

# Lecture 5: Basic Machine Learning using Scikit-learn.

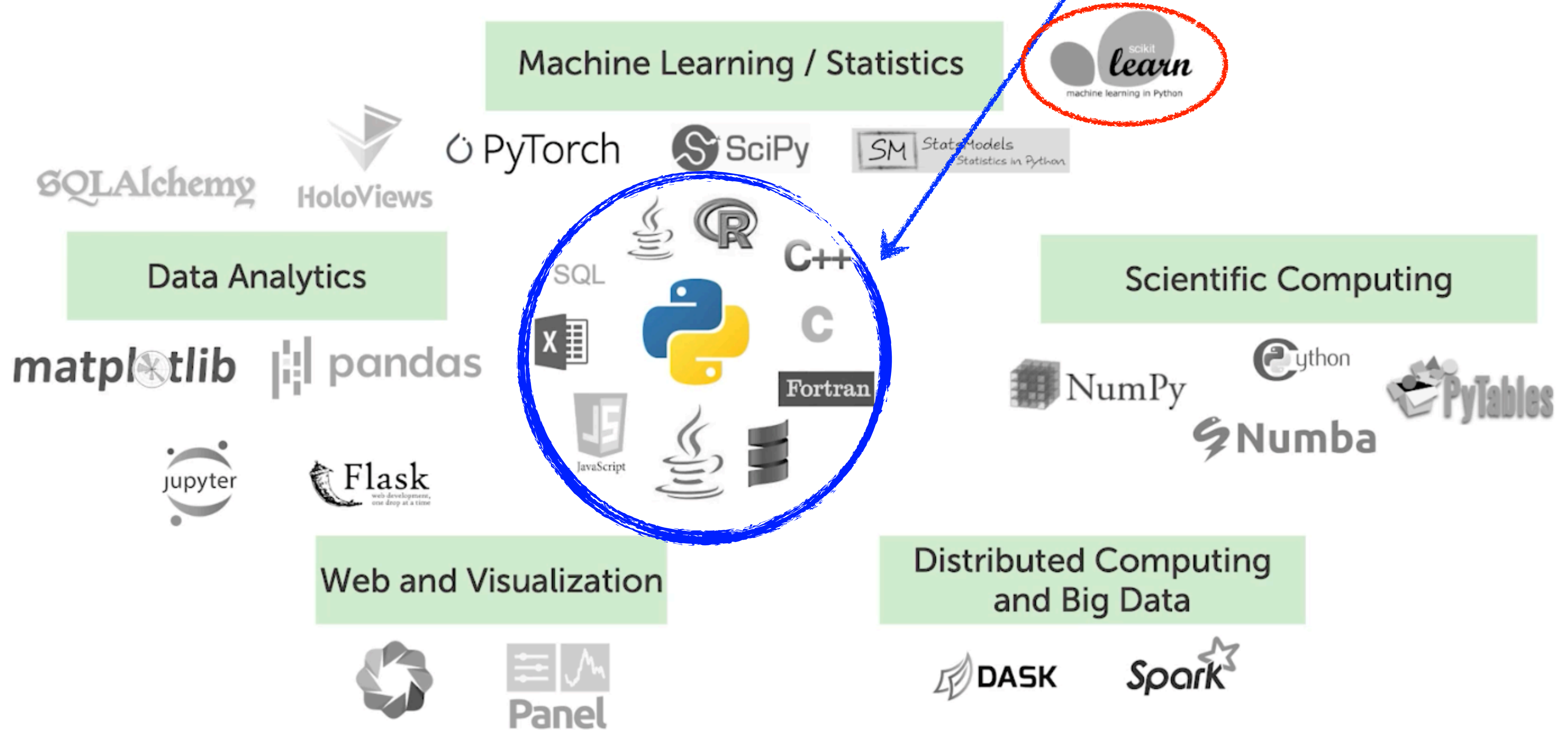
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Hanoi, 2021

# Lecture 5: Basic Machine Learning using Scikit-learn.

In this class I will briefly introduce Scikit-learn tool using for machine learning and data analysis.

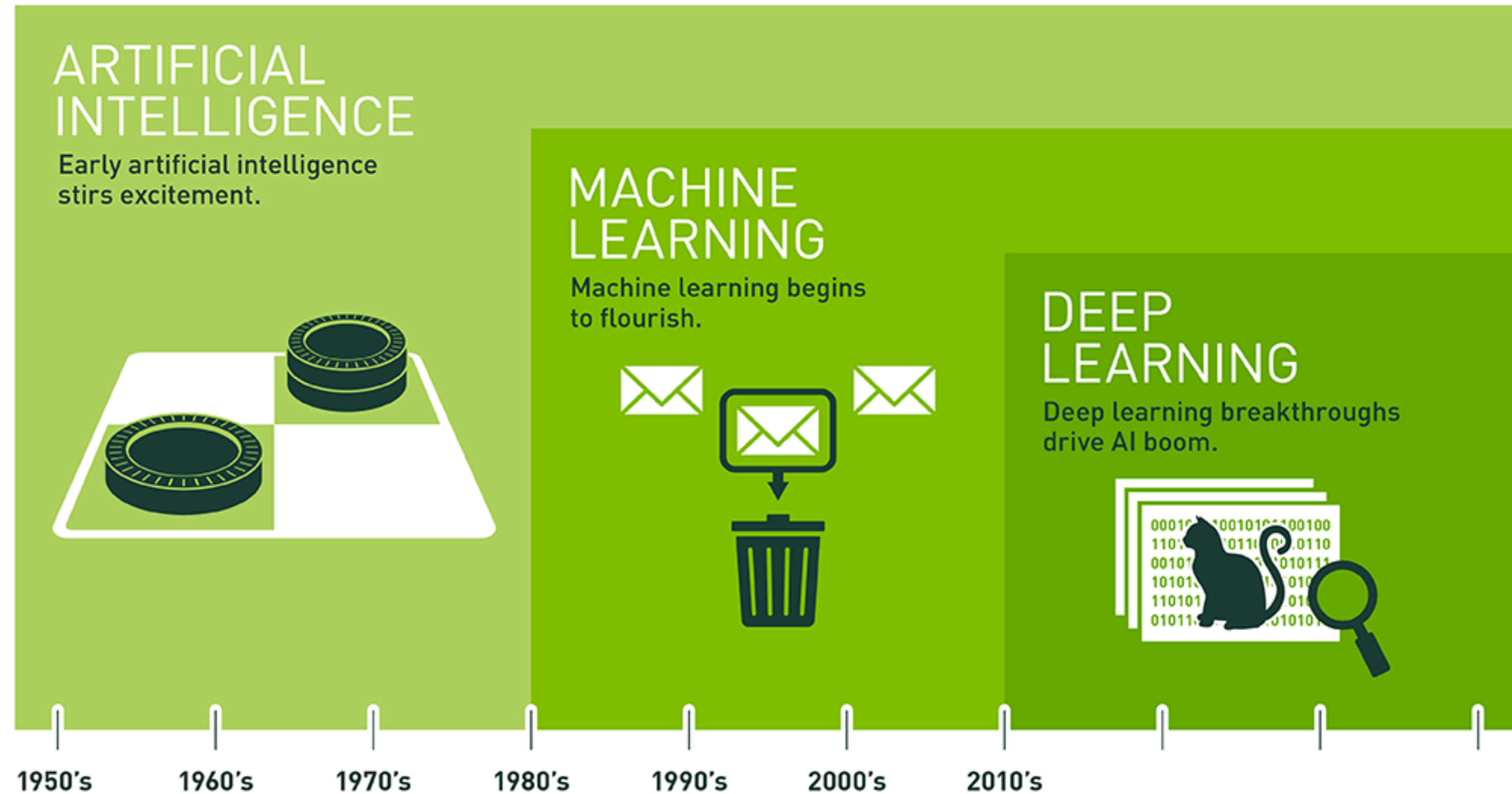
# Data Science & Anaconda





# AI, ML, DL


The relationship of AI, Machine Learning (ML), and Deep Learning (DL).



Since an early flush of optimism in the 1950s, smaller subsets of artificial intelligence – first machine learning, then deep learning, a subset of machine learning – have created ever larger disruptions.

Source: <https://blogs.nvidia.com/blog/2016/07/29/whats-difference-artificial-intelligence-machine-learning-deep-learning-ai/>

# Scikit-learn


[Install](#)
[User Guide](#)
[API](#)
[Examples](#)
[More ▾](#)

## scikit-learn

*Machine Learning in Python*

[Getting Started](#)
[Release Highlights for 0.24](#)
[GitHub](#)

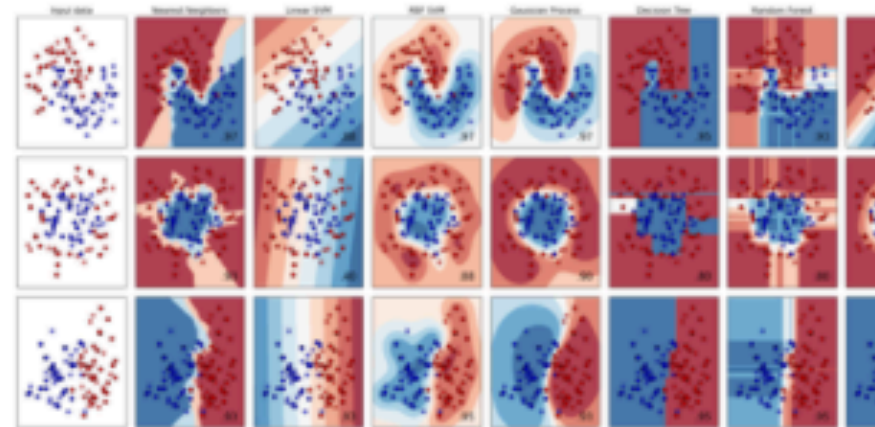
- Simple and efficient tools for predictive data analysis
- Accessible to everybody, and reusable in various contexts
- Built on NumPy, SciPy, and matplotlib
- Open source, commercially usable - BSD license

## Classification

Identifying which category an object belongs to.

**Applications:** Spam detection, image recognition.

**Algorithms:** [SVM](#), [nearest neighbors](#), [random forest](#), and [more...](#)



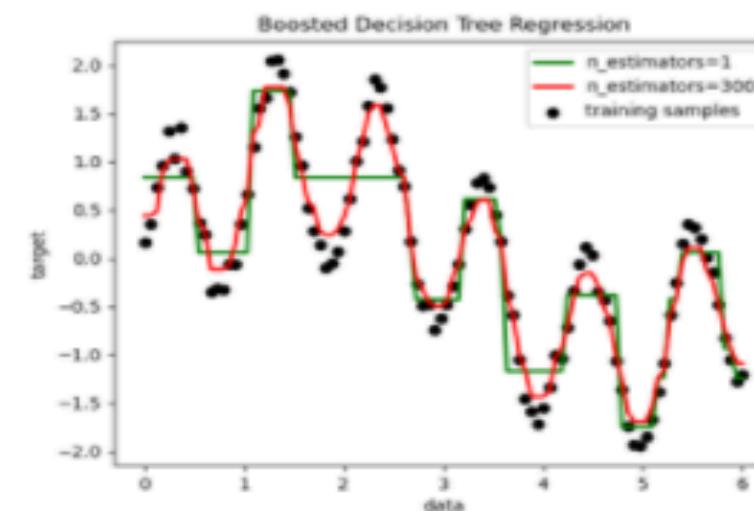
Examples

## Regression

Predicting a continuous-valued attribute associated with an object.

**Applications:** Drug response, Stock prices.

**Algorithms:** [SVR](#), [nearest neighbors](#), [random forest](#), and [more...](#)



Examples

## Clustering

Automatic grouping of similar objects into sets.

**Applications:** Customer segmentation, Grouping experiment outcomes

**Algorithms:** [k-Means](#), [spectral clustering](#), [mean-shift](#), and [more...](#)



Examples

## Dimensionality reduction

Reducing the number of random variables to consider.

**Applications:** Visualization, Increased efficiency

**Algorithms:** [k-Means](#), [feature selection](#), [non-negative matrix factorization](#), and [more...](#)

## Model selection

Comparing, validating and choosing parameters and models.

**Applications:** Improved accuracy via parameter tuning

**Algorithms:** [grid search](#), [cross validation](#), [metric-based selection](#), and [more...](#)

## Preprocessing

Feature extraction and normalization.

**Applications:** Transforming input data such as text for use with machine learning algorithms.

**Algorithms:** [preprocessing](#), [feature extraction](#), and [more...](#)

Source: <https://scikit-learn.org/stable/index.html>



# Scikit-learning for Machine Learning

- Scikit-learn is a tool for data analysis. In general, a learning problem considers a set of  $N$  samples data and then predict unknown quantities. Learning a problem can be classified into:

1. **Supervised learning:** Data have additional attributes, and the problem either:

- **Classification:** Identifying which category of an object belongs to. Example: Spam detection
- **Regression:** Predicting an attribute associated with an object. Example: Parameter estimations, minimum chi-square, stock price prediction.

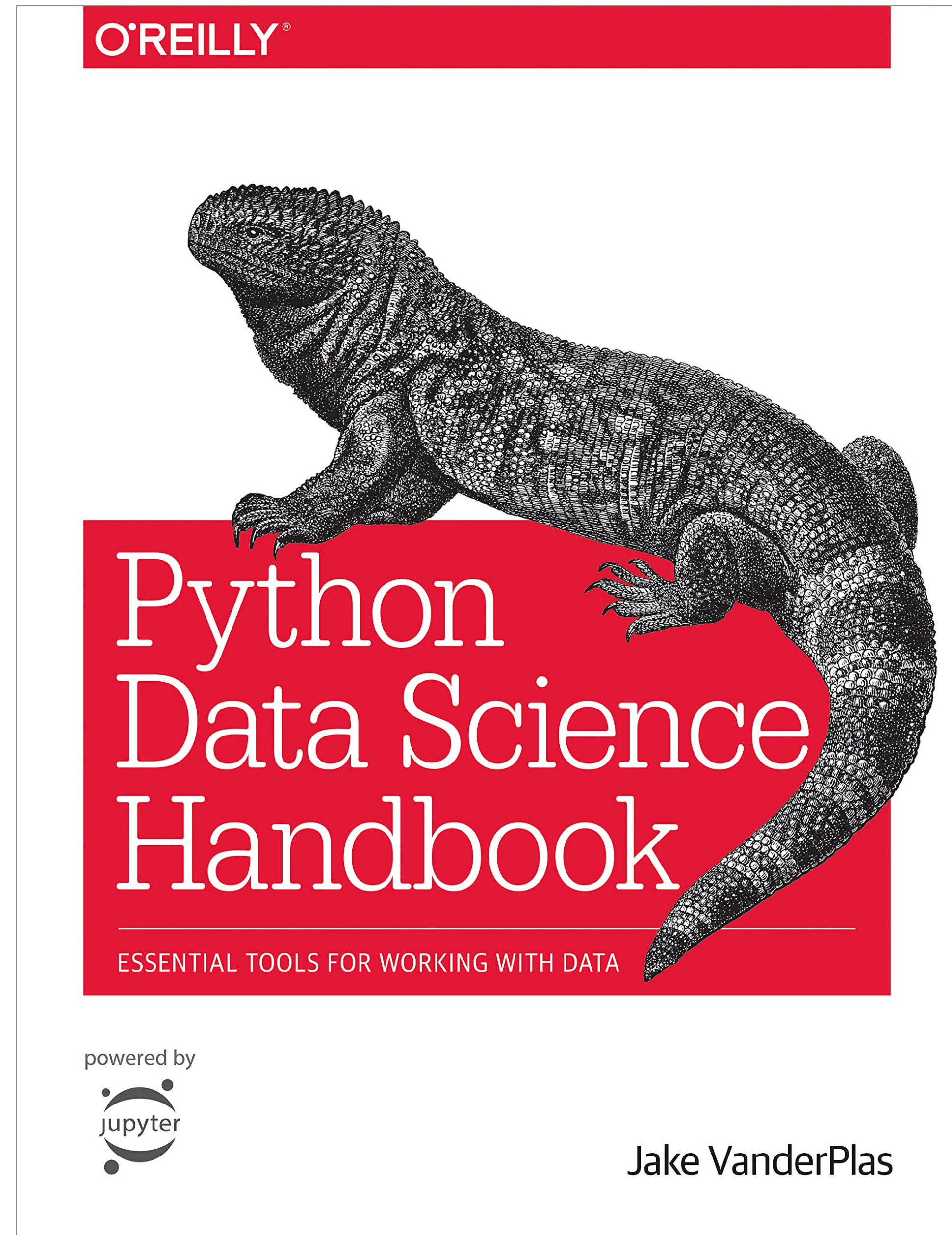
2. **Unsupervised learning:** Training data to discover groups of similar examples.

- **Clustering:** Automatic group of similar onbjects into sets, example: Customer segmentation. Methods: K-means.
- **Model selection:** Comparing, validating, choosing parameters and model, example PCA method.



# Scikit-learning book

GitHub: <https://github.com/jakevdp/PythonDataScienceHandbook>





# Scikit-learning example

The K-means algorithm divides a set of  $N$  samples of variable  $X$  into  $K$  number of clusters  $C$ , each cluster described by the mean  $\mu_j$  (centroids).

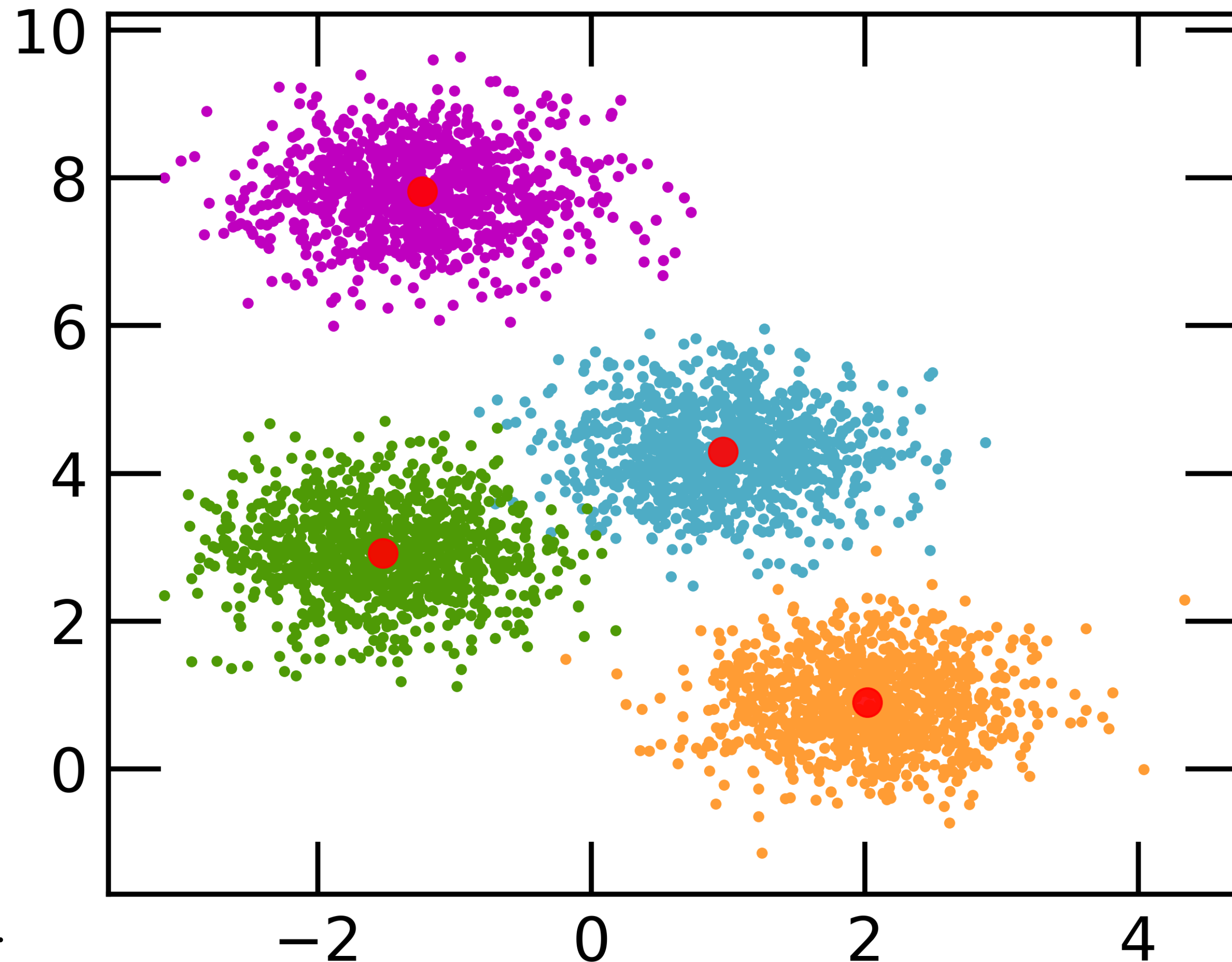
$$\sum_{i=0}^n \min_{\mu_j \in C} (||x_i - \mu_j||^2)$$

Notation:  $|| \cdot ||$  is norm of vector.

Approach:

1. Select random  $K$  as first centers.
2. Distribute data points to cluster that have closest center.
3. If the data distribution is stabled. The method is done.
4. Update centers for each cluster by estimate the mean again of all the data points in the current clusters after distribution step 2.
5. Redo step 2.

• Example of clustering K-Means





# Summary

- **Lecture 0:** Introduction to Jupyter notebook and Python.

In this class I review tools for data analysis and visualization: Python and Notebook.

- **Lecture 1:** Fundamental statistics quantities: Mean, median, standard deviation (variance), correlation.

In this class I review statistic quantities as mean, median, variance, and associated errors. The correlation between two random variables.

- **Lecture 2:** Probability distributions: Binomial, Uniform, Normal (or Gaussian), Poisson, Gamma, T-Student's, Chi-square, the central limit theorem.

In this class I review several special probability distributions and simple applications.

# Summary

- **Lecture 3:** Hypothesis testing, model fitting and parameter estimation.

In this class we studied the use of the  $\chi^2$  statistic as a (1) hypothesis test of a model and (2) estimate the best fit parameter estimation.

- **Lecture 4:** Principal components analysis (PCA) and Bayesian methods.

In this class We focus on PCA dimensionality reduction to model variables and Bayesian statistic applications.

- **Lecture 5:** Basic Machine Learning using Scikit-learn tool.

In this class I briefly introduce Scikit-learn tool using for machine learning and data analysis.