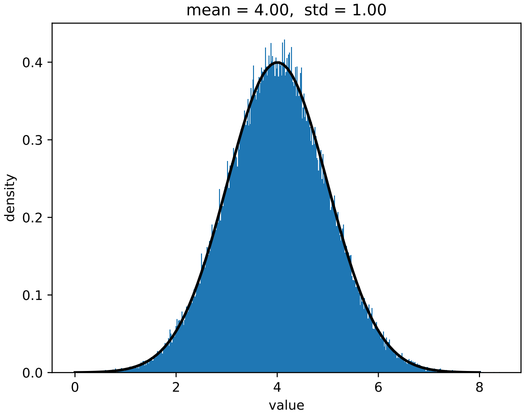
Machine learning II, unsupervised learning and agents: overview of mathematical tools

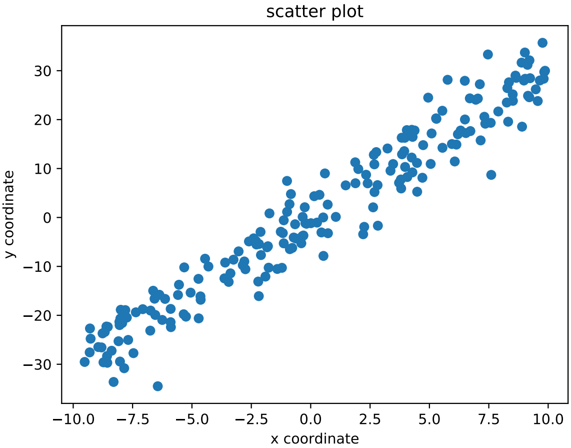


Probabilities and statistics

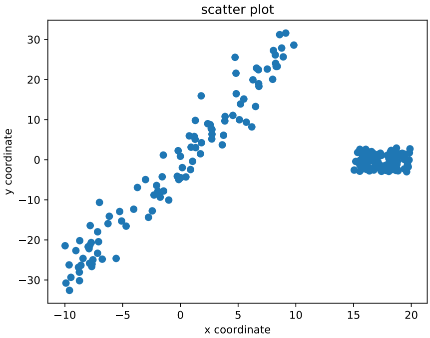
Optimization

To have a solid understanding of machine learning, it is necessary to be familiar with elementary probabilities and statistics.

Random variables



Random variables



We want to analyse how the data are **distributed.** For instance the x coordinate, the y coordinate.

Random variables

- (informal definition) A **random variable** is a quantity that can take several values, with some randomness.
- https://en.wikipedia.org/wiki/Random_variable

Random variables

- A **random variable** is a quantity that can take several values
- For instance :
  - the result of a dice throw

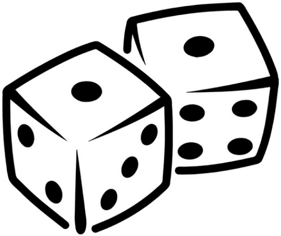


Figure – Dice

Random variables

- A **random variable** is a quantity that can take several values
- For instance :
  - the result of a dice throw
  - waiting time with RATP

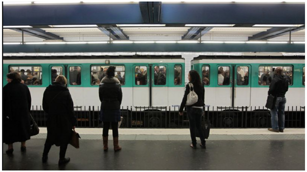


Figure – Some metro station

Random variables

- A **random variable** is a quantity that can take several values
- For instance :
  - the result of a dice throw
  - waiting time with RATP
  - weather

Figure – Weather in November

Random variables

- A **random variable** is a quantity that can take several values
- For instance :
  - the result of a dice throw
  - waiting time with RATP
  - weather
  - number of cars taking the periphérique at the same time

Why are random variables important ?

- most datasets encountered in machine learning can be considered as sampled from random variables.
- this is important for theoretical studies, and hence for applications : a better theoretical understanding of a problem allows to choose the best algorithm to solve it.
- theoretical results are sometimes precise in the sense that they allow to estimate the order of magnitude of the statistical error (e.g. the prediction error) as a function of d (dimension of the samples) and n (number of samples)
- a subdomain of machine learning is "statistical learning"

Random variables

- Some are random variables **continuous**, others **discrete**
- **continuous** : weather, RATP
- **discrete** : dice (6 possibilities), number of cars (> 10000)

Probability distributions

- A random variable is linked to a **probability distribution**, which is a function P
- It quantifies the probability of observing one outcome.
- For a discrete variable : each possible outcome is associated with a number between 0 and 1

Probability distributions

- For a dice game, the possible outcomes are in the set $\{1, 2, 3, 4, 5, 6\}$
- For a dice game : $P(1) = ?$ $P(2) = ?$ $P(3) = ?$ $P(4) = ?$ $P(5) = ?$ $P(6) = ?$

Probability distributions

- For a dice game, the possible outcomes are in the set $\{1, 2, 3, 4, 5, 6\}$
- For a dice game : $P(1) = \frac{1}{6}$, $P(2) = \frac{1}{6}$, $P(3) = \frac{1}{6}$, $P(4) = \frac{1}{6}$, $P(5) = \frac{1}{6}$, $P(6) = \frac{1}{6}$
- This is called a **uniform distribution**

Probability distributions

- Periphérique : probably a time-dependent very complicated distribution

Continuous variables

- The situation is different for continuous random variables.
- The distribution is given by a **probability density function**. Informally, the probably of being between x and $x + dx$ is $p(x)dx$.
- https://en.wikipedia.org/wiki/Probability_density_function
- Not that some variables are neither discrete nor continuous.

Uniform discrete

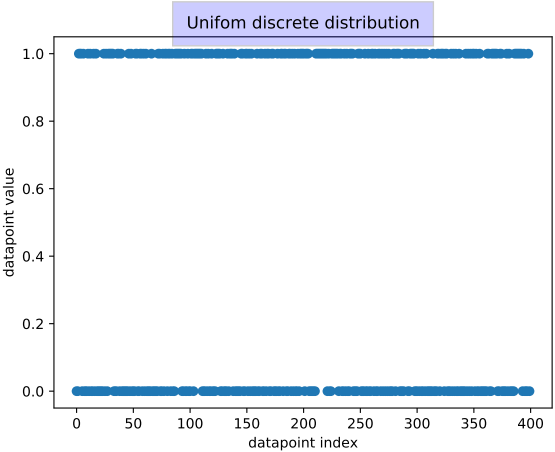


Figure – Uniform discrete distribution with 2 values

Uniform discrete

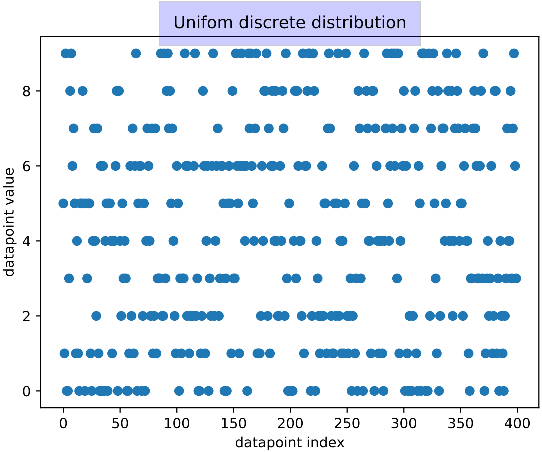


Figure – Uniform discrete distribution with 10 values

Bernoulli

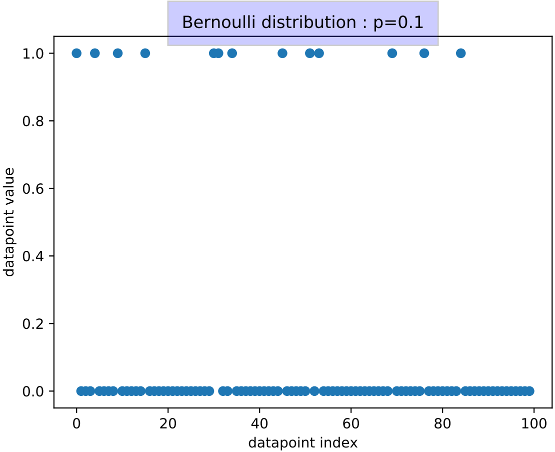


Figure – Bernoulli distribution

Bernoulli p

- With probability p, $X = 1$
- With probability $1 - p$, $X = 0$

Bernoulli

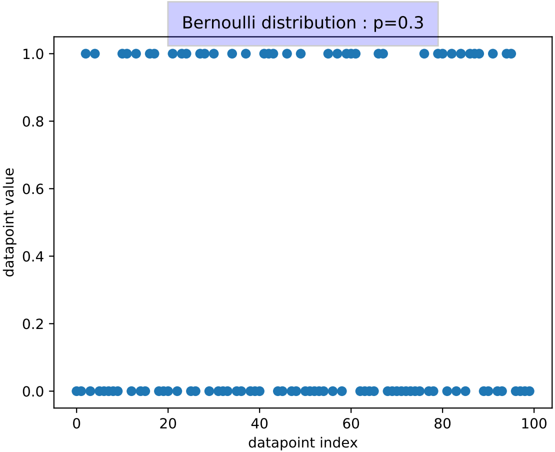


Figure – Bernoulli Distribution

Bernoulli

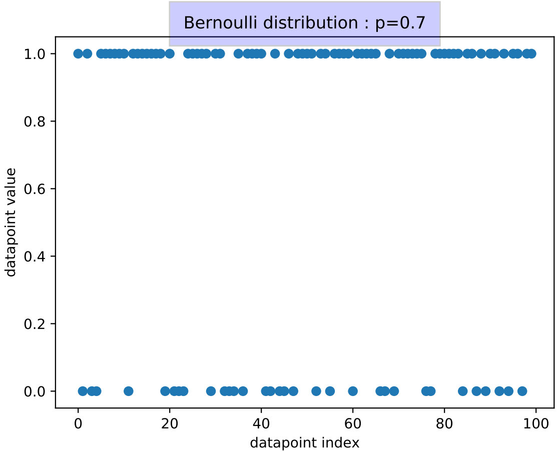


Figure – Bernoulli Distribution

Uniform continuous

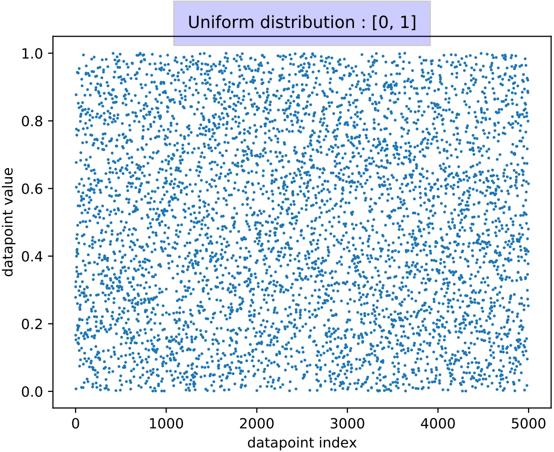


Figure – Uniform continuous distribution

Uniform continuous

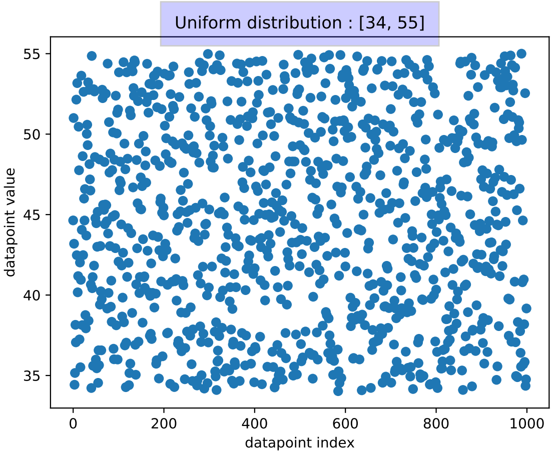


Figure – Uniform continuous distribution

Uniform continuous

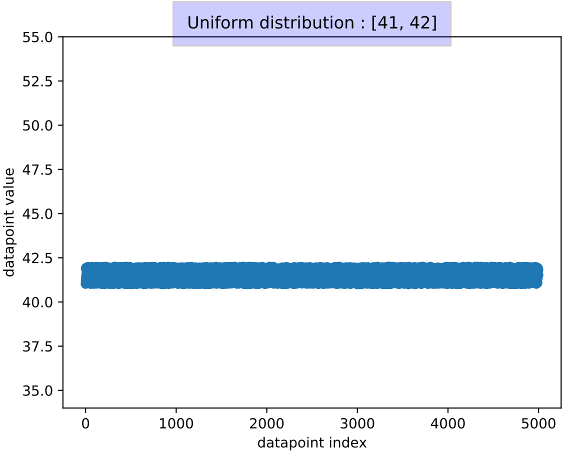


Figure – Uniform continuous distribution

Normal

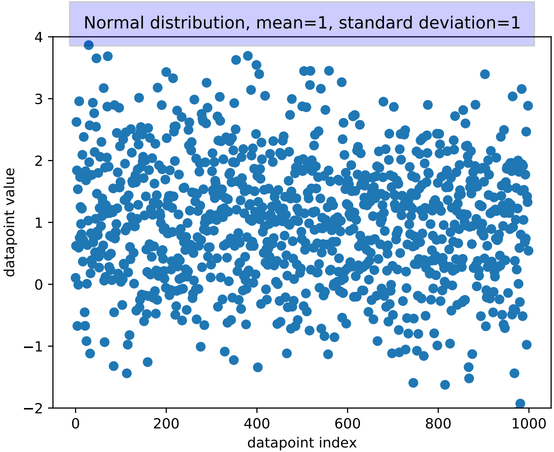


Figure – Normal distribution

Normal

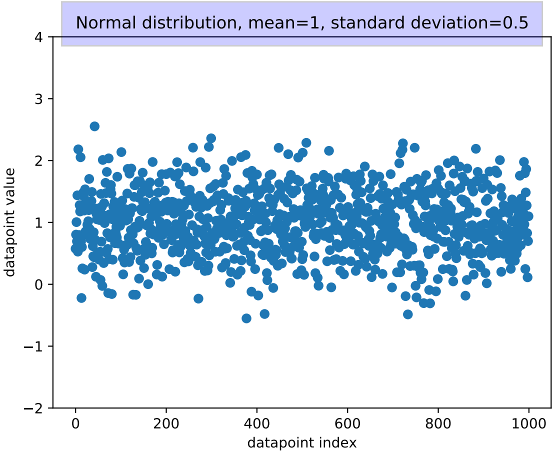


Figure – Normal distribution

Normal

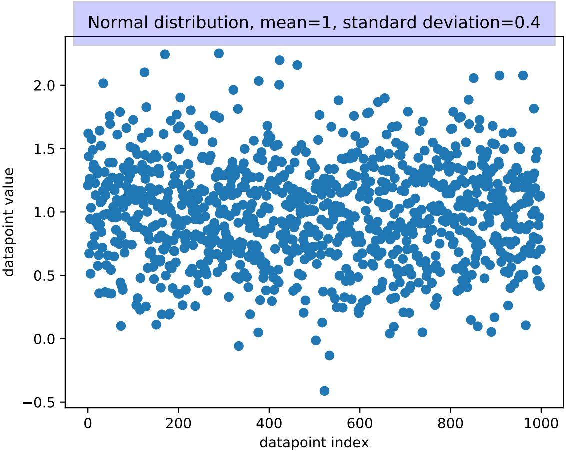


Figure – Normal distribution

White noise

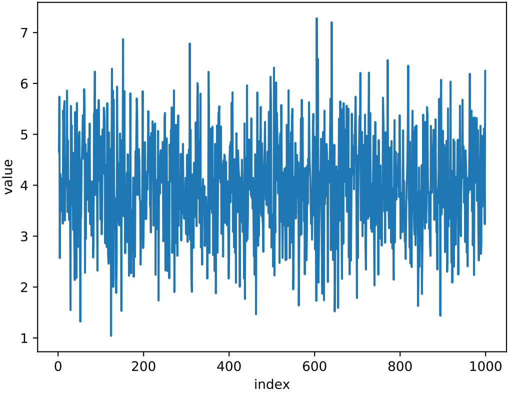


Figure – White noise

Histograms

Histograms are an alternative representation of the results of a (one-dimensional) random variable.

Uniform discrete

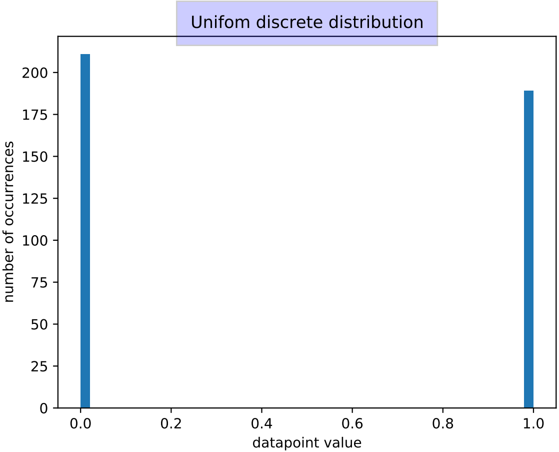


Figure – Historgram 1

Uniform discrete

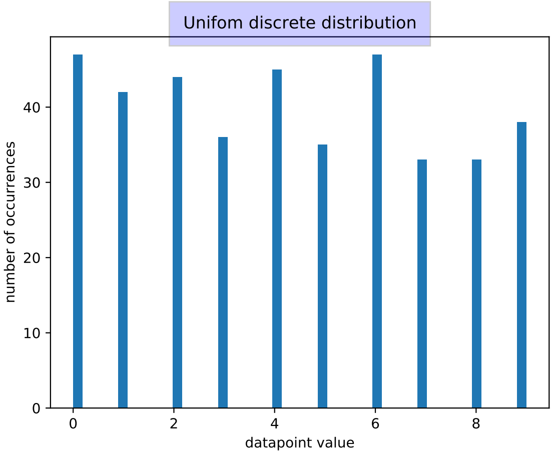


Figure – Historgram 1

Bernoulli

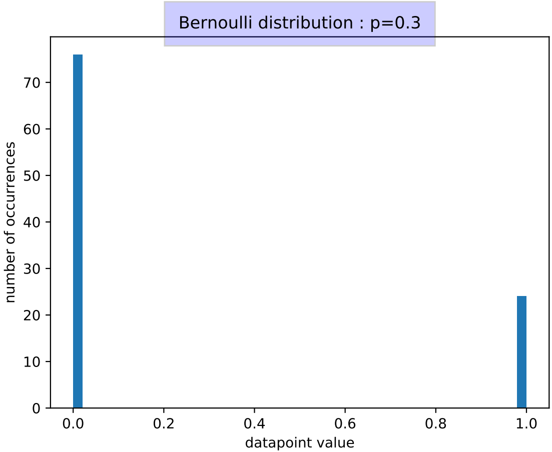


Figure – Historgram 2

Uniform continuous

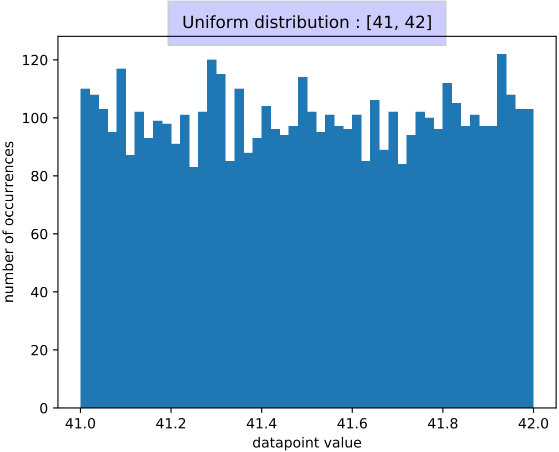


Figure – Historgram 3

Normal

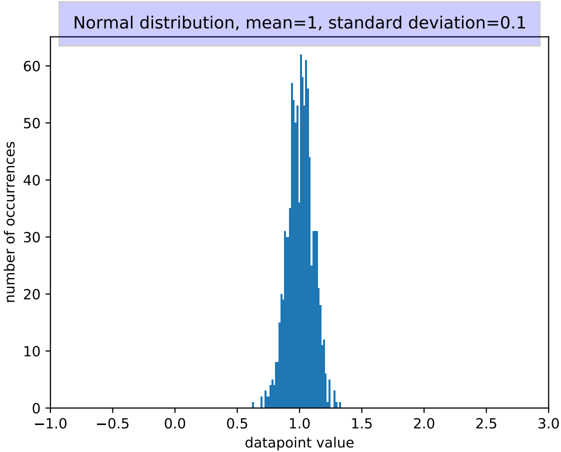


Figure – Historgram 4

Normal

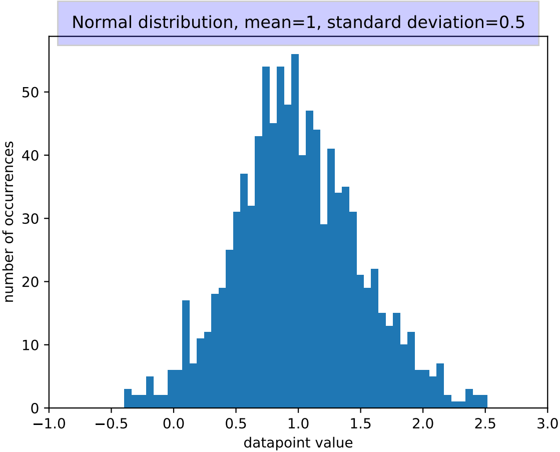


Figure – Historgram 4

cd distributions/
We can use the files **analyze_distribution_1.py** and **analyze_distribution_2.py** to analyze and plot some simple datasets, stored in **csv_files/**

Distribution 1

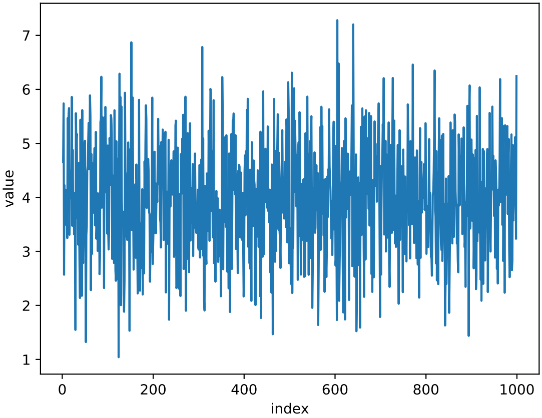


Figure – The data we analyze

histograms

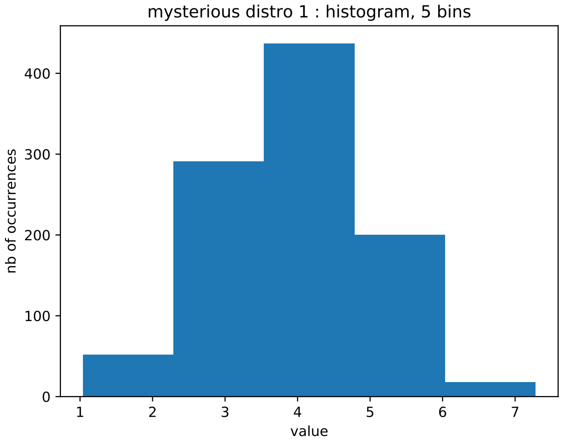


Figure – 5 bins

histograms

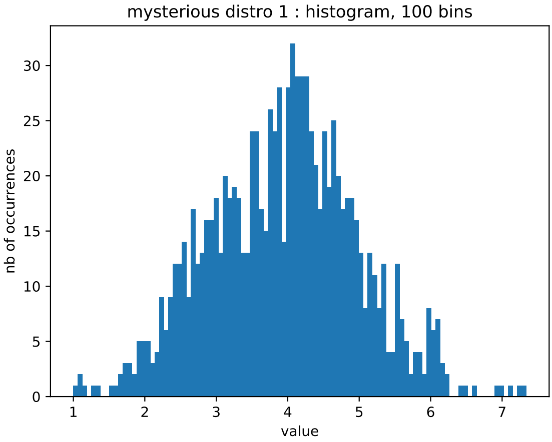


Figure – 100 bins

histograms

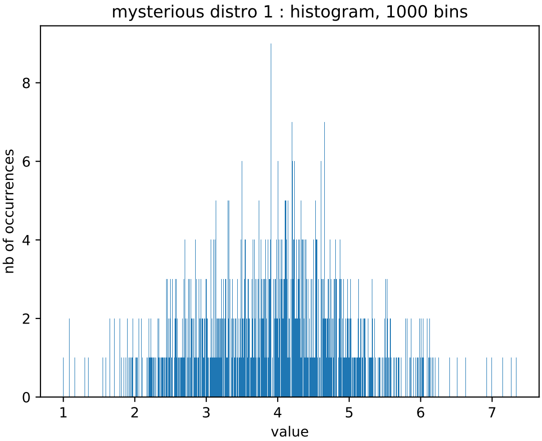


Figure – 1000 bins (too many)

Normal distribution

```python
import csv
import numpy as np

file_name = 'mysterious_distro_1.csv'

mean = 4
std_dev = 1
nb_point = 1000

with open('csv_files/' + file_name, 'w') as csvfile:
    filewriter = csv.writer(csvfile, delimiter=',')
    for point in range(1, nb_point):
        random_variable = np.random.normal(loc=mean, scale=std_dev)
        filewriter.writerow([str(point), str(random_variable)])
```

Figure – **create_normal.py** : Creation of the distribution

Second example

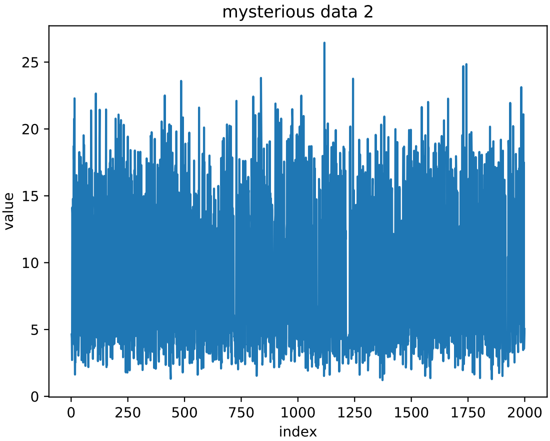


Figure – Second distribution

Exercice 1 : Create a one-dimensional dataset with a histogram that looks like this one !

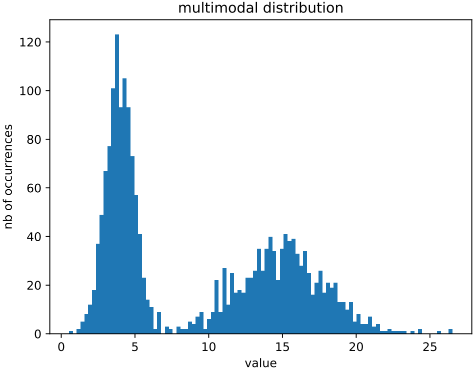


Figure – This distribution has several **modes**

Expected value

- The **expected value** of a random variable is its probablistic average.
- Under the condition that this probabilistic average is correctly defined.

Expected value (espérance)

- For a discrete random variable X that takes the values x_i with probability p_i :

$$E(X) = \sum_{i=1}^{n} p_i x_i \qquad (1)$$

- For a continuous random variable X with density p :

$$E(X) = \int xp(x)dx \qquad (2)$$

Note that X may have values in $\mathbb{R}^d$, with $d \geq 1$.

Expected value (espérance)

Exercice 2 : Computing an expected value

- For a discrete random variable X that takes the values x_i with probability p_i :

$$E(X) = \sum_{i=1}^{n} p_i x_i \tag{3}$$

- For a continuous random variable X with density :

$$E(X) = \int x p(x) dx \tag{4}$$

Compute the expected value of the dice game.

Variance

The variance is a measure of the dispersion of a random real variable.
https://en.wikipedia.org/wiki/Variance

$$var(X) = E\left(\left(X - E(X)\right)^2\right) \tag{5}$$

Note that we can also define the variance of a multidimensonial random variable (which means a random vector). In that case, it is a matrix.

Multidimensional vectors

We often consider random variables and data that live in spaces with a higher dimension than 2 (random vectors).

- images
- sensor that receives **multimodal information**

Correlation

Random vectors with correlated components are common statistical objects.

- In physics, temperature and pressure, measured by some sensores are correlated.
- In a dataset of customers of a company, some dimensions are likely to be correlted.

To study the statistical relationship between components, we can compute the **covariance** of the two components, or the **correlation**, (normalized covariance (see below)).
`https://en.wikipedia.org/wiki/Correlation`

Covariance

The covariance is a measure of the relationship between the variations of two random variables.

$$cov(X, Y) = E\Big((X - E(X))(Y - E(Y))\Big) \tag{6}$$

Correlation

The correlation is the covariance divided by the square roots of the variances.

$$corr(X, Y) = \frac{cov(X, Y)}{\sqrt{Var(X)}\sqrt{Var(Y)}} \tag{7}$$

Example

The data in **csv_files/distribution_3.csv** contain samples of a random variable with 5 dimensions (random vector). Some of these dimensions are correlated.

Covariance

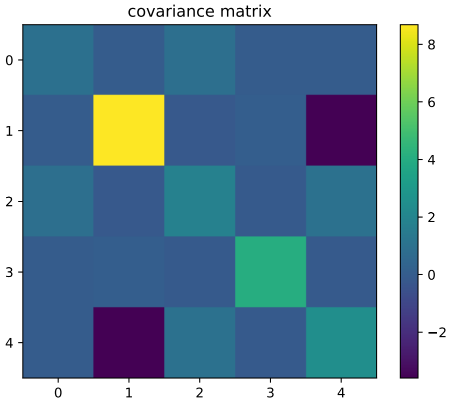


Figure – Covariance matrix of the random vector.

Correlation matrix

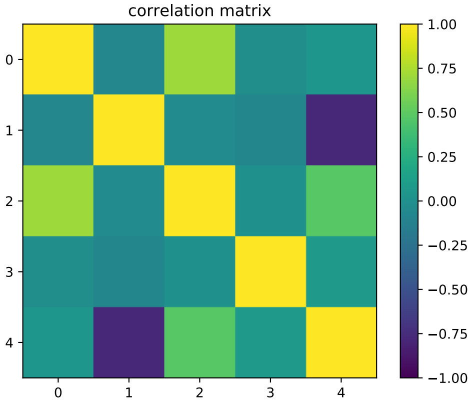


Figure – Correlation matrix for the distribution, note the difference in the scale.

Pandas, scikit-learn

- https://pandas.pydata.org/
- https://scikit-learn.org/stable/datasets/toy_dataset.html
- pandas demo with iris and distribution 3.

Minimization of a function

Optimization is another core aspect of machine learning.

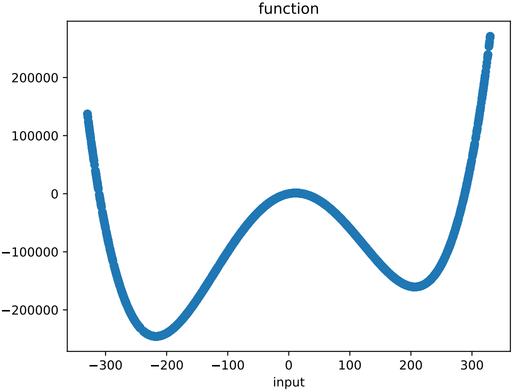


Figure – Loss function

Optimization in machine learning

The loss function typically represents the quality of a set of parameters to solve a problem.

- in supervised learning, typically a measure of the prediction error on the dataset
- in clustering, typically a distorsion
- in density estimation, a likelihood

Analytic minimization

Exercice 3 : What is the minimum of the function

$$f : x \to (x-1)^2 + 3.5 \tag{8}$$

And for what value x is it obtained ?

Iterative algorithms

However, in most applications of machine learning, it is not possible to use an analytical solution, either because :

- we do not know the analytical solution
- we know how to compute it, but the computation is too costly for practical use.

Instead, we use **iterative algorithms** (gradient descent, coordinate descent, etc.)

Gradient algorithms

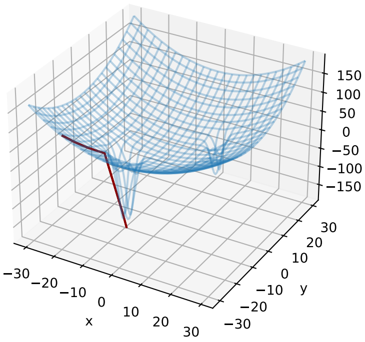


Figure – Optimization trajectory.