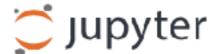
Linear Regression with Python

Tools

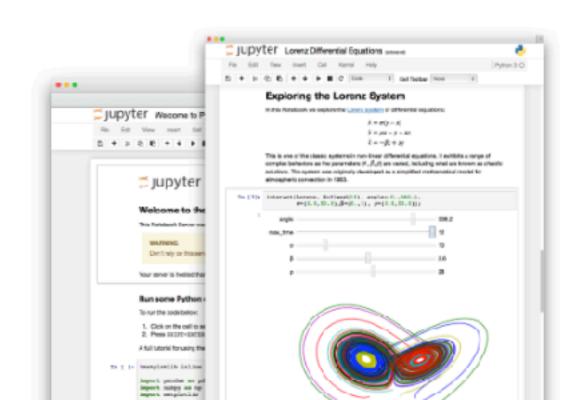
INSTALL

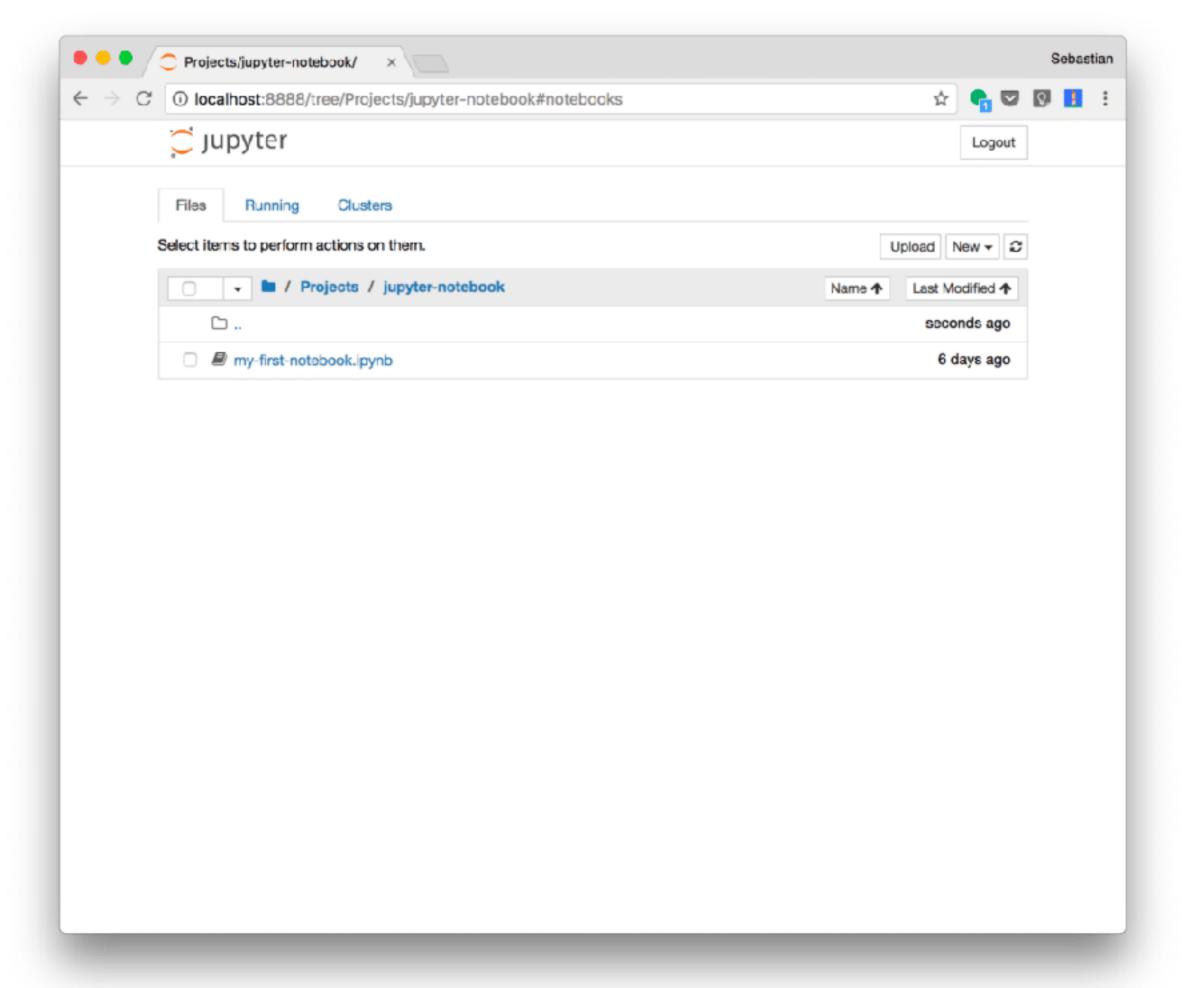


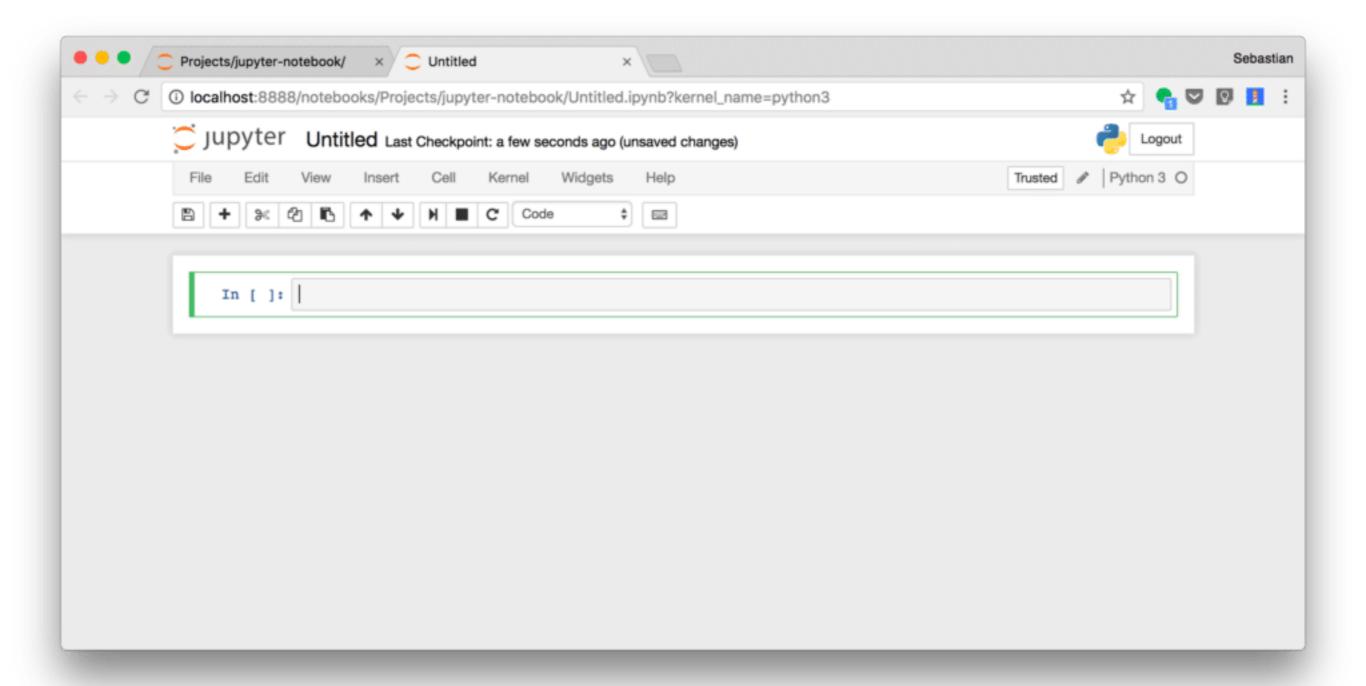


Jupyter Notebook

The Jupyter Notebook is a web-based interactive computing platform that allows users to author data- and code-driven narratives that combine live code, equations, narrative text, visualizations, interactive dashboards and other media.







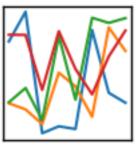
```
In [1]: print('Hello World')
Hello World

In [2]: i = 1
while i <= 10:
    print(i)
    i = i + 1

1
2
3
4
5
6
7
8
9
10</pre>
In []:
```









pandas

- open source library
- high-performance, easy-to-use data structures and data analysis tools

https://pandas.pydata.org/

Scipy.org

NumPy

NumPy is the fundamental package for scientific computing with Python. It contains among other things:

- a powerful N-dimensional array object
- · sophisticated (broadcasting) functions
- · tools for integrating C/C++ and Fortran code
- useful linear algebra, Fourier transform, and random number capabilities

Besides its obvious scientific uses, NumPy can also be used as an efficient multi-dimensional container of generic data. Arbitrary data-types can be defined. This allows NumPy to seamlessly and speedily integrate with a wide variety of databases.

NumPy is licensed under the BSD license, enabling reuse with few restrictions.

Getting Started

- · Getting NumPy
- Installing the SciPy Stack
- NumPy and SciPy documentation page
- NumPy Tutorial
- NumPy for MATLAB© Users
- · NumPy functions by category
- · NumPy Mailing List

For more information on the SciPy Stack (for which NumPy provides the fundamental array data structure), see scipy.org.

Donate to Numpy

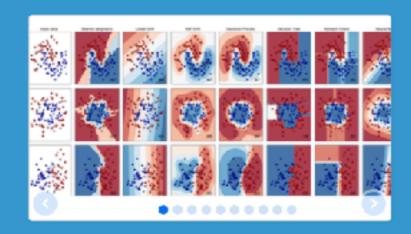
Open Hub

NumPy

About NumPy

License

Old array packages



scikit-learn

Machine Learning in Python

- · Simple and efficient tools for data mining and data analysis
- Accessible to everybody, and reusable in various contexts.
- Built on NumPy, SciPy, and matplotlib
- Open source, commercially usable BSD license

Classification

Identifying to which category an object belongs to.

Applications: Spam detection, Image recog-

Algorithms: SVM, nearest neighbors, random forest, ... — Examples

Regression

Predicting a continuous-valued attribute associated with an object.

Applications: Drug response, Stock prices. Algorithms: SVR, ridge regression, Lasso,

Examples

Clustering

Automatic grouping of similar objects into sets.

Applications: Customer segmentation,

Grouping experiment outcomes

Algorithms: k-Means, spectral clustering, mean-shift, ... — Examples

Dimensionality reduction

Reducing the number of random variables to consider.

Applications: Visualization, Increased efficiency

Algorithms: PCA, feature selection, nonnegative matrix factorization. — Examples

Model selection

Comparing, validating and choosing parameters and models.

Goal: Improved accuracy via parameter tun-

ing

Modules: grid search, cross validation, metrics. — Examples

Preprocessing

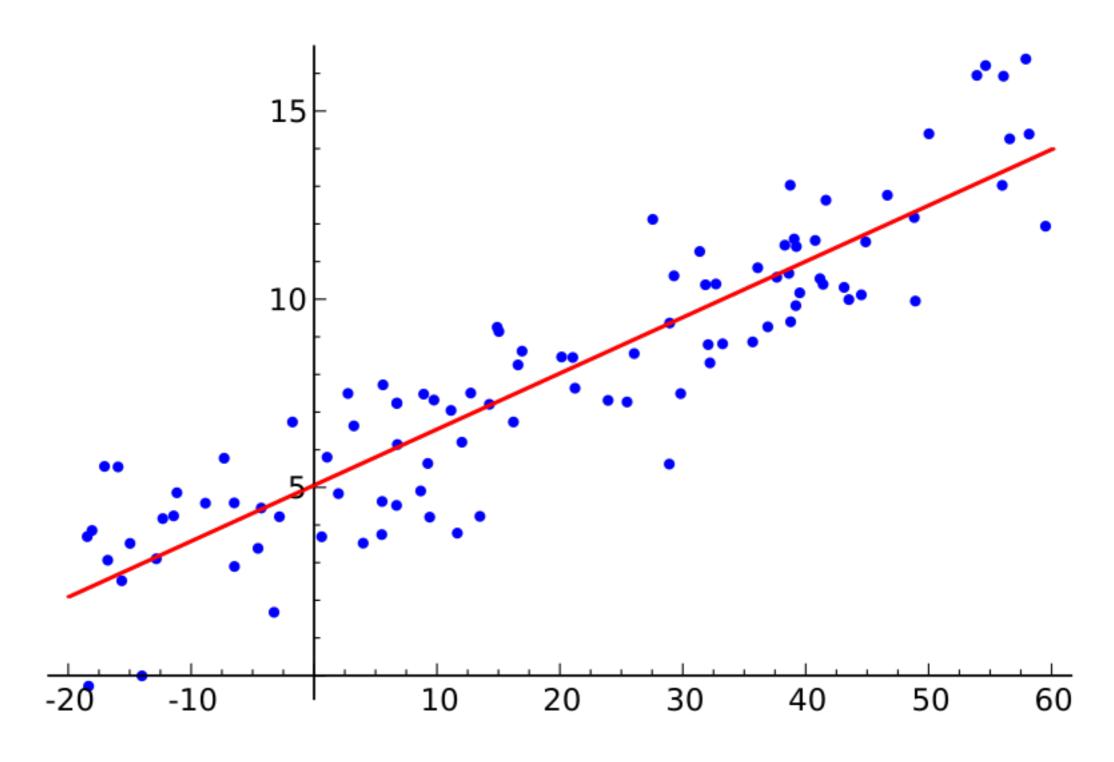
Feature extraction and normalization.

Application: Transforming input data such as text for use with machine learning algorithms. **Modules:** preprocessing, leature extraction.

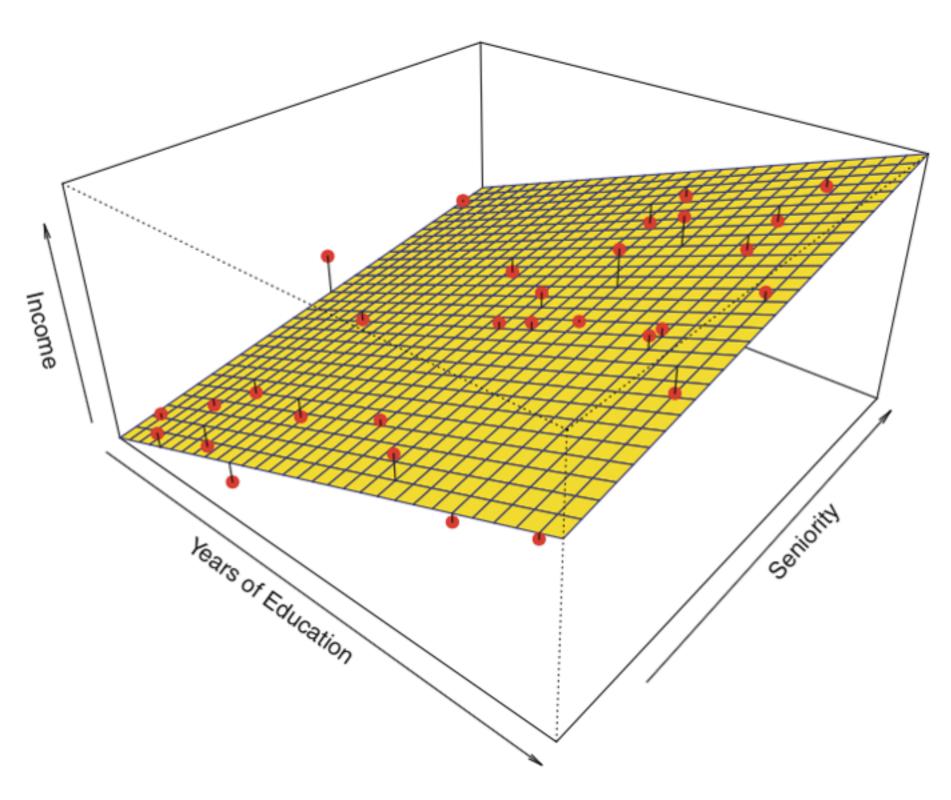
Examples

Linear Regression





y = a x + b y + c



http://sphweb.bumc.bu.edu/otlt/MPH-Modules/BS/R/R5_Correlation-Regression/R5_Correlation-Regression4.html

General formulation:

$$y(\mathbf{x}, \mathbf{w}) = w_0 + w_1 x_1 + \dots + w_D x_D$$

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The key property of this model is that it is a linear function of the parameters!

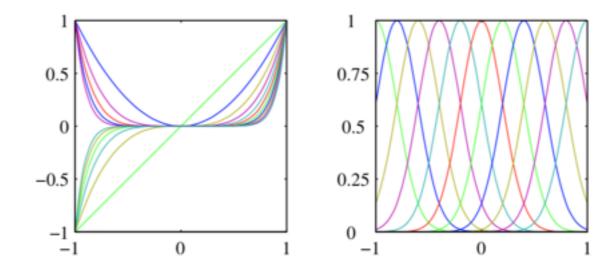
$$y(\mathbf{x},\mathbf{w})=w_0+\sum_{j=1}^{M-1}w_jx_j$$
 Linear in \mathbf{x} and \mathbf{w}
$$y(\mathbf{x},\mathbf{w})=w_0+\sum_{j=1}^{M-1}w_j\phi_j(\mathbf{x})$$
 Linear in \mathbf{w} and phi(\mathbf{x})

The key property of this model is that it is a linear function of the parameters!

$$y(\mathbf{x},\mathbf{w}) = w_0 + \sum_{j=1}^{M-1} w_j x_j$$
 Linear in \mathbf{x} and \mathbf{w}

$$y(\mathbf{x},\mathbf{w}) = w_0 + \sum_{j=1}^{M-1} w_j \overline{\phi_j(\mathbf{x})}$$
 Linear in w and phi(x)

Basis functions



$$y(\mathbf{x}, \mathbf{w}) = w_0 + \sum_{j=1}^{M-1} w_j \phi_j(\mathbf{x})$$

$$\phi_0(\mathbf{x}) = 1$$

 $y(\mathbf{x}, \mathbf{w}) = \mathbf{w}^{\top} \phi(\mathbf{x})$

$$y(\mathbf{x}, \mathbf{w}) = w_0 + \sum_{j=1}^{M-1} w_j x_j$$

- goal: Predicting a continuous y as a function of the variables x and the parameters w
- Model assumption:
 - data generated by $t = y(\mathbf{x}, \mathbf{w}) + \epsilon$
 - Error ∈ ~ normal distribution with zero mean
- Loss function: sum-of-squares error between t and prediction y (least squares)