PG TECH WEEKEND

WnCC

Debriefing Data

IIT BOMBAY





GET STARTED WITH EXPLORATORY DATA ANALYSIS

See how we can extract valuable information from data and use it to infer stuff



LEARN LIBRARIES FOR DATA HANDLING

Numpy and Pandas to filter and manage huge repositories of data



VISUALISE EXTRACTED DATA

Using libraries like matplotlib to plot and find visual correlations between data samples



NUMPY

WHY SHOULD I USE NUMPY? AM I NOT DOING WELL ENOUGH WITH LOOPS AND LISTS?

NO.

Vectorisation over loops. Numpy arrays over lists.

Data analysis is computation-heavy. Using libraries like numpy optimises a lot of operations - for example matrix multiplication. Numpy arrays also take less memory than python lists.





Installation

We'll be using these Libraries

pip install numpy

pip install scipy

pip install pandas

pip install matplotlib

Getting started with numpy

with help from our trusty friend, python lists!

mumpy.array(object)

Where object is array-like (a python list, tuple, a nested python list or tuple) returns a numpy array constructed from this object

We can represent matrices using numpy arrays, where the object is a list of lists, like [[1, 2], [3, 4]]. The array formed is a 2D array, and we can make arrays of n dimensions similarly. These arrays are of the type <numpy.ndarray>





arange() and linspace()

Similar to the range(start, stop, step) in python

np.arange() returns a numpy array of evenly spaced values in the
 given interval [start, stop). Precision over the step

When the step sizes are non-integer, such as 0.1, it is best to use linspace instead of arange.









ones and zeros - shapes

np.zeros(shape) returns an array of specified shape, filled with zeros.

np.ones(shape) returns an array of specified shape, filled with ones.

np.reshape(array, newshape) or array.reshape(newshape) changes the shape of an array without changing its data.

This gives us good matrix, or an N-dimensional numpy array of a shape of choice to play and experiment with.





Random Numbers















MMMM





We use the **square bracket** to select a part of the ndarray object.

arr[8] # gives the element at index 8

arr[:5] # returns the array from the start to one before the end index arr[3:] # returns the array from the start index (included) to the last one

INDEXING FOR 2D ARRAYS

Square brackets again. This time, the elements might be arrays themselves matrix[i] # returns the ith row of the 2D matrix matrix[i][j] #returns the element at row i, column j matrix[i:] #returns all the rows after and including the ith row

CONDITIONAL SELECTION

myarray <operator> value*
returns a boolean array of
shape similar to myarray whose
elements are the condition
applied to every element in
myarray

```
>>> bool_arr =
np.array([1, 2, 3], [4, 5, 6]) > 3
>>> bool_arr
array([False, False, False], [True, True,
True])
```

*VALUE CAN BE ARRAY-LIKE, SEE BROADCASTING

ndarray[boolean array] returns
the elements of the array for
which the corresponding element
in the boolean array is true.

```
>>> arr = np.array([1, 2, 3, 4, 5],
[6, 7, 8, 9, 10])
>>> arr[arr%2==0]
[2, 4, 6, 8, 10]
```

Arithmetic Operations

You can add two arrays if they are of the same shape using just +

Matrix Multiplication:

If a and b are arrays, their matrix multiplication is done by a@b

Elementwise Product:
Done using *







With numpy you can apply functions for a single element for the entire array very efficiently.

Numpy does this automatically.

```
def addOne(a):
return a+1
```

```
x = numpy.ones(15, 40)
y = addOne(x)
```

Each element in y would be 1 + the corresponding element in x





Pandas

























SERIES DATA TYPE

The series data type is quite similar to a numpy array. The difference is that a **Series** object can have axis labels, meaning it can be indexed by a label, instead of just a number.

pandas.Series(data, index)

data is array-like and contains the data stored in the series index is array-like, of same length as data and contains the labels for the data values

```
Eg. data = [10, 20, 30]
index = ['A', 'B', 'C']
```

Thus the elements in data can be indexed by their corresponding labels too.

```
>>> my_series = pandas.Series(data=[1, 2, 3], index=['A', 'B', 'C'])
```

>>> my_series['A']







DATAFRAMES

DataFrames are a crucial data type in the Pandas library. They can be thought of as a bunch of Series objects put together to share the same **index**.

pandas.DataFrame(data, index, columns)

data is an ndarray or some other iterable and contains the data stored in the DataFrame

index is array-like, and specifies the indexing for the DataFrame
columns specifies the column labels for the data



```
Eg. data = [[90, 80], ['A', 'B']]
  index = ['Marks', 'Grade']
  columns = ['Alice', 'Bob']
```

Here columns are a labelled Series object which share the same indices, namely 'Marks' and 'Grade'







SELECTION OF ELEMENTS

To select a Series object, use **square brackets** [] with the label of the column.

>>> my_dataFrame[labels] # labels can be a list of column names

To remove a row or a column, we use the **drop(labels, axis) function**, where labels is the list of labels to remove, and axis specifies whether these labels are indices or of columns.

To select rows, we use the .loc[labels] syntax.

df.loc['Marks'] # the row with index 'Marks', returned as a Series object
 with column labels as indices





MISSING DATA



We can drop the rows or columns entirely which have a missing value.

dropna(axis, how, thresh) can be used.

Look at the Jupyter Notebook for parameter info.



CONDITIONAL SELECTION

Look at the Jupyter Notebook







INPUT AND OUTPUT



Pandas can read a variety of file types using its pandas.read_ methods.

CSV FILES

```
my_dataFrame = pandas.read_csv(Input_Filepath)
my_dataFrame.to_csv(Output_Filepath)
```

EXCEL SHEETS

```
my_dataFrame = pandas.read_excel(Input_Filepath, sheet_name)
my_dataFrame.to_excel(Output_Filepath, sheet_name)
```





Matplotlib

Make beautiful graphs to display your data import matplotlib.pyplot as plt











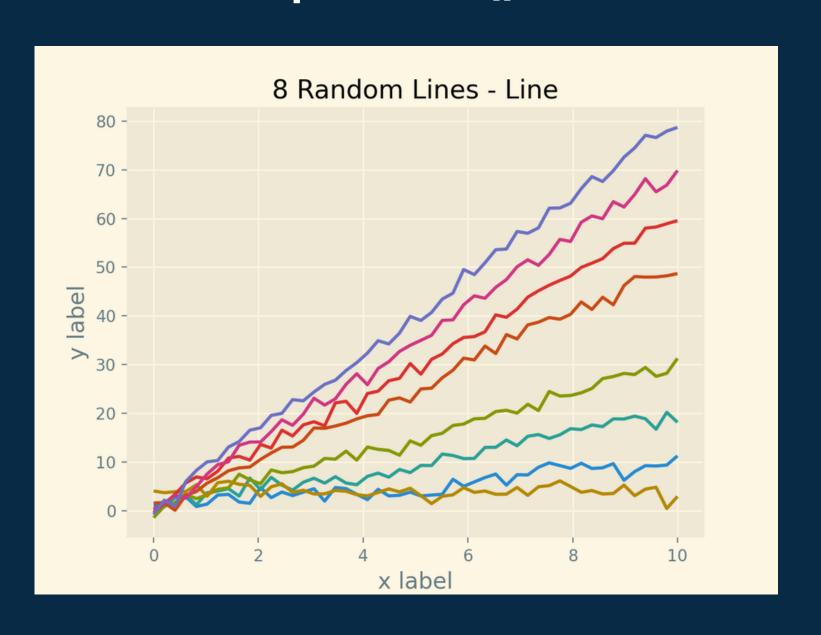






PLOTTING DATA









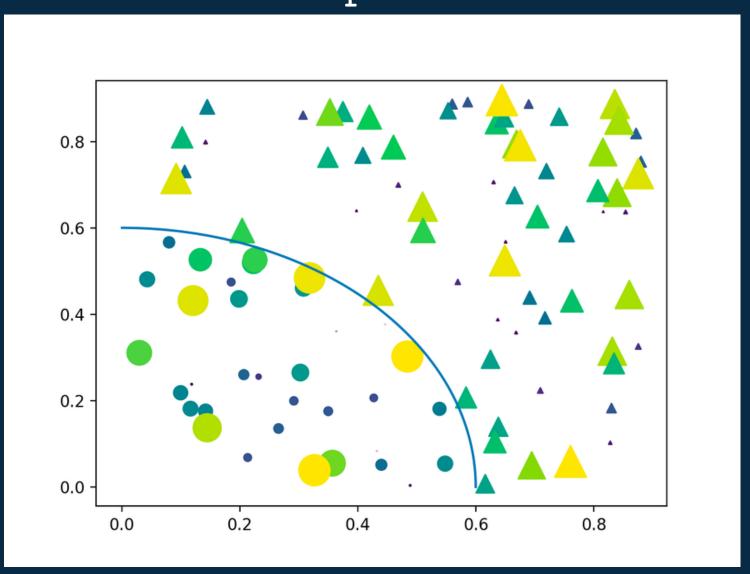






plt.scatter(x, y)

Some better properties than plt.plot if you just want to plot some points







HISTOGRAMS

Use the **plt.hist()** function to plot a histogram

plt.hist(x, bins, range, density)

x is the the data in form of an array
bins specifies the number of bins to divide the data into
range specifies the lower and upper range of the bins, can
be used to narrow down data points
density when True, will draw and return the probability
density: the histogram will be normalised such that the
areas add to 1



Tell me and I forget, teach me and I may remember, involve me and I learn.

- Benjamin Franklin



