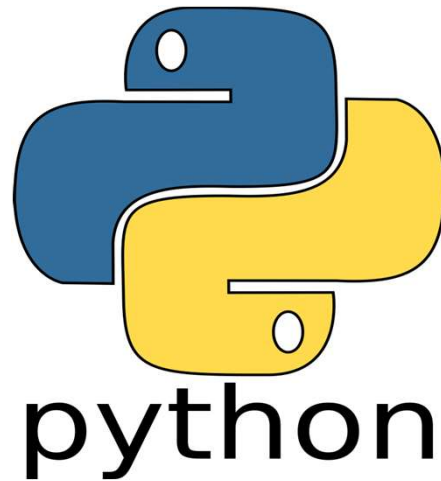


BIG DATA TOOLS FOR MANAGERS

Unit-4 & 5 : Introduction to Python Pandas



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Comparison with Python Data structure

Lists	Tuple	Set	Dictionary
Lists are mutable	Tuples are <u>immutable</u> .	Sets are mutable	Dictionary are mutable
Lists are enclosed within square braces. []	Tuples are enclosed within parenthesis. ()	Sets are enclosed in curly brackets. { }	Dictionaries are enclosed in curly brackets with key-value pairs. { key : value }
List element can be accessed using index/ range of index	Tuple element can be accessed using index/range of index	Have use iteration like for, while loop to access element	Dictionary element can be accessible using its key
len() function used to get length/size of list	len() function used to get length/size of Tuple	len() function used to get length/size of set	len() function used to get length/size of dictionary
Easy to combine two or more lists with + (plus) operator	Easy to combine two or more Tuple with + (plus) operator	union to be used for combining two set	update function used to combine two dictionary

Exercise

Create a List with given element [10,20,30,40,50,60,70,80,90,100]

Write a python code for :

1. Create a List
2. Print element using print()
3. Print element using iteration (For loop)
4. Multiply list elements with number 2
5. Display first element of list
6. Display last element of list
7. Display first 3 elements of list
8. Display last 3 elements of list

Exercise

Create a dictionary for Employee data

employee_Name : John

employee_City : Bangalore

employee_Mobile: 9876512345

employee_Email : john@gmail.com

Write a python code for :

1. Create a employee dictionary
2. Display all the key present in dictionary
3. Display all the value present in dictionary
4. Print Dictionary element
5. Access dictionary element using employee_Name
6. Access dictionary element using employee_Email

About Pandas

- Pandas is an open-source Python library providing high-performance data manipulation and analysis tool using its powerful data structure.
- The name panda is derived from the word **Panel Data**.
- In 2008 panda library introduced for high performance, flexible tool for analysis of data.
- Python with Pandas is used in a wide range of fields including academic and commercial domains including finance, economics, Statistics, analytics, etc.

Key Features of Pandas

- Fast and efficient DataFrame object with default and customized indexing.
- Tools for loading data into in-memory data objects from different file formats.
- Data alignment and integrated handling of missing data.
- Label-based slicing, indexing and sub-setting of large data sets.
- Columns from a data structure can be deleted or inserted.
- Group by data for aggregation and transformations.
- High performance merging and joining of data.
- Time Series functionality.

Where to get pandas library?



Where to get pandas library?

- Download pandas library from an internet using pip package manager

pip install pandas

how to import pandas library?

import statement used to import the pandas or any other libraries in Python.

```
import pandas
```

or

```
import pandas as pd
```

or

```
import pandas as xyz
```

- **as** keyword helps to create **alias name** of the package.
- This helpful when package name is too large.

Data Frames in R

- Creation of data frame

```
my_data <- data.frame(  
  name   = c("A","B","C","D"),  
  age    = c(40,45,70,60),  
  gender = c("F","M","F","M")  
)
```

name, age, gender are the vectors

Output:

Name	Age	gender
A	40	F
B	45	M
C	70	F
D	60	M

Data Frames in Python

- Creation of data frame in Python

```
import pandas as pd
```

```
my_data = pd.DataFrame(  
    {  
        "name"      : ["A","B","C","D"],  
        "age"       : [40,45,70,60],  
        "gender"    : ["F","M","F","M"]  
    }  
)
```

Output:

	name	age	gender
0	A	40	F
1	B	45	M
2	C	70	F
3	D	60	M

name, age, gender are the dictionary key

Import files using Pandas

- Pandas library has provided different methods for loading datasets with many different formats onto DataFrame.
- *** Once data loaded into memory pandas creates Pandas DataFrame automatically for easy to manipulate data.
- File format pandas support:
csv, json, excel, table, fwf(fixed with format)
- Read function looks like:
read_csv()
read_excel()
read_table()

Import csv file using Pandas

- Pandas has read_csv functions to read the csv file.

Example:

Download dataset and save in C:/dataset location

<https://drive.google.com/file/d/1EPQhI0wVCZnNP1vx7BE43phOUljzZXGD/view?usp=sharing>

Import csv file using Pandas

- Pandas has read_csv functions to read the csv file.

Example:

```
import pandas as pd
```

```
pd.read_csv("C:/dataset/VEHICLE_PARK.csv")
```

or

```
import pandas
```

```
pandas.read_csv("C:/dataset/VEHICLE_PARK.csv")
```

Import csv file using Pandas

- Pandas has **read_csv()** functions to read the csv file.
- After importing csv file, store the data in any variable.

Example:

```
import pandas as pd  
data = pd.read_csv("C:/dataset/VEHICLE_PARK.csv")
```

- Now, data variable having entire csv file data in form of rows & cols.

Python Pandas

print() function to display the data frame values

Example:

print(data)

Output:

```
print(data)
```

	YEAR	VEHICLE_TYPE	BRAND	VEHICLE_COUNT	AGE_GROUP	AGE	\
0	2000	TRUCK	SCANIA	3208	0-1	0	
1	2000	TRUCK	MAN	7486	0-1	0	
2	2000	TRUCK	Tata	8021	0-1	0	
3	2000	TRUCK	Ashok Leyland	2673	0-1	0	
4	2000	TRUCK	VOLVO	1069	0-1	0	
...
22545	2022	OTHERS	HITACHI	4277	22-23	22	
22546	2022	OTHERS	KOMASTU	5882	22-23	22	
22547	2022	OTHERS	XCBG	2673	22-23	22	
22548	2022	OTHERS	CATERPILLAR	534	22-23	22	
22549	2022	OTHERS	VOLVO	4812	22-23	22	

	RTO_REGISTRATION_YEAR
0	2000
1	2000
2	2000
3	2000
4	2000
...	...
22545	2000
22546	2000
22547	2000
22548	2000
22549	2000

[22550 rows x 7 columns]

Python Pandas

info() Function to get the structure of Pandas DataFrame

Output:

Example:

data.info()

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 22550 entries, 0 to 22549
Data columns (total 7 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   YEAR                                22550 non-null  int64
1   VEHICLE_TYPE                        22550 non-null  object
2   BRAND                              22550 non-null  object
3   VEHICLE_COUNT                       22550 non-null  int64
4   AGE_GROUP                           22550 non-null  object
5   AGE                                22550 non-null  int64
6   RTO_REGISTRATION_YEAR               22550 non-null  int64
dtypes: int64(4), object(3)
memory usage: 1.2+ MB
```

Python Pandas

shape is properties to get the no of rows & columns of DataFrame

Output:

Example:

data.shape

```
data.shape
: (22550, 7)
```

Python Pandas

head(n) & **tail(n)** are the functions to display top and bottom rows from the pandas DataFrame.

Output:

Example:

data.head(10)

data.tail(10)

```
data.head(10)
```

	YEAR	VEHICLE_TYPE	BRAND	VEHICLE_COUNT	AGE_GROUP	AGE	RTO_REGISTRATION_YEAR
0	2000	TRUCK	SCANIA	3208	0-1	0	2000
1	2000	TRUCK	MAN	7486	0-1	0	2000
2	2000	TRUCK	Tata	8021	0-1	0	2000
3	2000	TRUCK	Ashok Leyland	2673	0-1	0	2000
4	2000	TRUCK	VOLVO	1069	0-1	0	2000
5	2000	TRUCK	MAZ	4277	0-1	0	2000
6	2000	TRUCK	Asia MotorWorks (AMW)	5882	0-1	0	2000
7	2000	TRUCK	PACCAR	2673	0-1	0	2000
8	2000	TRUCK	Force Motors	534	0-1	0	2000
9	2000	TRUCK	BharatBenz	4812	0-1	0	2000

Python Pandas

Column selection

- `['col-name']` – for selecting single column
- `[['col-name-1','col-name-2',.....'col-name-n']]` - for selecting multiple columns

Example:

`data['YEAR']` #One column

`data[['YEAR', 'AGE_GROUP']]` # Multiple columns

Python Pandas

value_counts(normalize=False) function Gives frequency of each unique value in a column

normalize=True will calculate the percentage of total frequency.

Example:

`data['VEHICLE_TYPE'].value_counts()`

```
data['VEHICLE_TYPE'].value_counts()
: TRUCK          5412
  BUSE           5412
  FOUR WHEELER   4510
  OTHERS         4510
  TWO WHEELER    2706
Name: VEHICLE_TYPE, dtype: int64
```

Python Pandas

value_counts(normalize=False) function Gives frequency of each unique value in a column

normalize=True will calculate the percentage of total frequency.

Example:

`data['VEHICLE_TYPE'].value_counts(normalize=True)`

```
data['VEHICLE_TYPE'].value_counts(normalize=True)
```

```
TRUCK      0.24  
BUSE       0.24  
FOUR WHEELER  0.20  
OTHERS     0.20  
TWO WHEELER  0.12  
Name: VEHICLE_TYPE, dtype: float64
```

Python Pandas

crosstab() function Gives frequency of two columns.

Example:

`pd.crosstab(data['VEHICLE_TYPE'], data['AGE_GROUP'])`

```
pd.crosstab(data['VEHICLE_TYPE'], data['AGE_GROUP'])
```

AGE_GROUP	0-1	1-2	10-11	11-12	12-13	13-14	14-15	15-16	16-17	17-18	...	20-21	21-22	22-23	3-4	4-5	5-6	6-7	7-8	8-9	9-10
VEHICLE_TYPE																					
BUSE	276	276	276	264	252	240	228	216	204	192	...	156	144	132	276	276	276	276	276	276	276
FOUR WHEELER	230	230	230	220	210	200	190	180	170	160	...	130	120	110	230	230	230	230	230	230	230
OTHERS	230	230	230	220	210	200	190	180	170	160	...	130	120	110	230	230	230	230	230	230	230
TRUCK	276	276	276	264	252	240	228	216	204	192	...	156	144	132	276	276	276	276	276	276	276
TWO WHEELER	138	138	138	132	126	120	114	108	102	96	...	78	72	66	138	138	138	138	138	138	138

5 rows × 23 columns

Python Pandas

crosstab() function Gives frequency of two columns.

Example:

`pd.crosstab(data['VEHICLE_TYPE'], data['AGE_GROUP'])`

```
pd.crosstab(data['VEHICLE_TYPE'], data['AGE_GROUP'])
```

AGE_GROUP	0-1	1-2	10-11	11-12	12-13	13-14	14-15	15-16	16-17	17-18	...	20-21	21-22	22-23	3-4	4-5	5-6	6-7	7-8	8-9	9-10
VEHICLE_TYPE																					
BUSE	276	276	276	264	252	240	228	216	204	192	...	156	144	132	276	276	276	276	276	276	276
FOUR WHEELER	230	230	230	220	210	200	190	180	170	160	...	130	120	110	230	230	230	230	230	230	230
OTHERS	230	230	230	220	210	200	190	180	170	160	...	130	120	110	230	230	230	230	230	230	230
TRUCK	276	276	276	264	252	240	228	216	204	192	...	156	144	132	276	276	276	276	276	276	276
TWO WHEELER	138	138	138	132	126	120	114	108	102	96	...	78	72	66	138	138	138	138	138	138	138

5 rows × 23 columns

Python Pandas

sort_values(col-name, ascending=False) function sort the DataFrame based on the specified columns.

By default, sorting is done in ascending order.

Example (Single Column):

```
data.sort_values(["BRAND"], ascending=[False])
```

or

```
data.sort_values("BRAND", ascending=False)
```

Python Pandas

sort_values(col-name, ascending=False) function sort the DataFrame based on the specified columns.

By default, sorting is done in ascending order.

Example (Multi Columns):

```
data.sort_values(  
    ['BRAND', 'VEHICLE_COUNT'],  
    ascending=[True, False]  
)
```

Python Pandas

describe() function to display quick summary of the DataFrame. By default, it gives summary for numeric data only.

Example

data.describe()

```
data.describe()
```

	YEAR	VEHICLE_COUNT	AGE	RTO_REGISTRATION_YEAR
count	22550.000000	22550.000000	22550.000000	22550.000000
mean	2012.268293	27447.555344	9.731707	2002.536585
std	6.262782	62575.624461	6.262782	7.622657
min	2000.000000	129.000000	0.000000	1990.000000
25%	2007.000000	2422.000000	4.000000	1997.000000
50%	2013.000000	6415.000000	9.000000	2002.000000
75%	2018.000000	18363.000000	15.000000	2008.000000
max	2022.000000	605882.000000	22.000000	2022.000000

Python Pandas

describe() function to display quick summary of the DataFrame. By default, it gives summary for numeric data only.

describe(include='all') function with include='all' gives summary for all the columns available in DataFrame

Example

data.describe(include='all')

```
data.describe(include='all')
```

	YEAR	VEHICLE_TYPE	BRAND	VEHICLE_COUNT	AGE_GROUP	AGE	RTO_REGISTRATION_YEAR
count	22550.000000	22550	22550	22550.000000	22550	22550.000000	22550.000000
unique	NaN	5	38	NaN	23	NaN	NaN
top	NaN	TRUCK	VOLVO	NaN	0-1	NaN	NaN
freq	NaN	5412	1804	NaN	1150	NaN	NaN
mean	2012.268293	NaN	NaN	27447.555344	NaN	9.731707	2002.536585
std	6.262782	NaN	NaN	62575.624461	NaN	6.262782	7.622657
min	2000.000000	NaN	NaN	129.000000	NaN	0.000000	1990.000000
25%	2007.000000	NaN	NaN	2422.000000	NaN	4.000000	1997.000000

Python Pandas

describe() function to display quick summary of the DataFrame. By default, it gives summary for numeric data only.

describe(include='all') function with include='all' gives summary for all the columns available in DataFrame

Example

data.describe(include='all')

```
data.describe(include='all')
```

	YEAR	VEHICLE_TYPE	BRAND	VEHICLE_COUNT	AGE_GROUP	AGE	RTO_REGISTRATION_YEAR
count	22550.000000	22550	22550	22550.000000	22550	22550.000000	22550.000000
unique	NaN	5	38	NaN	23	NaN	NaN
top	NaN	TRUCK	VOLVO	NaN	0-1	NaN	NaN
freq	NaN	5412	1804	NaN	1150	NaN	NaN
mean	2012.268293	NaN	NaN	27447.555344	NaN	9.731707	2002.536585
std	6.262782	NaN	NaN	62575.624461	NaN	6.262782	7.622657
min	2000.000000	NaN	NaN	129.000000	NaN	0.000000	1990.000000
25%	2007.000000	NaN	NaN	2422.000000	NaN	4.000000	1997.000000

Python Pandas

describe() function to display quick summary of the DataFrame. By default, it gives summary for numeric data only.

describe(include='all') function with include='all' gives summary for all the columns available in DataFrame

Example

data.describe(include='all')

```
data.describe(include='all')
```

	YEAR	VEHICLE_TYPE	BRAND	VEHICLE_COUNT	AGE_GROUP	AGE	RTO_REGISTRATION_YEAR
count	22550.000000	22550	22550	22550.000000	22550	22550.000000	22550.000000
unique	NaN	5	38	NaN	23	NaN	NaN
top	NaN	TRUCK	VOLVO	NaN	0-1	NaN	NaN
freq	NaN	5412	1804	NaN	1150	NaN	NaN
mean	2012.268293	NaN	NaN	27447.555344	NaN	9.731707	2002.536585
std	6.262782	NaN	NaN	62575.624461	NaN	6.262782	7.622657
min	2000.000000	NaN	NaN	129.000000	NaN	0.000000	1990.000000
25%	2007.000000	NaN	NaN	2422.000000	NaN	4.000000	1997.000000

Python Pandas

Aggregate functions like count, sum, min, max, mean...etc can be applied on DataFrame.

Example

#Single Column to Count number of elements present

```
data['YEAR'].count()
```

#Multiple Cols, Count number of elements present

```
data[['YEAR', 'AGE']].count()
```

#all the columns

```
data.count()
```

Python Pandas

Aggregate functions like count, sum, min, max, mean...etc can be applied on DataFrame.

Example

#Single Column, check Avg values for Age column

```
data['AGE'].mean()
```

#Multiple Cols, check Avg values for Age, Vehicle Count columns

```
data[['AGE', 'VEHICLE_COUNT']].mean()
```

#all the columns (categorical values get excluded)

```
data.mean()
```


Python Pandas

groupby() functions to group the records based on specified columns and then apply for aggregation (max, min, sum, mean...etc)

Example

#Single Column, get the total number of vehicles by VEHICLE_TYPE

data\

```
.groupby("VEHICLE_TYPE")\  
.aggregate({"VEHICLE_COUNT": "sum"})
```

```
data\  
.groupby("VEHICLE_TYPE")\  
.aggregate({"VEHICLE_COUNT": "sum"})
```

VEHICLE_COUNT	
VEHICLE_TYPE	
BUSE	13600671
FOUR WHEELER	90155336
OTHERS	39306204
TRUCK	40503771
TWO WHEELER	435376391

Python Pandas

groupby() functions to group the records based on specified columns and then apply for aggregation (max, min, sum, mean...etc)

Example

#Multi Column, get the total number of vehicles by VEHICLE_TYPE, BRAND

data\

```
.groupby(["VEHICLE_TYPE", "BRAND"])\n.aggregate({"VEHICLE_COUNT": "sum"})
```

```
data\  
.groupby(["VEHICLE_TYPE", "BRAND"])\n.aggregate({"VEHICLE_COUNT": "sum"})
```

		VEHICLE_COUNT
VEHICLE_TYPE	BRAND	
BUSE	Ashok Leyland	1085440
	Asia MotorWorks (AMW)	1010703
	BYD	1002228
	BharatBenz	1086966
	EICHER MOTOR	1140108
	FOTON	1475555
	ISUZU	978069

Python Pandas

merge() functions to join multiple DataFrame based on the column columns and nature of join.

Before doing join, first create two sample dataframe.

DataFrame : **Orders**

OrderID	CustomerID	OrderDate
10308	2	2022-08-15
10309	1	2022-08-26
10310	2	2022-09-01

DataFrame : **Customers**

CustomerID	CustomerName	Country
1	John Todd	Germany
2	Dominic Dom	Mexico
3	Paul S	Mexico

Python Pandas

Step-1 Create dictionary

```
orders_dictionary ={  
    "OrderID"      : [10308, 10309, 10310],  
    "CustomerID"   : [2,1,2],  
    "OrderDate"    : ["2022-08-15", "2022-08-26", "2022-09-01"]  
}
```

Python Pandas

Step-1 Create dictionary

```
orders_dictionary = {  
    "OrderID"      : [10308, 10309, 10310],  
    "CustomerID"   : [2,1,2],  
    "OrderDate"    : ["2022-08-15", "2022-08-26", "2022-09-01"]  
}
```

Step-2 Create Pandas DataFrame

```
Orders = pd.DataFrame(orders_dictionary)
```

Python Pandas

Step-1 Create dictionary

```
customer_dictionary = {  
    "CustomerID" : [1,2,3],  
    "CustomerName" : ["John Todd", "Dominic Dom", "Paul S"],  
    "Country" : ["Germany", "Maxico", "Maxico"]  
}
```

Step-2 Create Pandas DataFrame

```
Customers = pd.DataFrame(customer_dictionary)
```

Python Pandas

Step-3 Display the Orders & Customers DataFrame

`print(Customers)`

`print(Orders)`

```
| print(Customers)
```

	CustomerID	CustomerName	Country
0	1	John Todd	Germany
1	2	Dominic Dom	Maxico
2	3	Paul S	Maxico

```
| print(Orders)
```

	OrderID	CustomerID	OrderDate
0	10308	2	2022-08-15
1	10309	1	2022-08-26
2	10310	2	2022-09-01

Python Pandas

Step-4 Perform inner join

```
pd.merge(Orders, Customers, on=['CustomerID'], how="inner")
```

```
pd.merge(Orders, Customers, on=['CustomerID'], how="inner")
```

	OrderID	CustomerID	OrderDate	CustomerName	Country
0	10308	2	2022-08-15	Dominic Dom	Maxico
1	10310	2	2022-09-01	Dominic Dom	Maxico
2	10309	1	2022-08-26	John Todd	Germany

Python Pandas

Step-5 Perform left join

```
pd.merge(Orders, Customers, on=['CustomerID'], how="left")
```

```
| pd.merge(Orders, Customers, on=['CustomerID'], how="left")
```

	OrderID	CustomerID	OrderDate	CustomerName	Country
0	10308	2	2022-08-15	Dominic Dom	Maxico
1	10309	1	2022-08-26	John Todd	Germany
2	10310	2	2022-09-01	Dominic Dom	Maxico

Python Pandas

Step-6 Perform right join

`pd.merge(Orders, Customers, on=['CustomerID'], how="right")`

```
pd.merge(Orders, Customers, on=['CustomerID'], how="right")
```

	OrderID	CustomerID	OrderDate	CustomerName	Country
0	10309.0	1	2022-08-26	John Todd	Germany
1	10308.0	2	2022-08-15	Dominic Dom	Maxico
2	10310.0	2	2022-09-01	Dominic Dom	Maxico
3	NaN	3	NaN	Paul S	Maxico

Python Pandas

Step-7 Perform full join

`pd.merge(Orders, Customers, on=['CustomerID'], how="outer")`

```
pd.merge(Orders, Customers, on=['CustomerID'], how="outer")
```

	OrderID	CustomerID	OrderDate	CustomerName	Country
0	10308.0	2	2022-08-15	Dominic Dom	Maxico
1	10310.0	2	2022-09-01	Dominic Dom	Maxico
2	10309.0	1	2022-08-26	John Todd	Germany
3	NaN	3	NaN	Paul S	Maxico

Python Pandas

rename() functions to rename the columns in DataFrame

data-frame.rename({"col-name": "new-col-name"}, axis=1)

- axis=1 represent column axis
- axis=0 represent row axis

Example:

data.rename({"VEHICLE_COUNT": "SALES"}, axis=1)

```
data.rename({"VEHICLE_COUNT": "SALES"}, axis=1)|
```

	YEAR	VEHICLE_TYPE	BRAND	SALES	AGE_GROUP	AGE	RTO_REGISTRATION_YEAR
0	2000	TRUCK	SCANIA	3208	0-1	0	2000
1	2000	TRUCK	MAN	7486	0-1	0	2000
2	2000	TRUCK	Tata	8021	0-1	0	2000
3	2000	TRUCK	Ashok Leyland	2673	0-1	0	2000

Python Pandas

rename() functions to rename the columns in DataFrame.

Inplace additional parameters to specify with rename function, to rename column name permanently.

Syntax:

```
data-frame.rename({"col-name": "new-col-name"},  
                  axis=1,  
                  inplace=True)
```

Example:

```
data.rename({"VEHICLE_COUNT": "SALES"},  
            axis=1,  
            inplace=True)
```

Python Pandas

Example:

```
data.rename({"VEHICLE_COUNT": "SALES"},  
            axis=1,  
            inplace=True)
```

```
print(data)
```

```
print(data)
```

	YEAR	VEHICLE_TYPE	BRAND	SALES	AGE_GROUP	AGE	\
0	2000	TRUCK	SCANIA	3208	0-1	0	
1	2000	TRUCK	MAN	7486	0-1	0	
2	2000	TRUCK	Tata	8021	0-1	0	
3	2000	TRUCK	Ashok Leyland	2673	0-1	0	
4	2000	TRUCK	VOLVO	1069	0-1	0	
...
22545	2022	OTHERS	HITACHI	4277	22-23	22	
22546	2022	OTHERS	KOMASTU	5882	22-23	22	
22547	2022	OTHERS	XCBG	2673	22-23	22	
22548	2022	OTHERS	CATERPILLAR	534	22-23	22	
22549	2022	OTHERS	VOLVO	4812	22-23	22	

	RTO_REGISTRATION_YEAR
0	2000
1	2000
2	2000

Python Pandas

drop() functions to delete columns from DataFrame.

Inplace additional parameters to specify with drop function, to delete column name permanently.

Syntax:

```
data-frame.drop({"col-name": "new-col-name"},  
                axis=1,  
                inplace=True)
```

Example:

```
data.drop(["YEAR", "BRAND"], axis=1, inplace=True)
```

Python Pandas

Example:

```
data.drop(["YEAR", "BRAND"], axis=1, inplace=True)
```

```
print(data)
```

```
print(data)
```

	VEHICLE_TYPE	SALES	AGE_GROUP	AGE	RTO_REGISTRATION_YEAR
0	TRUCK	3208	0-1	0	2000
1	TRUCK	7486	0-1	0	2000
2	TRUCK	8021	0-1	0	2000
3	TRUCK	2673	0-1	0	2000
4	TRUCK	1069	0-1	0	2000
...
22545	OTHERS	4277	22-23	22	2000
22546	OTHERS	5882	22-23	22	2000
22547	OTHERS	2673	22-23	22	2000
22548	OTHERS	534	22-23	22	2000
22549	OTHERS	4812	22-23	22	2000

Python Pandas

Panda DataFrame allows to write conditions to filter out the data.

Syntax:

data-frame[conditions]

Example: display 10 years old vehicles from DataFrame

data[data['AGE']==10]

Python Pandas

Panda DataFrame allows to write conditions to filter out the data.

Multiple conditions can be combine with the help of logical operator ie. AND (&), OR (|)

Syntax:

data-frame[conditions]

data-frame[(conditions-1) & (conditions-2)] #AND(&) condition

data-frame[(conditions-1) | (conditions-2)] #OR(|) condition

Example: display 10 years old FOUR WHEELER vehicles from DataFrame

```
data[ (data['AGE']==10) &  
      (data['VEHICLE_TYPE']=='FOUR WHEELER')  
      ]
```

Data Visualization

- Python Libraries used for generating graphs
 - **Matplotlib**
 - **Seaborn**
- Download data visualization libraries
 - `pip install matplotlib`
 - `pip install seaborn`

Data Visualization

- Importing Matplotlib & Seaborn libraries

```
import matplotlib.pyplot as plt  
import seaborn as sn  
%matplotlib inline
```

Data Visualization Functions

- Bar Graphs Syntax

```
sn.barplot(x=, y=, data=)
```

```
sn.barplot(x=, y=, hue= ,data=)
```

Data Visualization

- **Histogram**

`plt.hist(x, bins=)`

- **Distribution or Density Plot**

`sn.distplot(x=)`

- **Boxplot**

`sn.boxplot(x=, data=)`

`sn.boxplot(x=, y=, data=)`

Data Visualization

- **Pairplot**

```
sn.pairplot(df[selected-columns])
```

```
sn.pairplot(df)          #For all the columns
```

- **Heatmap**

```
sn.heatmap(df[selected-columns].corr(), annot=True)
```

```
***annotate show the data values on graph
```