CS343 Introduction to Artificial Intelligence Spring 2012

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Welcome to a **fun**, but **challenging** course.

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Goal

• Learn about Artificial Intelligence

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 - Increase your Al Literacy
 - Prepare you for Topics Courses and/or Research

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- Breadth over Depth

• Artificial Intelligence is....

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Textbook: Autonomous Agents

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No generally accepted definition

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- No generally accepted definition
- I know one when I see one...

Artificial Intelligence is....

Textbook: Autonomous Agents

- No generally accepted definition
- I know one when I see one...
- ... By the end of this course, so will you

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– What is a mind?

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 Is a running computer (just) a physical object?
 - Can we build a mind?
 - Can trying to build one teach us what a mind is?

Today

- 1. An introduction to What AI can Do
- 2. A walk through the syllabus

A Goal of Al

Robust, **fully autonomous** agents in the real world

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How?

A Goal of Al

Robust, **fully autonomous** agents in the real world

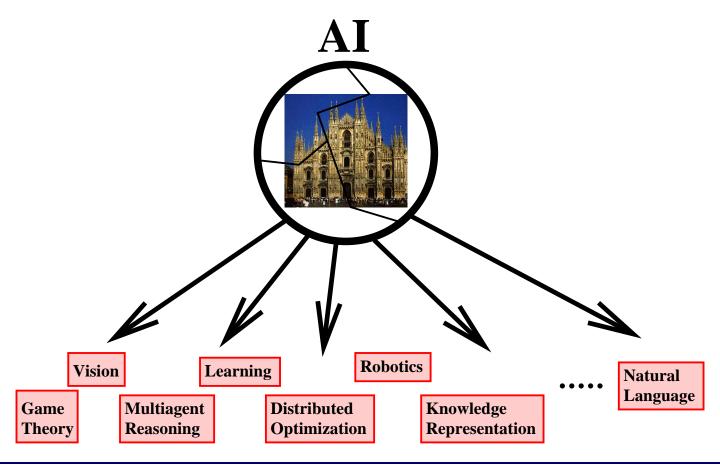
How?

Bottom-Up Metaphor

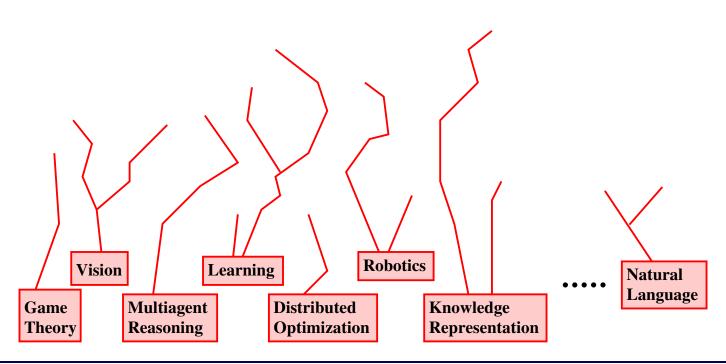
Russell, '95

"Theoreticians can produce the AI equivalent of bricks, beams, and mortar with which AI architects can build the equivalent of cathedrals."

Dividing the Problem

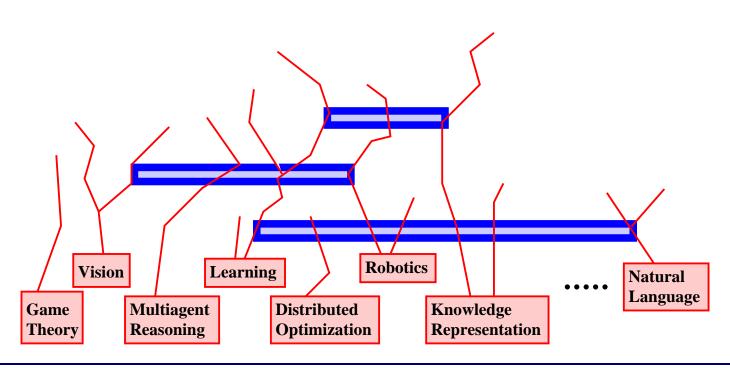


The Bricks

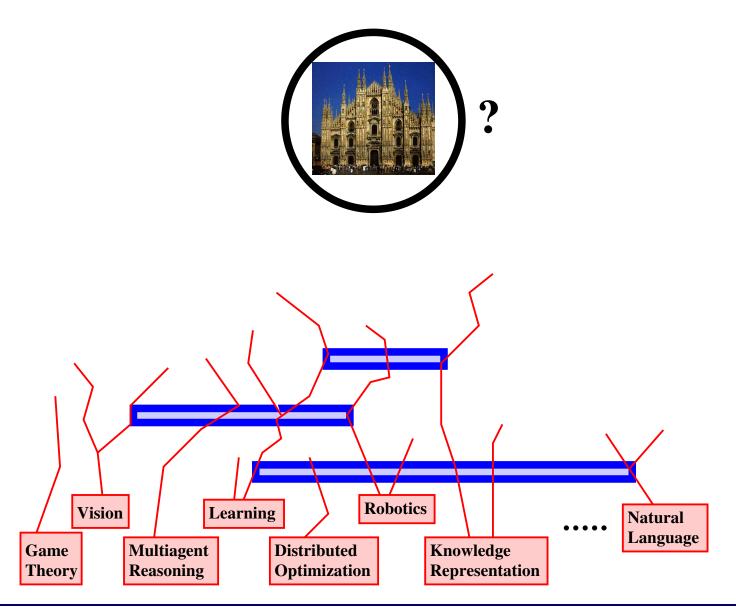




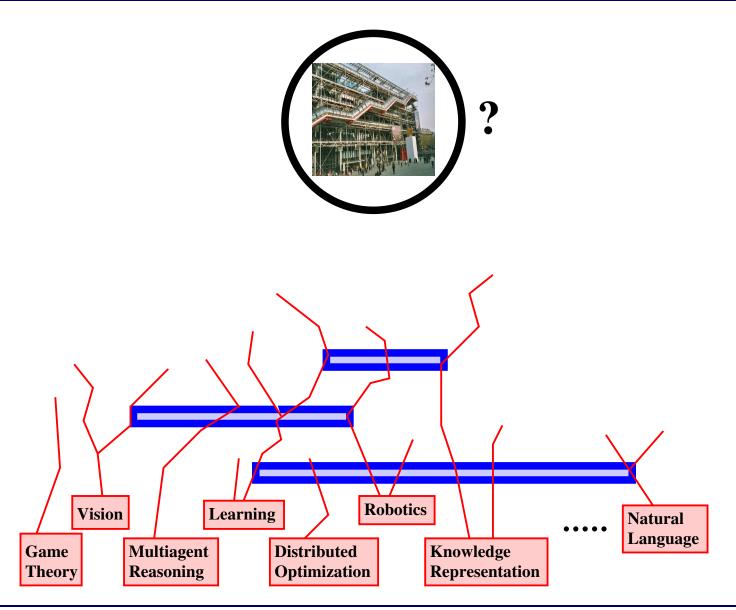
The Beams and Mortar



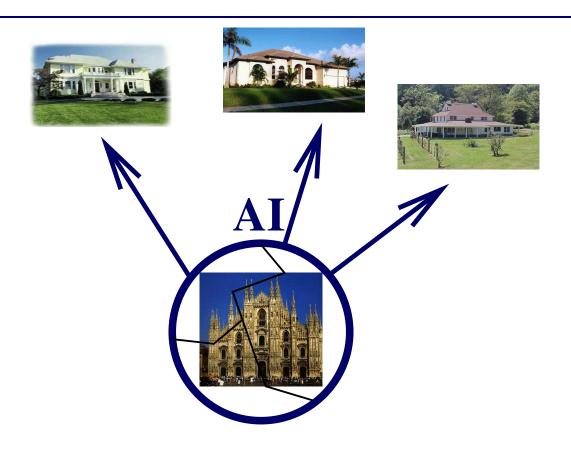
Towards a Cathedral?



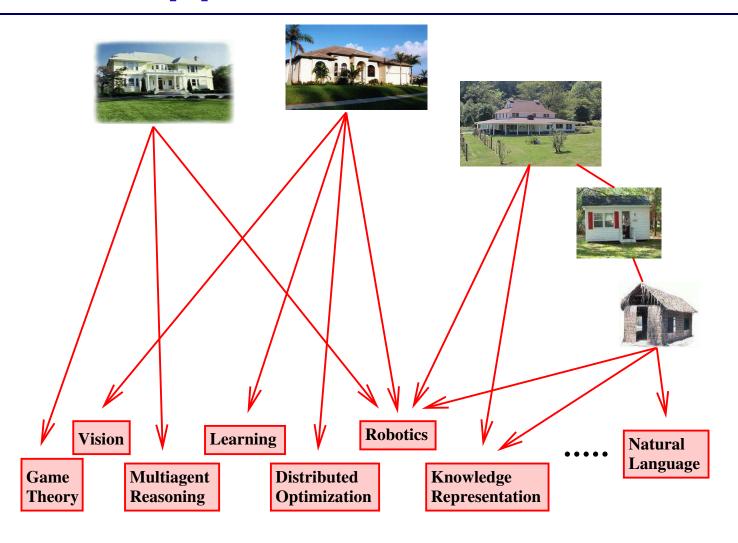
Or Something Else?



A Different Problem Division

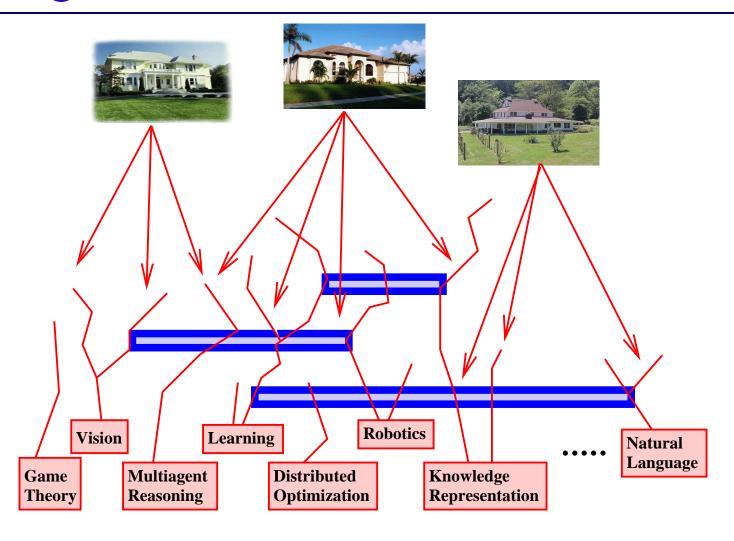


Top-Down Approach



"Good problems ... produce good science" [Cohen, '04]

Meeting in the Middle



Good Problems Produce Good Science

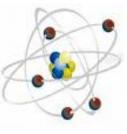
Manned flight



Apollo mission



Manhattan project



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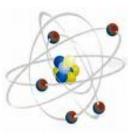
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RoboCup soccer

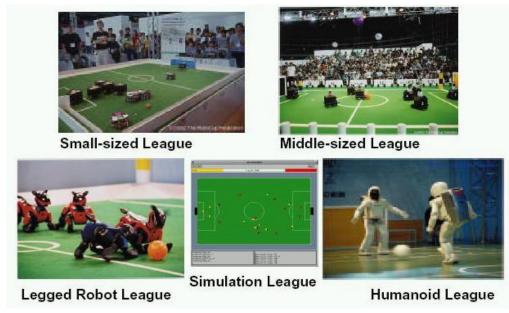


Goal: By the year 2050, a team of humanoid robots that can beat the human World Cup champion team.

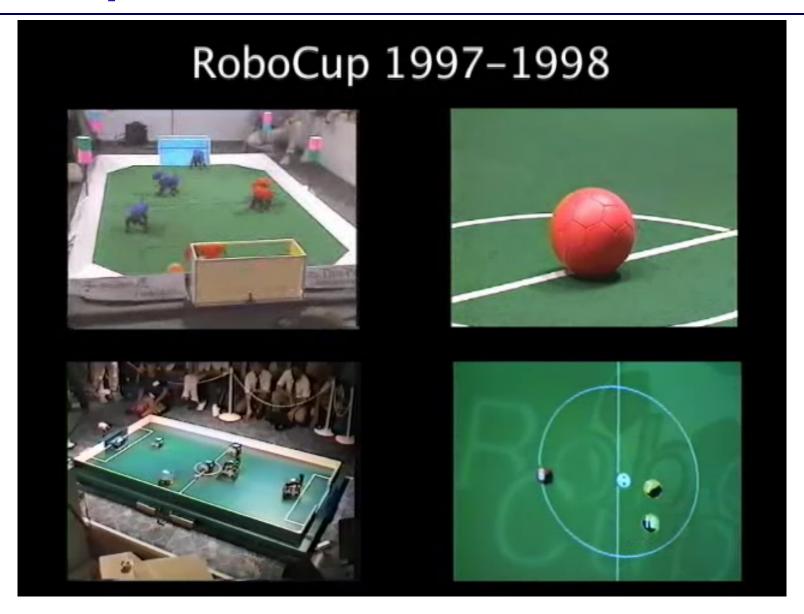
[Kitano, '97]

RoboCup Soccer

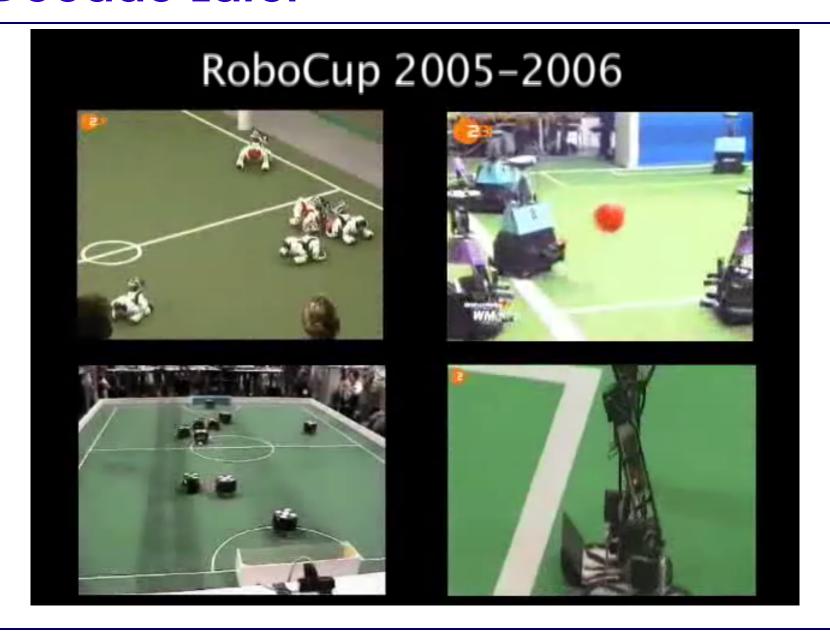
- Still in the early stages
- Many virtues:
 - Incremental challenges, closed loop at each stage
 - Relatively easy entry
 - Multiple robots possible
 - Inspiring to many
- Visible progress



The Early Years



A Decade Later

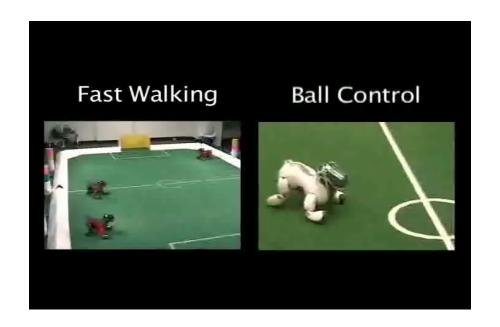


Learning in RoboCup

1999 Champion Simulation team

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Learning in RoboCup

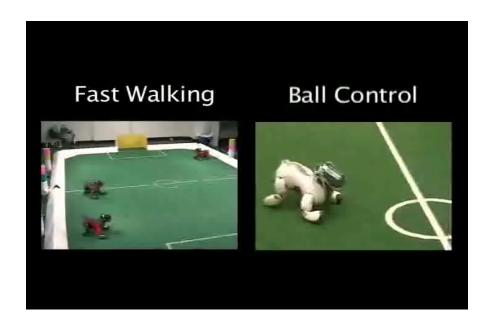
1999 Champion Simulation team



2011 Champion Simulation team

Learning in RoboCup

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2011 Champion Simulation team

Initial walk, During learning, Final walk, Game highlight

- Computer vision
 - Shape modeling, object recognition, face detection...
- Robot vision



Mobile camera, limited computation, color features

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- Mobile camera, limited computation, color features
- Object detection in real-time, on-board a robot



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Other Good AI Challenges





Autonomous vehicles



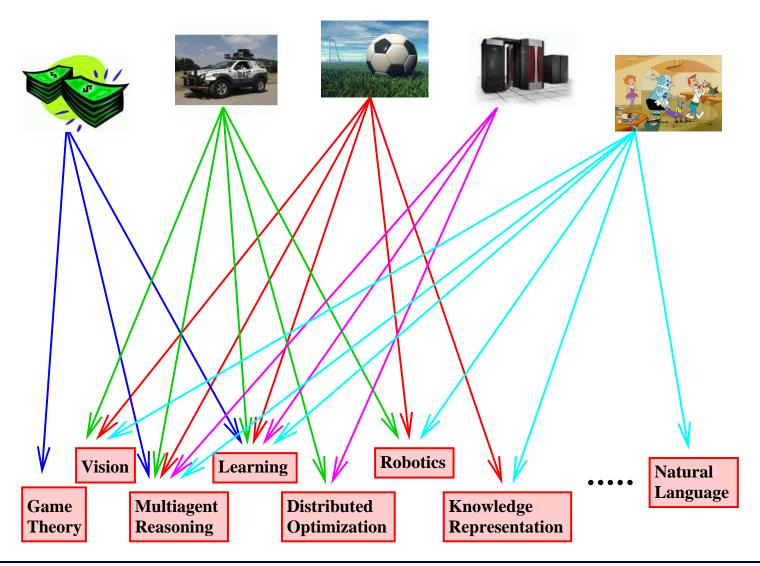
Autonomic computing



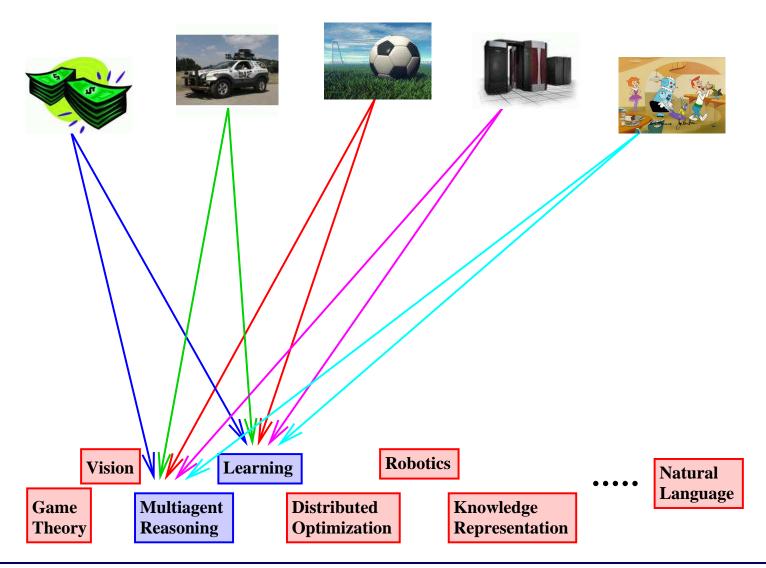
Socially assistive robots



Challenge Problems Drive Research



Learning and Multiagent Reasoning





Machine Learning

- Backgammon [Tesauro, '94]
- Helicopter control [Ng et al., '03]





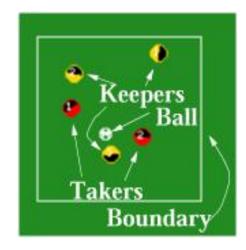
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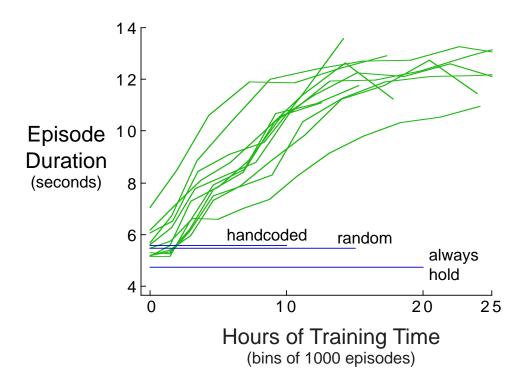


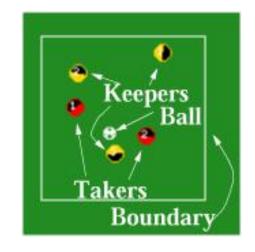


• RoboCup Soccer Keepaway [Stone & Sutton, '01]



After Learning



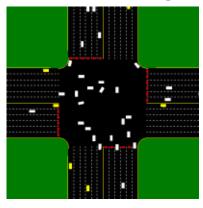


- Once there is one, there will soon be many
- To coexist, agents need to interact

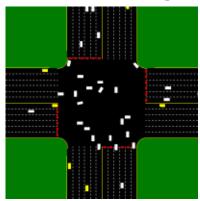
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Autonomous Bidding Agents

ATTac: champion travel agent



- Learns model of auction closing prices from past data
- Novel algorithm for conditional density estimation

TacTex: champion SCM agent



- Adapts procurement strategy based on recent data
- Predictive planning and scheduling algorithms

TacTex'09: champion Ad-Auctions agent

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- Watson wins at Jeopardy

Ethics/Implications

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What happens when we achieve this goal

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A Walk through the Syllabus

Official syllabus is on-line

Readings due at least once per week

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Brief written responses for every reading

10%

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Class participation

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Assignments (mostly programming)
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Pacman! Including tournament



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- Final 25%

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