

# Lectures on machine learning |

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# Several words before start

- Xiaohu gave a set of lectures about statistics several years ago ([Link](#)) before he left. Now it is my turn to present ML!
- Machine learning is not a one-day lesson.
- This tutorial could be a weekly tutorial before I leave for a new position.
- This set of tutorials will include algorithms, code of implementation and applications in various areas.

# Outline

- 1 What is machine learning
- 2 Typical method of machine learning
  - Top 10 algorithms in data mining and TMVA
  - Deep learning
- 3 Examples
  - kNN
  - Decision Tree
  - Neural Network
- 4 Tools and packages
  - Python package and platform
- 5 Applications
- 6 Remarks

# What is machine learning([twiki](#))

Machine learning (ML) is the scientific study of algorithms and statistical models that computer systems use to perform a specific task without using explicit instructions, relying on patterns and inference instead.

The type of machine learning:

- Supervised learning
- Un-supervised learning
- Re-inforcement learning

# How ML is used in HEP

- Classification:
  - Particle identification
  - Flavor tagging
  - Event classification
- Regression:
  - Energy calibration
  - Reconstruction : pattern recognition

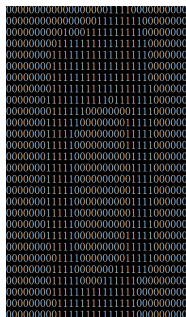
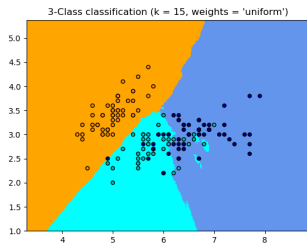
# Typical method of machine learning

- Top 10 algorithms in data mining([Link](#))
- C4.5
- K-means
- Support Vector Machine(SVM)
- Apriori
- EM
- What is in TMVA?
  - Cut, Likelihood, kNN, Linear Discriminant
  - FDA(Function Discriminant Analysis)
  - Neural Network, SVM, BDT
- Deep learning : RNN, CNN etc
- PageRank
- AdaBoost
- kNN
- Naive Bayes
- Classification and Regression Tree(CART)

# k-nearest neighboring

## Algorithm

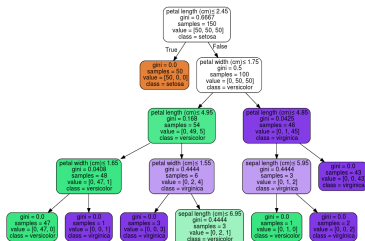
- Given the labeled dataset.
- Calculate the distance between the test point and the labeled event.
- Select the k-nearest event.
- Calculate the frequent label and it is assigned to the test point.
- Parameters : **K**



# Decision Tree

## Algorithm

- Give a dataset with  $\vec{X}$
- From each node, separate the sample to two parts by cutting on the variables, to maximize the significance(entropy, gini-index)
- ending node :
  - Number(Fraction) of events is smaller enough
  - The improvement of significance is smaller enough
  - Maximum number of nodes or depth
- Check the over-training
- Parameters : type of figure of merit, criteria of ending

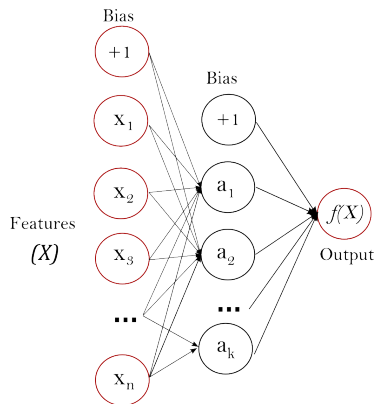




# Neural Network—Multi-Layer Perceptron

## Algorithm

- Give a dataset with character  $\vec{X}$  (n-dimensional vector)
- Hidden layer :  $\vec{a} = f(\mathbf{W}_1\vec{X})$ ,  $f(x)$  is propagation function
- Output :  $y = \mathbf{W}_2\vec{a}$
- Parameters : **Weight Matrix, number of hidden layers,  $f(x)$**



# Tools and packages

- Python packages
  - Scikit-learn([Link](#)), Keras([Link](#)), Theano
- Platform
  - Tensorflow(Google)([Link](#))
  - PyTorch(Facebook)([Link](#))
- Example codes from Wisconsin
  - DNN : <https://github.com/laserkaplan/ttHyyML>
  - XGBoost : [https://gitlab.cern.ch/wisc\\_atlas/ttHyyML](https://gitlab.cern.ch/wisc_atlas/ttHyyML)

# Applications

- Speech recognition
- Handwriting recognition
- Natural Language Processing(NLP)
- Computer vision : face recognition, medical diagnosis
- Online advertising : I also know HEP PhD is doing electronic business

# Remarks

- How to learn machine learning
  - Do you want to be a “man of parameter-tuning”



- Want to go to industry?
  - I talked to a HEP PhD from USTC working in HuaWei and he already recruited two of my undergraduate classmates who are also HEP PhD.
  - You need to be familiar with:
    - the detail of traditional ML algorithms
    - related python packages
    - deep learning
    - one mature platform Tensorflow or PyTorch

# To be continued

- Let's start with BDT next week.
- It will includes:
  - BDTA, BDTG, BDTB, BDTD, BDTF, XGBoost BDT
  - algorithms, implementation and comparison
- See you next time!