Numpy — fast & convenient.

```
import numpy as np
l = range(1000)
```

```
l1 = range(size)
l2 = range(size)
a1 = np.arange(size)
a2 = np.arange(size)
result = [(x+y) for x,y in zip(l1,l2)]
```

Advantage:
1- Convenient.
2- Faster
3- Less memory

```
a = np.array([[1,2],[3,4],[5,6]])
a.ndim
```
↳ To print dimension.

a.itemsize
↳ Byte size.

```
a = np.array([[1,2],[3,4],[5,6]])
```
↳ a.size → 6.
a.shape = (3,2).

np.zeros((3,4)) = 3x4 zero matrix.
np.ones((3,4)) = 3x4 1 matrix.
np.arange(1,5) → array([1,2,3,4])
np.arange(1,5,2) → array([1,3]).
a.ravel() → Make 1-D
a.min(), a.max(), a.sum(), a.sum(axis=0).
np.sqrt(a)
np.std(a) → standard deviation

Slicing, Indexing etc:

```
n = [6,7,8]
n[0:2]
```
↳ [6,7]

a[0:2,2]
0-1 → 2nd element
([8,3])

np.vstack((a,b))
↳ one over other

np.hsplit(a,3) → split into 3 array.

Iterate numpy array using nditer :-

```
for row in a:
    for cell in row:
        print cell.
```

→

```
0
1
2
3
1
10
```

```
for all in a.flatten():
    print(all)
```

→

```
0
1
2
1
10
```

Also, `for x in np.nditer(a, order = 'F'):`
`print(x).`

C order: ⊇

Fortran order : (F) ∪

Pandas

- Pandas is a python module that makes data science extremely easy & effective.
- Process of cleaning messy data is called data munging or data wrangling.

→

```
import pandas as pd
df = pd.read_csv(—)
df.
```

Different ways of creating df :-

1 - `df = pd.read_csv(—)` or 2-`pd.read_excel(—)`

3- `df = pd.DataFrame(df, columns = [- - -]).`
`df` ↳ List

4- ↳ But instead of List - Dict -

Reading / Writing in Excel or CSV

```
import pandas as pd
df = pd.read_csv(—)
df
```

```
pd.read_csv(—, skiprows=1)
df.
```

```
pd.read_csv(—, header=None)
```
↳ 0-header column.

```
, na_values = ("not available", "n.a.")
```

```
, header = False
```
→ Header deleted.

Read_Excel

```
import pandas as pd
df = pd.read_excel(—, "sheet1")
df.
```

Similarly we use excelwriter to import excel file etc.

Handling missing data

1) fillna() → df.fillna(0) → Fill null with 0.

```
df.fillna({
    'temp': 0,
    'wind': 0,
    'event': no})
```

```
df.fillna(method = "bfill")
```
↳ Fill prev value either horizontally or vertically.

df.interpolate()
↳ Give interpolate value to blank/null.

df.dropna() → Drop null value.

df.dropna(how="all") → whose entire row is blank.

Replace function

```
df = df.replace(-999, np.NaN)
```

Regex → Regular expression

```
df.replace({'[A-Za-z]': ''}, regex=True)
```

↳ Remove alphabets

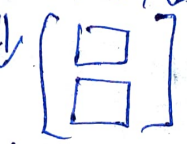
GroupBy

```
g → df.groupby('col')
g.max(), g.mean(), g.describe → Give all details.
```

Concat in Pandas

→ `df = pd.concat([ind_wise, us_wise], ignore_index=True)`

↳ Continuous index



`, key=[ind, us], axis = 0 or 1`

~~→ s=pd.Series(name = "Event")~~

→ `s = pd.Series(["_", "_", "_"], name="Event")`

`df.concat([temp_df, s], axis=1)`

Merge in Python

→ `df = pd.merge(df1, df2, on = "city", how="outer")`

→ ✓

Pivot & Pivot_table

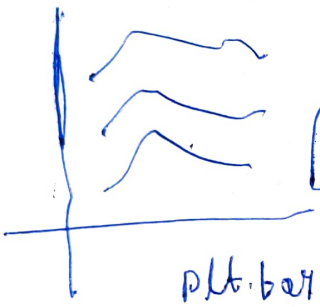
→ df.pivot(index='humidity', columns = "city")
, aggfunc = "sum"

Also margins use.

Matplotlib

import matplotlib.pyplot as plt

x = [1, 2, ___, 7]
y = [S0, S1, S2, - - -, 96]
plt.plot (x, y, color = "green", linewidth=5, linestyle='dotted')
→ plt.label ('Day')
plt.ylabel ('Temp')



→ plt.plot (loc = "best")
, shadow = "True"

plt.bar (—, →) ← for bar chart.
plt.bar (—, label = "Revenue")
plt.bar (—, label = "Profit") >

Histograms

blood_sugar = [100, 110, 102, —]
plt.hist (blood_sugar, bin = 3, rwidth = 0.95)
, label = [men, women], orientation = 'horizontal')

Pie Chart

```
exp_val = [1400, 600, 800, 700]
exp_labels = ["Home", "Wash", "Bike"—]

plt.axis("equal")
plt.pie(exp_values, labels = exp_labels, radius = 1.5,
        autopct = —)
```