

# DDA3020: Machine Learning - Tutorial 1

- Python, Scikit-Learn Basis & Gradient Descent

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- ① Tutorial Information
- ② Python Basis
- ③ Numpy & Scikit-Learn
- ④ Gradient Decent

## Course Information

### Tutorial:

**Venue:** TB102 (126 seats), Tuesday 18:00 - 21:00pm from Week 2 -Week 13.  
S1:18:00-18:50pm(for T02,05), S2: 19:00-19:50pm(for T03,04), S3: 20:00-20:50pm(for T01,06).

All the 3 sessions will give the same tutorial content, and you can select a convenient time to attend. The Zoom link will be given on BB before each tutorial and the tutorial recordings will also be provided.

### TAs:

- **Xudong Wang:** xudongwang@link.cuhk.edu.cn  
**OH:** Tue 4:30-5:30pm, Seat 70, SDS Lab, 4F ZX
- **Dan Qiao:** danqiao@link.cuhk.edu.cn  
**OH:** Fri 4:00-5:00pm, Seat 1, SDS Lab, 4F, ZX
- **Fei Yu:** feiyu1@link.cuhk.edu.cn  
**OH:** Thu 3:30-4:30pm, Seat 14, SDS Lab, 4F, ZX
- **Sho Inoue:** shoinoue@link.cuhk.edu.cn  
**OH:** Tue 5:30-6:30pm, ZOOM: 759 394 8941 (pw123456)
- **UStaff TAs:** TBA

**Add/Drop: September 5th (00:00 Monday) to September 16th (23:59 Friday)**

# Python Checklist

- **Install:** Python Version  $\geq 3.7$ , Choose a IDE as you like: VSCode, PyCharm, Spyder, **Jupyter Notebook/Lab** (Web-based, interactive)...
- **Basis Knowledge:**
  - Values Types: int, float, str, byte ...
  - Values Assignment and Operations
  - Input/output Operations
  - Control Flow in Python Conditional statements : if ... else ...  
Loop statements: for ... / while ...
  - Function call: def function:
  - Data Structure(list, tuple, dict...)
  - Python Classes and Objects: OOD
  - Others: import, comments

Recap the CSC1001 materials or Ref the Python3 Cheat Sheet.

## Python Packages for ML

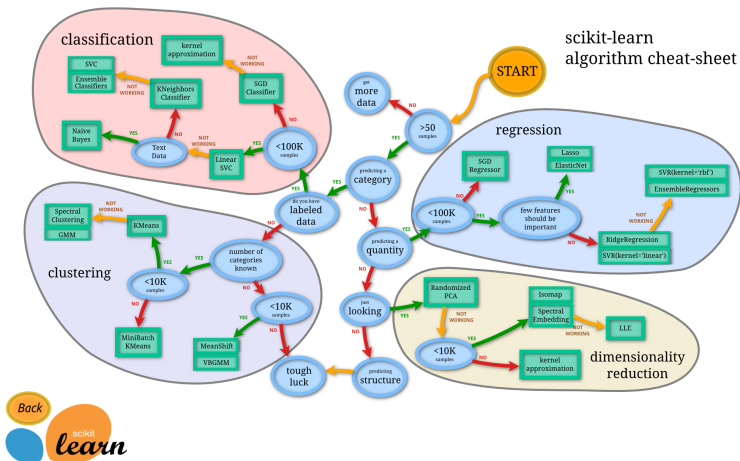
- Numpy + Scipy:
  - i) Useful data structure: ndarray, matrix, array
  - ii) Algorithms: statistics, scientific computation
- Pandas:  
Processing table-like dataset (such as Excel), DataFrame, Series
- Scikit-learn: Various data mining/machine learning functions
- NLTK: Natural Language Toolkit
- Visualization: Matplotlib, Seaborn...
- Deep Learning: TensorFlow (Google), PyTorch (Facebook), Caffe, Keras, MXNeT (Amazon)

Recap the lecture content, we can express the data in vector/matrix formula: one data sample with  $d$  features  $\mathbf{x}_i \in R^d$ , total  $n$  sample we can stack a matrix  $X = [\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n] \in R^{d \times n}$

In this tutorial, we will briefly introduce the **Numpy: ndarray** and **Scikit-learn**.

# Scikit-Learn: Various data mining/machine learning functions

Strongly recommend to ref the Scikit-Learn official website:  
[https://scikit-learn.org/stable/user\\_guide.html](https://scikit-learn.org/stable/user_guide.html)



# Numpy & Scikit-Learn

Move to Jupyter Notebook

# Gradient

In vector calculus, the gradient of a **scalar-valued differentiable function**  $f$  of several variables is the vector field (or vector-valued function)  $\nabla f$  whose value at a point  $\mathbf{x}$  is the vector whose components are the partial derivatives of  $f$  at  $\mathbf{x}$ . That is, for:

$$f : \mathbb{R}^d \rightarrow \mathbb{R}$$

its gradient:

$$\nabla f : \mathbb{R}^d \rightarrow \mathbb{R}^d$$

is defined at the point:

$$\mathbf{x} = (x_1, x_2, \dots, x_d)^T$$

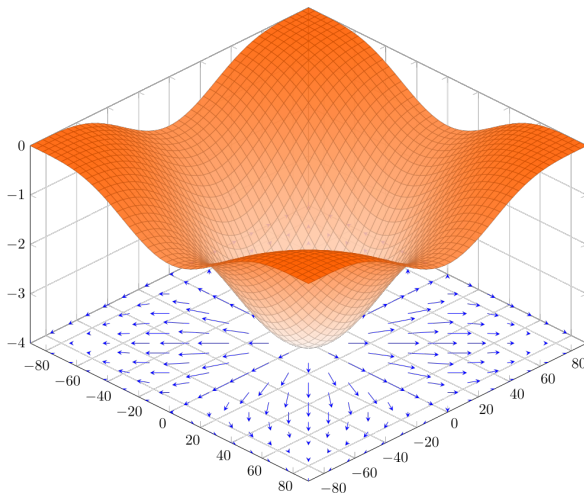
in  $n$ -dimensional space as the vector

$$\nabla f(\mathbf{x}) = \begin{bmatrix} \frac{\partial f}{\partial x_1}(\mathbf{x}) \\ \vdots \\ \frac{\partial f}{\partial x_d}(\mathbf{x}) \end{bmatrix}$$

We can find that the  $\nabla f(\mathbf{x})$  have the same shape of  $\mathbf{x}$ . Actually, it gives a direction.



# Gradient Direction: "Steepest" direction to increase function value



**Figure 1:** The gradient of the function  $f(x, y) = -(\cos^2 x + \cos^2 y)2$  depicted as a projected vector field on the bottom plane.

# Gradient Descent Algorithm

GD is a "fundamental" algorithm in **continuous optimization**, both historically and mathematically. Consider the unconstrained problem:

$$\min_{\theta} f(\theta)$$

where  $f: R^d \rightarrow R$  A basic form of the gradient descent algorithm (constant-stepsize) is:

$$\theta_{k+1} = \theta_k - \eta \nabla f(\theta_k), k = 0, 1, 2, \dots$$

where  $\eta$  is a constant called "stepsize" (a.k.a. learning rate in machine learning community), and  $\nabla f(\cdot)$  denotes the gradient of  $f$ .

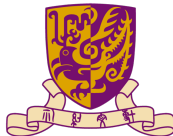
It means that at each iteration, the iterate will move along the negative gradient direction and how far it moves is controlled by the stepsize  $\eta$ .

GD is an **iterative** method. Using **heuristic** metrics: function value(loss), gradient... to determine the stop criteria.

# Summary

- Python Basis
- Numpy ndarray
- Scikit-learn Package Usage
- Gradient Descent (Solve, optimize model parameters)

# Thank you for listening!



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