

" Last Unit i.e  
Society Law and Ethics "

Read completely  
from  
previously shared  
" Must Read " notes

They are useful for not only  
Exams but also day to day  
activities of cyber world.

These summarised notes are  
to be read after doing one  
Revision from book. Then only  
these pages will work like  
Handy Quick Review  
Capsule.

Best wishes  
from

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Important statements

- import numpy as np
- import pandas as pd
- from pandas import DataFrame

→ pivot()  
→ needs 3 arguments  
index, col, values

pivot\_table()  
→ need 4 parameters  
index, col, values, agg func

- index becomes row
- col becomes column
- value shown as data

} → aggregate function of value shown as data.

Ex df-P = df.pivot\_table(index='item', column='company', values='Rupees', aggfunc=np.mean)

Dataframes are size mutable and also value mutable

properties of Dataframe

- index = row
- columns = cols
- dtypes = dtypes of data
- size = total no. of elements in df

- shape = dimension (r, c) (m, n)
- ndim = tot axes → 2 for every dataframe

\* leave this & you will get pivot.

→ Notation in Dataframe

• T = transpose of df → row to col, col to row

df.columnname  
df["colname"]  
df[['col1', 'col2', 'col3', ...]]

→ use of loc & iloc → to select a subset from DF

df.loc [start : end, start : end]  
row : row, col : col

df.iloc [start : end, start : end]  
index : index, index : index, col : col, col : col  
→ end index not included

# Dataframes

portion from book  
page 59 Add, delete

→ Add a col ⇒ `df['col'] = [val1, val2, ... valn]`

→ delete a col ⇒

`df.drop(['col1', 'col2'], axis=1)` } drop is dataframe  
`df.drop(columns=['col1', 'col2'])` } function.  
`del df['colname']` → del is command of python

delete a Row

`df.drop(['row1', 'row2'])`

by default axis  
is 0.

★ Dataframe functions

`df.min()` } similarly `mode()`, `sum()`, `count()`  
`df.max()` } `median()`, `corr()`, `std()`, `mean()`  
`var()`, `quantile()`

by default axis = 0 ⇒ and calculation done on each  
column

write axis=1 to calculate along rows.

Sorting of dataframe (By Value, By index) `sort_values()`

ex ⇒ `df1.sort_value('col', ascending=True/False)`

`df1.sort_values(by=['col', 'col'])`

`df1.sort_values('col', 'col')`

`df1.sort_values`

⇒ `df1.sort_value('col', axis=0/1, asc=T/F, na_position='last')`

★ Pipe function ⇒ `df.pipe(func1).pipe(func2).pipe(func3)`

ex) how it works (sandwiching of functions).  
`(func3 (func2 (func1)))` order of execution  
3 2 1

## Dataframes Contd...

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⇒ apply & applymap()

apply() → applies the function given as argument to one row or col.

applymap() → applies the function to all individual element <sup>every</sup>

ex → df.apply(function, axis=0/1) → 0-index/row  
1-column

df.applymap(function) → no need to give axis.

axis = 0  
↓ [1]  
↓ [2]  
↓ [3]  
heading

axis = 1  
→ [A]  
→ [B]  
→ [C]  
→ [D]  
heading

applymap() → result spread over every element (cumsum)

groupby() → df.groupby(by=label or list of labels, axis=0/1)

ex → df1.groupby('tutor') or df1.groupby(['col1', 'col2'])  
one column multi column

then group attributes can be used after saving groupby results in some Dataframe.

ex ⇒ df1 = df.groupby('tutor') then we can use following attributes.

→ 1) groups → give the list of all items in the group

\* 2) get\_group(group) → show the data of particular group  
get\_group('x')

→ 3) size → shows the group size (How many items in grp)

→ 4) count → count of non-Na value of each column in group

→ 5) head → list any specified col in detail  
→ df1[ 'col1' ].head()

⇒ Aggregation can be done in a grouped dataframe as you do in normal dataframe, using agg()

df1.agg([np.mean(), np.median(), np.mode()])

⇒ aggregation & groupby can be combined to get result in one single command.

df1.groupby('col').agg([np.mean, ...])

Transform → agg function reduces the data by giving summary. Transform() function returns the transformed version of data by adding repeated rows of result so that shape of original data & transformed data matches.

ex:- if aggregate data has only 4 rows. Transformed data will original data has 12 rows then convert agg data also to 12 rows by repeating result at same groupby col data.

```
dfnew = df1.groupby(['f1', 'f2']).transform(np.mean)
```

reindex\_like() → df.reindex\_like(dfnew)   
 ↑ will copy index from dfnew

\* Reindex, Reshape, Rename, reindex\_like() (small letters)

⇒ Rename → rename() function renames the existing index/columns provide the data in form of dictionary where {key1: value1, key2: value2, key3: value3...}   
 ↑ old index      ↑ new index

ex ⇒ df.rename({'dict data'}, axis=0/1, inplace=True/false)

→ df.rename({'Q1:A1, Q2:A2, Q3:A3'}, inplace=True)   
 All change in df.

→ df1 = df.rename({'Q1:A1, Q2:A2, Q3:A3'}, axis=1, inplace=False)   
 change in column, changes saved in df1. df is as it is.

⇒ reindex() ⇒ change the order of existing index or recreate index

```
df.reindex(['q4', 'q1', 'q3', 'q2'])
```

for row { or df.reindex(index = [new index])

for col → df.reindex(columns = [new values]   
 for index

→ if new col/row is defined then use fill\_value to give new value   
 ex original df has only 2 col, new df after reindex has 4

```
→ df.reindex(columns = [1, 2, 3, 4], fill_value = 50)   
 ↑ new cells → filled by 50
```

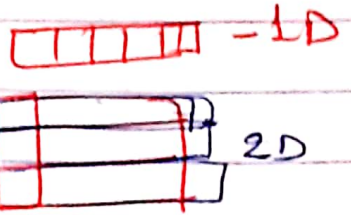
# Numpy (Unit -1 topic 2)

import numpy as np → library linking

→ Arrays  $\begin{cases} 1D - \text{vectors} - \text{single row/col} \\ 2D \rightarrow \text{multi dimension} \end{cases}$  MR/MC if  $M=2$ , 2D Array

→ homogenous, continuous set of element.

→ 2D arrays stored in Row Major / Col Major pattern.



## → Creating 2D Array

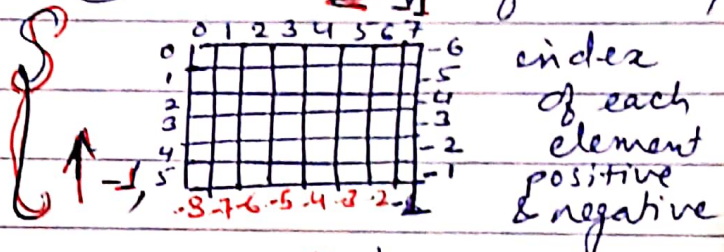
```
a = np.array([ [ , , ], [ , , ] ]) → ①
```

```
b = np.array([ , , , , , , ]) → ②
```

```
c = np.reshape(b, (m, n)) → new shape of 2D Array
```

## → Slicing A [start:stop:step]

B [start:stop:step, start:stop:step]



default → start=0, stop=dimension, step=1 \* step=-1 (negative means reverse traverse)

→ X = A [ :: -1 ] → reverse of A (1D array)

Y = A [ :: -1, :: -1 ] → reverses both row & col in any 2D array → bottom-up / Right-left

Z = A [ 0 : m ] ⇒ traversal from 0 to m-1

properties → shape, dtype, itemsize  $\left\{ \begin{array}{l} a.shape \rightarrow (2, 3) \\ a.dtype \rightarrow \text{type of data of Array} \\ a.itemsize \rightarrow \text{size in bytes of each elements} \end{array} \right.$

## \* difference b/w numpy Array & list \*

- | Array                                       | List                               |
|---|------------------------------------|
| → size not changeable                       | → size can be change               |
| → homogenous items                          | → different type of datatypes item |
| → less space                                | → more space in memory             |
| → vectorized operation effects each element | → not allowed vector operation     |

split: subset of Arrays

- \* split(A, (m,n))
- \* hsplit(A, n)
- \* vsplit(A, n)

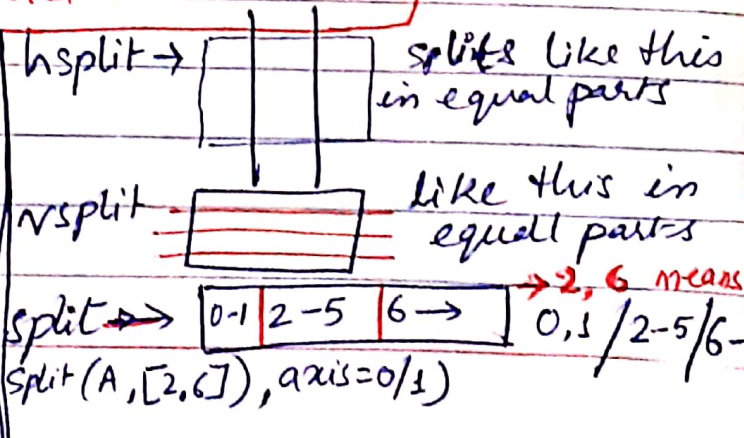
$L \times 2 \Rightarrow L, L \rightarrow L[1,2] \rightarrow [1,2,1,2]$

elements extended

then [2,4] elements multiplied



Concatenate()  $\rightarrow np.concatenate((A1, A2), axis=0/1)$



1D, 2D Array

Array slices, joins, subset

Arithmetic operation on 2D

Covariance, Correlation, Linear Regression

Plot  $\rightarrow$  bar, histogram, frequency polygons  
boxplot, scatter plots

## numpy Contd...

★ → Arithmetic operations +, -, \*, /, % etc works like mathematical operation in vector sequence. ex

$$A = \begin{bmatrix} A1 & A2 \\ A3 & A4 \end{bmatrix} + B = \begin{bmatrix} B1 & B2 \\ B3 & B4 \end{bmatrix} = \begin{bmatrix} A1+B1 & A2+B2 \\ A3+B3 & A4+B4 \end{bmatrix}$$

'+' is distributed →  $A+B$

★ Covariance ⇒ np.cov() method

→ np.cov(arr1, arr2) ⇒ 02 arrays needed to find covariance by elements of arr1 & arr2.

→ if array is 2D or multidimension → np.cov(ARR)

ND array

np.cov(a,b)

cov (a,b)	cov*
cov*	cov
(b,a)	(b,b)

cov(a,b) = cov(b,a)  
both are equal.

cov(a,a) ⇒ is called variance of a  
cov(b,b) ⇒ is variance of b

covariance zero (0) means not related  
-ve means ⇒ negatively related, +ve ⇒ positively related

★ Correlation → it is also called as normalised covariance

np.corrcoef(arr1, arr2) → correlation coefficient is calculated

output

corrcoef	corrcoef*
a, a*	a, b
corrcoef	corrcoef
b, a	b, b*

→ both matrix (array) must have same shape.

\* ⇒ is always 1

→ positive correlation of 1 means highly related data

→ negative means when one data increases other decreases.

★ → linear regression ⇒ A method to find relationship b/w dependant variable and a set of independant variables.

① → np.polyfit(x, y, deg)

↳ degree of freedom.

★ Random numbers random module → import random

random() → any random number b/w 0 to 1 (less than 1) < 100

randint(a,b) → generated integer N such that a) = N <= b (a,b included)

randrange() → randrange(start, stop, step)  
Any number between a & b with step value s.

for math function → import math



# Plots and Graphs Data Visualisation

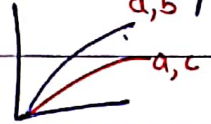
Library needed  $\rightarrow$  `import matplotlib.pyplot as plt`  
install it using `pip install matplotlib`

pyplot  $\rightarrow$  collection of methods to plot 2D graphs/plots, available under matplotlib.

matplotlib  $\rightarrow$  in idle do a pip installation.  
 $\hookrightarrow$  in Anaconda, it is pre-installed  
 $\hookrightarrow$  It is a high quality plotting library of python.

## plotting

1) line chart  $\rightarrow$  `plt.plot(a, b)`  $a, b$  are arrays.  
*\*(not mentioned in syllabus) still study it too.*



in same plot if you want to plot another line use `plt.plot(a, c)`

attributes  $\rightarrow$  you can change color of line, type of marker, its size etc

`plt.plot(a, b, color=code, marker=code, markersize=integer,  
markeredgecolor=code)`  $\leftarrow$  also linestyle, linewidth etc

similarly for every graph you can add xlabel, ylabel, title etc  
 $\rightarrow$  `plt.xlabel("your label")`, `plt.ylabel("your label")`

`plt.title("title of chart")`

$\rightarrow$  these attributes and method are generally for all plots

2) scatter plots  $\left[ \begin{array}{l} \text{using plot()} \\ \text{using scatter()} \end{array} \right.$  } Scatter chart is a graph of scattered points b/w 02 data sets.

in plot method use ~~marker~~ " " line color and marker style

in following manner  $\Rightarrow$  "r+", "b+", "bo" or any marker type without giving linestyle attributes you will get scattered chart. ex- `plt.plot(a, b, "o")`  
ex- `plt.plot(a, b, "r+")`

plt.show()  
to show the plot

in scatter method  $\rightarrow$  `plt.scatter(a, b, s=_, c=_, marker=)`

s  $\rightarrow$  marker size  
c  $\rightarrow$  marker color  
marker  $\rightarrow$  marker style  
ex  $\rightarrow$  `plt.scatter(a, b, marker="+")`

★ Creating Bar / histograms

Bar graph → Bars of different height are used to display the data → `plt.bar(a, b)`

first sequence forms x axis. use xlabel, ylabel  
second sequence forms y axis. title methods to draw better.

→ `plt.bar(a, b, width = floatval)` → for common width  
`plt.bar(a, b, width = [0.5, 0.6, 0.8, 0.9])` for variable width.

plotting 02 bars in same plots, [a small trick]

A = [2, 4, 6, 8]    B = [3, 5, 7, 9]    X = [1, 2, 3, 4]

`plt.bar(X, A, color='red', width=0.35)` first bar

normally we <sup>but</sup> change this as → `plt.bar(X+0.35, B, color='b', width=0.35)`  
Plot → `plt.bar(X, B, color='blue', width=0.35)` 2nd bar.

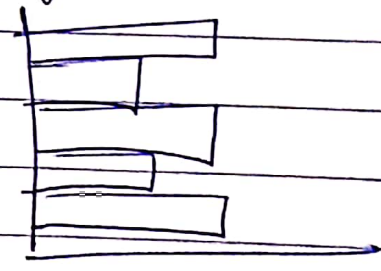
\* if we plot in normal way bars will overlap to add the width of 1st graph in x array of 2nd graph



→ now

`plt.barh(a, b)`

will create a horizontal bar plot



★ Histogram → `df.hist(column='ari')` → simplest statement  
continuous data set / no gaps in bar → it will create histogram of bin size = 10.

`df.hist()` is available under Dataframe and creates histogram for all numeric column if given as `df.hist()`

in pyplot. it has following syntax

`plt.hist(x, bins = __, weight = __, cumulative = T/F, color = __, edgecolor = __, histtype = __)`

when `histtype` → `bar` = histogram  
→ `step` → frequency polygon.

few examples of histograms

Histogram  $\rightarrow$  `pl.hist(x, bins=30, cumulative=True)`

frequency polygon  $\rightarrow$  `pl.hist(x, bins=20, histtype=step)`

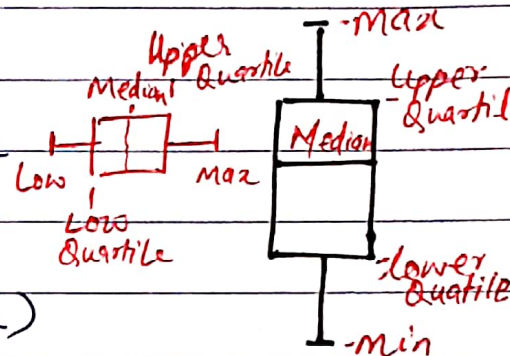
two nd arrays  $\rightarrow$  `pl.hist([x, y])`

horizontal orientation  $\rightarrow$  `pl.hist(x, bins=50, orientation='horizontal')`  
 $\rightarrow$  horizontal orientation

### Box Plot

Box plot displays

- min range value
- max range value
- Upper Quartiles
- Lower Quartiles
- median



`pl.boxplot(Data, showmeans=True)`

properties

- `notch` -  notched box when True,  rectangular box when False
- `vertical` -  `vert=True`,  `vert=False`
- `meanline` -  shows mean as a line
- `showbox` - by default True  $\rightarrow$   True,  False
- `showmeans` -  True,  False

will show mean point also when True

No box is shown when `showbox` is False

### SQL and Database

install  $\rightarrow$  `import mysql.connector`  
 $\rightarrow$  `pip install mysql-connector-python`

1 how to connect.

2 import package

`import mysql.connector as sqltor`

3 open a connection

`mycon = sqltor.connect(host="localhost", user="root",  
password="yourMySQL password", database="yourdb")`

4 create cursor

to test connection  $\rightarrow$  `if mycon.is_connected():` suitable msg  
 $\rightarrow$  `cursor1 = mycon.cursor()`

5 Execute a Query

`cursor1.execute("SQL Query as per Question")`

It contains methods like - `fetchall()`, `fetchone()`, `fetchmany(2)`, `cursor.rowcount`

6 Extract data

`data = cursor1.fetchall()`  
`data = cursor1.fetchmany(5)`  
`data = cursor1.fetchone()`

works after fetching  
`data = cursor1.fetchall()`  
for row in data:  
print(row)  
if row

and proceed like tuple of python  $\rightarrow$

7 clean Environment  
`cursor1.close()`  
 $\rightarrow$  `mycon.close()`

# DJANGO pronounced Jango-h

OSS, web application framework

→ Instagram, pinterest

→ deep, dynamic sites with web enabled application.

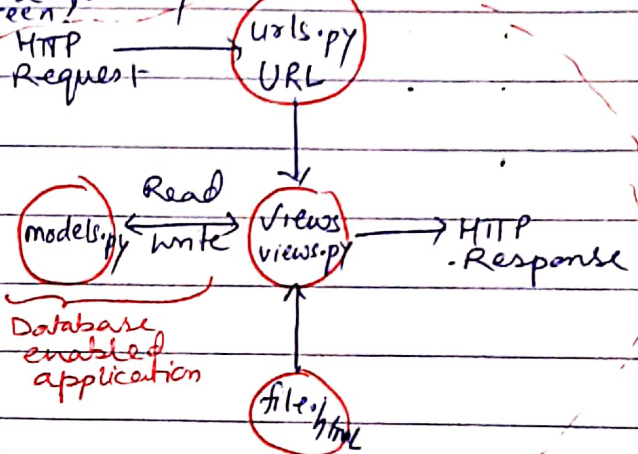
Advantages of Using Django (It is written in Python)

- ↳ Easy and fast web application development
- ↳ Compatible with major OS and databasess.
- ↳ less coding required
- ↳ Easy to extend
- ↳ possesses scalability

How Django works

Frontend → User end

It send request  
ex. Browser, or any  
user Interface screen



Backend: End which responds to the user's request usually any DBMS or RDBMS.

HTTP Get Request :- a way of retrieving information from a web server through some URL.

HTTP Post Request :- a way to send data to server, may be via HTML forms.

- Install the django in system
- Creating a project in Django (poj is project)
- run the Django server (127.0.0.1:8000)\* or (localhost:8000)
- Creating any application (myapp)

- pip install django
- django-admin startproject poj
- manage.py runserver
- manage.py startapp myapp

Virtual environment, is not necessary but if any user creates it ensures co-existence of multiple python environment of different Django projects in isolated manner.

- To create Virtual Environment
- pip install virtualenv
- Activate Virtual Environment (env\Scripts\activate)
- virtualenv venv
- create venv at Virtual environment

file in outer project folder → manage.py

- files in inner project folder
  - \_\_init\_\_.py
  - settings.py → most of settings
  - urls.py → stores into of URLs
  - wsgi.py

- files in app folder
  - admin.py → Configuration file for Django Admin
  - apps.py → configuration file for app
  - models.py → DB model defined here
  - tests.py → app specific tests
  - views.py → resolve http requests
  - migrations / → updates DB.

## Get & Post Method

DJANGO. contd

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2 Type of HTTP Request  
Get ← → Post

→ limited data can be sent → large data can be sent

→ data sent in header → data sent in body of page

→ Not secure as data is visible in header → secure as data not shown in URL.

→ Can be bookmarked → can't be bookmarked

→ Request remains in browser history → do not remain in browser history

\* CSV ⇒ comma separated values \*

↓  
Read & write in CSV  
May be helpful in few programs

Django  
Follows

Model

View

Template

MVT

paradigm  
app gets data from Model,

View works on data and template displays the final processed data.

# Society, law and Ethics

→ IPR - Intellectual Property Rights: Rights of the Owner of Information to decide how much info to be shared, exchanged or distributed & to decide cost/price for the share/distribute / Exchange info. <sup>Why to protect →</sup> ensures distributi

Plagiarism <sup>Why →</sup> It has to be protected for following reason

- ① It encourages creation of new s/w & ideas, improves.
- ② Ensures wide distribution of innovative idea/techno.
- ③ promotes investment in National Economy

Solutions → Patent, Copyright, TradeMarks,

<sup>-stealing  
-cheating  
-academic offence</sup> Plagiarism → stealing & pretending own property. → It is unauthorized use of someone else's thoughts, Intellectual work and representing it as own work, without citing sources or original generator name.

→ To avoid plagiarism → GIVE DUE CREDIT TO OWNER.

\* Plagiarism can be planned or may happen accidentally.

→ DRM → Digital Rights Management: - Assets you own in digital form is called digital property/estate.

DRM is a scheme that protects or controls the access to the Copyrighted materials by using technology. It can be done by encrypting, coding, password protections, time bound sharing, obtaining license, putting legal clauses, limiting the sharing, Anti-temper solutions.

Few threats to DRM: - ① Digital s/w penetration tools.  
② plagiarism or stealing.

→ Licensing: - It is a document that provides legal bindings and guidelines to the person to perform Digital Right mgmt and secure IPR. It includes rules like: -

① fair use of s/w ② Limitation of liability, warranty disclaimers & protections.

GPL → (GNU) → General public licence → free s/w licence means freedom for use, not necessarily free from price.  
ex - Wordpress).