

## 2.Fundamentals of python

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## Why Python?

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- Simple and easy to use
- free software and open source
- Interpreted
- Dynamically typed
- Extensible
- Embedded
- Extensive Libraries
- Usability
  - Desk top and web applications
  - Data base applications
  - Networking applications
  - Data Analysis (Data Science)
  - Machine Learning
  - IoT and AI Applications
  - Games



## ***Objectives***

---

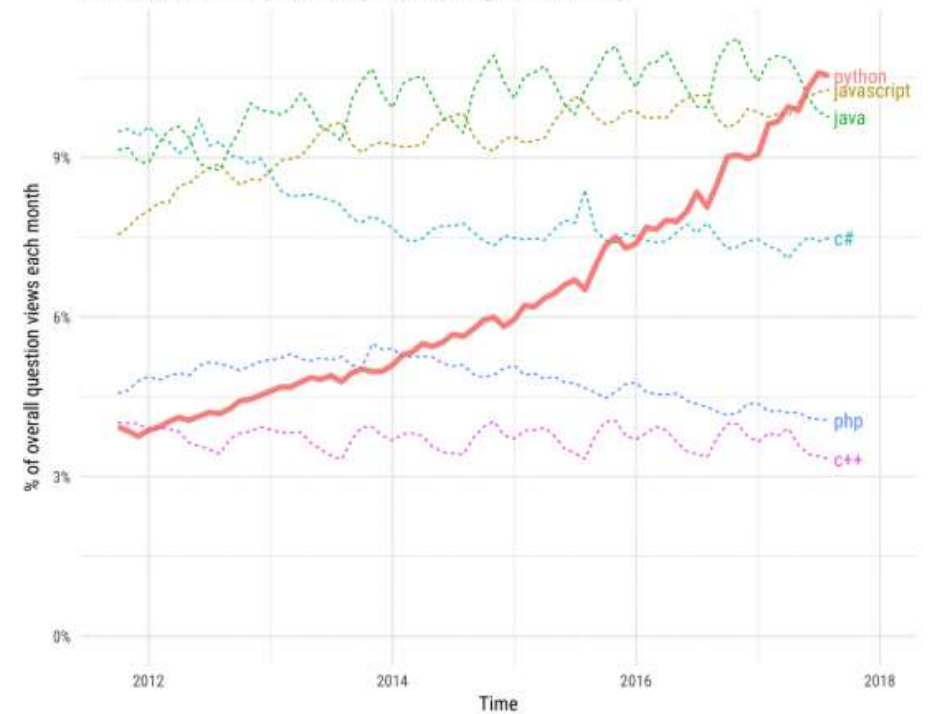
- Installing python
- Fundamentals of python
- Data visualization

## Top Companies using python



### Growth of major programming languages

Based on Stack Overflow question views in World Bank high-income countries



## ***Jupyter Note Book***

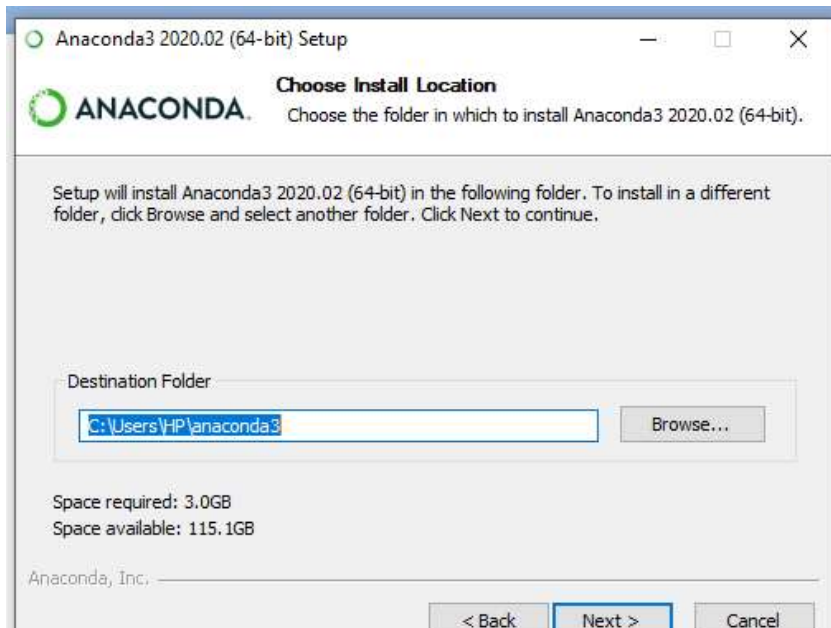
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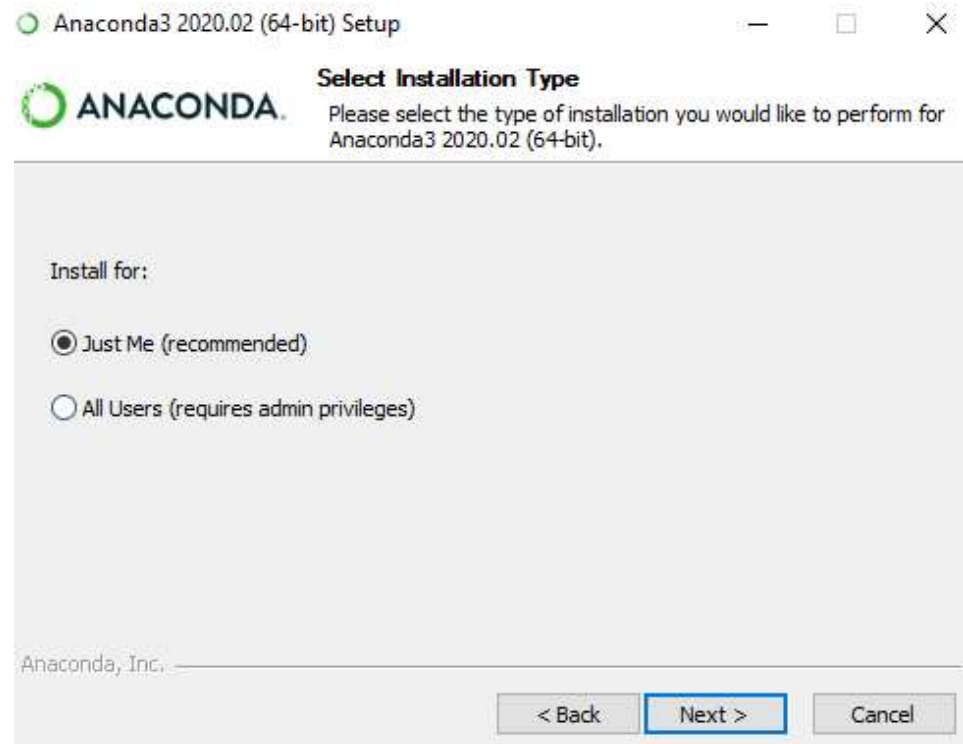
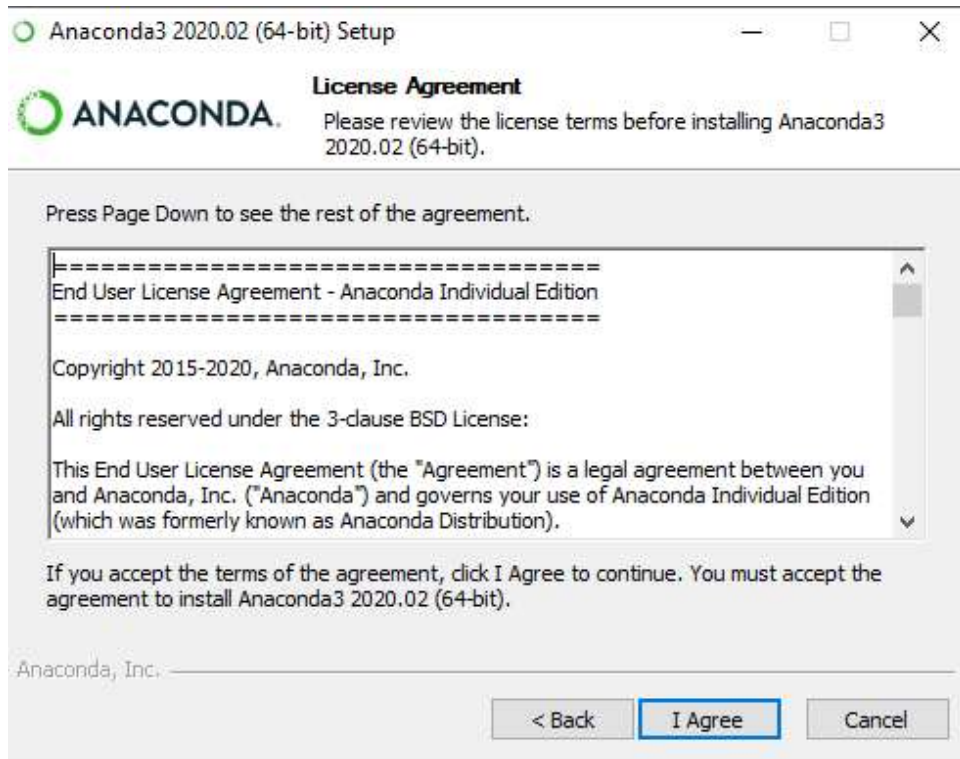
- client-server application
- Edit code on the web browser
- Easy in documentation
- Easy in demonstration
- User friendly



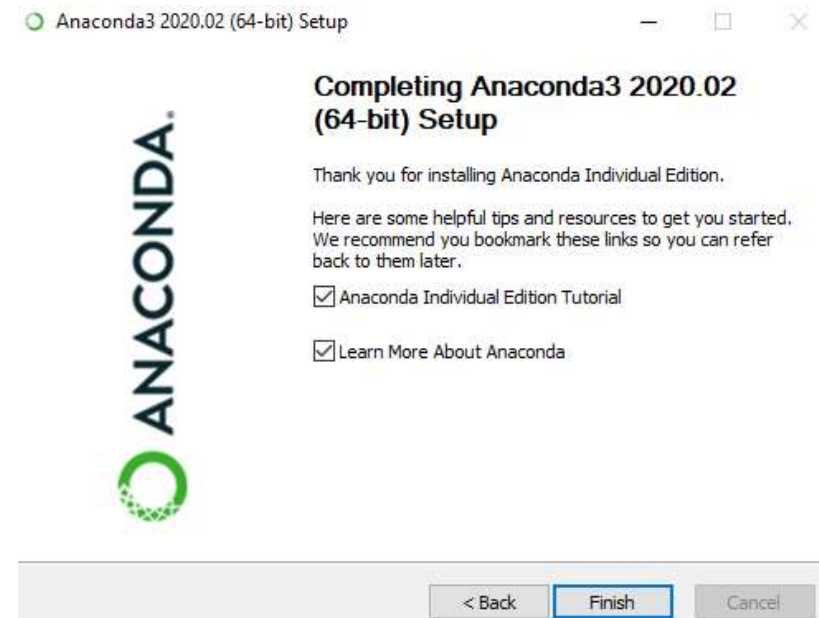
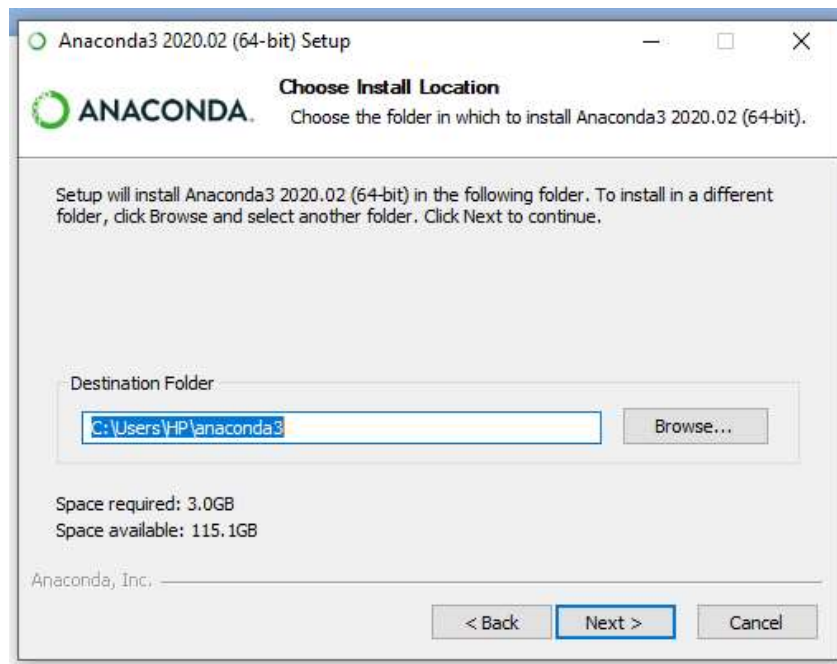
# Installation of Python

- Installation Process
- Step 1: Type [https://:www.anaconda.com](https://www.anaconda.com) at the address bar of the web browser
- Step 2: click on download button
- Step 3: download python 3.7 version for windows OS
- Step 4: Double click on file to run the application
- Step 5: Follow the instructions until completion of the process



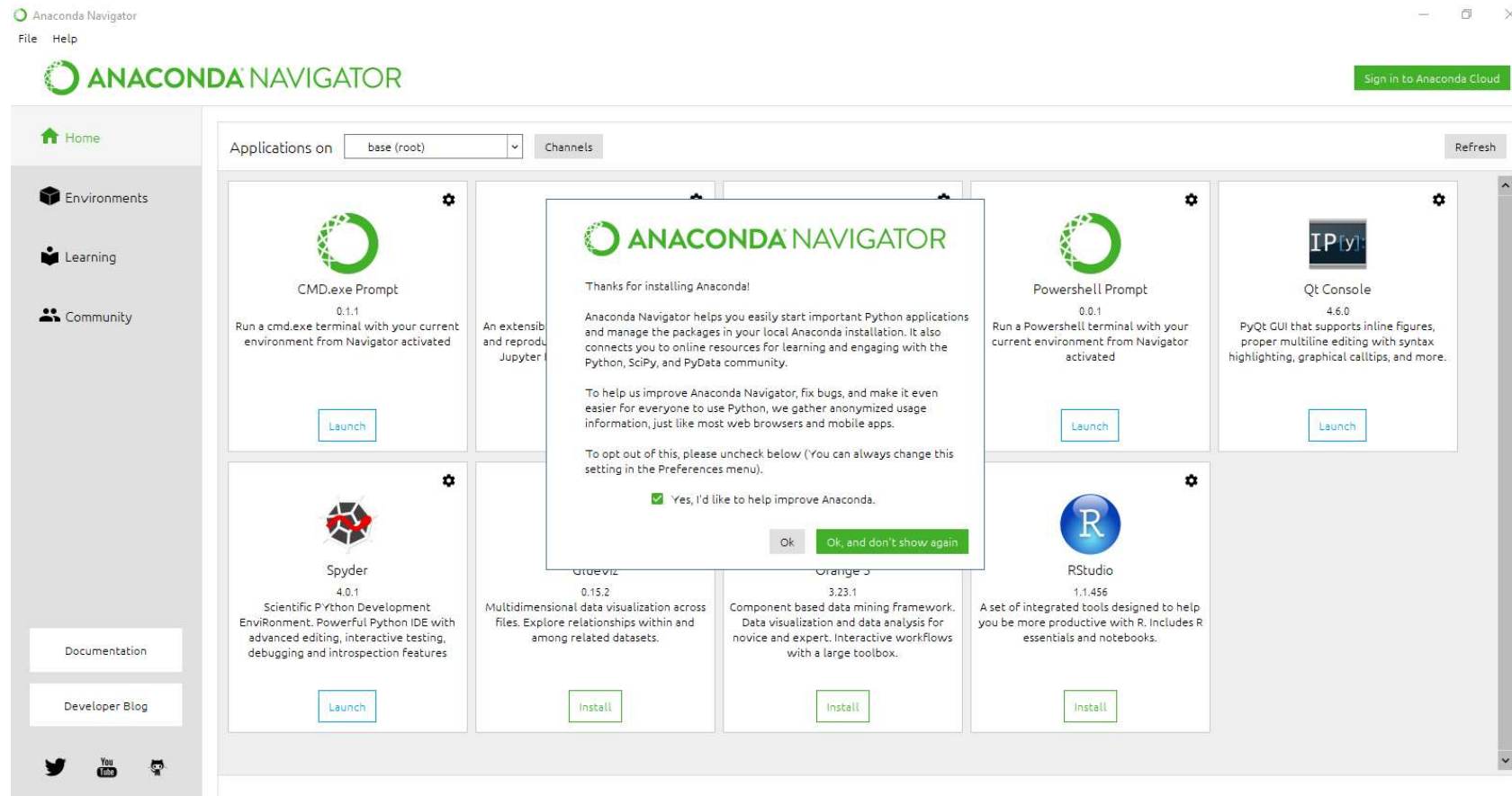


## Python Installation Process





# Python Installation Process



# Python Installation Process

Anaconda Navigator

File Help

ANACONDA NAVIGATOR

Sign in to Anaconda Cloud

Home




Environments

Learning


Community

Documentation

Developer Blog




Applications on base (root) Channels Refresh



CMD.exe Prompt  
0.1.1

Run a cmd.exe terminal with your current environment from Navigator activated


Launch



JupyterLab  
1.2.6

An extensible environment for interactive and reproducible computing, based on the Jupyter Notebook and Architecture.


Launch



Notebook  
6.0.3

Web-based, interactive computing notebook environment. Edit and run human-readable docs while describing the data analysis.


Launch



Powershell Prompt  
0.0.1

Run a Powershell terminal with your current environment from Navigator activated


Launch



Qt Console  
4.6.0

PyQt GUI that supports inline figures, proper multiline editing with syntax highlighting, graphical calltips, and more.


Launch



Spyder  
4.0.1

Scientific Python Development Environment. Powerful Python IDE with advanced editing, interactive testing, debugging and introspection features


Launch



Glueviz  
0.15.2

Multidimensional data visualization across files. Explore relationships within and among related datasets.


Install



Orange 3  
3.23.1

Component based data mining Framework. Data visualization and data analysis for novice and expert. Interactive workflows with a large toolbox.

Install



RStudio  
1.1.456

A set of integrated tools designed to help you be more productive with R. Includes R essentials and notebooks.

Install

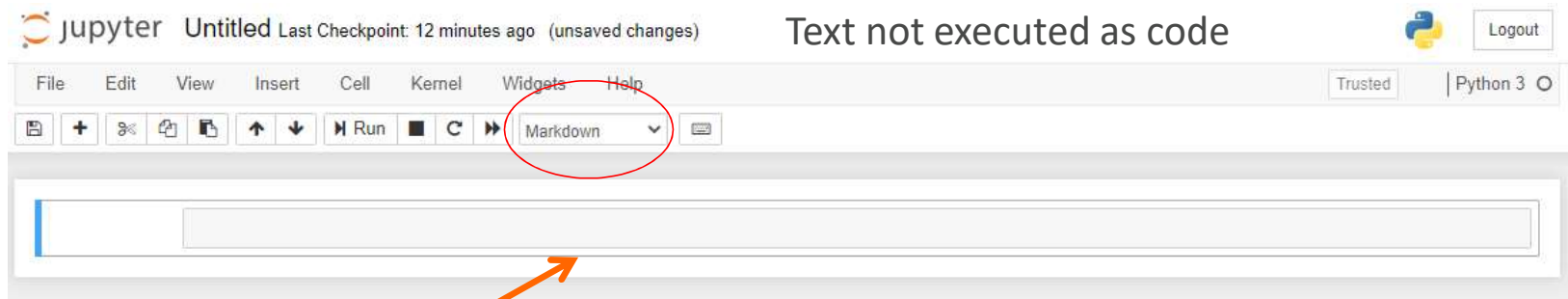
10

# About Jupyter Notebook

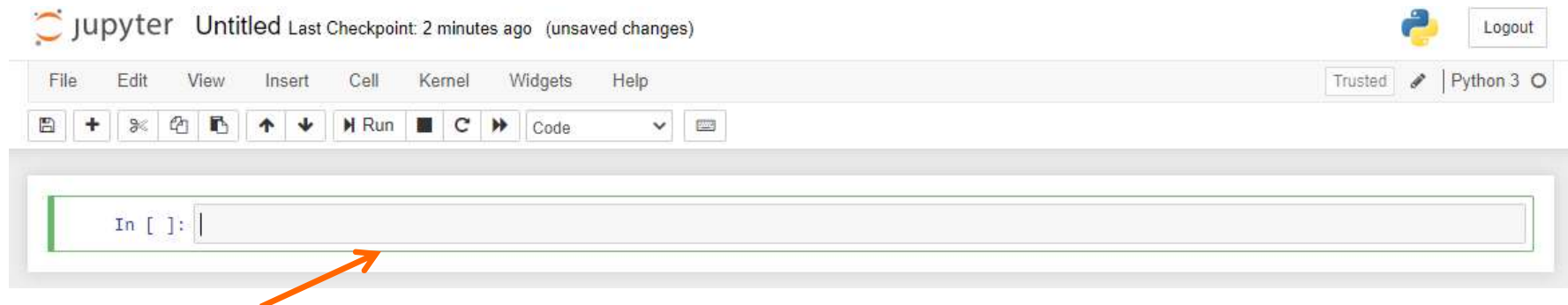


## About Jupyter Notebook

It contains documentation  
Text not executed as code



Cell<-Access using enter key



Input Field<-Green color indicated Edit Mode  
<-Blue color indicates command mode

## About Jupyter Notebook

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- Command allows to edit notebook as whole
- To close edit mode- press escape Key
- Execution three ways
  - Ctrl + Enter (output field can not be modified)
    - run selected cells
  - Shift + enter (output field is modified)
    - run the current cell, select below
  - Run Button on Jupiter Interface
  - Alt + Enter run the current cell, insert below
- Comment line is written preceding with #symbol
- A insert cell above
- B insert cell below
- D, D (press the key twice) delete selected cells
- Y change the cell type to *Code*
- M change the cell type to *Markdown*

## ***Fundamentals of python***

---

- Loading a simple delimited data file
- Counting how many rows and columns are loaded
- Determining which type of data is loaded
- Looking at different parts of the data by subsetting rows and columns
  
- [Data Source: https://github.com/jennybc/gapminder/blob/master/data/gapminder.rdata](https://github.com/jennybc/gapminder/blob/master/data/gapminder.rdata)
- [www.github.com/jennybc/gapminder](https://www.github.com/jennybc/gapminder)
- pip install pyreadr
- import pandas as pd
- import numpy as np
- import matplotlib.pyplot as plt
- import pyreadr
- df=pyreadr.read\_r('E:/IIITDM/Courses/Data\_Analytics/Data/gapminder.rdata')
- df1=pd.read\_csv(" E:/IIITDM/Courses/Data\_Analytics/Data/gapminder.csv ")

## Fundamentals of python

---

```
write.csv(gapminder,"E:/IIITDM/Courses/Data_Analytics/Data/gapminder.csv", row.names = FALSE)
```

```
df1=pd.read_csv("E:/IIITDM/Courses/Data_Analytics/Data/gapminder.csv")
```

```
In [10]: df1=pd.read_csv("E:/IIITDM/Courses/Data_Analytics/Data/gapminder.csv")
In [11]: df1
Out[11]:
```

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314
1	Afghanistan	Asia	1957	30.332	9240934	820.853030
2	Afghanistan	Asia	1962	31.997	10267083	853.100710
3	Afghanistan	Asia	1967	34.020	11537966	836.197138
4	Afghanistan	Asia	1972	36.088	13079460	739.981106
...	...	...	...	...	...	...
1699	Zimbabwe	Africa	1987	62.351	9216418	706.157306
1700	Zimbabwe	Africa	1992	60.377	10704340	693.420786
1701	Zimbabwe	Africa	1997	46.809	11404948	792.449960
1702	Zimbabwe	Africa	2002	39.989	11926563	672.038623
1703	Zimbabwe	Africa	2007	43.487	12311143	469.709298

```
[1704 rows x 6 columns]
```

## Fundamentals of python

---

- To know the header of the table
- `print(df1.head())`
- Head function gives first five rows of the table

```
In [21]: print(df1.head())
```

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314
1	Afghanistan	Asia	1957	30.332	9240934	820.853030
2	Afghanistan	Asia	1962	31.997	10267083	853.100710
3	Afghanistan	Asia	1967	34.020	11537966	836.197138
4	Afghanistan	Asia	1972	36.088	13079460	739.981106

- To show the size of the Table (Number of rows and columns)
- `print(df1.shape)`

```
In [22]: print(df1.shape)
```

```
(1704, 6)
```



## Fundamentals of python

---

- TO get the column Names

`print(df1.columns)`

```
In [24]: print(df1.columns)
```

```
Index(['country', 'continent', 'year', 'lifeExp', 'pop', 'gdpPercap'], dtype='object')
```

To get the data types of each column

```
In [25]: print (df1.dtypes)
```

```
country      object
continent    object
year          int64
lifeExp      float64
pop           int64
gdpPercap    float64
dtype: object
```

## Fundamentals of python

---

### Panda types Vs Python Types

Pandas Type	Python Type	Description
object	string	Most common data type
int64	int	Whole numbers
float64	float	Numbers with decimals
datetime64	datetime	datetime is found in the Python standard library (i.e., it is not loaded by default and needs to be imported)

Source: [Pandas for Everyone](#) , Daniel Y. Chen,

## Fundamentals of python

---

To get more information about the data

```
In [27]: print(df1.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1704 entries, 0 to 1703
Data columns (total 6 columns):
#   Column      Non-Null Count  Dtype
---  -
0   country     1704 non-null   object
1   continent   1704 non-null   object
2   year        1704 non-null   int64
3   lifeExp     1704 non-null   float64
4   pop         1704 non-null   int64
5   gdpPercap   1704 non-null   float64
dtypes: float64(2), int64(2), object(2)
memory usage: 80.0+ KB
None
```

## Looking at Columns, Rows, and Cells

---

- **Subsetting Columns**

- If we want to examine multiple columns, we can specify them by names, positions, or ranges.

### Subsetting Columns by Name

just get the country column and save it  
to its own variable

```
country_df = df['country']
```

show the first 5 observations

```
print(country_df.head())
```

```
In [28]: country_df = df1['country']
```

```
In [29]: country_df
```

```
Out[29]: 0      Afghanistan
1      Afghanistan
2      Afghanistan
3      Afghanistan
4      Afghanistan
...
1699   Zimbabwe
1700   Zimbabwe
1701   Zimbabwe
1702   Zimbabwe
1703   Zimbabwe
Name: country, Length: 1704, dtype: object
```

```
In [30]: country_df.head(5)
```

```
Out[30]: 0      Afghanistan
1      Afghanistan
2      Afghanistan
3      Afghanistan
4      Afghanistan
Name: country, dtype: object
```

## Looking at Columns, Rows, and Cells

---

- *show the last 5 observations*

```
In [31]: print(country_df.tail())
```

1699	Zimbabwe
1700	Zimbabwe
1701	Zimbabwe
1702	Zimbabwe
1703	Zimbabwe

Name: country, dtype: object

- *Looking at country, continent, and year*
- `subset = df[['country', 'continent', 'year']]`  
`print(subset.head())`

```
In [32]: subset = df1[['country', 'continent', 'year']]
```

```
In [33]: print(subset.head(5))
```

	country	continent	year
0	Afghanistan	Asia	1952
1	Afghanistan	Asia	1957
2	Afghanistan	Asia	1962
3	Afghanistan	Asia	1967
4	Afghanistan	Asia	1972

`print(subset.tail())`

```
In [34]: print(subset.tail(5))
```

	country	continent	year
1699	Zimbabwe	Africa	1987
1700	Zimbabwe	Africa	1992
1701	Zimbabwe	Africa	1997
1702	Zimbabwe	Africa	2002
1703	Zimbabwe	Africa	2007

## Subsetting Rows

---

- Rows can be subset in multiple ways, by row name or row index

Subset method	Description
<code>loc</code>	Subset based on index label (row name)
<code>iloc</code>	Subset based on row index (row number)

### Subset Rows by Index Label: `loc`

```
In [36]: print(df1.head())
```

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314
1	Afghanistan	Asia	1957	30.332	9240934	820.853030
2	Afghanistan	Asia	1962	31.997	10267083	853.100710
3	Afghanistan	Asia	1967	34.020	11537966	836.197138
4	Afghanistan	Asia	1972	36.088	13079460	739.981106

### Python counts from 0<sup>th</sup> Row

```
In [38]: print(df1.loc[0])
```

country	Afghanistan
continent	Asia
year	1952
lifeExp	28.801
pop	8425333
gdpPercap	779.445
Name: 0, dtype: object	

## Subsetting Rows

---

- To get the 100<sup>th</sup> row
- `print(df1.loc[99])`

```
In [39]: print(df1.loc[99])
```

```
country    Bangladesh
continent   Asia
year        1967
lifeExp     43.453
pop         62821884
gdpPercap   721.186
Name: 99, dtype: object
```

## Subsetting Rows

---

- *# get the last row*  
*# this will cause an error*
- **print(df.loc[-1])**

```
In [40]: print(df1.loc[-1])
```

```
-----  
ValueError                                Traceback (most recent call last)  
~\anaconda3\lib\site-packages\pandas\core\indexes\range.py in get_loc(self, key, method, tolerance)  
    349         try:  
--> 350             return self._range.index(new_key)  
    351         except ValueError:  
  
ValueError: -1 is not in range  
  
During handling of the above exception, another exception occurred:  
  
KeyError                                Traceback (most recent call last)  
<ipython-input-40-450ed06f07d6> in <module>  
----> 1 print(df1.loc[-1])  
  
~\anaconda3\lib\site-packages\pandas\core\indexing.py in __getitem__(self, key)  
    1765  
    1766         maybe_callable = com.apply_if_callable(key, self.obj)  
-> 1767         return self._getitem_axis(maybe_callable, axis=axis)  
    1768  
    1769     def _is_scalar_access(self, key: Tuple):
```



## Subsetting Rows

---

- *# get the last row (correctly)*  
*# use the first value given from shape to get the number of rows*

```
In [41]: number_of_rows = df1.shape[0]
```

```
In [42]: number_of_rows
```

```
Out[42]: 1704
```

```
In [43]: last_row_index = number_of_rows - 1
```

- *# now do the subset using the index of the last row*  
**print(df1.loc[last\_row\_index])**

```
In [45]: print(df1.loc[last_row_index])
```

```
country    Zimbabwe
continent   Africa
year        2007
lifeExp     43.487
pop        12311143
gdpPercap   469.709
Name: 1703, dtype: object
```

```
print(df1.tail(n=1))
```

```
In [47]: print(df1.tail(n=1))
```

	country	continent	year	lifeExp	pop	gdpPercap
1703	Zimbabwe	Africa	2007	43.487	12311143	469.709298

## Subsetting Rows

---

Notice that when we used `tail()` and `loc`, the results were printed out differently.

```
In [48]: subset_loc=df1.loc[0]
subset_head=df1.head(n=1)
```

```
In [49]: subset_loc
```

```
Out[49]: country      Afghanistan
continent      Asia
year          1952
lifeExp       28.801
pop           8425333
gdpPercap     779.445
Name: 0, dtype: object
```

```
In [50]: subset_head
```

```
Out[50]:
```

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314

## Subsetting Multiple Rows

---

*# select the first, 100th, and 1000th rows*

*# note the double square brackets similar to the syntax used to subset multiple columns*

```
print(df1.loc[[0,99,999]])
```

```
In [51]: print(df1.loc[[0,99,999]])
```

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314
99	Bangladesh	Asia	1967	43.453	62821884	721.186086
999	Mongolia	Asia	1967	51.253	1149500	1226.041130

## Subset Rows by Row Number: *iloc*

---

- # get the 2nd row `print(df.iloc[0])`

```
In [52]: print(df1.iloc[0])
```

```
country    Afghanistan
continent   Asia
year        1952
lifeExp     28.801
pop         8425333
gdpPercap   779.445
Name: 0, dtype: object
```

- ## get the 100th row `print(df.iloc[99])`

```
In [53]: print(df1.iloc[99])
```

```
country    Bangladesh
continent   Asia
year        1967
lifeExp     43.453
pop         62821884
gdpPercap   721.186
Name: 99, dtype: object
```

## Subset Rows by Row Number: *iloc*

---

- With *iloc*, we can pass in the -1 to get the last row—something we couldn't do with *loc*.
- *# using -1 to get the last row*  
`print(df.iloc[-1])`

```
In [54]: print(df1.iloc[-1])
```

country	Zimbabwe
continent	Africa
year	2007
lifeExp	43.487
pop	12311143
gdpPercap	469.709
Name: 1703, dtype: object	

---

## Subset Rows by Row Number: *iloc*

---

- *## get the first, 100th, and 1000th rows*

```
print(df.iloc[[0, 99, 999]])
```

```
In [55]: print(df1.iloc[[0, 99, 999]])
```

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314
99	Bangladesh	Asia	1967	43.453	62821884	721.186086
999	Mongolia	Asia	1967	51.253	1149500	1226.041130

## Mixing It Up

---

- The loc and iloc attributes can be used to obtain subsets of columns, rows, or both.
- The general syntax for loc and iloc uses square brackets with a comma. The part to the left of the comma is the row values to subset;
- the part to the right of the comma is the column values to subset.
- `df1.loc[[rows], [columns]]` or `df1.iloc[[rows], [columns]]`
- `df1.loc[[0], ['year']]`

```
In [58]: df1.loc[[0], ['year']]
```

```
Out[58]:
```

	year
0	1952

## Subsetting Columns

---

- must use Python's slicing syntax
- The Python slicing syntax uses a colon, :
- colon, the attribute refers to everything
- to subset the column(s). `df.loc[:, [columns]]`

- *# subset columns with loc*  
*# note the position of the colon*  
*# it is used to select all rows*

- `subset = df1.loc[:, ['year', 'pop']]`
- `print(subset.head())`

```
In [60]: subset = df1.loc[:, ['year', 'pop']]  
print(subset.head())
```

	year	pop
0	1952	8425333
1	1957	9240934
2	1962	10267083
3	1967	11537966
4	1972	13079460



## Subsetting Columns

---

- *# subset columns with iloc*  
*# iloc will allow us to use integers*  
*# -1 will select the last column*

```
subset = df.iloc[:, [2, 4, -1]]  
print(subset.head())
```

```
In [61]: subset = df1.iloc[:, [2, 4, -1]]  
print(subset.head())
```

	year	pop	gdpPercap
0	1952	8425333	779.445314
1	1957	9240934	820.853030
2	1962	10267083	853.100710
3	1967	11537966	836.197138
4	1972	13079460	739.981106

## Subsetting Columns

---

- *# subset columns with loc*  
*# but pass in integer values*  
*# this will cause an error*

```
subset = df1.loc[:, [2, 4, -1]]  
print(subset.head())
```

```
Traceback (most recent call last):  
  File "<ipython-input-1-719bcb04e3c1>", line  
2, in <module>  
    subset = df.loc[:, [2, 4, -1]]  
KeyError: 'None of [[2, 4, -1]] are in the  
[columns]'
```

- *# subset columns with iloc*  
*# but pass in index names*  
*# this will cause an error*

```
subset = df1.iloc[:, ['year', 'pop']]  
print(subset.head())
```

```
Traceback (most recent call last):  
  File "<ipython-input-1-43f52fceb49>", line  
2, in <module>  
    subset = df.iloc[:, ['year', 'pop']]  
TypeError: cannot perform reduce with flexible  
type
```

## Subsetting Columns by Range

---

- built-in range function to create a range of values in Python
- range(5) is called, five integers are returned: 0 – 4.

- *# create a range of integers from 0 to 4 inclusive*

- `small_range = list(range(5))`  
`print(small_range)`

```
In [64]: small_range = list(range(5))  
print(small_range)
```

```
[0, 1, 2, 3, 4]
```

- *# subset the dataframe with the range*

```
subset = df.iloc[:, small_range]  
print(subset.head())
```

```
In [65]: subset = df1.iloc[:, small_range]  
print(subset.head())
```

	country	continent	year	lifeExp	pop
0	Afghanistan	Asia	1952	28.801	8425333
1	Afghanistan	Asia	1957	30.332	9240934
2	Afghanistan	Asia	1962	31.997	10267083
3	Afghanistan	Asia	1967	34.020	11537966
4	Afghanistan	Asia	1972	36.088	13079460

## Subsetting Columns by Range

---

*# create a range from 3 to 5 inclusive*

```
small_range = list(range(3, 6))  
print(small_range)
```

the values are specified in a way  
such that the range is inclusive  
on the left, and exclusive on the right

```
In [66]: small_range = list(range(3, 6))  
print(small_range)  
  
[3, 4, 5]
```

```
subset = df1.iloc[:, small_range]  
print(subset.head())
```

```
In [67]: subset = df1.iloc[:, small_range]  
print(subset.head())
```

	lifeExp	pop	gdpPercap
0	28.801	8425333	779.445314
1	30.332	9240934	820.853030
2	31.997	10267083	853.100710
3	34.020	11537966	836.197138
4	36.088	13079460	739.981106

## Subsetting Columns by Range

---

```
# create a range from 0 to 5 inclusive, every other integer
small_range = list(range(0, 6, 2))
subset = df.iloc[:, small_range]
print(subset.head())
```

```
In [67]: subset = df1.iloc[:, small_range]
         print(subset.head())
```

	lifeExp	pop	gdpPercap
0	28.801	8425333	779.445314
1	30.332	9240934	820.853030
2	31.997	10267083	853.100710
3	34.020	11537966	836.197138
4	36.088	13079460	739.981106

## Subsetting Rows and Columns

---

- *# using loc*

```
print(df.loc[42, 'country'])
```

- Angola

- *# using iloc*

```
print(df.iloc[42, 0])
```

- Angola

- *# will cause an error*

```
print(df.loc[42, 0])
```

```
In [72]: print(df1.loc[42, 'country'])
```

Angola

```
In [73]: print(df1.iloc[42, 0])
```

Angola

Traceback (most recent call last):

File "<ipython-input-1-2b69d7150b5e>", line 2, in <module>

```
print(df.loc[42, 0])
```

TypeError: cannot do label indexing on <class 'pandas.core.indexes.base.Index'> with these indexers [0] of <class

## Subsetting Multiple Rows and Columns

---

*# get the 1st, 100th, and 1000th rows  
# from the 1st, 4th, and 6th columns  
# the columns we are hoping to get are  
# country, lifeExp, and gdpPercap*

```
print(df.iloc[[0, 99, 999], [0, 3, 5]])
```

```
In [74]: print(df1.iloc[[0, 99, 999], [0, 3, 5]])
```

	country	lifeExp	gdpPercap
0	Afghanistan	28.801	779.445314
99	Bangladesh	43.453	721.186086
999	Mongolia	51.253	1226.041130

*# if we use the column names directly,  
# it makes the code a bit easier to read  
# note now we have to use loc, instead of iloc*

```
print(df.loc[[0, 99, 999], ['country', 'lifeExp', 'gdpPercap']])
```

```
print(df1.loc[[0, 99, 999], ['country', 'lifeExp', 'gdpPercap']])
```

	country	lifeExp	gdpPercap
0	Afghanistan	28.801	779.445314
99	Bangladesh	43.453	721.186086
999	Mongolia	51.253	1226.041130

using absolute indexes can lead to problems if the column order gets changed for some reason

## Subsetting Multiple Rows and Columns

---

- `print(df.loc[10:13, ['country', 'lifeExp', 'gdpPercap']])`

```
In [76]: print(df1.loc[10:13, ['country', 'lifeExp', 'gdpPercap']])
```

	country	lifeExp	gdpPercap
10	Afghanistan	42.129	726.734055
11	Afghanistan	43.828	974.580338
12	Albania	55.230	1601.056136
13	Albania	59.280	1942.284244



## Grouped and Aggregated Calculations

---

- `print(df1.head(n=10))`

In [77]: `print(df1.head(n=10))`

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314
1	Afghanistan	Asia	1957	30.332	9240934	820.853030
2	Afghanistan	Asia	1962	31.997	10267083	853.100710
3	Afghanistan	Asia	1967	34.020	11537966	836.197138
4	Afghanistan	Asia	1972	36.088	13079460	739.981106
5	Afghanistan	Asia	1977	38.438	14880372	786.113360
6	Afghanistan	Asia	1982	39.854	12881816	978.011439
7	Afghanistan	Asia	1987	40.822	13867957	852.395945
8	Afghanistan	Asia	1992	41.674	16317921	649.341395
9	Afghanistan	Asia	1997	41.763	22227415	635.341351

- 1. For each year in our data, what was the average life expectancy? What is the average life expectancy, population, and GDP?
- 2. What if we stratify the data by continent and perform the same calculations?
- 3. How many countries are listed in each continent?

**To answer the questions just posed, we need to perform a grouped (i.e., aggregate) calculation.**

## Grouped Means

---

*# For each year in our data, what was the average life expectancy?*

*# To answer this question,*

*# we need to split our data into parts by year*

*# then we get the 'lifeExp' column and calculate the mean*

**print(df.groupby('year')['lifeExp'].mean())**

```
In [78]: print(df1.groupby('year')['lifeExp'].mean())
```

```
year
1952    49.057620
1957    51.507401
1962    53.609249
1967    55.678290
1972    57.647386
1977    59.570157
1982    61.533197
1987    63.212613
1992    64.160338
1997    65.014676
2002    65.694923
2007    67.007423
Name: lifeExp, dtype: float64
```

## Grouped Means

---

- *# the backslash allows us to break up 1 long line of Python code  
# into multiple lines*  
*# df.groupby(['year', 'continent'])[['lifeExp', 'gdpPercap']].mean()*  
*# is the same as the following code*  

```
multi_group_var = df.\n    groupby(['year', 'continent'])\n    [['lifeExp', 'gdpPercap']].\n    mean()\nprint(multi_group_var)
```
- (OR)
- ```
multi_group_var = df1.groupby(['year', 'continent'])[['lifeExp', 'gdpPercap']].mean()
```
- ```
print(multi_group_var)
```

## Grouped Means

---

		lifeExp	gdpPercap
year	continent		
1952	Africa	39.135500	1252.572466
	Americas	53.279840	4079.062552
	Asia	46.314394	5195.484004
	Europe	64.408500	5661.057435
	Oceania	69.255000	10298.085650
1957	Africa	41.266346	1385.236062
	Americas	55.960280	4616.043733
	Asia	49.318544	5787.732940
	Europe	66.703067	6963.012816
	Oceania	70.295000	11598.522455
1962	Africa	43.319442	1598.078825
	Americas	58.398760	4901.541870
	Asia	51.563223	5729.369625
	Europe	68.539233	8365.486814
	Oceania	71.085000	12696.452430
1967	Africa	45.334538	2050.363801
	Americas	60.410920	5668.253496
	Asia	54.663640	5971.173374
	Europe	69.737600	10143.823757
	Oceania	71.310000	14495.021790

## Grouped Means

---

- If you need to “flatten” the dataframe, you can use the `reset_index` method.
- `flat = multi_group_var.reset_index()`  
`print(flat.head(15))`

```
In [85]: flat = multi_group_var.reset_index()
         print(flat.head(15))
```

	year	continent	lifeExp	gdpPercap
0	1952	Africa	39.135500	1252.572466
1	1952	Americas	53.279840	4079.062552
2	1952	Asia	46.314394	5195.484004
3	1952	Europe	64.408500	5661.057435
4	1952	Oceania	69.255000	10298.085650
5	1957	Africa	41.266346	1385.236062
6	1957	Americas	55.960280	4616.043733
7	1957	Asia	49.318544	5787.732940
8	1957	Europe	66.703067	6963.012816
9	1957	Oceania	70.295000	11598.522455
10	1962	Africa	43.319442	1598.078825
11	1962	Americas	58.398760	4901.541870
12	1962	Asia	51.563223	5729.369625
13	1962	Europe	68.539233	8365.486814
14	1962	Oceania	71.085000	12696.452430

## Grouped Frequency Counts

---

- use the `nunique` and `value_counts` methods, respectively, to get counts of unique values and frequency counts on a Pandas Series.
- `print(df1.groupby('continent')['country'].nunique())`

```
print(df1.groupby('continent')['country'].nunique())
```

```
continent
Africa      52
Americas    25
Asia        33
Europe      30
Oceania      2
Name: country, dtype: int64
```

- `print(df1.groupby('continent')['country'].value_counts())`

```
In [88]: print(df1.groupby('continent')['country'].value_counts())
```

```
continent country
Africa  Algeria      12
        Angola      12
        Benin       12
        Botswana    12
        Burkina Faso 12
        ..
Europe  Switzerland  12
        Turkey       12
        United Kingdom 12
Oceania Australia    12
        New Zealand   12
Name: country, Length: 142, dtype: int64
```

## Basic Plot

- Visualizations are extremely important in almost every step of the data process. They help us identify trends in data when we are trying to understand and clean the data, and they help us convey our final findings.
- `global_yearly_life_expectancy = df.groupby('year')['lifeExp'].mean()`
- `print(global_yearly_life_expectancy)`
- `global_yearly_life_expectancy.plot()`

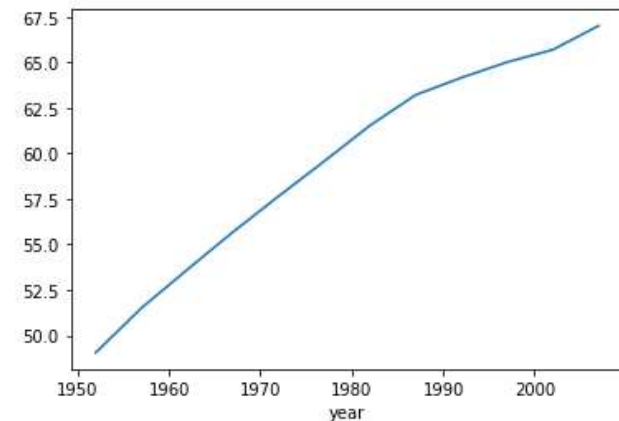
```
In [89]: global_yearly_life_expectancy = df1.groupby('year')['lifeExp'].mean()
```

```
In [90]: print(global_yearly_life_expectancy)
```

```
year
1952    49.057620
1957    51.507401
1962    53.609249
1967    55.678290
1972    57.647386
1977    59.570157
1982    61.533197
1987    63.212613
1992    64.160338
1997    65.014676
2002    65.694923
2007    67.007423
Name: lifeExp, dtype: float64
```

```
In [92]: global_yearly_life_expectancy.plot()
```

```
Out[92]: <matplotlib.axes._subplots.AxesSubplot at 0x215cfde14c8>
```





# நன்றி