

Deep Learning: definition and remarks

Carlos Perales-González¹

¹Universidad Loyola Andalucía

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Overview

Introduction

Differences with Machine Learning

How to program Machine Learning?

How does a Neural Network work?

What is Jupyter Notebook?

Code

Now your job starts

What is Machine Learning?(I)

This field of knowledge within Computer Science explores how computers can perform tasks on their own, rather than programming them with sets of rules about how to act[1]. More specifically, Machine Learning studies what can be predicted through the data, with a series of modeling hypotheses, and with what reliability.

What is Machine Learning? (II)

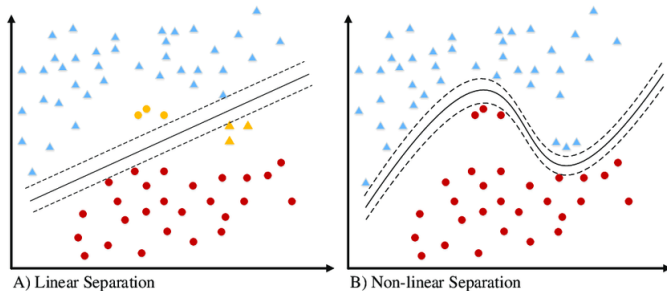


Figure: SVM linear and kernel for a binary classification problem

What is Deep Learning?

While Machine Learning task is to infer statistical rules from data, Deep Learning focus on extract features from the original data variables, in order to outperform classical machine learning algorithms.

Same, but different.

Machine Learning vs Deep Learning (I)

Is a Multilayer Neural Network already a Deep Learning architecture? Still controversy of what is Deep Learning and what is Machine Learning. Here is a proposal:

DEEP LEARNING IS
FOCUSED ON FEATURES

Machine Learning vs Deep Learning (II)

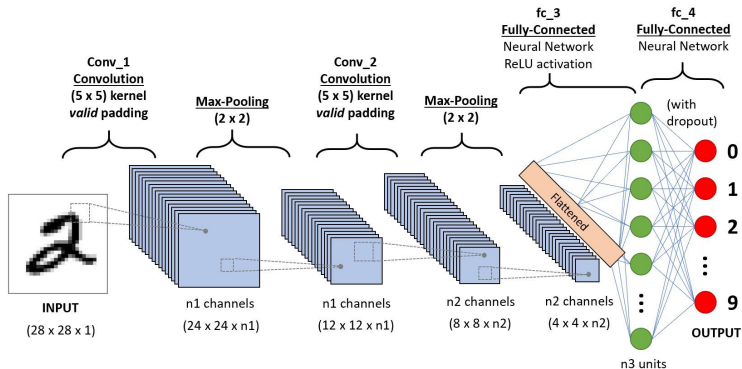


Figure: MNIST problem with Deep Learning

Machine Learning vs Deep Learning (III)

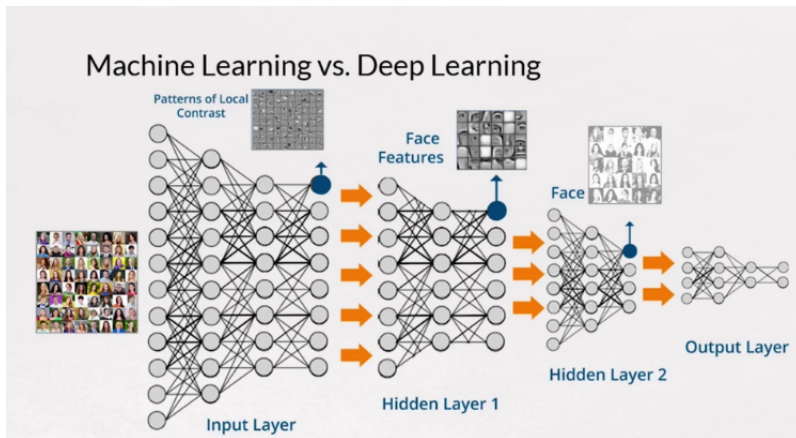


Figure: Multilayer Deep Learning

Machine Learning vs Deep Learning (IV)

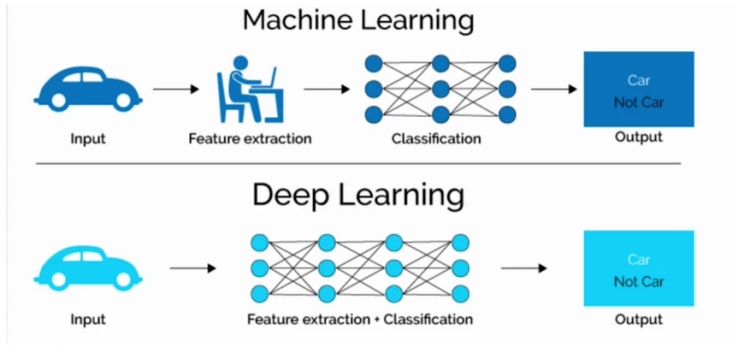


Figure: Machine Learning vs Deep Learning

Historical Advances

1. **1960s** Perceptron.
2. **1970s** Multilayer.
3. **1991** Autoencoder by Kramer et al.
4. **1997** Long short-term memory (LSTM), an RNN with feedback connections, Schmidhuber et al.
5. **1998** Convolutional Neural Networks by Yann LeCun et al.
6. **2012** Deep Convolutional Neural Networks (AlexNet) by Alex Krizhevsky et al.
7. **2014** Generative adversarial network by Ian Goodfellow.

Several languages, several libraries. **We will work in Python here.**

- **Mathematical representation** in Python, strong math libraries (*numpy*[2], *scipy*[3], ...).
- **Much work already done** in Python (*Sklearn*[4], *Tensorflow* / *Keras*[5], *PyTorch*[6], ...).
- **Object Oriented Programming** (OOP), more freedom to do stuff (*Have you worked with Sklearn?*).
- **APIs** for ulterior work (Twitter, Facebook, ...).

You already know this. So let's see some visual example

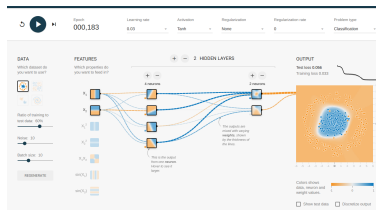


Figure: Playground of Tensorflow library

<https://playground.tensorflow.org/>

- The Jupyter Notebook is an open-source web application that allows you to create and share documents that contain live code.
- It saves the results of the code already run.
- It allows to mix real text, equations and code.

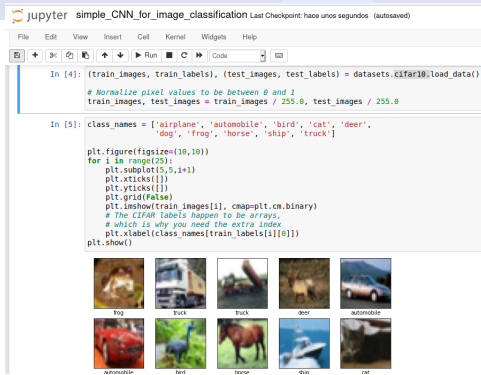


Figure: How Jupyter Notebook looks like.

Let's do some code!

- Create your own Python environment.
- Install Jupyter and Tensorflow.
- Explore example with *CIFAR10* dataset[7].

https://github.com/cperales/tensorflow_examples

MNIST dataset

- Create your own Python environment.
- Install Jupyter and Tensorflow.
- Download MNIST dataset.
- Play with Tensorflow.

Overview of datasets in Tensorflow:

<https://www.tensorflow.org/datasets/overview>

References I



T. M. Mitchell, *The discipline of machine learning*, vol. 9. Carnegie Mellon University, School of Computer Science, Machine Learning . . . , 2006.



C. R. Harris, K. J. Millman, S. J. van der Walt, R. Gommers, P. Virtanen, D. Cournapeau, E. Wieser, J. Taylor, S. Berg, N. J. Smith, R. Kern, M. Picus, S. Hoyer, M. H. van Kerkwijk, M. Brett, A. Haldane, J. F. del R'io, M. Wiebe, P. Peterson, P. G'erard-Marchant, K. Sheppard, T. Reddy, W. Weckesser, H. Abbasi, C. Gohlke, and T. E. Oliphant, "Array programming with NumPy," *Nature*, vol. 585, pp. 357–362, Sept. 2020.

References II



P. Virtanen, R. Gommers, T. E. Oliphant, M. Haberland, T. Reddy, D. Cournapeau, E. Burovski, P. Peterson, W. Weckesser, J. Bright, S. J. van der Walt, M. Brett, J. Wilson, K. J. Millman, N. Mayorov, A. R. J. Nelson, E. Jones, R. Kern, E. Larson, C. J. Carey, Í. Polat, Y. Feng, E. W. Moore, J. VanderPlas, D. Laxalde, J. Perktold, R. Cimrman, I. Henriksen, E. A. Quintero, C. R. Harris, A. M. Archibald, A. H. Ribeiro, F. Pedregosa, P. van Mulbregt, and SciPy 1.0 Contributors, “SciPy 1.0: Fundamental Algorithms for Scientific Computing in Python,” *Nature Methods*, vol. 17, pp. 261–272, 2020.

References III



F. Pedregosa, G. Varoquaux, A. Gramfort, V. Michel, B. Thirion, O. Grisel, M. Blondel, P. Prettenhofer, R. Weiss, V. Dubourg, J. Vanderplas, A. Passos, D. Cournapeau, M. Brucher, M. Perrot, and E. Duchesnay, “Scikit-learn: Machine learning in Python,” *Journal of Machine Learning Research*, vol. 12, pp. 2825–2830, 2011.



M. Abadi, A. Agarwal, P. Barham, E. Brevdo, Z. Chen, C. Citro, G. S. Corrado, A. Davis, J. Dean, M. Devin, S. Ghemawat, I. Goodfellow, A. Harp, G. Irving, M. Isard, Y. Jia, R. Jozefowicz, L. Kaiser, M. Kudlur, J. Levenberg, D. Mané, R. Monga, S. Moore, D. Murray, C. Olah, M. Schuster, J. Shlens, B. Steiner, I. Sutskever, K. Talwar,

References IV

P. Tucker, V. Vanhoucke, V. Vasudevan, F. Viégas, O. Vinyals, P. Warden, M. Wattenberg, M. Wicke, Y. Yu, and X. Zheng, “TensorFlow: Large-scale machine learning on heterogeneous systems,” 2015.

Software available from [tensorflow.org](https://www.tensorflow.org).



A. Paszke, S. Gross, F. Massa, A. Lerer, J. Bradbury, G. Chanan, T. Killeen, Z. Lin, N. Gimeshein, L. Antiga, A. Desmaison, A. Kopf, E. Yang, Z. DeVito, M. Raison, A. Tejani, S. Chilamkurthy, B. Steiner, L. Fang, J. Bai, and S. Chintala, “Pytorch: An imperative style, high-performance deep learning library,” in *Advances in Neural Information Processing Systems 32* (H. Wallach, H. Larochelle,

References V

A. Beygelzimer, F. d'Alché-Buc, E. Fox, and R. Garnett, eds.), pp. 8024–8035, Curran Associates, Inc., 2019.



A. Krizhevsky, “Learning Multiple Layers of Features from Tiny Images,” ... *Science Department, University of Toronto, Tech. ...*, 2009.