Unit-1 Introduction to Python Pandas

```
In [1]:
          1 pip install pandas
         Requirement already satisfied: pandas in c:\programdata\anaconda3\lib\site-packages (1.1.3)
         Requirement already satisfied: pytz>=2017.2 in c:\programdata\anaconda3\lib\site-packages (from pandas) (2020.1)
         Requirement already satisfied: python-dateutil>=2.7.3 in c:\programdata\anaconda3\lib\site-packages (from pandas) (2.8.
         1)
         Requirement already satisfied: numpy>=1.15.4 in c:\programdata\anaconda3\lib\site-packages (from pandas) (1.19.2)
         Requirement already satisfied: six>=1.5 in c:\programdata\anaconda3\lib\site-packages (from python-dateutil>=2.7.3->pan
         das) (1.15.0)
         Note: you may need to restart the kernel to use updated packages.
 In [2]:
           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 [3]:
           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
          15 | # dtype: int64
         0
              1
              2
         1
              3
         dtype: int64
In [10]:
           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
              1
               2
         1
         dtype: int64
         3
```

```
In [6]:
          1 # For another data type
          2 import pandas as pd
          3 \mid 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
             2.0
        1
             3.0
        2
        dtype: float64
        1.0
        2.0
        3.0
In [7]:
          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
             2.0
        2
             3.0
        dtype: float64
        1.0
        2.0
        3.0
In [8]:
         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
        1
             2
        dtype: object
        1.0
        2
        а
In [9]:
         1 | import pandas as pd
          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
        TypeError
                                                   Traceback (most recent call last)
        <ipython-input-9-b86b612605b6> 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
            298
                             elif isinstance(data, (set, frozenset)):
                                 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 [13]:
           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'])
           7 print(myvar['C'])
         Α
              1
              2
              3
         C
         dtype: int64
         1
         2
         3
In [14]:
          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'])
         Α
              [1, 2]
         В
                   2
         C
                   3
         dtype: object
         [1, 2]
         2
         3
          1 import pandas as pd
In [16]:
           2 \mid 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-16-349654d3951c> 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):
          --> 313
                                          raise ValueError(
                                              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 [17]:
           1 import pandas as pd
           2 \mid a = [1,2,3]
           3 myvar = pd.Series(a,index=['x','y','z'])
           4 print(myvar)
           6 # Output :
             # X
                     1
           8 | # y
                     2
           9 # z
                     3
          10 | # dtype: int64
              1
         Х
              2
         У
              3
         dtype: int64
```

```
In [20]:
           1 import pandas as pd
           calories = {'day1':420,'day2':380,'day3':390}
           3 myvar = pd.Series(calories)
             print(myvar)
           6 # Output :
           7
             # day1
                       420
             # day2
                       380
          9 # day3
                       390
          10 | # dtype: int64
                 420
         day1
                 380
         day2
         day3
                 390
         dtype: int64
In [23]:
          1 import pandas as pd
           calories = {'day1':420,'day2':380,'day3':390}
           3 myvar = pd.Series(calories,index=['x','y','z'])
           4 print(myvar)
           6
             # Output :
          7 | # x
                   NaN
           8 |# y
                   NaN
          9 | # z NaN
          10 | # dtype: float64
             NaN
         Х
             NaN
         У
             NaN
         dtype: float64
In [24]:
          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.
             # Output :
           6
           7
             # X
                         NaN
             # y
           8
                         NaN
          9 |# z
                         NaN
          10 # day1 420.0
          11 # dtype: float64
                   NaN
         Х
                   NaN
         У
                   NaN
         day1
                420.0
         dtype: float64
In [25]:
          1 | import pandas as pd
           calories = {'day1':420,'day2':380,'day3':390}
          myvar = pd.Series(calories,index=['day2','day1'])
           4 | print(myvar) # value must be same of particular key.
         day2
                 380
                 420
         day1
         dtype: int64
In [30]:
          1 \mid 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[30]: 0
         2
              3
```

DataFrame :(2D)

dtype: int64

• many rows many columns

380

390

1

40

45

```
In [31]:
           1 import pandas as pd
           2 data = {'calories':[420,380,390],'duration':[50,40,45]}
           3 df = pd.DataFrame(data)
             print(df)
           6
             #
                   calories duration
           7
             # 0
                       420
                                   50
           8
             # 1
                        380
                                   40
             # 2
                        390
                                   45
                      duration
            calories
         0
                 420
```

```
loc & iloc(integer location)

    loc : accepts labels as well as int

           • iloc: accepts only integer not a string
In [33]:
           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 [3]:
           1 | import pandas as pd
             data = {'calories':[420,380,390],'duration':[50,40,45]}
             df = pd.DataFrame(data)
             print(df)
             calories
                       duration
         0
                  420
                             50
         1
                  380
                             40
         2
                  390
                             45
 In [9]:
           1 import pandas as pd
           2 data = {'calories':[420,380,390],'duration':[50,40,45]}
           3 | df = pd.DataFrame(data,index=['day1','day2','day3'])
           4 | # print(df['calories'].loc[0]) # give key error because index is change.
             print(df['calories'].loc['day1']) # 420
           6
           7
              # Output :
           8
             #
                      calories duration
           9
             # day1
                           420
                                       50
          10 # day2
                           380
                                      40
          11 # day3
                           390
                                      45
         420
In [10]:
           1 import pandas as pd
           2 data = {'calories':[420,380,390],'duration':[50,40,45]}
           3 df = pd.DataFrame(data)
             print(df['calories'].iloc[0]) # 420
         420
In [12]:
          1 | import pandas as pd
           2 | data = {'calories':[420,380,390],'duration':[50,40,45]}
           3 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

Out[14]:

For CSV File.

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

398 rows × 9 columns

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

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

```
Non-Null Count Dtype
#
     Column
---
0
                   398 non-null
                                    float64
     mpg
1
     cylinders
                   398 non-null
                                    int64
                   398 non-null
2
     displacement
                                    float64
                   398 non-null
3
     horsepower
                                    object
                   398 non-null
4
     weight
                                    int64
                   398 non-null
5
     acceleration
                                    float64
     model year
                   398 non-null
                                    int64
 6
 7
     origin
                   398 non-null
                                    int64
     car name
                   398 non-null
                                    object
 8
dtypes: float64(3), int64(4), object(2)
```

In [17]: 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 [18]: 1 import pandas as pd
```

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

Out[18]:

	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

In [19]:

- 1 import pandas as pd
- 2 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[19]:

	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 [20]:

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

Out[20]:

	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 [21]:
- 1 import pandas as pd
- df = pd.read_csv('auto-mpg.csv') # converting into dataframe
- 3 df.tail(10)

Out[21]:

	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 l
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[22]:

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

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

- df = pd.read_csv('auto-mpg.csv') # converting into dataframe
- 3 df.loc[[34,56]] # for particular row show then we pass list

Out[24]:

	mpg	cylinders	displacement	norsepower	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[27]: 34 16.0 56 26.0

Name: mpg, dtype: float64

Out[29]:

	mpg	displacement
34	16.0	225.0
56	26.0	91.0

```
In [31]:
           1 import pandas as pd
           2 df = pd.read_csv('auto-mpg.csv')
           3 df[['mpg','cylinders']].loc[0:20]
Out[31]:
             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
                         4
          15 22.0
                         6
          16 18.0
                         6
          17 21.0
                         6
          18 27.0
          19 26.0
                         4
          20 25.0
In [32]:
          1 import pandas as pd
           2 df = pd.read_csv('auto-mpg.csv')
           3 df.shape # (398, 9)
```

```
In [36]:
           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)
             354
         --> 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:
         KeyError
                                                    Traceback (most recent call last)
         <ipython-input-36-83f047b6aa65> in <module>
               1 import pandas as pd
               2 df = pd.read_csv('auto-mpg.csv')
         ----> 3 df.loc[-1]
         C:\ProgramData\Anaconda3\lib\site-packages\pandas\core\indexing.py in __getitem__(self, key)
             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)
                     def _get_label(self, label, axis: int):
            1057
            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)
                              loc, new_index = self.index.get_loc_level(key, drop_level=drop_level)
             3489
             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 [37]:
           1 import pandas as pd
           2 | df = pd.read_csv('auto-mpg.csv')
           3 | df.loc[:-1] # only give column name.
Out[37]:
            mpg cylinders displacement horsepower weight acceleration model year origin car name
           1 import pandas as pd
In [42]:
             df = pd.read_csv('auto-mpg.csv')
             print(df.loc[:-1])
             # Empty DataFrame
           6 # Columns: [mpg, 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[44]:

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

Statistics All Operations:

Out[45]:

	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[1]:

	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[11]:

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

Out[3]:

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

Out[4]:

	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[5]:

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

Out[7]:

```
mpg cylinders displacement
                                              weight acceleration model year origin
       6.000000
                                6.000000
                                            6.000000
                                                         6.000000
                       6.0
                                                                          6.0
                                                                                  6.0
count
mean 16.500000
                       8.0
                             335.000000 3642.666667
                                                         11.166667
                                                                         70.0
                                                                                  1.0
       1.378405
                       0.0
                              49.363954
                                          355.980149
                                                         0.816497
                                                                          0.0
                                                                                  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
 50% 16.500000
                       8.0
                             312.500000 3476.500000
                                                         11.250000
                                                                         70.0
                                                                                  1.0
                             342.000000 3645.750000
 75% 17.750000
                       8.0
                                                         11.875000
                                                                         70.0
                                                                                  1.0
 max 18.000000
                       8.0
                             429.000000 4341.000000
                                                         12.000000
                                                                         70.0
                                                                                  1.0
```

Out[12]:

	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[8]: 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 [9]: 1 import pandas as pd
import numpy as np
df = pd.read_csv('auto-mpg.csv')
df.describe(percentiles=[0.3,0.57,0.83])
```

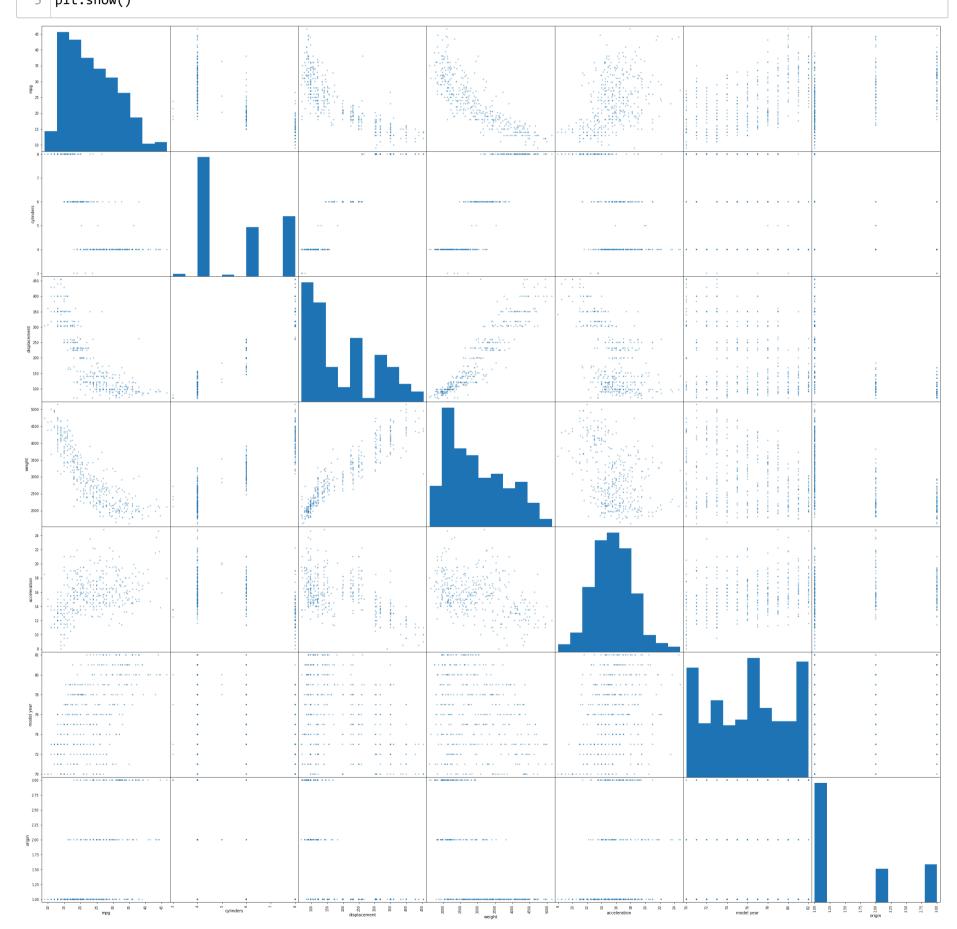
Out[9]:

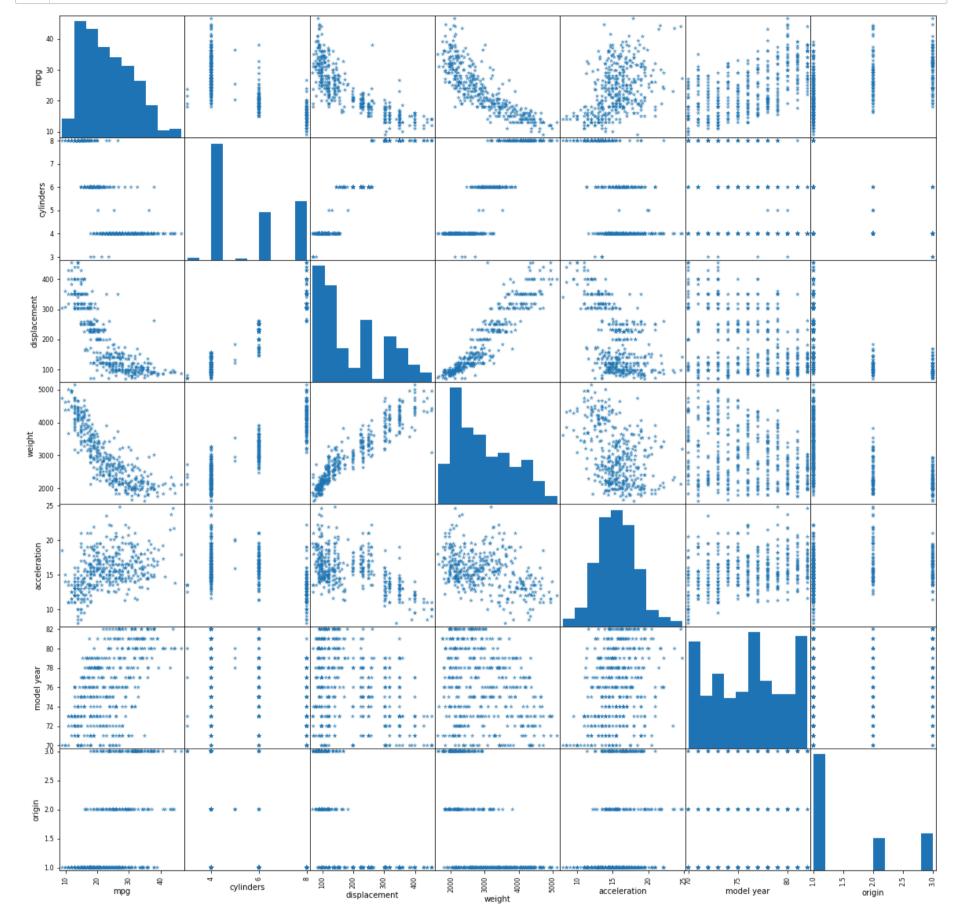
		mpg	cylinders	displacement	weight	acceleration	model year	origin
C	ount	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000	398.000000
n	nean	23.514573	5.454774	193.425879	2970.424623	15.568090	76.010050	1.572864
	std	7.815984	1.701004	104.269838	846.841774	2.757689	3.697627	0.802055
	min	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	7.0%	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

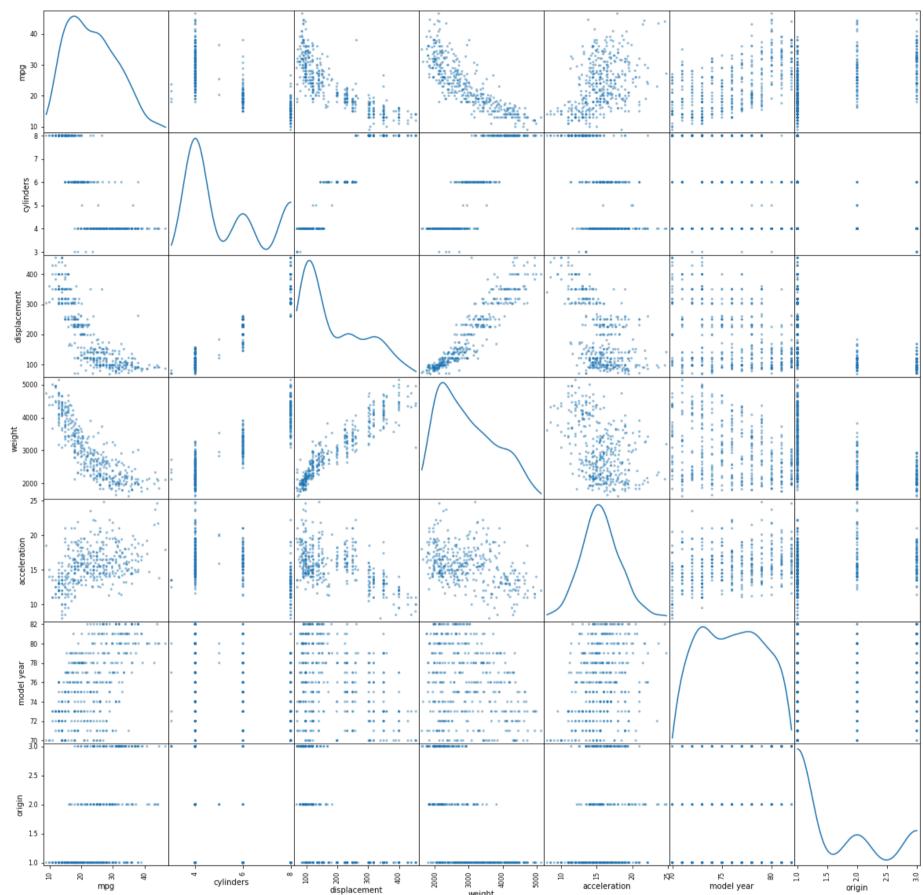
Corr:- corelation of cofficient

Out[13]:

	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

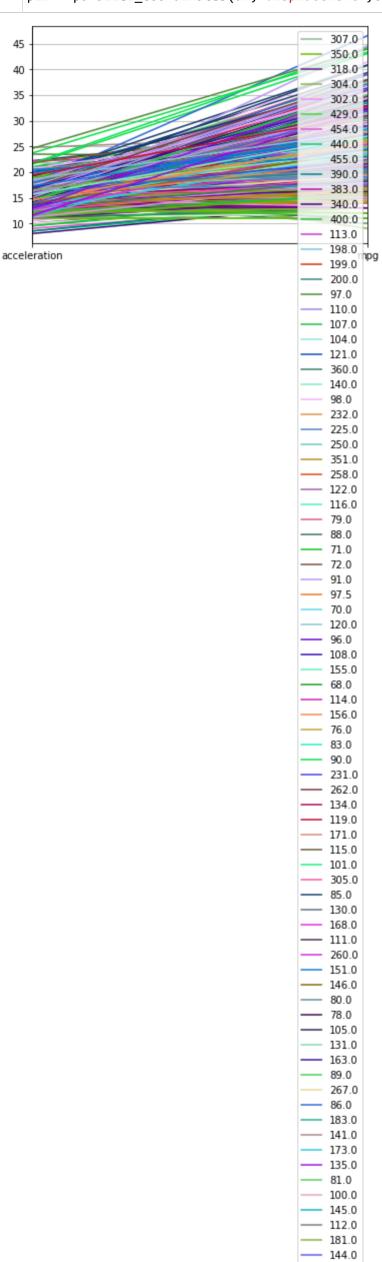


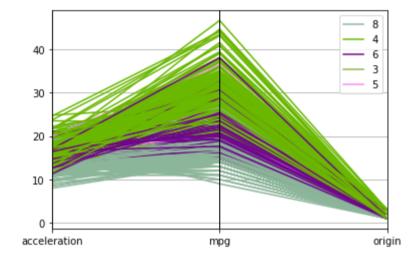




mpg

acceleration





Out[34]:

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

cylinders

```
      3
      0
      0
      1
      1
      0
      0
      0
      1
      0
      0
      1
      0
      0
      1
      0
      0

      4
      7
      13
      14
      11
      15
      12
      15
      14
      17
      12
      25
      21
      28

      5
      0
      0
      0
      0
      0
      0
      1
      1
      1
      0
      0

      6
      4
      8
      0
      8
      7
      12
      10
      5
      12
      6
      2
      7
      3

      8
      18
      7
      13
      20
      5
      6
      9
      8
      6
      10
      0
      1
      0
```

Data cleaning :-

Out[35]:

	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

```
Ch.-1_Sem-4_T-1_Yash - Jupyter Notebook
In [36]:
           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]})
           8 sales_data.isna() # boolean type
Out[36]:
             name region sales expense
             False
                    False False
                                  False
              True
                     True
                           True
                                   True
                    False False
                                  False
              True
             False
                    False
                          True
                                   True
                     True False
                                  False
           4 False
In [37]:
              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.isna().sum()
```

Out[37]: name 2 region 2 sales 2 2 expense dtype: int64

```
In [38]:
             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]})
           8 | sales_data.dropna()
```

Out[38]:

```
name region sales expense
0
                 10.0
                         50.0
           ma
```

```
In [39]:
              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.dropna(how="all") # how=all :- je row ma badha nan hoy to ej row kadhe.
```

Out[39]:

	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

```
In [40]:
             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[40]:

```
name region sales expense
              10.0
                       50.0
         ma
```

```
In [41]:
           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],
           7
                                          'expense':[50,np.nan,70,np.nan,90]})
           8 sales_data.dropna(subset=["sales"])
```

Out[41]:

```
name region sales expense
              10.0
                       50.0
         ma
 NaN
         mp
              30.0
                       70.0
              50.0
                       90.0
        NaN
   е
```

```
In [42]:
          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],
                                         'sales':[10,np.nan,30,np.nan,50],
                                          'expense':[50,np.nan,70,np.nan,90]})
           8 sales_data.dropna(subset=["sales","region"])
```

Out[42]:

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

In []: