

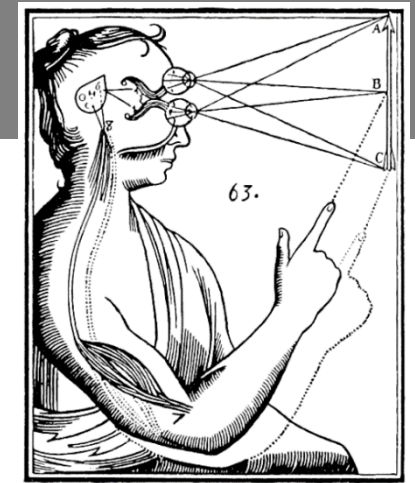
Artificial Intelligence

A course about foundations

Introduction

Marco Piastra

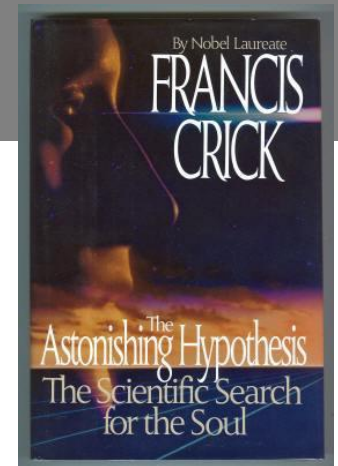
Artificial Mind?



(from Wikipedia)

- [Descartes, R., *Discours de la méthode pour bien conduire sa raison, et chercher la vérité dans les sciences*, 1637]

“I had after this described the **reasonable soul**, and shown that *it could by no means be educed from the power of matter*, as the other things of which I had spoken, but that it must be expressly created; and that it is not sufficient that it be lodged in the human body exactly like a pilot in a ship, unless perhaps to move its members, but that it is necessary for it to be joined and united more closely to the body, in order to have sensations and appetites similar to ours, and thus constitute a real **man**” [English version from Project Gutenberg]

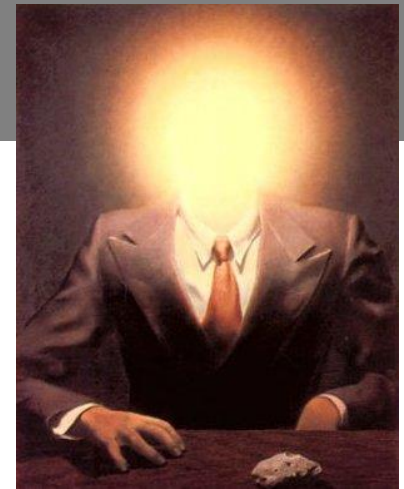


(from Wikipedia)

- [Crick, F., *The Astonishing Hypothesis*, 1994]

“You, your joys and your sorrows, your memories and your ambitions, your sense of personal identity and free will, are in fact no more than the behavior of a vast assembly of nerve cell and their associated molecules.”

*Artificial Brain:
can machines think?*



(from Wikipedia)

- [Searle, J. R., *Minds, Brain and Science*, 1986]

“Because we do not understand the brain very well we are constantly tempted to use the latest technology as a model for trying to understand it.

In my childhood we were always assured that the brain was a telephone switchboard (*‘What else could it be?’*).

I was amused to see that Sherrington, the great British neuroscientist, thought that the brain worked like a telegraph system. Freud often compared the brain to hydraulic and electro-magnetic systems. Leibniz compared it to a mill, and I am told some of the ancient Greeks thought the brain functions like a catapult.

At present, obviously, the metaphor is the digital computer.”

Turing Machine (A. Turing, 1937)

- Informal description (*more to come, later on*)

A **tape**, made up of individual **cells**

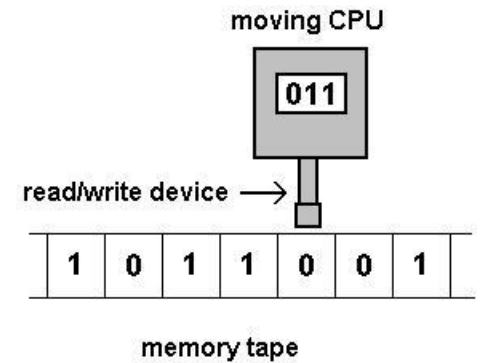
Each cell contains a **symbol**, from a finite **alphabet**

A **read/write head**, which can move in each direction - one cell at time

A **state register** that keeps the current **state**, from a finite set

A **transition table**, i.e. a set of *entries* like this:

{ $\langle \text{current_state}, \text{input_symbol} \rangle \rightarrow \langle \text{next_state}, \text{output_symbol}, \text{move} \rangle$ }



The **transition table** describes a *finite state machine*

Each *transition* is governed by the input symbol, the current state and the corresponding entry in the transition table

The next state is stored into the state register

The output is written to the cell

Then the head moves (i.e. *Left, None, Right*)

Turing Machine (A. Turing, 1937)

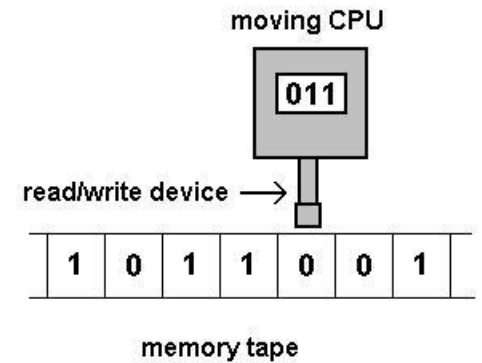
- What is the meaning of this?

The Turing Machine is a mathematical model of a physical computing device

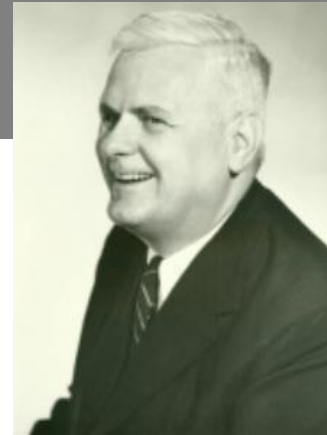
Any given problem for which there is a Turing Machine that computes the solution is clearly computable by a physical machine

Is the vice-versa also true?

(Whenever a problem is computable by a physical machine, does it exist a Turing Machine for it?)



Church-Turing Thesis



Caution: there is no such a thesis in the original writings of either author. Its formulation could be extrapolated from both, hence the attribution (made by others)

- A possible formulation (from Wikipedia):

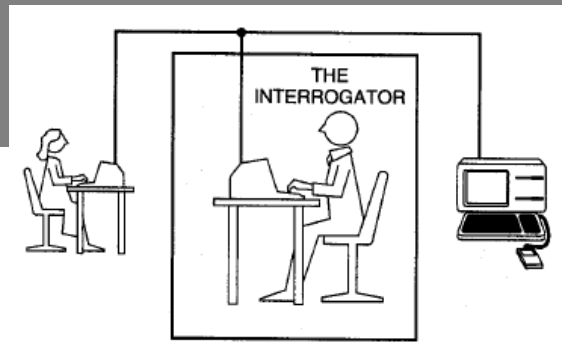
“Every 'function which would naturally be regarded as computable' can be computed by a Turing machine.”

The vagueness in the above sentence gives raise to different interpretations. One of these (though not entirely equivalent) is (from Wikipedia):

“Every 'function that could be physically computed' can be computed by a Turing machine.”

Searle: “... At present, obviously, the metaphor is the digital computer.”

Can machines think? (the Turing Test)



(from Wikipedia)

- Turing, A., *Computing Machinery and Intelligence*, 1950

“[The ‘imitation game’] is played with three people, a man (A), a woman (B), and an interrogator (C) who may be of either sex.

The interrogator stays in a room apart from the other two.

The object of the game for the interrogator is to determine which of the other two is the man and which is the woman.

He knows them by labels X and Y, and at the end of the game he says either ‘X is A and Y is B’ or ‘X is B and Y is A’

The interrogator is allowed to put **questions** to A and B. [...]

We now ask the question,

‘What will happen when a machine takes the part of A in this game?’

Will the interrogator decide wrongly as often when the game is played like this as he does when the game is played between a man and a woman?

These questions replace our original, ‘Can machines think?’ ”

*Artificial Brain:
can machines think?*

*Artificial Brain:
can machines think?*

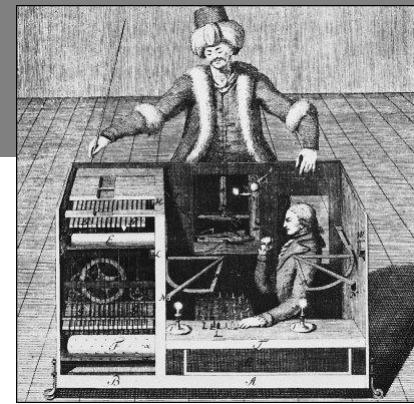
*Do answers, however partial,
change the original question?*

Can machines play chess?

In 1945 A. Turing mentions playing chess as an example of intelligent human activity that some days machines could perform

In 1946 A. Turing defines the first *algorithm* for playing chess

In 1948 C. Shannon wrote a famous article on the possible strategies for playing chess *automatically*



(from Wikipedia)

■ Programming a Computer for Playing Chess [Shannon, 1948]

More than 10^{43} different legitimate chessboard configurations

More than 10^{120} possible games

Strategy A

Starting from the current position, the machine *looks forward* by exploring all possible positions in the game not farther away than k moves

The computer chooses its move by **backward induction** using a value function (MINIMAX method)

Strategy B

“A good human player examines only **a few selected variations** and carries these out to a reasonable stopping point”

Strategy A or Strategy B?

[Shannon, 1948]

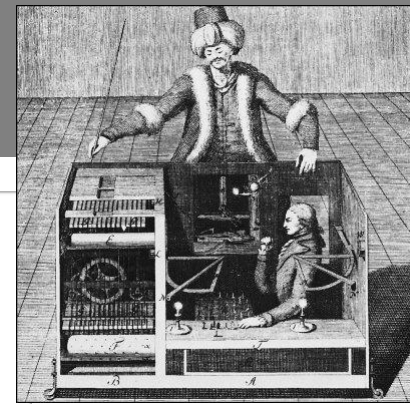
Due to the high computational complexity of Strategy A, he foresees a progressive development of Strategy B

(i.e. “Computer can improve by emulating humans”)

How did it go, in reality?

- At the early stages of computer chess technology, Strategy B was preferred
- During the period 1959-1962 a first ‘credible’ player was developed (Kotok-McCarthy) (at the *beginner* level)
- In 1973 the developers of the soon-to-be world champion in computer chess players abandoned Strategy B in favor of Strategy A
- *Since then, Strategy A – with significant improvements – dominates the scene*

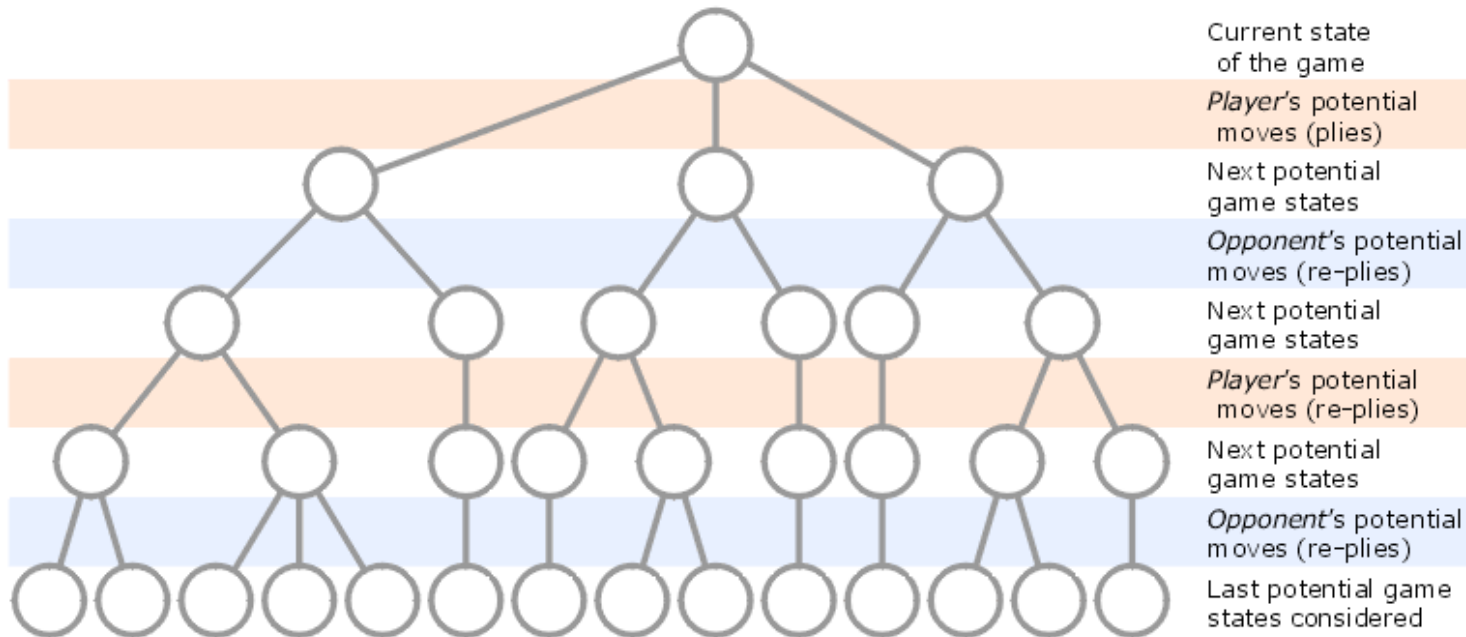
Can machines play chess?



(from Wikipedia)

Strategy A

Minimax on a two-person game tree of 4 plies



(from Wikipedia)

Deep Blue

In 1945 A. Turing mentions playing chess as an example of intelligent human activity that some days machines could perform

In 1946 A. Turing defines the first *algorithm* for playing chess

In 1948 C. Shannon wrote a famous article on the possible strategies for playing chess *automatically*

In 1997 the *Deep Blue* system, made by IBM, beats the world chess champion Gary Kasparov



(from Wikipedia)

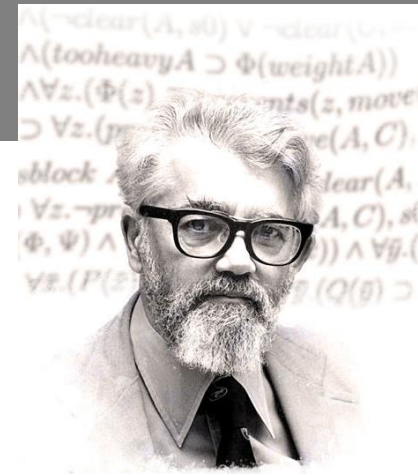
■ Deep Blue, 1997 [Campbell, Hoane, Hsu, F., 2001]

- 30 standard CPUs (120Mhz) + 480 special-purpose CPUs ('chess search engines', each evaluating >2.5M moves per second)
- Three-layered hardware architecture, 30 GB of RAM
- Software written in C
- Dedicated team of software and hardware engineers, 10 year of development
- Wide usage of a large database of recorded games played by grand masters

(A *supercomputer* for those times - *It was turned off at the end of the match*)

Artificial Intelligence: the beginning

"Artificial Intelligence" (first appearance of the term)



(from Wikipedia)

- [John McCarthy et al., 1955]

“We propose that a two-month, ten man study of **artificial intelligence** carried out during the summer of 1956 [...]

The study is to proceed on the basis of the conjecture that every aspect of learning or any other feature of **intelligence** can in principle be *so precisely described* that a machine can be made to *simulate* it. [...]

It may be speculated that a large part of human thought consists of manipulating **words** according to **rules of reasoning** and **rules of conjecture.**”

The Physical Symbol System Hypothesis (PSSH)

[Newell, A., Simon, H., *Computer Science as Empirical Inquiry Symbols and Search*, 1976]

“A **physical symbol system** consists of a set of entities, called **symbols**, which are physical patterns that can occur as components of another type of entity called an **expression** (or symbol structure).

Thus, a **symbol structure** is composed of a number of **instances** (or tokens) of symbols related in some physical way (such as one token being next to another).

At any instant of time the system will contain a collection of these symbol structures.

Besides these structures, the system also contains a collection of **processes** that operate on *expressions* to produce other *expressions*: processes of creation, modification, reproduction and destruction.”

Do elephants play chess?

[Brooks, R., *Elephants Don't Play Chess*, 1990]

- Criticism of *intelligence* intended as the manipulation of *symbols*

- A unique and synchronous control system

- Studies on cerebral lesions report evidence of parallel processing

- A unique, *general purpose* and neutral computational device

- Studies about human visual perception show clear preferences towards some interpretations over others

- A unique language for the internal representation of reality

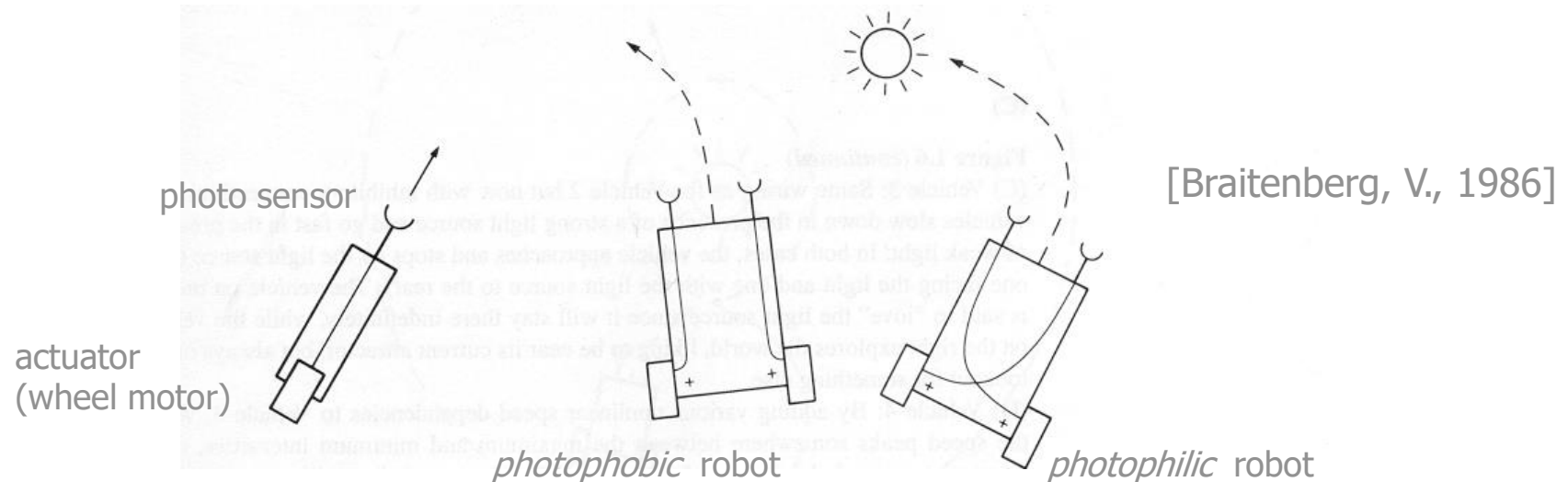
- Human beings do it differently – e.g. *change blindness*
[O'Reagan, J. K., Rensink, R. A., Clark, J. J., 1999]

- Total separation between the thinker and its hardware (*disembodiment*)

- Hence excluding all forms of non-symbolical intelligence

(Besides, how could it possibly *evolve* such a form of intelligence?)

Does this behavior look intelligent?

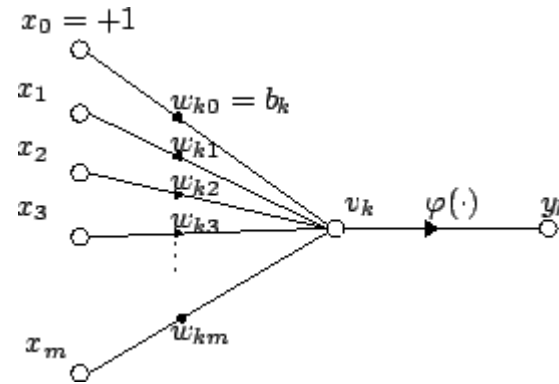
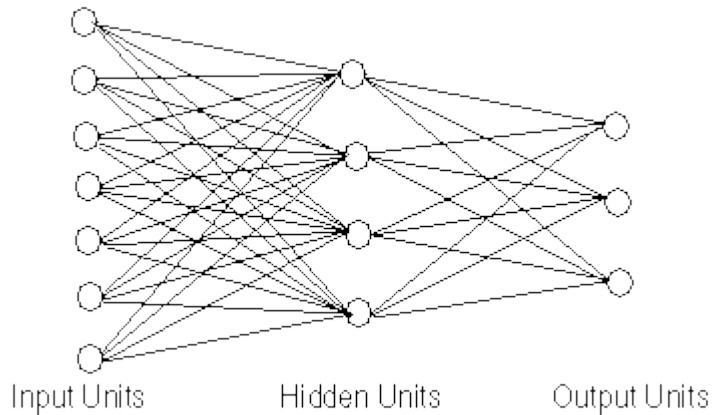


■ Direct connection

These robots by V. Braitenberg have just a *reactive* behavior, i.e. no 'thought in between': since sensors are directly connected to actuators

The resulting behavior is remarkable anyway ... ("*intelligence is in the eye of the beholder*")

Emulation or simulation? Connectionism



(from Wikipedia)

“In our view, people are smarter than today’s computers because the brain employs a basic computational architecture that is more suited to deal with a central aspect of the natural information processing tasks that people are so good at.”

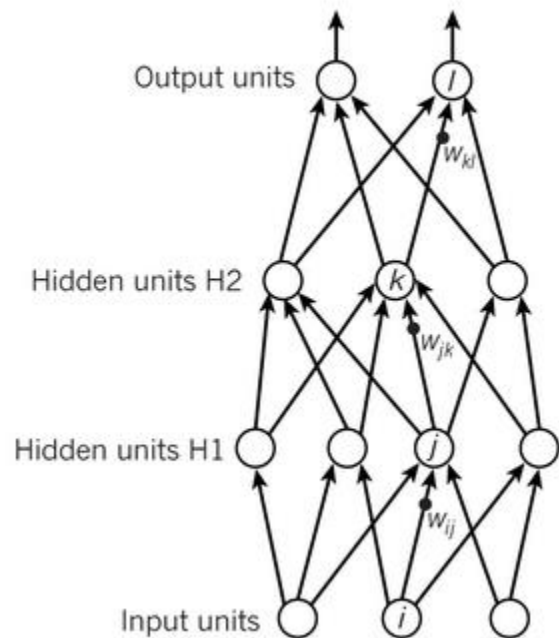
[Rumelhart, D.E., J.L. McClelland and the PDP Research Group (1986)
Parallel Distributed Processing: Explorations in the Microstructure of Cognition]

■ **Basic assumption**

Mental phenomena can be described by interconnected networks of simple and often uniform units

Artificial Neural Network

c



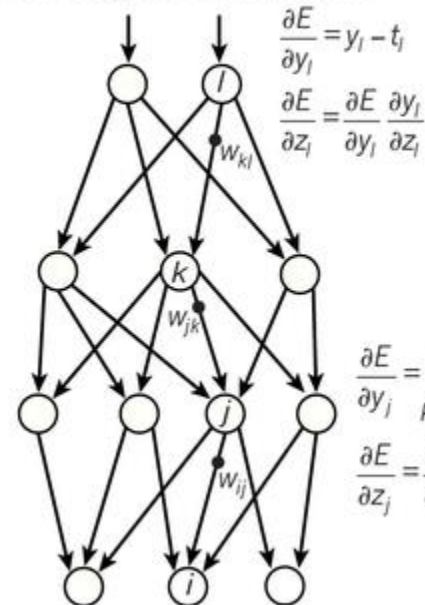
$$y_l = f(z_l)$$
$$z_l = \sum_{k \in H2} w_{kl} y_k$$

$$y_k = f(z_k)$$
$$z_k = \sum_{j \in H1} w_{jk} y_j$$

$$y_j = f(z_j)$$
$$z_j = \sum_{i \in \text{Input}} w_{ij} x_i$$

d

Compare outputs with correct answer to get error derivatives



$$\frac{\partial E}{\partial y_l} = y_l - t_l$$
$$\frac{\partial E}{\partial z_l} = \frac{\partial E}{\partial y_l} \frac{\partial y_l}{\partial z_l}$$

$$\frac{\partial E}{\partial y_k} = \sum_{l \in \text{out}} w_{kl} \frac{\partial E}{\partial z_l}$$

$$\frac{\partial E}{\partial z_k} = \frac{\partial E}{\partial y_k} \frac{\partial y_k}{\partial z_k}$$

$$\frac{\partial E}{\partial y_j} = \sum_{k \in H2} w_{jk} \frac{\partial E}{\partial z_k}$$

$$\frac{\partial E}{\partial z_j} = \frac{\partial E}{\partial y_j} \frac{\partial y_j}{\partial z_j}$$

[figure from LeCun, Bengio, Hinton, *Deep Learning*, Nature 521, 2015]

Function approximation

Basically, this is what an artificial neural network does

Supervised learning

The parameters (i.e. *weights*) are "learnt" from a dataset of inputs and expected outputs pairs

Incremental optimization — a.k.a. "backward propagation"

Weights are progressively corrected to reduce *the difference* between actual and expected outputs

Artificial Intelligence *now*:
a long jump ahead

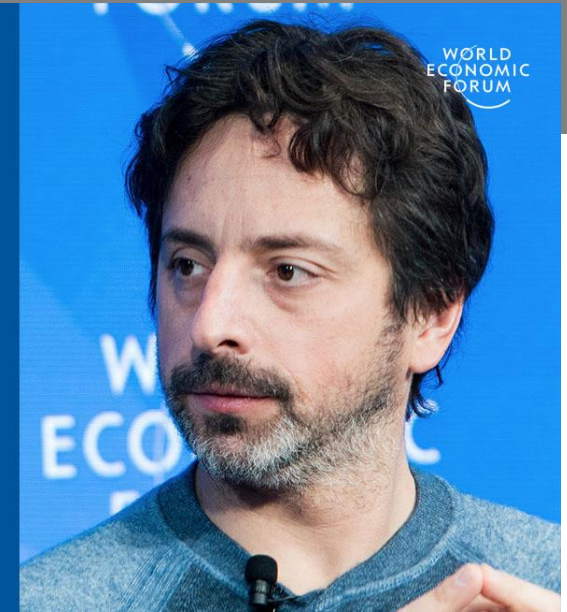


[Image from <https://www.tomgauld.com/portfolio>]

AI strikes back...

The revolution in AI has been profound, it definitely surprised me, even though I was sitting right there.

Sergey Brin
Google co-founder



- **Sergey Brin** [Google Co-Founder, January 2017]

"I didn't pay attention to it [i.e. Artificial Intelligence] at all, to be perfectly honest."

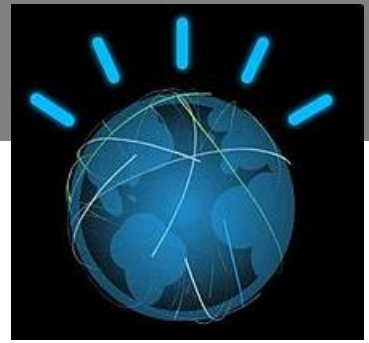
"Having been trained as a computer scientist in the 90s, everybody knew that AI didn't work.

People tried it, they tried neural nets and none of it worked."

[Quote and image from <https://www.weforum.org/agenda/2017/01/google-sergey-brin-i-didn-t-see-ai-coming/>]

Question Answering

DeepQA (a.k.a. "Watson")



(from Wikipedia)

- *Jeopardy!*: a quiz game

Category: General Science

Clue: When hit by electrons, a phosphor gives off electromagnetic energy in this form.

Answer: Light (or Photons)

Category: Diplomatic Relations

Clue: Of the four countries in the world that the United States does not have diplomatic relations with, the one that's farthest north.

Answer: North Korea

Category: Rhyme Time

Clue: It's where Pele stores his ball.

Answer: soccer locker

Category: Lincoln Blogs

Clue: Secretary Chase just submitted this to me for the third time; guess what, pal. This time I'm accepting it.

Answer: his resignation

DeepQA (a.k.a. "Watson")

■ DeepQA, 2010 [Ferrucci, D., et al. 2010]

The Event (14-18/02/2011)

In a sequence of three "*Jeopardy!*" games, Watson beats in a very convincing way the all-times human champions

- Brad Rutter, winner of the highest amount of money
- Ken Jennings, winner of the longest string of games

Jeopardy!: a quiz game

In the real game, questions can also be about images, audio or video displays

DeepQA can only accept spoken text as input

Autonomous search, local memory

The rules of the challenge forbid connecting to Internet during the game:

DeepQA must use its local memory only

It does use Internet during training

Conventional hardware, massive parallelism

High Performance system, with 2880 standard CPUs (no specialized hardware required)

Linux SUSE ES 11, Software in Java and C++, with Apache Hadoop and Apache UIMA

(IBM makes business on Watson, now)

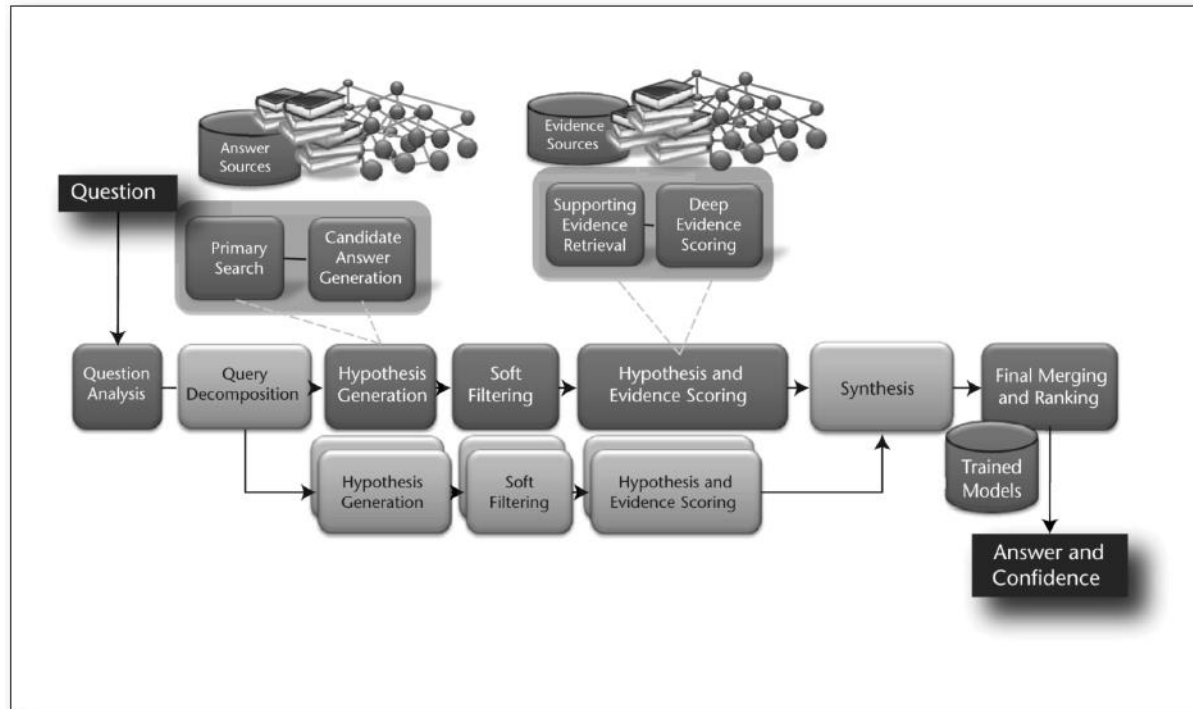


(from Wikipedia)

DeepQA (a.k.a. "Watson")

- How does it work?

Very little is known...



(from Wikipedia)

(from [Ferrucci, D, et al. 2010])

Processing stream

"They used nearly every trick in the book.." (from a video on YouTube)

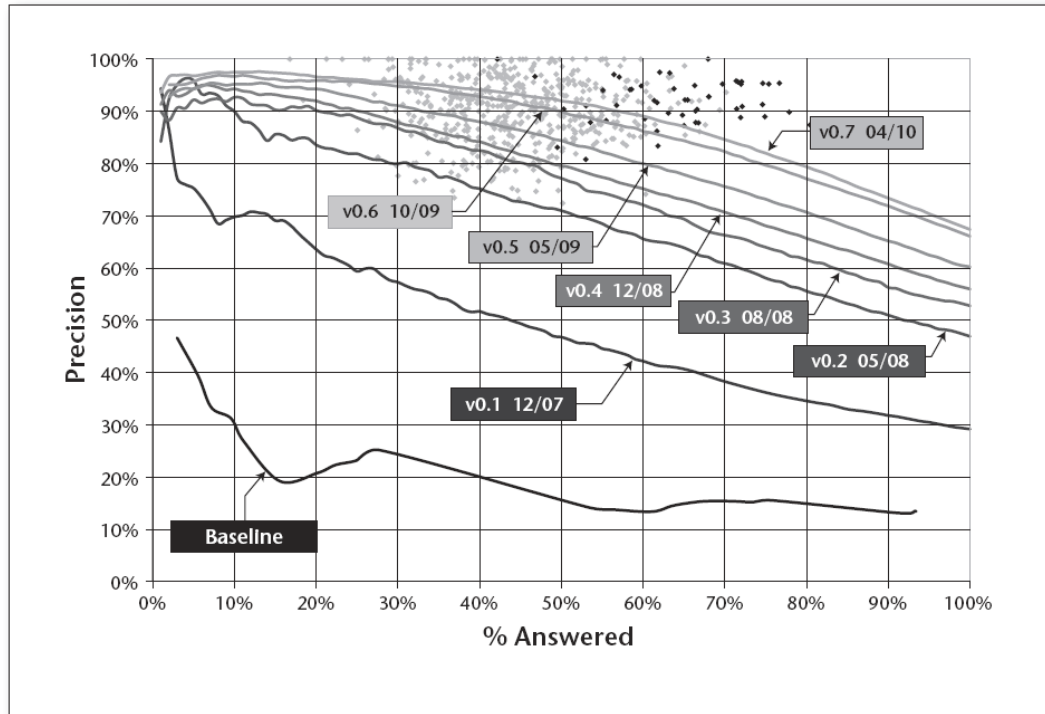
Several competing streams in parallel

Each stream scores a 'degree of confidence': the best answer is chosen, at the end

DeepQA (a.k.a. "Watson")

- How does it work?

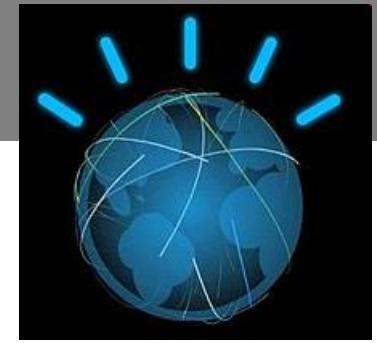
Very little is known...



(from [Ferrucci, D, et al. 2010])

Progressive, incremental training

Vast usage of *machine learning* techniques



(from Wikipedia)

Is Watson intelligent?

■ “Does Watson Think?”

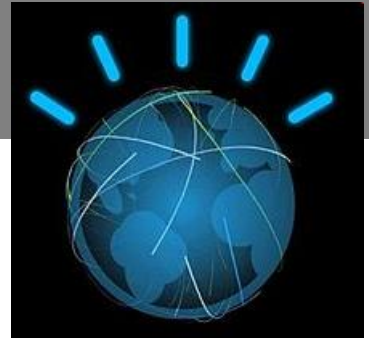
[D. Ferrucci, transcript from video
<http://www.ted.com/webcast/archive/event/ibmwatson>]

“Huh, hmm, what’s my favorite response on that?
(Do submarines swim?) [...]

I’d like to look at it as a sort of task-based view:
when you think of Watson playing Jeopardy!
it is acting like an intelligent Jeopardy! player,
if you deconstruct its intelligence
you’re gonna find lots of different algorithms
no one of them you would look at and say
“Wow! That’s really intelligent! It really understands the question!”
[...]

You have this holistic effect,
where it’s solving a problem that you formally think
that takes you *think*, to solve that problem, ...
Watson is doing it in a perhaps different way.
[...]

And I think ultimately of it as a tool, that helps humans solving problems... “

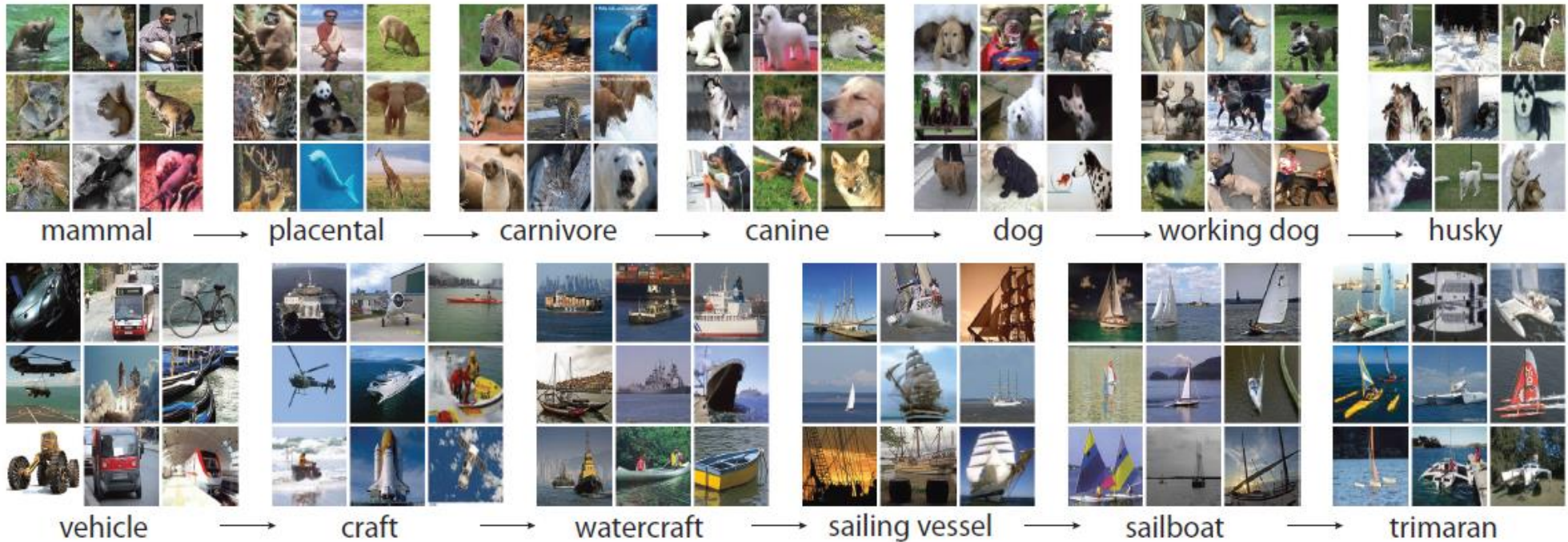


(from Wikipedia)

Image Classification via Deep Convolutional Neural Networks

ImageNet Challenge

■ The ImageNet Large Scale Visual Recognition Challenge



1,461,406 full resolution images

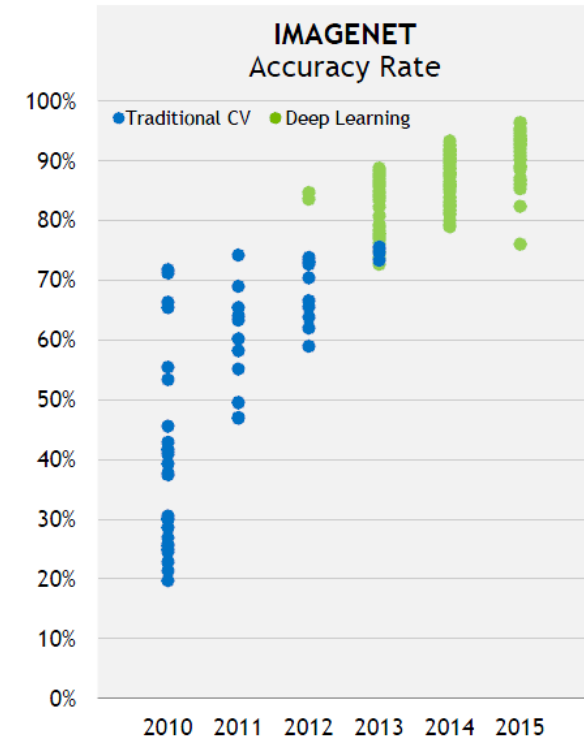
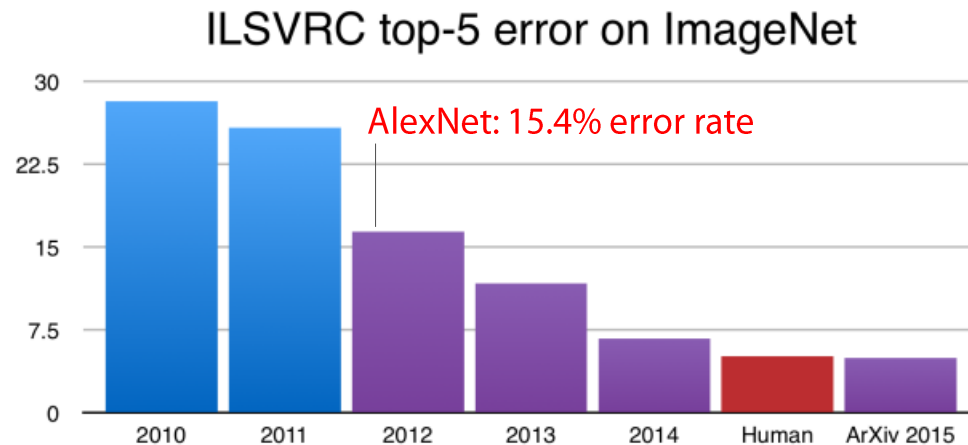
Complex and multiple textual annotation,
hierarchy of 1000 object classes along several dimensions

The image classification challenge is run annually since 2010

[figures from www.nvidia.com]

ImageNet Challenge

■ The ImageNet Large Scale Visual Recognition Challenge



1,461,406 full resolution images

Complex and multiple textual annotation,
hierarchy of 1000 object classes along several dimensions

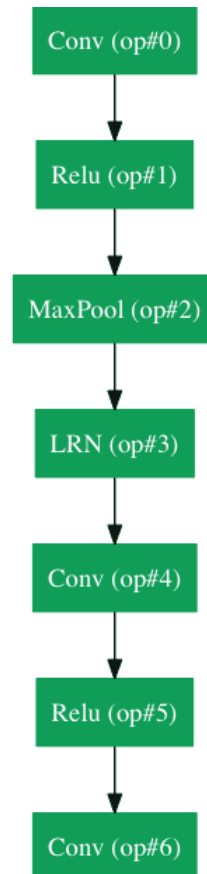
The image classification challenge was run annually from 2010 to 2017

[figures from www.nvidia.com]

Inception Architecture

- The ImageNet Large Scale Visual Recognition Challenge

How deep is a deep neural network, for a task like this?



GoogLeNet (Inception v4) winner of two out of three categories in 2014: 154 network layers

Game Playing with Reinforcement Learning

Generalization: DeepMind

■ Playing Atari with Deep Reinforcement Learning

[2013, V. Mnih, K. Kavukcuoglu, D. Silver, A. Graves, I. Antonoglou, D. Wierstra, M. Riedmiller, <http://www.nature.com/nature/journal/v518/n7540/full/nature14236.html>]

A software system only

Runs on virtually any Linux-based system, it contains optional provisions for GPU

Sophisticated machine-learning techniques:

Deep Reinforcement Learning

Reinforcement learning (RL - *see later in this course*)

in combination with convolutional neural networks (CNN)

Same configuration, multiple games

Same configuration applied to arcade games

It learned to play 7 (2013) or 49 (2015) different games

It is *autonomous*

It learns by itself, it receives no human expertise as input

In many cases, it outperforms human players

It's open source

<https://github.com/kuz/DeepMind-Atari-Deep-Q-Learner>



(from GitHub)

Beyond Chess: AlphaGo

Image from: <https://nikcheerla.github.io/deeplearningschool/2018/01/01/AlphaZero-Explained/>



- Mastering the game of Go with deep neural networks and tree search, [2016, D. Silver, et al. (22 authors), <http://www.nature.com/nature/journal/v529/n7587/full/nature16961.html>]

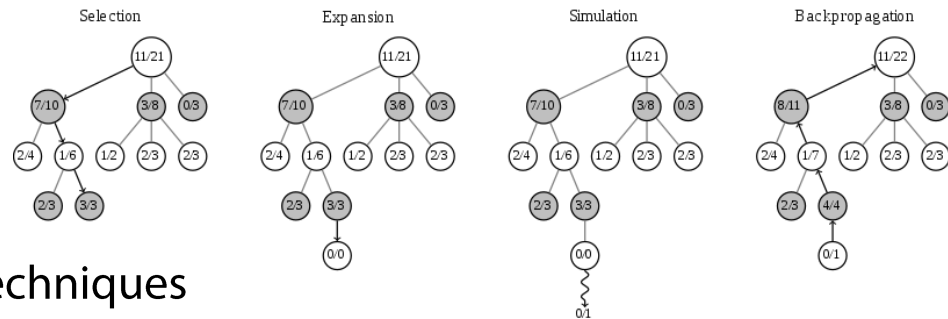
"The game of Go originated in China more than 2,500 years ago.

The rules of the game are simple: Players take turns to place black or white stones on a board, trying to capture the opponent's stones or surround empty space to make points of territory.

*As simple as the rules are, Go is a game of profound complexity. **There are more possible positions in Go than there are atoms in the universe.**"* [<https://deepmind.com/research/alphago/>]

Monte-Carlo Tree Search (MCTS)

Image from: https://en.wikipedia.org/wiki/Monte_Carlo_tree_search



Sophisticated machine-learning techniques

Deep neural networks (trained on human matches) for selecting the moves in the 'playouts'
Subsequent autonomous self-training: playing against itself and improving via reinforcement learning (RL - *see later in this course*)

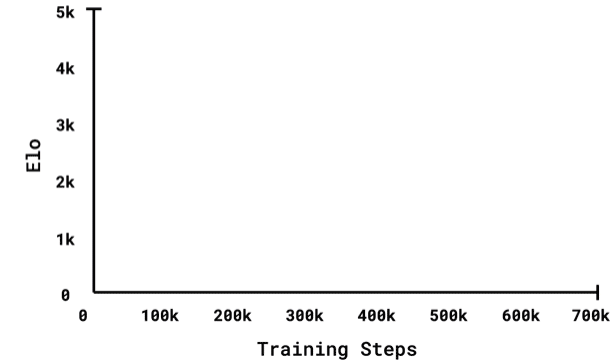
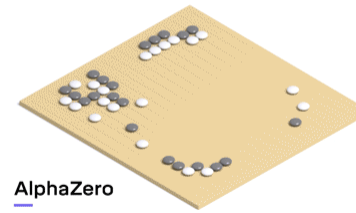
It is a very strong player

On March 2016, AlphaGo won 4-1 against the legendary Lee Sedol, the top Go player in the world over the past decade

Beyond Emulating Humans: AlphaZero

Image from: <https://deepmind.com/blog/article/alphazero-shedding-new-light-grand-games-chess-shogi-and-go>

DeepBlue and AlphaGo were heavily reliant on the experience of human players



■ AlphaZero learns by itself

[2018, D. Silver, et al. (13 authors), <https://science.sciencemag.org/content/362/6419/1140.full>]

Basic Knowledge Only

It just knows the basic rules of the games

Learning via Self-Play

It plays against a (frozen) copy of itself

MCTS and DCNN in a closed loop

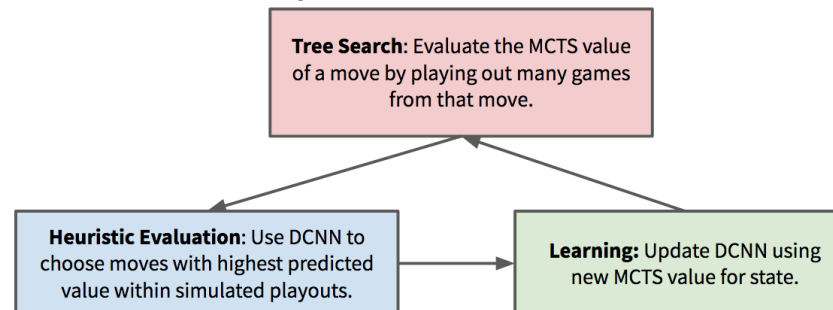


Image from: <https://nikcheerla.github.io/deeplearningschool/2018/01/01/AlphaZero-Explained/>



Artificial Intelligence: *short plan of this course*

Artificial Intelligence: *part 1*

- *Reasoning with symbols*

Propositional logic then first-order logic, logic programming (hints)

Representation: language and semantics (i.e. formulae and their meaning)

Inference: entailment among formulae

Automation: can machines compute entailment?

Plausible reasoning: beyond the scope of entailment

Artificial Intelligence: *part 2*

- *Reasoning with numbers*

 - Probability and machine learning**

 - Representation*: probability, graphical models

 - Inference*: answers to queries from joint probability distributions

 - Supervised learning*: learning from complete and well-formed datasets

 - Unsupervised learning*: when some of the data are either missing or *hidden*

 - Reinforcement learning*: learning from experience (even *online*)

 - Self-organization*:

 - the system changes its configuration in reaction to inputs (even *online*)