Untitled185

August 15, 2024

1 INTERVIEW QUESTIONS

```
[5]: import pandas as pd
 [7]: import pandas as pd
      # Sample DataFrames
      db_employee = pd.DataFrame({
          'employee_id': [1, 2, 3, 4],
          'name': ['Alice', 'Bob', 'Charlie', 'David'],
          'department_id': [10, 20, 30, 40],
          'salary': [30000,40000,50000,60000]
      })
      db_dept = pd.DataFrame({
          'id': [10, 20, 30, 40],
          'department_name': ['HR', 'IT', 'Finance', 'Marketing'],
          'salary': [30000,40000,50000,70000]
      })
      # Merge DataFrames
      df = db_employee.merge(db_dept, how='left', left_on='department_id',__

¬right_on='id')
      # Display the merged DataFrame
      print(df)
        employee_id
                        name
                               department_id
                                              salary_x
                                                        id department_name
                                                                             salary_y
                                                 30000
                                                                                30000
     0
                  1
                        Alice
                                          10
                                                        10
                                                                         HR
                  2
                         Bob
                                          20
                                                 40000
                                                        20
                                                                         IT
                                                                                40000
     1
     2
                  3
                     Charlie
                                          30
                                                 50000
                                                        30
                                                                                50000
                                                                    Finance
     3
                       David
                                          40
                                                 60000 40
                                                                                70000
                                                                  Marketing
[15]: df.groupby('department_id')['salary_x'].max().reset_index()
[15]:
         department_id
                        salary_x
      0
                    10
                           30000
```

```
40000
      1
                    20
      2
                            50000
                    30
      3
                    40
                            60000
[16]: ## what are the diffirence between the list and tuple
      both are the data structure in python list is the mutable tuple is imutable
      the item list can be change where as tuple can not be change
[19]: ## what is the indexing
      \textit{\#\#indexing is the escending item from the data struce li}
      list1=['a','b','c','d']
      list1[1:3]
[19]: ['b', 'c']
 []: ## if and else statement
      ##if statement allows us to execute based on subconditions if the statement is \Box
       satified its execute oterwise it not execute
      #the else statement can be used with if statement to execute the block of code,
       →when the if condition is not satisfied
[65]: | ## we want to check weather the person is capable of voting or not
      age = 20
      if age < 18:
          print("you con't vote")
      else:
          print("you can vote")
     you can vote
[21]: ## how do we wright a function
      ##biggining of line we wright def ofter we put the instance and give the
       \hookrightarrow paranthesis
      ##inside the parathesis give parameter leave it no problem and give the colon:
      def simple():
          print("my first function")
[22]: ##cal the function
      simple()
     my first function
 []: ## what are the libraries used in python.
```

[]: | ## how is numpy and pandas used in analytics

numpy, pandas, seabon, matplotlib, plotly.sklearn

```
numpy is used for numerical analysis and linear algebra operation and matrix
       →operation
      pandas is used for data wrangling, data profiling, data analysis.
 []: ## what is the linear regression and assumption of linear regression
      Linear regression is used to predict the numerical and continuous values we fit
       →the model with equitiony=mx+c
     minimize the sum of squire errors
 []: ## What is the life cycles of data science projects
      understand the bussiness requirement from clints
      data collection from the various sources
      data analysis such as EDA data preparation, data cleaning
      feature engineering and feature sellection
      hyper parameter tuning
      model building and training data
      testing data and devolopment
 []: ## what is the data warehouse why do we need it
      data warehouse are used store the data for analytics
      OLAP AND REAL TIME TRANSACTION
[23]: | ##. Write a program to demonstrate different number data types in Python.
      a = 10
      b=1.5
      c=2.05i
[24]: ## delet one item from list
      list=['v','i','k','a','s']
      del list[2]
[25]: list
[25]: ['v', 'i', 'a', 's']
[28]: ## delet multiple item from list
      del list[1:5]
[29]: list
[29]: ['v']
[30]: ##delet entire list
      del list
```

<class 'list'>

[31]: print(list)

```
[34]: ##. Write a program to demonstrate working with tuples in python
      ## empty tuple
      tuple=()
      print(tuple)
[35]: ## tuple having integer
      tuple=(1,2,3)
      print(tuple)
     (1, 2, 3)
[36]: ## tuple with mixed data type
      tuple=(1,"hello",3.5)
      print(tuple)
     (1, 'hello', 3.5)
[37]: ## nested tuple
      tuple = ("mouse", [2,3,4], (1,5,7))
      print(tuple)
     ('mouse', [2, 3, 4], (1, 5, 7))
[38]: ##. Write a python program to find largest of three numbers
      num1=10
      n_{11}m_{2}=14
      num3=12
      if(num1>=num2)and(num1>=num3):
          lagest=num1
      elif(num2>=num1)and(num2>=num3):
          largest=num2
      else:
          largest=num3
      print("print lagest number is",largest)
     print lagest number is 14
[46]: | #. Write a Python program to construct the following pattern, using a
      #nested for loop
      n = 5
      # First Part: Increasing Stars
      for i in range(n):
          for j in range(i):
              print('* ', end="")
          print("")
      # Second Part: Decreasing Stars
```

```
for i in range(n, 0, -1):
          for j in range(i):
              print('* ', end="")
          print("")
[49]: | ##Python program to check if a string is palindrome or not
      def ispalindrome(s):
          return s == s[::-1]
      s = "malayalam"
      ans = ispalindrome(s)
      if ans:
         print('yes')
      else:
          print('no')
     yes
 []: ## what are the class and object in python
      class is a blue print and object is houses
      cal the class creat an object
[51]: class human:
          name=None
          age=None
          def get_name(self):
              print("enter the name")
              self.name=input()
          def get_age(self):
              print("enter the age")
              self.age=input()
          def put_name(self):
              print("your name is",self.name)
          def put_age(self):
              print("your age is",self.age)
```

```
[52]: person1=human()
[53]: person1.get_name()
     enter the name
     vikas
[54]: person1.get_age()
     enter the age
     30
[55]: person1.put_age(),person1.put_name()
     your age is 30
     your name is vikas
[55]: (None, None)
 []: ## what do we understand by __init__()methode in python, give an example of it.
      --init__ is the special methode in python classes
      it is the constructor methode of the python classes
      __init is when ever object of the class is constructor
      it is used to initialize the variables
[66]: class student:
          def __init__(self,name,age,branch):
              self.name=name
              self.age=age
              self.branch=branch
          def print_student(self):
              print("name", self.name)
              print("age", self.age)
              print("branch",self.branch)
[67]: student1 = student("vikas",30,"engineering")
[68]: student1.print_student()
     name vikas
     age 30
     branch engineering
[70]: | ## what do you understand by inherritence in python, give example of it
      ##one class is inheriting the properties of another class
      class fruit:
          def __init__(self):
              print('IAM A FRUIT')
      class citrus(fruit):
```

```
def __init__(self):
              super().__init__()
              print("iam a citrus")
      lemon = citrus()
     IAM A FRUIT
     iam a citrus
 []: ## whst is the numoy how can you create 1darray 2d array
      numpy is widly used python library and linear algebra and used to performing
       →mathemetical & logical functions
[71]: import numpy as np
[72]: a=np.array([1,2,3])
[73]: a
[73]: array([1, 2, 3])
[74]: b=np.array([[1,2,3],[4,5,6]])
[75]: b
[75]: array([[1, 2, 3],
             [4, 5, 6]])
[76]: ## how can we initialize 5*5 numpy array comparising of all zeros
      a=np.zeros((5,5))
[77]: a
[77]: array([[0., 0., 0., 0., 0.],
             [0., 0., 0., 0., 0.]
             [0., 0., 0., 0., 0.]
             [0., 0., 0., 0., 0.]
             [0., 0., 0., 0., 0.]])
[78]: ## add the two numpy array
      a=np.array([1,2,3])
      b=np.array([4,5,6])
      np.sum((a,b),axis=0)
[78]: array([5, 7, 9])
[79]: a=np.array([1,2,3])
      b=np.array([4,5,6])
```

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np.sum((a,b),axis=1)
[79]: array([ 6, 15])
[86]: ## in numpy array find the 2lagest values
      x=np.array([12,43,2,100,54,5,68])
[87]: x
[87]: array([ 12, 43,
                         2, 100, 54,
                                        5,
                                           68])
[88]: x[np.argsort(x)[-2:][::1]]
[88]: array([ 68, 100])
[89]: import pandas as pd
      ## give an examples creating a dataframe list and dictionary.
      data=[1,2,3,4]
      df=pd.DataFrame(data)
[90]: df
[90]:
         0
      0 1
      1 2
      2 3
      3 4
[92]: | dict1 = {'fruit':['apple', 'mango', 'banana'], 'count':[10,20,30]}
      df=pd.DataFrame(dict1)
[93]: df
[93]:
          fruit count
      0
          apple
                    10
                    20
      1
         mango
      2 banana
                    30
[94]: | ## extract the sapel lenght greater then 5 amd sapel widthgreater then 3,
      iris =pd.read_csv('iris.csv')
[95]: iris.head()
[95]:
         sepal_length sepal_width petal_length petal_width species
      0
                  5.1
                               3.5
                                             1.4
                                                           0.2 setosa
                  4.9
                               3.0
                                             1.4
                                                           0.2 setosa
      1
      2
                  4.7
                               3.2
                                             1.3
                                                           0.2 setosa
      3
                  4.6
                               3.1
                                             1.5
                                                           0.2 setosa
```

4 5.0 3.6 1.4 0.2 setosa

[96]: iris[(iris['sepal_length']>5)&(iris['sepal_width']>3)]

[96]:	sepal_length	sepal_width	petal_length	petal_width	species	
0	5.1	3.5	1.4	0.2	setosa	
5	5.4	3.9	1.7	0.4	setosa	
10	5.4	3.7	1.5	0.2	setosa	
14	5.8	4.0	1.2	0.2	setosa	
15	5.7	4.4	1.5	0.4	setosa	
16	5.4	3.9	1.3	0.4	setosa	
17	5.1	3.5	1.4	0.3	setosa	
18	5.7	3.8	1.7	0.3	setosa	
19	5.1	3.8	1.5	0.3	setosa	
20	5.4	3.4	1.7	0.2	setosa	
21	5.1	3.7	1.5	0.4	setosa	
23	5.1	3.3	1.7	0.5	setosa	
27	5.2	3.5	1.5	0.2	setosa	
28	5.2	3.4	1.4	0.2	setosa	
31	5.4	3.4	1.5	0.4	setosa	
32	5.2	4.1	1.5	0.1	setosa	
33	5.5	4.2	1.4	0.2	setosa	
36	5.5	3.5	1.3	0.2	setosa	
39	5.1	3.4	1.5	0.2	setosa	
44	5.1	3.8	1.9	0.4	setosa	
46	5.1	3.8	1.6	0.2	setosa	
48	5.3	3.7	1.5	0.2	setosa	
50	7.0	3.2	4.7	1.4	versicolor	
51	6.4	3.2	4.5	1.5	versicolor	
52	6.9	3.1	4.9	1.5	versicolor	
56	6.3	3.3	4.7	1.6	versicolor	
65	6.7	3.1	4.4	1.4	versicolor	
70	5.9	3.2	4.8	1.8	versicolor	
85	6.0	3.4	4.5	1.6	versicolor	
86	6.7	3.1	4.7	1.5	versicolor	
100	6.3	3.3	6.0	2.5	virginica	
109	7.2	3.6	6.1	2.5	virginica	
110	6.5	3.2	5.1	2.0	virginica	
115	6.4	3.2	5.3	2.3	virginica	
117	7.7	3.8	6.7	2.2	virginica	
120	6.9	3.2	5.7	2.3	virginica	
124	6.7	3.3	5.7	2.1	virginica	
125	7.2	3.2	6.0	1.8	virginica	
131	7.9	3.8	6.4	2.0	virginica	
136	6.3	3.4	5.6	2.4	virginica	
137	6.4	3.1	5.5	1.8	virginica	
139	6.9	3.1	5.4	2.1	virginica	

```
140
                     6.7
                                   3.1
                                                 5.6
                                                              2.4
                                                                     virginica
       141
                     6.9
                                   3.1
                                                 5.1
                                                              2.3
                                                                     virginica
       143
                     6.8
                                   3.2
                                                 5.9
                                                                     virginica
                                                              2.3
                     6.7
                                   3.3
                                                 5.7
       144
                                                              2.5
                                                                     virginica
       148
                     6.2
                                   3.4
                                                 5.4
                                                              2.3
                                                                     virginica
 [97]: ## how can you intruduce the Nan values in the first ten rows sapel length and
        ⇔sapel width
       iris.head()
 [97]:
          sepal_length sepal_width petal_length petal_width species
                                               1.4
       0
                   5.1
                                 3.5
                                                            0.2 setosa
       1
                   4.9
                                 3.0
                                               1.4
                                                            0.2 setosa
       2
                   4.7
                                3.2
                                               1.3
                                                            0.2 setosa
       3
                   4.6
                                 3.1
                                               1.5
                                                            0.2 setosa
                   5.0
                                               1.4
       4
                                 3.6
                                                            0.2 setosa
 [98]: iris1=iris
 [99]: iris1.iloc[0:10,1]=np.NAN
[100]: iris1.head()
[100]:
          sepal_length sepal_width petal_length petal_width species
                   5.1
       0
                                NaN
                                               1.4
                                                            0.2 setosa
       1
                   4.9
                                NaN
                                               1.4
                                                            0.2 setosa
                   4.7
                                NaN
                                               1.3
                                                            0.2 setosa
                   4.6
       3
                                NaN
                                               1.5
                                                            0.2 setosa
                   5.0
                                NaN
                                               1.4
                                                            0.2 setosa
[101]: iris1.iloc[0:10,2]=np.NAN
[102]: iris1.head()
[102]:
          sepal_length sepal_width petal_length petal_width species
                   5.1
                                                            0.2 setosa
       0
                                 NaN
                                               NaN
       1
                   4.9
                                 NaN
                                               NaN
                                                            0.2 setosa
       2
                   4.7
                                NaN
                                               NaN
                                                            0.2 setosa
                   4.6
       3
                                NaN
                                               NaN
                                                            0.2 setosa
                   5.0
                                NaN
                                               NaN
                                                            0.2 setosa
[104]: ## how can i count NAN values
       iris1.isna().sum()
[104]: sepal length
                        0
                       10
       sepal_width
       petal_length
                       10
       petal_width
                        0
```

species 0 dtype: int64

[121]: ## how can i read the file ## "r"represent opening the file read mode f=open("C:/Users/Vikas/Desktop/iris.csv","r")

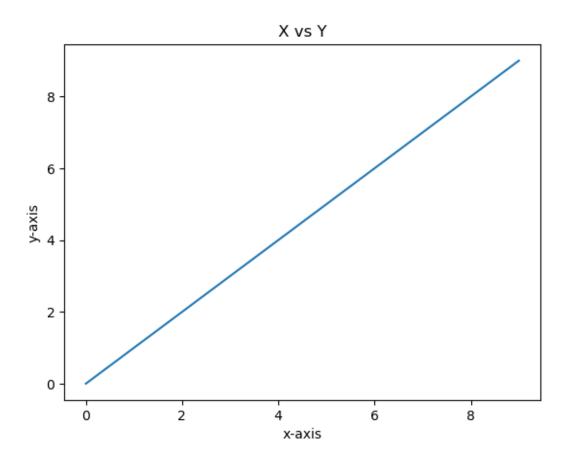
[122]: f.read()

[122]: 'sepal_length, sepal_width, petal_length, petal_width, species\n5.1,3.5,1.4,0.2, seto $sa\n4.9,3.0,1.4,0.2,setosa\n4.7,3.2,1.3,0.2,setosa\n4.6,3.1,1.5,0.2,setosa\n5.0,$ 3.6,1.4,0.2, setosa n5.4,3.9,1.7,0.4, setosa n4.6,3.4,1.4,0.3, setosa n5.0,3.4,1.5, $sa^1.6,0.2,setosa^1.4,0.1,setosa^1$ $4.0,1.2,0.2,setosa\n5.7,4.4,1.5,0.4,setosa\n5.4,3.9,1.3,0.4,setosa\n5.1,3.5,1.4,$ $sa\n5.1,3.7,1.5,0.4,setosa\n4.6,3.6,1.0,0.2,setosa\n5.1,3.3,1.7,0.5,setosa\n4.8,$ 3.4,1.9,0.2, setosa n5.0,3.0,1.6,0.2, setosa n5.0,3.4,1.6,0.4, setosa n5.2,3.5,1.5,0.2, setosa n5.2, 3.4, 1.4, 0.2, setosa n4.7, 3.2, 1.6, 0.2, setosa n4.8, 3.1, 1.6, 0.2, setosa n5.2, 3.4, 1.4, 0.2, setosa n5.2, 3.4, 1.4, 0.2, setosa n5.2, 3.4, 1.4, 0.2, setosa n6.2, 1.6, 0.2, $sa\n5.4,3.4,1.5,0.4,setosa\n5.2,4.1,1.5,0.1,setosa\n5.5,4.2,1.4,0.2,setosa\n4.9,$ $3.1,1.5,0.1,setosa\n5.0,3.2,1.2,0.2,setosa\n5.5,3.5,1.3,0.2,setosa\n4.9,3.1,1.5,$ 0.1, setosa n4.4, 3.0, 1.3, 0.2, setosa n5.1, 3.4, 1.5, 0.2, setosa n5.0, 3.5, 1.3, 0.3, setosa n6.1, 3.4, 1.5, 0.2, setosa n6.1, 3.5, 1.3, 0.3, $sa\n4.5,2.3,1.3,0.3,setosa\n4.4,3.2,1.3,0.2,setosa\n5.0,3.5,1.6,0.6,setosa\n5.1,$ 3.8,1.9,0.4, setosa n4.8,3.0,1.4,0.3, setosa n5.1,3.8,1.6,0.2, setosa n4.6,3.2,1.4,0.2, setosa n5.3, 3.7, 1.5, 0.2, setosa n5.0, 3.3, 1.4, 0.2, setosa n7.0, 3.2, 4.7, 1.4, versicolor n6.4, 3.2, 4.5, 1.5, versicolor n6.9, 3.1, 4.9, 1.5, versicolor n5.5, 2.3, 4.0, 1.3, $versicolor \n6.5, 2.8, 4.6, 1.5, versicolor \n5.7, 2.8, 4.5, 1.3, versicolor \n6.3, 3.3, 4.7,$ $1.6, versicolor \n 4.9, 2.4, 3.3, 1.0, versicolor \n 6.6, 2.9, 4.6, 1.3, versicolor \n 5.2, 2.7,$ $3.9,1.4, versicolor \n 5.0,2.0,3.5,1.0, versicolor \n 5.9,3.0,4.2,1.5, versicolor \n 6.0,$ $2.2,4.0,1.0, \text{versicolor} \ 1,2.9,4.7,1.4, \text{versicolor} \ 5.6,2.9,3.6,1.3, \text{versicolor} \ 1$ 6.7,3.1,4.4,1.4, versicolor\n5.6,3.0,4.5,1.5, versicolor\n5.8,2.7,4.1,1.0, versicol $or\n6.2,2.2,4.5,1.5,versicolor\n5.6,2.5,3.9,1.1,versicolor\n5.9,3.2,4.8,1.8,versicol$ $icolor \ n6.1, 2.8, 4.0, 1.3, versicolor \ n6.3, 2.5, 4.9, 1.5, versicolor \ n6.1, 2.8, 4.7, 1.2,$ $versicolor \n6.4, 2.9, 4.3, 1.3, versicolor \n6.6, 3.0, 4.4, 1.4, versicolor \n6.8, 2.8, 4.8,$ $1.4, versicolor \ 6.7, 3.0, 5.0, 1.7, versicolor \ 6.0, 2.9, 4.5, 1.5, versicolor \ 7, 2.6,$ 3.5,1.0, versicolor n5.5,2.4,3.8,1.1, versicolor n5.5,2.4,3.7,1.0, versicolor n5.8,2.7,3.9,1.2, versicolor $\n5.4,3.0,4.5,1.5,$ versicolor \n 6.0,3.4,4.5,1.6, versicolorn6.7,3.1,4.7,1.5, versicolorn6.3,2.3,4.4,1.3, versicol $or\n5.6,3.0,4.1,1.3,versicolor\n5.5,2.5,4.0,1.3,versicolor\n5.5,2.6,4.4,1.2,vers$ $icolor \ n6.1, 3.0, 4.6, 1.4, versicolor \ n5.8, 2.6, 4.0, 1.2, versicolor \ n5.0, 2.3, 3.3, 1.0,$ versicolor\n5.6,2.7,4.2,1.3,versicolor\n5.7,3.0,4.2,1.2,versicolor\n5.7,2.9,4.2, $1.3, versicolor \n 6.2, 2.9, 4.3, 1.3, versicolor \n 5.1, 2.5, 3.0, 1.1, versicolor \n 5.7, 2.8,$ $4.1, 1.3, \text{versicolor} \land 6.3, 3.3, 6.0, 2.5, \text{virginica} \land 5.8, 2.7, 5.1, 1.9, \text{virginica} \land 7.1, 3.$ 0,5.9,2.1, virginica n6.3,2.9,5.6,1.8, virginica n6.5,3.0,5.8,2.2, virginica n7.6,3 $.0,6.6,2.1,virginica\n4.9,2.5,4.5,1.7,virginica\n7.3,2.9,6.3,1.8,virginica\n6.7,$ $2.5, 5.8, 1.8, \text{virginica} \ n7.2, 3.6, 6.1, 2.5, \text{virginica} \ n6.5, 3.2, 5.1, 2.0, \text{virginica} \ n6.4$,2.7,5.3,1.9,virginica n6.8,3.0,5.5,2.1,virginica n5.7,2.5,5.0,2.0,virginica n5.8,2.8,5.1,2.4,virginica\n6.4,3.2,5.3,2.3,virginica\n6.5,3.0,5.5,1.8,virginica\n7

 $.7,3.8,6.7,2.2, \text{virginica} \ n7.7,2.6,6.9,2.3, \text{virginica} \ n6.0,2.2,5.0,1.5, \text{virginica} \ n6.9,3.2,5.7,2.3, \text{virginica} \ n5.6,2.8,4.9,2.0, \text{virginica} \ n7.7,2.8,6.7,2.0, \text{virginica} \ n6.3,2.7,4.9,1.8, \text{virginica} \ n6.7,3.3,5.7,2.1, \text{virginica} \ n7.2,3.2,6.0,1.8, \text{virginica} \ n6.2,2.8,4.8,1.8, \text{virginica} \ n6.1,3.0,4.9,1.8, \text{virginica} \ n6.4,2.8,5.6,2.1, \text{virginica} \ n7.2,3.0,5.8,1.6, \text{virginica} \ n7.4,2.8,6.1,1.9, \text{virginica} \ n7.9,3.8,6.4,2.0, \text{virginica} \ n6.4,2.8,5.6,2.2, \text{virginica} \ n6.3,2.8,5.1,1.5, \text{virginica} \ n6.1,2.6,5.6,1.4, \text{virginica} \ n7.7,3.0,6.1,2.3, \text{virginica} \ n6.3,3.4,5.6,2.4, \text{virginica} \ n6.4,3.1,5.5,1.8, \text{virginica} \ n6.0,3.0,4.8,1.8, \text{virginica} \ n6.9,3.1,5.4,2.1, \text{virginica} \ n6.7,3.1,5.6,2.4, \text{virginica} \ n6.9,3.1,5.1,2.3, \text{virginica} \ n5.8,2.7,5.1,1.9, \text{virginica} \ n6.8,3.2,5.9,2.3, \text{virginica} \ n6.7,3.3,5.7,2.5, \text{virginica} \ n6.7,3.0,5.2,2.3, \text{virginica} \ n6.3,2.5,5.0,1.9, \text{virginica} \ n6.5,3.0,5.2,2.0, \text{virginica} \ n6.2,3.4,5.4,2.3, \text{virginica} \ n5.9,3.0,5.1,1.8, \text{virginica} \ n6.5,3.0,5.2,2.0, \text{virginica} \ n6.2,3.4,5.4,2.3, \text{virginica} \ n5.9,3.0,5.1,1.8, \text{virginica} \ n6.5,3.0,5.2,2.0, \text{virginica} \ n6.2,3.4,5.4,2.3, \text{virginica} \ n5.9,3.0,5.1,1.8, \text{virginica} \ n6.1,2.6,2.4,2.3, \text{virginica} \ n6.2,3.2,2.0,2.3, \text{virginica} \ n6.2,3.2,2.3, \text{virginica} \ n6.3,2.5,3.0,5.1,1.8, \text{virginica} \ n6.2,3.2,2.3, \text{virginica} \ n6.3,2.5,3.0,5.1,1.8, \text{virginica} \ n6.2,3.2,2.3, \text{virginica} \ n6.2,3.2,2.3, \text{virginica} \ n6.3,2.3,2.5,3.0,5.1,1.8, \text{virginica} \ n6.2,3.2,2.3, \text{virginica} \ n6.2,3.2,2.3, \text{virginica} \ n6.3,2.3,2.5,3.0,5.1,1.8, \text{virginica} \ n6.2,3.2,2.3, \text$

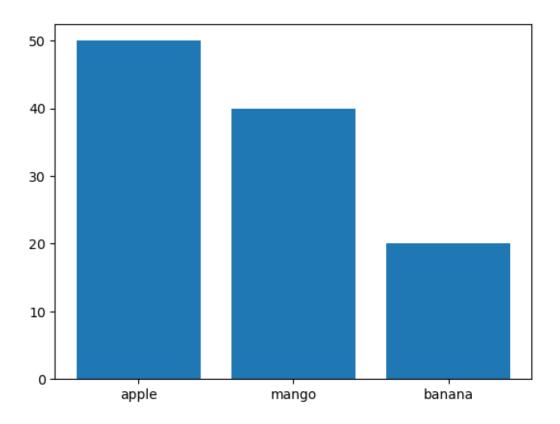
```
[126]: ## use lambdafunction print 10 numbers
       x=lambda a: a+10
[128]: x(100)
[128]: 110
[129]: ## below mentioned points used create line plot
       x=np.arange(0,10,1)
[130]: x
[130]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
[131]: y=np.arange(0,10,1)
[132]: y
[132]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
[133]: import matplotlib.pyplot as plt
[136]: plt.plot(x,y)
       plt.xlabel("x-axis")
       plt.ylabel("y-axis")
       plt.title("X vs Y")
       plt.show
```

[136]: <function matplotlib.pyplot.show(close=None, block=None)>



```
[141]: ## CREATE A SIMPLE BAR PLOT WHERE X-AXIS NAMES HAS FRUITS Y-AXIS COST OF FRUITS
data = {'apple':50,'mango':40,'banana':20}
names=list(data.keys())
values=list(data.values())
plt.bar(names,values)
```

[141]: <BarContainer object of 3 artists>



```
[142]: from random import shuffle
[143]: x=['mary', 'had', 'a', 'little', 'lamb']
[144]: shuffle(x)
[145]: x
[145]: ['had', 'mary', 'a', 'little', 'lamb']
[146]: ## find the length string
    a = "vikas"
    count=0
    for i in a:
        count=count+1
    print(count)

5
[147]: len("vikas")
```

```
[151]: ## replace all odd value with -1 the numbers in numpy array -1.
       arr=np.arange(0,10)
[152]: arr
[152]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
[153]: arr[arr%2==1] =-1
       ###0 divided by2 check remender wherever remender is equal to 1 changing to -1
[154]: arr
[154]: array([ 0, -1, 2, -1, 4, -1, 6, -1, 8, -1])
[155]: ## how can you get common items between two numpy array.
       a=np.array([1,2,3,4,5,6])
       b=np.array([2,3,5,6,7,9,0])
       np.intersect1d(a,b)
[155]: array([2, 3, 5, 6])
[156]: | ## how can you convert first character of each element ipandas series to | |
        \hookrightarrowuppercase
       ser=pd.Series(['marry', 'had', 'a', 'little', 'lamb'])
[157]: ser
[157]: 0
             marry
       1
               had
       2
       3
            little
              lamb
       dtype: object
[158]: ser.map(lambda x: x.title())
[158]: 0
             Marry
               Had
       1
                 Α
       3
            Little
              Lamb
       dtype: object
[159]: | ##how woild you caluclate the number of character in each word in series
       ser = pd.Series(['maryy', 'had', 'a', 'little', 'lamb'])
       ser.map(lambda a:len(a))
```

```
[159]: 0
       1
            3
       2
            1
       3
            6
       4
            4
       dtype: int64
[160]: ## in iris dataset change the col name sepel legth to s length
       iris
[160]:
                          sepal_width petal_length petal_width
            sepal_length
                                                                       species
                     5.1
                                   NaN
                                                  NaN
                                                               0.2
                                                                       setosa
       1
                     4.9
                                   NaN
                                                  NaN
                                                               0.2
                                                                        setosa
                     4.7
       2
                                   NaN
                                                  NaN
                                                               0.2
                                                                        setosa
                     4.6
       3
                                   NaN
                                                  NaN
                                                               0.2
                                                                       setosa
       4
                     5.0
                                   NaN
                                                               0.2
                                                  NaN
                                                                       setosa
       145
                     6.7
                                   3.0
                                                  5.2
                                                               2.3 virginica
       146
                     6.3
                                   2.5
                                                  5.0
                                                               1.9 virginica
       147
                     6.5
                                   3.0
                                                  5.2
                                                               2.0 virginica
                     6.2
                                   3.4
       148
                                                  5.4
                                                               2.3 virginica
       149
                     5.9
                                   3.0
                                                  5.1
                                                               1.8 virginica
       [150 rows x 5 columns]
[162]: | iris1=iris.rename(columns={'sepal_length':'s_legth'})
[164]: iris1.head()
[164]:
          s_legth sepal_width petal_length petal_width species
       0
              5.1
                            NaN
                                          NaN
                                                        0.2
                                                            setosa
       1
              4.9
                            NaN
                                          NaN
                                                        0.2 setosa
       2
              4.7
                            NaN
                                          NaN
                                                        0.2 setosa
       3
              4.6
                            NaN
                                          NaN
                                                        0.2 setosa
       4
              5.0
                            NaN
                                          NaN
                                                        0.2 setosa
  []: ## Build a lenear regression model on this Boston dataset where the independent
        ⇔vareable is 'rm' dependent varieble is 'medv'
       ## the train and test set need to be 80:20
[165]: ## Loading the required package
       import pandas as pd
[166]: ## load the boston dataset
       boston = pd.read_csv('Boston1.csv')
[167]: boston.head()
```

```
[167]:
            crim
                        indus chas
                                                            dis rad tax ptratio \
                    zn
                                       nox
                                               rm
                                                    age
      0 0.00632 18.0
                         2.31
                                  0
                                     0.538
                                           6.575 65.2 4.0900
                                                                   1
                                                                      296
                                                                              15.3
      1 0.02731
                   0.0
                         7.07
                                  0 0.469
                                            6.421 78.9 4.9671
                                                                   2
                                                                      242
                                                                              17.8
      2 0.02729
                   0.0
                         7.07
                                  0 0.469
                                            7.185 61.1 4.9671
                                                                   2 242
                                                                              17.8
      3 0.03237
                   0.0
                         2.18
                                  0 0.458
                                            6.998 45.8 6.0622
                                                                      222
                                                                              18.7
                                                                   3
      4 0.06905
                   0.0
                         2.18
                                  0 0.458 7.147 54.2 6.0622
                                                                   3 222
                                                                              18.7
          black lstat
                        medv
      0 396.90
                 4.98
                        24.0
      1 396.90
                  9.14
                        21.6
      2 392.83
                  4.03
                        34.7
      3 394.63
                  2.94
                        33.4
      4 396.90
                  5.33 36.2
[168]: ## getting features from the dataset
      x= pd.DataFrame(boston['rm'])
      y=pd.DataFrame(boston['medv'])
[169]: ## split the train and test
      from sklearn.model_selection import train_test_split
      x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.20)
[171]: ## build the model
      from sklearn.linear_model import LinearRegression
      regressor=LinearRegression()
      regressor.fit(x_train,y_train)
[171]: LinearRegression()
[172]: ## Predicting the value
      y_pred = regressor.predict(x_test)
[175]: y_pred
[175]: array([[24.54588025],
              [22.84168803],
              [27.95426471],
              [31.05444418],
              [14.36605115],
              [21.62699782],
              [24.05637823],
              [17.5478143],
              [21.18282006],
              [30.89127684],
              [15.79829781],
              [19.83215707],
              [17.5478143],
              [36.4842907],
```

```
[24.25580498],
```

- [28.11743205],
- [18.45429953],
- [3.27067197],
- [26.03251603],
- [17.9466678],
- [23.61220047],
- [29.05111183],
- [27.61886517],
- [22.22527807],
- [19.22481197],
- [36.64745804],
- [26.340721],
- [23.10456874],
- [21.25533888],
- [24.63652878],
- [18.98912581],
- [33.64699193],
- [22.23434292],
- [26.10503484],
- [10.23247852],
- [19.93187045],
- [17.09457169],
- [23.27680093],
- [7.93000604],
- [23.39464401],
- [28.9242039],
- [25.36171696],
- [38.58733642],
- [18.85315303],
- [22.22527807],
- [20.72957745],
- [28.77010141],
- [35.47809209],
- [25.543014],
- [20.52108584],
- [29.34118711],
- [12.77063715],
- [19.97719471],
- [21.66325723],
- [21.01058787],
- [23.91134059],
- [22.19808351],
- [32.45949629],
- [23.18615241],
- [19.5602115],
- [19.69618429],

```
[21.01965272],
```

- [21.7448409],
- [23.25867123],
- [32.05157793],
- [28.28059939],
- [24.1198322],
- [21.52728445],
- [24.00198912],
- [6.72438069],
- [0:72400000]
- [14.24820807],
- [22.04398103],
- [30.15702381],
- [22.77823406],
- [31.6980487],
- [21.91707309],
- [26.7395745],
- [20.28539969],
- [27.22907653],
- [25.75150561],
- [20.10100001],
- [23.24054153],
- [18.45429953],
- [19.56927636],
- [19.02538522],
- [17.67472223],
- [18.24580792],
- [18.82595847],
- [21.66325723],
- [22.42470482],
- [23.5306168],
- [23.82069207],
- [31.39890857],
- [21.31879284],
- [25.1260308],
- [29.76723516],
- [20.18568631],
- [21.52728445],
- [20.88367994],
- [18.66279113],
- [24.44616688],
- [18.18235396],
- [16.59600481]])

2 TOP 20 FUNCTION IN PANDAS

```
[177]: ## Reading a file
      data =pd.read_csv('iris.csv')
[178]: ## get the top and bottom of dataset
      data.head()
[178]:
         sepal_length sepal_width petal_length petal_width species
      0
                  5.1
                               3.5
                                             1.4
                                                          0.2 setosa
      1
                  4.9
                               3.0
                                             1.4
                                                          0.2 setosa
      2
                  4.7
                               3.2
                                             1.3
                                                          0.2 setosa
                                             1.5
      3
                  4.6
                               3.1
                                                          0.2 setosa
                  5.0
                               3.6
                                             1.4
                                                          0.2 setosa
[179]: data.tail()
[179]:
           sepal_length sepal_width petal_length petal_width
                                                                   species
      145
                    6.7
                                 3.0
                                               5.2
                                                            2.3 virginica
      146
                    6.3
                                 2.5
                                               5.0
                                                            1.9 virginica
      147
                    6.5
                                 3.0
                                               5.2
                                                            2.0 virginica
      148
                    6.2
                                 3.4
                                               5.4
                                                            2.3 virginica
      149
                    5.9
                                 3.0
                                               5.1
                                                            1.8 virginica
[180]: | ## qet the complete information about all the columns in dataset
      data.info()
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 150 entries, 0 to 149
      Data columns (total 5 columns):
                         Non-Null Count Dtype
       #
          Column
                         _____
           sepal_length 150 non-null
       0
                                         float64
           sepal_width
                         150 non-null
                                         float64
       1
       2
           petal_length 150 non-null
                                         float64
                         150 non-null
       3
           petal_width
                                         float64
       4
           species
                         150 non-null
                                         object
      dtypes: float64(4), object(1)
      memory usage: 6.0+ KB
[181]: ## see the shape of the dataset
      data.shape
[181]: (150, 5)
[182]: data.size
[182]: 750
```

```
[183]: ## get the standard methematical analysis each columns
       data.describe()
[183]:
              sepal_length
                            sepal_width petal_length petal_width
                             150.000000
                                            150.000000
                150.000000
                                                         150.000000
       count
      mean
                  5.843333
                               3.054000
                                              3.758667
                                                           1.198667
       std
                  0.828066
                               0.433594
                                              1.764420
                                                           0.763161
      min
                  4.300000
                               2.000000
                                              1.000000
                                                           0.100000
       25%
                  5.100000
                               2.800000
                                              1.600000
                                                           0.300000
       50%
                  5.800000
                               3.000000
                                              4.350000
                                                           1.300000
       75%
                  6.400000
                               3.300000
                                              5.100000
                                                           1.800000
                               4.400000
                                             6.900000
      max
                  7.900000
                                                           2.500000
[184]: data.describe().T
                                                    ## T means "transpose" row to col
        ⇒col to row convert
[184]:
                     count
                                            std
                                                min 25%
                                                            50%
                                                                 75%
                                                                      max
                                mean
       sepal_length
                    150.0
                            5.843333
                                      0.828066
                                                4.3
                                                     5.1
                                                           5.80
                                                                 6.4
                                                                      7.9
                                                 2.0 2.8 3.00 3.3 4.4
                     150.0
       sepal_width
                            3.054000
                                      0.433594
       petal_length
                     150.0
                                                 1.0
                                                      1.6 4.35
                                                                 5.1
                                                                      6.9
                            3.758667
                                      1.764420
      petal width
                                      0.763161
                                                0.1 0.3 1.30 1.8 2.5
                     150.0
                            1.198667
[185]: ## find the no of distinct value in categorical col
       data.nunique()
[185]: sepal_length
                       35
       sepal_width
                       23
       petal_length
                       43
      petal_width
                       22
       species
                        3
       dtype: int64
[187]: ## how to find the missing values
       data.isnull().sum()
[187]: sepal_length
                       0
       sepal_width
                       0
       petal_length
                       0
       petal_width
                       0
       species
                       0
       dtype: int64
[188]: data.isnull().any()
[188]: sepal_length
                       False
       sepal_width
                       False
       petal_length
                       False
      petal_width
                       False
```

```
dtype: bool
[189]: ## find the column names
       data.columns
[189]: Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
              'species'],
             dtype='object')
[191]: ## get the nsmalest and nlargest values in column
       data.nsmallest(10, 'petal_length')
                         sepal_width petal_length petal_width species
[191]:
           sepal_length
                    4.6
                                  3.6
                                                1.0
       22
                                                              0.2
                                                                  setosa
                    4.3
                                  3.0
                                                1.1
       13
                                                              0.1 setosa
       14
                    5.8
                                  4.0
                                                1.2
                                                              0.2 setosa
       35
                    5.0
                                  3.2
                                                1.2
                                                              0.2
                                                                   setosa
       2
                    4.7
                                  3.2
                                                1.3
                                                              0.2 setosa
       16
                    5.4
                                  3.9
                                                1.3
                                                              0.4
                                                                   setosa
       36
                    5.5
                                  3.5
                                                1.3
                                                              0.2 setosa
                    4.4
                                  3.0
                                                1.3
                                                              0.2
                                                                   setosa
       38
                    5.0
                                                1.3
       40
                                  3.5
                                                              0.3
                                                                   setosa
       41
                    4.5
                                  2.3
                                                1.3
                                                              0.3
                                                                   setosa
[192]: data.nlargest(10, 'petal_length')
[192]:
            sepal_length sepal_width petal_length petal_width
                                                                      species
       118
                     7.7
                                   2.6
                                                 6.9
                                                               2.3 virginica
                     7.7
                                   3.8
       117
                                                 6.7
                                                               2.2 virginica
       122
                     7.7
                                   2.8
                                                 6.7
                                                               2.0 virginica
       105
                     7.6
                                   3.0
                                                 6.6
                                                               2.1 virginica
       131
                     7.9
                                   3.8
                                                 6.4
                                                               2.0 virginica
       107
                     7.3
                                   2.9
                                                 6.3
                                                               1.8 virginica
       109
                                   3.6
                     7.2
                                                 6.1
                                                               2.5 virginica
       130
                     7.4
                                   2.8
                                                 6.1
                                                               1.9 virginica
       135
                     7.7
                                   3.0
                                                 6.1
                                                               2.3 virginica
       100
                     6.3
                                   3.3
                                                 6.0
                                                               2.5 virginica
[194]: ## loc and iloc
       data.loc[1:6,['sepal_length','sepal_width']]
                                                                                       Ш
        ⇔## lloc 6 is inclussive
[194]:
          sepal_length sepal_width
                   4.9
                                 3.0
       1
                   4.7
       2
                                 3.2
       3
                   4.6
                                 3.1
                   5.0
                                 3.6
```

species

False

```
5
                   5.4
                                 3.9
       6
                   4.6
                                 3.4
[197]: data.iloc[1:6,[0,1]]
                                                               ## iloc 6 is exclussive
[197]:
          sepal_length sepal_width
                   4.9
                                 3.0
       1
       2
                   4.7
                                 3.2
       3
                   4.6
                                 3.1
       4
                   5.0
                                 3.6
       5
                   5.4
                                 3.9
  []: ### groupby it means we can group the perticular columns
       data[['age','fare']].groupby(data['pclass']).mean()
       data[['age', 'fare']].groupby(data['pclass']).sum()
[198]: ## sort values
       data.sort_values(by='sepal_length',ascending =False)
[198]:
            sepal_length sepal_width petal_length petal_width
                                                                       species
       131
                     7.9
                                   3.8
                                                  6.4
                                                               2.0 virginica
       135
                      7.7
                                   3.0
                                                  6.1
                                                               2.3 virginica
       122
                     7.7
                                   2.8
                                                  6.7
                                                               2.0 virginica
       117
                     7.7
                                   3.8
                                                               2.2 virginica
                                                  6.7
                     7.7
       118
                                   2.6
                                                  6.9
                                                               2.3 virginica
       . .
       41
                      4.5
                                   2.3
                                                  1.3
                                                               0.3
                                                                        setosa
       42
                      4.4
                                   3.2
                                                  1.3
                                                               0.2
                                                                        setosa
       38
                      4.4
                                   3.0
                                                  1.3
                                                               0.2
                                                                        setosa
                      4.4
                                   2.9
                                                               0.2
       8
                                                  1.4
                                                                        setosa
       13
                     4.3
                                   3.0
                                                  1.1
                                                               0.1
                                                                        setosa
       [150 rows x 5 columns]
[203]: ## query in dataframe###
                                                                                 Ш
        →[columns:rows] [1:2]
       data.query('sepal_length > 4.4')[:4]
[203]:
          sepal_length sepal_width petal_length petal_width species
                   5.1
                                 3.5
                                                1.4
                                                             0.2 setosa
                   4.9
                                 3.0
                                                1.4
                                                             0.2 setosa
       1
       2
                   4.7
                                 3.2
                                                1.3
                                                             0.2 setosa
       3
                   4.6
                                                1.5
                                                             0.2 setosa
                                 3.1
[204]: ## unique in dataframe
       data['sepal_length'].unique()
```

```
[204]: array([5.1, 4.9, 4.7, 4.6, 5., 5.4, 4.4, 4.8, 4.3, 5.8, 5.7, 5.2, 5.5,
              4.5, 5.3, 7., 6.4, 6.9, 6.5, 6.3, 6.6, 5.9, 6., 6.1, 5.6, 6.7,
              6.2, 6.8, 7.1, 7.6, 7.3, 7.2, 7.7, 7.4, 7.9
[206]: data.memory_usage()
[206]: Index
                        128
      sepal_length
                       1200
       sepal_width
                       1200
      petal_length
                       1200
      petal_width
                       1200
      species
                       1200
       dtype: int64
 []: ## what is the difference between the python array and list
       array in python can only contain element of same data type
       dtatype of array should be homogeneous.
       list in python can contain element ofdifferent datatype
       datatype of list it is heterogeneous
 []: ## what is the lambda in python
       lambda is anonymous function in python that can accept any number of ___
        ⇔orguments, can only have a single expression
       lambda function can be used either two way
[209]: mul = lambda a, b: a * b
       # Use the lambda function to multiply 2 and 5, then print the result
       print(mul(2, 5))
      10
[210]: def mywrapper(n):
           return lambda a : a * n
       mulfive=mywrapper(5)
       print(mulfive(2))
      10
 []: ## explain split(), join() function in python
       split can used split the string base don the delimeters.
       join the list of string based o the delimeters
[213]: string="this is a string"
       string_list = string.split(' ')
       print(string_list)
      ['this', 'is', 'a', 'string']
```

```
[214]: print(' '.join(string_list))
      this is a string
 []: ## what are the built in data types in python,
       Numbers:-they includs the integers, floating point numbers, and complex numbers.
              :-An order sequence of items is called list the element of list may u
       List
        ⇒belong to different datatype
       Tuple :-it is also an order sequence of element, unlikelist, immutable, means it ⊔
       ⇔canot be change
       String ;-A sequence of character is called string they are declared within.
       ⇔single, double quote " "
              :it is set of unique item, that are not in order,
       Set
       Dictionary :- Dictionary is a stored the value keys and values, each value can
       ⇔eccess though its keys.
       Boolean :-there are two boolean value True and False
[216]: class employees:
           def __init__(self,firstname,lastname,salary):
               self.firstname=firstname
              self.lastname=lastname
              self.salary=salary
               self.email=self.firstname + "." + self.lastname + "@gmail.com"
[219]: ## reverse list
       my_list = [2, 4, 6, 7, 8]
       reversed_list = my_list[::-1]
       print(reversed_list)
      [8, 7, 6, 4, 2]
[220]: import re
                                             ### re means regex
[221]: def add(a,b):
                                      ## here we used 2 line code
          return a+b
       print(add(4,5))
[222]: add=lambda a,b:a+b
                                 ## hre we used one line code
       print(add(4,5))
[223]: twice = lambda x:x*2
       print(twice(6))
```

```
[]: ## what is the OS modulu
       it is stands for operating system OS module provides various function for \Box
        →interacting
       import os in anaconda prompt
       os.name
       it gives 'nt'is os name
                    it means cwd 'current working dirrectory'
       os.getcwd()
       ## i want to change my dirrectory
       os.chdir('c://')
       os.getcwd()
       os.chdir('c://projects//')
       ## i want to make dirrectory
       os.mkdir('new')
       ## i want to remove the dirrectory
       os.rmdir('new')
       ## i want folder within folder
       os.makedirs('new//new2')
       ## i want to know the directory perticulat folder
       os.listdir()
       #3 i want clear my current screen
       clear=lambda:os.system('cls')
       clear()
       ## i want to know all dirrectory
       ## i want to know weather this dirrectory exist or not
       os.path.exists('c://')
[224]: | ## wap to read afile with multiple line& split data bylines and also split intou
       fb = open("C:\\Users\\Vikas\\Desktop\\iris.csv")
       data = fb.read()
       print(data)
      sepal_length, sepal_width, petal_length, petal_width, species
      5.1,3.5,1.4,0.2,setosa
      4.9,3.0,1.4,0.2,setosa
      4.7,3.2,1.3,0.2,setosa
      4.6,3.1,1.5,0.2,setosa
      5.0,3.6,1.4,0.2,setosa
      5.4,3.9,1.7,0.4,setosa
      4.6,3.4,1.4,0.3,setosa
      5.0,3.4,1.5,0.2,setosa
      4.4,2.9,1.4,0.2,setosa
      4.9,3.1,1.5,0.1,setosa
      5.4,3.7,1.5,0.2,setosa
      4.8,3.4,1.6,0.2,setosa
```

- 4.8,3.0,1.4,0.1,setosa
- 4.3,3.0,1.1,0.1,setosa
- 5.8,4.0,1.2,0.2,setosa
- 5.7,4.4,1.5,0.4,setosa
- 5.4,3.9,1.3,0.4,setosa
- 5.1,3.5,1.4,0.3,setosa
- 5.7,3.8,1.7,0.3,setosa
- 5.1,3.8,1.5,0.3,setosa
- 5.4,3.4,1.7,0.2,setosa
- 5.1,3.7,1.5,0.4,setosa
- 4.6,3.6,1.0,0.2,setosa
- 5.1,3.3,1.7,0.5,setosa
- 4.8,3.4,1.9,0.2,setosa
- 5.0,3.0,1.6,0.2,setosa
- 5.0,3.4,1.6,0.4,setosa
- 5.2,3.5,1.5,0.2,setosa
- 5.2,3.4,1.4,0.2,setosa
- 4.7,3.2,1.6,0.2,setosa
- 4.8,3.1,1.6,0.2,setosa
- 5.4,3.4,1.5,0.4,setosa
- 5.2,4.1,1.5,0.1,setosa
- 5.5,4.2,1.4,0.2,setosa
- 0.0,4.2,1.4,0.2,50050
- 4.9,3.1,1.5,0.1,setosa
- 5.0,3.2,1.2,0.2,setosa
- 5.5,3.5,1.3,0.2,setosa
- 4.9,3.1,1.5,0.1,setosa 4.4,3.0,1.3,0.2,setosa
- 5.1,3.4,1.5,0.2,setosa
- 5.0,3.5,1.3,0.3,setosa
- 4.5,2.3,1.3,0.3,setosa
- 4.4,3.2,1.3,0.2,setosa
- 5.0,3.5,1.6,0.6,setosa
- 5.1,3.8,1.9,0.4,setosa
- 4.8,3.0,1.4,0.3,setosa
- 5.1,3.8,1.6,0.2,setosa
- 4.6,3.2,1.4,0.2,setosa
- 1.0,0.2,1.1,0.2,60006
- 5.3,3.7,1.5,0.2,setosa
- 5.0,3.3,1.4,0.2,setosa
- 7.0, 3.2, 4.7, 1.4, versicolor
- 6.4, 3.2, 4.5, 1.5, versicolor
- 6.9,3.1,4.9,1.5,versicolor
- 5.5, 2.3, 4.0, 1.3, versicolor
- 6.5,2.8,4.6,1.5,versicolor
- 5.7,2.8,4.5,1.3, versicolor 6.3,3.3,4.7,1.6, versicolor
- 0.3,3.3,4.7,1.0,Versicolor
- 4.9,2.4,3.3,1.0,versicolor
- 6.6,2.9,4.6,1.3,versicolor
- 5.2,2.7,3.9,1.4, versicolor

5.0,2.0,3.5,1.0,versicolor 5.9,3.0,4.2,1.5, versicolor 6.0,2.2,4.0,1.0,versicolor 6.1,2.9,4.7,1.4, versicolor 5.6,2.9,3.6,1.3, versicolor 6.7,3.1,4.4,1.4, versicolor 5.6,3.0,4.5,1.5, versicolor 5.8,2.7,4.1,1.0, versicolor 6.2,2.2,4.5,1.5, versicolor 5.6,2.5,3.9,1.1,versicolor 5.9,3.2,4.8,1.8, versicolor 6.1,2.8,4.0,1.3, versicolor 6.3,2.5,4.9,1.5, versicolor 6.1,2.8,4.7,1.2, versicolor 6.4,2.9,4.3,1.3, versicolor 6.6,3.0,4.4,1.4, versicolor 6.8,2.8,4.8,1.4, versicolor 6.7,3.0,5.0,1.7,versicolor 6.0,2.9,4.5,1.5, versicolor 5.7,2.6,3.5,1.0, versicolor 5.5,2.4,3.8,1.1,versicolor 5.5,2.4,3.7,1.0, versicolor 5.8,2.7,3.9,1.2, versicolor 6.0,2.7,5.1,1.6,versicolor 5.4,3.0,4.5,1.5, versicolor 6.0,3.4,4.5,1.6,versicolor 6.7,3.1,4.7,1.5, versicolor 6.3,2.3,4.4,1.3, versicolor 5.6,3.0,4.1,1.3, versicolor 5.5,2.5,4.0,1.3, versicolor 5.5,2.6,4.4,1.2, versicolor 6.1,3.0,4.6,1.4, versicolor 5.8,2.6,4.0,1.2, versicolor 5.0,2.3,3.3,1.0,versicolor 5.6,2.7,4.2,1.3, versicolor 5.7,3.0,4.2,1.2, versicolor 5.7,2.9,4.2,1.3, versicolor 6.2,2.9,4.3,1.3, versicolor 5.1,2.5,3.0,1.1,versicolor 5.7,2.8,4.1,1.3, versicolor 6.3,3.3,6.0,2.5, virginica 5.8,2.7,5.1,1.9, virginica 7.1,3.0,5.9,2.1, virginica 6.3,2.9,5.6,1.8, virginica 6.5,3.0,5.8,2.2,virginica 7.6,3.0,6.6,2.1, virginica 4.9,2.5,4.5,1.7, virginica 7.3,2.9,6.3,1.8, virginica

```
6.7,2.5,5.8,1.8, virginica
7.2,3.6,6.1,2.5,virginica
6.5,3.2,5.1,2.0,virginica
6.4,2.7,5.3,1.9, virginica
6.8, 3.0, 5.5, 2.1, virginica
5.7,2.5,5.0,2.0, virginica
5.8,2.8,5.1,2.4, virginica
6.4,3.2,5.3,2.3,virginica
6.5,3.0,5.5,1.8, virginica
7.7,3.8,6.7,2.2,virginica
7.7,2.6,6.9,2.3, virginica
6.0,2.2,5.0,1.5,virginica
6.9,3.2,5.7,2.3,virginica
5.6,2.8,4.9,2.0, virginica
7.7,2.8,6.7,2.0, virginica
6.3,2.7,4.9,1.8, virginica
6.7,3.3,5.7,2.1,virginica
7.2,3.2,6.0,1.8, virginica
6.2,2.8,4.8,1.8, virginica
6.1,3.0,4.9,1.8, virginica
6.4,2.8,5.6,2.1,virginica
7.2,3.0,5.8,1.6, virginica
7.4,2.8,6.1,1.9,virginica
7.9,3.8,6.4,2.0, virginica
6.4,2.8,5.6,2.2, virginica
6.3,2.8,5.1,1.5,virginica
6.1,2.6,5.6,1.4,virginica
7.7,3.0,6.1,2.3, virginica
6.3,3.4,5.6,2.4, virginica
6.4,3.1,5.5,1.8, virginica
6.0,3.0,4.8,1.8, virginica
6.9,3.1,5.4,2.1, virginica
6.7, 3.1, 5.6, 2.4, \text{virginica}
6.9,3.1,5.1,2.3, virginica
5.8,2.7,5.1,1.9, virginica
6.8,3.2,5.9,2.3, virginica
6.7, 3.3, 5.7, 2.5, virginica
6.7,3.0,5.2,2.3, virginica
6.3,2.5,5.0,1.9, virginica
6.5, 3.0, 5.2, 2.0, virginica
6.2,3.4,5.4,2.3,virginica
5.9,3.0,5.1,1.8, virginica
```

```
[225]: lines = data.split("\n")
print(lines)
```

['sepal_length, sepal_width, petal_length, petal_width, species',

```
'5.1,3.5,1.4,0.2,setosa', '4.9,3.0,1.4,0.2,setosa', '4.7,3.2,1.3,0.2,setosa',
'4.6,3.1,1.5,0.2,setosa', '5.0,3.6,1.4,0.2,setosa', '5.4,3.9,1.7,0.4,setosa',
'4.6,3.4,1.4,0.3,setosa', '5.0,3.4,1.5,0.2,setosa', '4.4,2.9,1.4,0.2,setosa',
'4.9,3.1,1.5,0.1,setosa', '5.4,3.7,1.5,0.2,setosa', '4.8,3.4,1.6,0.2,setosa',
'4.8,3.0,1.4,0.1,setosa', '4.3,3.0,1.1,0.1,setosa', '5.8,4.0,1.2,0.2,setosa',
'5.7,4.4,1.5,0.4,setosa', '5.4,3.9,1.3,0.4,setosa', '5.1,3.5,1.4,0.3,setosa',
'5.7,3.8,1.7,0.3,setosa', '5.1,3.8,1.5,0.3,setosa', '5.4,3.4,1.7,0.2,setosa',
'5.1,3.7,1.5,0.4,setosa', '4.6,3.6,1.0,0.2,setosa', '5.1,3.3,1.7,0.5,setosa',
'4.8,3.4,1.9,0.2,setosa', '5.0,3.0,1.6,0.2,setosa', '5.0,3.4,1.6,0.4,setosa',
'5.2,3.5,1.5,0.2,setosa', '5.2,3.4,1.4,0.2,setosa', '4.7,3.2,1.6,0.2,setosa',
'4.8,3.1,1.6,0.2,setosa', '5.4,3.4,1.5,0.4,setosa', '5.2,4.1,1.5,0.1,setosa',
'5.5,4.2,1.4,0.2,setosa', '4.9,3.1,1.5,0.1,setosa', '5.0,3.2,1.2,0.2,setosa',
'5.5,3.5,1.3,0.2,setosa', '4.9,3.1,1.5,0.1,setosa', '4.4,3.0,1.3,0.2,setosa',
'5.1,3.4,1.5,0.2,setosa', '5.0,3.5,1.3,0.3,setosa', '4.5,2.3,1.3,0.3,setosa',
'4.4,3.2,1.3,0.2,setosa', '5.0,3.5,1.6,0.6,setosa', '5.1,3.8,1.9,0.4,setosa',
'4.8,3.0,1.4,0.3,setosa', '5.1,3.8,1.6,0.2,setosa', '4.6,3.2,1.4,0.2,setosa',
'5.3,3.7,1.5,0.2,setosa', '5.0,3.3,1.4,0.2,setosa',
'7.0,3.2,4.7,1.4, versicolor', '6.4,3.2,4.5,1.5, versicolor',
'6.9,3.1,4.9,1.5, versicolor', '5.5,2.3,4.0,1.3, versicolor',
'6.5,2.8,4.6,1.5,versicolor', '5.7,2.8,4.5,1.3,versicolor',
'6.3,3.3,4.7,1.6,versicolor', '4.9,2.4,3.3,1.0,versicolor',
'6.6,2.9,4.6,1.3,versicolor', '5.2,2.7,3.9,1.4,versicolor',
'5.0,2.0,3.5,1.0,versicolor', '5.9,3.0,4.2,1.5,versicolor',
'6.0,2.2,4.0,1.0,versicolor', '6.1,2.9,4.7,1.4,versicolor',
'5.6,2.9,3.6,1.3,versicolor', '6.7,3.1,4.4,1.4,versicolor',
'5.6,3.0,4.5,1.5,versicolor', '5.8,2.7,4.1,1.0,versicolor',
'6.2,2.2,4.5,1.5,versicolor', '5.6,2.5,3.9,1.1,versicolor',
'5.9,3.2,4.8,1.8,versicolor', '6.1,2.8,4.0,1.3,versicolor',
'6.3,2.5,4.9,1.5,versicolor', '6.1,2.8,4.7,1.2,versicolor',
'6.4,2.9,4.3,1.3,versicolor', '6.6,3.0,4.4,1.4,versicolor',
'6.8,2.8,4.8,1.4, versicolor', '6.7,3.0,5.0,1.7, versicolor',
'6.0,2.9,4.5,1.5,versicolor', '5.7,2.6,3.5,1.0,versicolor',
'5.5,2.4,3.8,1.1,versicolor', '5.5,2.4,3.7,1.0,versicolor',
'5.8,2.7,3.9,1.2,versicolor', '6.0,2.7,5.1,1.6,versicolor',
'5.4,3.0,4.5,1.5, versicolor', '6.0,3.4,4.5,1.6, versicolor',
'6.7,3.1,4.7,1.5, versicolor', '6.3,2.3,4.4,1.3, versicolor',
'5.6,3.0,4.1,1.3,versicolor', '5.5,2.5,4.0,1.3,versicolor',
'5.5,2.6,4.4,1.2,versicolor', '6.1,3.0,4.6,1.4,versicolor',
'5.8,2.6,4.0,1.2,versicolor', '5.0,2.3,3.3,1.0,versicolor',
'5.6,2.7,4.2,1.3,versicolor', '5.7,3.0,4.2,1.2,versicolor',
'5.7,2.9,4.2,1.3,versicolor', '6.2,2.9,4.3,1.3,versicolor',
'5.1,2.5,3.0,1.1,versicolor', '5.7,2.8,4.1,1.3,versicolor',
'6.3,3.3,6.0,2.5, virginica', '5.8,2.7,5.1,1.9, virginica',
'7.1,3.0,5.9,2.1, virginica', '6.3,2.9,5.6,1.8, virginica',
'6.5,3.0,5.8,2.2,virginica', '7.6,3.0,6.6,2.1,virginica',
'4.9,2.5,4.5,1.7, virginica', '7.3,2.9,6.3,1.8, virginica',
'6.7,2.5,5.8,1.8, virginica', '7.2,3.6,6.1,2.5, virginica',
'6.5,3.2,5.1,2.0,virginica', '6.4,2.7,5.3,1.9,virginica',
```

```
'6.8,3.0,5.5,2.1,virginica', '5.7,2.5,5.0,2.0,virginica',
      '5.8,2.8,5.1,2.4, virginica', '6.4,3.2,5.3,2.3, virginica',
      '6.5,3.0,5.5,1.8,virginica', '7.7,3.8,6.7,2.2,virginica',
      '7.7,2.6,6.9,2.3, virginica', '6.0,2.2,5.0,1.5, virginica',
      '6.9,3.2,5.7,2.3, virginica', '5.6,2.8,4.9,2.0, virginica',
      '7.7,2.8,6.7,2.0, virginica', '6.3,2.7,4.9,1.8, virginica',
      '6.7,3.3,5.7,2.1, virginica', '7.2,3.2,6.0,1.8, virginica',
      '6.2,2.8,4.8,1.8,virginica', '6.1,3.0,4.9,1.8,virginica',
      '6.4,2.8,5.6,2.1,virginica', '7.2,3.0,5.8,1.6,virginica',
      '7.4,2.8,6.1,1.9,virginica', '7.9,3.8,6.4,2.0,virginica',
      '6.4,2.8,5.6,2.2, virginica', '6.3,2.8,5.1,1.5, virginica',
      '6.1,2.6,5.6,1.4, virginica', '7.7,3.0,6.1,2.3, virginica',
      '6.3,3.4,5.6,2.4, virginica', '6.4,3.1,5.5,1.8, virginica',
      '6.0,3.0,4.8,1.8,virginica', '6.9,3.1,5.4,2.1,virginica',
      '6.7,3.1,5.6,2.4, virginica', '6.9,3.1,5.1,2.3, virginica',
      '5.8,2.7,5.1,1.9, virginica', '6.8,3.2,5.9,2.3, virginica',
      '6.7,3.3,5.7,2.5, virginica', '6.7,3.0,5.2,2.3, virginica',
      '6.3,2.5,5.0,1.9,virginica', '6.5,3.0,5.2,2.0,virginica',
      '6.2,3.4,5.4,2.3, virginica', '5.9,3.0,5.1,1.8, virginica', '']
[226]: lines = data.split("\n")
       for line in lines:
           print(line)
      sepal_length, sepal_width, petal_length, petal_width, species
      5.1,3.5,1.4,0.2,setosa
      4.9,3.0,1.4,0.2,setosa
      4.7,3.2,1.3,0.2,setosa
      4.6,3.1,1.5,0.2,setosa
      5.0,3.6,1.4,0.2,setosa
      5.4,3.9,1.7,0.4,setosa
      4.6,3.4,1.4,0.3,setosa
      5.0,3.4,1.5,0.2,setosa
      4.4,2.9,1.4,0.2,setosa
      4.9,3.1,1.5,0.1,setosa
      5.4,3.7,1.5,0.2,setosa
      4.8,3.4,1.6,0.2,setosa
      4.8,3.0,1.4,0.1,setosa
      4.3,3.0,1.1,0.1,setosa
      5.8,4.0,1.2,0.2,setosa
      5.7,4.4,1.5,0.4,setosa
      5.4,3.9,1.3,0.4,setosa
      5.1,3.5,1.4,0.3,setosa
      5.7,3.8,1.7,0.3,setosa
      5.1,3.8,1.5,0.3,setosa
      5.4,3.4,1.7,0.2,setosa
      5.1,3.7,1.5,0.4,setosa
```

- 4.6,3.6,1.0,0.2,setosa
- 5.1,3.3,1.7,0.5,setosa
- 4.8,3.4,1.9,0.2,setosa
- 5.0,3.0,1.6,0.2,setosa
- 5.0,3.4,1.6,0.4,setosa
- 5.2,3.5,1.5,0.2,setosa
- 5.2,3.4,1.4,0.2,setosa
- 4.7,3.2,1.6,0.2,setosa
- 4.8,3.1,1.6,0.2,setosa
- 5.4,3.4,1.5,0.4,setosa
- 5.2,4.1,1.5,0.1,setosa
- 5.5,4.2,1.4,0.2,setosa
- 4.9,3.1,1.5,0.1,setosa
- 5.0,3.2,1.2,0.2,setosa
- 5.5,3.5,1.3,0.2,setosa
- 4.9,3.1,1.5,0.1,setosa
- 4.4,3.0,1.3,0.2,setosa
- 4.4,0.0,1.0,0.2,50050
- 5.1,3.4,1.5,0.2,setosa
- 5.0,3.5,1.3,0.3,setosa
- 4.5,2.3,1.3,0.3,setosa
- 4.4,3.2,1.3,0.2,setosa
- 5.0,3.5,1.6,0.6,setosa
- 5.1,3.8,1.9,0.4,setosa
- 4.8,3.0,1.4,0.3,setosa
- 5.1,3.8,1.6,0.2,setosa
- 4.6,3.2,1.4,0.2,setosa
- 5.3,3.7,1.5,0.2,setosa
- 5.0,3.3,1.4,0.2,setosa
- 7.0, 3.2, 4.7, 1.4, versicolor
- 6.4, 3.2, 4.5, 1.5, versicolor
- 6.9, 3.1, 4.9, 1.5, versicolor
- 5.5,2.3,4.0,1.3, versicolor
- 6.5,2.8,4.6,1.5,versicolor 5.7,2.8,4.5,1.3,versicolor
- 6.3,3.3,4.7,1.6,versicolor
- 4.9,2.4,3.3,1.0, versicolor
- 6.6,2.9,4.6,1.3, versicolor
- 5.2,2.7,3.9,1.4,versicolor
- 5.0,2.0,3.5,1.0,versicolor
- 5.9,3.0,4.2,1.5,versicolor
- 6.0,2.2,4.0,1.0,versicolor
- 6.1,2.9,4.7,1.4, versicolor
- 0.1,2.9,4.7,1.4, Versicolor
- 5.6,2.9,3.6,1.3,versicolor 6.7,3.1,4.4,1.4,versicolor
- 5.6,3.0,4.5,1.5,versicolor
- 5.8,2.7,4.1,1.0,versicolor
- 6.2,2.2,4.5,1.5,versicolor
- 5.6,2.5,3.9,1.1,versicolor

5.9,3.2,4.8,1.8, versicolor 6.1,2.8,4.0,1.3, versicolor 6.3,2.5,4.9,1.5, versicolor 6.1,2.8,4.7,1.2, versicolor 6.4,2.9,4.3,1.3, versicolor 6.6,3.0,4.4,1.4, versicolor 6.8,2.8,4.8,1.4, versicolor 6.7,3.0,5.0,1.7,versicolor 6.0,2.9,4.5,1.5, versicolor 5.7,2.6,3.5,1.0, versicolor 5.5,2.4,3.8,1.1,versicolor 5.5,2.4,3.7,1.0, versicolor 5.8,2.7,3.9,1.2, versicolor 6.0,2.7,5.1,1.6,versicolor 5.4,3.0,4.5,1.5, versicolor 6.0,3.4,4.5,1.6, versicolor 6.7,3.1,4.7,1.5, versicolor 6.3,2.3,4.4,1.3, versicolor 5.6,3.0,4.1,1.3, versicolor 5.5,2.5,4.0,1.3, versicolor 5.5,2.6,4.4,1.2, versicolor 6.1,3.0,4.6,1.4, versicolor 5.8,2.6,4.0,1.2, versicolor 5.0,2.3,3.3,1.0,versicolor 5.6,2.7,4.2,1.3, versicolor 5.7,3.0,4.2,1.2, versicolor 5.7,2.9,4.2,1.3, versicolor 6.2,2.9,4.3,1.3, versicolor 5.1,2.5,3.0,1.1,versicolor 5.7,2.8,4.1,1.3, versicolor 6.3,3.3,6.0,2.5, virginica 5.8,2.7,5.1,1.9, virginica 7.1,3.0,5.9,2.1, virginica 6.3,2.9,5.6,1.8, virginica 6.5,3.0,5.8,2.2,virginica 7.6,3.0,6.6,2.1,virginica 4.9,2.5,4.5,1.7, virginica 7.3,2.9,6.3,1.8, virginica 6.7,2.5,5.8,1.8, virginica 7.2,3.6,6.1,2.5,virginica 6.5,3.2,5.1,2.0, virginica 6.4,2.7,5.3,1.9, virginica 6.8,3.0,5.5,2.1,virginica 5.7,2.5,5.0,2.0, virginica 5.8,2.8,5.1,2.4, virginica 6.4,3.2,5.3,2.3, virginica 6.5,3.0,5.5,1.8, virginica 7.7,3.8,6.7,2.2, virginica

```
7.7,2.6,6.9,2.3, virginica
      6.0,2.2,5.0,1.5,virginica
      6.9,3.2,5.7,2.3, virginica
      5.6,2.8,4.9,2.0, virginica
      7.7, 2.8, 6.7, 2.0, virginica
      6.3,2.7,4.9,1.8, virginica
      6.7, 3.3, 5.7, 2.1, virginica
      7.2,3.2,6.0,1.8, virginica
      6.2,2.8,4.8,1.8, virginica
      6.1,3.0,4.9,1.8, virginica
      6.4,2.8,5.6,2.1, virginica
      7.2,3.0,5.8,1.6, virginica
      7.4,2.8,6.1,1.9, virginica
      7.9,3.8,6.4,2.0, virginica
      6.4,2.8,5.6,2.2,virginica
      6.3,2.8,5.1,1.5,virginica
      6.1,2.6,5.6,1.4,virginica
      7.7,3.0,6.1,2.3, virginica
      6.3,3.4,5.6,2.4, virginica
      6.4,3.1,5.5,1.8, virginica
      6.0,3.0,4.8,1.8,virginica
      6.9,3.1,5.4,2.1,virginica
      6.7,3.1,5.6,2.4, virginica
      6.9,3.1,5.1,2.3, virginica
      5.8,2.7,5.1,1.9, virginica
      6.8,3.2,5.9,2.3, virginica
      6.7, 3.3, 5.7, 2.5, \text{virginica}
      6.7,3.0,5.2,2.3, virginica
      6.3,2.5,5.0,1.9,virginica
      6.5,3.0,5.2,2.0,virginica
      6.2,3.4,5.4,2.3, virginica
      5.9,3.0,5.1,1.8,virginica
[236]: lines = data.split("\n")
       for line in lines:
           words = line.split(" ")
           for word in words:
               print(word)
      sepal_length, sepal_width, petal_length, petal_width, species
      5.1,3.5,1.4,0.2,setosa
      4.9,3.0,1.4,0.2,setosa
      4.7,3.2,1.3,0.2,setosa
      4.6,3.1,1.5,0.2,setosa
      5.0,3.6,1.4,0.2,setosa
      5.4,3.9,1.7,0.4,setosa
      4.6,3.4,1.4,0.3,setosa
```

- 5.0,3.4,1.5,0.2,setosa
- 4.4,2.9,1.4,0.2,setosa
- 4.9,3.1,1.5,0.1,setosa
- 5.4,3.7,1.5,0.2,setosa
- 4.8,3.4,1.6,0.2,setosa
- 4.8,3.0,1.4,0.1,setosa
- 4.3,3.0,1.1,0.1,setosa
- 5.8,4.0,1.2,0.2,setosa
- 5.7,4.4,1.5,0.4,setosa
- 5.4,3.9,1.3,0.4,setosa
- 5.1,3.5,1.4,0.3,setosa
- 5.7,3.8,1.7,0.3,setosa
- 5.1,3.8,1.5,0.3,setosa
- 5.4,3.4,1.7,0.2,setosa
- 5.1,3.7,1.5,0.4,setosa
- 4.6,3.6,1.0,0.2,setosa
- 5.1,3.3,1.7,0.5,setosa
- 4.8,3.4,1.9,0.2,setosa
- 5.0,3.0,1.6,0.2,setosa
- 5.0,3.4,1.6,0.4,setosa
- 5.2,3.5,1.5,0.2,setosa
- 5.2,3.4,1.4,0.2,setosa
- 4.7,3.2,1.6,0.2,setosa
- 4.8,3.1,1.6,0.2,setosa
- 5.4,3.4,1.5,0.4,setosa
- 5.2,4.1,1.5,0.1,setosa 5.5,4.2,1.4,0.2,setosa
- 4.9,3.1,1.5,0.1,setosa
- 5.0,3.2,1.2,0.2,setosa
- 5.5,3.5,1.3,0.2,setosa
- 4.9,3.1,1.5,0.1,setosa
- 4.4,3.0,1.3,0.2,setosa
- 5.1,3.4,1.5,0.2,setosa
- 5.0,3.5,1.3,0.3,setosa
- 4.5,2.3,1.3,0.3,setosa
- 4.4,3.2,1.3,0.2,setosa
- 5.0,3.5,1.6,0.6,setosa
- 5.1,3.8,1.9,0.4,setosa
- 4.8,3.0,1.4,0.3,setosa
- 5.1,3.8,1.6,0.2,setosa
- 4.6,3.2,1.4,0.2,setosa
- 5.3,3.7,1.5,0.2,setosa
- 5.0,3.3,1.4,0.2,setosa
- 7.0,3.2,4.7,1.4, versicolor
- 6.4, 3.2, 4.5, 1.5, versicolor
- 6.9,3.1,4.9,1.5, versicolor
- 5.5,2.3,4.0,1.3, versicolor
- 6.5,2.8,4.6,1.5, versicolor

5.7,2.8,4.5,1.3, versicolor 6.3,3.3,4.7,1.6, versicolor 4.9,2.4,3.3,1.0, versicolor 6.6,2.9,4.6,1.3, versicolor 5.2,2.7,3.9,1.4, versicolor 5.0,2.0,3.5,1.0,versicolor 5.9,3.0,4.2,1.5, versicolor 6.0,2.2,4.0,1.0,versicolor 6.1,2.9,4.7,1.4, versicolor 5.6,2.9,3.6,1.3, versicolor 6.7,3.1,4.4,1.4, versicolor 5.6,3.0,4.5,1.5, versicolor 5.8,2.7,4.1,1.0, versicolor 6.2,2.2,4.5,1.5, versicolor 5.6,2.5,3.9,1.1,versicolor 5.9,3.2,4.8,1.8, versicolor 6.1,2.8,4.0,1.3, versicolor 6.3,2.5,4.9,1.5,versicolor 6.1,2.8,4.7,1.2, versicolor 6.4,2.9,4.3,1.3, versicolor 6.6,3.0,4.4,1.4, versicolor 6.8,2.8,4.8,1.4, versicolor 6.7,3.0,5.0,1.7, versicolor 6.0,2.9,4.5,1.5, versicolor 5.7,2.6,3.5,1.0, versicolor 5.5,2.4,3.8,1.1,versicolor 5.5,2.4,3.7,1.0, versicolor 5.8,2.7,3.9,1.2, versicolor 6.0,2.7,5.1,1.6, versicolor 5.4,3.0,4.5,1.5, versicolor 6.0,3.4,4.5,1.6, versicolor 6.7,3.1,4.7,1.5, versicolor 6.3,2.3,4.4,1.3, versicolor 5.6,3.0,4.1,1.3, versicolor 5.5,2.5,4.0,1.3, versicolor 5.5,2.6,4.4,1.2, versicolor 6.1,3.0,4.6,1.4, versicolor 5.8,2.6,4.0,1.2, versicolor 5.0,2.3,3.3,1.0,versicolor 5.6,2.7,4.2,1.3, versicolor 5.7,3.0,4.2,1.2, versicolor 5.7,2.9,4.2,1.3, versicolor 6.2,2.9,4.3,1.3, versicolor 5.1,2.5,3.0,1.1,versicolor 5.7,2.8,4.1,1.3, versicolor 6.3,3.3,6.0,2.5, virginica 5.8,2.7,5.1,1.9, virginica 7.1,3.0,5.9,2.1, virginica

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```
[237]: ## how to merge one dictionary to another dictionary
       dict1 = {1:'apple',2:'banana'}
       dict2 = {3:'orange',4:'pomogranate'}
       dict1.update(dict2)
       print(dict1)
      {1: 'apple', 2: 'banana', 3: 'orange', 4: 'pomogranate'}
[238]: dict3 = {**dict1,**dict2}
       print(dict3)
      {1: 'apple', 2: 'banana', 3: 'orange', 4: 'pomogranate'}
[240]: ## Remove duplicates from list
       ## remove duplicate by useing 5 methodes
       arr=[1,2,2,3,3,4,4,5,6,7,8,9,9,]
[241]: ## remove the duplicates using set function
       arr2=set(arr)
       print(arr2)
      {1, 2, 3, 4, 5, 6, 7, 8, 9}
[244]: ## remove the duplicate useing lambda
       re_duplicate_fun=lambda arr:set(arr)
       print(re_duplicate_fun(arr))
      {1, 2, 3, 4, 5, 6, 7, 8, 9}
[245]: 4+3%5
[245]: 7
 []: ## Why is the python called interpreted language
       execute instruction dirrectly by line by line
       does not compile and run a complete programe at once
       this make the programm esier to debug
       we can just wright a python code and run it dirrectly
       this makes python fast to develop
 []: ## do you need a indedation in python
       yes indendation is a must in python and is part of the languages syntax
       indedation provides the readability to the code
 []: ## what are the classes and object in python
       class is the blueprint creating a object
       object is instance of class
```

```
[]: ## diffirence between the list and tuple
    list are mutable they can be modified and alter and creation(adding ,changing_
     item)
    list are defined useing squire bracket
    list take up more memory due to there ability to change and grow
    tuple is immutable once it created content can not be changed
    tuple are defined using parenthesis()
    tuple are more memory efficient usually faster because they are immutable.
[]: # what is the shallow copy and deep copy in python
    a shallow copy is a copy of object that stored the reference of the original
    so if we make any changes to the copy of the objects, it wiil reflected in u
      original object
[]: ### fibonecci generator
    def fibo_gen():
        a = 0
        b = 1
        while True:
            c = a
            a = b
            b = c + a
            yield c
    # Create a generator instance
    f = fibo_gen()
    # Generate and print the first two numbers of the Fibonacci sequence
    print(next(f)) # Output: 0
    print(next(f)) # Output: 1
[]: ## how to reverse list in python
    list1 = [4, 5, 6, 7]
    list1.reverse()
    print(list1)
                          ##[start vlue:end value:sep]
[]: ## slicing methode
    list2=list1[::-1]
    print(list2)
[]: list2=[]
    list1=[1,2,3,4]
    for i in range(len(list1)-1,-1-1):
        list2.append(list1[i])
    print(list2)
```

```
[]: ## find the current date time
     (base) C:\Users\Vikas>python
     Python 3.10.9 | packaged by Anaconda, Inc. | (main, Mar 1 2023, 18:18:15) [MSC_{\sqcup}
     \rightarrowv.1916 64 bit (AMD64)] on win32
     Type "help", "copyright", "credits" or "license" for more information.
     >>> import datetime
     >>> datetime.datetime.now()
     datetime.datetime(2024, 8, 12, 19, 53, 27, 237943)
     >>> today=datetime.datetime.now()
     >>> today.strftime("%y-%m-%d %H:%M:%S")
     '24-08-12 19:54:52'
     >>>
[]: ## check weather the email id is correct or wrong.
     import re
[]: import re
     def isvalidemail(email):
         regex = r"^[A-Z0-9._%+-]+0[A-Z0-9.-]+\.[A-Z]{2,}$"
         if len(email) > 7:
             if re.match(regex, email, re.IGNORECASE) is not None:
                 return True
         return False
     # Test the function
     if isvalidemail("netsetos@gmail.com"):
         print("Valid email address")
     else:
         print("Invalid email address")
[]: def add(a,b):
         return a+b
     print(add(4,5))
[]: Add =lambda a,b:a+b
     print(Add(4,5))
[]:
```