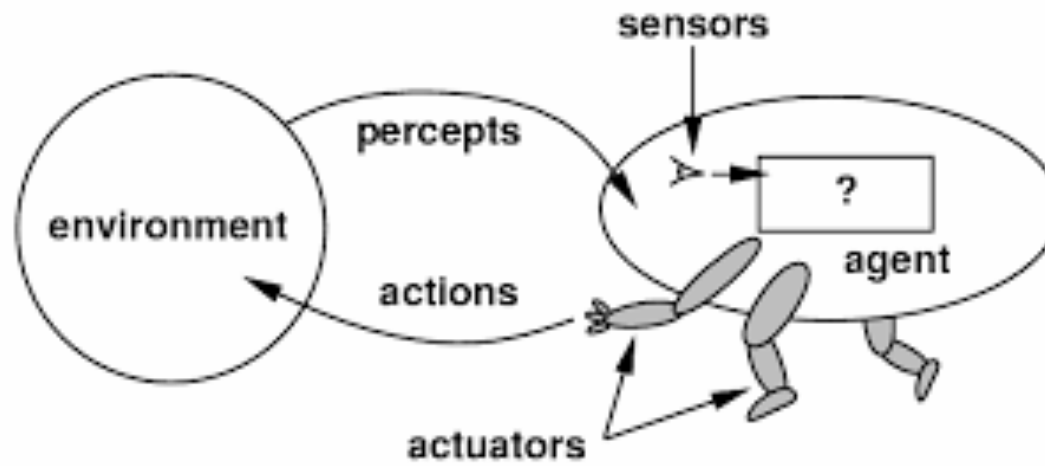
# State-based models

- Search problems
- Markov decision processes
- Adversarial games

# Agent and environments

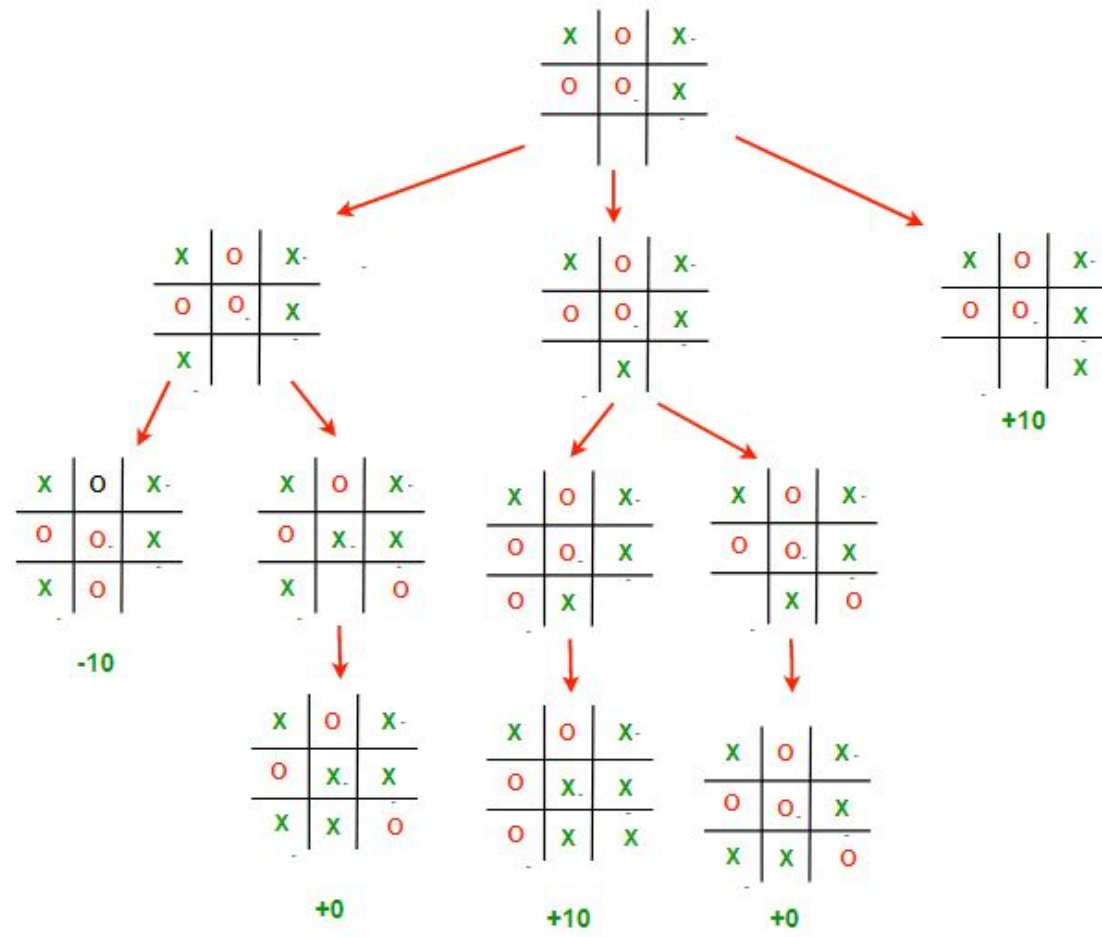
The agent function maps percept sequence to actions

$$f: P \rightarrow A$$

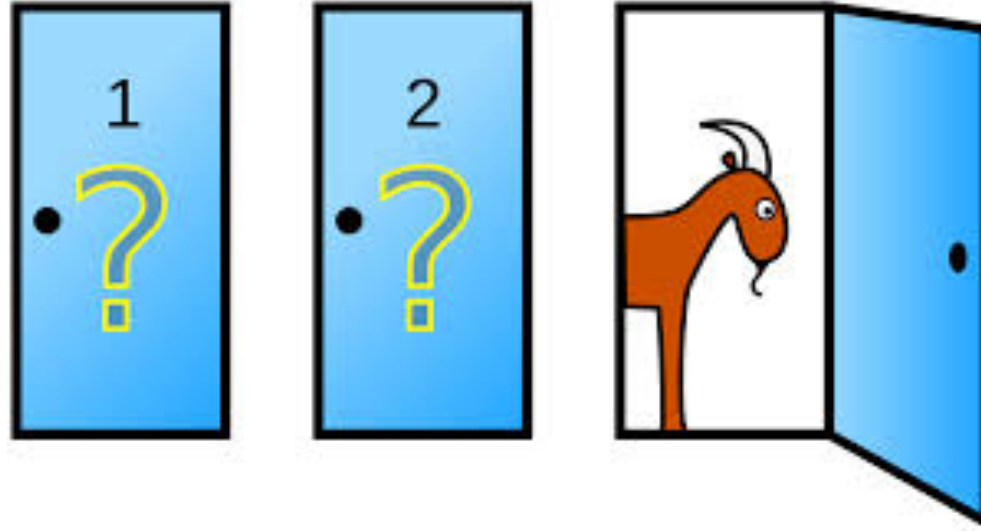


**Perception Action Cycle**

# Tic Tac Toe

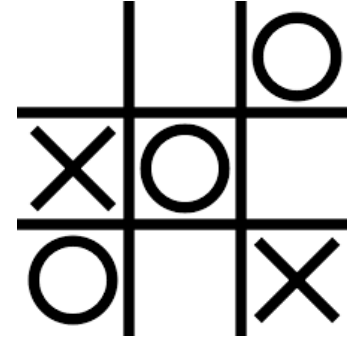


# Monty Hall Gameshow

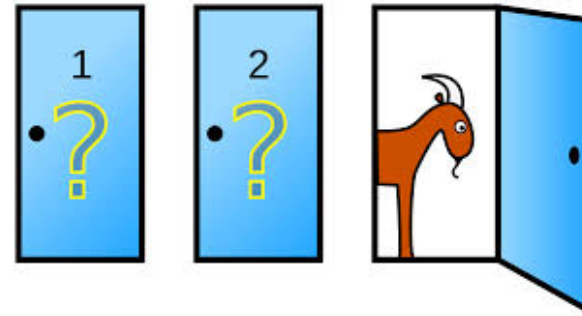


# Environment States

- Fully Observable
- Partially Observable
- Deterministic
- Stochastic
- Discrete
- Continuous
- Benign
- Adversarial



Monty Hall Gameshow



# Example

Is playing a game of Poker any of the following?

- Partially observable
- Stochastic
- Discrete
- Adversarial

Is the task of recognizing handwritten text any of the following?

- fully observable
- Stochastic
- Continuous
- Benign

Driving on the Road

- Partially observable
- Stochastic
- Continuous
- Benign

Playing Chess

- Fully observable
- Deterministic
- Discrete
- Adversarial

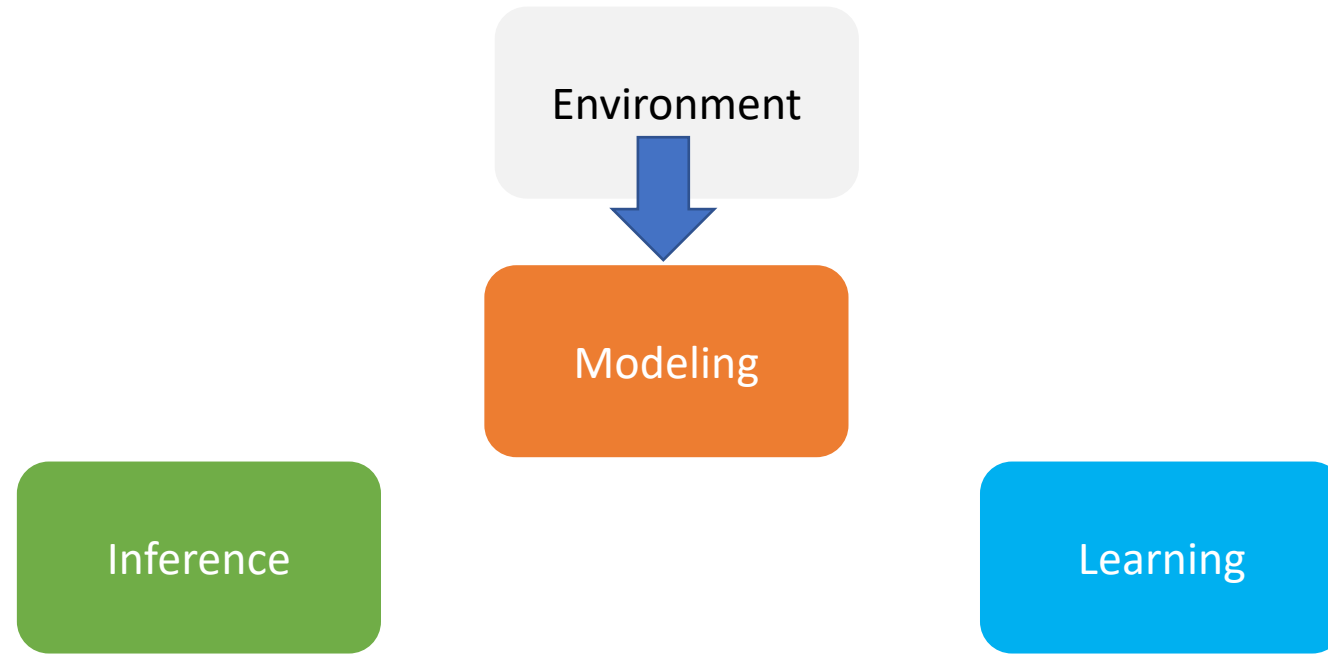
# Definition of Intelligence

- Rational Behavior – an intelligent agent it takes action to **maximize** its expected utility given a desired goal
- Consider constraints (such as computation, time, ...)
- Rational Behavior and Bounded Optimality



# Paradigm

## Graphical Models (Inference and Learning)



Graphical models (or probabilistic graphical models) provide a powerful paradigm to jointly exploit probability theory and graph theory for solving complex real-world problems.

# Optimization

- Discrete optimization

Algorithmic tool: dynamic programming

**min Distance (p)**     $p = \text{a discrete object}$   
 $p \in \text{Paths}$

- Continuous optimization

Algorithmic tool: gradient decent

**min Loss (w)**     $w = \text{a vector of real numbers}$   
 $w \in \mathbb{R}^n$

# Problem: computing edit distance

- Input: two strings, s and t
- Output: minimum number of character insertions, deletions, and substitutions it takes to change s into t.
- Examples:

s	t	distance
Cat	Cat	0
cat	dog	3
cat	at	1
cat	cats	1
a cat!	the cats!	4