

In [1]:

```
1  #Program 1
2  from collections import deque
3
4  class Graph:
5      def __init__(self, adjac_lis):
6          self.adjac_lis = adjac_lis
7
8      def get_neighbors(self, v):
9          return self.adjac_lis[v]
10
11     def h(self, n):
12         H = {
13             'A': 1,
14             'B': 1,
15             'C': 1,
16             'D': 1
17         }
18         return H[n]
19
20     def a_star_algorithm(self, start, stop):
21         open_lst = set([start])
22         closed_lst = set([])
23
24         poo = {}
25         poo[start] = 0
26
27         par = {}
28         par[start] = start
29
30         while len(open_lst) > 0:
31             n = None
32
33             for v in open_lst:
34                 if n == None or poo[v] + self.h(v) < poo[n] + self.h(n):
35                     n = v;
36
37             if n == None:
38                 print('Path does not exist!')
39                 return None
40
41             if n == stop:
```

```

42         reconst_path = []
43
44         while par[n] != n:
45             reconst_path.append(n)
46             n = par[n]
47
48         reconst_path.append(start)
49
50         reconst_path.reverse()
51
52         print('Path found: {}'.format(reconst_path))
53         return reconst_path
54
55     for (m, weight) in self.get_neighbors(n):
56         if m not in open_lst and m not in closed_lst:
57             open_lst.add(m)
58             par[m] = n
59             poo[m] = poo[n] + weight
60
61         else:
62             if poo[m] > poo[n] + weight:
63                 poo[m] = poo[n] + weight
64                 par[m] = n
65
66             if m in closed_lst:
67                 closed_lst.remove(m)
68                 open_lst.add(m)
69
70         open_lst.remove(n)
71         closed_lst.add(n)
72
73     print('Path does not exist!')
74     return None
75
76
77     adjac_lis = {
78         'A': [('B', 1), ('C', 3), ('D', 7)],
79         'B': [('D', 5)],
80         'C': [('D', 12)]
81     }
82     graph1 = Graph(adjac_lis)
83     graph1.a_star_algorithm('A', 'D')

```

```
Path found: ['A', 'B', 'D']
```

```
Out[1]: ['A', 'B', 'D']
```

In [2]:

```
1  #Program 2
2  class Graph:
3      def __init__(self, graph, heuristicNodeList, startNode):
4
5          self.graph = graph
6          self.H=heuristicNodeList
7          self.start=startNode
8          self.parent={}
9          self.status={}
10         self.solutionGraph={}
11
12     def applyA0Star(self):
13         self.aoStar(self.start, False)
14
15     def getNeighbors(self, v):
16         return self.graph.get(v, '')
17
18     def getStatus(self,v):
19         return self.status.get(v,0)
20
21     def setStatus(self,v, val):
22         self.status[v]=val
23
24     def getHeuristicNodeValue(self, n):
25         return self.H.get(n,0)
26
27     def setHeuristicNodeValue(self, n, value):
28         self.H[n]=value
29
30     def printSolution(self):
31         print("FOR GRAPH SOLUTION, TRAVERSE THE GRAPH FROM THE START NODE:",self.start)
32         print("-----")
33         print(self.solutionGraph)
34         print("-----")
35
36     def computeMinimumCostChildNodes(self, v):
37         minimumCost=0
38         costToChildNodeListDict={}
39         costToChildNodeListDict[minimumCost]=[]
40         flag=True
41
```

```
42     for nodeInfoTupleList in self.getNeighbors(v):
43         cost=0
44         nodeList=[]
45         for c, weight in nodeInfoTupleList:
46             cost=cost+self.getHeuristicNodeValue(c)+weight
47             nodeList.append(c)
48
49         if flag==True:
50             minimumCost=cost
51             costToChildNodeListDict[minimumCost]=nodeList
52             flag=False
53         else:
54             if minimumCost>cost:
55                 minimumCost=cost
56                 costToChildNodeListDict[minimumCost]=nodeList
57
58     return minimumCost, costToChildNodeListDict[minimumCost]
59
60
61 def aoStar(self, v, backTracking):
62     print("HEURISTIC VALUES  :", self.H)
63     print("SOLUTION GRAPH    :", self.solutionGraph)
64     print("PROCESSING NODE    :", v)
65     print("-----")
66
67     if self.getStatus(v) >= 0:
68         minimumCost, childNodeList = self.computeMinimumCostChildNodes(v)
69         self.setHeuristicNodeValue(v, minimumCost)
70         self.setStatus(v, len(childNodeList))
71
72         solved=True
73         for childNode in childNodeList:
74             self.parent[childNode]=v
75             if self.getStatus(childNode)!=-1:
76                 solved=solved & False
77
78         if solved==True:
79             self.setStatus(v, -1)
80             self.solutionGraph[v]=childNodeList
81
82
83     if v!=self.start:
```

```

84         self.aoStar(self.parent[v], True)
85
86         if backTracking==False:
87             for childNode in childNodeList:
88                 self.setStatus(childNode,0)
89                 self.aoStar(childNode, False)
90
91 h1 = {'A': 1, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
92 graph1 = {
93     'A': [[('B', 1), ('C', 1)], [('D', 1)]],
94     'B': [[('G', 1)], [('H', 1)]],
95     'C': [[('J', 1)]],
96     'D': [[('E', 1), ('F', 1)]],
97     'G': [[('I', 1)]]
98 }
99 G1= Graph(graph1, h1, 'A')
100 G1.applyA0Star()
101 G1.printSolution()

```

```

HEURISTIC VALUES : {'A': 1, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {}
PROCESSING NODE  : A

```

```

-----
HEURISTIC VALUES : {'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {}
PROCESSING NODE  : B

```

```

-----
HEURISTIC VALUES : {'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {}
PROCESSING NODE  : A

```

```

-----
HEURISTIC VALUES : {'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 5, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {}
PROCESSING NODE  : G

```

```

-----
HEURISTIC VALUES : {'A': 10, 'B': 6, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 8, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {}
PROCESSING NODE  : B

```

```

-----
HEURISTIC VALUES : {'A': 10, 'B': 8, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 8, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {}
PROCESSING NODE  : A

```

```
-----
HEURISTIC VALUES : {'A': 12, 'B': 8, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 8, 'H': 7, 'I': 7, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {}
PROCESSING NODE   : I
-----
```

```
HEURISTIC VALUES : {'A': 12, 'B': 8, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 8, 'H': 7, 'I': 0, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {'I': []}
PROCESSING NODE   : G
-----
```

```
HEURISTIC VALUES : {'A': 12, 'B': 8, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 1, 'H': 7, 'I': 0, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {'I': [], 'G': ['I']}
PROCESSING NODE   : B
-----
```

```
HEURISTIC VALUES : {'A': 12, 'B': 2, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 1, 'H': 7, 'I': 0, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {'I': [], 'G': ['I'], 'B': ['G']}
PROCESSING NODE   : A
-----
```

```
HEURISTIC VALUES : {'A': 6, 'B': 2, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 1, 'H': 7, 'I': 0, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {'I': [], 'G': ['I'], 'B': ['G']}
PROCESSING NODE   : C
-----
```

```
HEURISTIC VALUES : {'A': 6, 'B': 2, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 1, 'H': 7, 'I': 0, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {'I': [], 'G': ['I'], 'B': ['G']}
PROCESSING NODE   : A
-----
```

```
HEURISTIC VALUES : {'A': 6, 'B': 2, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 1, 'H': 7, 'I': 0, 'J': 1, 'T': 3}
SOLUTION GRAPH   : {'I': [], 'G': ['I'], 'B': ['G']}
PROCESSING NODE   : J
-----
```

```
HEURISTIC VALUES : {'A': 6, 'B': 2, 'C': 2, 'D': 12, 'E': 2, 'F': 1, 'G': 1, 'H': 7, 'I': 0, 'J': 0, 'T': 3}
SOLUTION GRAPH   : {'I': [], 'G': ['I'], 'B': ['G'], 'J': []}
PROCESSING NODE   : C
-----
```

```
HEURISTIC VALUES : {'A': 6, 'B': 2, 'C': 1, 'D': 12, 'E': 2, 'F': 1, 'G': 1, 'H': 7, 'I': 0, 'J': 0, 'T': 3}
SOLUTION GRAPH   : {'I': [], 'G': ['I'], 'B': ['G'], 'J': [], 'C': ['J']}
PROCESSING NODE   : A
-----
```

```
-----
FOR GRAPH SOLUTION, TRAVERSE THE GRAPH FROM THE START NODE: A
-----
```

```
{ 'I': [], 'G': ['I'], 'B': ['G'], 'J': [], 'C': ['J'], 'A': ['B', 'C'] }
```

```

In [3]: 1 #Program 3
2 import csv
3 with open("trainingdata.csv") as f:
4     csv_file=csv.reader(f)
5     data=list(csv_file)
6
7     s=data[1][: -1]
8     g=["?" for i in range(len(s))] for j in range(len(s))]
9
10    for i in data:
11        if i[-1]=="Yes":
12            for j in range(len(s)):
13                if i[j]!=s[j]:
14                    s[j]="?"
15                    g[j][j]="?"
16
17            elif i[-1]=="No":
18                for j in range(len(s)):
19                    if i[j]!=s[j]:
20                        g[j][j]=s[j]
21                else:
22                    g[j][j]="?"
23            print("\nStep",data.index(i)+1)
24            print(s)
25            print(g)
26
27    gh=[]
28
29    for i in g:
30        for j in i:
31            if j!="?":
32                gh.append(i)
33                break
34    print("\nFinal specific hypothesis:\n",s)
35    print("\nFinal general hypothesis:\n",gh)

```

Step 1

```

['Sunny', 'Warm', 'Normal', 'Strong', 'Warm', 'Same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]

```


Step 2

```
['Sunny', 'Warm', 'Normal', 'Strong', 'Warm', 'Same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
```

Step 3

```
['Sunny', 'Warm', '?', 'Strong', 'Warm', 'Same']
[['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
```

Step 4

```
['Sunny', 'Warm', '?', 'Strong', 'Warm', 'Same']
[['Sunny', '?', '?', '?', '?', '?'], ['?', 'Warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', 'Same']]
```

Step 5

```
['Sunny', 'Warm', '?', 'Strong', '?', '?']
[['Sunny', '?', '?', '?', '?', '?'], ['?', 'Warm', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?'], ['?', '?', '?', '?', '?', '?']]
```

Final specific hypothesis:

```
['Sunny', 'Warm', '?', 'Strong', '?', '?']
```

Final general hypothesis:

```
[['Sunny', '?', '?', '?', '?', '?'], ['?', 'Warm', '?', '?', '?', '?']]
```

In [4]:

```
1 #Program 4
2 import numpy as np
3 import math
4 import csv
5
6 def read_data(filename):
7     with open(filename, 'r') as csvfile:
8         datareader = csv.reader(csvfile, delimiter=',')
9         headers = next(datareader)
10        metadata = []
11        traindata = []
12        for name in headers:
13            metadata.append(name)
14        for row in datareader:
15            traindata.append(row)
16
17    return (metadata, traindata)
18
19 class Node:
20     def __init__(self, attribute):
21         self.attribute = attribute
22         self.children = []
23         self.answer = ""
24
25     def __str__(self):
26         return self.attribute
27
28 def subtables(data, col, delete):
29     dict = {}
30     items = np.unique(data[:, col])
31     count = np.zeros((items.shape[0], 1), dtype=np.int32)
32
33     for x in range(items.shape[0]):
34         for y in range(data.shape[0]):
35             if data[y, col] == items[x]:
36                 count[x] += 1
37
38     for x in range(items.shape[0]):
39         dict[items[x]] = np.empty((int(count[x]), data.shape[1]), dtype="|S32")
40         pos = 0
41         for y in range(data.shape[0]):
```

```
42         if data[y, col] == items[x]:
43             dict[items[x]][pos] = data[y]
44             pos += 1
45         if delete:
46             dict[items[x]] = np.delete(dict[items[x]], col, 1)
47
48     return items, dict
49
50 def entropy(S):
51     items = np.unique(S)
52
53     if items.size == 1:
54         return 0
55
56     counts = np.zeros((items.shape[0], 1))
57     sums = 0
58
59     for x in range(items.shape[0]):
60         counts[x] = sum(S == items[x]) / (S.size * 1.0)
61
62     for count in counts:
63         sums += -1 * count * math.log(count, 2)
64     return sums
65
66 def gain_ratio(data, col):
67     items, dict = subtables(data, col, delete=False)
68     total_size = data.shape[0]
69     entropies = np.zeros((items.shape[0], 1))
70     intrinsic = np.zeros((items.shape[0], 1))
71
72     for x in range(items.shape[0]):
73         ratio = dict[items[x]].shape[0]/(total_size * 1.0)
74         entropies[x] = ratio * entropy(dict[items[x]][:, -1])
75         intrinsic[x] = ratio * math.log(ratio, 2)
76
77     total_entropy = entropy(data[:, -1])
78     iv = -1 * sum(intrinsic)
79
80     for x in range(entropies.shape[0]):
81         total_entropy -= entropies[x]
82
83     return total_entropy / iv
```

```
84
85 def create_node(data, metadata):
86     if (np.unique(data[:, -1])).shape[0] == 1:
87         node = Node("")
88         node.answer = np.unique(data[:, -1])[0]
89         return node
90
91     gains = np.zeros((data.shape[1] - 1, 1))
92
93     for col in range(data.shape[1] - 1):
94         gains[col] = gain_ratio(data, col)
95
96     split = np.argmax(gains)
97     node = Node(metadata[split])
98     metadata = np.delete(metadata, split, 0)
99     items, dict = subtables(data, split, delete=True)
100
101     for x in range(items.shape[0]):
102         child = create_node(dict[items[x]], metadata)
103         node.children.append((items[x], child))
104
105     return node
106
107 def empty(size):
108     s = ""
109     for x in range(size):
110         s += "    "
111     return s
112
113 def print_tree(node, level):
114     if node.answer != "":
115         print(empty(level), node.answer)
116         return
117     print(empty(level), node.attribute)
118     for value, n in node.children:
119         print(empty(level + 1), value)
120         print_tree(n, level + 2)
121
122 metadata, traindata = read_data("tennis.csv")
123 data = np.array(traindata)
124 node = create_node(data, metadata)
125 print_tree(node, 0)
```

Outlook

Overcast

b'Yes'

Rainy

Windy

b'FALSE'

b'Yes'

b'TRUE'

b'No'

Sunny

Humidity

b'High'

b'No'

b'Normal'

b'Yes'

In [5]:

```
1  #Program 5
2  import numpy as np
3
4  def sigmoid (x):
5      return 1/(1 + np.exp(-x))
6
7  def sigmoid_derivative(x):
8      return x * (1 - x)
9
10
11 inputs = np.array([[0,0],[0,1],[1,0],[1,1]])
12 expected_output = np.array([[0],[1],[1],[0]])
13
14 epochs = 10000
15 lr = 0.1
16 inputLayerNeurons, hiddenLayerNeurons, outputLayerNeurons = 2,2,1
17
18 hidden_weights = np.random.uniform(size=(inputLayerNeurons,hiddenLayerNeurons))
19 hidden_bias =np.random.uniform(size=(1,hiddenLayerNeurons))
20 output_weights = np.random.uniform(size=(hiddenLayerNeurons,outputLayerNeurons))
21 output_bias = np.random.uniform(size=(1,outputLayerNeurons))
22
23 print("Initial hidden weights: ",end='')
24 print(*hidden_weights)
25 print("Initial hidden biases: ",end='')
26 print(*hidden_bias)
27 print("Initial output weights: ",end='')
28 print(*output_weights)
29 print("Initial output biases: ",end='')
30 print(*output_bias)
31
32
33 for _ in range(epochs):
34     hidden_layer_activation = np.dot(inputs,hidden_weights)
35     hidden_layer_activation += hidden_bias
36     hidden_layer_output = sigmoid(hidden_layer_activation)
37     output_layer_activation = np.dot(hidden_layer_output,output_weights)
38     output_layer_activation += output_bias
39     predicted_output = sigmoid(output_layer_activation)
40
41     error = expected_output - predicted_output
```

```
42 d_predicted_output = error * sigmoid_derivative(predicted_output)
43 error_hidden_layer = d_predicted_output.dot(output_weights.T)
44 d_hidden_layer = error_hidden_layer * sigmoid_derivative(hidden_layer_output)
45
46 output_weights += hidden_layer_output.T.dot(d_predicted_output) * lr
47 output_bias += np.sum(d_predicted_output,axis=0,keepdims=True) * lr
48 hidden_weights += inputs.T.dot(d_hidden_layer) * lr
49 hidden_bias += np.sum(d_hidden_layer,axis=0,keepdims=True) * lr
50
51 print("Final hidden weights: ",end='')
52 print(*hidden_weights)
53 print("Final hidden bias: ",end='')
54 print(*hidden_bias)
55 print("Final output weights: ",end='')
56 print(*output_weights)
57 print("Final output bias: ",end='')
58 print(*output_bias)
59
60 print("\nOutput from neural network after 10,000 epochs: ",end='')
61 print(*predicted_output)
```

Initial hidden weights: [0.76817719 0.45708773] [0.98412098 0.25020245]

Initial hidden biases: [0.5978023 0.6549754]

Initial output weights: [0.38790814] [0.75967532]

Initial output biases: [0.98548079]

Final hidden weights: [5.82563358 3.6504641] [5.79200572 3.64398326]

Final hidden bias: [-2.41296209 -5.58175845]

Final output weights: [7.41817367] [-8.06799813]

Final output bias: [-3.32796279]

Output from neural network after 10,000 epochs: [0.06017549] [0.94427096] [0.94437921] [0.06034169]

In [6]:

```
1  #Program 6
2  import pandas as pd
3  from sklearn.model_selection import train_test_split
4  from sklearn.naive_bayes import GaussianNB
5  from sklearn.metrics import accuracy_score
6
7  DB = pd.read_csv('pg5.csv')
8  print(DB.columns)
9  len(DB)
10 DB.head(3)
11 X = DB.values[:,0:4]
12 Y = DB.values[:,4]
13 X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.30,random_state=10)
14
15 clf = GaussianNB()
16 clf.fit(X_train,Y_train)
17 Y_pred = clf.predict(X_test)
18 accuracy_score(Y_test,Y_pred,normalize=True)
```

Index(['Day', 'Temperature', 'Humidity', 'Windy', 'Play'], dtype='object')

Out[6]: 1.0

In [7]:

```
1  #Program 7
2  from sklearn import datasets
3  from sklearn.model_selection import train_test_split
4  from sklearn import metrics
5
6  iris = datasets.load_iris()
7  X_train,X_test,Y_train,Y_test = train_test_split(iris.data,iris.target)
8
9  from sklearn.cluster import KMeans
10 model = KMeans(n_clusters=3)
11 model.fit(X_train,Y_train)
12 model.score
13 acc1=metrics.accuracy_score(Y_test,model.predict(X_test))
14 print(acc1)
15
16 from sklearn.mixture import GaussianMixture
17 model2 = GaussianMixture(n_components=3)
18 model2.fit(X_train,Y_train)
19 model2.score
20 metrics
21 acc2=metrics.accuracy_score(Y_test,model.predict(X_test))
22 print(acc2)
```

0.10526315789473684

0.10526315789473684

In [8]:

```

1  #Program 8
2  from sklearn.model_selection import train_test_split
3  from sklearn.neighbors import KNeighborsClassifier
4  from sklearn.metrics import classification_report, confusion_matrix
5  from sklearn import datasets
6
7  iris=datasets.load_iris()
8  iris_data=iris.data
9  iris_labels=iris.target
10 x_train,x_test,y_train,y_test=train_test_split(iris_data,iris_labels,test_size=0.30)
11
12 classifier=KNeighborsClassifier(n_neighbors=5)
13 classifier.fit(x_train,y_train)
14 y_pred=classifier.predict(x_test)
15 print('Confusion matrix is as follows')
16 print(confusion_matrix(y_test,y_pred))
17 print('Accuracy Matrics')
18 print(classification_report(y_test,y_pred))

```

Confusion matrix is as follows

```

[[16  0  0]
 [ 0 11  2]
 [ 0  1 15]]

```

Accuracy Matrics

	precision	recall	f1-score	support
0	1.00	1.00	1.00	16
1	0.92	0.85	0.88	13
2	0.88	0.94	0.91	16
accuracy			0.93	45
macro avg	0.93	0.93	0.93	45
weighted avg	0.93	0.93	0.93	45

```
In [9]: 1 #Program 9
2 import matplotlib.pyplot as plt
3 import pandas as pd
4 import numpy as np
5
6 def kernel(point,xmat, k):
7     m,n= np1.shape(xmat)
8     weights = np1.mat(np1.eye((m)))
9
10    for j in range(m):
11        diff = point - X[j]
12        weights[j,j] = np1.exp(diff*diff.T/(-2.0*k**2))
13    return weights
14
15 def localWeight(point,xmat,yamat,k):
16     wei = kernel(point,xmat,k)
17     W = (X.T*(wei*X)).I*(X.T*(wei*yamat.T))
18     return W
19
20 def localWeightRegression(xmat,yamat,k):
21     m,n = np1.shape(xmat)
22     ypred = np1.zeros(m)
23
24     for i in range(m):
25         ypred[i] = xmat[i]*localWeight(xmat[i],xmat,yamat,k)
26     return ypred
27
28 data = pd.read_csv('prg9tips.csv')
29 bill = np1.array(data.total_bill)
30 tip = np1.array(data.tip)
31 mbill = np1.mat(bill)
32 mtip = np1.mat(tip)
33 m= np1.shape(mbill)[1]
34 print("*****",m)
35
36 one = np1.mat(np1.ones(m))
37 X= np1.hstack((one.T,mbill.T))
38 ypred = localWeightRegression(X,mtip,2)
39 SortIndex = X[:,1].argsort(0)
40 xsort = X[SortIndex][:,0]
41
```

```
42 fig = plt.figure()
43 ax = fig.add_subplot(1,1,1)
44 ax.scatter(bill,tip, color='blue')
45 ax.plot(xsort[:,1],ypred[SortIndex], color = 'red', linewidth=1)
46 plt.xlabel('Total bill')
47 plt.ylabel('Tip')
48 plt.show();
```

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