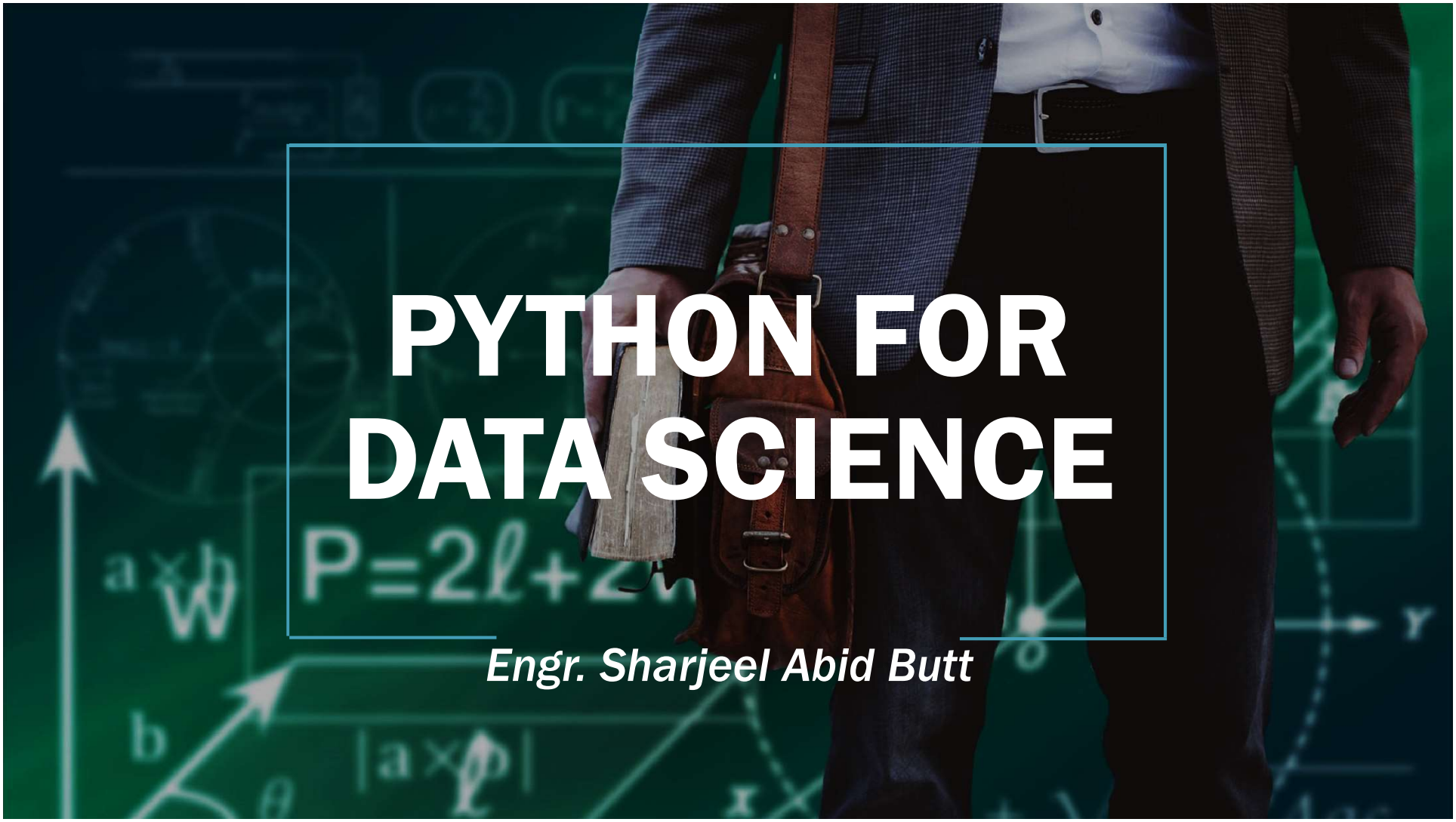

PYTHON FOR
DATA SCIENCE
Engr. Sharjeel Abid Butt

About me

B.E. Electrical Engineering (2009)

Instructor @ Department of Electrical Engineering, IIUI since January 2010

MS Electronic Engineering (2014)

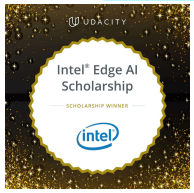
PhD Electronic Engineering (in progress)

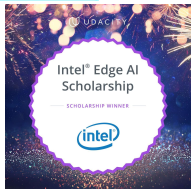
2019 Recipient **Facebook Secure & Private AI Challenge** Scholarship

2019 Recipient **Udacity Deep Learning Nano Degree** Scholarship

2019 Recipient **Intel Edge AI Challenge** Scholarship

2020 Recipient **Udacity Intel Edge AI Nano Degree** Scholarship





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- IPython / Jupyter: Beyond Normal Python
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What is Data Science?

There are a few existing definitions

Obtain, **Scrub**, **Explore**, **Model**, and **iNterpret** (**OSEMN**)

Mason and Wiggins, 2010

The "ability to [create] **prototype-level** versions of ... the steps needed to derive **new insights** or build **data products**"

Analyzing the Analyzers, 2013

Data Science exists to drive better outcomes

Using **multidisciplinary methods** to understand and have a positive **impact** on a **business process** or **product**

- **Route optimization** in a supply chain
- **Conjoint analysis** for product ideation
- **Attribution modeling** for connecting marketing spend to outcomes
- **Marketing spend optimization** for efficient outreach given a budget
- **Effectiveness testing** for creative or offers
- **Detecting fraud** in insurance claims
- Predicting and influencing **employee or customer retention**
- Understanding **who is likely to vote**

How do we *do* Data Science?

We **collaborate** across disciplines.

Not only do we need to speak the same **language of mathematics** we must **share similar processes and tools** to produce impactful data science.

Some of these processes and tools come from agile **product development** and **software engineering**.

Processes like **design sprints**, **project planning**, **planning poker**, and **daily standups**.

Tools like **version control**, **open source languages**, and linux **software containers.**

Why Python?

Python is one of these open source languages that you may **choose** to use.

It's a **full-featured** language with **many, many packages** for making data science tasks easier.

There are robust libraries and services for **testing** your code and methods

It makes it easy to write **defensive code**.

Readability counts and **style matters**.

Straightforward to go **from prototype to production**.

A **large community** of disciplined, helpful, and seasoned programmers.

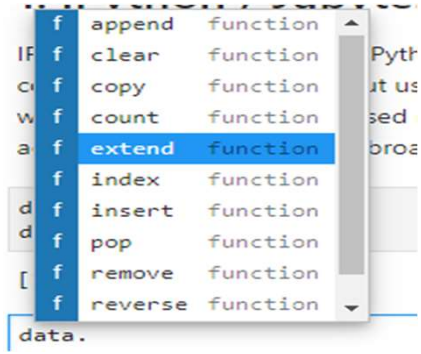
IPython / Jupyter

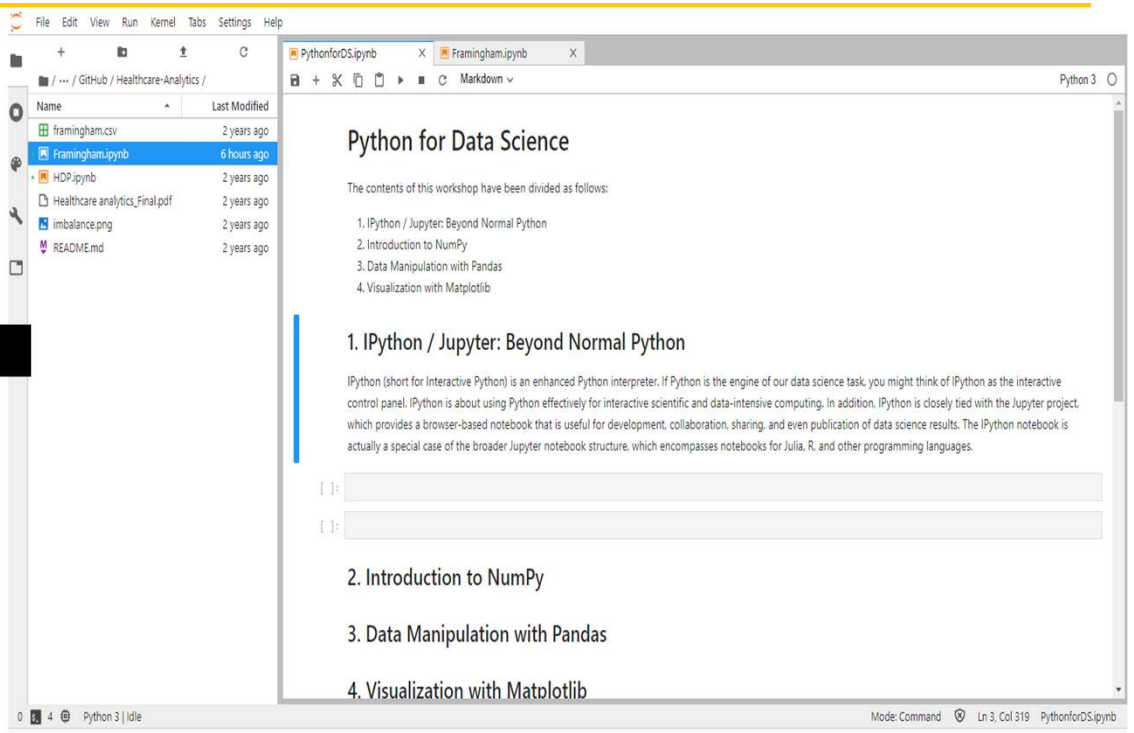
Environment:

Anaconda 3.7

Command:

```
(base) C:\Users\Sharjeel>jupyter lab
```





For detail about Markdown cells:
https://jupyter-notebook.readthedocs.io/en/stable/examples/Notebook/Working%20With%20Markdown%20Cells.html

Python --> NumPy

Data in Python can be broadly classified as:

Basic data types

- Numbers: Integers and floats work as you would expect from other languages
- Booleans: Python implements all of the usual operators for Boolean logic
- Strings: Python has great support for strings

Containers

- Lists: A list is the Python equivalent of an array
- Dictionaries: stores (key, value) pairs
- Sets: A set is an unordered collection of distinct elements
- Tuples: A tuple is an (immutable) ordered list of values.

NumPy is a wrapper around a library implemented in C allowing mathematical operations not directly / easily possible in Python.

NumPy

- **Core library for scientific computing**
- main object is the homogeneous multidimensional array which is a table of elements (usually numbers), all of the same type
- NumPy functions, being compiled, execute much faster than their Python counterparts

```
import numpy as np
d1 = np.array([1,2])          # 1D array
d2 = np.array([[1,2],[10,20]]) # 2D array
```

Pandas

- Python package for providing fast, flexible, and expressive data structures designed to make working with “relational” or “labeled” data both easy and intuitive. It aims to be the fundamental high-level building block for doing practical, real world data analysis in Python.

- **Pandas Series** is like generalized 1D NumPy array or a specialization of a Python dictionary.

- **Pandas DataFrame** is like generalized 2D NumPy array.

```
import pandas as pd
data = pd.Series([0.25, 0.5, 0.75, 1.0], index=['a', 'b', 'c', 'd’])
df = pd.DataFrame(np.random.rand(3, 2), columns=['foo', 'bar’], index=['a', 'b', 'c'])
```

Pandas --> Read a CSV File

CSV = Comma Separated Values
TSV = Tab Separated Values
JSON = JavaScript Object Notation

```
csv_file = open("IMDB-Movie-Data.csv")
reader = csv.reader(csv_file)
line = next(reader)
pprint(line)
```

VS

```
movies  = pd.read_csv("IMDB-Movie-Data.csv", delimiter=',')
```

Pandas --> Functions + Data Extraction

Functions:

.head()

.tail()

.info()

.describe()

.isnull()

.dropna()

Data Extraction:

Column:

```
data = movies[['Genre', 'Rating']]
```

Row:

- **.loc - loc**ates by name
- **.iloc- loc**ates by numerical **I**ndex

```
data = moviesT.loc['Prometheus']
data = moviesT.iloc[2]
```

Pandas --> SQL for Python

SQL:	Pandas:
SELECT TOP 5 * FROM movies	movies.head(5)
SELECT * FROM movies	movies
SELECT Title FROM movies	movies[[‘Title’]]
SELECT Title, Genre FROM movies	movies[[‘Title’, ‘Genre’]]
SELECT * FROM movies WHERE Year = 2014	movies[movies[‘Year’] == 2014]
SELECT * FROM movies where Year = 2014 AND Rating > 8	movies[(movies['Year'] == 2014) & (movies['Rating'] > 8)]
SELECT * FROM movies WHERE Year = 2014 OR Rating > 8	movies[(movies['Year'] == 2014) | (movies['Rating'] > 8)]

Matplotlib

- Most widely used Python visualization library/package.
- Cross-Platform
- Large Number of backends and Outputs
- Advanced usage is achieved by using Higher-level package like **Seaborn**
- Simplest plots are Line plot and Scatter plot

```
import matplotlib.pyplot as plt
plt.plot(x, y)
plt.scatter(x, y)
```

Final Words + Question/Answers