

Deep Learning

01-Introduction

Marco Piastra

This presentation can be downloaded at: http://vision.unipv.it/DL

Deep Learning: 01-Introduction [1]

Prologue: "Deep Learning"?

Deep Learning: 01-Introduction [2]

Deep Learning in the Artificial Intelligence Cosmos

Artificial Intelligence

automated reproduction of human cognitive activities

Machine Learning

automated extraction of generalized knowledge from data and experience

Deep Learning

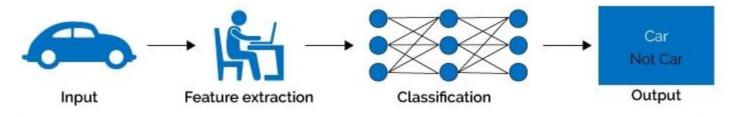
automated extraction of generalized knowledge from data and experience ...

... using multi-layered representations

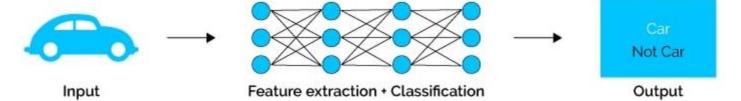
Deep Learning: 01-Introduction [3]

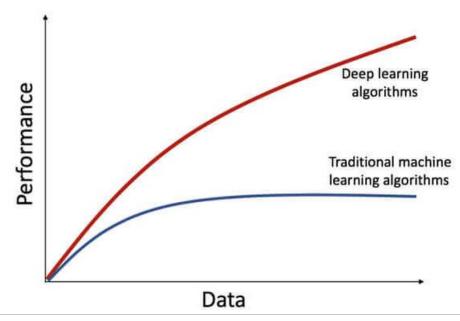
Machine Learning vs Deep Learning

Machine Learning



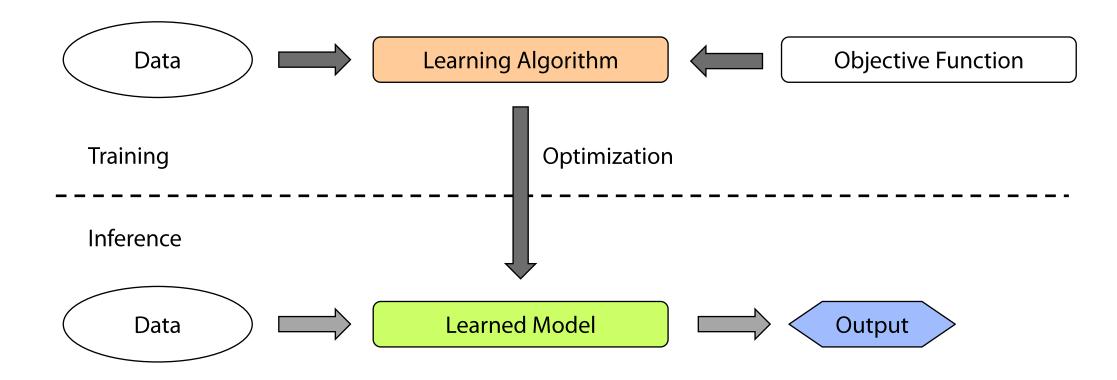
Deep Learning





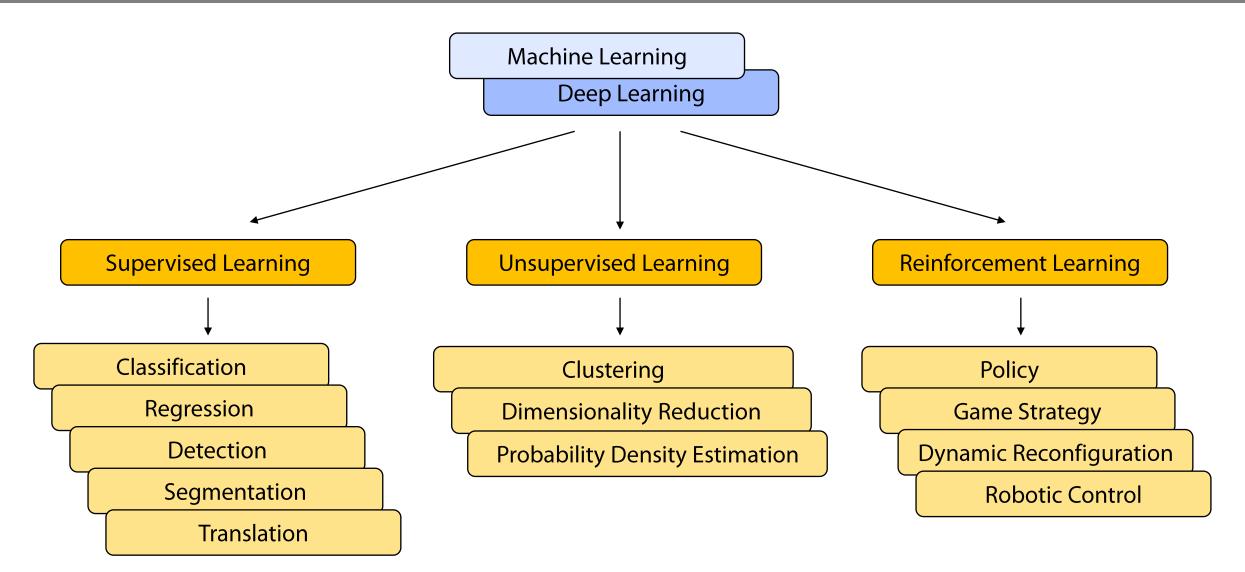
Deep Learning: 01-Introduction [4]

Machine Learning?



Deep Learning: 01-Introduction [5]

Types of learning algorithms and typical applications

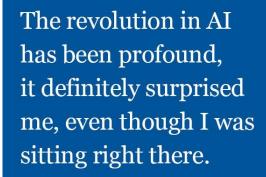


Deep Learning: 01-Introduction [6]

Yet Another Prologue: "Al Strikes Back?"

Deep Learning: 01-Introduction [7]

Al strikes back?



Sergey Brin
Google co-founder



Sergey Brin [Google Co-Founder, January 2017]

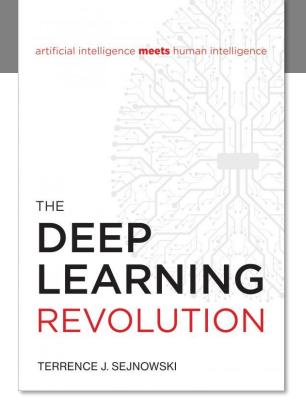
"I didn't pay attention to it [i.e. Artificial Intelligence] at all, to be perfectly honest."

"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/]

Deep Learning: 01-Introduction [8]

The Deep Learning Revolution



■ Terrence J. Sejnowski [President of the Neural Information Processing (NIPS) Foundation, October 2018]

"How deep learning—from Google Translate to driverless cars to personal cognitive assistants—is changing our lives and transforming every sector of the economy."

"Al is now awakening and transforming our world.

Driving these breakthroughs is the deep learning revolution [...]"

[Quote and image from https://mitpress.mit.edu/books/deep-learning-revolution]

Deep Learning: 01-Introduction [9]

Artificial Intelligence Overhype





10,693 views | Feb 15, 2019, 02:15pm

3 Reasons AI Is Way Overhyped



Peter Cohan Contributor ①
Markets

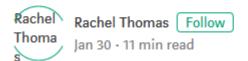
- 1. Many CEOs Are Being Scared Into Caring Too Much About AI
- 2. There Are Very Few Examples of High Payoff AI Applications
- 3. Very Few Companies Can Afford or Find Good Uses For AI

[Quote from https://www.forbes.com/sites/petercohan/2019/02/15/3-reasons-ai-is-way-overhyped/#3d3fef8c5a6a/

Deep Learning: 01-Introduction [10]

Artificial Intelligence Risks

I'm an AI researcher, and here's what scares me about AI



AI is being increasingly used to make important decisions. Many AI experts (including <u>Jeff Dean</u>, head of AI at Google, and <u>Andrew Ng</u>, founder of Coursera and deeplearning.ai) say that warnings about sentient robots are overblown, but other harms are not getting enough attention. I agree. I am an AI researcher, and <u>I'm worried</u> about some of the societal impacts that we're already seeing. In particular, these 5 things scare me about AI:

- 1. Algorithms are often implemented without ways to address mistakes.
- 2. AI makes it easier to not feel responsible.
- 3. AI encodes & magnifies bias.
- 4. Optimizing metrics above all else leads to negative outcomes.
- 5. There is no accountability for big tech companies.

[Quote from https://medium.com/@racheltho/im-an-ai-researcher-and-here-is-what-scares-me-about-ai-909a406e4a71

Deep Learning: 01-Introduction [11]

Al on the Rise: is that Good?



I'm an Al researcher, and here's what scares me about Al

AI is being increasingly used to make important decisions. Many AI experts (including Jeff Dean, head of AI at Google, and Andrew Ng, founder of Coursera and deeplearning.ai) say that warnings about sentient robots are overblown, but other harms are not getting enough attention. I agree. I am an AI researcher, and I'm worried about some of the societal impacts that we're already seeing. In particular, these 5 things scare me about AI:

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[Quote from https://medium.com/@racheltho/im-an-ai-researcher-and-here-is-what-scares-me-about-ai-909a406e4a71

Deep Learning: 01-Introduction [12]

Al on the Rise: is that Good?

[Quote from https://www.nytimes.com/2020/12/03/technology/google-researcher-timnit-gebru.html1

The New York Times

An Unsettling Chat With Bing Read the Conversation How Chatbots Work Spotting A.I.-Generated Text

Google Researcher Says She Was Fired Over Paper Highlighting Bias in A.I.

Timnit Gebru, one of the few Black women in her field, had voiced exasperation over the company's response to efforts to increase minority hiring.











Timnit Gebru, a respected researcher at Google, questioned biases built into artificial intelligence systems. Cody O'Loughlin for The New York Times





By Cade Metz and Daisuke Wakabayashi

Dec. 3, 2020

Deep Learning: 01-Introduction

Artificial Intelligence Hysteria?



Al isn't as scary as we imagine. AndreyZH/Shutterstock

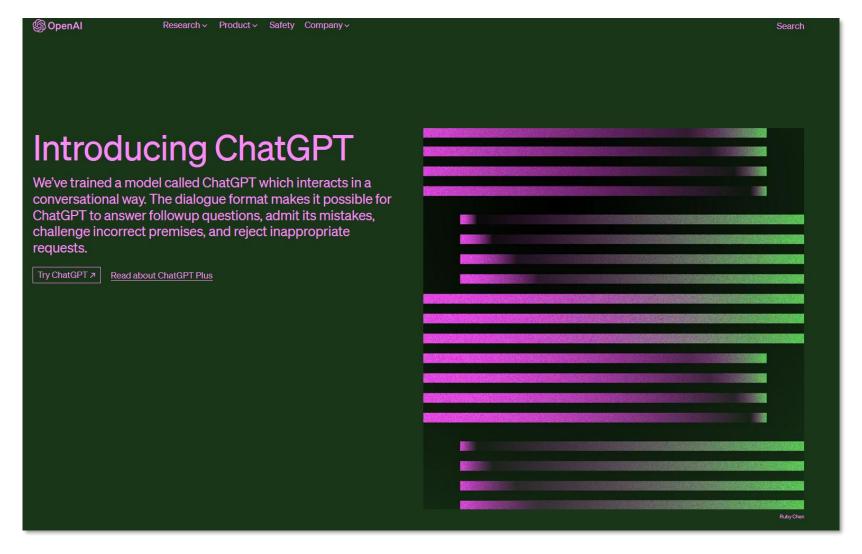
The reality of AI is currently very different, particularly when you look at the threat of automation. Back in 2013, <u>researchers estimated</u> that, in the following ten to 20 years, 47% of jobs in the US could be automated. Six years later, instead of a trend towards mass joblessness, we're in fact seeing US unemployment at <u>a</u> historic low.

Current AI is good at finding patterns in large datasets, and not much else.

[Quote from https://theconversation.com/ais-current-hype-and-hysteria-could-set-the-technology-back-by-decades-120514]

Deep Learning: 01-Introduction [14]

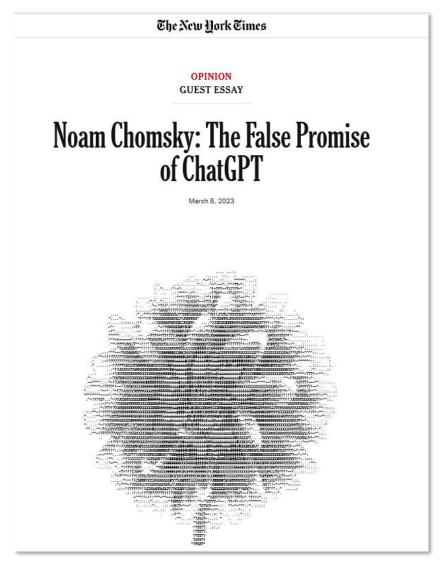
Is Artificial Intelligence Here to Stay?



[Image from https://openai.com/blog/chatqpt, 09/03/2023]

Deep Learning: 01-Introduction [15]

Is Artificial Intelligence Intelligent?



[Image from https://www.nytimes.com/2023/03/08/opinion/noam-chomsky-chatgpt-ai.html]

Deep Learning: 01-Introduction [16]

An aside: There might be more patterns about us than we may want to admit...

Deep Learning: 01-Introduction [17]

The Cambridge Analytical Scandal

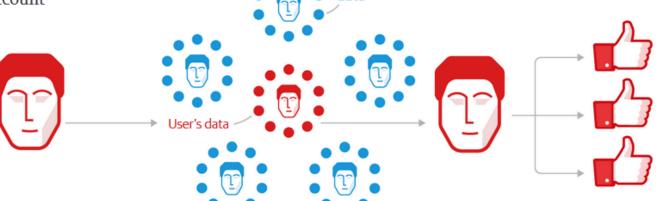
Cambridge Analytica: how 50m Facebook records were hijacked

Approx. 320,000 US
voters ('seeders') were
paid \$2-5 to take a
detailed personality/
political test that
required them to log in
with their Facebook
account

2 The app also collected data such as likes and personal information from the testtaker's Facebook account ...

The personality quiz results were paired with their Facebook data - such as likes - to seek out psychological patterns

4
Algorithms combined the data
with other sources such as voter
records to create a superior set
of records (initially 2m people in
11 key states*), with hundreds
of data points per person



... as well their friends' data,

raw Facebook data

amounting to over 50m people's

Friends'



These individuals could then be targeted with highly personalised advertising based on their personality data

[Graphics from https://www.theguardian.com/technology/2018/mar/17/facebook-cambridge-analytica-kogan-data-algorithm]

Deep Learning: 01-Introduction [18]

The Cambridge Analytical Scandal

Scientific foundations: the method

Two well-known articles by Kosinski et al.



Private traits and attributes are predictable from digital records of human behavior

Michal Kosinskia,1, David Stillwella, and Thore Graepelb

^aFree School Lane, The Psychometrics Centre, University of Cambridge, Cambridge CB2 3RQ United Kingdom; and ^bMicrosoft Research, Cambridge CB1 2FB, United Kingdom

Edited by Kenneth Wachter, University of California, Berkeley, CA, and approved February 12, 2013 (received for review October 29, 2012)



Computer-based personality judgments are more accurate than those made by humans

Wu Youyou^{a,1,2}, Michal Kosinski^{b,1}, and David Stillwell^a

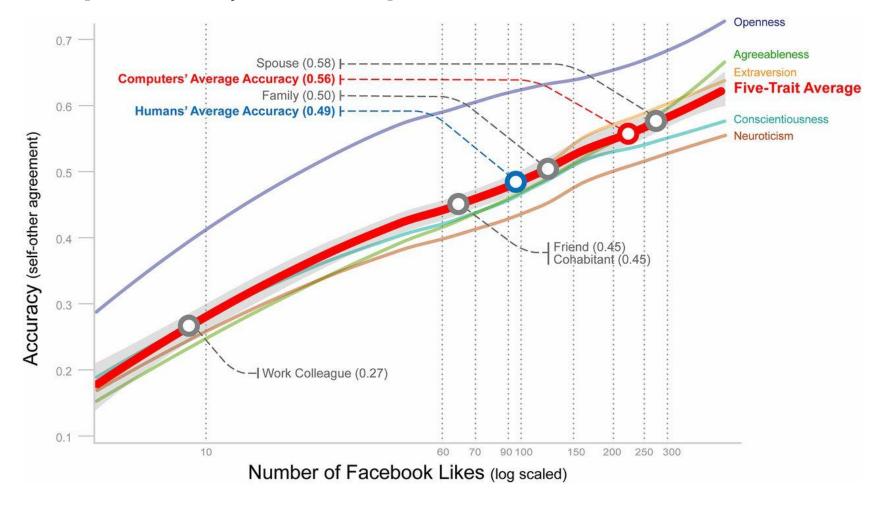
^aDepartment of Psychology, University of Cambridge, Cambridge CB2 3EB, United Kingdom; and ^bDepartment of Computer Science, Stanford University, Stanford, CA 94305

Edited by David Funder, University of California, Riverside, CA, and accepted by the Editorial Board December 2, 2014 (received for review September 28, 2014)

Deep Learning: 01-Introduction [19]

The Cambridge Analytical Scandal

The "Big Five" personality traits are predictable from Facebook likes



Wu Youyou et al. PNAS 2015;112:4:1036-1040

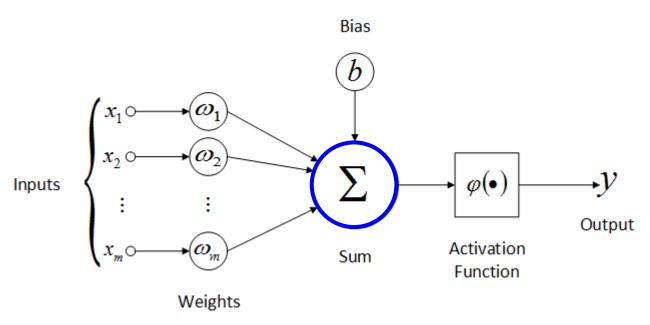
©2015 by National Academy of Sciences

Deep Learning: 01-Introduction [20]

Artificial Neural Networks: The Origins

Deep Learning: 01-Introduction [21]

Emulation or simulation? Connectionism







[Images from Wikipedia]

[Rumelhart, D.E., J.L. McClelland 1986]

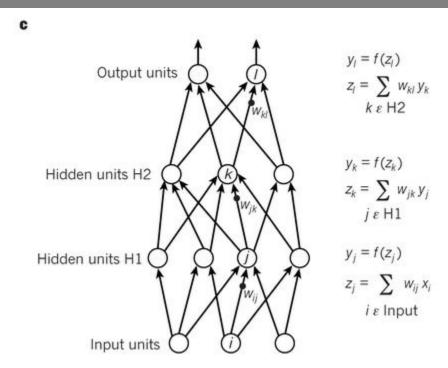
Basic assumption

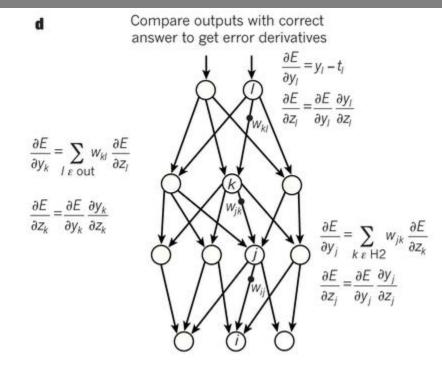
Mental phenomena can be described by interconnected networks of simple and often uniform units

"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]

Deep Learning: 01-Introduction [22]





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

Function approximation

Basically, this is what a 'classical' 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

Deep Learning: 01-Introduction [23]

From shallow to deep networks

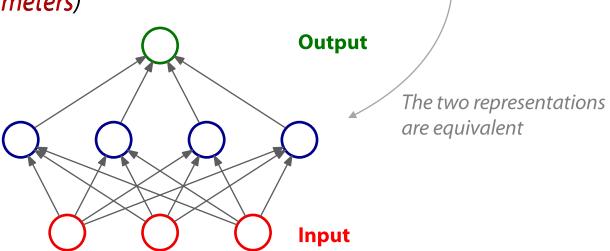
A feed-forward neural network with one hidden layer

$$\tilde{y} = \boldsymbol{w} \cdot g(\boldsymbol{W}^{[1]}\boldsymbol{x} + \boldsymbol{b}^{[1]}) + b$$

It can approximate any target function

$$y = f^*(\boldsymbol{x}), \ \boldsymbol{x} \in \mathbb{R}^d$$

(given enough units and proper parameters)



Deep Learning systems

use this representation

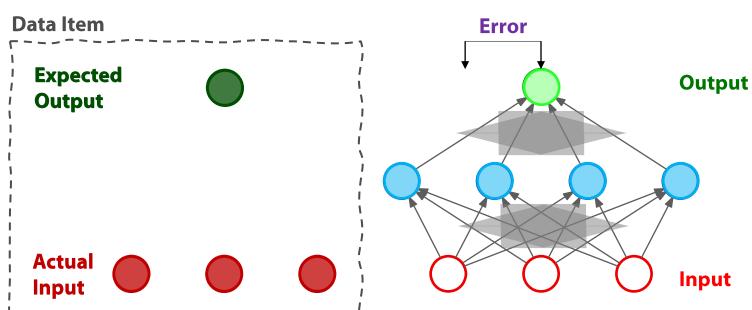
(e.g. TensorFlow, PyTorch)

Deep Learning: 01-Introduction [24]

Learning is a parameter optimization process

Using a large dataset of input-output pairs (data items)

$$\tilde{y} = \boldsymbol{w} \cdot g(\boldsymbol{W}\boldsymbol{x} + \boldsymbol{b}) + b$$



Feed Data Item(s)

Improve

Repeat

Several million times...

Propagate Input to compute Output

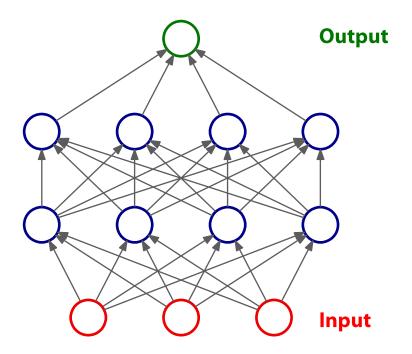
Propagate Error to improve the parameters

Deep Learning: 01-Introduction [25]

From shallow to deep networks

A feed-forward neural network with two hidden layers

$$\tilde{y} = w \cdot g(W^{[2]}g(W^{[1]}x + b^{[1]}) + b^{[2]}) + b$$

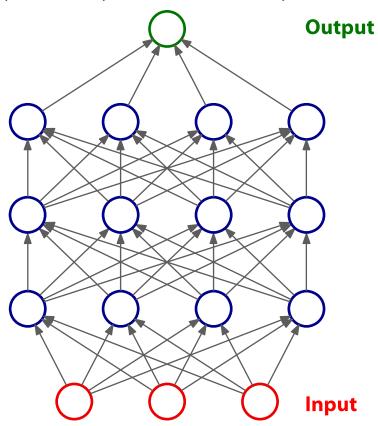


Deep Learning: 01-Introduction [26]

From shallow to deep networks

A feed-forward neural network with three hidden layers

$$\tilde{y} = w \cdot g(W^{[3]}g(W^{[2]}g(W^{[1]}x + b^{[1]}) + b^{[2]}) + b^{[3]}) + b$$



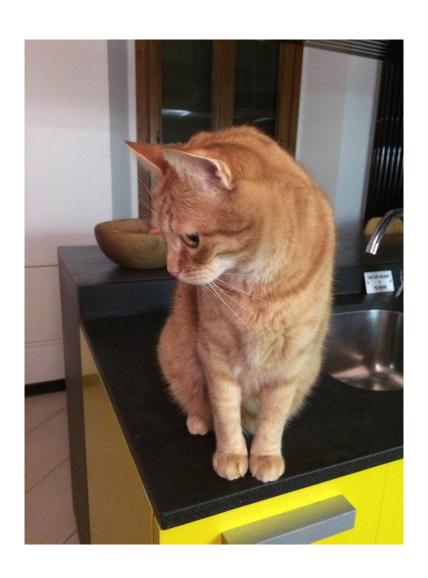
Deep Learning: 01-Introduction [27]

Finding Patterns in Large Datasets

Deep Learning: 01-Introduction [28]

One Giant Leap for Mankind

Is there a cat in this picture?

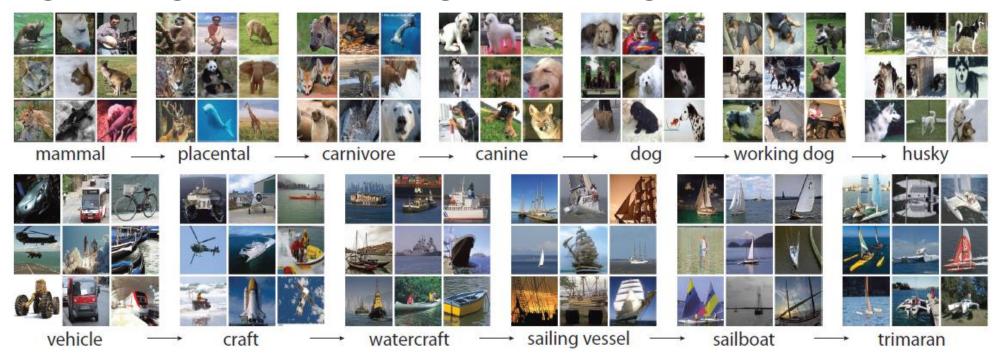


[this is my cat, Rabarbaro]

Deep Learning: 01-Introduction [29]

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

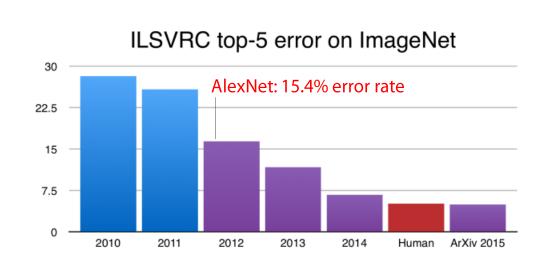
The image classification challenge was run annually from 2010 to 2017

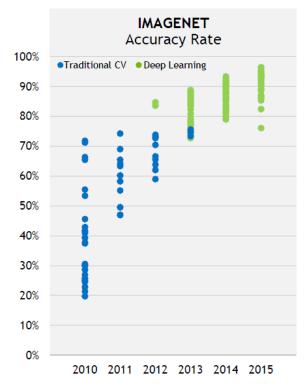
[figures from www.nvidia.com]

Deep Learning: 01-Introduction [30]

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

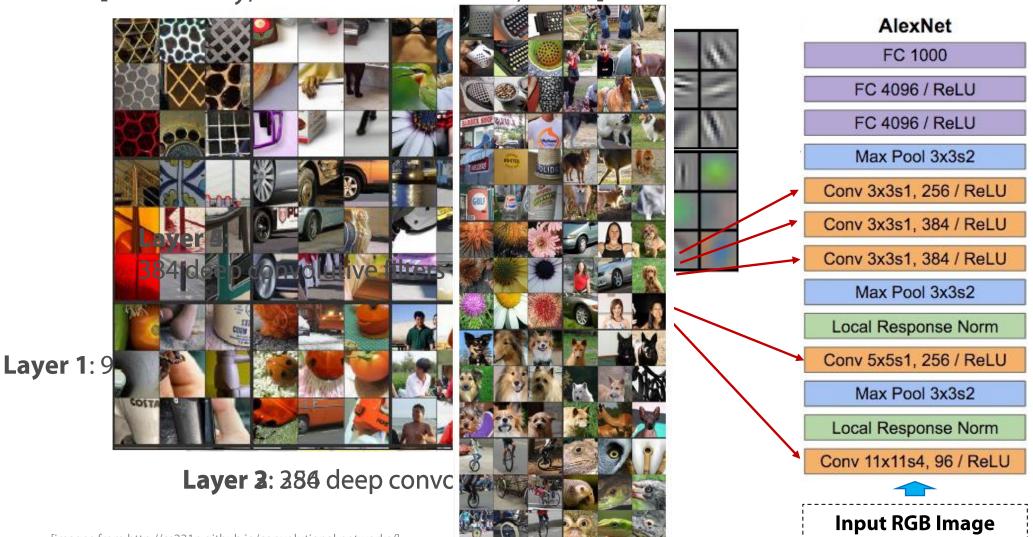
The image classification challenge was run annually from 2010 to 2017

[figures from www.nvidia.com]

Deep Learning: 01-Introduction [31]

Deep Convolutional Neural Networks (DCNN)

AlexNet [Krizhevsky, Sutskever & Hinton, 2012]



[images from http://cs231n.github.io/convolutional-networks/]

Deep Learning: 01-Introduction [32]

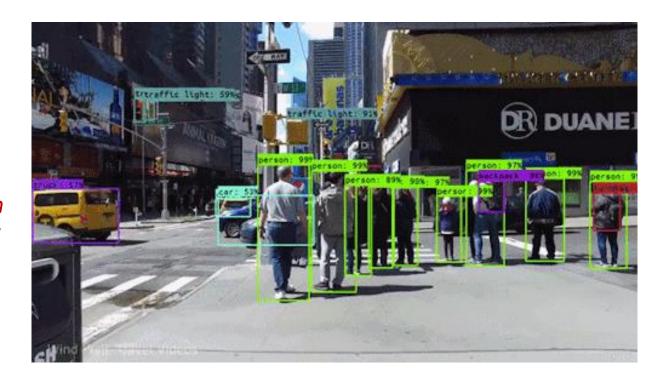
Object (and People) Real-Time Detection

Deep Convolutional Neural Networks have evolved since then ...

Now these system can identify objects and persons from videos, in real time

NOTE:

According to the recent EU Proposal for a Regulation about Al, **remote biometric identification** (RBI) in public places will require a special authorization



Deep Learning: 01-Introduction [33]

Image Segmentation

Deep Convolutional Neural Networks have evolved since then ...

They can perform a complete scene analysis, from videos, in real time



At present, DCNN work on a frame-by-frame basis

Deep Learning: 01-Introduction [34]

Well, it's just a function anyway (contd.)...

Deep Learning: 01-Introduction [35]

Games of strategy: AlphaGo (2016)

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]

There are more possible positions in Go than there are atoms in the universe

Sophisticated machine-learning techniques

Strategy selection via Monte Carlo Tree Search (MCTS)

Deep neural networks (trained on human matches) for both guidance and learning

Adversarial self-training:

playing again itself and improving via reinforcement learning

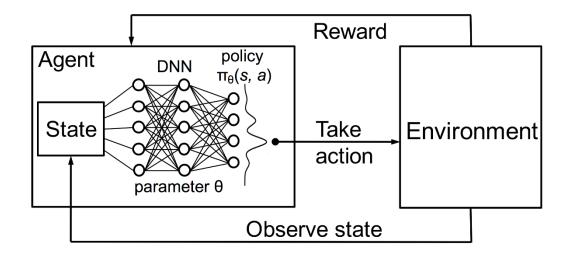
Super-human player?

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

Deep Learning: 01-Introduction [36]

Deep Reinforcement Learning (DRL)

A Deep Neural Network learns a policy



The agent interacts with an environment (it could be a copy of itself)

It selects an **action** in each **state** and receives a **reward** (possibly deferred) as a function of the results obtained

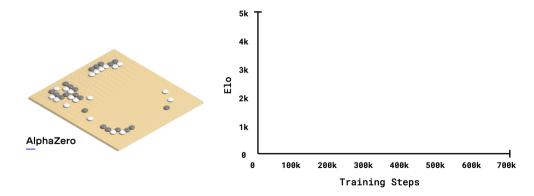
The DRL system optimizes its policy

Deep Learning: 01-Introduction [37]

Beyond Emulating Humans: AlphaZero (2018)

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

AlphaGo is 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



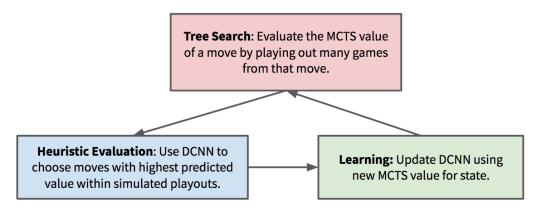
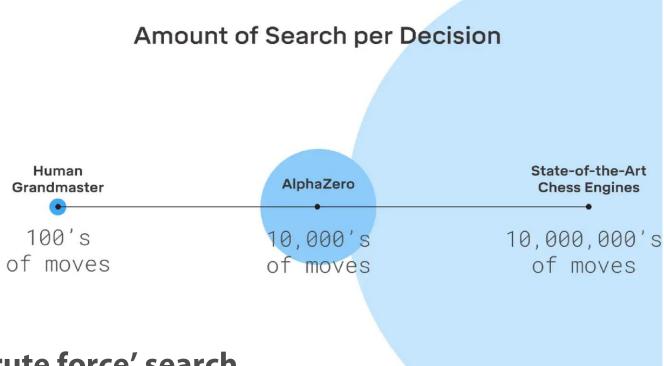


Image from: https://nikcheerla.github.io/deeplearningschool/2018/01/01/AlphaZero-Explained/

Beyond Emulating Humans: AlphaZero (2018)

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



AlphaZero uses much less 'brute force' search

When playing, the search process is driven by its neural network

It acts like a memory of past experiences

While training, it learns through a huge amount of self-playing

But it is a faster learner than Alpha Go

Deep Learning: 01-Introduction [39]

Transition Functions Predict Dynamics

Deep Learning: 01-Introduction [40]

AlphaFold (2020)

Images from https://deepmind.com/blog/article/alphafold-a-solution-to-a-50-year-old-grand-challenge-in-biology

Predicting how proteins fold

Several Neural Networks

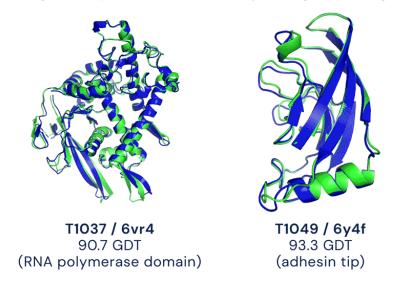
Trained separately, working together

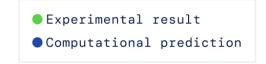
A big step ahead

Substantial improvement over the best actual systems yet

Median Free-Modelling Accuracy









It's open source

In the version used for the CASP13 challenge

https://github.com/deepmind/deepmind-research/tree/master/alphafold_casp13

Deep Learning: 01-Introduction [41]

Autogressive Models Generate Language

Deep Learning: 01-Introduction [42]

GPT-3 (2020)

Image from https://www.theguardian.com/commentisfree/2020/sep/08/robot-wrote-this-article-gpt-3



Deep Learning: 01-Introduction [43]

GPT-3 (2020)

Image from https://www.nytimes.com/2020/07/29/opinion/gpt-3-ai-automation.html

Fortier fed GPT-3 a strange prompt: "Below is a transcript from an interview where Barack Obama explained why he was banned from Golden Corral for life." The system then filled in the rest of the interview, running with the concept that Obama had been banned from an all-you-can-eat buffet.

Obama: Yes. It's true. I am no longer allowed in Golden Corral.

Interviewer: Is this because of your extensive shrimp-n-crab legs policy?

Obama: Absolutely.

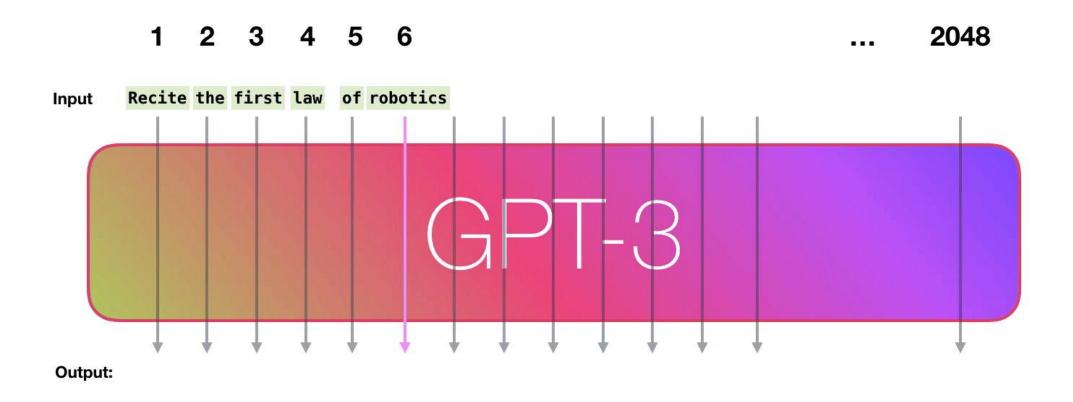
Interviewer: What is your extensive shrimp-n-crab legs policy?

Obama: Oh, well, in brief, they were offering an all-you-can-eat shrimp-n-crab leg buffet, and I did not hesitate. After I ate so much shrimp and crab that my stomach hurt, I would quietly sneak in and throw more shrimp and crab onto my plate. I did this over and over again until I had cleaned out the buffet and was full of shrimp-n-crab.

Deep Learning: 01-Introduction [44]

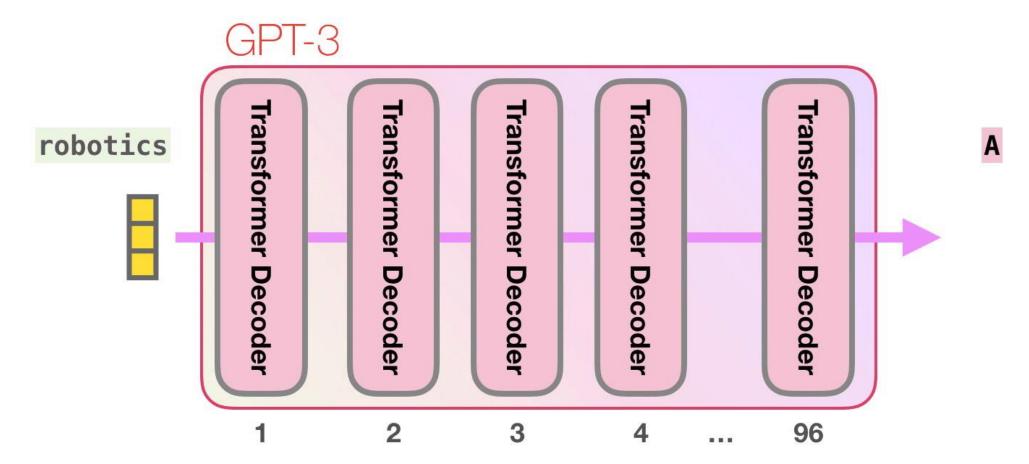
Image from http://jalammar.github.io/how-gpt3-works-visualizations-animations/





Deep Learning: 01-Introduction [45]

Image from http://jalammar.github.io/how-gpt3-works-visualizations-animations/



One of the biggest Neural Networks yet

GPT-3 has 175 <u>Billion</u> parameters (AlexNet has 64 <u>Million</u>)

Deep Learning: 01-Introduction [46]

AlphaCode (2022)

Solving coding problems using transformers



Problem (input)

D.Backspace

You are given two strings s and t, both consisting of lowercase English letters. You are going to type the string s character by character, from the first character to the last one.

When typing a character, instead of pressing the button corresponding to it, you can press the 'Backspace' button. It deletes the last character you have typed among those that aren't deleted yet (or does nothing if there are no characters in the current string). For example, if s is "abcbd" and you press Backspace instead of typing the first and the fourth characters, you will get the string "bd" (the first press of Backspace deletes no character 'c'). Another example, if s is "abcaa" and you press Backspace instead of the last two letters, then the resulting text is "a".

Your task is to determine whether you can obtain the string t, if you type the string s and press "Backspace" instead of typing several (maybe zero) characters of s.

Input

The first line contains a single integer q ($1 \le q \le 10^5$) — the number of test cases.

The first line of each test case contains the string s ($1 \le |s| \le 10^5$). Each character of s is a lowercase English letter.

The second line of each test case contains the string t ($1 \le |t| \le 10^5$). Each character of t is a lowercase English letter.

It is guaranteed that the total number of characters in the strings over all test cases does not exceed 2·10^s.

Output

For each test case, print "YES" if you can obtain the string t by typing the string s and replacing some characters with presses of "Backspace" button, or "NO" if you cannot.

You may print each letter in any case (YES, yes, Yes will all be recognized as positive answer, NO, no and nO will all be recognized as negative answer).

Input

abab ba abab

bb

aaa aaaa aaba

Output

YE NO NO

Note

Consider the example test from the statement.

In order to obtain "ba" from "ababa", you may press Backspace instead of typing the first and the fourth characters.

There's no way to obtain "bb" while typing "ababa".

There's no way to obtain "aaaa" while typing "aaa".

In order to obtain "ababa" while typing "aababa", you have to press Backspace instead of typing the first character, then type all the remaining characters.

1

AlphaCode is presented with a problem, in this case to figure out if it's possible to convert one phrase to another by pressing backspace instead of typing some letters.





Solution (output)

Deep Learning : 01-Introduction [47]

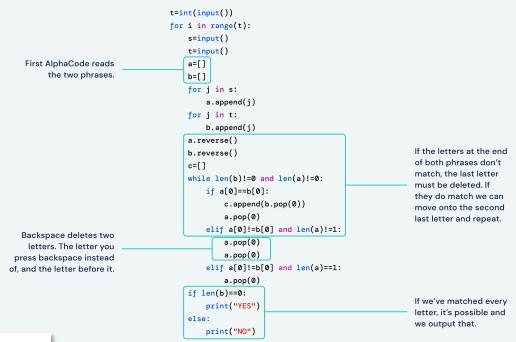
AlphaCode (2022)

Solving coding problems using transformers



2

Solution (output)



(2)

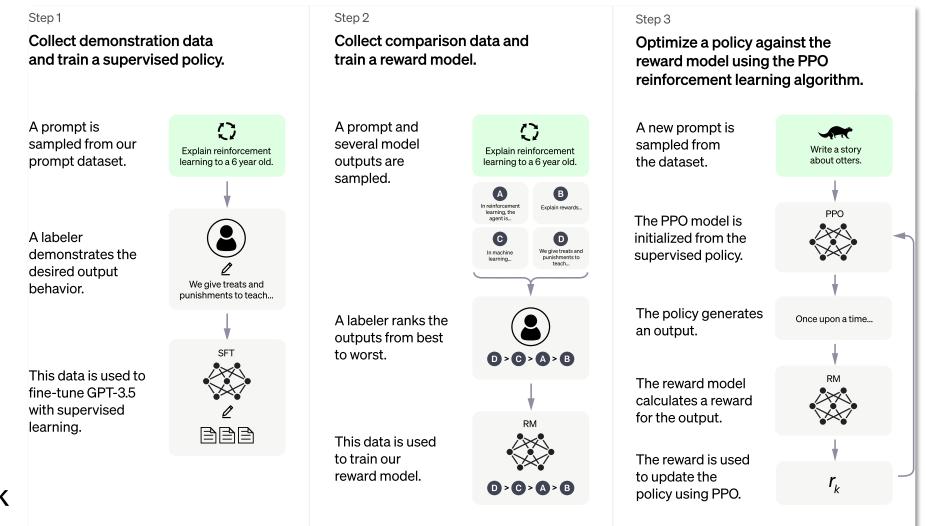
AlphaCode reads the whole problem statement and produces code, analogous to how a human would approach the problem by reading it, coding a solution, and submitting.

Deep Learning: 01-Introduction [48]

Chat GPT (2023)



Putting together all the tricks in the book (as known so far)



Deep Learning: 01-Introduction [49]

OK, about this course ...

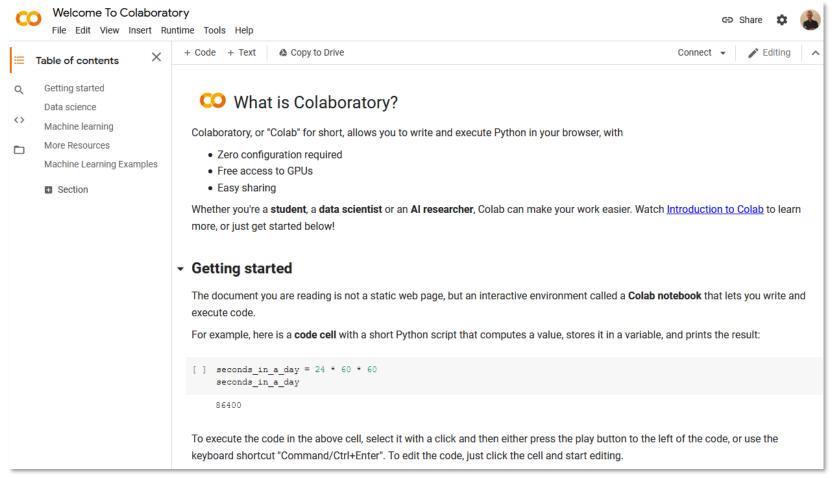
Deep Learning: 01-Introduction [50]

This Course in a Nutshell

- 1) Fundamentals
- 2) Deep Supervised Learning
- 3) Deep Convolutional Neural Networks
- 4) Deep Recurrent Networks
- 5) Deep Reinforcement Learning

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Labs with Google Colab



Make sure you have a look!

<u>Better yet</u>: follow the tutorial at https://colab.research.google.com/notebooks/intro.ipynb

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PyCharm (optional)

However, if you really want to see how things work...

Set it up on your computer

Python 3.8+ Numpy 1.19+ TensorFlow 2.+

... and, above all ...

Install PyCharm (Community Edition)

https://www.jetbrains.com/pycharm/

It's free

Learn using the <u>debugger</u>:

it will change the way you learn (and code)

CAUTION: no assistance whatsoever will be provided for this ... "Legions of students succeeded in doing this on their own: you can make it too!"



The Python IDE for Professional Developers

DOWNLOAD

Full-fledged Professional or Free Community

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The Final Exam

1) Choose and propose a final project

The topic could be any of your choice
The techniques adopted must be (strongly) related with the course
Groupwork is allowed, with a maximum of two (no exceptions)

2) Develop and submit your projects

Each final project must be submitted as a Google Colab notebook plus dataset (*if required*) Submission must be made at least <u>two</u> days before the exams

3) Be strong on theory, nonetheless

Alone, not even a *phenomenal* final project will give you a final grade (sorry) Appropriate knowledge about theory will be fundamental

4) Final Exam

Oral interview, about both final project and theory Relative weights: final project 40%, theory 60%

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Some References Books for the Course

Deep Learning

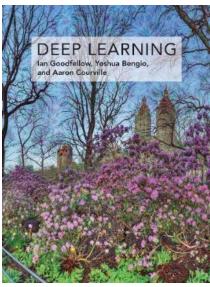
Ian Goodfellow, Yoshua Bengio and Aaron Courville *MIT Press, 2017*

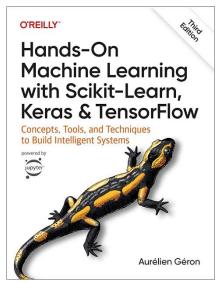
See also

http://www.deeplearningbook.org/

 Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow: Concepts, Tools, and Techniques to Build Intelligent Systems, 3rd Edition

Aurélien Géron O'Reilly, 2022





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