

# Introduction to Artificial Intelligence (AI)

Computer Science cpsc502, Lecture 1

Sep, 8, 2011

2

# People

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This course is designed as a "breadth" introduction to AI.

It is suitable for those with

- no AI background ,
- or with only one undergraduate course in AI (or Machine Learning)

# Today Sept 8

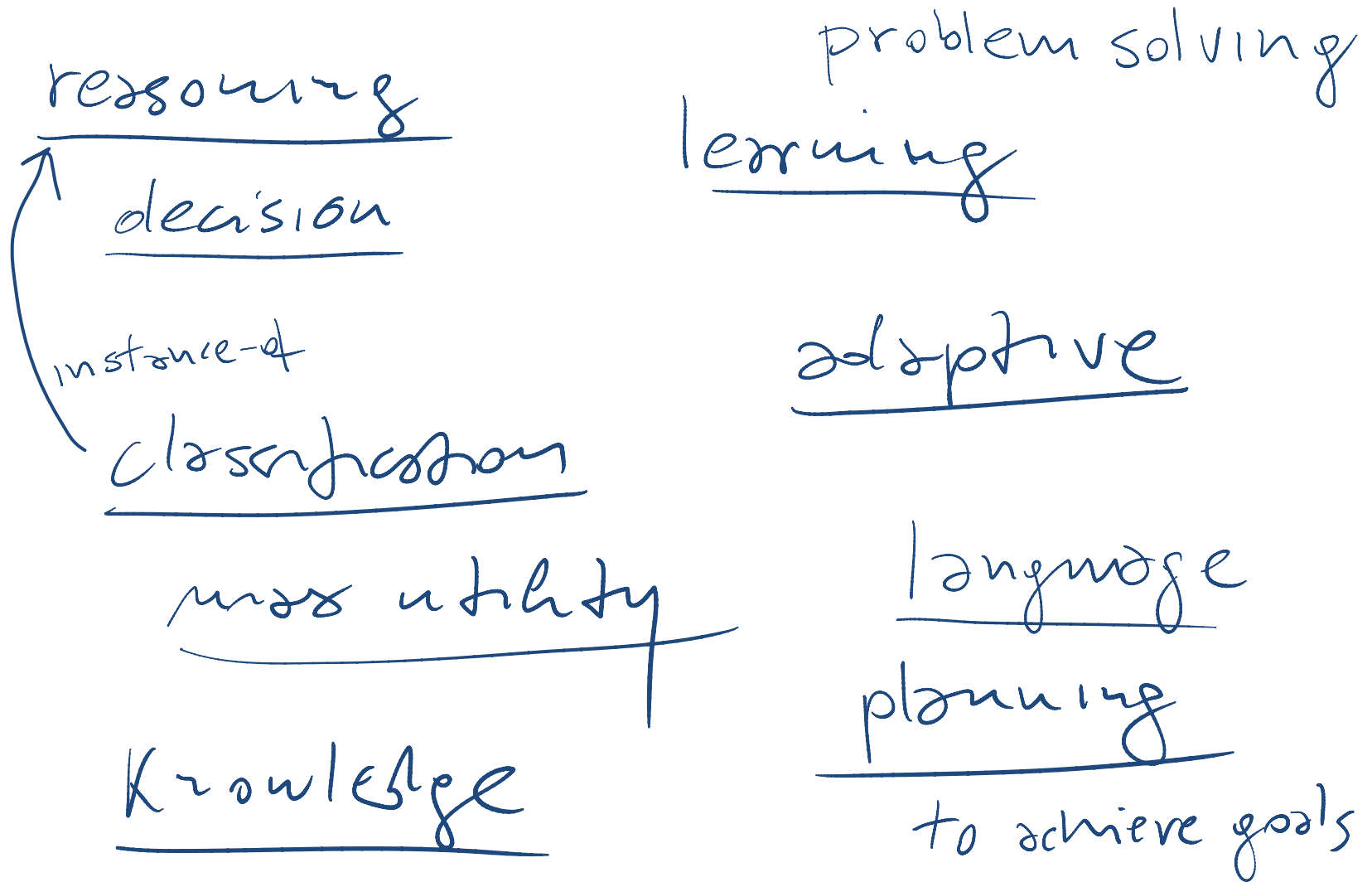
Overview of the field – Key definitions

Overview of course

- Background knowledge
- Topics
- Activities and Grading
- Administrative Stuff

Assignment-0

# What is Intelligence?



# Thinking and Acting Humanly

## Model the cognitive functions of human beings

- Humans are our only example of intelligence: we should use that example!

## Problems:

- But... humans often think/act in ways that we don't consider intelligent (**why?**)

emotions

cognitive limitations

memory  
tired

incorrect  
missing  
knowledge

- And... detailed model of how people's minds operate not yet available ←

# Thinking Rationally

**Rationality:** an abstract “ideal” of intelligence, rather than “whatever humans think/do”

- Ancient Greeks invented *syllogisms*: argument structures that always yield correct conclusions given correct premises
  - This led to **logic**, and **probabilistic reasoning** which we'll discuss in this course
- But correct sound reasoning is not always enough “to survive” “to be useful”...

# Acting (&thinking) Rationally

This course will emphasize a view of AI as building **agents**: artifacts that are able to think and act rationally in their environments

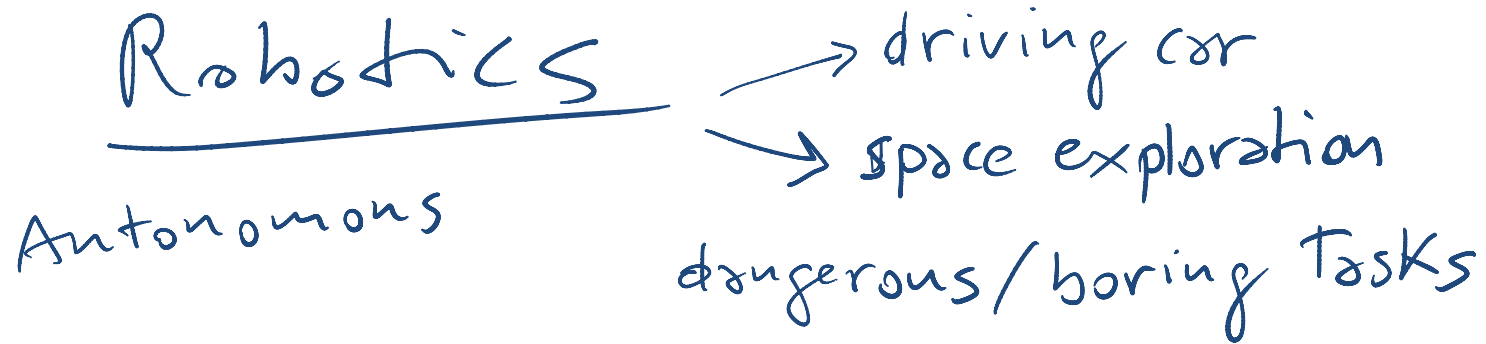
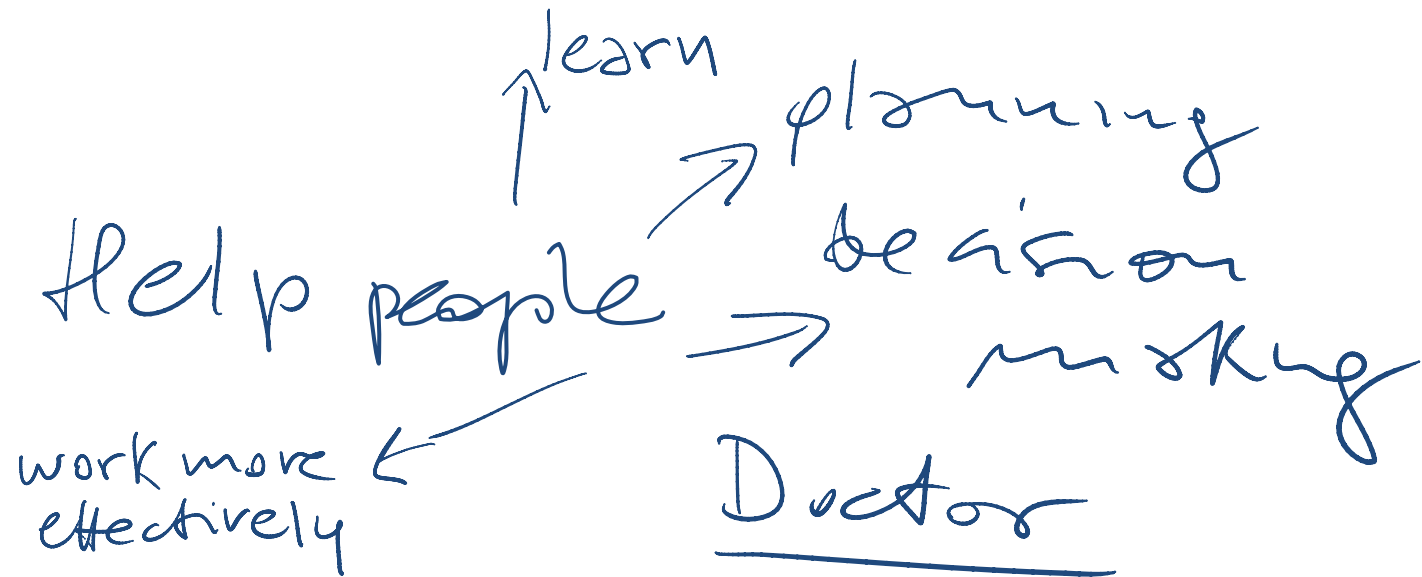
Rationality is **more cleanly defined** than human behavior, so it's a better design objective

(Eg: "intelligent" vacuum cleaner: maximize area cleaned, minimize noise and electricity consumption)

Agents that can **answer queries**, **plan actions** and **solve complex problems**

And when you have a rational agent you can always **tweak it to make it irrational!**

# Why do we need intelligent agents?



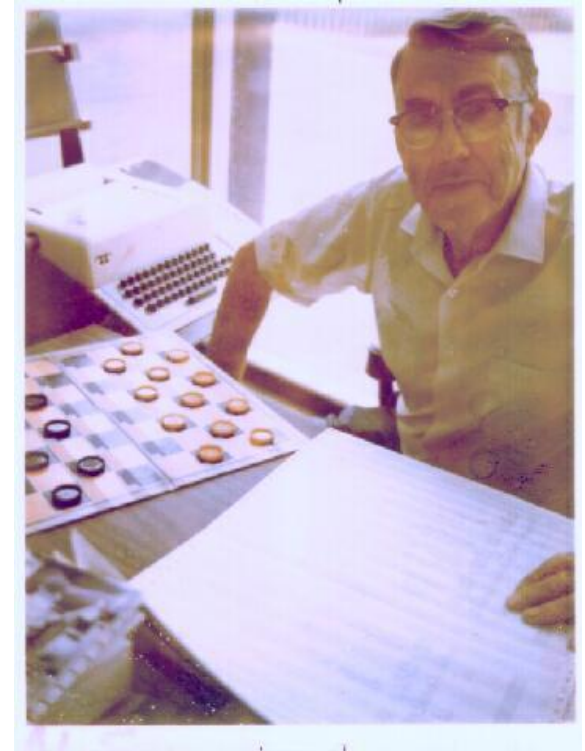


# (Adversarial) Search: Checkers

Game playing was one of the first tasks undertaken in AI

**Arthur Samuel** at IBM wrote programs to play checkers (1950s)

- initially, they played at a strong amateur level
- however, they used some (simple) machine learning techniques, and soon outperformed Samuel



Source: *IBM Research*

**Chinook's program** was declared the Man-Machine World Champion in checkers in 1994!

...and completely solved by a program in 2007!

# (Adversarial) Search: Chess

In 1996 and 1997, Gary Kasparov, the world chess grandmaster played two tournaments against Deep Blue, a program written by researchers at IBM



Source: *IBM Research*



# (Adversarial) Search: Chess

Deep Blue's Results in the first tournament:

- won 1 game, lost 3 and tied 1
  - ✓ first time a reigning world champion lost to a computer



Source: CNN

# (Adversarial) Search: Chess

Deep Blue's Results in the second tournament:

- second tournament: won 3 games, lost 2, tied 1



- 30 CPUs + 480 chess processors
- Searched 126.000.000 nodes per sec
- Generated 30 billion positions per move reaching depth 14 routinely

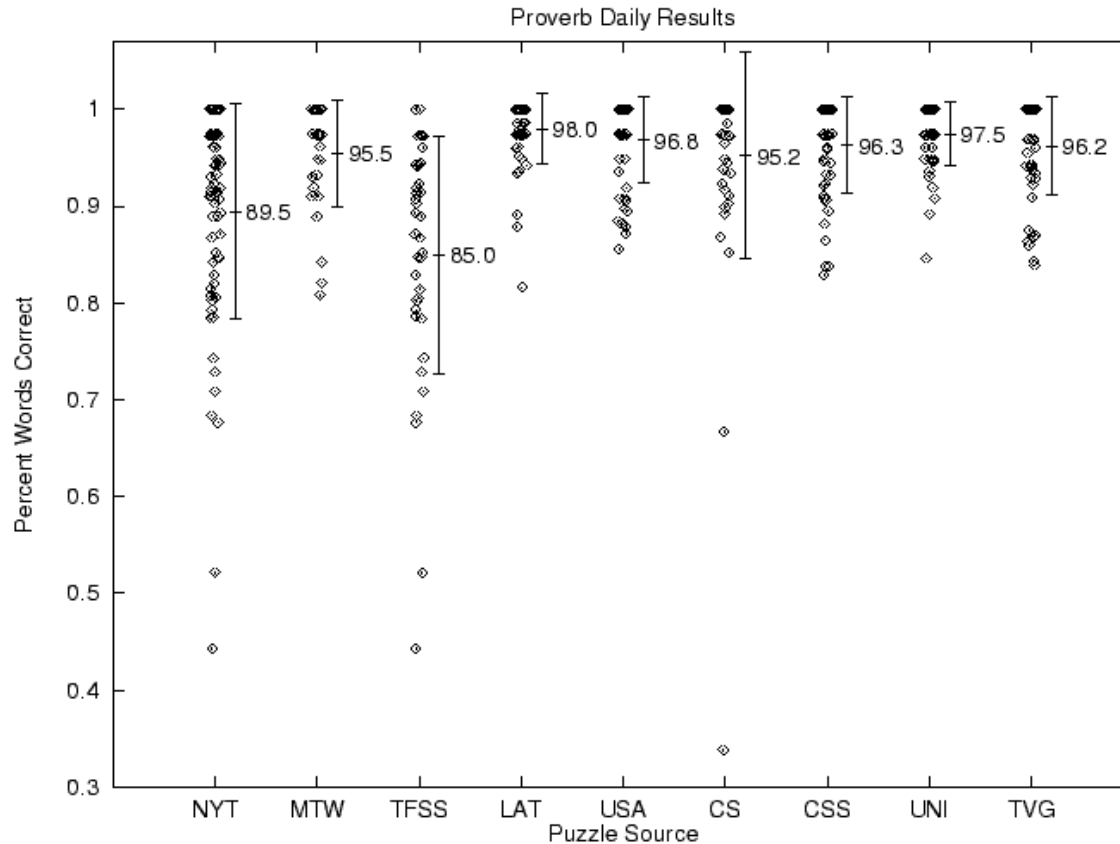
# CSPs: Crossword Puzzles

## Daily Puzzles

370 puzzles from 7 sources.

Summary statistics:

- ◆ 95.3% words correct (miss three or four words per puzzle)
- ◆ 98.1% letters correct
- ◆ 46.2% puzzles completely correct



P	O	L	O	N	E		P	A	L	O	M	I	N	O			
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S	L	E	E	V	E		T	H	W	A	R	T	E	D			
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Source: *Michael Littman*

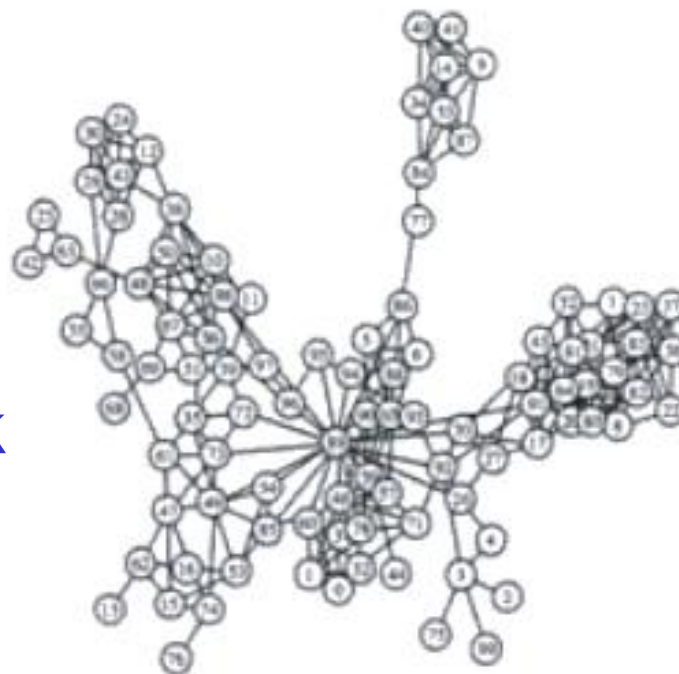
# CSPs: Radio link frequency assignment

Assigning frequencies to a set of radio links defined between pairs of sites in order to **avoid interferences**.

Constraints on frequency depend on **position of the links** and on **physical environment**.

Source: *INRIA*

Sample Constraint network



CPS

# Example: SLS for RNA secondary structure design

RNA strand made up of four bases: cytosine (C), guanine (G), adenine (A), and uracil (U)

2D/3D structure RNA strand folds into is important for its **function**

Predicting structure for a strand is “easy”:  $O(n^3)$

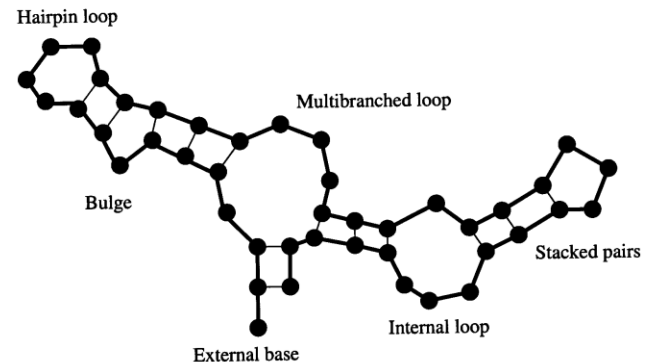
But what if we want a strand that folds into a certain structure?

- Local search over strands
  - ✓ Search for one that folds into the right structure
- Evaluation function for a strand
  - ✓ Run  $O(n^3)$  prediction algorithm
  - ✓ Evaluate how different the result is from our target structure
  - ✓ Only defined implicitly, but can be evaluated by running the prediction algorithm

RNA strand  
GUCCCAUAGGAUGUCCCAUAGGA

↓ Easy ↑ Hard

Secondary structure



Best algorithm to date: Local search algorithm RNA-SSD **developed at UBC**  
[Andronescu, Fejes, Hutter, Condon, and Hoos, Journal of Molecular Biology, 2004]

# Constraint optimization problems

Optimization under side constraints (similar to CSP)

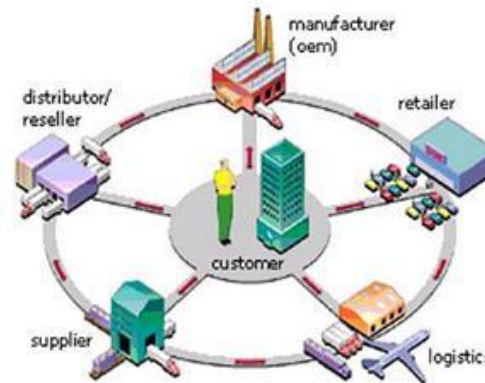
E.g. mixed integer programming (software: **IBM CPLEX**)

- **Linear** program:  $\max c^T x$  such that  $Ax \leq b$
- **Mixed integer** program: additional constraints,  $x_i \in \mathbb{Z}$  (integers)
- NP-hard, widely used in operations research and in industry



**Transportation/Logistics:**

SNCF, United Airlines  
UPS, United States  
Postal Service, ...



**Supply chain management software:**  
Oracle,  
SAP, ...



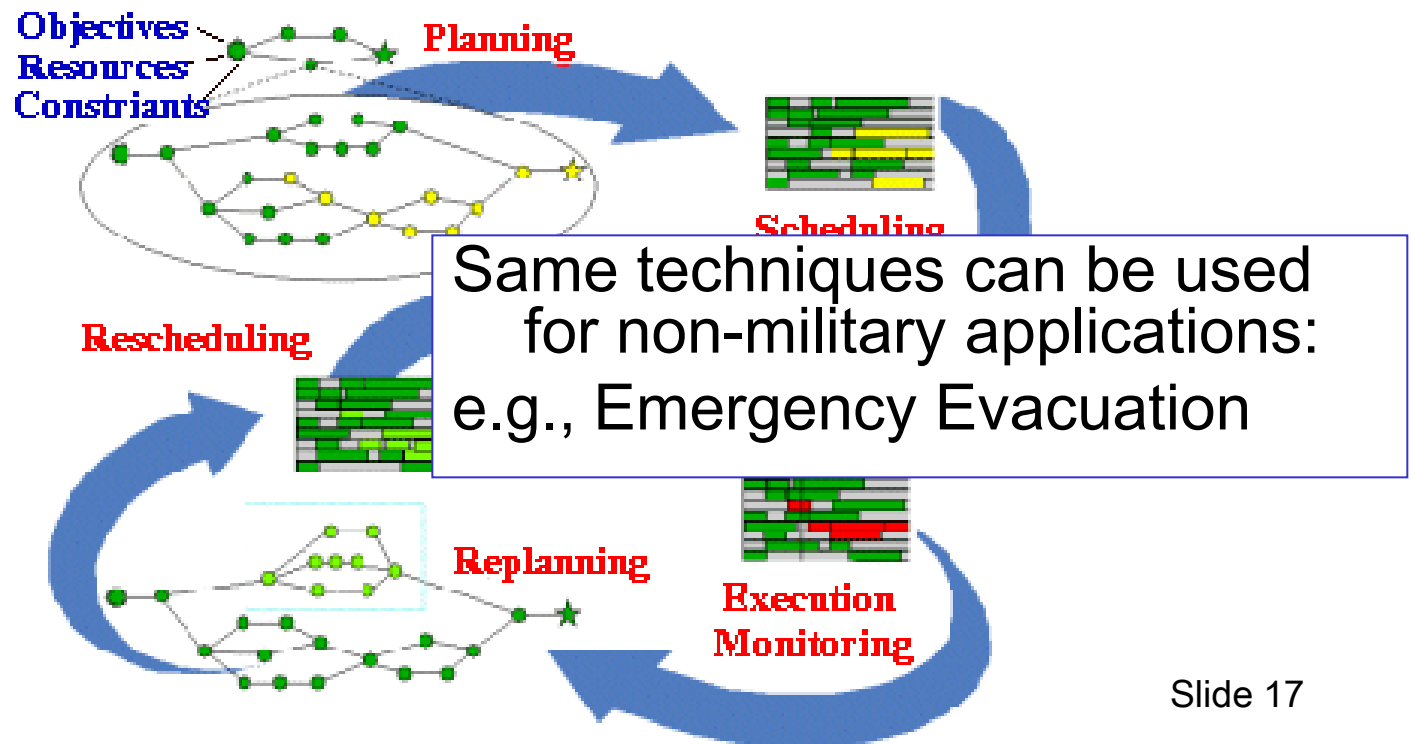
**Production planning and optimization:**  
Airbus, Dell, Porsche,  
Thyssen Krupp,  
Toyota, Nissan, ...



# Planning & Scheduling: Logistics

## Dynamic Analysis and Replanning Tool (Cross & Walker)

- logistics planning and scheduling for military transport
- used in the 1991 Gulf War by the US
- problems had 50,000 entities (e.g., vehicles); different starting points and destinations



# CSP/logic: formal verification



Hardware verification  
(e.g., IBM)



Software verification  
(small to medium programs)

Most progress in the last 10 years based on:  
Encodings into propositional satisfiability (SAT)

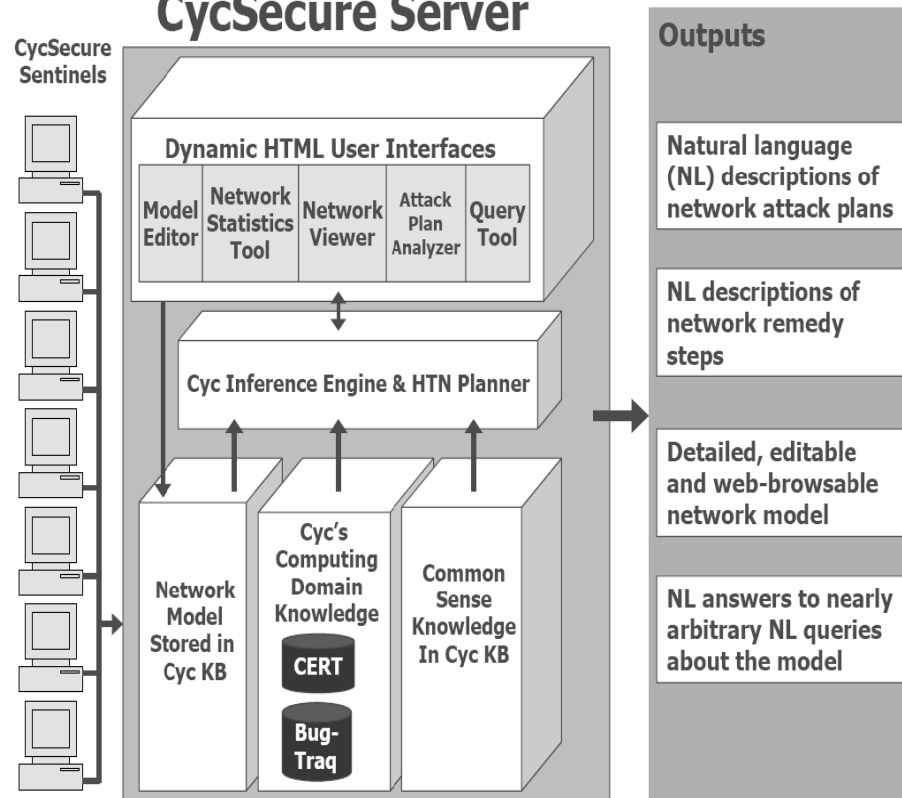
# Logic: CycSecure

“scans a computer network to build a formal representation of the network, based on Cyc’s pre-existing ontology of networking, security, and computing concepts:

Excerpted from: *Shepard et al., 2005*  
**CycSecure Server**

This formal representation also allows users to interact directly with the model of the network, allowing testing of proposed changes.”

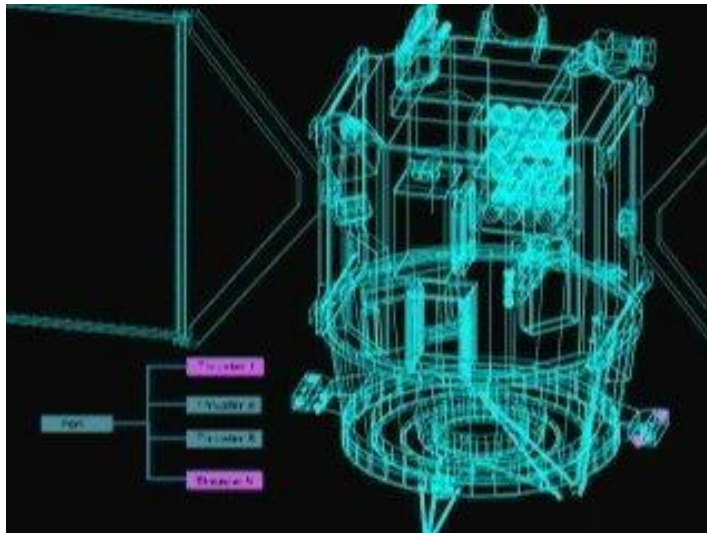
- Knowledge Representation
- Semantic Web !



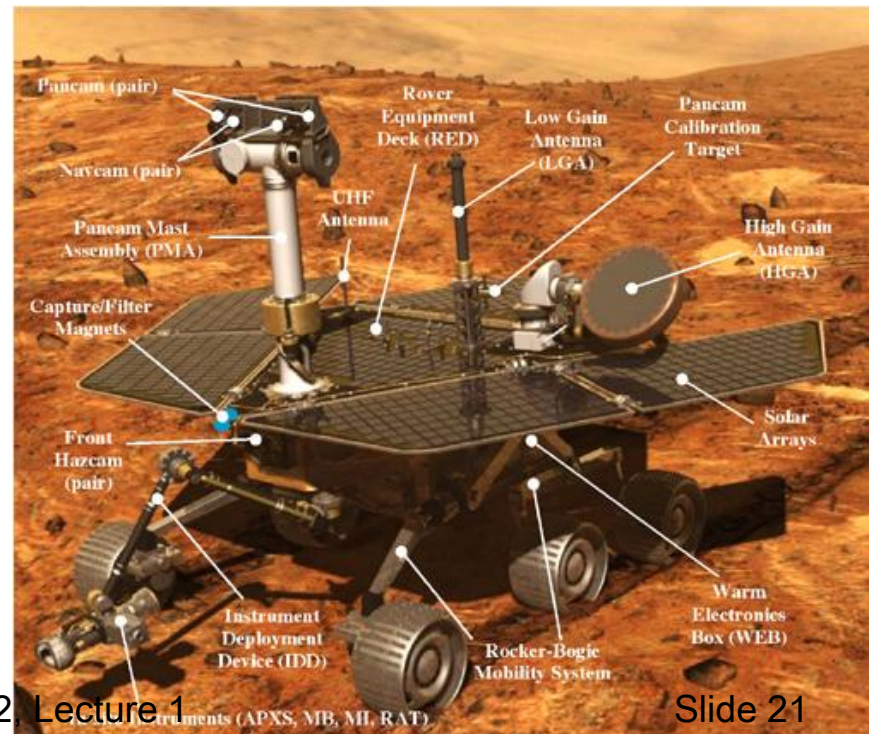
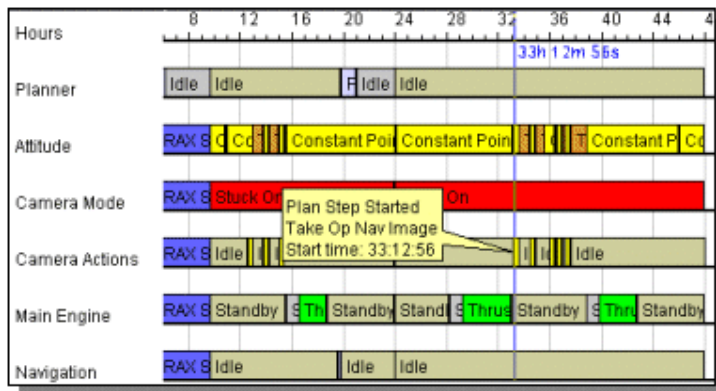
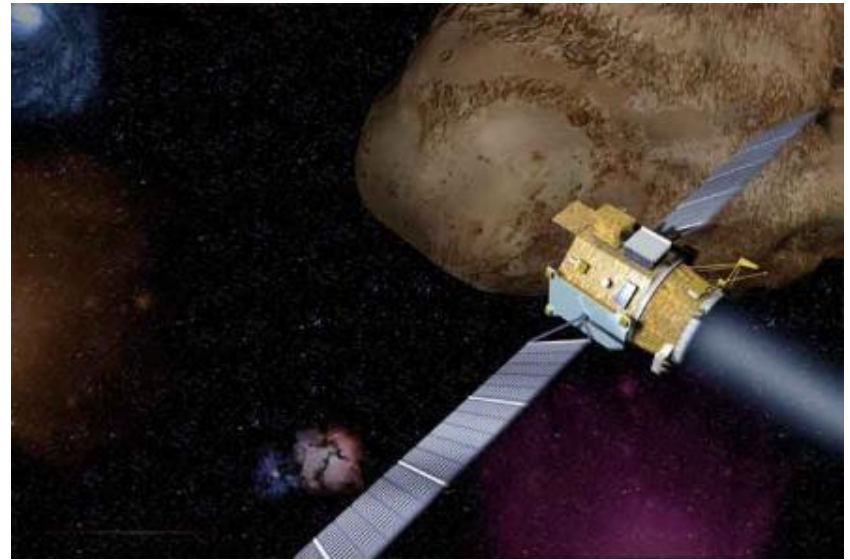
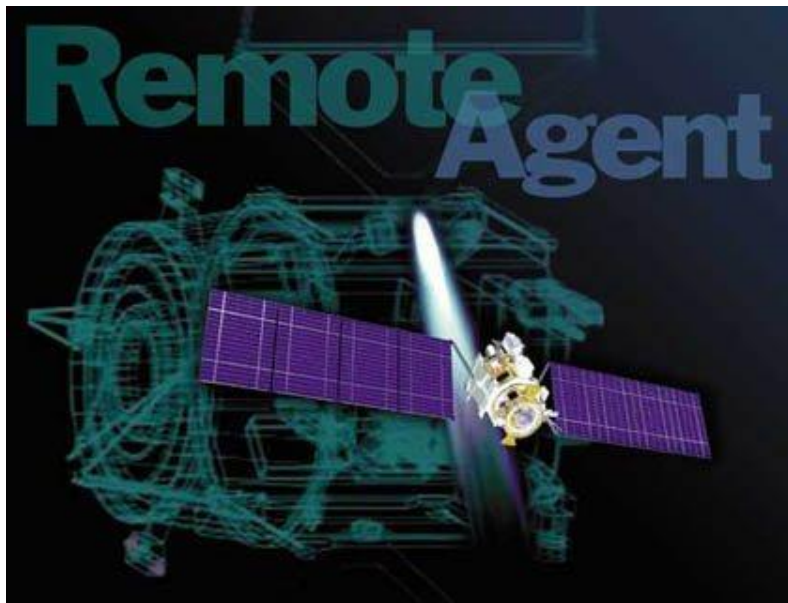
# Planning: Spacecraft Control

NASA: Deep Space One spacecraft operated autonomously for two days in May, 1999:

- determined its precise position using stars and asteroids
  - ✓ despite a malfunctioning ultraviolet detector
- planned the necessary course adjustment
- fired the ion propulsion system to make this adjustment

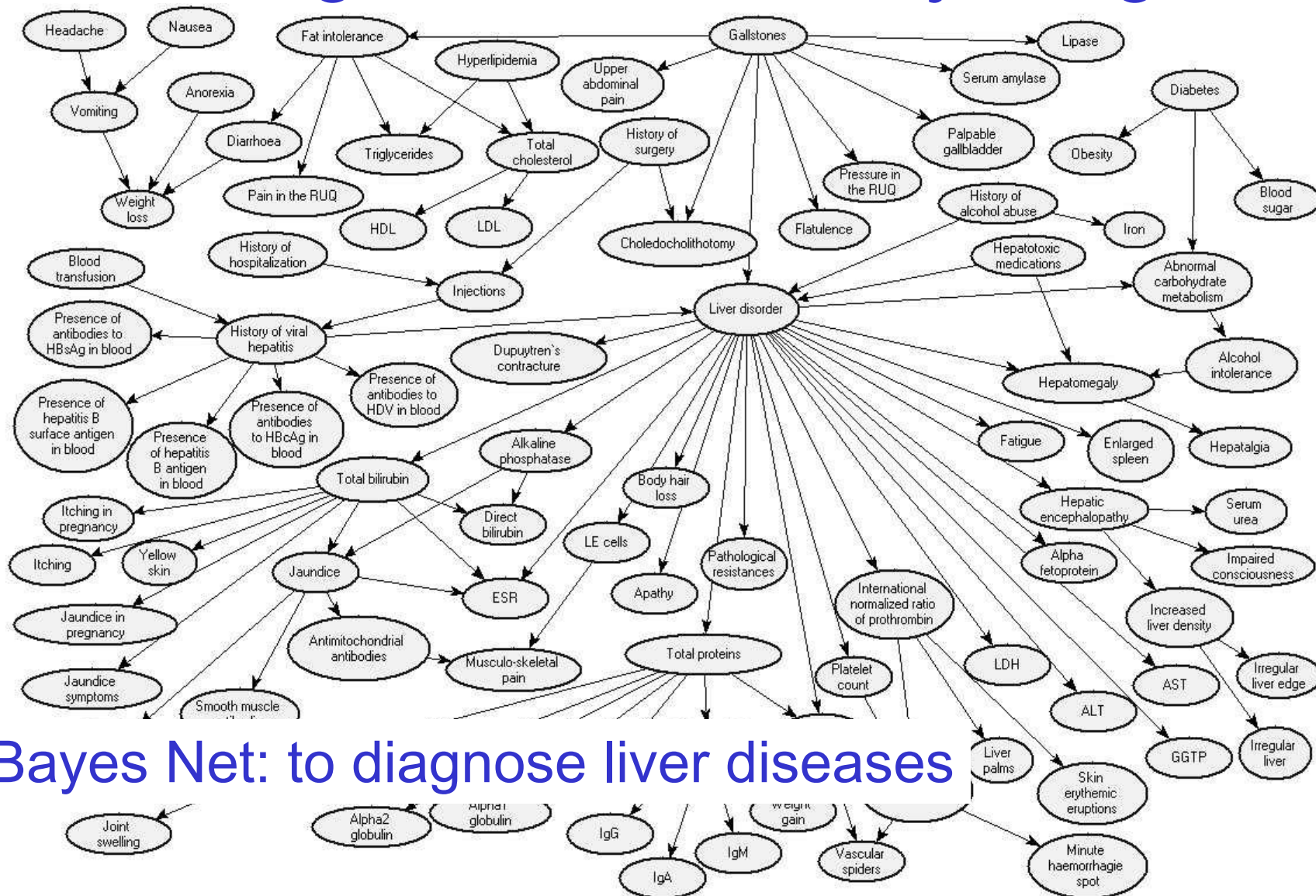


For another space application see the Spike system for the Hubble telescope



Source:  
cs221 stanford

# Reasoning under Uncertainty: Diagnosis



## Bayes Net: to diagnose liver diseases

# Reasoning Under Uncertainty

## Texture classification using Support Vector Machines

- foliage, building, sky, water



Source: *Mike Cora, UBC*

# Reasoning Under Uncertainty

E.g. motion tracking: track a hand and estimate activity:

- drawing, erasing/shading, other



Source:  
*Kevin Murphy,*  
Slide 24 *UBC*



# Computer Vision (not just for robots!)

## Jing, Baluja, Rowley, Google: Finding Canonical Images

Web Images Maps News Shopping Gmail more

Sign in



mona lisa

Search Images

Search the Web

Advanced Image Search  
Preferences

Strict SafeSearch is on

New! Google Image Labeler

Images Showing: All image sizes

Results 1 - 21 of about 343,000 for mona lisa with Safesearch on. (0.04 seconds)



Word has it that **Mona Lisa** wasn't a ...  
320 x 366 - 21k - jpg  
uk.gizmodo.com



da Vinci: **Mona Lisa**  
340 x 472 - 10k - gif  
www.enchantedlearning.com



**Mona Lisa** We have examined the topic ...  
379 x 589 - 63k - jpg  
thesituationist.wordpress.com



**Mona Lisa** right  
282 x 795 - 59k - jpg  
www.museumldv.com



**Mona Lisa** made from train tickets --  
468 x 296 - 67k - jpg  
www.pinktentacle.com



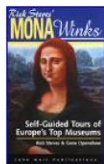
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commons.wikimedia.org



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Study Page: **Mona Lisa** in Book  
Cover ...  
360 x 595 - 85k - gif  
www.studiolo.org



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www.cartoonstock.com



**Mona Lisa**  
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**Mona Lisa** - Joint Poster  
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www.allposters.com



"**Mona Lisa**"  
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www.oregoncoastradio.com



**Mona Lisa** is **Lisa** Gherardini  
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yedda.com



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www.paris.org



Sir Joshua's **Mona Lisa**  
502 x 502 - 50k - jpg  
www.moviespring.com



Complete history of **Mona Lisa**  
450 x 328 - 22k - jpg  
www.simplonpc.co.uk



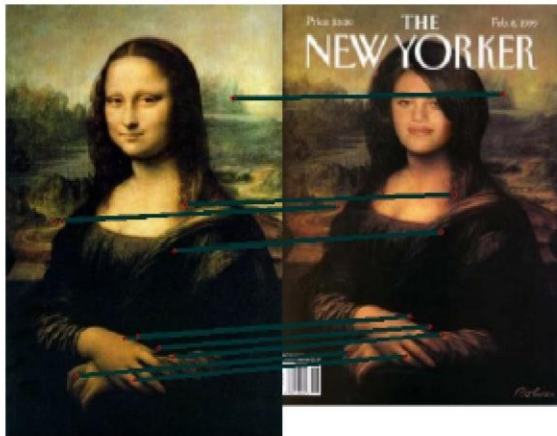
**Mona Lisa** Magnet by Leonardo da  
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348 x 450 - 29k - jpg  
www.allposters.com



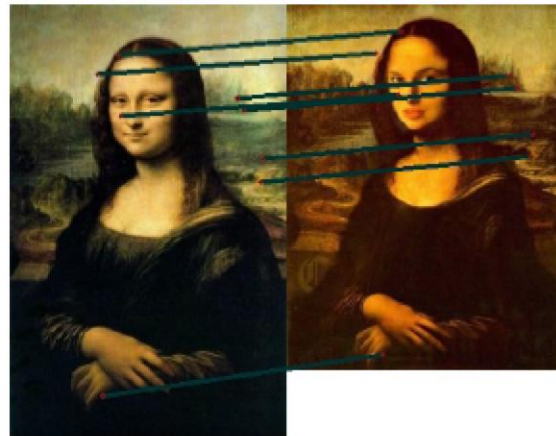
CPSC 502, Lecture 1

New! Want to help improve Google Image Search? Try [Google Image Labeler](#).

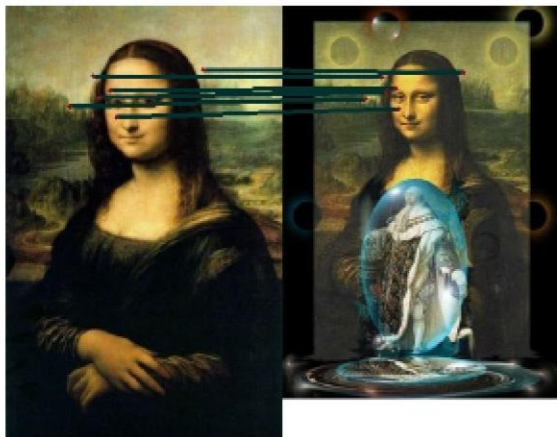
# Compare low level features



(a) A v.s. B



(b) A v.s. C



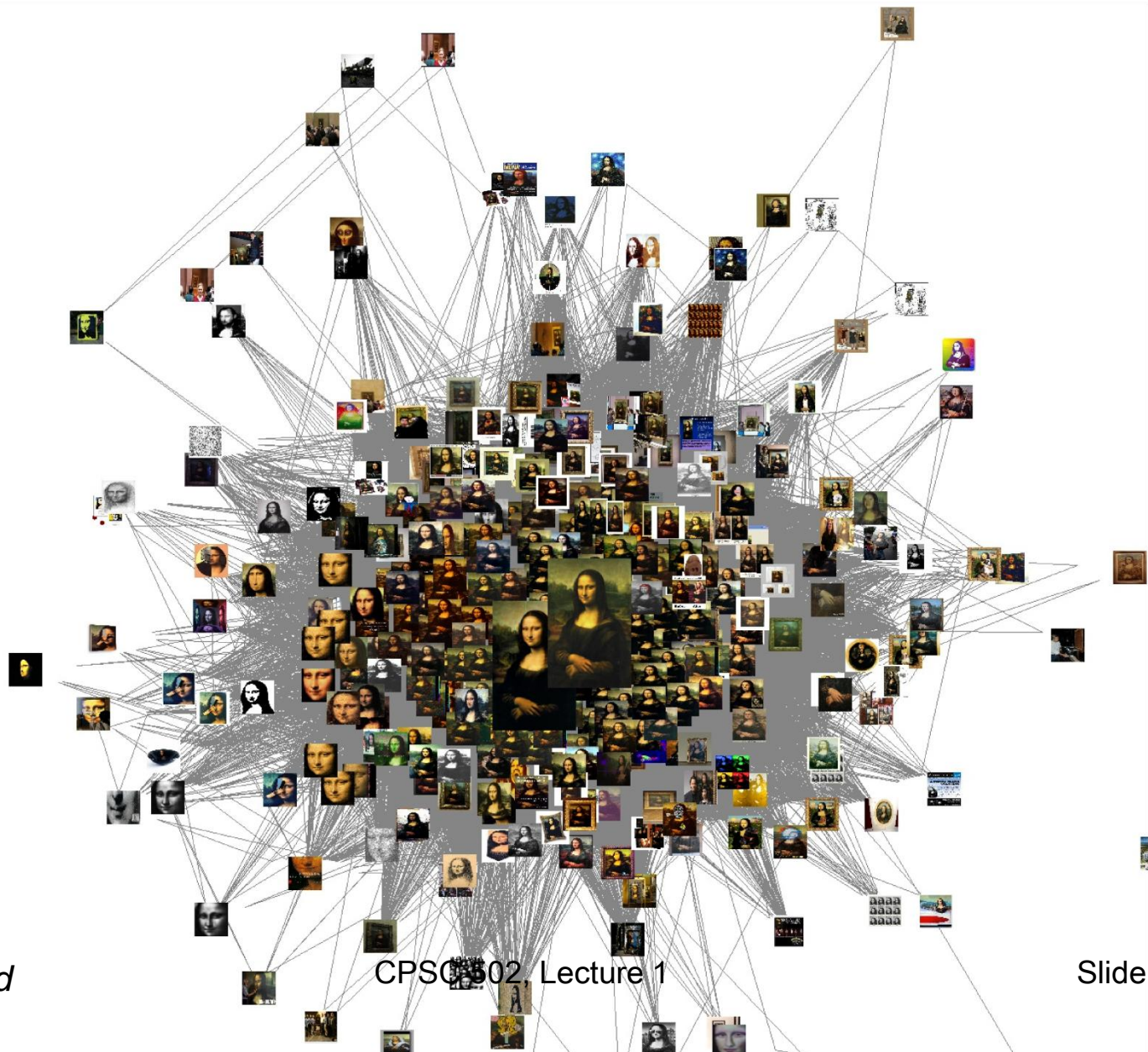
(c) A v.s. D



(d) B v.s. C



# Induced Graph



Source:  
*cs221 stanford*

CPS002, Lecture 1

Slide 28

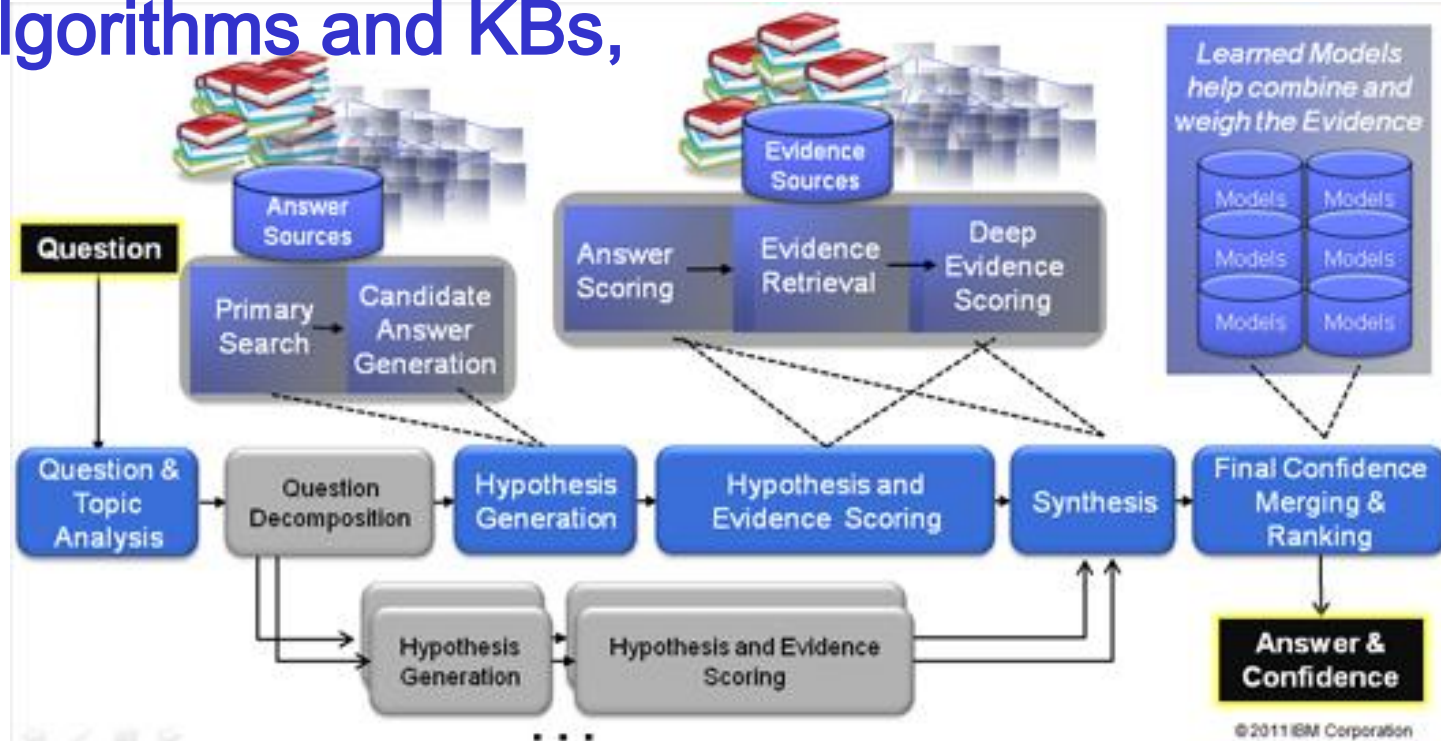
# AI - Machine Learning @google

- Spam/Porn Detection
- Which ad to place given a query
- Train Speech to search on mobile
- Machine Translation
- .....
- Highly Parallelizable EM + Map Reduce (simple code to write)
- Stochastic Gradient Descent

**Watson** : analyzes natural language questions and content well enough and fast enough to compete and win against champion players at Jeopardy!

*“This Drug has been shown to relieve the symptoms of ADD with relatively few side effects.”* • **3 secs**

- **1000s of algorithms and KBs,**



Source:  
IBM

# Planning Under Uncertainty

Learning and Using POMDP models of Patient-Caregiver Interactions During Activities of Daily Living

**Goal:** Help Older adults living with cognitive disabilities (such as Alzheimer's) when they:

- forget the proper sequence of tasks that need to be completed
- they lose track of the steps that they have already completed.



# Military applications: ethical issues

- Robot soldiers
  - Existing: robot dog carrying heavy materials for soldiers in the field
  - The technology is there
- Unmanned airplanes
- Missile tracking
- Surveillance
- ...



# Planning Under Uncertainty

Helicopter control: MDP, reinforcement learning

**States:** all possible positions, orientations, velocities and angular velocities

Final solution involves  
**Deterministic search!**



Source: *Andrew Ng 2004*



# Decision Theory: Decision Support Systems

E.g., **Computational Sustainability**

New interdisciplinary field, **AI** is a key component

- Models and methods for **decision making** concerning the **management and allocation of resources**
- to solve most challenging problems related to **sustainability**

Often **constraint optimization problems**. E.g.

- **Energy**: when and where to produce green energy most economically?
- Which parcels of land to purchase to **protect endangered species**?
- **Urban planning**: how to use budget for best development in 30 years?



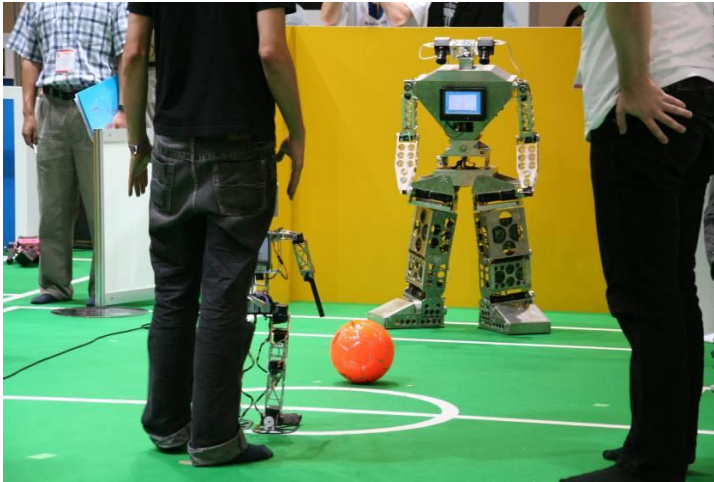
# Multiagent Systems: Poker



**Search Space:** 1.2  
quintillion nodes

“In full 10-player games Poki is better than a typical low-limit casino player and wins consistently; however, not as good as most experts  
New programs being developed for the 2-player game are quite a bit better, and we believe they will very soon surpass all human players”

# Multiagent Systems: Robot Soccer



## Extremely complex

- Stochastic
- Sequence of actions
- Multiagent

robotic soccer competition was proposed by LCI (UBC) in 1992 (which became *Robocup* in 1997).

Source: *RoboCup* web site

# Statistical Machine Translation

SEHR GEEHRTER GAST!  
KUNST, KULTUR UND  
KOMFORT IM HERZEN  
BERLIN.

DEAR GUESTS,  
ART, CULTURE AND  
LUXURY IN THE HEART  
OF BERLIN.



DIE ÖRTLICHE  
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BEI 50 HERTZ.

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IS 220/240 VOLTS 50 HZ.



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# Zite: a personalized magazine

... that gets smarter as you use it

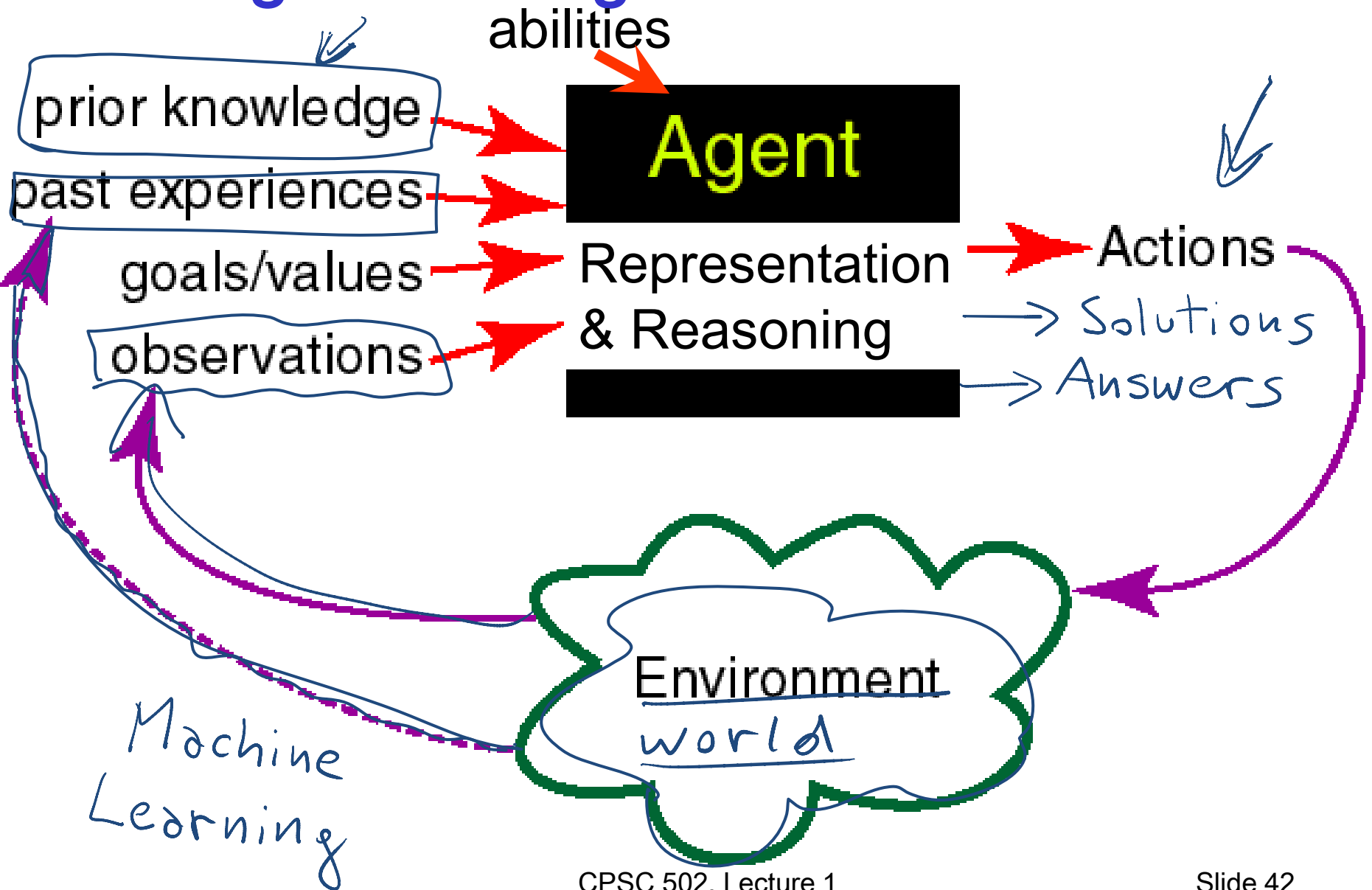




See the AI timeline and more at [www.aaai.org/AIlandscape](http://www.aaai.org/AIlandscape)

# The AI Landscape

# Agents acting in an environment





# What is an agent?

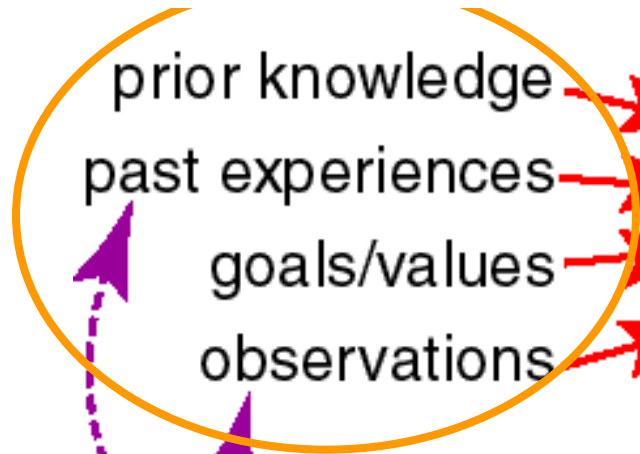
It has the following characteristics:

- It is situated in some **environment**
  - does not have to be the real world---can be an abstracted electronic environment *Medical test / Eye tracking* ↙
- It can make **observations** (*perhaps imperfectly*)
- It is able to **act** (*provide an answer, buy a ticket*)
- It has **goals or preferences** (*possibly of its user*) ↙  
*real estate advisor*
- It may have **prior knowledge or beliefs**, and some way of **updating beliefs** based on new experiences (to reason, to make inferences)

# Intelligent Agents in the World

**Knowledge Representation**  
**Machine Learning**

abilities



**Reasoning +  
Decision Theory**

Actions

**Natural Language  
Generation**

**Natural Language  
Understanding**

+

**Computer Vision  
Speech Recognition**

+

**Physiological Sensing  
Mining of Interaction Logs**



+

**Robotics**

+

**Human Computer  
/Robot  
Interaction**

# Today Sept 9

Overview of the field – Key definitions

Overview of course

- Background knowledge
- **Topics**
- Activities and Grading
- Administrative Stuff

Assignment 0

# 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:** *electric circuit* sum of current into a node = 0
  - **Causal:** *medicine* what are the causes and the effects of brain disorders?
  - **Actions** preconditions and effects: *when can I press this button? What happens if I press it?*

# Corresponding Reasoning Tasks / Problems

## STATIC

- **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?*

## SEQUENTIAL

- **Planning** – Find sequence of actions to reach a goal state / maximize utility. *E.g.*, *Navigate through and environment to reach a particular location*

# Representation and Reasoning System

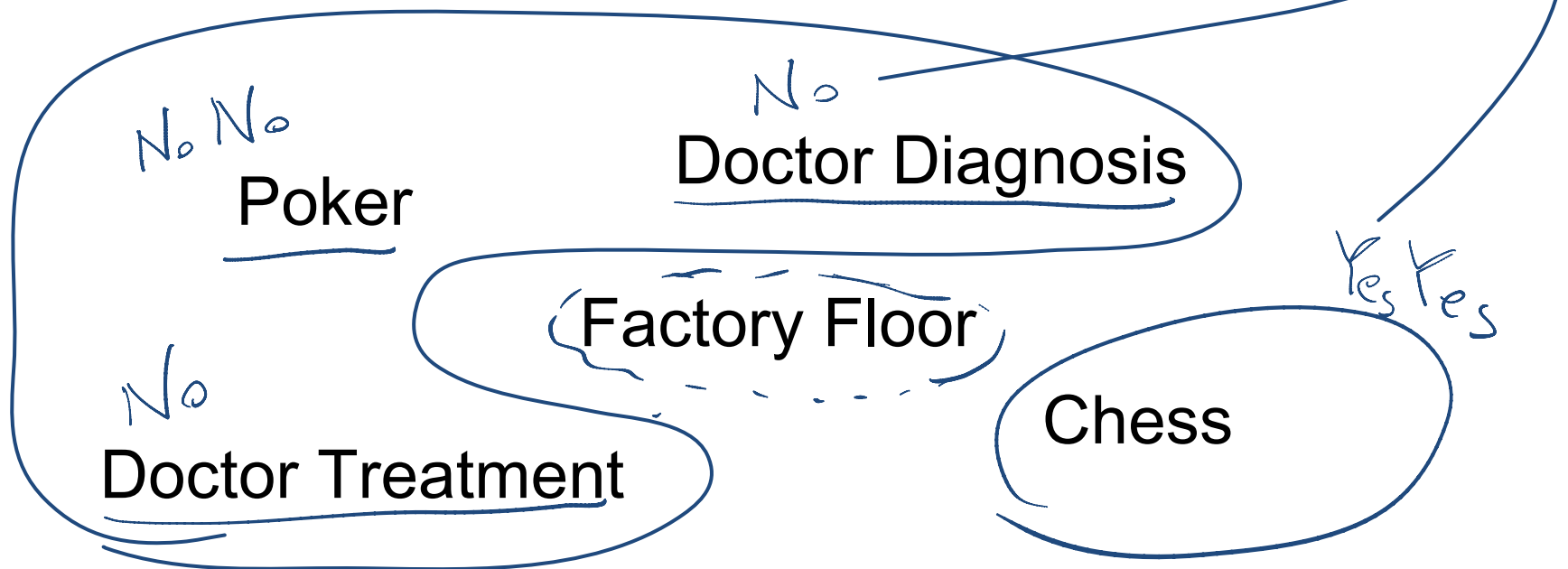
- A (**representation**) language in which the environment and how it works can be described
- Computational (**reasoning**) procedures to compute a solution to a problem in that environment (an answer, a sequence of actions)



**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?



# Deterministic vs. Stochastic Domains

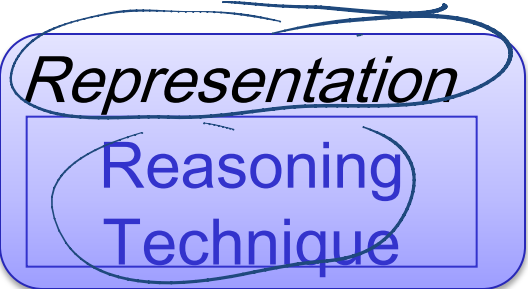
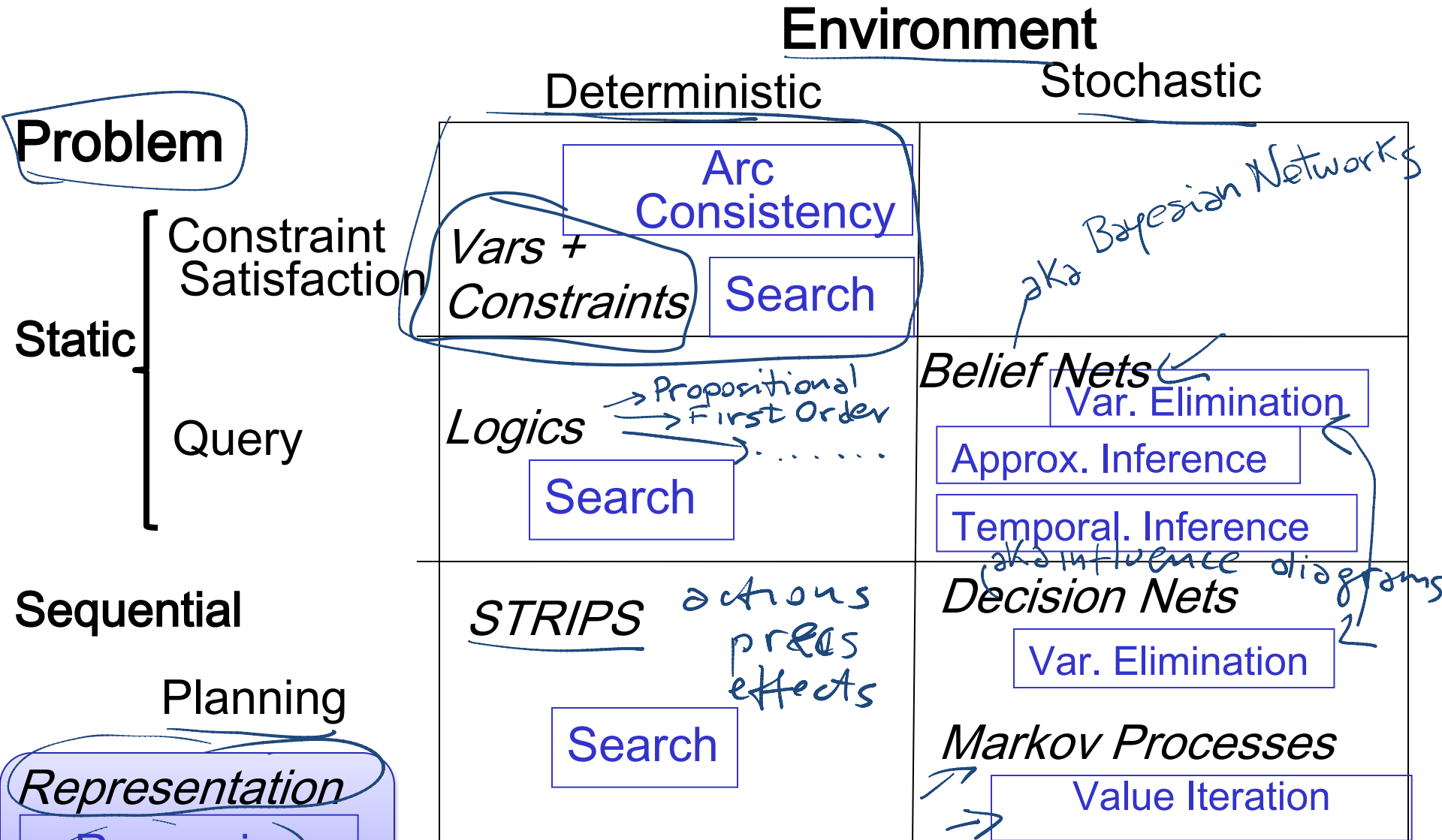
Historically, AI has been divided into two camps:

- those who prefer representations based on **logic**
- those who prefer **probability**.

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



# R&Rsys we'll cover in this course



# 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**
- **Supervised Machine Learning**
- **Unsupervised Machine Learning**
- **Reinforcement Learning**

# Dimensions of Representational Complexity

We've already discussed:

- Reasoning tasks (Static vs. Sequential )
- Deterministic versus stochastic domains
- Knowledge given versus knowledge learned from experience (Machine Learning)



Some other important dimensions of complexity:

- Explicit state or <sup>features</sup> (propositions) or relations
- Flat or hierarchical <sub>or binary features</sub>
- 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. *S<sub>1</sub> S<sub>2</sub> S<sub>3</sub> ... efficiency*
- A state can be described in terms of **features** *much more concise*
  - 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.

## Mars Explorer Example

⇒ Weather { *sunny* S, *cloudy* C }  
⇒ Temperature { -40, +40 }  
*longitude* Long 0-359 *latitude* Lati 0 179

*one possible state* { *S<sub>1</sub>*, +35, 30°, 110° }

$2 * 81 * 360 * 180$   
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

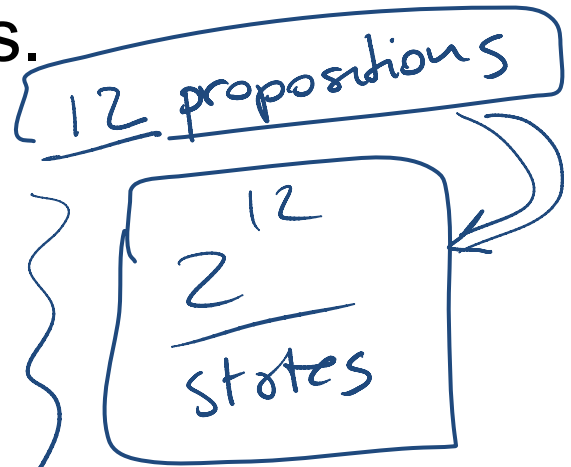
$Registered(S, C)$

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

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

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

individuals/objects

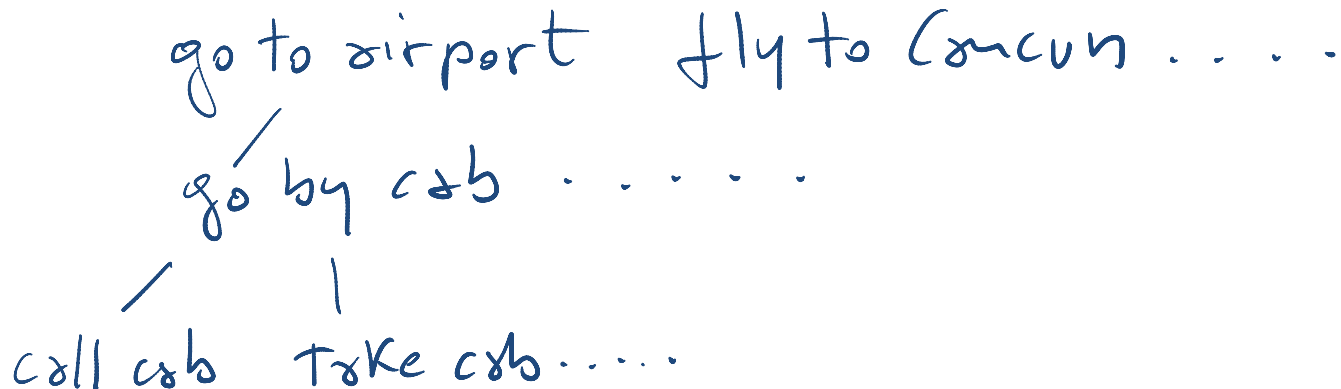


- Textbook example: One binary relation and 10 individuals can represent  $10^2=100$  propositions and  $2^{100}$  states!

# 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



# 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

An agent may have **preferences**  $[0, 1]$  <sup>worst</sup> <sup>best</sup>

- 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
- Other Agents can be: **cooperative, competitive, or a bit of both**



# Dimensions of Representational Complexity

- Reasoning tasks (Constraint Satisfaction / Logic&Probabilistic Inference / Planning)
- Deterministic versus stochastic domains
- Knowledge given versus knowledge learned from experience

## Some other important dimensions of complexity:

- Explicit state or features or relations
- Flat or ~~hierarchical~~
- Goals vs. (~~complex~~) preferences
- Single-agent vs. ~~multi-agent~~

grad course

# Today Sept 8

Overview of the field – Key definitions

Overview of course

- Background knowledge
- Topics
- **Activities and Grading**
- **Administrative Stuff**

**Assignment 0**

# Activities

- For the first 9 or 10 weeks we will cover an overview of the material in **lectures** and **assignments**
- There will be a **midterm** on the material covered in class
- The last few weeks will cover **current research topics presented by students**
- Students will **write a review paper** based on the lecture they gave (+ additional relevant papers), and **peer review** the papers of other students

# Tentative Grading Scheme

- 5% Class Participation
- 25% Assignments
- 30% Midterm exam
- 10% Your Presentation
- 30% Your Essay

# Textbook - Readings

## Required

- Selected Chapters of **Artificial Intelligence: foundations of computational agents** by D. Poole and A. Mackworth, Cambridge University Press, 2010
- Between 15 and 20 research papers from recent conferences/journals in Artificial Intelligence

## Reference

- **Artificial Intelligence : A Modern Approach**, by Russell and Norvig, 3rd Edition (Prentice-Hall, 2010)
- **Synthesis Lectures in Artificial Intelligence** [webpage](#)

# Class Forum: Piazza

Join the class **asap** via the signup link below.

**<http://www.piazza.com/ubc.ca/fall2011/cpsc502>**

You need a **ubc.ca** or **cs.ubc.ca** email address to sign up. If you do not have one, please send an email to **rjoty@cs.ubc.ca**

# TODO for next Tue

**Read Chp 1 and 3 of textbook**

**Assignment 0: available from the course web page**

- Join **piazza** (the class discussion forum)
- Read **Course web-pages:**

[www.cs.ubc.ca/~carenini/TEACHING/CPSC502-11/502-11.html](http://www.cs.ubc.ca/~carenini/TEACHING/CPSC502-11/502-11.html)

**WebSearch: Giuseppe Carenini**

This is where most information about the course will be posted,  
most handouts (e.g., slides) will be distributed, etc.

**CHECK IT OFTEN!**