Unit-1 Introduction to Python Pandas

```
In [ ]:
           1 pip install pandas
In [39]:
           1 import pandas as pd
           2 print(pd.__version__) # 1.1.3
         1.1.3
         Series:
           • List: support
           • Tuple : support
           • Set : not support
           • Dictnory: support
           • if ?: that given object
In [40]:
           1 # Series :- 1 column, many rows & 1D
           2 import pandas as pd
           3 \mid a = [1,2,3]
           4 myvar = pd.Series(a)
           5 print(myvar)
           6 print(myvar[0])
           7 print(myvar[1])
           8 print(myvar[2])
          10  # Output :
          11
          12 # 0
                     1
          13 | # 1
                     2
          14 # 2
                     3
          15 | # dtype: int64
         0
              1
              2
         2
         dtype: int64
         1
         2
         3
In [41]:
           1 # Task For Tuple
           2 import pandas as pd
           3 \mid a = (1,2,3)
           4 myvar = pd.Series(a)
           5 print(myvar)
           6 print(myvar[0])
           7 print(myvar[1])
           8 print(myvar[2])
           9 | # print(myvar[3]) # Key Error
         0
              1
              2
         1
              3
         dtype: int64
         1
         2
         3
           1 # For another data type
           2 import pandas as pd
           3 a = (1.0, 2.0, 3.0)
           4 myvar = pd.Series(a)
           5 print(myvar)
           6 print(myvar[0])
           7 print(myvar[1])
           8 print(myvar[2])
         0
              1.0
         1
              2.0
              3.0
         dtype: float64
         1.0
         2.0
         3.0
```

```
In [43]:
           1 import pandas as pd
           2 | a = (1.0,2,3.0) # convert to float
           3 myvar = pd.Series(a)
           4 print(myvar)
           5 print(myvar[0])
           6 print(myvar[1])
           7 print(myvar[2])
         0
              1.0
         1
              2.0
              3.0
         dtype: float64
         1.0
         2.0
         3.0
In [44]:
           1 import pandas as pd
           2 | a = (1.0,2,'a') # give object datatype if any character input
           3 myvar = pd.Series(a)
           4 print(myvar)
           5 print(myvar[0])
           6 print(myvar[1])
           7 print(myvar[2])
         0
              1
              2
         1
         dtype: object
         1.0
         2
         а
          1 | import pandas as pd
In [45]:
           2 \mid a = \{1.0, 2, 'a'\}
           3 myvar = pd.Series(a)
           4 print(myvar)
           5 print(myvar[0])
           6 | print(myvar[1])
           7 print(myvar[2]) # TypeError: 'set' type is unordered
                                                    Traceback (most recent call last)
         TypeError
         <ipython-input-45-5074ab57916c> in <module>
               1 import pandas as pd
               2 a = \{1.0, 2, 'a'\}
         ----> 3 myvar = pd.Series(a)
               4 print(myvar)
               5 print(myvar[0])
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\series.py in __init__(self, data, index, dtype, name, copy, fast
         path)
             297
                                  pass
                              elif isinstance(data, (set, frozenset)):
             298
                                  raise TypeError(f"'{type(data).__name__}' type is unordered")
          --> 299
             300
                              else:
             301
                                  data = com.maybe_iterable_to_list(data)
         TypeError: 'set' type is unordered
In [46]:
           1 import pandas as pd
           2 | a = {'A':1,'B':2,'C':3}
           3 myvar = pd.Series(a)
           4 print(myvar)
           5 | print(myvar['A'])
           6 | print(myvar['B'])
             print(myvar['C'])
         Α
              1
         В
              2
              3
         C
         dtype: int64
         1
         2
         3
```

```
In [47]:
           1 import pandas as pd
           2 \mid a = \{'A':[1,2],'B':2,'C':3\} # if we pass dictniory in list give object.
           3 myvar = pd.Series(a)
           4 print(myvar)
           5 print(myvar['A'])
           6 print(myvar['B'])
           7 print(myvar['C'])
              [1, 2]
         C
                   3
         dtype: object
         [1, 2]
         3
In [48]:
          1 | import pandas as pd
           2 a = [1,2,3]
           3 myvar = pd.Series(a,index=['x','y']) # we have must pass 3 values.
           4 print(myvar) # ValueError: Length of passed values is 3, index implies 2.
         ValueError
                                                    Traceback (most recent call last)
         <ipython-input-48-349654d3951c> in <module>
               1 import pandas as pd
               2 a = [1,2,3]
         ----> 3 myvar = pd.Series(a,index=['x','y']) # we have must pass 3 values.
               4 print(myvar) # ValueError: Length of passed values is 3, index implies 2.
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\series.py in __init__(self, data, index, dtype, name, copy, fast
         path)
             311
                                  try:
             312
                                     if len(index) != len(data):
                                          raise ValueError(
         --> 313
                                              f"Length of passed values is {len(data)}, "
             314
                                              f"index implies {len(index)}."
             315
         ValueError: Length of passed values is 3, index implies 2.
In [49]:
          1 import pandas as pd
           2 \mid a = [1,2,3]
           3 | myvar = pd.Series(a,index=['x','y','z'])
           4 print(myvar)
           5
           6 # Output :
           7 | # x
                    1
           8 # y
                     2
          9 # z
                    3
          10 | # dtype: int64
              1
         Χ
              2
         У
         dtype: int64
In [50]:
          1 import pandas as pd
           2 | calories = {'day1':420,'day2':380,'day3':390}
           3 myvar = pd.Series(calories)
           4 print(myvar)
           6 # Output :
           7 | # day1
                        420
           8 # day2
                        380
           9 # day3
                        390
          10 | # dtype: int64
         day1
                 420
         day2
                 380
         day3
                 390
         dtype: int64
```

```
In [51]:
          1 import pandas as pd
           calories = {'day1':420,'day2':380,'day3':390}
           3 myvar = pd.Series(calories,index=['x','y','z'])
            print(myvar)
           6 # Output:
          7
                   NaN
             # X
            # y
                   NaN
          9 # z NaN
          10 # dtype: float64
            NaN
         У
             NaN
            NaN
         dtype: float64
In [52]:
          1 import pandas as pd
          calories = {'day1':420,'day2':380,'day3':390}
          3 | myvar = pd.Series(calories,index=['x','y','z','day1'])
          4 print(myvar) # NaN convert automacally float.
          6
             # Output :
          7 | # x
                         NaN
          8 | # y
                         NaN
          9 # z
                         NaN
          10 # day1
                       420.0
          11 # dtype: float64
                   NaN
         Χ
         У
                   NaN
                   NaN
         day1
                420.0
         dtype: float64
In [53]:
          1 import pandas as pd
          calories = {'day1':420,'day2':380,'day3':390}
          myvar = pd.Series(calories,index=['day2','day1'])
             print(myvar) # value must be same of particular key.
                 380
         day2
         day1
                 420
         dtype: int64
In [54]:
          1 a = [1,2,3,4,5,6]
          2 myvar = pd.Series(a)
          3 myvar[[0,1,3]] # when we pass multiple value then using list.
          4 | # myvar[0,1,3]
          5 myvar[0::2]
Out[54]: 0
              1
              3
         2
         dtype: int64
```

DataFrame:(2D)

• many rows many columns

```
In [55]:
          1 import pandas as pd
           2 data = {'calories':[420,380,390],'duration':[50,40,45]}
           3 df = pd.DataFrame(data)
             print(df)
           5
           6
                  calories duration
           7
             # 0
                        420
                                   50
           8
             # 1
                        380
                                   40
           9 # 2
                       390
                                   45
                      duration
            calories
         0
                 420
                            50
```

- loc & iloc(integer location)

380

390

- loc : accepts labels as well as int
- iloc: accepts only integer not a string

40

45

1

2

```
In [56]:
          1 import pandas as pd
           2 data = {'calories':[420,380,390],'duration':[50,40,45]}
          3 df = pd.DataFrame(data)
          4 print(df['calories'][0])
          5 print(df['calories'].loc[0])
           6 print(df['duration'].loc[1])
         420
         420
         40
In [57]:
          1 import pandas as pd
           2 data = {'calories':[420,380,390],'duration':[50,40,45]}
          3 df = pd.DataFrame(data)
          4 print(df)
            calories
                      duration
         0
                 420
                            50
                 380
                            40
         2
                 390
                            45
In [58]:
          1 import pandas as pd
          2 data = {'calories':[420,380,390],'duration':[50,40,45]}
          df = pd.DataFrame(data,index=['day1','day2','day3'])
          4 # print(df['calories'].loc[0]) # give key error because index is change.
          5 print(df['calories'].loc['day1']) # 420
          7 # Output :
                     calories duration
          8 #
             # day1
          9
                          420
                                     50
          10 | # day2
                          380
                                     40
                          390
                                     45
          11 # day3
         420
In [59]:
          1 import pandas as pd
          2 data = {'calories':[420,380,390],'duration':[50,40,45]}
          3 df = pd.DataFrame(data)
          4 print(df['calories'].iloc[0]) # 420
         420
In [60]:
          1 import pandas as pd
          2 | data = {'calories':[420,380,390],'duration':[50,40,45]}
          df = pd.DataFrame(data,index=['day1','day2','day3'])
          4 | # print(df['calories'].iloc['day1']) # TypeError:Cannot index by location index with a non-integer key
```

For CSV File.

Out[61]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
2	18.0	8	318.0	150	3436	11.0	70	1	plymouth satellite
3	16.0	8	304.0	150	3433	12.0	70	1	amc rebel sst
4	17.0	8	302.0	140	3449	10.5	70	1	ford torino
393	27.0	4	140.0	86	2790	15.6	82	1	ford mustang gl
394	44.0	4	97.0	52	2130	24.6	82	2	vw pickup
395	32.0	4	135.0	84	2295	11.6	82	1	dodge rampage
396	28.0	4	120.0	79	2625	18.6	82	1	ford ranger
397	31.0	4	119.0	82	2720	19.4	82	1	chevy s-10

398 rows × 9 columns

```
In [62]:
         1 import pandas as pd
         2 df = pd.read_csv('auto-mpg.csv') # converting into dataframe
         3 df.info()
         4
         5
            # Output :
         6
         7 # <class 'pandas.core.frame.DataFrame'>
         8 # RangeIndex: 398 entries, 0 to 397
         9 # Data columns (total 9 columns):
         10 # # Column Non-Null Count Dtype
                  -----
                             -----
         11 # ---
                  mpg 398 non-null float64
cylinders 398 non-null int64
         12 # 0
         13 # 1
         14 # 2
                  displacement 398 non-null float64
                  horsepower 398 non-null object
         15 # 3
                             398 non-null int64
         16 # 4
                  weight
                  acceleration 398 non-null float64
         17 # 5
         18 # 6
                  model year 398 non-null
                                             int64
         19 # 7
                  origin
                               398 non-null
                                             int64
         20 # 8 car name
                              398 non-null
                                             object
         21 # dtypes: float64(3), int64(4), object(2)
         22 # memory usage: 28.1+ KB
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 398 entries, 0 to 397
Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype					
0	mpg	398 non-null	float64					
1	cylinders	398 non-null	int64					
2	displacement	398 non-null	float64					
3	horsepower	398 non-null	object					
4	weight	398 non-null	int64					
5	acceleration	398 non-null	float64					
6	model year	398 non-null	int64					
7	origin	398 non-null	int64					
8	car name	398 non-null	object					
<pre>dtypes: float64(3), int64(4), object(2)</pre>								
memo	memory usage: 28.1+ KB							

In [63]: 1 help(pd)

```
Help on package pandas:
```

NAME

pandas

DESCRIPTION

pandas - a powerful data analysis and manipulation library for Python

pandas is a Python package providing fast, flexible, and expressive data structures designed to make working with "relational" or "labeled" data both easy and intuitive. It aims to be the fundamental high-level building block for doing practical, **real world** data analysis in Python. Additionally, it has the broader goal of becoming **the most powerful and flexible open source data analysis / manipulation tool available in any language**. It is already well on its way toward this goal.

Main Features

Here are just a few of the things that pandas does well:

- Easy handling of missing data in floating point as well as non-floating point data.
- Size mutability: columns can be inserted and deleted from DataFrame and higher dimensional objects
- Automatic and explicit data alignment: objects can be explicitly aligned to a set of labels, or the user can simply ignore the labels and let `Series`, `DataFrame`, etc. automatically align the data for you in computations.
- Powerful, flexible group by functionality to perform split-apply-combine operations on data sets, for both aggregating and transforming data.
- Make it easy to convert ragged, differently-indexed data in other Python and NumPy data structures into DataFrame objects.
- Intelligent label-based slicing, fancy indexing, and subsetting of large data sets.
- Intuitive merging and joining data sets.
- Flexible reshaping and pivoting of data sets.
- Hierarchical labeling of axes (possible to have multiple labels per tick).
- Robust IO tools for loading data from flat files (CSV and delimited),
 Excel files, databases, and saving/loading data from the ultrafast HDF5 format.
- Time series-specific functionality: date range generation and frequency conversion, moving window statistics, date shifting and lagging.

```
PACKAGE CONTENTS
    _config (package)
    _libs (package)
    _testing
    _typing
    _version
    api (package)
    arrays (package)
    compat (package)
    conftest
    core (package)
    errors (package)
    io (package)
    plotting (package)
    testing
    tests (package)
    tseries (package)
    util (package)
SUBMODULES
    _hashtable
    _lib
    _tslib
    offsets
FUNCTIONS
    __getattr__(name)
DATA
    IndexSlice = <pandas.core.indexing._IndexSlice object>
    NA = \langle NA \rangle
    NaT = NaT
    __docformat__ = 'restructuredtext'
    git version = 'db08276bc116c438d3fdee492026f8223584c477'
    describe_option = <pandas._config.config.CallableDynamicDoc object>
    get_option = <pandas._config.config.CallableDynamicDoc object>
    options = <pandas._config.config.DictWrapper object>
    reset_option = <pandas._config.config.CallableDynamicDoc object>
    set_option = <pandas._config.config.CallableDynamicDoc object>
VERSION
```

1.1.3

- head() & tail()

In [64]:

- 1 import pandas as pd
- df = pd.read_csv('auto-mpg.csv') # converting into dataframe
- 3 df.head() # given first 5 row print by default when args not pass.

Out[64]:

car name	origin	model year	acceleration	weight	horsepower	displacement	cylinders	mpg	
chevrolet chevelle malibu	1	70	12.0	3504	130	307.0	8	18.0	0
buick skylark 320	1	70	11.5	3693	165	350.0	8	15.0	1
plymouth satellite	1	70	11.0	3436	150	318.0	8	18.0	2
amc rebel sst	1	70	12.0	3433	150	304.0	8	16.0	3
ford torino	1	70	10.5	3449	140	302.0	8	17.0	4

In [65]:

- 1 import pandas as pd
- df = pd.read_csv('auto-mpg.csv') # converting into dataframe
- 3 df.tail() # given last 5 row print by default when args not pass.

Out[65]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
393	27.0	4	140.0	86	2790	15.6	82	1	ford mustang gl
394	44.0	4	97.0	52	2130	24.6	82	2	vw pickup
395	32.0	4	135.0	84	2295	11.6	82	1	dodge rampage
396	28.0	4	120.0	79	2625	18.6	82	1	ford ranger
397	31.0	4	119.0	82	2720	19.4	82	1	chevy s-10

In [66]:

- 1 **import** pandas **as** pd
- 2 | df = pd.read_csv('auto-mpg.csv') # converting into dataframe
- 3 df.head(10)

Out[66]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
2	18.0	8	318.0	150	3436	11.0	70	1	plymouth satellite
3	16.0	8	304.0	150	3433	12.0	70	1	amc rebel sst
4	17.0	8	302.0	140	3449	10.5	70	1	ford torino
5	15.0	8	429.0	198	4341	10.0	70	1	ford galaxie 500
6	14.0	8	454.0	220	4354	9.0	70	1	chevrolet impala
7	14.0	8	440.0	215	4312	8.5	70	1	plymouth fury iii
8	14.0	8	455.0	225	4425	10.0	70	1	pontiac catalina
9	15.0	8	390.0	190	3850	8.5	70	1	amc ambassador dpl

- In [67]:
- 1 import pandas as pd
- df = pd.read_csv('auto-mpg.csv') # converting into dataframe
- 3 df.tail(10)

Out[67]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
388	26.0	4	156.0	92	2585	14.5	82	1	chrysler lebaron medallion
389	22.0	6	232.0	112	2835	14.7	82	1	ford granada I
390	32.0	4	144.0	96	2665	13.9	82	3	toyota celica gt
391	36.0	4	135.0	84	2370	13.0	82	1	dodge charger 2.2
392	27.0	4	151.0	90	2950	17.3	82	1	chevrolet camaro
393	27.0	4	140.0	86	2790	15.6	82	1	ford mustang gl
394	44.0	4	97.0	52	2130	24.6	82	2	vw pickup
395	32.0	4	135.0	84	2295	11.6	82	1	dodge rampage
396	28.0	4	120.0	79	2625	18.6	82	1	ford ranger
397	31.0	4	119.0	82	2720	19.4	82	1	chevy s-10

Out[68]:

car name	origin	model year	acceleration	weight	horsepower	displacement	cylinders	mpg	
plymouth satellite custom	1	71	15.5	3439	105	225.0	6	16.0	34
chevrolet chevelle malibu	1	71	15.5	3329	100	250.0	6	17.0	35
ford torino 500	1	71	15.5	3302	88	250.0	6	19.0	36
amc matador	1	71	15.5	3288	100	232.0	6	18.0	37
chevrolet impala	1	71	12.0	4209	165	350.0	8	14.0	38
pontiac catalina brougham	1	71	11.5	4464	175	400.0	8	14.0	39
ford galaxie 500	1	71	13.5	4154	153	351.0	8	14.0	40
plymouth fury iii	1	71	13.0	4096	150	318.0	8	14.0	41
dodge monaco (sw)	1	71	11.5	4955	180	383.0	8	12.0	42
ford country squire (sw)	1	71	12.0	4746	170	400.0	8	13.0	43
pontiac safari (sw)	1	71	12.0	5140	175	400.0	8	13.0	44
amc hornet sportabout (sw)	1	71	13.5	2962	110	258.0	6	18.0	45
chevrolet vega (sw)	1	71	19.0	2408	72	140.0	4	22.0	46
pontiac firebird	1	71	15.0	3282	100	250.0	6	19.0	47
ford mustang	1	71	14.5	3139	88	250.0	6	18.0	48
mercury capri 2000	1	71	14.0	2220	86	122.0	4	23.0	49
opel 1900	2	71	14.0	2123	90	116.0	4	28.0	50
peugeot 304	2	71	19.5	2074	70	79.0	4	30.0	51
fiat 124b	2	71	14.5	2065	76	88.0	4	30.0	52
toyota corolla 1200	3	71	19.0	1773	65	71.0	4	31.0	53
datsun 1200	3	71	18.0	1613	69	72.0	4	35.0	54
volkswagen model 111	2	71	19.0	1834	60	97.0	4	27.0	55
plymouth cricket	1	71	20.5	1955	70	91.0	4	26.0	56

Out[69]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
34	16.0	6	225.0	105	3439	15.5	71	1	plymouth satellite custom
56	26.0	4	91.0	70	1955	20.5	71	1	plymouth cricket

- df: df ni bajuma hamesha column ave.
- loc: loc ni bajuma hamesha row ave.

Out[70]: 34 16.0 56 26.0

Name: mpg, dtype: float64

Out[71]:

	mpg	displacement
34	16.0	225.0
56	26.0	91.0

Out[72]:

```
mpg cylinders
0 18.0
               8
1 15.0
               8
2 18.0
               8
3 16.0
               8
4 17.0
               8
5 15.0
               8
6 14.0
               8
7 14.0
               8
8 14.0
               8
9 15.0
               8
10 15.0
               8
11 14.0
               8
12 15.0
               8
13 14.0
               8
14 24.0
15 22.0
               6
16 18.0
               6
17 21.0
18 27.0
19 26.0
               4
20 25.0
```

Out[73]: (398, 9)

```
In [74]:
           1 import pandas as pd
           2 df = pd.read_csv('auto-mpg.csv')
           3 | df.loc[-1] # ValueError
         ValueError
                                                    Traceback (most recent call last)
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexes\range.py in get_loc(self, key, method, tolerance)
         --> 355
                                      return self._range.index(new_key)
                                  except ValueError as err:
             356
         ValueError: -1 is not in range
         The above exception was the direct cause of the following exception:
                                                    Traceback (most recent call last)
         KeyError
         <ipython-input-74-124c5c46c56e> in <module>
               1 import pandas as pd
               2 df = pd.read_csv('auto-mpg.csv')
         ----> 3 df.loc[-1] # ValueError
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexing.py in __getitem__(self, key)
             877
             878
                              maybe_callable = com.apply_if_callable(key, self.obj)
          --> 879
                              return self._getitem_axis(maybe_callable, axis=axis)
             880
             881
                      def _is_scalar_access(self, key: Tuple):
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexing.py in _getitem_axis(self, key, axis)
            1108
                         # fall thru to straight lookup
                          self._validate_key(key, axis)
            1109
          -> 1110
                         return self._get_label(key, axis=axis)
            1111
            1112
                      def _get_slice_axis(self, slice_obj: slice, axis: int):
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexing.py in _get label(self, label, axis)
            1057
                      def _get_label(self, label, axis: int):
            1058
                          # GH#5667 this will fail if the label is not present in the axis.
         -> 1059
                         return self.obj.xs(label, axis=axis)
            1060
            1061
                      def _handle_lowerdim_multi_index_axis0(self, tup: Tuple):
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\generic.py in xs(self, key, axis, level, drop_level)
             3489
                              loc, new_index = self.index.get_loc_level(key, drop_level=drop_level)
             3490
                          else:
          -> 3491
                              loc = self.index.get_loc(key)
             3492
            3493
                              if isinstance(loc, np.ndarray):
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexes\range.py in get_loc(self, key, method, tolerance)
             355
                                      return self._range.index(new_key)
             356
                                  except ValueError as err:
          --> 357
                                      raise KeyError(key) from err
             358
                              raise KeyError(key)
             359
                         return super().get_loc(key, method=method, tolerance=tolerance)
         KeyError: -1
In [75]:
           1 import pandas as pd
           2 | df = pd.read_csv('auto-mpg.csv')
           3 | df.loc[:-1] # only give column name.
Out[75]:
            mpg cylinders displacement horsepower weight acceleration model year origin car name
         1 | import pandas as pd
           2 df = pd.read_csv('auto-mpg.csv')
           3 print(df.loc[:-1])
           5 # Empty DataFrame
           6 # Columns: [mpq, cylinders, displacement, horsepower, weight, acceleration, model year, origin, car name]
           7 | # Index: []
         Empty DataFrame
         Columns: [mpg, cylinders, displacement, horsepower, weight, acceleration, model year, origin, car name]
         Index: []
```

Statistics

not analysis of object only Integer & Float.

Out[77]:

	mpg	cylinders	displacement	weight	acceleration	model year	origin
count	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000
mean	23.514573	5.454774	193.425879	2970.424623	15.568090	76.010050	1.572864
std	7.815984	1.701004	104.269838	846.841774	2.757689	3.697627	0.802055
min	9.000000	3.000000	68.000000	1613.000000	8.000000	70.000000	1.000000
25%	17.500000	4.000000	104.250000	2223.750000	13.825000	73.000000	1.000000
50%	23.000000	4.000000	148.500000	2803.500000	15.500000	76.000000	1.000000
75%	29.000000	8.000000	262.000000	3608.000000	17.175000	79.000000	2.000000
max	46.600000	8.000000	455.000000	5140.000000	24.800000	82.000000	3.000000

Statistics All Operations:

Out[78]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
count	398.000000	398.000000	398.000000	398	398.000000	398.000000	398.000000	398.000000	398
unique	NaN	NaN	NaN	94	NaN	NaN	NaN	NaN	305
top	NaN	NaN	NaN	150	NaN	NaN	NaN	NaN	ford pinto
freq	NaN	NaN	NaN	22	NaN	NaN	NaN	NaN	6
mean	23.514573	5.454774	193.425879	NaN	2970.424623	15.568090	76.010050	1.572864	NaN
std	7.815984	1.701004	104.269838	NaN	846.841774	2.757689	3.697627	0.802055	NaN
min	9.000000	3.000000	68.000000	NaN	1613.000000	8.000000	70.000000	1.000000	NaN
25%	17.500000	4.000000	104.250000	NaN	2223.750000	13.825000	73.000000	1.000000	NaN
50%	23.000000	4.000000	148.500000	NaN	2803.500000	15.500000	76.000000	1.000000	NaN
75%	29.000000	8.000000	262.000000	NaN	3608.000000	17.175000	79.000000	2.000000	NaN
max	46.600000	8.000000	455.000000	NaN	5140.000000	24.800000	82.000000	3.000000	NaN

Out[79]:

	mpg	cylinders	displacement	weight	acceleration	model year	origin
count	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000
mean	23.514573	5.454774	193.425879	2970.424623	15.568090	76.010050	1.572864
std	7.815984	1.701004	104.269838	846.841774	2.757689	3.697627	0.802055
min	9.000000	3.000000	68.000000	1613.000000	8.000000	70.000000	1.000000
25%	17.500000	4.000000	104.250000	2223.750000	13.825000	73.000000	1.000000
50%	23.000000	4.000000	148.500000	2803.500000	15.500000	76.000000	1.000000
75%	29.000000	8.000000	262.000000	3608.000000	17.175000	79.000000	2.000000
max	46.600000	8.000000	455.000000	5140.000000	24.800000	82.000000	3.000000

Out[80]:

	horsepower	car name
count	398	398
unique	94	305
top	150	ford pinto
freq	22	6

Out[81]:

```
        korsepower
        car name

        count
        398
        398

        unique
        94
        305

        top
        150
        ford pinto

        freq
        22
        6
```

Out[82]:

	mpg	cylinders	displacement	weight	acceleration	model year	origin
count	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000
mean	23.514573	5.454774	193.425879	2970.424623	15.568090	76.010050	1.572864
std	7.815984	1.701004	104.269838	846.841774	2.757689	3.697627	0.802055
min	9.000000	3.000000	68.000000	1613.000000	8.000000	70.000000	1.000000
25%	17.500000	4.000000	104.250000	2223.750000	13.825000	73.000000	1.000000
50%	23.000000	4.000000	148.500000	2803.500000	15.500000	76.000000	1.000000
75%	29.000000	8.000000	262.000000	3608.000000	17.175000	79.000000	2.000000
max	46.600000	8.000000	455.000000	5140.000000	24.800000	82.000000	3.000000

Out[83]:

	mpg	cylinders	displacement	weight	acceleration	model year	origin
count	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000
mean	23.514573	5.454774	193.425879	2970.424623	15.568090	76.010050	1.572864
std	7.815984	1.701004	104.269838	846.841774	2.757689	3.697627	0.802055
min	9.000000	3.000000	68.000000	1613.000000	8.000000	70.000000	1.000000
25%	17.500000	4.000000	104.250000	2223.750000	13.825000	73.000000	1.000000
50%	23.000000	4.000000	148.500000	2803.500000	15.500000	76.000000	1.000000
75%	29.000000	8.000000	262.000000	3608.000000	17.175000	79.000000	2.000000
max	46.600000	8.000000	455.000000	5140.000000	24.800000	82.000000	3.000000

Out[84]: count

```
398.000000
          23.514573
mean
           7.815984
std
min
           9.000000
25%
          17.500000
50%
          23.000000
          29.000000
75%
          46.600000
max
Name: mpg, dtype: float64
```

Out[85]:

```
mpg cylinders displacement
                                               weight acceleration model year origin
       6.000000
                                6.000000
                                             6.000000
                                                          6.000000
                                                                           6.0
                                                                                  6.0
count
                       6.0
      16.500000
                       8.0
                              335.000000 3642.666667
                                                         11.166667
                                                                          70.0
                                                                                  1.0
mean
       1.378405
                       0.0
                                          355.980149
                                                          0.816497
                                                                           0.0
                               49.363954
                                                                                  0.0
  std
 min 15.000000
                       8.0
                              302.000000 3433.000000
                                                         10.000000
                                                                          70.0
                                                                                  1.0
 25% 15.250000
                       8.0
                              304.750000 3439.250000
                                                         10.625000
                                                                          70.0
                                                                                  1.0
                                                         11.250000
      16.500000
                       8.0
                              312.500000 3476.500000
                                                                          70.0
                                                                                  1.0
                       8.0
                              342.000000 3645.750000
                                                         11.875000
                                                                          70.0
 75% 17.750000
                                                                                  1.0
 max 18.000000
                       8.0
                              429.000000 4341.000000
                                                         12.000000
                                                                          70.0
                                                                                  1.0
```

Out[86]:

	mpg	cylinders	displacement	weight	acceleration	model year	origin
count	5.00000	5.0	5.000000	5.000000	5.00000	5.0	5.0
mean	16.80000	8.0	316.200000	3503.000000	11.40000	70.0	1.0
std	1.30384	0.0	19.879638	110.006818	0.65192	0.0	0.0
min	15.00000	8.0	302.000000	3433.000000	10.50000	70.0	1.0
25%	16.00000	8.0	304.000000	3436.000000	11.00000	70.0	1.0
50%	17.00000	8.0	307.000000	3449.000000	11.50000	70.0	1.0
75%	18.00000	8.0	318.000000	3504.000000	12.00000	70.0	1.0
max	18.00000	8.0	350.000000	3693.000000	12.00000	70.0	1.0

```
Out[87]: count
                    6.000000
                   16.500000
         mean
                    1.378405
         std
         min
                   15.000000
         25%
                   15.250000
         50%
                   16.500000
         75%
                   17.750000
                   18.000000
         max
         Name: mpg, dtype: float64
```

```
In [88]: 1 import pandas as pd 2 import numpy as np
```

df = pd.read_csv('auto-mpg.csv')
df.describe(percentiles=[0.3,0.57,0.83])

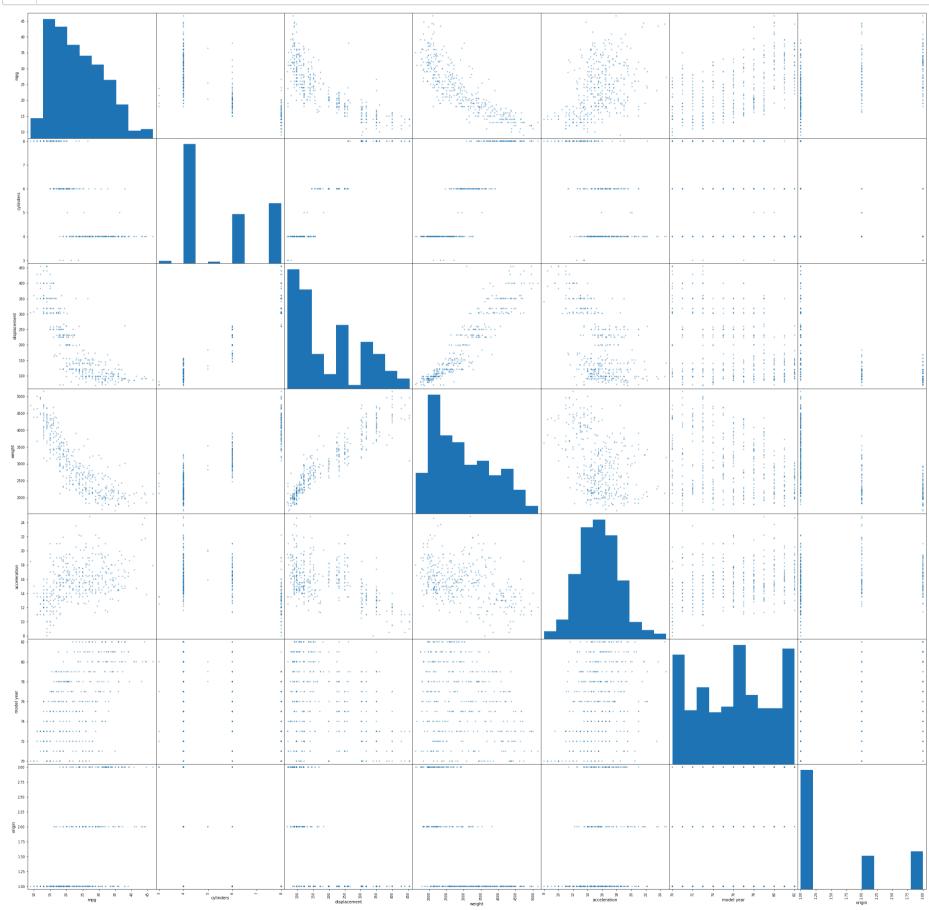
Out[88]:

	mpg	cylinders	displacement	weight	acceleration	model year	origin
count	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000
mean	23.514573	5.454774	193.425879	2970.424623	15.568090	76.010050	1.572864
std	7.815984	1.701004	104.269838	846.841774	2.757689	3.697627	0.802055
min	9.000000	3.000000	68.000000	1613.000000	8.000000	70.000000	1.000000
30%	18.000000	4.000000	112.000000	2301.000000	14.200000	73.000000	1.000000
50%	23.000000	4.000000	148.500000	2803.500000	15.500000	76.000000	1.000000
57.0%	24.358000	6.000000	187.350000	2969.060000	15.900000	77.000000	1.000000
83%	31.951000	8.000000	318.000000	3940.000000	18.151000	80.000000	3.000000
max	46.600000	8.000000	455.000000	5140.000000	24.800000	82.000000	3.000000

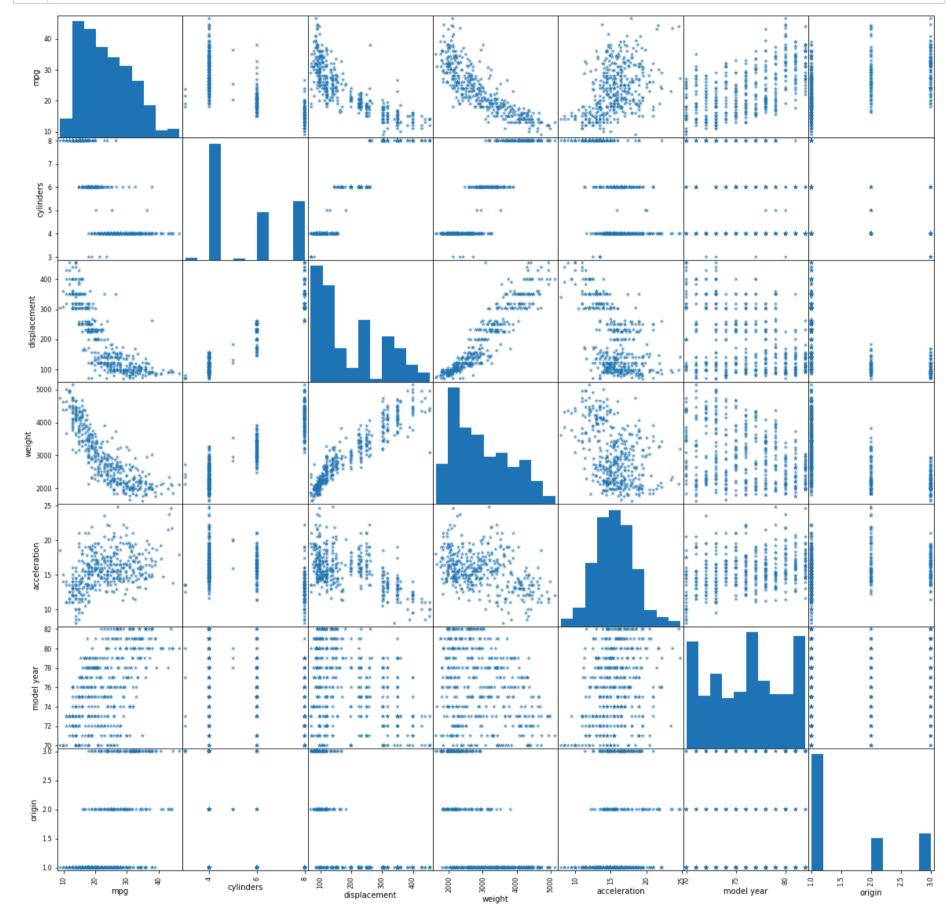
Corr:- corelation of cofficient

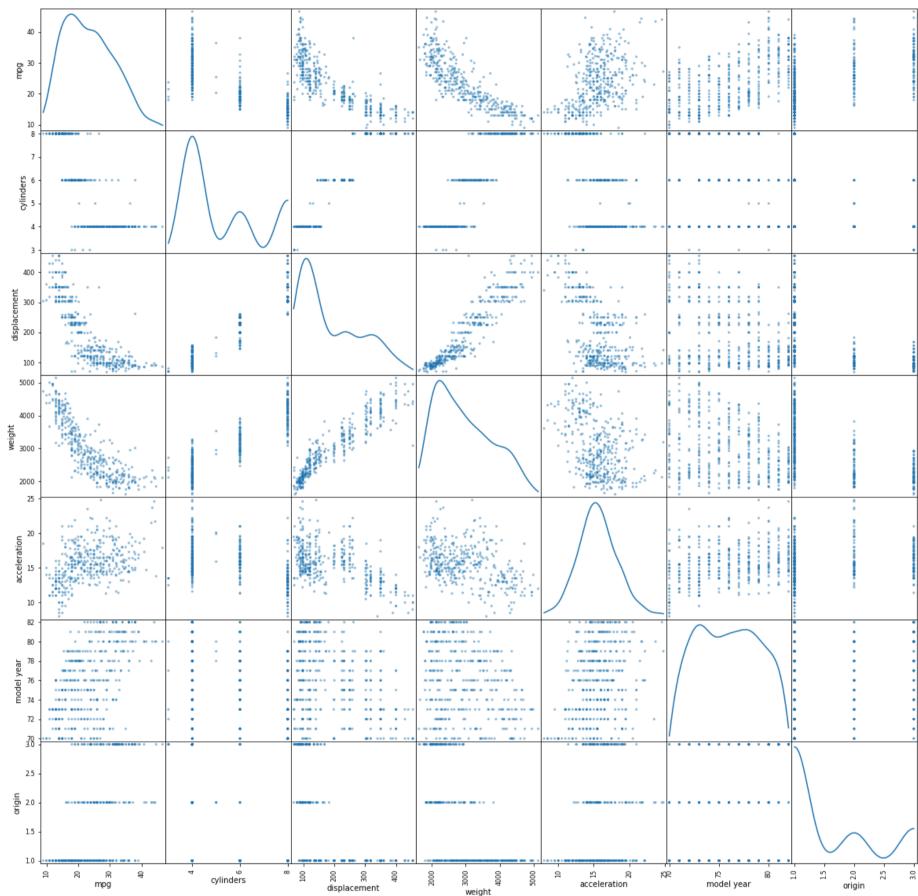
Out[3]:

	mpg	cylinders	displacement	weight	acceleration	model year	origin
mpg	1.000000	-0.775396	-0.804203	-0.831741	0.420289	0.579267	0.563450
cylinders	-0.775396	1.000000	0.950721	0.896017	-0.505419	-0.348746	-0.562543
displacement	-0.804203	0.950721	1.000000	0.932824	-0.543684	-0.370164	-0.609409
weight	-0.831741	0.896017	0.932824	1.000000	-0.417457	-0.306564	-0.581024
acceleration	0.420289	-0.505419	-0.543684	-0.417457	1.000000	0.288137	0.205873
model year	0.579267	-0.348746	-0.370164	-0.306564	0.288137	1.000000	0.180662
origin	0.563450	-0.562543	-0.609409	-0.581024	0.205873	0.180662	1.000000



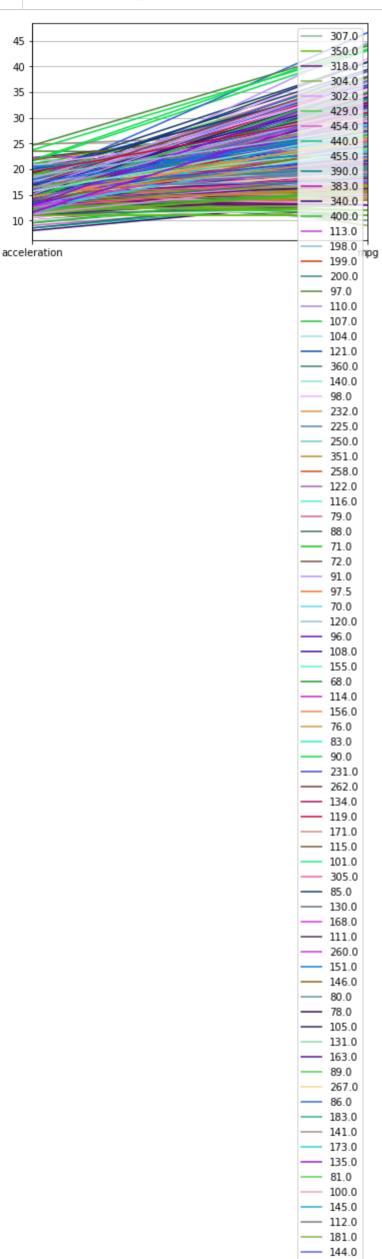
```
In [4]: 1 import pandas as pd
2 import matplotlib.pyplot as plt
3 df = pd.read_csv('auto-mpg.csv')
4 pd.plotting.scatter_matrix(df,figsize=[20,20],marker="*",alpha=0.7)
5 plt.show()
```

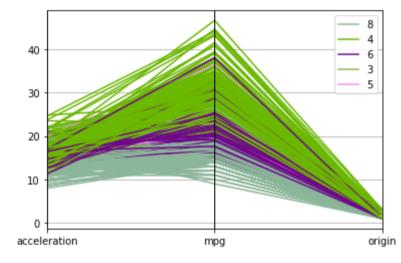




mpg

acceleration





Out[9]:

```
        model year
        70
        71
        72
        73
        74
        75
        76
        77
        78
        79
        80
        81
        82

        cylinders
        ...
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        ...
        ...
```

7 13 20 5

Data cleaning :-

Out[10]:

	name	region	sales	expense
0	а	ma	10.0	50.0
1	NaN	NaN	NaN	NaN
2	NaN	mp	30.0	70.0
3	d	gu	NaN	NaN
4	е	NaN	50.0	90.0

```
In [11]:
           1 import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                            'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                            'sales':[10,np.nan,30,np.nan,50],
                                            'expense':[50,np.nan,70,np.nan,90]})
             sales_data.isna() # boolean type
Out[11]:
             name region sales expense
            False
                   False False
                                 False
              True
                          True
                                  True
                    True
             True
                   False False
                                 False
                   False
                          True
                                  True
             False
                    True False
                                 False
          4 False
In [12]:
             import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                            'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                            'sales':[10,np.nan,30,np.nan,50],
           7
                                            'expense':[50,np.nan,70,np.nan,90]})
             sales_data.isna().sum()
Out[12]: name
                     2
         region
                     2
         sales
         expense
                     2
         dtype: int64
In [13]:
              import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                            'region':['ma',np.nan,'mp','gu',np.nan],
           5
           6
                                            'sales':[10,np.nan,30,np.nan,50],
                                            'expense':[50,np.nan,70,np.nan,90]})
             sales_data.dropna() # nan value row remove.
Out[13]:
             name region sales expense
          0
                          10.0
                                  50.0
                     ma
In [14]:
              import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                            'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                            'sales':[10,np.nan,30,np.nan,50],
           7
                                            'expense':[50,np.nan,70,np.nan,90]})
             sales_data.dropna(how="all") # how=all :- je row ma badha nan hoy to ej row kadhe.
Out[14]:
             name region sales expense
          0
                          10.0
                                  50.0
                а
                     ma
              NaN
                          30.0
                                  70.0
                     mp
                d
                          NaN
                                  NaN
                      gu
                    NaN
                          50.0
                                  90.0
In [15]:
              import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                            'region':['ma',np.nan,'mp','gu',np.nan],
                                            'sales':[10,np.nan,30,np.nan,50],
           6
                                            'expense':[50,np.nan,70,np.nan,90]})
             sales_data.dropna(how="any")
```

```
Out[15]:
```

```
name region sales expense

o a ma 10.0 50.0
```

```
        0
        a
        ma
        10.0
        50.0

        2
        NaN
        mp
        30.0
        70.0

        4
        expense

        a
        ma
        10.0
        50.0

        90.0
        90.0
```

Out[17]:

```
        name
        region
        sales
        expense

        0
        a
        ma
        10.0
        50.0

        2
        NaN
        mp
        30.0
        70.0
```

Out[18]:

	name	region	sales	expense
0	а	ma	10.0	50.0
2	NaN	mp	30.0	70.0
4	6	NaN	50.0	90.0

Out[19]:

	name	region	sales	expense
0	а	ma	10.0	50.0
2	NaN	mp	30.0	70.0
3	d	gu	NaN	NaN
4	е	NaN	50.0	90.0

Out[20]:

```
name region sales expense

o a ma 10.0 50.0
```

Out[21]:

```
        name

        0
        a

        1
        b

        2
        c

        3
        d
```

fillna()

Out[22]:

	name	region	sales	expense
0	а	ma	10.0	50.0
1	0	0	0.0	0.0
2	0	mp	30.0	70.0
3	d	gu	0.0	0.0
4	е	0	50.0	90.0

Out[23]:

	name	region	sales	expense
0	а	ma	10.0	50.0
1	30	30	30.0	30.0
2	30	mp	30.0	70.0
3	d	gu	30.0	30.0
4	е	30	50.0	90.0

Out[24]: 0 10.0

1 30.0

2 30.0

3 30.0

4 50.0

Name: sales, dtype: float64

```
In [25]:
           1 import pandas as pd
           2 | import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                           'sales':[10,np.nan,30,np.nan,50],
                                           'expense':[50,np.nan,70,np.nan,90]})
             sales_data['sales'].fillna(sales_data["sales"].median())
Out[25]: 0
              10.0
               30.0
         1
               30.0
         2
               30.0
         3
         4
               50.0
         Name: sales, dtype: float64
In [26]:
           1 | import pandas as pd
             import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                           'sales':[10,np.nan,30,np.nan,50],
           7
                                           'expense':[50,np.nan,70,np.nan,90]})
           8 sales_data['sales'].mode()
Out[26]: 0
              10.0
               30.0
         2
               50.0
         dtype: float64
In [27]:
              import pandas as pd
              import numpy as np
           3
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
                                           'sales':[10,np.nan,30,np.nan,50],
           6
           7
                                           'expense':[50,np.nan,70,np.nan,90]})
              sales_data['sales'].fillna(sales_data["sales"].mode()[0])
Out[27]: 0
              10.0
         1
               10.0
         2
               30.0
               10.0
               50.0
         Name: sales, dtype: float64
In [28]:
              import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
                                           'sales':[10,np.nan,30,np.nan,50],
                                           'expense':[50,np.nan,70,np.nan,90]})
              sales_data['sales'].fillna(sales_data["sales"].mode()[1])
Out[28]: 0
               10.0
               30.0
         1
         2
               30.0
         3
               30.0
               50.0
         Name: sales, dtype: float64
In [29]:
             import pandas as pd
             import numpy as np
           4
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
           5
                                           'region':['ma',np.nan,'mp','gu',np.nan],
                                           'sales':[10,np.nan,30,np.nan,50],
           6
           7
                                           'expense':[50,np.nan,70,np.nan,90]})
              sales_data['sales'].fillna(sales_data["sales"].mode()[2])
Out[29]: 0
               10.0
               50.0
         2
               30.0
               50.0
         3
         4
               50.0
         Name: sales, dtype: float64
```

```
In [30]:
           1 import pandas as pd
             import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                           'sales':[10,30,30,np.nan,50],
                                           'expense':[50,np.nan,70,np.nan,90]})
           8 sales_data
Out[30]:
             name region sales expense
                          10.0
                                  50.0
                     ma
              NaN
                          30.0
                    NaN
                                  NaN
                                  70.0
                          30.0
              NaN
                     mp
                d
                      gu
                          NaN
                                  NaN
                          50.0
                                  90.0
                е
                    NaN
In [31]:
           1 import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                           'sales':[10,30,30,np.nan,50],
           7
                                           'expense':[50,np.nan,70,np.nan,90]})
             sales_data['sales'].mode()
Out[31]: 0
              30.0
         dtype: float64
In [32]:
              import pandas as pd
              import numpy as np
           3
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
                                           'sales':[10,30,30,np.nan,50],
                                           'expense':[50,np.nan,70,np.nan,90]})
           7
              sales_data['sales'].fillna(sales_data["sales"].mode()[0])
Out[32]: 0
               10.0
         1
               30.0
         2
               30.0
         3
               30.0
               50.0
         Name: sales, dtype: float64
In [33]:
              import pandas as pd
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
           5
                                           'region':['ma',np.nan,'mp','gu',np.nan],
                                           'sales':[10,20,30,np.nan,50],
                                           'expense':[50,np.nan,70,np.nan,90]})
             sales_data['sales'].mode()
              sales_data
Out[33]:
                  region sales expense
             name
                          10.0
                                  50.0
                а
                     ma
              NaN
                    NaN
                          20.0
                                  NaN
              NaN
                     mp
                          30.0
                                  70.0
                d
                          NaN
                                  NaN
                      gu
                    NaN 50.0
                                  90.0
In [34]:
              import pandas as pd
              import numpy as np
             sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
                                           'sales':[10,20,30,np.nan,50],
           6
                                           'expense':[50,np.nan,70,np.nan,90]})
             sales_data.dropna(inplace=True)
              sales_data
Out[34]:
             name region sales expense
                          10.0
                                  50.0
                     ma
```

```
Ch.-1_Sem-4_T-1_Yash - Jupyter Notebook
In [35]:
           1 import pandas as pd
           2 import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
                                           'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                            'sales':[10,20,30,np.nan,50],
                                            'expense':[50,np.nan,70,np.nan,90]})
             sales_data = sales_data.dropna()
              sales_data
Out[35]:
             name region sales expense
          0
                          10.0
                                  50.0
              import pandas as pd
In [36]:
              import numpy as np
              sales_data = pd.DataFrame({'name':['a',np.nan,np.nan,'d','e'],
           5
                                            'region':['ma',np.nan,'mp','gu',np.nan],
           6
                                            'sales':[10,20,30,np.nan,50],
                                           'expense':[50,np.nan,70,np.nan,90]})
             sales_data['sales'].fillna(30,inplace=True)
             sales_data
Out[36]:
             name region sales expense
                          10.0
                                  50.0
                а
                     ma
              NaN
                    NaN
                          20.0
                                  NaN
              NaN
                          30.0
                                  70.0
                     mp
                d
                          30.0
                                  NaN
                      gu
                    NaN
                          50.0
                е
In [37]:
              import pandas as pd
                  'A':['TA','TB','TB','TC','TA'],
           3
           4
                   'B':[50,40,40,30,50],
           5
                  'C':[True,False,False,True]
           6 }
             df = pd.DataFrame(data)
           8 dups = df.duplicated()
              print(dups)
         0
               False
         1
               False
         2
               True
         3
               False
               True
         dtype: bool
In [38]:
              import pandas as pd
              data = {
           2
                  'A':['TA','TB','TB','TC','TA'],
           3
           4
                  'B':[50,40,40,30,50],
           5
                  'C':[True,False,False,False,True]
           6
             }
             df = pd.DataFrame(data)
             dups = df.duplicated()
              print(dups)
           9
          10
          11 | df = df.drop_duplicates()
```

```
12 df
```

0 False 1 False

2 True 3 False

4 True

dtype: bool

Out[38]:

```
A B
           С
0 TA 50 True
1 TB 40 False
3 TC 30 False
```

```
In [7]:
           1 import pandas as pd
           2
              data = {
           3
                   'A':['TA','TB','TB','TC','TA'],
                   'B':[50,40,40,30,50],
           4
                   'C':[True,False,False,True]
           5
           6
             }
           7
              df = pd.DataFrame(data)
              dups = df.duplicated()
           9
              print(dups)
          10
          11
             df = df.reset_index(drop=True)
          12
              df
         0
               False
         1
               False
         2
               True
         3
               False
               True
         dtype: bool
Out[7]:
              A B
                       С
          0 TA 50
                     True
            TB 40 False
            TB 40 False
          3 TC 30 False
          4 TA 50
                     True
In [8]:
              import pandas as pd
              import numpy as np
           3
              df = pd.read_csv('auto-mpg.csv')
Out[8]:
               mpg cylinders displacement horsepower weight acceleration model year origin
                                                                                                      car name
               18.0
                                                                                 70
            0
                           8
                                    307.0
                                                 130
                                                       3504
                                                                    12.0
                                                                                        1 chevrolet chevelle malibu
            1
               15.0
                           8
                                    350.0
                                                       3693
                                                                                 70
                                                  165
                                                                    11.5
                                                                                        1
                                                                                                 buick skylark 320
               18.0
                                                                                 70
            2
                           8
                                    318.0
                                                 150
                                                       3436
                                                                    11.0
                                                                                        1
                                                                                                 plymouth satellite
               16.0
                                    304.0
                                                       3433
                                                                                 70
            3
                           8
                                                  150
                                                                    12.0
                                                                                        1
                                                                                                    amc rebel sst
               17.0
                           8
                                    302.0
                                                       3449
                                                                                 70
            4
                                                  140
                                                                    10.5
                                                                                        1
                                                                                                      ford torino
          393
               27.0
                           4
                                    140.0
                                                  86
                                                       2790
                                                                    15.6
                                                                                 82
                                                                                        1
                                                                                                  ford mustang gl
               44.0
                                                       2130
                                                                                        2
          394
                           4
                                     97.0
                                                  52
                                                                    24.6
                                                                                 82
                                                                                                      vw pickup
                                                       2295
          395
               32.0
                                    135.0
                                                  84
                                                                    11.6
                                                                                 82
                                                                                        1
                                                                                                  dodge rampage
               28.0
                                    120.0
                                                  79
                                                       2625
          396
                           4
                                                                    18.6
                                                                                 82
                                                                                        1
                                                                                                      ford ranger
                                    119.0
                                                       2720
                                                                                                      chevy s-10
          397 31.0
                                                  82
                                                                    19.4
                                                                                 82
                                                                                        1
         398 rows × 9 columns
In [9]:
              import pandas as pd
              import numpy as np
           3 df = pd.read_csv('auto-mpg.csv')
           4 df['horsepower']=="?"
Out[9]: 0
                 False
                 False
         1
                 False
         2
         3
                 False
         4
                 False
         393
                 False
         394
                 False
         395
                 False
                 False
         396
         397
                 False
         Name: horsepower, Length: 398, dtype: bool
```

```
In [10]:
          1 import pandas as pd
          2 import numpy as np
          3 df = pd.read_csv('auto-mpg.csv')
          4 df[df['horsepower']=="?"]
```

Out[10]:

	mpg	cylinaers	aispiacement	norsepower	weight	acceleration	model year	origin	car name
32	25.0	4	98.0	?	2046	19.0	71	1	ford pinto
126	21.0	6	200.0	?	2875	17.0	74	1	ford maverick
330	40.9	4	85.0	?	1835	17.3	80	2	renault lecar deluxe
336	23.6	4	140.0	?	2905	14.3	80	1	ford mustang cobra
354	34.5	4	100.0	?	2320	15.8	81	2	renault 18i
374	23.0	4	151.0	?	3035	20.5	82	1	amc concord dl

In [13]:

```
1 import pandas as pd
2 import numpy as np
4 df = pd.read_csv('auto-mpg.csv')
5 df.loc[[32,126,330,336,354,374]]
```

Out[13]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
32	25.0	4	98.0	?	2046	19.0	71	1	ford pinto
126	21.0	6	200.0	?	2875	17.0	74	1	ford maverick
330	40.9	4	85.0	?	1835	17.3	80	2	renault lecar deluxe
336	23.6	4	140.0	?	2905	14.3	80	1	ford mustang cobra
354	34.5	4	100.0	?	2320	15.8	81	2	renault 18i
374	23.0	4	151.0	?	3035	20.5	82	1	amc concord dl

```
In [15]:
          1 import pandas as pd
          2 import numpy as np
```

4 df = pd.read_csv('auto-mpg.csv') 5 df[df['horsepower']!="?"]

Out[15]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
2	18.0	8	318.0	150	3436	11.0	70	1	plymouth satellite
3	16.0	8	304.0	150	3433	12.0	70	1	amc rebel sst
4	17.0	8	302.0	140	3449	10.5	70	1	ford torino
393	27.0	4	140.0	86	2790	15.6	82	1	ford mustang gl
394	44.0	4	97.0	52	2130	24.6	82	2	vw pickup
395	32.0	4	135.0	84	2295	11.6	82	1	dodge rampage
396	28.0	4	120.0	79	2625	18.6	82	1	ford ranger
397	31.0	4	119.0	82	2720	19.4	82	1	chevy s-10

392 rows × 9 columns

Out[16]:

	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	8	350.0	165	3693	11.5	70	1	buick skylark 320
2	8	318.0	150	3436	11.0	70	1	plymouth satellite
3	8	304.0	150	3433	12.0	70	1	amc rebel sst
4	8	302.0	140	3449	10.5	70	1	ford torino
393	4	140.0	86	2790	15.6	82	1	ford mustang gl
394	4	97.0	52	2130	24.6	82	2	vw pickup
395	4	135.0	84	2295	11.6	82	1	dodge rampage
396	4	120.0	79	2625	18.6	82	1	ford ranger
397	4	119.0	82	2720	19.4	82	1	chevy s-10

398 rows × 8 columns

Out[17]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
3	16.0	8	304.0	150	3433	12.0	70	1	amc rebel sst
4	17.0	8	302.0	140	3449	10.5	70	1	ford torino
5	15.0	8	429.0	198	4341	10.0	70	1	ford galaxie 500
393	27.0	4	140.0	86	2790	15.6	82	1	ford mustang gl
394	44.0	4	97.0	52	2130	24.6	82	2	vw pickup
395	32.0	4	135.0	84	2295	11.6	82	1	dodge rampage
396	28.0	4	120.0	79	2625	18.6	82	1	ford ranger
397	31.0	4	119.0	82	2720	19.4	82	1	chevy s-10

397 rows × 9 columns

Outlier:

```
In [25]:
          1 import pandas as pd
             import numpy as np
             def find_outlier(ds,col):
                 Q1 = ds[col].quantile(0.25)
                 Q3 = ds[col].quantile(0.75)
           6
           7
                 IQR = Q3 - Q1
                 low_val = Q1 - (1.5*IQR)
                 high_val = Q3 + (1.5*IQR)
                 ds = ds.loc[(ds[col]<low_val)|(ds[col]>high_val)]
          10
          11
                 return ds
          12
          df = pd.read_csv('auto-mpg.csv')
          14 find_outlier(df,"mpg")
```

```
Out[25]: mpg cylinders displacement horsepower weight acceleration model year origin car name

322 46.6 4 86.0 65 2110 17.9 80 3 mazda glc
```

```
In [24]: 1 import pandas as pd
import numpy as np

df = pd.read_csv('auto-mpg.csv')
df.describe()
```

Out[24]:

```
mpg
                    cylinders displacement
                                                 weight acceleration model year
                                                                                      origin
count 398.000000
                  398.000000
                                                                     398.000000
                                                                                 398.000000
                                398.000000
                                             398.000000
                                                          398.000000
       23.514573
                    5.454774
                                193.425879 2970.424623
                                                           15.568090
                                                                      76.010050
                                                                                    1.572864
mean
                                                                                   0.802055
        7.815984
                    1.701004
                                104.269838
                                             846.841774
                                                            2.757689
                                                                       3.697627
  std
                                                                      70.000000
        9.000000
                    3.000000
                                 68.000000 1613.000000
                                                            8.000000
                                                                                    1.000000
 min
 25%
       17.500000
                    4.000000
                                104.250000 2223.750000
                                                           13.825000
                                                                       73.000000
                                                                                    1.000000
 50%
       23.000000
                    4.000000
                                148.500000 2803.500000
                                                           15.500000
                                                                       76.000000
                                                                                    1.000000
       29.000000
                    8.000000
                                                                       79.000000
                                                                                   2.000000
 75%
                                262.000000 3608.000000
                                                           17.175000
       46.600000
                    8.000000
                                455.000000 5140.000000
                                                           24.800000
                                                                      82.000000
                                                                                    3.000000
 max
```

```
In [26]:
           1 import pandas as pd
             import numpy as np
           3
             def find_outlier(ds,col):
                 Q1 = ds[col].quantile(0.25)
           5
                 Q3 = ds[col].quantile(0.75)
           6
           7
                 IQR = Q3 - Q1
                 low_val = Q1 - (1.5*IQR)
           8
                 high_val = Q3 + (1.5*IQR)
           9
          10
                 ds = ds.loc[(ds[col]<low_val)|(ds[col]>high_val)]
          11
                  return ds
          12
          df = pd.read_csv('auto-mpg.csv')
          14 find_outlier(df,"acceleration")
```

Out[26]:

car name	origin	model year	acceleration	weight	horsepower	displacement	cylinders	mpg	
plymouth fury ii	1	70	8.5	4312	215	440.0	8	14.0	7
amc ambassador dp	1	70	8.5	3850	190	390.0	8	15.0	9
plymouth 'cuda 340	1	70	8.0	3609	160	340.0	8	14.0	11
volkswagen type 3	2	72	23.5	2254	54	97.0	4	23.0	59
chevrolet chevette	1	76	22.2	2035	52	85.0	4	29.0	195
peugeot 504	2	79	24.8	3190	71	141.0	4	27.2	299
oldsmobile cutlass salon brougham	1	79	22.2	3420	90	260.0	8	23.9	300
vw dasher (diesel)	2	80	23.7	2335	48	90.0	4	43.4	326
vw pickup	2	82	24.6	2130	52	97.0	4	44.0	394

```
In [27]:
           1 import pandas as pd
           2 import numpy as np
             def find_outlier(ds,col):
                  Q1 = ds[col].quantile(0.25)
                  Q3 = ds[col].quantile(0.75)
           6
           7
                  IQR = Q3 - Q1
           8
                  low_val = Q1 - (1.5*IQR)
           9
                  high_val = Q3 + (1.5*IQR)
                  ds = ds.loc[(ds[col]>low_val)&(ds[col]<high_val)]</pre>
          10
                  return ds
          11
          12
          13 df = pd.read_csv('auto-mpg.csv')
          14 find_outlier(df,"mpg")
```

Out[27]:

m	npg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0 18	8.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1 15	5.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
2 18	8.0	8	318.0	150	3436	11.0	70	1	plymouth satellite
3 16	6.0	8	304.0	150	3433	12.0	70	1	amc rebel sst
4 17	7.0	8	302.0	140	3449	10.5	70	1	ford torino
393 27	7.0	4	140.0	86	2790	15.6	82	1	ford mustang gl
394 44	4.0	4	97.0	52	2130	24.6	82	2	vw pickup
395 32	2.0	4	135.0	84	2295	11.6	82	1	dodge rampage
396 28	0.8	4	120.0	79	2625	18.6	82	1	ford ranger
397 31	1.0	4	119.0	82	2720	19.4	82	1	chevy s-10

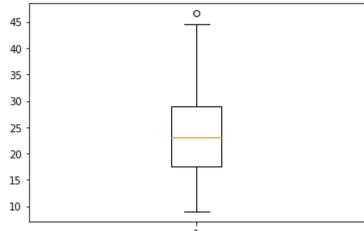
397 rows × 9 columns

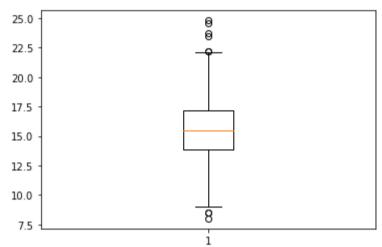
```
In [29]:
             import pandas as pd
           2 import numpy as np
             def find_outlier(ds,col):
                  Q1 = ds[col].quantile(0.25)
                  Q3 = ds[col].quantile(0.75)
           6
           7
                 IQR = Q3 - Q1
           8
                  low_val = Q1 - (1.5*IQR)
           9
                 high_val = Q3 + (1.5*IQR)
          10
                  ds = ds.loc[(ds[col]>low_val)&(ds[col]<high_val)]</pre>
                  return ds
          11
          12
          df = pd.read_csv('auto-mpg.csv')
          14 find_outlier(df,"acceleration")
```

Out[29]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
2	18.0	8	318.0	150	3436	11.0	70	1	plymouth satellite
3	16.0	8	304.0	150	3433	12.0	70	1	amc rebel sst
4	17.0	8	302.0	140	3449	10.5	70	1	ford torino
392	27.0	4	151.0	90	2950	17.3	82	1	chevrolet camaro
393	27.0	4	140.0	86	2790	15.6	82	1	ford mustang gl
395	32.0	4	135.0	84	2295	11.6	82	1	dodge rampage
396	28.0	4	120.0	79	2625	18.6	82	1	ford ranger
397	31.0	4	119.0	82	2720	19.4	82	1	chevy s-10

389 rows × 9 columns





```
In [ ]: 1
```