Machine Learning for Engineering and Science Applications

OF TECK

Introduction to the Course
History of Artificial Intelligence

What makes these possible? Your recently viewed items and featured recommendations Inspired by your browsing history MISBEHAVING Benjamin Franklii American Life > Walter Isaacson ★★★★☆ 14 Paperback ₹ 1,028.00 yprin tically deleted. Delete all spam messages now 12th, 2018 - Tai: Peace, This is a go ... 7:35 AM ☐ ☆ ∑ King, Darryl IMPORTANT OFFER - IMPORTANT OFFER There is a donation fo... 5:35 AM Go-gle

Simplistic Definition -- Machine Learning aims to replicate activities requiring human cognition

Course Aims

- To understand basic Machine Learning models thoroughly
 - Specific emphasis on Deep Learning
- 2. To be familiar with some modern ML techniques
- To apply these techniques hands-on to problems in Engineering
 - Examples Medical Image Diagnosis, Turbulence Modeling
 - You will have to learn and program in Python.

Course Prerequisites

- Mathematical "sophistication"
 - We will be introducing Linear Algebra, Probability and Optimization ideas required but....
 - You must be comfortable with Mathematics
- Degree of comfort with programming
 - We will be using examples from Python, MATLAB so....
 - You must understand Python, MATLAB syntax
 - And must be comfortable with programming if you wish to take the course for credit

Course Outline

Course will be in three broad parts

Part 1: Artificial Neural Networks and Deep Learning

Including CNNs and RNNs

Part 2: Other techniques

 Includes tree based methods, kernel methods, probabilistic methods, etc

Part 3: Modern/ Advanced methods

- Autoencoders, Generative Models, Reinforcement Learning, etc
- Applications assigned/discussed throughout the course

Syllabus for the course

Week	Lessons/Topics
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Reference Books

- Deep Learning, Goodfellow, Bengio, Courville, MIT Press, 2016*
- Pattern Recognition and Machine Learning, Bishop, Springer, 2009*
- Deep Learning With Python, Chollet, Manning Publications, 2017

^{*} The first two books have been made available for free by the publishers on their websites

History of Artificial Intelligence

Those who cannot learn from history are doomed to repeat it.

Prehistory (Till 1900)

- 11th century Robots that could replicate human speech and motion (Raja Bhoja)
- Realistic automatons in several parts of the world
- Liebnitz All human ideas are combinations of a few thoughts

1837 – Charles Babbage –
 Analytical Engine





Birth of the field of Artificial Intelligence (1900-1960)

- 1914 First chess playing machine (KR-K endings)
- 1925 Radio-controlled driverless car (Francis Houdina)
- 1940s Pitts and McCulloch First Artificial Neuron

Alan Turing – Theory of Computation, Imitation Game Shannon (Information Theory)

- 1950 Wiener Cybernetics
- 1951 Minsky- (SNARC) First Neural Net Machine
- 1955 Simon and Newell Logic Theorist Theorem Proving machine
- 1956 Dartmouth conference The term "A.I" is coined with the aim to build thinking machines. Formal birth of Artificial Intelligence

1957 Rosenblatt – Perceptron – Two-layer Artificial Neural Network

Machine Learning is the field of study that gives computers the ability to learn without being explicitly programmed. – Arthur Samuel, 1959

The golden years (1960-74)

- Appearance of Expert Systems Explicit, <u>rule based</u>, programs
 - Playing Chess
 - Helping in constructing organic chemistry models
 - Solving word problems in algebra
 - Understanding natural language
 - General purpose mobile robot
 - Identifying infections and recommending antibiotics
- Theoretical Progress Backpropagation 1969 (Bryson and Ho)

Optimism

"Machines will be capable, within twenty years, of doing any work a man can do"

-- H.A. Simon (1965)

"In from three to eight years we will have a machine with the general intelligence of an average human being." – Marvin Minsky (1970)

.....etc, etc

The first A.I. winter (1974-80)

Problems with A.I

- Results were primarily for toy problems
- Low computational power
- Combinatorial explosion
- Commonsense is nearly impossible to program!
- Minsky's book Perceptrons showed limitations of simple neural networks
- Most importantly, loss of government funding in A.I.

A new seasonal cycle (1980-2000)

Boom – Spring -- (1980-87)

- Expert Systems used in businesses with specialized hardware
- First driverless car
- Hopfield networks, popularization of backpropagation
- Minsky (1984) "Winter is coming!"

Bust – 2nd A.I Winter – (1987-93)

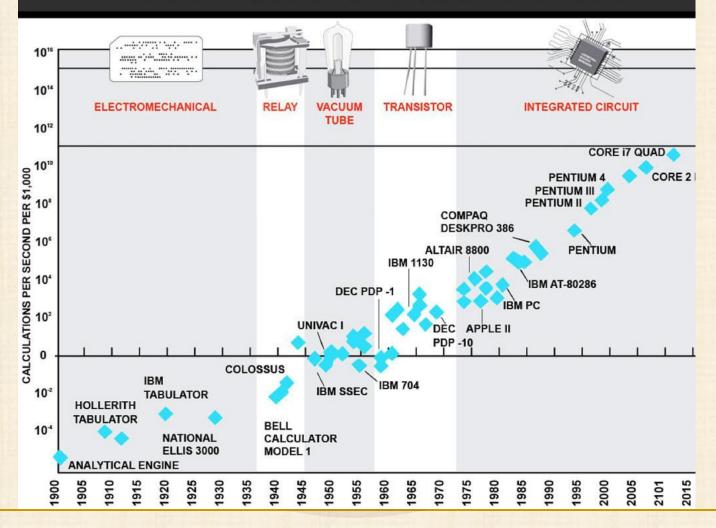
- Popularity of the PC.
- Disappointment in lack of spectacular results
- Brutal funding cuts

Consolidation – Summer – (1994-2000)

- 1997 Deep Blue beats Kasparov in chess
- Theory Including probability, information theory, optimization, etc
- Moore's Law Rapid growth of processing power

Moore's Law

115 Years of Moore's Law



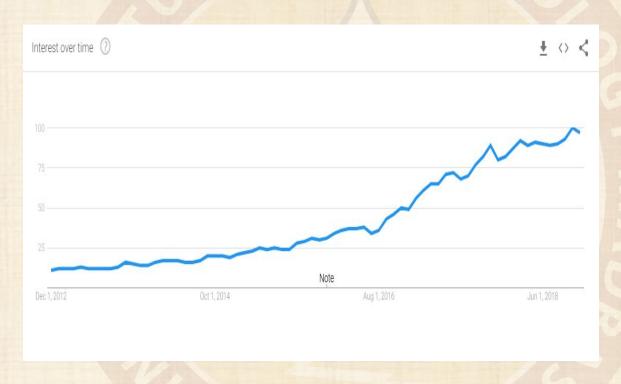
The quiet years (2000-12)

- Google is born.
- Internet Boom
- Shifted emphasis to big data statistical techniques
- Birth of Graphical Processing Units (GPUs)
- Good results in specific problems using deep networks
- Research focused on specific outcomes rather than general, all purpose, A.I
- 2005 Autonomous driving for 135 miles in desert
- IBM's Watson beat the Jeopardy champions

The A.I. Spring (2012 – ??)

- Lots of private funding
 - Google, IBM, Facebook, Microsoft....
- Rapid development in computational power
- Rapid growth of data
 - Data mining
 - Voluntary, distributed work (Amazon Turk, Captchas, Games)

Growth of Machine Learning/A.I



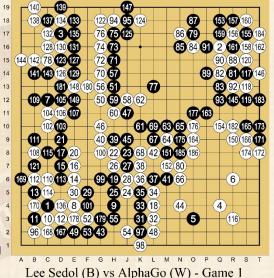
2012 Convolutional Neural Networks perform extremely well on image recognition challenge (ImageNet)

Machines vs Humans



https://upload.wikimedia.org/wikipedia/commons/thumb/5/51/IBM _Watson_w_Jeopardy.jpg/800px-IBM_Watson_w_Jeopardy.jpg

https://commons.wikimedia.org/wiki/File:Lee_Sedol_(B)_vs_AlphaGo_(W)_-_Game_1.svg



What is different this time

- Better technology
 - Exponentially huge computational power
- 2. (Really, really) Big Data
- Democratization of resources
 - Software and Hardware
- 4. Better algorithms

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