Artificial Intelligence: An Introduction

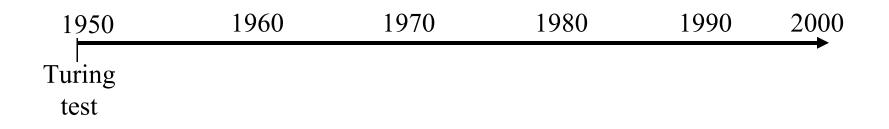
Mohsen Afsharchi

Strong Al

 An autonomous self-moving machine that acts and reasons like a human



- 1950: Alan Turing. The Turing test.
 - Can machines think? → Can we tell it's a machine from conversation?
 - text in / text out
 - demo: A.L.I.C.E. (http://www.alicebot.org/)
 - Turing, A.M. (1950). Computing machinery and intelligence. Mind, 59, 433-460
 - it also contains things like genetic algorithm, human cloning ...

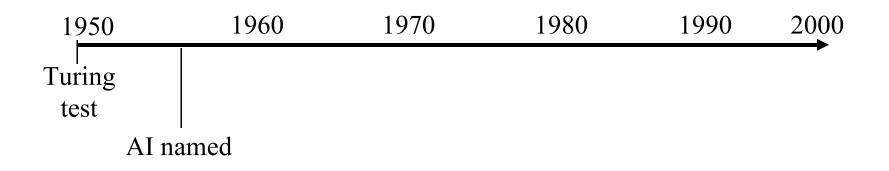




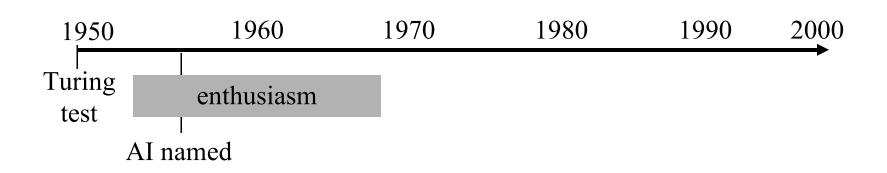
- 1956: Dartmouth summer workshop
 - Al named
 - big players introduced



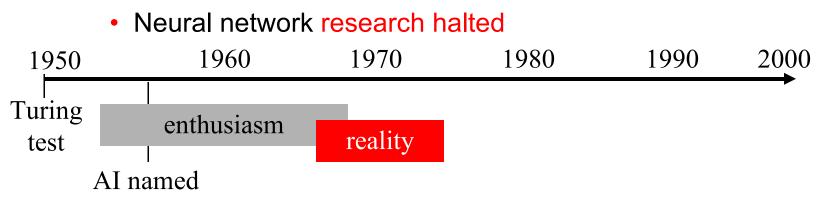
- John McCarthy, Marvin Minsky, Claude Shannon, Nathaniel Rochester, Trenchard More, Arthur Samuel, Ray Solomonoff, Oliver Selfridge, Allen Newell, Herbert Simon
- no consensus



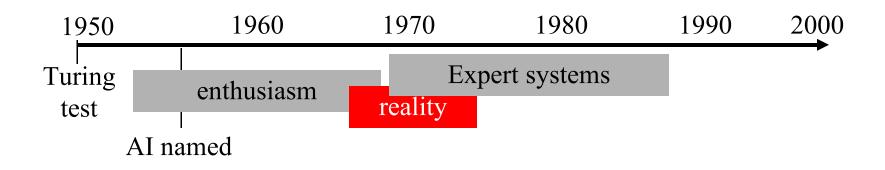
- 1952—1969: early enthusiasm: Computers can do X
 - X = solve puzzles, prove geometry theorems, play checker, Lisp, block world, ELIZA, perceptron...
 - but many are toy problems



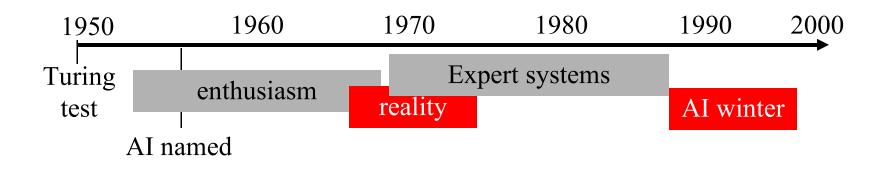
- 1966-1973: a dose of reality
 - syntactic without domain knowledge doesn't work
 - The spirit is willing but the flesh is weak
 - The vodka is good but the meat is rotten (US \rightarrow RU \rightarrow US)
 - US gov canceled funding for machine translation
 - intractability: exponential complexity
 - British gov ended AI support based on the Lighthill report
 - theoretic limit: perceptron can't do XOR



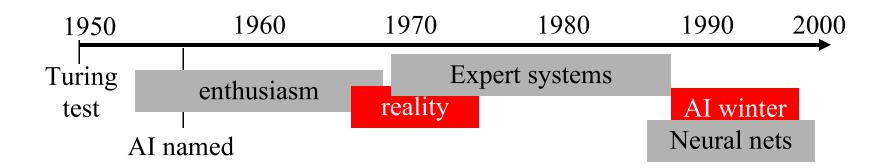
- 1969-1988: Knowledge-based systems
 - Add domain-specific knowledge to guide search
 - CYC: world = millions of rules. (cyc.com)
 - Expert systems commercialized in the 80's
 - One AI group in every major US company
 - Billions of \$\$\$ industry



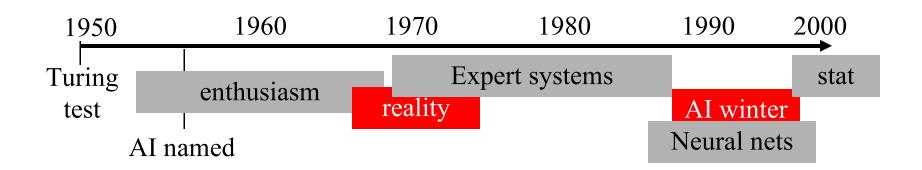
- 1988 not long ago: Al winter
 - Expert systems
 - Massive investment from venture capitalists
 - Extravagant promises
 - Bubble burst
 - AI funding dried up
 - AI companies down



- 1986 not long ago: neural networks
 - Multi-layer perceptron
 - Back propagation training algorithm rediscovered
 - Connectionists vs.
 - Symbolic models (Newell, Simon)
 - Logicist (McCarthy)
 - What it really is: statistical machine learning



- present: statistics
 - machine learning
 - Hidden Markov models (HMM), support vector machines (SVM), Gaussian processes, graphical models (Bayes networks, conditional random fields)
 - data mining



The Reality

• In the early period of AI it seemed plausible that new forms of symbolic computation, e.g., frames and semantic networks, made much of classical theory obsolete. This led to a form of isolationism in which AI became largely separated from the rest of computer science. This isolationism is currently being abandoned. There is a recognition that machine learning should not be isolated from information theory, that uncertain reasoning should not be isolated from stochastic modeling, that search should not be isolated from classical optimization and control, and that automated reasoning should not be isolated from formal methods. David McAllester 1998

Weak Al

- Machines *can* demonstrate intelligence, but do not necessarily have a mind, mental states or consciousness
- weak AI refers to the use of software to study or accomplish specific problem solving or reasoning tasks that do not encompass the full range of human cognitive abilities.

Al today

- Don't know how to do 98% of the intelligent things
- But the rest 2% can do quite well

There's no magic in AI. It's all about optimization, probability, algorithms.

School of Thoughts

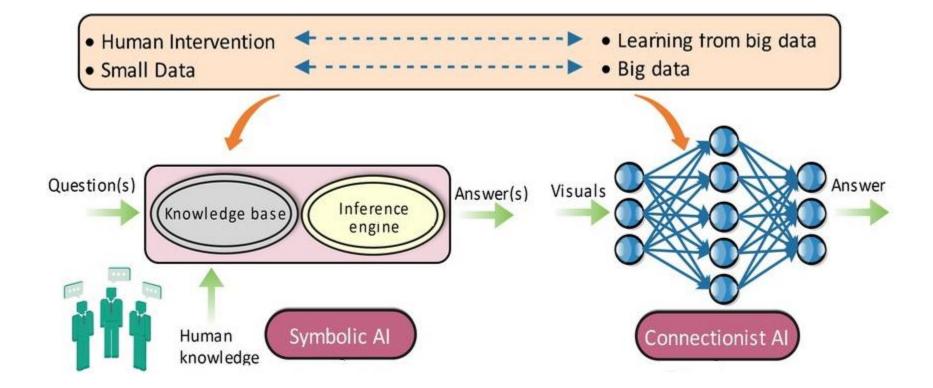
- Symbolic AI (Classical AI or Good Old-Fashioned AI GOFAI)
 - **Core Idea**: Intelligence can be achieved by manipulating symbols and rules that represent knowledge.
 - **Examples**: Expert systems, theorem provers, and knowledge-based systems.

Connectionism (Neural Networks and Deep Learning)

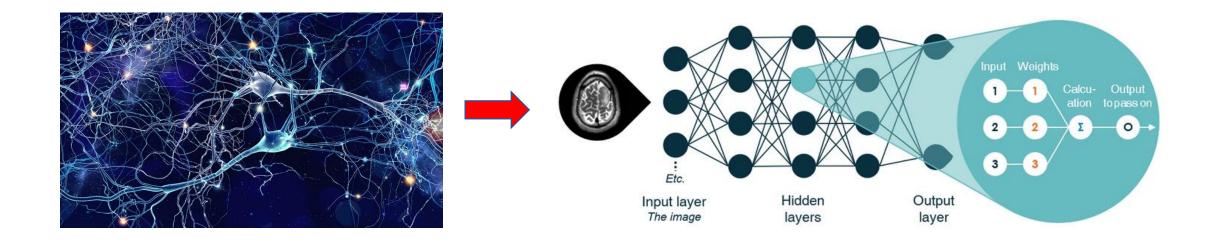
- **Core Idea**: Intelligence emerges from the interaction of simple processing units (neurons) in a network, inspired by the human brain.
- **Examples**: Deep learning models like convolutional neural networks (CNNs) and transformers (e.g., GPT, BERT).



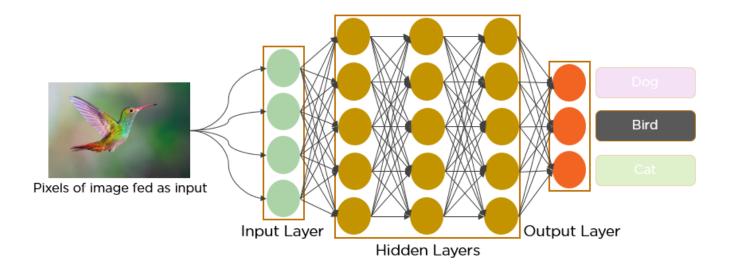
Connectionist vs Symbolic AI



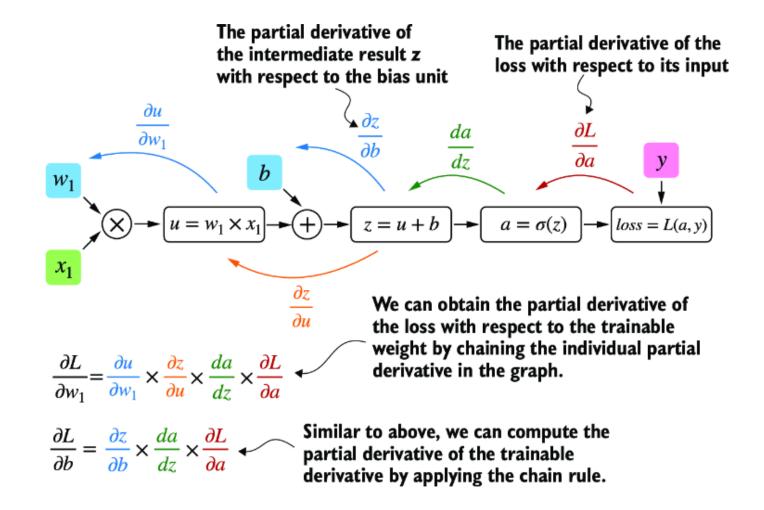
From Physical to Artificial Neural Networks



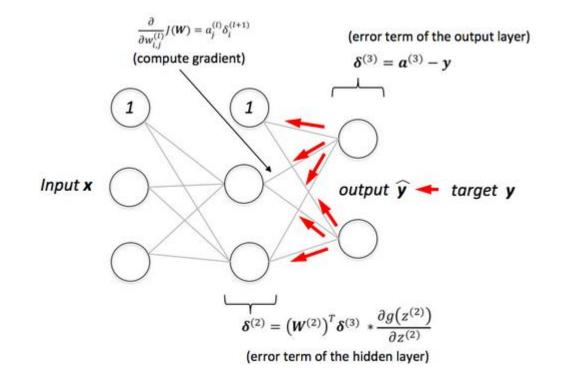
Training Artificial NNs – Backpropagation



Training Artificial NNs – Backpropagation

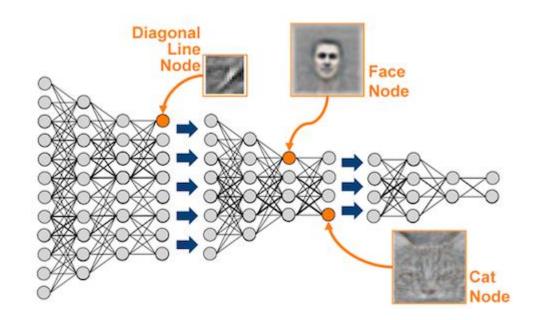


Training Artificial NNs – Backpropagation



Al-revolutionary change- since 2010

- More computational power
 - Graphics processing unit GPU
 - Clusters
- Better learning algorithms
 - Learning data representations;
 - Deep neural networks
 - Deep reinforcement learning



AI Today

- Don't know how to do 98% of the intelligent things
- But the rest 2% can do quite well

There's no magic in AI. It's all about optimization, probability, algorithms. with more computational power and huge volume of data

Why is it more important now?

- Flood of available data (especially with the advent of the Internet)
- Increasing computational power
- Growing progress in available algorithms and theory developed by researchers
- Increasing support from industries

Al today: natural language: chatbot



Valerie

Valerie: CMU Robot Receptionist in Newell-Simon hall.

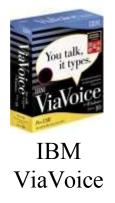
ALICE: 2004 Loebner Prize winner

Whisper OpenAl Automatic Speech Recognition (ASR)

Shallow?? natural language processing, pattern matching

Al today: natural language: speech recognition

- "speak or touch tone your card number" (tiny vocabulary, high accuracy needed)
- call routing: "how can I help you?" (large voc, low acc)
- dictation (large voc, high acc)





Dragon NaturallySpeaking

Hidden Markov Model, A* search, …

Al today: natural language: machine translation



The spirit is willing but the flesh is weak. (2005/6/29)

Дух охотно готов но плоть слабаSpirit is willingly ready but flesh it is weak精神是愿意的但骨肉是微弱的The spirit is wants but the flesh and blood is weak精神は喜んでであるが、肉は弱いMind is rejoicing,, but the meat is weakEl alcohol está dispuesto pero la carne es débil
The alcohol is arranged but the meat is weak

الكحول مستعدّة خير أنّ اللحمة ضعيفة. The alcohol is ready nevertheless the meat is weak.

- IBM statistical machine translation models
- US gov major consumer
 - Now? Neural Machine Translation

Use Foundation Models(like GPT or T5 Transformers)

Al today: natural language: question answering

What can I help with?

Message ChatGPT				
+ Gearch Q R	eason			•
🔀 Create image	Summarize text	Q Brainstorm	🕫 Help me write	d) Analyze data
	💡 Make a plan	Code	🕄 Surprise me	

Al today: game: from Chess to Go

- IBM Deep Blue vs. Kasparov, 1997/5
- 6 games: K, D, draw, draw, draw, D
- IBM stock up \$18 billion.



 Search, two-player zero-sum discrete finite games and Deep Reinforcement Learning.

AlphaGo



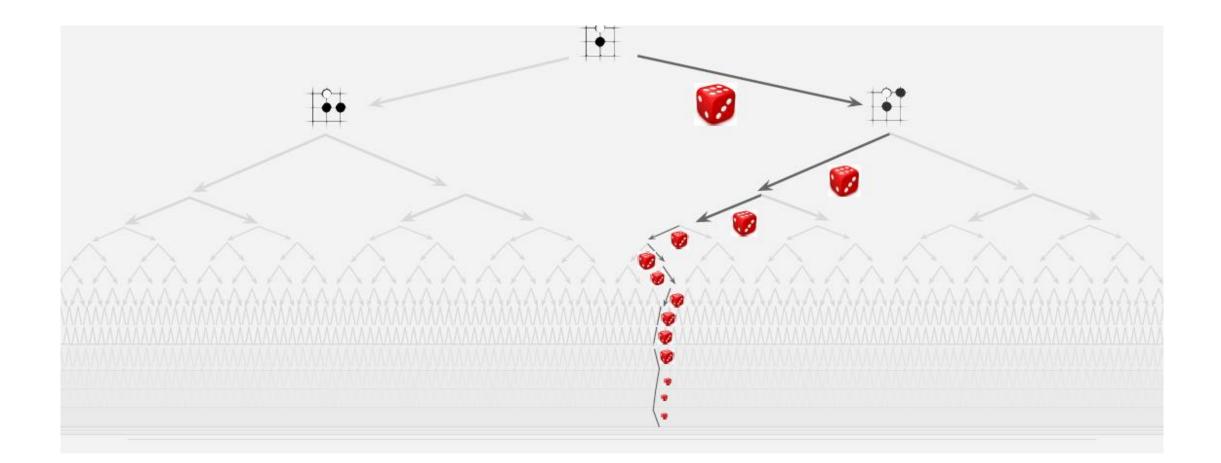
Brute force search intractable:

- 1. Search space is huge
- 2. "Impossible" for computers to evaluate who is winning

Game tree complexity = b^d



AlphaGo



Deep Blue vs AlphaGo

Deep Blue

AlphaGo

Handcrafted chess knowledge

Knowledge learned from expert games and self-play

Alpha-beta search guided by heuristic evaluation function

Monte-Carlo search guided by policy and value networks

200 million positions / second

60,000 positions / second



Al today: WWW: Google news

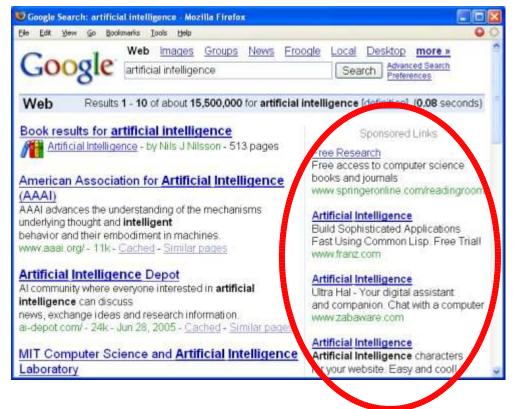
- Automatically selects / arranges news from multiple sources
- c.f. CNN



Unsupervised machine learning: clustering

Al today: WWW: ad

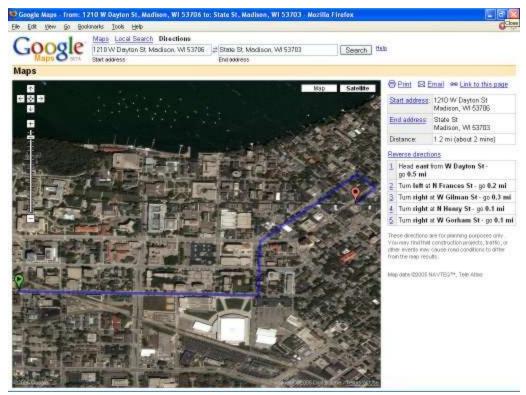
- "Sponsored links"
- Show ad based on relevance and money. Big business.



• Online algorithm, game, auction, multiple agents

Al today: WWW: driving directions

• From UW CS to state street



• search

Al today: WWW: collaborative filtering

- Recommendation based on other users' behavior
- e.g. Amazon.com



Availability: Usually ships within 24 hours. Ships from and so Want it delivered Friday, July 1? Order it in the next 8 hou choose One-Day Shipping at checkout. <u>See details</u> 76 used & new available from \$14.99

Share your own customer mnages Look inside another edition of this book

Customers who bought this book also bought

Introduction to Algorithms, Second Edition by Thomas H. Cormen

Machine Learning by Tom M. Mitchell

ANSI Common LISP by Paul Graham

Paradigms of Artificial Intelligence Programming : Case Studies in Common Lisp by Peter I

Operating System Concepts (Windows Xp Update) by Abraham Silberschatz

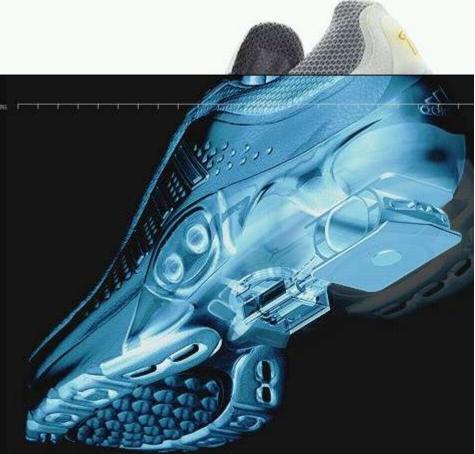
AI Application Programming (Programming Series) by M. Tim Jones

Explore Similar Items: in Books

Unsupervised learning

Al today: robotics: 'intelligent' shoes

• Adjust cushioning by speed, road surface (adidas_1)



Probably simple regression

Al today: robotics: robosoccer

• Robocup (http://www.robocup.org/)



Markov decision process, reinforcement learning

Al today: robotics: humanoid

Bipedal, human-like walking



Asimo (Honda)



QRIO (Sony)

Al today: robotics: Hubble telescope

- Scheduling: who gets to see what when
 - 30,000 observations per year
 - Many constraints, including
 - Earth blocks view every 95 minutes
 - Halts when in South Atlantic Ocean radiation belt
 - Avoid bright Sun, Moon, illuminated Earth
 - Disruption of plan for e.g. a supernova
- Search: Constraint satisfaction problem



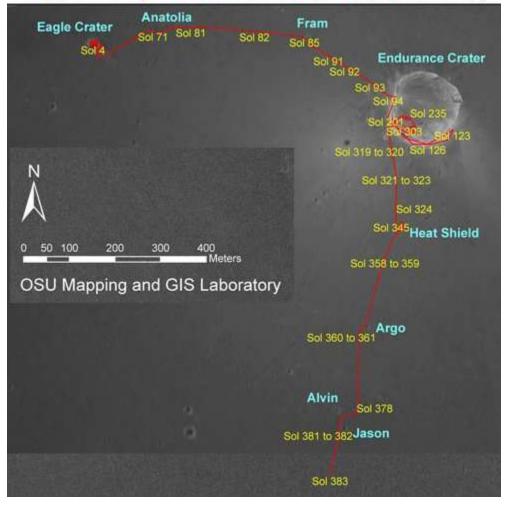
Al today: robotics: Mars Rovers

- Autonomous driving on Mars (part time)
- Robot motion planning



not always autonomously...

Opportunity Rover Traverse Map (Sol 383)



AI Today

Generative AI – Tools like ChatGPT and MidJourney create human-like text, images, and even music.

G Autonomous Al Agents – Al systems that can perform complex tasks with minimal human input.

Al in Work & Productivity – Automation is reshaping jobs, enhancing efficiency, and driving innovation.

Breakthroughs in Science & Medicine – AI accelerates drug discovery, medical diagnostics, and scientific research.

Are these intelligence?

Al defined

• Which one do you like?

	act	think
humanly	e.g. Turing test	How DO we think?
	1990, 1991	1985, 1978
rationally	agent 1998, 1998	Logic 1985, 1992

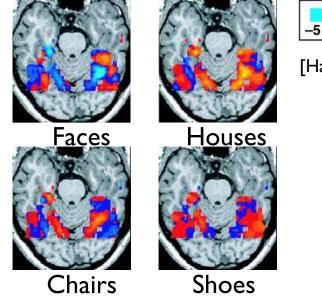
rational agent

- Sensors can be noisy or missing
- Actuators
 - may change the environment
 - can be inaccurate
- Performance measure
- Rational = optimize the performance measure
 - May not be perfect or even useful
 - e.g. "pick up as much dirt as possible"

How do we think? Mind reading



Which picture is seen? high accuracy [Cox & Savoy, Neuroimage 2003]





[Haxby et al, Science 2001]

 Also word meaning, picture vs. sentence, sentence ambiguity [Francisco Pereira, CMU Ph.D. thesis]