# 01\_slides\_jupyter

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## 1 AICE1006 - Data Analytics

#### 1.1 Lecture 1 - Introduction

**Zhiwu Huang** Lecturer (Assistant Professor) Vision, Learning and Control (VLC) Research Group School of Electronics and Computer Science (ECS) University of Southampton

Office Hour: Wed 2PM-3PM, Please book in advance. Zhiwu.Huang@soton.ac.uk

Credit: Marco Forgione, Researcher, SUPSI-USI

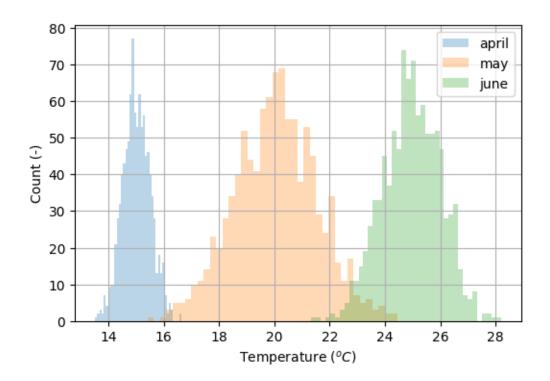
#### 2 Overview

Objective: \* To improve your programming skills with focus on data science applications \* To get acquainted with the most common data manipulation techniques \* To develop an intuitive understanding of real-world datasets

Main Topics: \* Data ingestion (how to load data) \* Data wrangling (concatenate, join, pivot,...) \* Data visualization (scatter plots, histograms, box plots,...) \* Basic data analysis (mean, standard deviation, correlation,...) \* Advanced data analysis (pricipal component analysis, manifold learning, clustering,...) \* More advanced (advanced machine/deep learning,...): NOT INCLUDED IN AICE1006

You may see similar concepts in the other modules like AICE1008: Mathematics. The focus of the Data Analytics module is on applications rather than theory!

#### 2.1 Data Visualization Examples

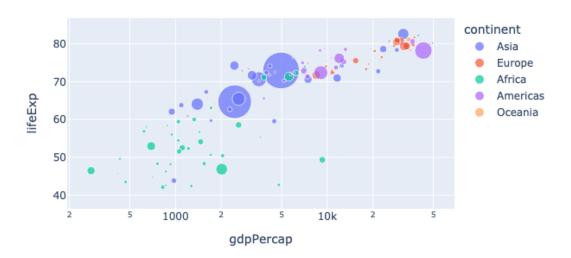


### 2.2 Data Visualization Examples

The signal y(t) follows a **sinuisoidal pattern** over time.

### 2.3 Data Visualization Examples

## GDP, life expectancy, continent, and population of countries in 2007



### 3 Software tools

- Python (programming language)
- Jupyter Notebook (web-based programming interface)
- Python libraries:
  - Numpy (numerical mathematics)
  - **Pandas** (tabular data handling)
  - Matplotlib, Pandas, and Seaborn (data visualization)

**–** ...

#### 3.1 Resources

Required: \* The course's slides \* Official documentation of python, pandas, matplotlib, ... \* Google, stack overflow, ...

Optional: \* The book "Python Data Science" by Jake Vanderplas (available on-line for free) \* The book "Python for Data Analysis" by Wes McKinney (link on the publisher's website)

# 4 Getting started

- Install all the required packages. Type in a terminal:
- \$ pip install numpy scipy pandas scikit-learn matplotlib jupyter
- \$ pip install rise
- \$ pip install plotly

• Start Jupyter notebook. Type in a terminal:

#### \$ jupyter notebook

A page should appear in your default browser. Otherwise, copy-paste the link from your terminal in the browser.

On Windows, Chrome seems to work better.

## 5 Introduction to Jupyter notebook

A web-based Python environment very good for: \* Rapid prototyping \* Small/medium data science projects \* Presenting results (these slides made with Jupyter!)

Jupyter allows mixing formatted text and python code. The cell below contains Python code!

```
[5]: print("Welcome to the data challenge module!")
#a = 1
#b = a + 1
#print(b)
```

Welcome to the data challenge module!

Don't use Jupyter for a large software development project! Go for an IDE like PyCharm or VSCode instead.

#### 5.1 Cell

The cell is the basic input unit in a Jupyter notebook. Common operations:

- Add a cell
  - Select an existing cell by clicking in the white area at its left
  - Click on Insert -> Insert Cell Above/Below (or click a/b)
- Remove a cell
  - Select a cell
  - Click on Edit -> Delete Cells (or click x)
- Move a cell
  - Select the cell (hold shift to select multiple cells)
  - Use the arrows to move the cell up and down

#### Cell type

- A cell may contain either Python code or Markdown (= formatted text).
- This slide is made up of Markdown cells!
- To change the cell type, select Python or Markdown in the drop-down menu

## 6 Python for data science

Python is an *interpreted*, dynamically typed programming language. \* No need to compile the code. A virtual machine interprets it on the fly. \* No need to declare variable types. They are inferred at run-time (dynamic typing).

```
[6]: # Sum numbers from 0 to 9
    result = 0
    for i in range(10):
        result += i
    print(result) # result = 0 + 1 + 2 + .. + 9 = 45
```

[7]: result = "foo" # now the type of the variable result changes dynamically to str

Conversely, C is a compiled, statically typed language.

```
// file main.c
// C language: sum numbers from 0 to 9
int result=0; // need to declare result as an integer variable
for(int i=0; i<10; i++){
    result += i;
}
printf("%d", result);
//result = 'foo' // this would result in a compile error!</pre>
```

In C, we need to first compile main.c, and then execute the generated program.

## 7 Python for data science

- Python is user-friendly, but it can be slower than a compiled language when used without care
- To get full speed, specialized libraries for the most intensive tasks must be used

In this course, we use Python for data science in conjunction with packages such as: \* numpy for number crunching \* pandas for handling tabular data \* matplotlib and plotly for data visualization \* scikit-learn for machine learning \* ...

Using these packages, we can tackle different data science tasks in Python without compromising performance.

But first, let us review some Python basics!

## 8 Data types

Python is a dynamically typed language. No need to define variable type. It is inferred at run-time.

```
[8]: counter = 1 # an integer
val = 3.14 # a float (floating point number)
message = "Hello, world" # a string. Also message = 'Hello, world' works
flag = False # a Boolean variable. Possible values: {True, False}
#val_c = 0.5 + 0.4j # a complex number (real + imaginary part)

type(message)
```

[8]: str

The variable type can change at run-time:

```
[9]: message = 42 # message is now an int!
type(message)
```

[9]: int

The variable message was first a string, then it was changed to int. This is OK for Python.

## 9 Basic operations

Basic arithmetic operations +, -, /, \* behave as expected on numeric types:

```
[10]: val1 = 3.14
val2 = 42
val3 = val1 + val2
val3
```

[10]: 45.14

```
[11]: type(val1), type(val2), type(val3)
```

[11]: (float, int, float)

```
[12]: val2/val1
```

[12]: 13.375796178343949

Basic operations + and \* are also defined on strings:

```
[13]: message1 = "Hello, "
  message2 = "World"
  message = message1 + message2 # string "sum"
  message
```

[13]: 'Hello, World'

```
[14]: message1 * 3 # string "multiplication"
```

[14]: 'Hello, Hello, Hello, '

### 10 Data structures

#### 10.1 Lists

A list is an ordered collection of python objects:

```
[15]: object_lst = [3.14, "pear", 42] # a float, a string, an integer object_lst
```

[15]: [3.14, 'pear', 42]

You can access its elements with square bracket notation and integer indexes 0,1,2...,len(object\_lst)-1

```
[16]: object_lst[2] # note: starting index is 0!
```

[16]: 42

You can modify the elements of a list:

```
[17]: object_lst[2] = 44 object_lst
```

[17]: [3.14, 'pear', 44]

You can append to a list using the append method:

```
[18]: object_lst.append(-5) object_lst
```

[18]: [3.14, 'pear', 44, -5]

### 11 Data structures

#### 11.1 Lists

You can access more elements at once and get a sub-list:

```
[19]: object_lst
```

[19]: [3.14, 'pear', 44, -5]

```
[20]: object_lst[1:3] # returns the elements from index 1 to 3-1
```

[20]: ['pear', 44]

The "sum" of two lists is the concatenation of its elements

```
[21]: list1 = ["A", "B", "C"]
list2 = ["D", "E", "F"]
list1 + list2
```

[21]: ['A', 'B', 'C', 'D', 'E', 'F']

The "multiplication" of a list times an integer is also defined:

```
[22]: ['A', 'B', 'C', 'A', 'B', 'C', 'A', 'B', 'C']
```

### 12 Data structures

#### 12.1 Tuples

A tuple is pretty much like a list, but it is immutable:

```
[23]: object_tup = ("A", "B", False)
object_tup

[23]: ('A', 'B', False)

[24]: object_tup[0]

[24]: 'A'

[25]: # Don't do this!
#object_tup[0] = True
```

### 13 Data structures

#### 13.1 Dictionaries

A dictionary is a collection of key-value pairs:

[26]: {'apple': 1.0, 'pear': 1.5, 'strawberry': 2.0}

"apple", "pear", "strawberry" are the keys, 1.0, 1.5, 2.0 are the values

You can access its elements using square bracket notation:

```
[27]: fruit_cost_dict["strawberry"] # dictionary[key] returns the corresponding value
```

[27]: 2.0

You can modify the value associated to a key:

```
[28]: fruit_cost_dict["strawberry"] = 2.5
```

You can even add a new key-value pair:

```
[29]: fruit_cost_dict["mango"] = 5.0
```

### 14 Control Flow

#### 14.1 If statement

Basic usage of the if statement: check a condition. If True, execute the code in the following block

```
[30]: x = 1
if x > 0:
    print("x is greater than 0")
```

x is greater than 0

The optional **else** block, it is executed when the condition is False:

```
[31]: x = 0
if x > 0:
    print("x is greater than 0")
else:
    print("x is smaller than (or equal to) 0")
```

x is smaller than (or equal to) 0

multiple elif blocks may be used to handle multiple cases:

x is equal 0

### 15 Control Flow

## 15.1 While loop

A while loop executes a block of code until a certain condition becomes False or a break statement is encountered:

```
[33]: result = 0
  idx = 0
  while idx < 5:
     result = result + idx # or result += idx
     idx = idx + 1
  result # = 0 + 1 + 2 + 3 + 4 = 10</pre>
```

[33]: 10

```
[34]: result = 0
    idx = 0
    while idx < 5:

        result = result + idx
        idx = idx + 1

        if result > 4:
            break

result # = 0 + 1 + 2 + 3 + 4 = 10
```

[34]: 6

## 16 Control Flow

### 16.1 For loop

A for loop iterates over a sequence (a collection like a list or a tuple, or an iterator).

```
[35]: name_lst = ["Marco", "Alice", "Dario", "Anna"]

for name in name_lst:
    print(name)
```

Marco

Alice

Dario

Anna

The **for** loop is generally more readable than the **while** equivalent:

```
[36]: idx = 0
while(idx < len(name_lst)):
    print(name_lst[idx])
    idx = idx + 1</pre>
```

Marco

Alice

Dario

Anna

## 16.2 The range iterator

It is very common to iterate over a numerical range. The range iterator comes in handy:

```
[37]: # Print the square of the numbers from 0 to 4
for idx in range(5): # no need to declare a list [0, 1, 2, 3, 4]
print(idx, idx**2) # print idx, idx squared
```

0 0

1 1

2 4

3 9

4 16

The code above is way more practical than explicitly constructing the sequence of values 0, 1, 2, 3, 4

In Python, range (5) is an **iterator** that behaves like the list [0, 1, 2, 3, 4]. It is more convenient to use the range iterator instead of explicitly constructing the equivalent list.

```
[38]: list(range(5)) # now the range is explicitly converted to an equivalent list
```

[38]: [0, 1, 2, 3, 4]

We can also specify a starting index:

```
[39]: list(range(1, 11)) # values from 1 to 10
```

[39]: [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]

## 17 Control Flow

#### 17.1 List comprehension

A very common task: filter the elements of a sequence according to a logic condition.

Example: take all the strings starting with A from a list

```
[40]: name_lst = ["Marco", "Alice", "Dario", "Anna"]

[41]: names_A = [] # result list
    for name in name_lst:
        if name.startswith("A"):
            names_A.append(name)
        names_A
```

[41]: ['Alice', 'Anna']

**List comprehension** offers a shorthand alternative:

```
[42]: names_A = [name for name in name_lst if name.startswith("A")]
names_A
```

[42]: ['Alice', 'Anna']

### 18 Control Flow

### 18.1 List comprehension cont'd

Sometimes we want to filter and transform the elements of the sequence.

Example: take the **square** of all the **odd numbers** from 1 to 10

```
[43]: squared_odds = [] # result_list
for num in range(1, 11):
    if num %2 == 1: # if the remainder of the division by 2 is zero...
        squared_odds.append(num**2)
squared_odds
```

[43]: [1, 9, 25, 49, 81]

This can also be done with list comprehension:

```
[44]: squared_odds = [el**2 for el in range(1, 11) if el%2 == 1] squared_odds
```

[44]: [1, 9, 25, 49, 81]

The general notation of **List comprehension** is thus: [expr for el in iterable if condition].

#### 19 Modules

### 19.1 The math module

In python, a **module** is just a .py python file (or a folder with more .py files). To use a module, you must first import it!

Example: the math module, with is part of the Python Standard Library.

```
[45]: import math # import the math module
```

```
[46]: math.pi, math.sin(math.pi/2), math.cos(math.pi/2), math.exp(0), math.log(1)
```

[46]: (3.141592653589793, 1.0, 6.123233995736766e-17, 1.0, 0.0)

Sometimes we want to import specific objects from a module:

```
[47]: from math import pi, sin, cos, exp, log
```

```
[48]: sin(pi/2), cos(pi/2), exp(0), log(1)
```

[48]: (1.0, 6.123233995736766e-17, 1.0, 0.0)

### 20 Modules

#### 20.1 The os module

Another useful module is os. It exposes functionalities from the operative system.

```
[49]: import os
[50]: os.listdir() # list of files in a folder (with no arguments, the working folder)
[50]: ['01_exercises.ipynb',
       '01_slides_jupyter.ipynb',
       '.ipynb checkpoints',
       'example_notebook.ipynb',
       '01 exercises solution.ipynb',
       'img']
     Example: find all the jupyter notebooks (.ipynbs) in a folder:
[51]: # find all the jupyter notebooks (.ipynb) in the current folder
      ipvnb lst = []
      for file in os.listdir():
          if file.endswith(".ipynb"):
              ipynb_lst.append(file)
      ipynb_lst
[51]: ['01_exercises.ipynb',
       '01_slides_jupyter.ipynb',
       'example_notebook.ipynb',
       '01_exercises_solution.ipynb']
[52]: | ipynb_lst = [file for file in os.listdir() if file.endswith(".ipynb")] # list_
       ⇔of all .ipynb files
```

# 21 Today's challenge

- Set up jupyter with all the required packages (Slide 9, getting started)
  - pip install all the required packages
  - Start jupyter
- Familiarize with Jupyter notebook:
  - Access jupyter from your favorite web browser
  - Create a new notebook
  - Create/remove/reorder/execute cells
  - Change cells type from code to markdown
  - Save notebook with name
- Familiarize with Python types and basic control flow
  - Strings, integers, floats
  - List, dictionaries
  - For loop, if statement