

CS10: The Beauty and Joy of Computing

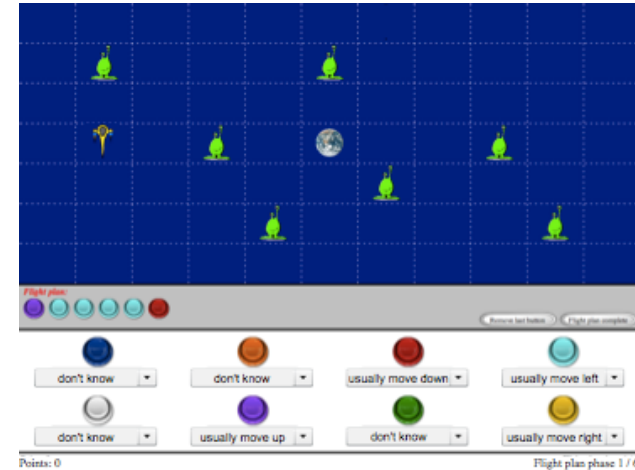
Artificial Intelligence



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(Slides adapted from Dan Garcia)
3 April 2013

What I Do...

- Model human learning using machine learning
- Adaptive instruction and feedback in computer-based educational environments
- E.g., diagnose a student's knowledge by watching her play a game



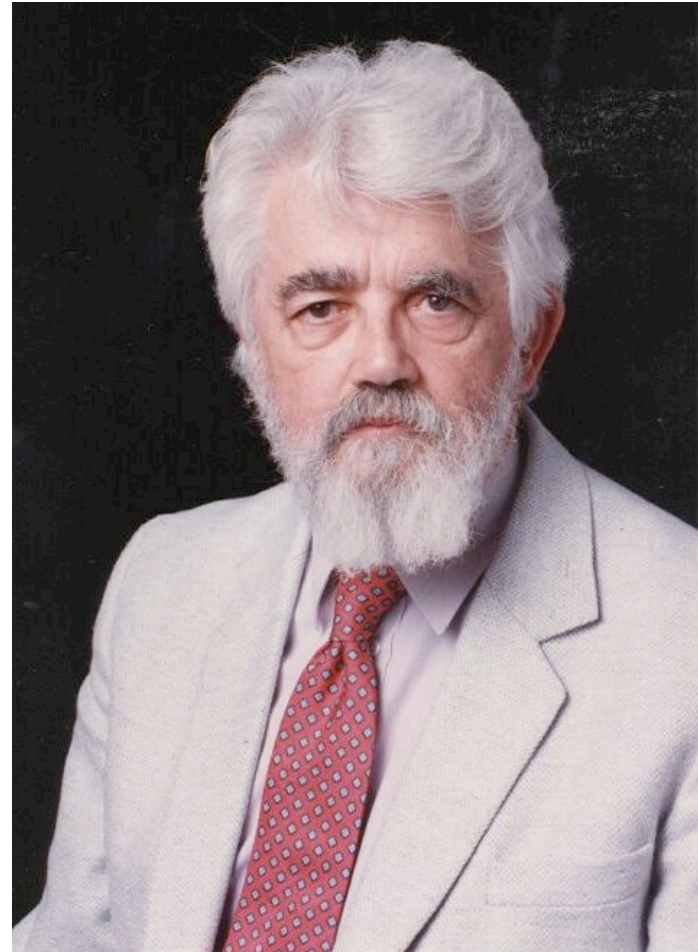
Lecture Overview

- Definition
- Some AI history
- Tour of areas of AI
- Turing Test and the Chinese room



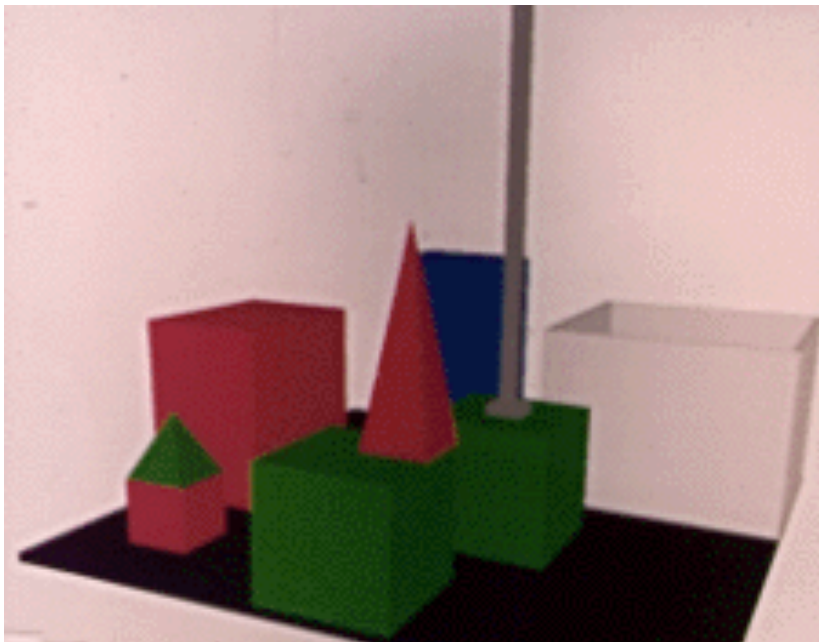
AI Definition by John McCarthy

- “Getting a computer to do things which, when done by people, are said to involve intelligence”
- Finesses the idea of whether a computer has consciousness, whether they have rights, etc.



A little history...

- Early AI (1956-early 1970s): symbolic reasoning and lots of optimism
- Neural nets (but very simple)



Person: PICK UP A BIG RED BLOCK.

Computer: OK. (does it)

Person: GRASP THE PYRAMID.

Computer: I DON'T UNDERSTAND WHICH PYRAMID YOU MEAN.

Person: FIND A BLOCK WHICH IS TALLER THAN THE ONE YOU ARE HOLDING AND PUT IT INTO THE BOX.

Computer: BY "IT", I ASSUME YOU MEAN THE BLOCK WHICH IS TALLER THAN THE ONE I AM HOLDING.

Computer: OK. (does it)



Clicker Question

- Which of these rules is true for all dogs?
 - (A) Has four legs
 - (B) Has fur
 - (C) Barks
 - (D) None of the above



(Image from: http://vision.stanford.edu/resources_links.html)

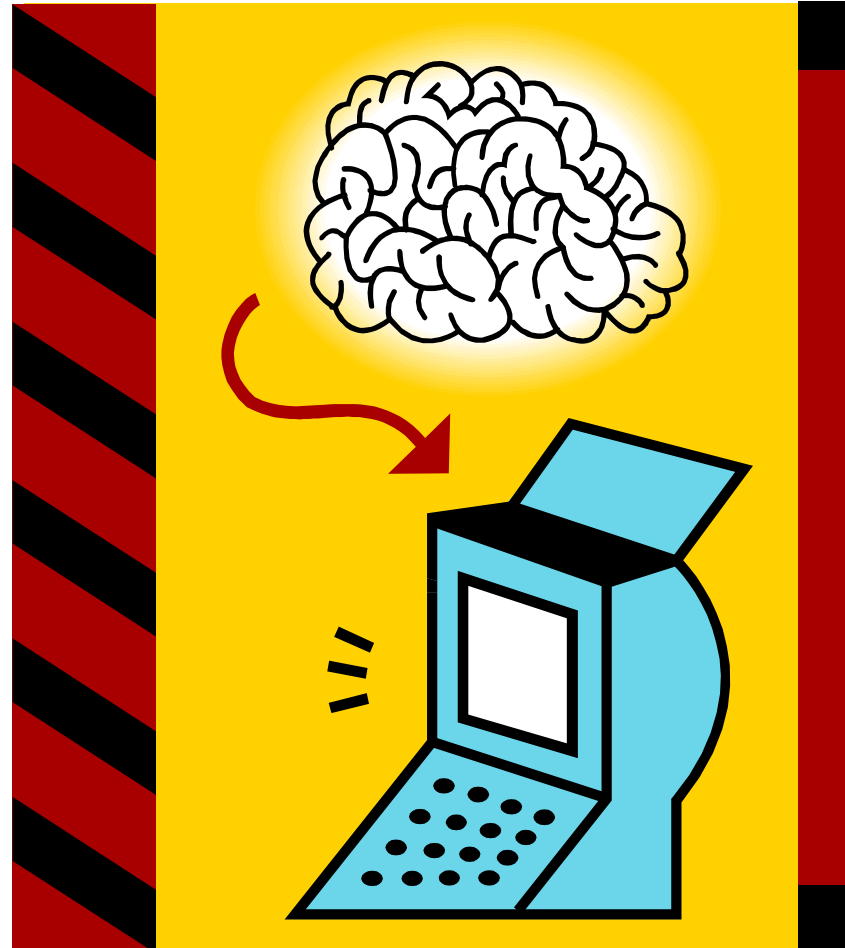
Revival of AI

- Brittle rules break down with complexity of real world
- Probability and uncertainty
- No “dog rule” – instead: what is the probability that the thing we’re seeing is a dog?



What intelligent things do people do?

- Planning
- (Machine) Learning
- Natural Language Processing
- Motion and manipulation
- Perception
- Creativity
- General Intelligence



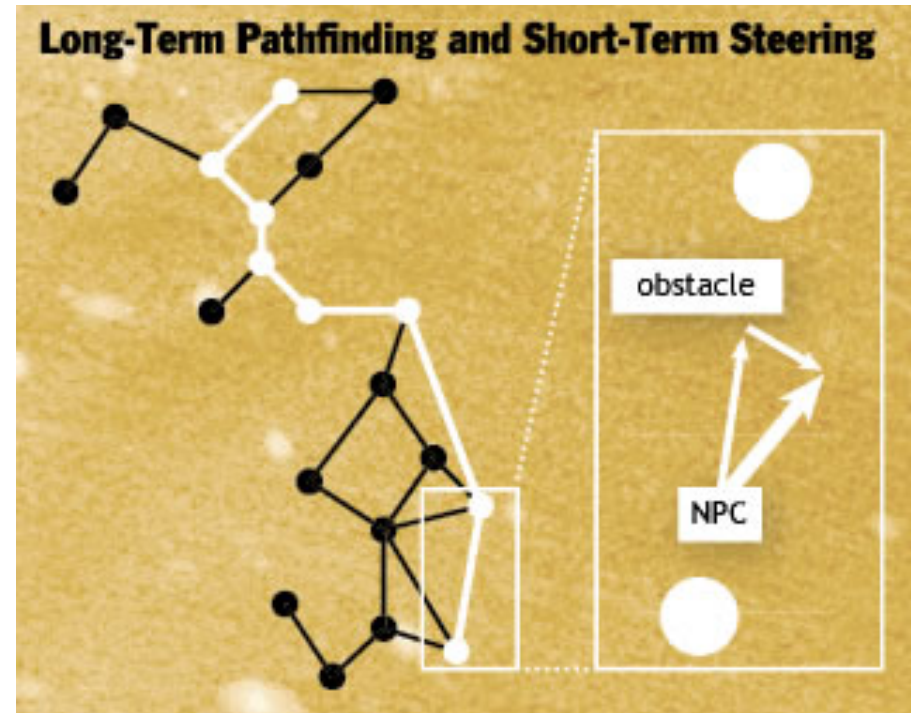
Tour of AI Applications

- Questions to keep in mind:
 - How would you evaluate how well a machine performed on the tasks we talk about?
 - How can blending artificial and human intelligence make tasks simpler, even if the AI isn't perfect?



Planning

- Range of intelligence
 - Low: simple heuristics
 - Medium: pathfinding
 - High: Learns from player
- Dynamic difficulty - adjust to player's skill
- Allocation of resources
 - E.g., choose what land resources to give to which conservation projects



www.businessweek.com/innovate/content/aug2008/id20080820_123140.htm
en.wikipedia.org/wiki/Dynamic_game_difficulty_balancing
en.wikipedia.org/wiki/Game_artificial_intelligence
queue.acm.org/detail.cfm?id=971593



Machine Learning

- “A program learns if, after an experience, it performs better”
- What are the right generalizations to make given the data you’ve seen and the task you’re completing?



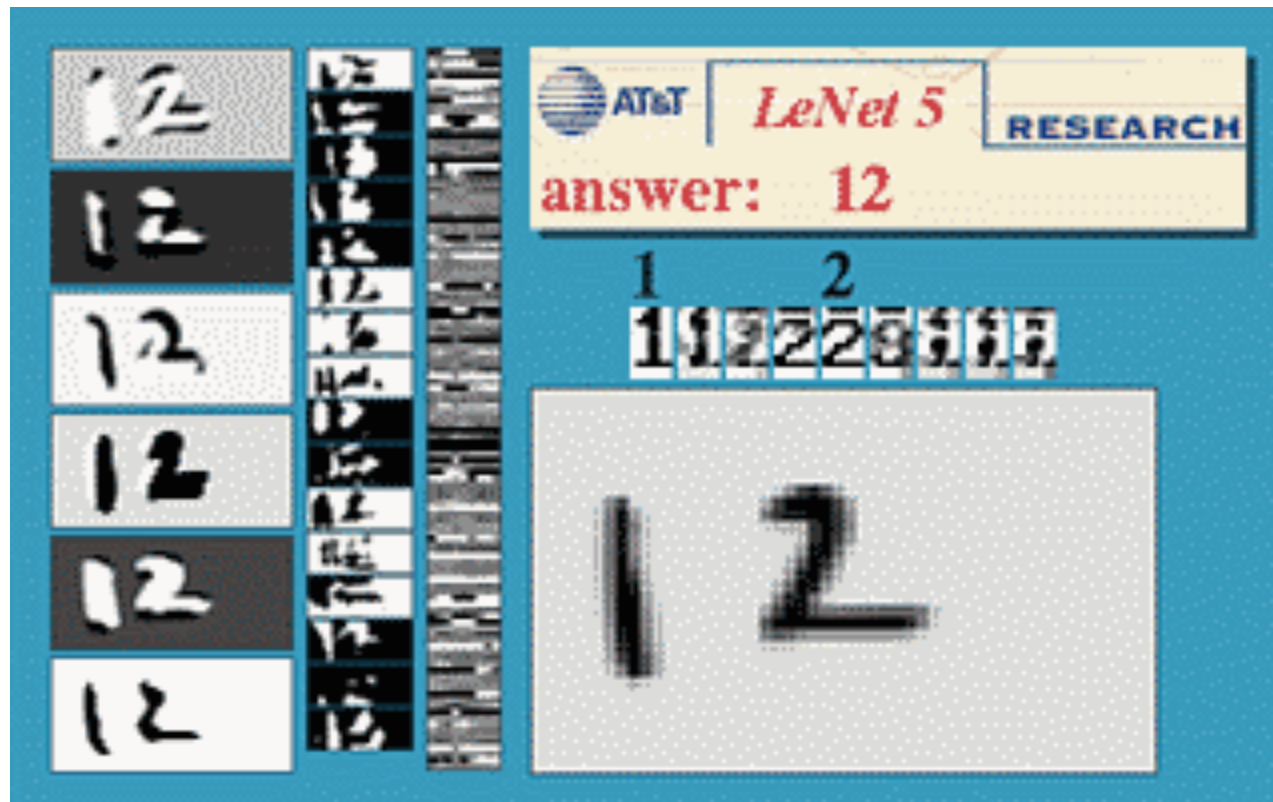
Machine Learning

- Algorithm Types
 - Supervised learning
 - Give a system input & output training data, and it produces a classifier
 - Unsupervised learning
 - Determine how data is organized or clustered
 - Reinforcement learning
 - No training data, real-time corrections adjust behavior



Example: Deep Learning

- Combines supervised and unsupervised learning:
Learn the right *representations* for input -> output



Benefiting from Big Data



Translation



Computer vision

Netflix Prize

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Leaderboard

Display top leaders.

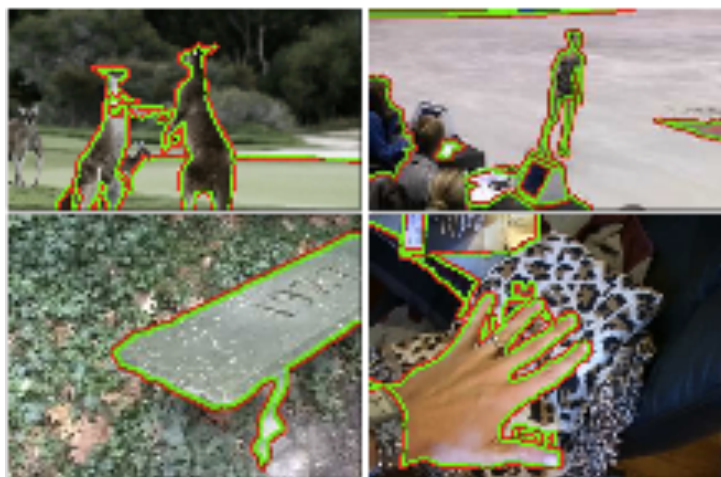
Rank	Team Name	Best Score	% Improvement	Last Submit Time
1	BellKor's Pragmatic Chaos	0.8558	10.05	2009-06-26 18:42:37
Grand Prize - RMSE <= 0.8563				
2	PragmaticTheory	0.8582	9.80	2009-06-25 22:15:51
3	BellKor in BigChaos	0.8590	9.71	2009-05-13 08:14:09
4	Grand Prize Team	0.8593	9.68	2009-06-12 08:20:24
5	Dace	0.8604	9.56	2009-04-22 05:57:03
6	BigChaos	0.8613	9.47	2009-06-23 23:06:52
Progress Prize 2008 - RMSE = 0.8616 - Winning Team: BellKor in BigChaos				
7	BellKor	0.8620	9.40	2009-06-24 07:16:02
8	Gravity	0.8634	9.25	2009-04-22 18:31:32
9	Opera Solutions	0.8638	9.21	2009-06-22 05:53:30
10	xlvector	0.8639	9.20	2009-06-26 13:49:04
11	xiangliang	0.8639	9.20	2009-06-26 07:47:34
12	BruceDengDaoCiyiYou	0.8641	9.18	2009-06-02 17:08:31
13	Ces	0.8642	9.17	2009-06-24 14:34:14
14	majia2	0.8642	9.17	2009-06-23 08:07:50

Recommendation



Vision

- Tasks related to understanding images/camera input



Figure/ground
segmentation



Pedestrian detection



Action
recognition

phoning



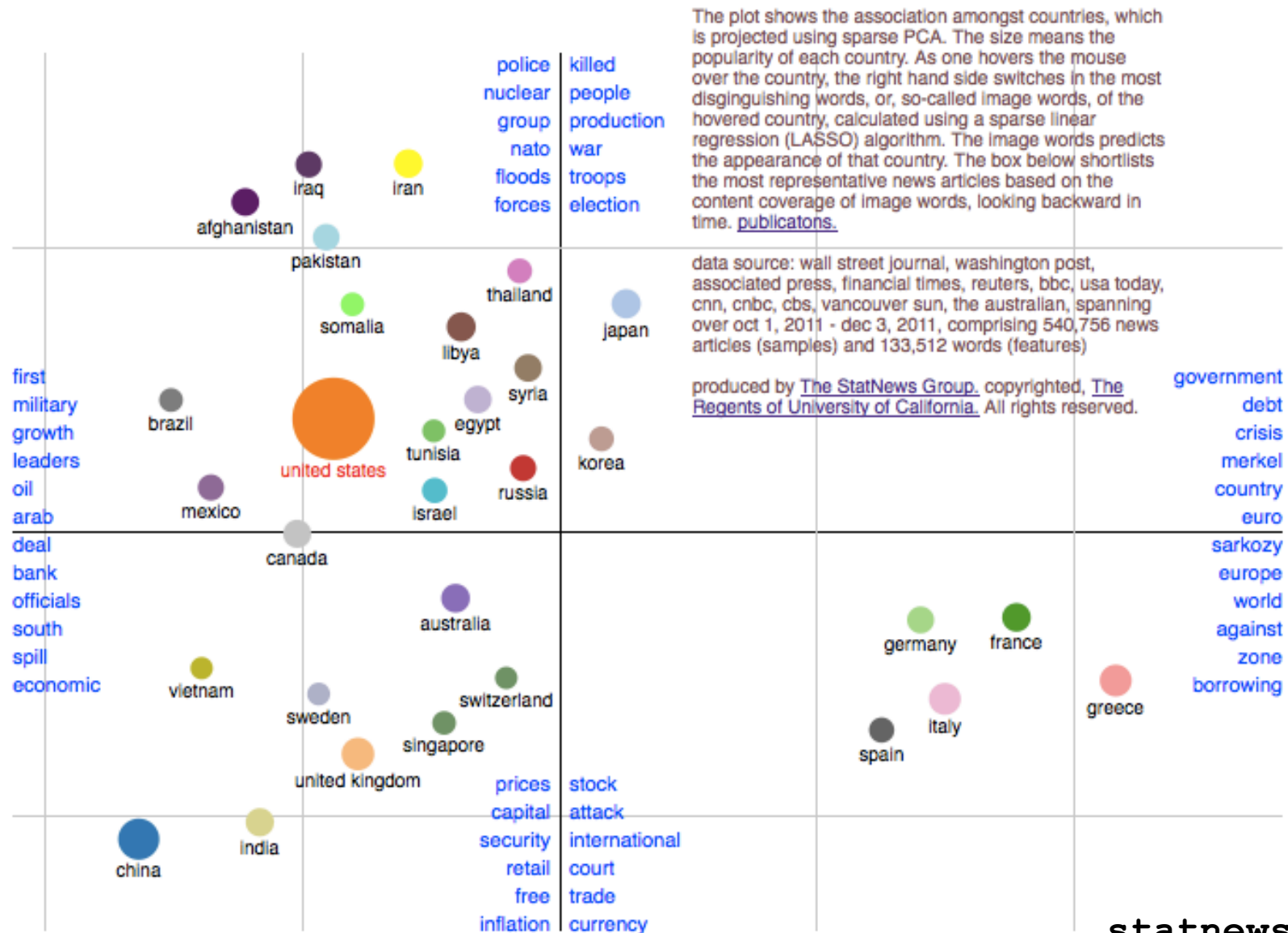
(Some images from Berkeley vision group)

Natural Language Processing

- Known as “AI-complete” problem
 - (Often) requires extensive knowledge of world
- Statistical NLP
 - Correcting/guessing text
 - Suggesting news stories
 - Finding articles that are similar to one another
 - Translate or paraphrase texts

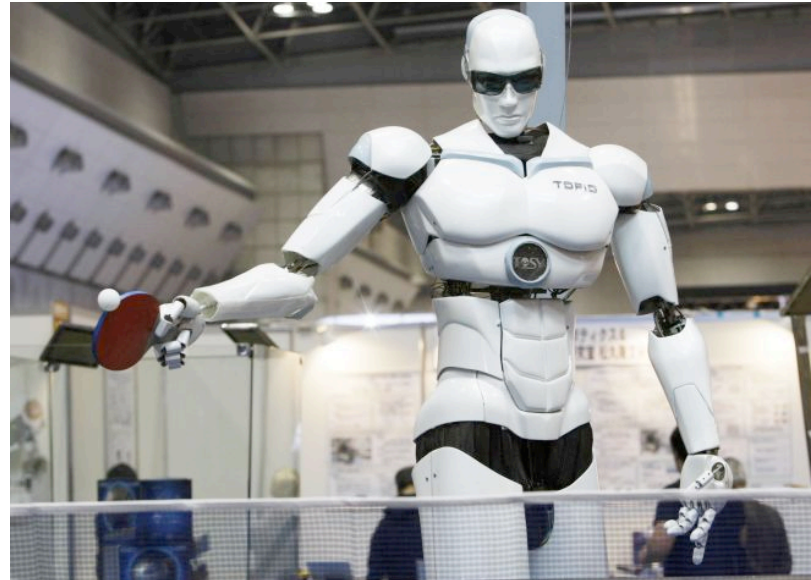


Unsupervised Learning Example



Robotics

- For many, the coolest and scariest part of AI
- Combines fields of AI/CS
 - Speech recognition
 - Synthetic voice
 - Machine vision
 - Planning
 - HCI

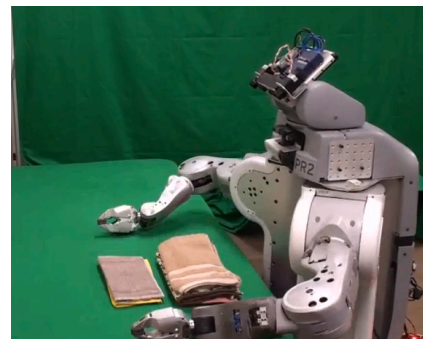


TOPIO, the ping-pong playing robot



Assistive robots

Surgical robots



UC Berkeley's towel-folder



Autonomous helicopter



en.wikipedia.org/wiki/Robotics

Recap

- All of these applications are tough because they require:
 - Knowing about context
 - Uncertainty about input
 - Intensive computations
- But AI has been relatively successful at making progress (and in some cases, better than people!)



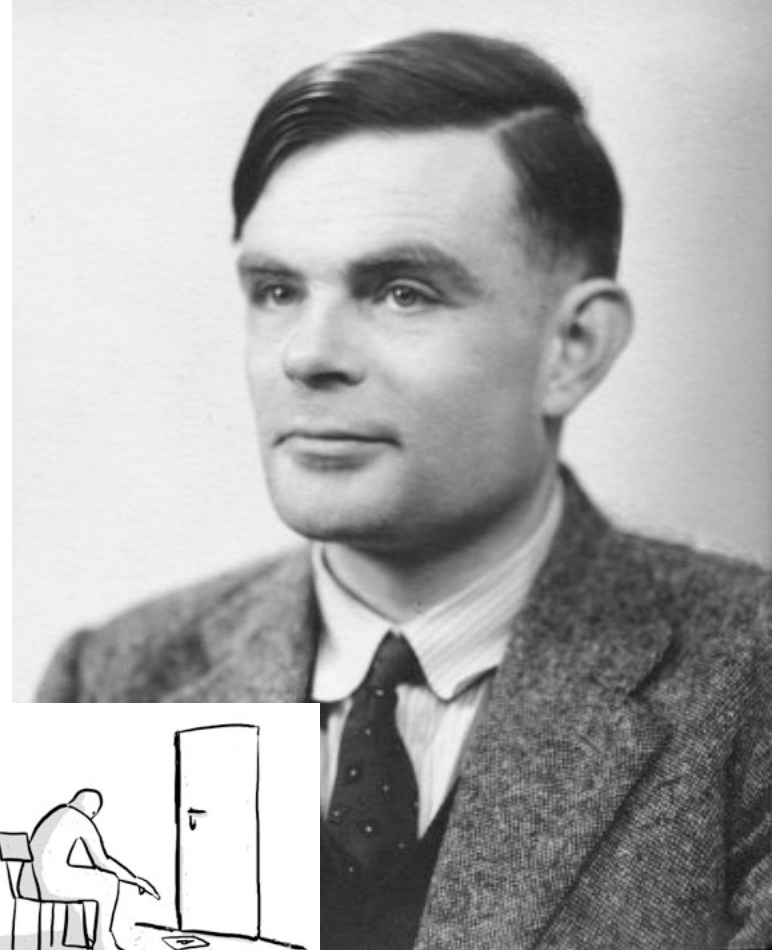
Clicker Question

- What would a “truly intelligent” AI system look like?
 - (A) Behaves in an optimal or rational manner
 - (B) Behaves similarly to people – when it makes errors, those errors are similar to people’s errors
 - (C) Carries out the same type of processing (mental representations) people do – i.e., thinks like people



Turing Test for Intelligence

- In 1950, Turing defined a test of whether a machine could “think”
- “A human judge engages in a natural language conversation with one human and one machine, each of which tries to appear human. If judge can’t tell, machine passes the Turing test”
- John Searle argued against the test via the Chinese room experiment, in which someone carries on a conversation by looking up phrases in a book. Does that person understand Chinese?



en.wikipedia.org/wiki/Turing_test



Clicker Question

- How would you respond to Searle's Chinese room experiment?
 - (A) The system as a whole understands Chinese
 - (B) The man doesn't understand Chinese, but if he had a way to connect with the outside world (rather than just receiving strings of symbols), he could understand Chinese
 - (C) We must be missing something about "understanding" since the argument implies that brains, which are collections of neurons, cannot understand



Summary

- AI systems excel in things computers are good at
 - Big data (using web to parse language)
 - Constrained worlds (chess, math)
- It's getting better at...
 - Language understanding
 - Real-time robotics
- Lots more applications that I didn't have time to talk about!
- CS188: Artificial Intelligence
 - One of the most popular courses on campus!
- CogSci131: Computational Models of Cognition



Thanks! Feel free to email me with questions at rafferty@cs.berkeley.edu

