Machine Intelligence

Lecture 1: Introduction to MI and Agents

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Course Organization

Teacher

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Literature

D. Poole and A. Mackworth: *Artificial Intelligence. Foundations of Computational Agents* (2nd edition)

http://artint.info/

Course Homepage

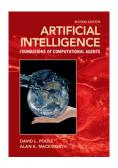
Can be found under Moodle

Times

- Mondays, 8.15-12.00 (typically Frb 7H, Aud): Exercises 8.15-10.00, Lecture 10.15-12.00
- Wednesdays 8.15–12.00 (group rooms): Extended exercise sessions

Exam

Exam in January



Several reasons, but:

We decided that it is better to clearly explain the foundations upon which more sophisticated techniques can be built, rather than present these more sophisticated techniques. This means that a larger gap may exist between what is covered in this book and the frontier of science. But it also means that the student will have a better foundation to understand current and future research.

Tentative course overview

Topics:

- Introduction
- Problem solving as search
- Constrained satisfaction problems
- Logic-based knowledge representation
- Representing domains endowed with uncertainty.
- Bayesian networks
- Machine learning
- Planning
- Reinforcement learning
- Multi-agent systems

Learning goals

After having followed the course you should

- have knowledge about basic techniques within the field of machine intelligence and computational agents.
- be able to apply key machine intelligence techniques to a specific problem domain.
- be able to reason about computational agents that can operate in domains of varying complexity.

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... and be well-prepared for the aMI course and the machine intelligence specialization!

ΑI

The Turing Test



A.M. Turing: "Computing Machinery and Intelligence", Mind Vol.59 (1950). Proposes an *imitation game*, which (slightly modified) has become known as the *Turing test*:

An *interrogator* is connected via one terminal to a real person, and by another terminal to a computer (both in another room). The interrogator does not know which terminal is connected to the machine. He or she can perform on both terminals a (natural language) dialogue with whatever is at the other end of the line. The machine passes the Turing test, if the interrogator is not able to identify, which terminal is connected to the machine.

⇒ The Turing test tests observable behavior, not cognitive processes!



Achievements: Deep Blue

- 30 IBM RS/6000 processors
- 480 custom chess processors
- able to examine 200 million moves per second
- database of 700.000 grandmaster games

endgame database (covering all 5 piece positions)

Results:

1996: Kasparov 4 Deep Blue 2 1997: Kasparov 2.5 Deep Blue 3.5

MI Autumn 2019 AI

... two decades later

AlphaGo was developed by Google Deepmind to play Go. In March 2016 it payed a five game match against Lee Sedol with a final score of 4-1 in favor of AlphaGo.

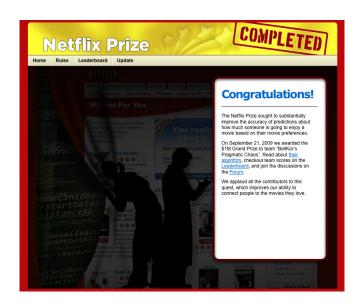


AlphaGo combines Monte Carlo tree search with (deep) artificial neural networks.

Recommender systems

Systems for recommending items that users are likely to find interesting





Jeopardy!: Watson

In 2011, Watson beat Brad Rutter, the biggest all-time money winner on Jeopardy!, and Ken Jennings, the record holder for the longest championship streak (75 days)



- Application of natural language processing, information retrieval, knowledge representation and reasoning, and machine learning
- Made up of a cluster of 90 IBM Power 750 servers with a total of 2880 POWER7 processor cores and 16 Terabytes of RAM. Each Power 750 server uses a 3.5 GHz POWER7 eight core processor, with four threads per core.

Taken from Wikipedia, August 25th, 2011

Jeopardy!: Watson

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New developments Watson has since evolved into a more general purpose cognitive platform (using natural language processing and machine learning) with applications in e.g. health care.



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Computer games

Al in computer games (e.g. Starcraft)





http://eis.ucsc.edu/StarCraftAICompetition

Computer games

Al in computer games (e.g. Starcraft)





This is mad. Over on StarCraft Z forum Teamliquid, a poster who goes by Lomilar has been talking about, a program he's coded cade EvolutionChamber. It uses generic agriorithms to find powerful build orders, meaning his program takes a population of build orders, list off the useless ones, and has the most successful ones reproduce asexually to create a new population, which tests listelf again, and so on. I'm taking all this from this blog post by programmer Louis Brandy, wherein he breaks down what Lomilar's done to hat lay folk can understand if use.

EvolutionChamber's already come up with one ludicrous build order, which I've posted beneath the jump.

- 10 extractor-trick to 11 11 overlord
- 11 spawning pool 15 extractor
- 16 queen (stop drones here)
- 18 overlord 18 roach warren
- 17 overlord (yes, two)
- spawn-larva on queen when she pops

roach x7

AIIDE 2011 StarCraft Competition

http://eis.ucsc.edu/StarCraftAICompetition

Computer games

The TrueSkill ranking system is a skill based ranking system for Xbox Live developed at Microsoft Research.



https://research.microsoft.com/en-us/projects/trueskill/

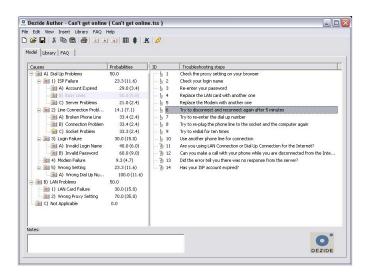


Figure borrowed from Dezide's homepage (www.deczide.com)

Automatic Translation

Achievements: Automatic Translation

Google translate:



Autonomous driving

Achievement: DARPA grand challenge 2005

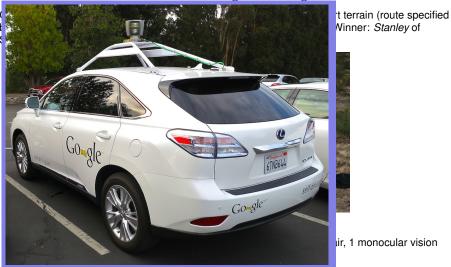
Competition for autonomous vehicles: navigate 132 miles through desert terrain (route specified by approx. 3000 "waypoints"). 5 out of 23 vehicles completed the task. Winner: *Stanley* of Stanford Racing Team in 6h 53m (19.2 mph).



- 7 Pentium M computers
- Sensors: 4 laser range finders, 1 radar system, 1 stereo camera pair, 1 monocular vision system, GPS, inertial measurement unit, wheel speed.

Autonomous driving

Achievement: DARPA grand challenge 2005



June 2016: approx. 2.8 mill. km. autonomous driving.

Image analysis/processing

Colorization of images



Zhang, Isola, Efros. Colorful Image Colorization. In ECCV, 2016.

Image captioning



"man in black shirt is playing guitar."



'construction worker in orange safety vest is working on road."



"two young girls are playing with lego toy."



"boy is doing backflip on wakeboard."



"girl in pink dress is jumping in



"black and white dog jumps over bar."



young girl in pink shirt is swinging on swing."



"man in blue wetsuit is surfing on wave."

Andrej Karpathy, Li Fei-Fei, Deep Visual-Semantic Alignments for Generating Image Descriptions, CVPR 2015

Other application areas

- Medical diagnosis and advisory systems
- Information processing and filtering,
- Display of information for time-critical decisions
- Spam filtering
- Optical character recognition
- Profiling/Credit scoring: profiling customers
- Bioinformatics
- Real estate: Prediction of house prices
- Computer networks: intrusion detection
- Alert and monitoring systems
- Speech recognition
- Face recognition, image annotation
- Action recognition (in video sequences)
- ...

Towards the Turing Test ...

Turing Test (Loebner Competition)

- Loebner Competition: (Non-scientific) competition for computer systems performing under Turing Test conditions.
- http://www.pandorabots.com/pandora/talk?botid=f5d922d97e345aa1

Online help

- Combine natural language interface with expert knowledge.
- Restricted Domain (Geography): CHAT-80 (1982)
- Broad Domain ("all factual knowledge"): Wolfram Alpha http://www.wolframalpha.com/

Wolfram Alpha: Reality

Some queries to try:

Examples by Topic

What can you ask Wolfram|Alpha about?







Word Properties - Dictionary Lookup · Word Puzzles Anagrams · Languages · Document Length · Morse Code · Soundex · Number Names · Character Encodings · ...



People · Genealogy · Names · Occupations · Political Leaders · Historical Events · Historical Periods · Historical Countries Historical Numerals · Historical US Money · Inventions · ...



Conversions - Calculations -Comparisons - Dimensional Analysis - Industrial & Construction -Batteries · Bulk Materials · Paint · Freight Containers - Display Formats · Ring Sizes · Shoe Sizes ·







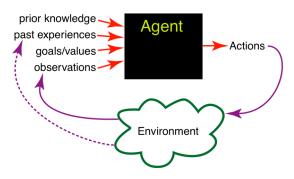
Automatic Analysis · Statistical Analysis · Time Series Analysis · Geographic Data · Data Visualization · ...

The Agent view of Al

Al and Agents

Al is the field that studies the synthesis and analysis of computational agents that act intelligently. [PM, p.3]

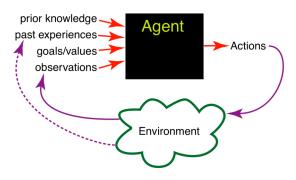
A coupling of perception, reasoning, and acting comprises an agent. [PM, p.10]



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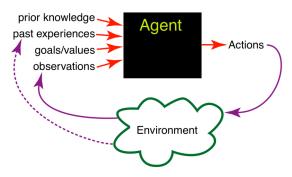
Some special flavors:

- Autonomous agents
- Intelligent agents
- Software agents
- Multi-agent systems

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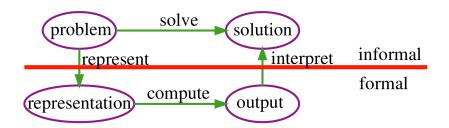
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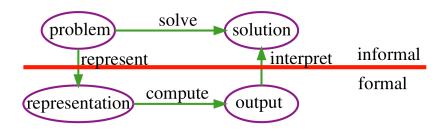
What is not an agent?

- ullet perception \sim input
- $\bullet \ \ \text{reasoning} \sim \text{computation} \\$
- $\bullet \ \ \text{acting} \sim \text{output} \\$
- "agent" a design metaphor, not a strict technical concept

21

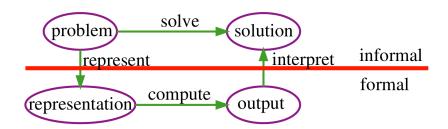
Represent and Compute





A representation should be:

- Sufficiently rich to encode the required knowledge.
- Be "close" to the problem.
- Amenable to efficient computation.
- Able to be acquired from people, data, or experience.

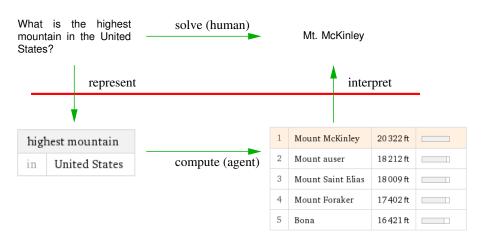


A representation should be:

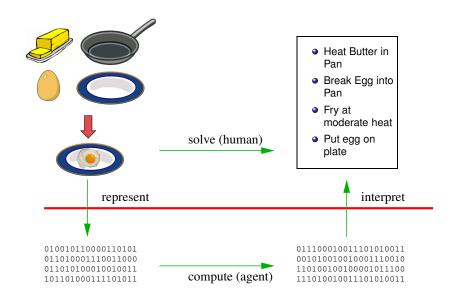
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Questions to be considered:

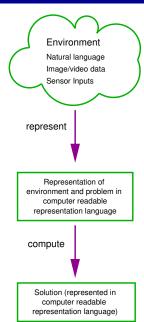
- What is a solution and how good should it be (optimal, satisficing, approximately, probable)?
- How can the problem be represented?
- How can an output be computed (what properties should a solution have)?

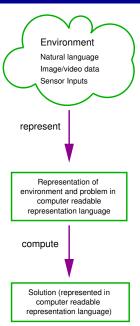


Represent and Compute

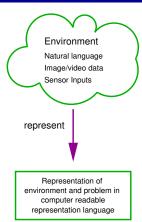


Problem Formalization





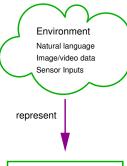
2. What formal representation languages can be used?



- **1.** How to interpret input/observations from environment?
- 2. What formal representation languages can be used?
- **3.** How to solve problems in the given representation language?

Solution (represented in computer readable representation language)

compute



Representation of environment and problem in computer readable representation language



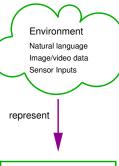
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Natural Language Processing
Computer Vision
...

Knowledge Representation

Problem Solving Automated Reasoning



Representation of environment and problem in computer readable representation language



Solution (represented in computer readable representation language)

- **1.** How to interpret input/observations from environment?
- 2. What formal representation languages can be used?
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Knowledge Representation

Problem Solving Automated Reasoning

In this course the focus is on 2, and 3.1

Representing the problem

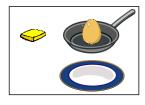
Levels of Representation

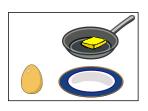
We consider 3 representation schemes

- State based
- Feature based
- Relational

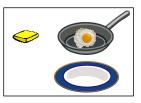
State based



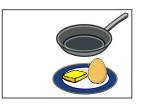




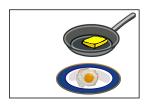
State 01



State 04



State 08

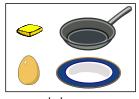


State 12

State 14

State 18

Feature based



egg=whole, butter_in=table, egg_in=table



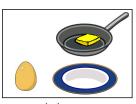
egg=broken, butter_in=table, egg_in=pan



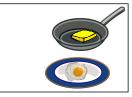
egg=whole, butter_in=table, egg_in=pan



egg=whole, butter_in=plate, egg_in=plate



egg=whole, butter_in=pan, egg_in=table



egg=broken, butter_in=pan, egg_in=plate

30 binary features represent $2^{30} = 1.073.741.824$ states.

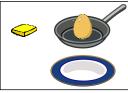
Relational



state(egg,whole), in(butter,table), in(egg,table)



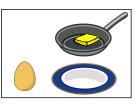
state(egg,broken), in(butter,table), in(egg,pan)



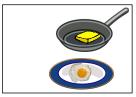
state(egg,whole), in(butter,table), in(egg,pan)



state(egg,whole), in(butter,plate), in(egg,plate)



state(egg,whole),in(butter,pan), in(egg,table)



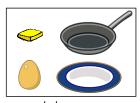
state(egg,broken), in(butter,pan), in(egg,plate)

29

1 binary relation and 100 individuals give $100^2 = 10.000$ boolean features, or 2^{10000} states.

Levels of Detail

Do we need to distinguish states



egg=whole, butter_in=table, egg_in=table, butter_position_to_pan=left



egg=whole, butter_in=table, egg_in=pan, butter_position_to_pan=right

- Not at "recipe level"
- Yes at robot control level: "move arm to left, grab butter, ..."
- $\bullet \ \ \, \hbox{$\leadsto$ may need hierarchical description of state space to reason at different levels of abstraction.}$

Modularity

• Flat, modular, hierarchical

Modularity

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Planning horizon

Non-planning, finite, indefinite, infinite

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Domain uncertainty

- Fully observable vs. partially observable world.
- Deterministic vs. stochastic.

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Preferences

- Achievement goals, maintenance goals.
- Complex ordinal or cardinal preferences.

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Number of agents

Single agent or multiple agents

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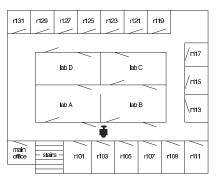
Computational limits

Perfect or bounded rationality

31

Prototypical Applications

Autonomous Delivery Robot



Inputs:

Prior knowledge, past experience, goals, observations

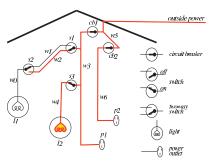
Outputs:

Motor control, speech, video display

Complexity issues:

- Hierarchical decomposition
- Planning horizon
- Goals
- Uncertainty
- ...
- Robot can move, pick up and put down objects
- Receives commands (in natural language)
- Can deliver packages, mail, coffee, ...
- Must interpret commands, develop and execute plans for action

Diagnostic Assistant



Inputs:

Prior knowledge, past experience, goals, observations

Outputs:

Recommendations on treatments and tests

Complexity issues:

- Sensing/effect uncertainty
- Knowledge representation
- Goal specification
- Advise a human about system, e.g. diagnose patient, troubleshoot electrical system, automobile, etc.
- Substitute of human expert

Trading Agent

Characteristics

- Automatically buy/sell goods for user/company (possibly at auction)
- Determine good strategy to procure necessary goods in time at best price

Inputs

Prior knowledge about goods, past experience, preferences, observations

Output

Proposals to the user

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