

# Advanced Methods in Data Analysis II

## Introduction

### A.I. Machine learning and Deep learning

Fabián Castiblanco

2021

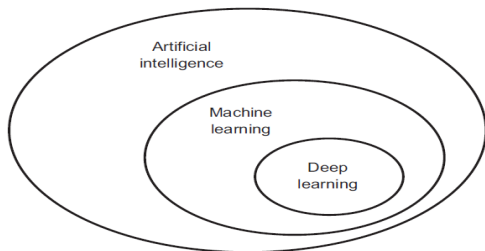
# Contents

- 1 A.I. M.L. D.L. How do they relate to each other?
- 2 Learning in machine learning
- 3 Deep Learning
- 4 A brief history of machine learning
- 5 Back to neural networks

# Contents

- 1 A.I. M.L. D.L. How do they relate to each other?
- 2 Learning in machine learning
- 3 Deep Learning
- 4 A brief history of machine learning
- 5 Back to neural networks

# A.I. M.L. D.L. How do they relate to each other?

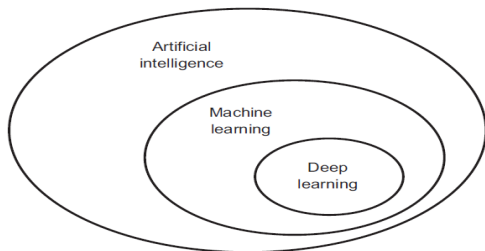


**Artificial intelligence,  
machine learning, and deep learning**

Source. Chollet (2018). Deep Learning with Python

- What is meant by learning in machine learning?

# A.I. M.L. D.L. How do they relate to each other?



**Artificial intelligence,  
machine learning, and deep learning**

Source. Chollet (2018). Deep Learning with Python

- What is meant by learning in machine learning?
- What does "deep" mean in Deep Learning?

# Contents

- 1 A.I. M.L. D.L. How do they relate to each other?
- 2 Learning in machine learning
- 3 Deep Learning
- 4 A brief history of machine learning
- 5 Back to neural networks

# Learning in machine learning

- A machine-learning system is trained rather than explicitly programmed.

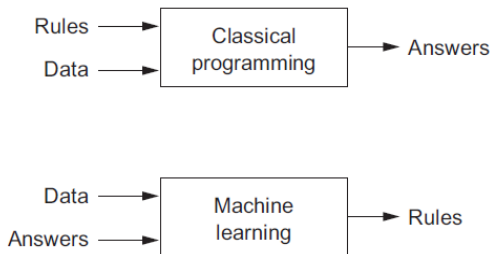
# Learning in machine learning

- A machine-learning system is trained rather than explicitly programmed.
- Machine learning discovers rules to execute a data-processing task, given examples of what's expected



# Learning in machine learning

- A machine-learning system is trained rather than explicitly programmed.
- Machine learning discovers rules to execute a data-processing task, given examples of what's expected



**Machine learning:  
a new programming paradigm**

Source. Chollet (2018). Deep Learning with Python

# Learning in machine learning

To do machine learning, we need three things:

- Input data points

# Learning in machine learning

To do machine learning, we need three things:

- Input data points
- Examples of the expected output

# Learning in machine learning

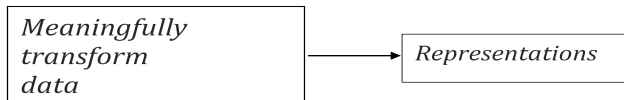
To do machine learning, we need three things:

- Input data points
- Examples of the expected output
- A way to measure whether the algorithm is doing a good job

# Learning in machine learning

To do machine learning, we need three things:

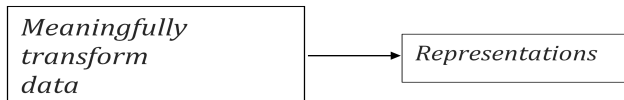
- Input data points
- Examples of the expected output
- A way to measure whether the algorithm is doing a good job



# Learning in machine learning

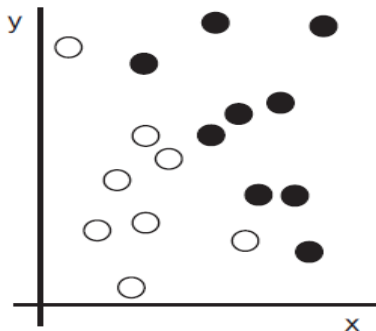
To do machine learning, we need three things:

- Input data points
- Examples of the expected output
- A way to measure whether the algorithm is doing a good job



- Machine-learning models are all about finding appropriate representations for their input data

## Representations



We want to develop an algorithm that can take the coordinates  $(x, y)$  of a point and output whether that point is likely to be black or to be white.

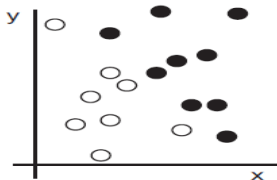
# Learning in machine learning

**Would be a coordinate change**

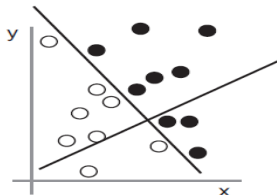


# Learning in machine learning

Would be a coordinate change



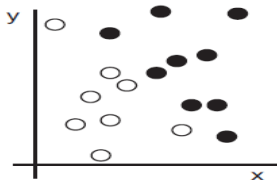
(a) Raw data.



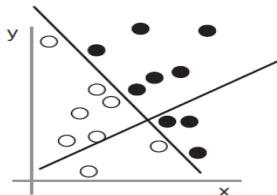
(b) Coordinate change

# Learning in machine learning

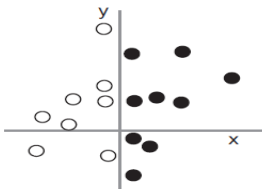
Would be a coordinate change



(a) Raw data.



(b) Coordinate change



(c) Better representation

# Learning in machine learning

All machine-learning algorithms consist of automatically finding such transformations that turn data into more-useful representations for a given task.

- Coordinate changes

# Learning in machine learning

All machine-learning algorithms consist of automatically finding such transformations that turn data into more-useful representations for a given task.

- Coordinate changes
- Linear projections (which may destroy information)

# Learning in machine learning

All machine-learning algorithms consist of automatically finding such transformations that turn data into more-useful representations for a given task.

- Coordinate changes
- Linear projections (which may destroy information)
- Translations

# Learning in machine learning

All machine-learning algorithms consist of automatically finding such transformations that turn data into more-useful representations for a given task.

- Coordinate changes
- Linear projections (which may destroy information)
- Translations
- Nonlinear operations

# Learning in machine learning

All machine-learning algorithms consist of automatically finding such transformations that turn data into more-useful representations for a given task.

- Coordinate changes
- Linear projections (which may destroy information)
- Translations
- Nonlinear operations

Predefined set of operations  $\rightarrow$  Hypothesis space.

# Learning in machine learning

All machine-learning algorithms consist of automatically finding such transformations that turn data into more-useful representations for a given task.

- Coordinate changes
- Linear projections (which may destroy information)
- Translations
- Nonlinear operations

Predefined set of operations  $\rightarrow$  Hypothesis space.

## Definition

*Searching for useful representations of some input data, within a predefined space of possibilities, using guidance from a feedback signal*



# Contents

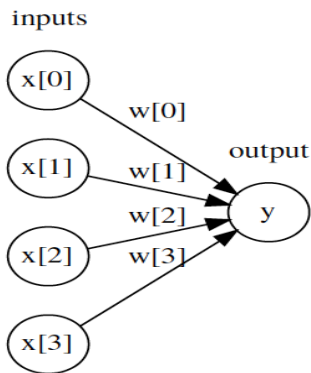
- 1 A.I. M.L. D.L. How do they relate to each other?
- 2 Learning in machine learning
- 3 Deep Learning**
- 4 A brief history of machine learning
- 5 Back to neural networks

# Deep Learning and neural networks

First idea

# Deep Learning and neural networks

First idea



$$y = w_1x_1 + \dots + w_dx_d + b$$

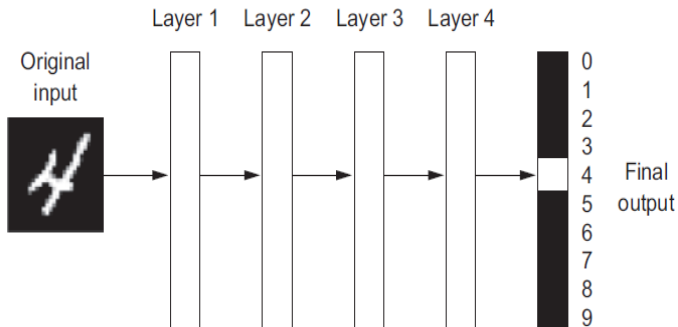
**What does "deep" mean in Deep Learning?**

## What does "deep" mean in Deep Learning?

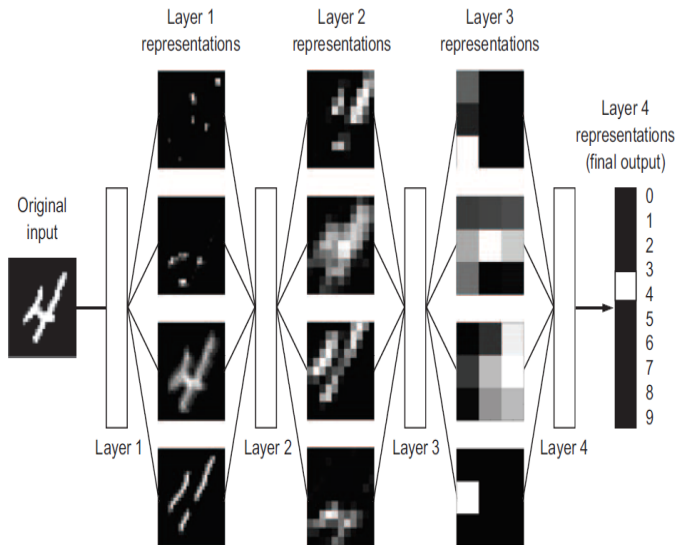
- Learning successive layers of increasingly meaningful representations

## What does "deep" mean in Deep Learning?

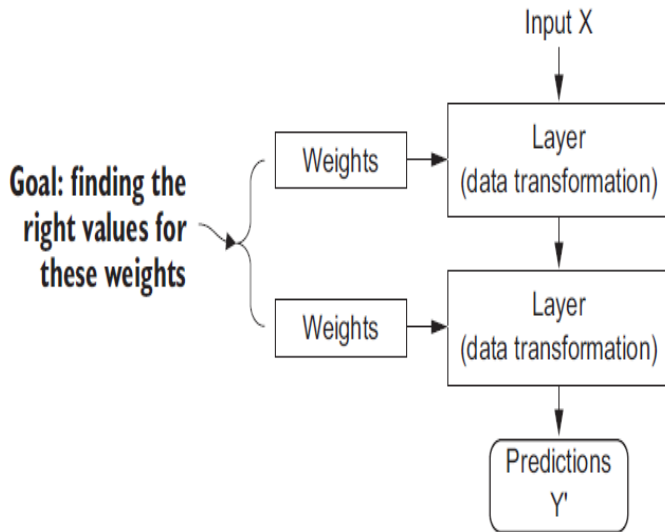
- Learning successive layers of increasingly meaningful representations



# Deep Learning

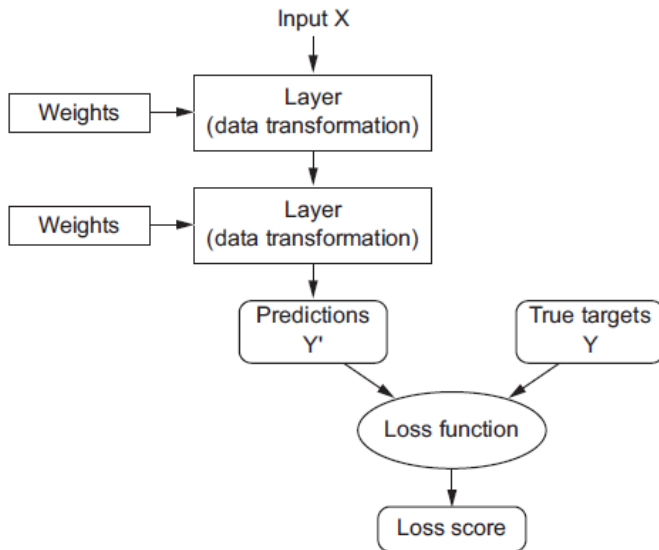


# Deep Learning

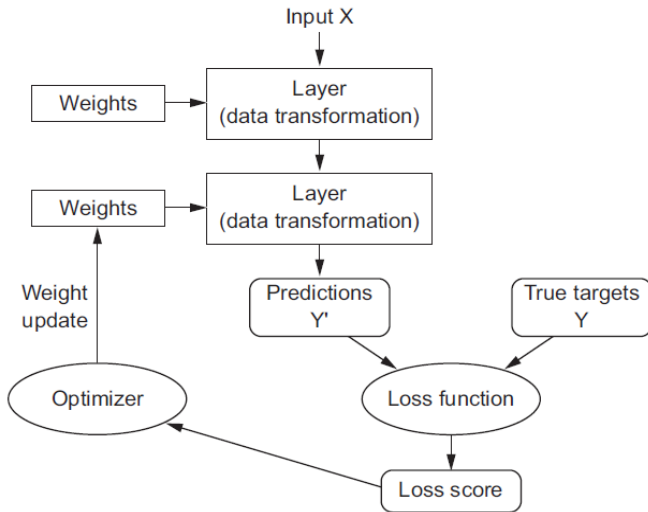




# Deep Learning



# Deep Learning



# Contents

- 1 A.I. M.L. D.L. How do they relate to each other?
- 2 Learning in machine learning
- 3 Deep Learning
- 4 A brief history of machine learning**
- 5 Back to neural networks

# A brief history

Deep learning isn't always the right tool for the job—sometimes there isn't enough data for deep learning to be applicable, and sometimes the problem is better solved by a different algorithm

Approach	algorithm	History
Probabilistic modeling	Naive Bayes logistic regression	The features in the input data are all independent
Early neural networks	LeNet	Emergence of SVM
Kernel methods	SVM	Hard to scale to large datasets
Decision trees	Decision trees, random forests, and gradient boosting machines	Weak prediction models

# Contents

- 1 A.I. M.L. D.L. How do they relate to each other?
- 2 Learning in machine learning
- 3 Deep Learning
- 4 A brief history of machine learning
- 5 Back to neural networks

# Back to neural networks

- ImageNet

# Back to neural networks

- ImageNet
- Deep convolutional neural networks

# Back to neural networks

- ImageNet
- Deep convolutional neural networks
- Computer vision tasks. All *perceptual* tasks



# Back to neural networks

- ImageNet
- Deep convolutional neural networks
- Computer vision tasks. All *perceptual* tasks
- Offered better performance on many problems

# Back to neural networks

- ImageNet
- Deep convolutional neural networks
- Computer vision tasks. All *perceptual* tasks
- Offered better performance on many problems
- Feature engineering

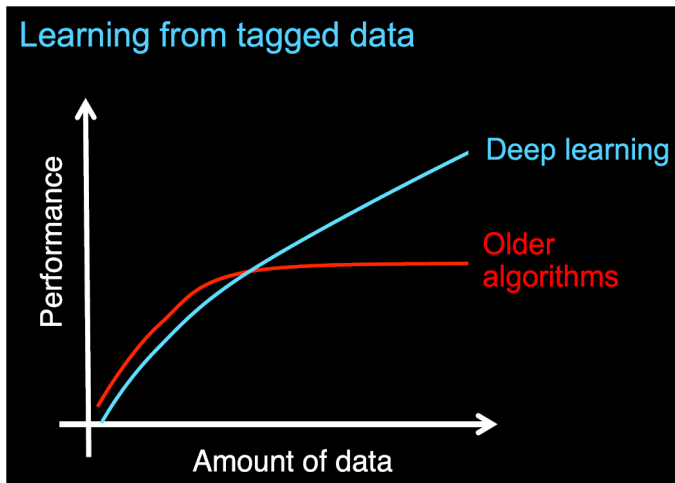
# Back to neural networks

- ImageNet
- Deep convolutional neural networks
- Computer vision tasks. All *perceptual* tasks
- Offered better performance on many problems
- Feature engineering
- Incremental, layer-by-layer way in which increasingly complex representations are developed

# Back to neural networks

- ImageNet
- Deep convolutional neural networks
- Computer vision tasks. All *perceptual* tasks
- Offered better performance on many problems
- Feature engineering
- Incremental, layer-by-layer way in which increasingly complex representations are developed
- Intermediate incremental representations are learned jointly

# Back to neural networks



(<http://cs229.stanford.edu/materials/CS229-DeepLearning.pdf>)

# Back to neural networks

- Near-human-level image classification
- Near-human-level speech recognition
- Near-human-level handwriting transcription
- Improved machine translation
- Improved text-to-speech conversion
- Digital assistants such as Google Now and Amazon Alexa
- Near-human-level autonomous driving
- Improved ad targeting, as used by Google, Baidu, and Bing
- Improved search results on the web
- Ability to answer natural-language questions

## KAGGLE

Kaggle was dominated by two approaches: gradient boosting machines and deep learning. Specifically, gradient boosting is used for problems where structured data is available, whereas deep learning is used for perceptual problems such as image classification. Practitioners of the former almost always use the excellent XGBoost library, which offers support for the two most popular languages of data science: Python and R. Meanwhile, most of the Kaggle entrants using deep learning use the Keras library, due to its ease of use, flexibility, and support of Python.

- Chollet. Deep learning with Python. Manning, 2018.
- Ch. Aggarwal. Recommender Systems. Springer, 2016.
- Goodfellow, Y. Bengio, A. Courville. Deep learning. MIT press, 2016.<https://www.deeplearningbook.org/>
- M. Nielsen. Neural Networks and Deep Learning, 2019.  
<http://neuralnetworksanddeeplearning.com/>