TP "Python for Al": a crash course focused on linear regression models

Name: Andrei Zinovyev

E-mail: Andrei.Zinovyev.U900@gmail.com

Page: <u>https://auranic.github.io/teaching/2021-python_for_ai</u>

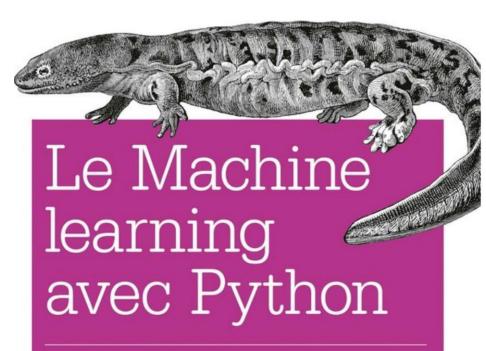
Python language



- Franca lingua for data science community today
- General purpose programming language (not specialized for a field like R)
- Rich set of standard libraries for data manipulation
- Rich set of standard libraries for numerical computations
- Rich set of standard libraries for machine learning
- Rich set of standard libraries for deep learning
- We will use Python 3!

What to read?

O'REILLY*



LA BIBLE DES DATA SCIENTISTS



Andreas C. Müller et Sarah Guido

Introduction au Machine Learning



Chloé-Agathe Azencott

DUNOD

Jupyter Python notebook

- To learn Python
- To write simple Python programs
- To develop complex Python program prototypes
- To exploit powerful computers remotely
- To do cloud computing
- To do data analysis and machine learning!
- To do data analysis without installing Python!



Colab: using Google computers and their GPUs, for your own projects

- Only a gmail account is needed
- Based on Jupyter notebooks
- Integrated with Google Drive
- Has Google machine learning products installed (such as TensorFlow)
- Free Cloud service with free GPU



Kaggle: open data scientist environment



- An online Community (3 million users) of Data Scientists and Machine Learners
- Enter competitions to solve data science challenges
- Data science online education
- Allows finding and publishing data sets

Other available services for online machine learning using Jupyter notebooks







Few Python libraries we will use in this TP

- pandas for manipulating data frames ('tables with headers' similar to Excel sheets)
- numpy for manipulating 1d and 2d arrays
- scipy for statistical calculations
- matplotlib for standard data plotting
- seaborn for easy data graphics with pandas
- scikit-learn for machine learning
- tensorflow for computing regression as a neural network
- lifelines for survival analysis

scikit-learn: the most popular library machine learning

• Almost any machine learning method, using the same code pattern

model = SomeModel(parameters)
model.fit(X,Y)
Y_predicted = model.predict(X)
model.score(X,Y)

- Excellent documentation: https://scikit-learn.org/
- Started and implemented in France by INRIA in 2010

tensorflow: Google open source library oriented to deep learning

- a free and open-source software library for machine learning
- Developed by Google Brain in 2017
- Based on <u>dataflow</u> and computational graph
- Based on <u>differentiable programming</u>
- Easiest way to use TensorFlow is by using Keras : a wrapper on top of tensorflow which allows one to construct deep learning models and fit them to data

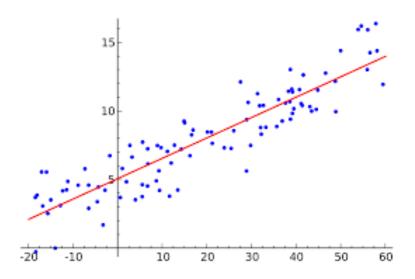


K Keras

Linear regression

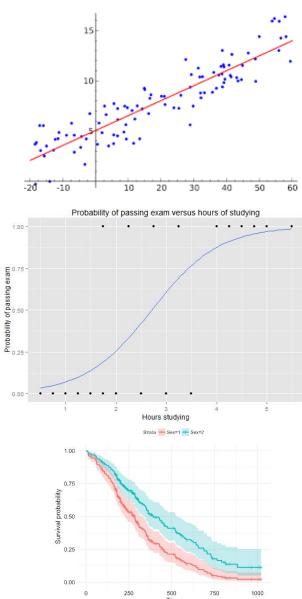
$$y_i = eta_0 + eta_1 x_{i1} + \dots + eta_p x_{ip} + arepsilon_i = \mathbf{x}_i^\mathsf{T} oldsymbol{eta} + arepsilon_i, \qquad i = 1, \dots, n,$$

- Connects many independent and one dependent variable via linear function
- Linear regression: the father of all supervised machine learning methods
- Linear regression: the most used machine learning method today
- Linear regression: the first machine learning method to apply, and see what it gives
- Linear regression: common problems and ways to deal with them (regularization)
- Linear regression can be used to produce nonlinear data models



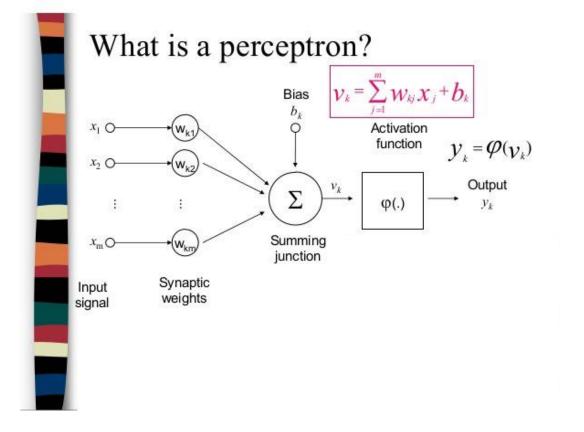
Three major flavors of linear regression in healthcare-related data science

- Simple Ordinary Least Square : when the target variable is continuous
- Logistic linear regression (logit): when the target variable is discrete (for example, binary)
- Survival Cox linear regression : when the target variable is a pair (follow up time + event)



Linear regression and a simple perceptron (formal neuron)

- Invented by Frank Rosenblatt in 1950s
- Elementary unit of any complex and deep neural network today



If $\phi(x) = x$, then it is simple linear regression model

If $\varphi(x)$ is a step-wise or sigmoidal function then it is a binary classifier just as logistic regression (even though they are trained with different algorithms!)

Linear regression is explainable ML model!

$$y_i = eta_0 + eta_1 x_{i1} + \dots + eta_p x_{ip} + arepsilon_i = \mathbf{x}_i^\mathsf{T} oldsymbol{eta} + arepsilon_i, \qquad i = 1, \dots, n,$$

Coefficients β_1 , β_2 , ..., β_p are comparable if independent variables are standardized (to z-scores) and have straingforward interpretation

It is possible to estimate statistical significance of β_i coefficients and provide p-value on the hypothesis that the coefficient is non-zero

This can help to simplify the regression

Other methods (such as regularization by lasso) for selecting important variables are readily avaialable

Objective of the TP: learn how to build simple regression models in Python

- Univariate and multi-variate regressions
- Validating the regression models in Python
- Building regression models using scikit-learn
- Building regression models using scipy
- Building regression models using tensorflow and train them using gradient-descent (similarly to building neural networks)

Plan of the TP

- Lesson I (~2.5 hours)
 - Connecting to Colab
 - Creating first Jupyter notebook
 - Working with a toy dataset
 - Visualizing simplest linear regression
 - Linear regression: using scikit-learn
 - Linear regression: using scipy statmodels
 - Loading a real-life dataset and exploring it
 - Simplifying the regression model

Plan of the TP

• Lesson II (~2.5 hours)

Testing and validating the linear regression

Implementing linear regression in tensorflow

- Basic things in tensorflow
- First computational graph
- Linear regression: using tensorflow and gradient descent

Ways to work on TP: all depends on your Python level/motivation!

- Some initial level of Python is assumed:
 - If not then a good start is at:

<u>https://wiki.python.org/moin/BeginnersGuide/NonProgrammers</u> <u>https://wiki.python.org/moin/BeginnersGuide/Programmers</u> <u>https://www.w3resource.com/python-exercises/python-basic-exercises.php</u>

- We will start with a simple warming up exercise
- We will have a basic code example on a tiny imaginary example
- Five real-life datasets from healthcare have been prepared (preprocessed) for this TP : try applying and validating linear regression on some of them
- In some tasks you will need to insert missing code indicated by '...'
- Do not hesitate to copy-paste (from anywhere, e.g. Stack Overflow)
- Do not hesitate to read Wikipedia