LECTURE 02

Computing Fundamentals

September 18, 2023

PBHLTH 198, Fall 2023 @ UC Berkeley

Andrew O'Connor



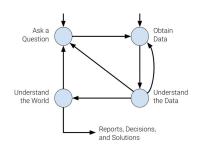
Class Outline

- Recap of lecture 1
- Motivation
- Python fundamentals + terminology
- Command Line
- Dependency Management
- Setup

Recap

- Public Health is about disease prevention in order to prevent disease, prolong health, promote health
 - Many ways to do this; Can do it through health policy, community engagement, getting involved with politics, or using statistics and epidemiological theory to help public officials make informed decisions to ultimately eliminate disease (this class!)
- Epidemiology seeks to understand disease within populations and design studies to explore the how, why, and where of disease occurrence
- Biostatistics aims to analyze health data using advanced statistical methods to derive valid conclusions
- Data analysis frameworks provide a <u>structured</u> and <u>efficient</u> approach to data analysis, ensuring consistency and quality in results while facilitating collaboration and compliance with regulations regarding privacy
 - Examples: PPDAC, Data Science Lifecycle (we will focus on this more)





Motivation

Main goals of this class

- Understand how to install Python locally for the purpose of data analysis in a Jupyter Notebook
- Understand the concept of dependency management and

Motivation - Workflow

- 1. Open up the command line
- 2. Activate a virtual environment

```
conda activate <name of virtual environment>
```

3. Install necessary libraries using the command line

```
conda install -c anaconda numpy

conda install -c conda-forge matplotlib

conda install -c anaconda pandas

conda install -c anaconda seaborn
```

4. Open JupyterLab

```
jupyter-lab
```

- 5. Open an existing Jupyter Notebook
- 6. Run cells, do data analysis, etc.
- 7. Document the changes made to the project (ignore for now)

Terminology

- **Software:** Applications or programs on your computer (e.g. Excel, Windows 10, Zoom)
- **Hardware:** Any physical part of a computer (e.g. CPU, GPU, RAM, hard drive)
- Machine: Computer or device capable of doing various tasks, computation, data processing;
 Can be physical or virtual
- Local/Locally: Referring to your own computer
- Environment: Refers to a specific type of setup or configuration (includes software, hardware, libraries, packages, etc.)
- **Operating System**: core software that manages hardware resources and provides a platform for running applications; controls tasks like file management, memory allocation, and user interface, allowing users and software to interact with the computer's hardware
 - examples: MacOS, Linux, Windows

Python



What do programs do?

- Programs work by manipulating values
- **Expressions** in programs evaluate to values
 - o Expression: 'a' + 'hoy'
 - o Value: 'ahoy'
- Every value has a certain data type

Data type	Example values			
Integers	2 44 -3			
Floats	3.14 4.5 -2.0			
Booleans	True False			
Strings	'¡hola!' 'its python time!'			

Expressions (Two Ways)

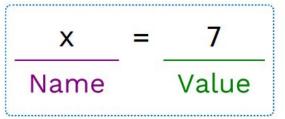
- Expressions can either use **operators** or **function calls** to obtain a value
- **Operators**: special symbols or keywords that are used to perform various operations on variables and values
 - Examples: + (addition), / (division), * (multiplication), ** (exponentiation), = (assignment)
- **Function Calls**: Pieces of predefined code we can reuse to return a value

How Python evaluates a call expression:

- 1. Evaluate the operator
- 2. Evaluate the operands
- 3. Apply the operator (a function) to the evaluated operands (arguments)

Names & Variables

- Names are bound to a value
 - Values can be expressions
- Names that are bound to values are also called variables



Functions

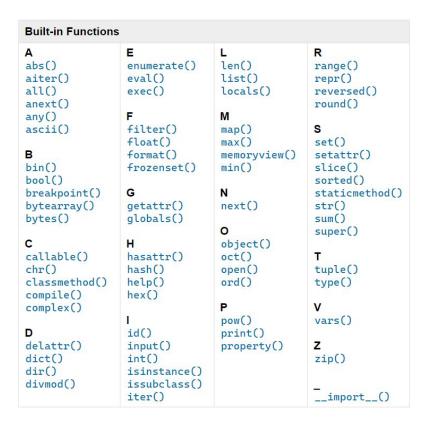
- **Functions** are sequences of code that perform a specific task that can easily be reused
 - The first line of a function is called the function signature
 - All lines after are considered the function body
 - Functions take in 0 or more **parameters**
 - Some functions have helpful comments that describe what the function does these are called **docstrings** and they appear on the second line

```
def <name>(<parameters>):  # ← Function signature
  return <return expression> # ← Function body

def add(num1, num2):  # ← Function signature
  return num1 + num2 # ← Function body
```

Built-in Functions

 Python has a list of **built-in functions** that can be called at any time without the need of installing/importing libraries



Conditionals

- Conditionals are statements that allow a program to make decisions and take different actions based on whether a specified condition evaluates to True or False
 - True and False values are called booleans

```
if num < 0:
    sign = "negative"
elif num > 0:
    sign = "positive"
else:
    sign = "neutral"
```

Syntax tips:

- Always start with if clause.
- Zero or more elif clauses.
- Zero or one else clause, always at the end.

```
if temperature < 0:
    clothing = "snowsuit"
elif temperature < 32:
    clothing = "jacket"
else:
    clothing = "shirt"</pre>
```

Iterables

- In Python, collections are containers or data structures that can hold multiple values or objects
 - They are used to group related data and provide convenient methods for manipulation and iteration
- Collections are iterables because are objects that can be looped/iterated over
- List[], Tuple(), Set (), Dictionary {}

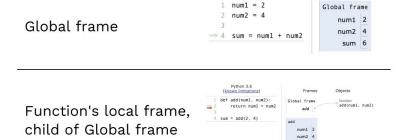
Data Structure	Ordered	Mutable	Constructor	Example
List	Yes	Yes	[] or list()	[5.7, 4, 'yes', 5.7]
Tuple	Yes	No	() or tuple()	(5.7, 4, 'yes', 5.7)
Set	No	Yes	{}* or set()	{5.7, 4, 'yes'}
Dictionary	No	Yes**	{ } or dict()	{'Jun': 75, 'Jul': 89}

Loops

- In Python, collections are containers or data structures that can hold multiple values or objects
 - They are used to group related data and provide convenient methods for manipulation and iteration
 - collections are considered **iterables** because are objects that can be looped/iterated over
- Two main types of loops
 - For loops
 - While loops

Environments

- All Python code is evaluated in the context of an environment
 - An environment is a sequence of frames
- Global frame: the top-level scope of your program, where variables and functions defined at the highest level are stored
 - Variables declared in the global frame are accessible from anywhere in your program
- Local frame: the local scope of a function
 - contains information about the function's parameters, local variables, for the execution of the function



Python Libraries

- Python libraries are collections of pre-written code that provide various functions, classes, and modules to help developers and analysts perform tasks without reinventing the wheel
- Common use cases for libraries
 - Data manipulation, Machine learning, Game development, Web development, Data visualization
- Python has a **standard library** where no installation is necessary,
 you just need to import them
- Other external libraries need to be installed using package management tools
 - Created by organizations, open-source, updated frequently











Importing Libraries

- To use functions from existing libraries we need to import them into our environment
 - Often, we shorten the name of a library to an **alias** so that it makes calling functions from the libraries easier

```
import numpy as np
import pandas as pd
import matplotlib.pylab as plt
%matplotlib inline
from matplotlib.pylab import rcParams
rcParams['figure.figsize']= 20,5
```

Function Calls From Libraries

When using functions from external libraries, we often use dot
 notation when calling functions

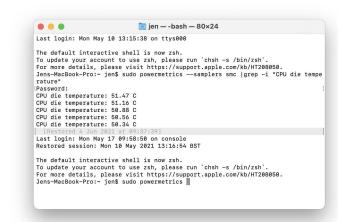
```
python lists
                                  numpy arrays
         a.sort()
                                  a.sort()
                                                 sorts in-place
         sorted(a)
                                  np.sort(a)
                                                 returns new sorted array
         a.sort(key=f)
                                                 key function
         a.sort(reversed=False)
                                                 ascending/descending
         python list
                                 VS
                                                   numpy array
a = [1, 2, 3]
                                            a = np.array([1, 2, 3])
b = a # no copy
                                            b = a # no copy
c = a[:] # copy
                                            c = a[:] # no copy!!!
d = a.copy() # copy
                                            d = a.copy() # copy
```

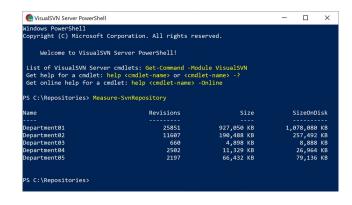
Command Line



Command Line

- For the purposes of this class, we won't go far into the details your computer's operating system or how it works
 - It's very complicated!
- Just know, in data science it is often used to
 - Make/Delete files/folders
 - Launch JupyterLab sessions
 - Interact with GitHub (using Git)
 - Install libraries/packages
- Command line is the broader concept of communicating with your computer's operating system but you will often hear "Terminal", "Shell", "Powershell" being used





Dependency Management

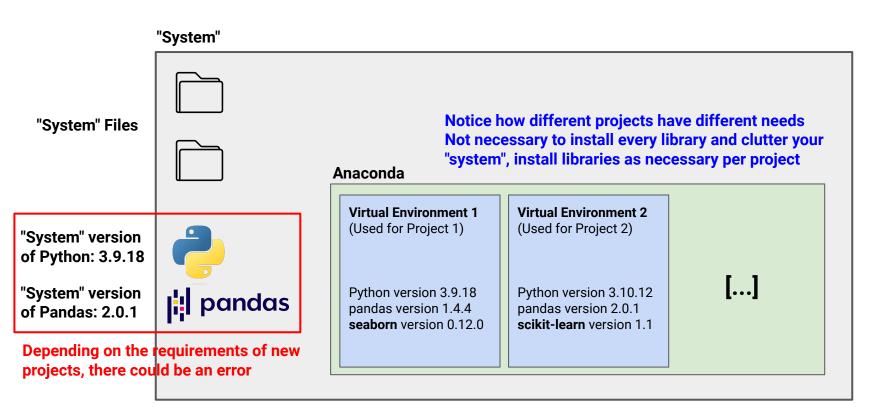


- Dependency management is the practice of handling external libraries, packages, and modules that your project relies on
 - Often as a data professional, you will work on different projects at different times – each with their own versions of libraries, configurations, etc.
- Not properly managing dependencies between projects will leave you with version conflicts, inconsistent computing environments, inconsistent results, errors, etc.
- Tools
 - o pip, Anaconda





"System" "System" Files Anaconda **Virtual Environment 1 Virtual Environment 2** (Used for Project 1) (Used for Project 2) "System" version of Python: 3.9.18 [...] "System" version Python version 3.10.12 Python version 3.9.18 pandas version 1.4.4 pandas version 2.0.1 of Pandas: 2.0.1 seaborn version 0.12.0 scikit-learn version 1.1



- Package managers allow users to easily install, update, and manage
 Python packages and libraries
- Virtual environments allow users to isolate and manage
 project-specific dependencies, ensuring that each project has its own
 clean and independent environment for installing packages and
 libraries
 - Isolation helps prevent conflicts between different projects and promotes consistency and reproducibility
 - Users can create, activate, deactivate, and delete virtual environments as needed, making it easier to work on multiple projects with different requirements on the same system

- Images from the scikit-learn <u>documentation</u>
 - If you don't manage your libraries, you'll
 manually have to update each dependency that
 does not meet the minimum requirements
- This is why using a package manager (pip, Anaconda)
 and creating new virtual environments (venv,
 Anaconda) for each project is considered a good
 practice
 - Anaconda can do both

Installing scikit-learn

Dependency	Minimum Version	Purpose	
numpy	1.17.3	build, install	
scipy	1.5.0	build, install	
joblib	1.1.1	install	
threadpoolctl	2.0.0	install	
cython	0.29.33	build	
matplotlib	3.1.3	benchmark, docs, examples, tests	
scikit-image	0.16.2	docs, examples, tests	
pandas	1.0.5	benchmark, docs, examples, tests	
seaborn	0.9.0	docs, examples	
memory_profiler	0.57.0	benchmark, docs	
pytest	7.1.2	tests	
pytest-cov	2.9.0	tests	
ruff	0.0.272	tests	
black	23.3.0	tests	
mypy	1.3	tests	
pyamg	4.0.0	tests	
sphinx	6.0.0	docs	
sphinx-copybutton	0.5.2	docs	
sphinx-gallery	0.10.1	docs	
numpydoc	1.2.0	docs, tests	
Pillow	7.1.2	docs	
pooch	1.6.0	docs, examples, tests	
sphinx-prompt	1.3.0	docs	
sphinxext-opengraph	0.4.2	docs	
plotly	5.14.0	docs, examples	
conda-lock	2.0.0	maintenance	

Warning: Scikit-learn 0.20 was the last version to support Python 2.7 and Python 3.4. Scikit-learn 0.21 supported Python 3.5-3.7. Scikit-learn 0.22 supported Python 3.5-3.8. Scikit-learn 0.23 - 0.24 require Python 3.6 or newer. Scikit-learn 1.0 supported Python 3.7-3.10. Scikit-learn 1.1 and later requires Python 3.8 or newer.