

# Representational Dimensions

Computer Science cpsc322, Lecture 2

*(Textbook Chpt1)*

May, 16, 2017



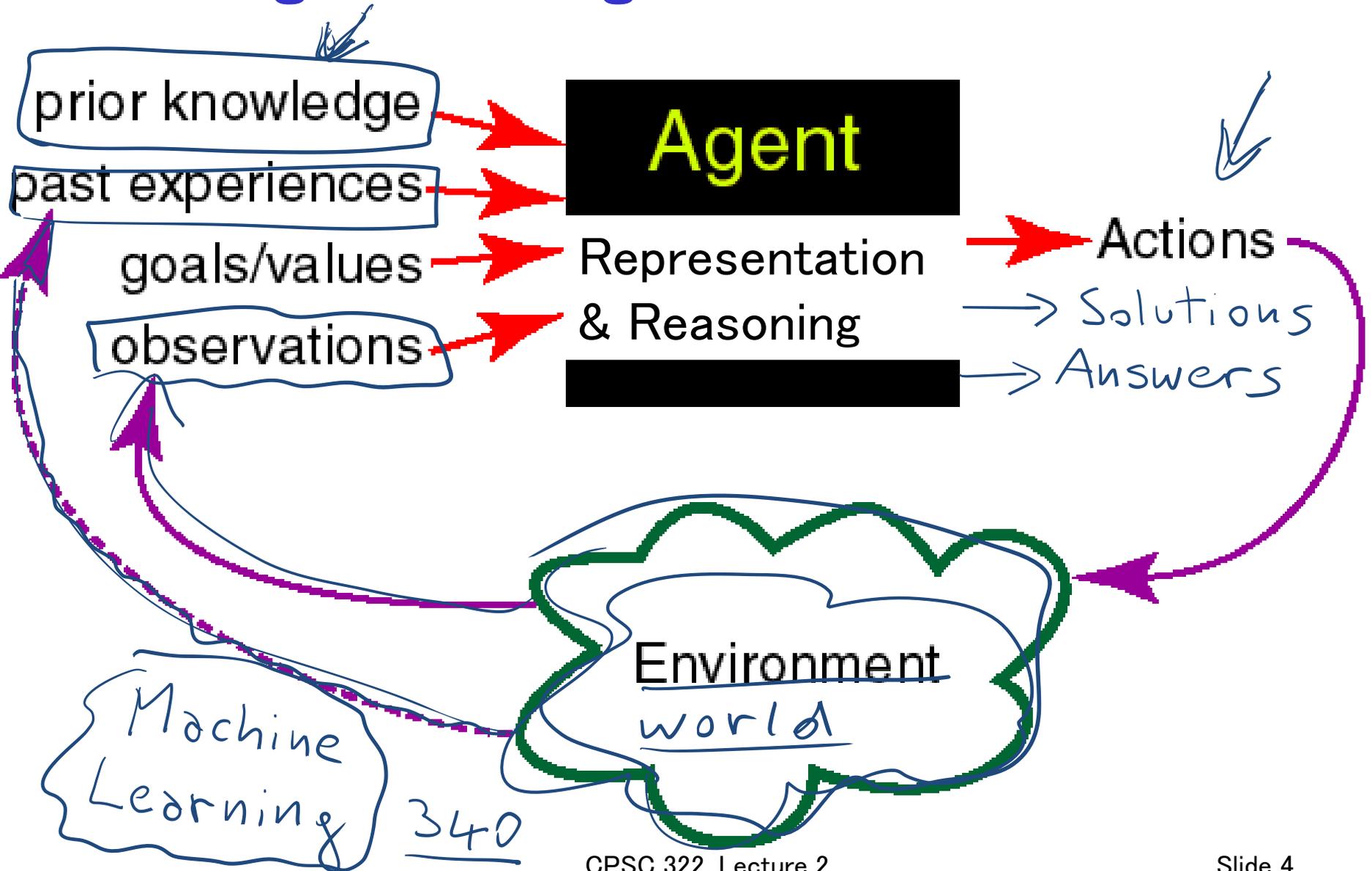
# Lecture Overview

- **Recap from last lecture**
- **Representation and Reasoning**
- **An Overview of This Course**
- **Further Dimensions of Representational Complexity**

# Course Essentials

- **Course web-page** : CHECK IT OFTEN!
- **Textbook**: Available online!
  - We will cover at least Chapters: 1, 3, 4, 5, 6, 8, 9
- **Connect**: discussion board, grades
- **AIspace** : online tools for learning Artificial Intelligence  
<http://aispace.org/> 
- **Lecture slides...**
- **Midterm exam**, planning to have in on Wed Jun 7 (will have a doodle on piazza)

# Agents acting in an environment



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# What do we need to represent ?

- **The environment / world** : What different configurations (**states / possible worlds**) can the world be in, and how do we denote them?

Chessboard, Info about a patient, Robot Location

- **How the world works** (*we will focus on*)
  - **Constraints:** *sum of current into a node = 0*  
*electric circuit*
  - **Causal:** *what are the causes and the effects of brain disorders?*  
*medicine*
  - **Actions** preconditions and effects: *when can I press this button? What happens if I press it?*

# Corresponding Reasoning Tasks / Problems

- **Constraint Satisfaction** – Find state that satisfies set of constraints. *E.g., What is a feasible schedule for final exams?*
- **Answering Query** – Is a given proposition true/likely given what is known? *E.g., Does this patient suffers from viral hepatitis?*
- **Planning** – Find sequence of actions to reach a goal state / maximize utility. *E.g., Navigate through and environment to reach a particular location. Collect gems and avoid monsters*

# Representation and Reasoning System

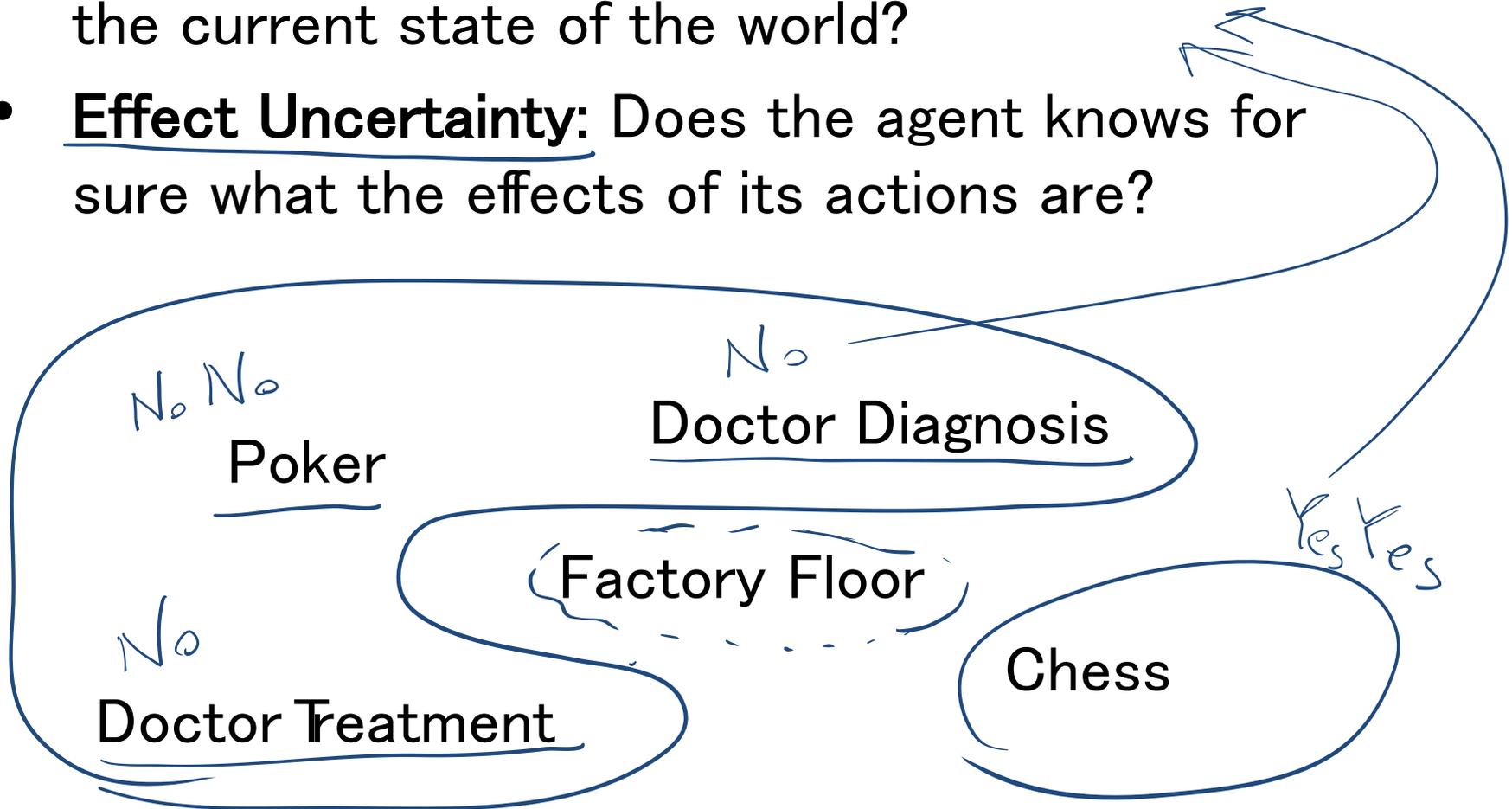
- A (**representation**) **language** in which the environment and how it works can be described
- Computational (**reasoning**) **procedures** to compute a solution to a problem in that environment (an answer a sequence of actions)



**But** the choice of an appropriate R&R system depends on a key property of the environment and of the agent's knowledge

# Deterministic vs. Stochastic (Uncertain) Domains

- Sensing Uncertainty: Can the agent fully observe the current state of the world?
- Effect Uncertainty: Does the agent know for sure what the effects of its actions are?



# Clicker Question: Chess and Poker

Stochastic if at least one of these is true

- **Sensing Uncertainty:** Can the agent fully observe the current state of the world?
- **Effect Uncertainty:** Does the agent know for sure what the effects of its actions are?

- A. Poker and Chess are both stochastic
- B. Chess is stochastic and Poker is deterministic
- C. Poker and Chess are both stochastic
- D. Chess is deterministic and Poker is stochastic

# Deterministic vs. Stochastic Domains

Historically, AI has been divided into two camps: those who prefer representations based on **logic** and those who prefer **probability**.

A few years ago, **CPSC 322** covered logic, while **CPSC 422** introduced probability:

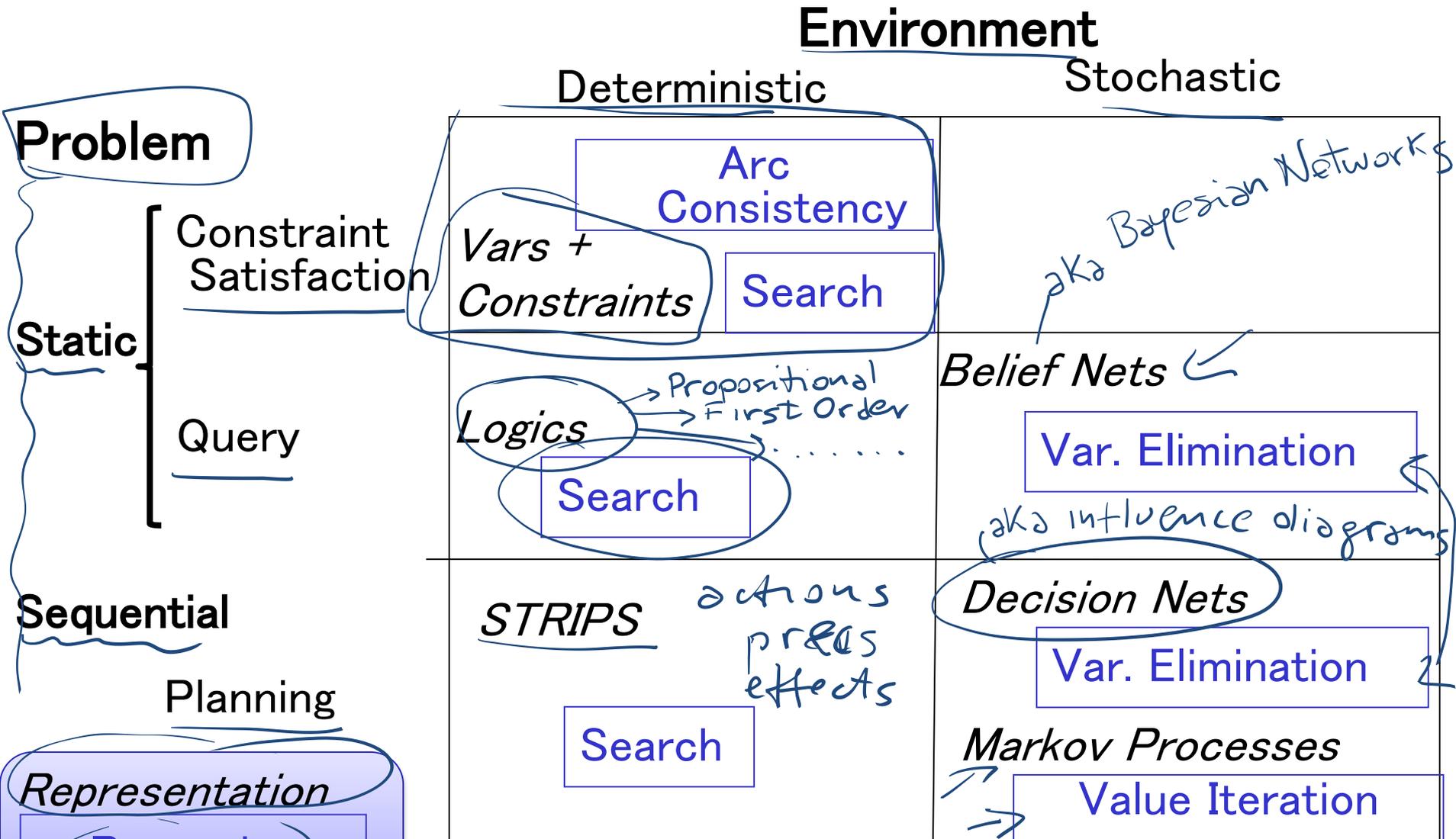
- now we introduce both representational families in 322, and 422 goes into more depth
- this should give you a better idea of what's included in AI

**Note:** Some of the most exciting current research in AI is actually building bridges between these camps.

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- Further Dimensions of Representational Complexity

# Modules we'll cover in this course: R&Rsys



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- Recap from last lecture
- Representation
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- **Further Dimensions of Representational Complexity**

# Dimensions of Representational Complexity

**We've already discussed:**

- Problems / Reasoning tasks (Static vs. Sequential)
- Deterministic versus stochastic domains

**Some other important dimensions of complexity:**

- Explicit state or <sup>features</sup> (propositions) or relations ←
- Flat or hierarchical — or binary features
- Knowledge given versus knowledge learned from experience
- Goals versus complex preferences
- Single-agent vs. multi-agent

# Explicit State or propositions

How do we model the environment?

- You can enumerate the **states** of the world.
- A state can be described in terms of **features**
  - Often it is more natural to describe states in terms of assignments of values to features (variables).
  - 30 binary features (also called propositions) can represent  $2^{30} = 1,073,741,824$  states.

$S_1 S_2 S_3 \dots$   
efficiency  
much more concise

## Mars Explorer Example

⇒ Weather {s, c}   
          <sup>sunny</sup>    <sup>cloudy</sup>

→ Temperature { -40, +40 }  
          <sup>longitude</sup>    <sup>latitude</sup>

LocX 0-359    LocY 0 179

one possible state  $\{S_1, +35, 30^\circ, 110^\circ\}$

$2 * 81 * 360 * 180$   
number of possible states  
mutually exclusive

# Explicit State or propositions or relations

- States can be described in terms of **objects** and **relationships**.
- There is a proposition for each relationship on each “possible” tuple of individuals.

## University Example

1 relationship

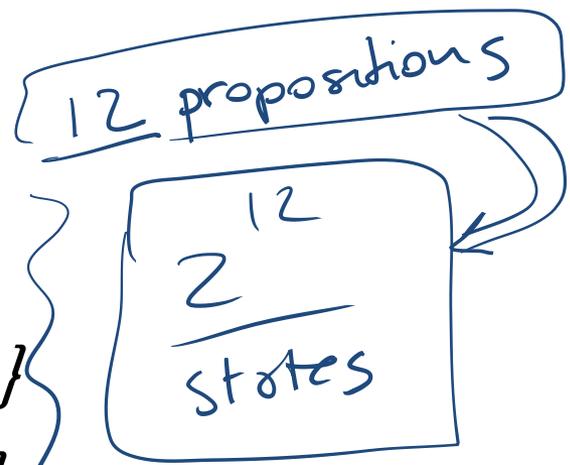
$Registered(S, C)$

$R(\underline{s_1}, \underline{c_2}) \in A_c$

$Students (S) = \{ s_1 s_2 s_3 s_4 \}$

$Courses (C) = \{ c_1 c_2 c_3 \}$

individuals/objects



# Clicker Question

One binary relation (e.g., *likes*) and 9 individuals (*people*). How many states?

- A.  $81^2$
- B.  $10^2$
- C.  $2^{81}$
- D.  $10^9$

I changed *same-nationality* to *likes* because if you reason on the meaning of *same-nationality* the states are less, they are  $2^{36}$

# Complete Example

$R \approx$  Registered  $R(s, c)$

Courses =  $\{c_1, c_2\}$   $|Courses| = 2$

Students =  $\{s_1, s_2, s_3\}$   $|Students| = 3$

All possible instantiations of R  $\left. \begin{matrix} R(s_1, c_1) \\ R(s_1, c_2) \\ R(s_2, c_1) \\ R(s_2, c_2) \\ R(s_3, c_1) \\ R(s_3, c_2) \end{matrix} \right\}$  each of these can be T=true or F=false

$3 * 2 = 6$

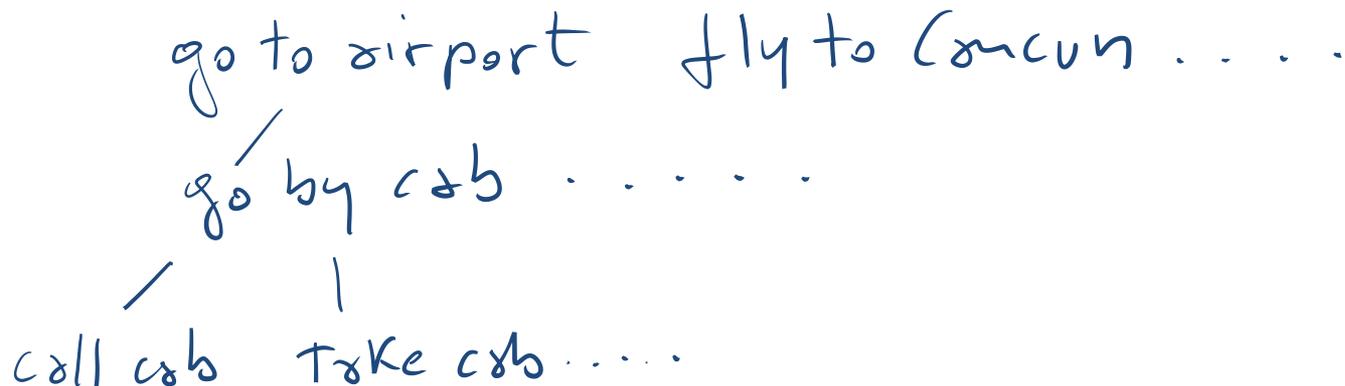
$R(s_1, c_1)$	T	T	T	T	F
$R(s_1, c_2)$	T	F	T	T	T
$R(s_2, c_1)$	T	T	F	T	F
$R(s_2, c_2)$	T	T	T	F	T
$R(s_3, c_1)$	T	T	T	T	T
$R(s_3, c_2)$	T	T	T	T	T

$2^6 = 64$  possible worlds

# Flat or hierarchical

Is it useful to model the whole world at the same level of abstraction?

- You can model the world at one level of abstraction: **flat**
- You can model the world at multiple levels of abstraction: **hierarchical**
- *Example: Planning a trip from here to a resort in Cancun, Mexico*



# Knowledge given vs. knowledge learned from experience

The agent is provided with a model of the world once and for all

- The agent **can learn** how the world works based on experience
  - in this case, the agent often still does start out with some **prior knowledge**

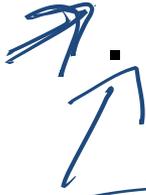
*not in this course*

# Goals versus (complex) preferences

An agent may have a **goal** that it wants to achieve



- e.g., there is some **state or set of states** of the world that the agent wants to be in
- e.g., there is some **proposition or set of propositions** that the agent wants to make true



$[0, 1]$   
worst / best

An agent may have **preferences**

- e.g., there is some **preference/utility function** that describes how happy the agent is in each state of the world; the agent's task is to reach a state which makes it as happy as possible



Preferences can be **complex**...

but Cappuccino takes 2mins  
Espresso takes 1min  
Agent must consider  
a TRADE-OFF

What beverage to order?

- *The sooner I get one the better*
- *Cappuccino better than Espresso*

# Single-agent vs. ~~Multiagent~~ domains

Does the environment include other agents?

Everything we've said so far presumes that there is only one agent in the environment.

- If there are other agents whose actions affect us, it can be useful to **explicitly model their goals and beliefs** rather than considering them to be part of the environment
- OtherAgents can be: **cooperative, competitive, or a bit of both**

# Dimensions of Representational Complexity in CPSC322

not in this  
course  
~

- Reasoning tasks (Constraint Satisfaction / Logic&Probabilistic Inference / Planning)
- Deterministic versus stochastic domains



## Some other important dimensions of complexity:

- Explicit state or features or relations
- Flat or ~~hierarchical~~
- Knowledge given versus knowledge learned from ~~experience~~
- Goals vs. (~~complex~~) preferences
- Single-agent vs. ~~multi-agent~~

simple

grad course

# In class activity

- Work in pair searching the web to find a **cool example of fielded (or experimental AI agents)** you found
- Try to find something different from the usual suspects (IBM Watson, Apple's Siri and Microsoft's Cortana)



# Try to answer the following questions (take notes)

1. What does the application do? (e.g., control a spacecraft, perform medical diagnoses, provide intelligent help for computer users, shop on eBay)
2. List some of the application : goals / preferences; observations that it needs about the environment; types of actions that it performs
3. What **AI technologies** does the application use (e.g., belief networks, Markov models, semantic networks, heuristic search, constraint satisfaction, planning)
4. Why is it intelligent? Which aspects make it an intelligent system?
5. Is it an experimental system or a fielded system (i.e., used in a real world setting)?
6. Is evidence provided on how well does the application perform?

## Next part

(SEE AI LANDSCAPE  
ON NEXT SLIDE)

- Assignment 0 due: **submit electronically** and **you can't use late days**
- **Hint: AAAI is the main AI association**
- Come to class ready to discuss the **two examples of fielded AI agents** you found  
*or experimental*
- I'll show some pictures of cool applications in that class

Read carefully Section 1.6 on textbook: “Example Applications”

- The Tutoring System
- The autonomous delivery robot
- The trading agent
- The diagnostic assistant



Autonomous Vehicle



AI Music Composition & Performance

AI Paintings

Robot Tour Guide

Characters for Virtual Worlds

Humanoid Robot

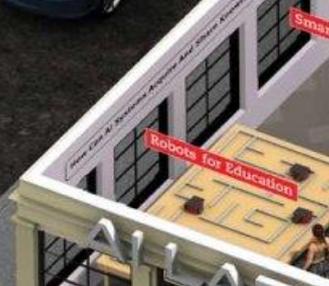
How Can AI Systems Learn to Perform Human Experiences?



Autonomous Space Exploration

Recommender System

When & Where Do We Need AI Systems?



Robots for Education



Smart Desk with Gesture Recognition

Disease Diagnosis

Drug Design

How Can AI Systems Solve Problems Creatively?



Robotic Surgery

Social Simulation Game

Smart Environmental Controls

Intelligent Tutoring System

Personalized TV Guide



Vehicle Navigation System

Robot Vacuum Cleaner



Aristotle

Leibniz

Whitehead

Russell

Turing

Lovelace

Descartes

Machine Translation

See the AI timeline and more at [www.aaai.org/AILandscape](http://www.aaai.org/AILandscape)

# The AI Landscape