

# model

January 16, 2023

## 1 Solution (using KNN classification)

```
[1]: # importing the libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: df = pd.read_excel('robot.xlsx')
```

```
[3]: df.head()
```

```
[3]:
```

	r_number	speed (m/s)	weight (kg)	qualified
0	1	2.50	600	No
1	2	3.75	800	No
2	3	2.25	550	No
3	4	3.25	825	No
4	5	2.75	750	No

```
[4]: # since the model only interprets numerical data, we replace the qualified
      ↪ column with numerical values
df['qualified'] = df['qualified'].replace(['No', 'Yes'], [0,1])
df.head()
```

```
[4]:
```

	r_number	speed (m/s)	weight (kg)	qualified
0	1	2.50	600	0
1	2	3.75	800	0
2	3	2.25	550	0
3	4	3.25	825	0
4	5	2.75	750	0

```
[5]: # taking the X and y values
X = df.iloc[:, [1,2]].values
print(X)
```

```
[[ 2.5  600. ]
 [ 3.75 800. ]
 [ 2.25 550. ]
```

```

[ 3.25 825. ]
[ 2.75 750. ]
[ 4.5  500. ]
[ 3.5  525. ]
[ 3.   325. ]
[ 4.   400. ]
[ 4.25 375. ]
[ 2.   200. ]
[ 5.   250. ]
[ 8.25 850. ]
[ 5.75 875. ]
[ 4.75 625. ]
[ 5.5  675. ]
[ 5.25 950. ]
[ 7.   425. ]
[ 7.5  800. ]
[ 7.25 575. ]]

```

```
[6]: y = df.iloc[:, -1].values
y
```

```
[6]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1],
      dtype=int64)
```

```
[7]: # import the needed sklearn libraries
from sklearn.neighbors import KNeighborsClassifier # importing the classifier
from sklearn import metrics # import the needed metrics to calculate the
    ↪ confusion matrix and accuracy
from sklearn.preprocessing import StandardScaler # to scale the X values
from sklearn.model_selection import train_test_split # splitting the data into
    ↪ training and testing sets

```

```
[22]: # splitting X and y data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    ↪ random_state=0)

```

```
[26]: # scaling the X values
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.fit_transform(X_test)

```

```
[27]: print(X_train)
```

```

[[-1.37320266 -1.71372656]
 [ 1.55629635 -0.70234695]
 [-0.49435296 -0.2528449 ]
 [ 0.8239216   1.32041227]
 [-0.93377781  0.75853471]

```

```

[-1.22672771 -0.14046939]
[ 0.09154684 -0.36522041]
[ 0.23802179  0.19665715]
[-0.05492811 -0.92709797]
[-0.78730286 -1.151849  ]
[ 0.5309717   1.6575388  ]
[ 0.38449675 -1.48897554]
[-0.64082791  1.09566124]
[-1.08025276  0.08428163]
[ 0.67744665  0.42140817]
[ 2.2886711   1.20803676]]

```

```
[28]: print(X_test)
```

```

[[ 1.06870575  0.92976   ]
 [-1.06870575  0.92976   ]
 [ 0.92621165 -0.4090944 ]
 [-0.92621165 -1.45042561]]

```

```

[37]: # create, train and test the model
K = []
train = []
test = []
score = {}
accuracy = []
for k in range(1,16):
    K.append(k)                                # fill the list
    #with the k values (1-10)
    clf = KNeighborsClassifier(n_neighbors=k)    # create the model
    clf.fit(X_train, y_train)                  # train the model

    y_pred = clf.predict(X_test)               # testing the
    #model

    train_score = clf.score(X_train, y_train)   # get the
    #training score
    test_score = clf.score(X_test, y_test)      # get the test
    #score
    acc = metrics.accuracy_score(y_test, y_pred) # get the
    #accuracy (which is same as test score)

    train.append(train_score)                  # put the train
    #scores in the train list
    test.append(test_score)                   # put the test
    #scores in the test list
    score[k] = [train_score, test_score]       # placing it in
    #the scores dictionary

```

```
accuracy.append(acc) # placing it in  
↳ the accuracy list
```

```
[38]: train
```

```
[38]: [1.0,  
      0.9375,  
      0.9375,  
      0.8125,  
      0.875,  
      0.8125,  
      0.8125,  
      0.6875,  
      0.6875,  
      0.6875,  
      0.6875,  
      0.6875,  
      0.6875,  
      0.6875,  
      0.6875]
```

```
[39]: test
```

```
[39]: [1.0,  
      0.75,  
      1.0,  
      1.0,  
      1.0,  
      0.75,  
      0.75,  
      0.75,  
      0.75,  
      0.5,  
      0.5,  
      0.5,  
      0.5,  
      0.5,  
      0.5]
```

```
[40]: accuracy
```

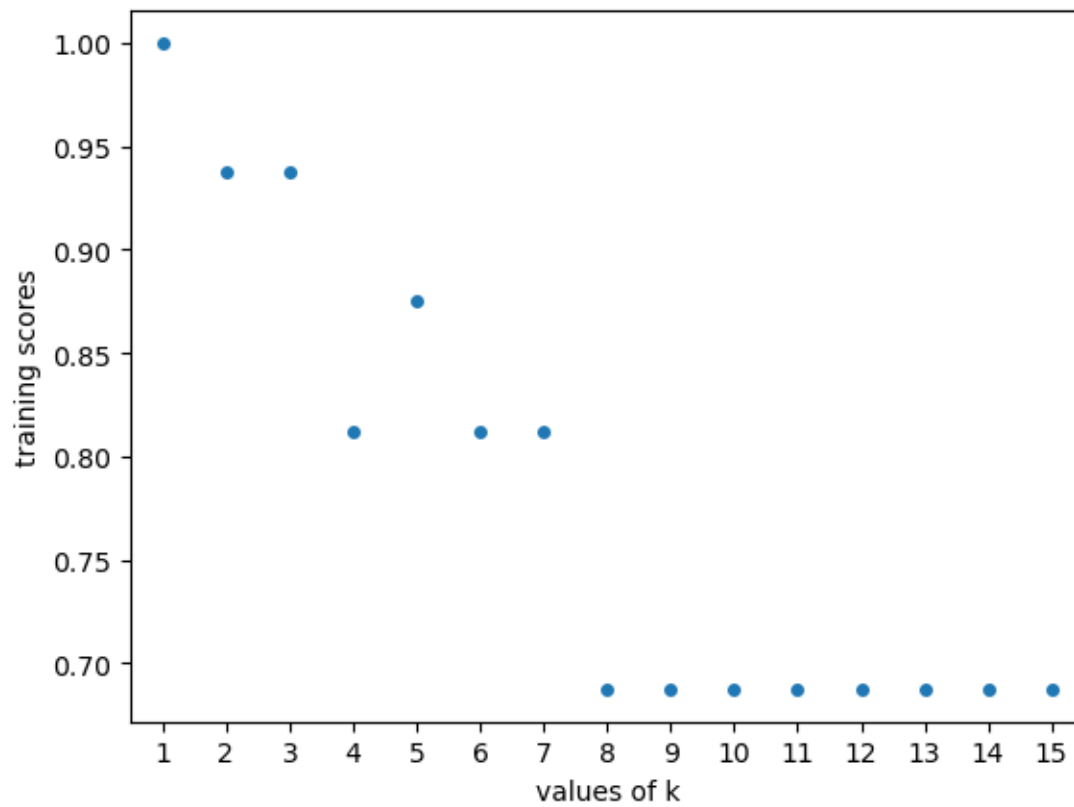
```
[40]: [1.0,  
      0.75,  
      1.0,  
      1.0,  
      1.0,  
      0.75,
```

```
0.75,  
0.75,  
0.75,  
0.5,  
0.5,  
0.5,  
0.5,  
0.5,  
0.5]
```

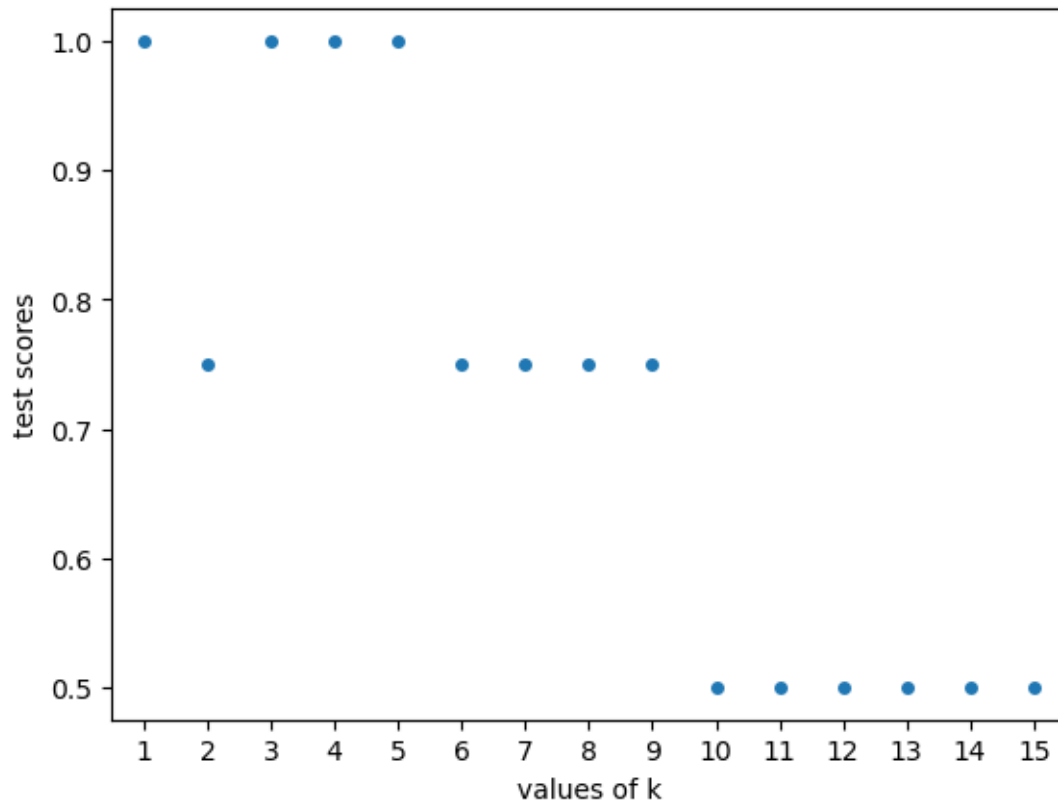
```
[41]: for keys, values in score.items():  
       print(f'{keys}:{values}')
```

```
1:[1.0, 1.0]  
2:[0.9375, 0.75]  
3:[0.9375, 1.0]  
4:[0.8125, 1.0]  
5:[0.875, 1.0]  
6:[0.8125, 0.75]  
7:[0.8125, 0.75]  
8:[0.6875, 0.75]  
9:[0.6875, 0.75]  
10:[0.6875, 0.5]  
11:[0.6875, 0.5]  
12:[0.6875, 0.5]  
13:[0.6875, 0.5]  
14:[0.6875, 0.5]  
15:[0.6875, 0.5]
```

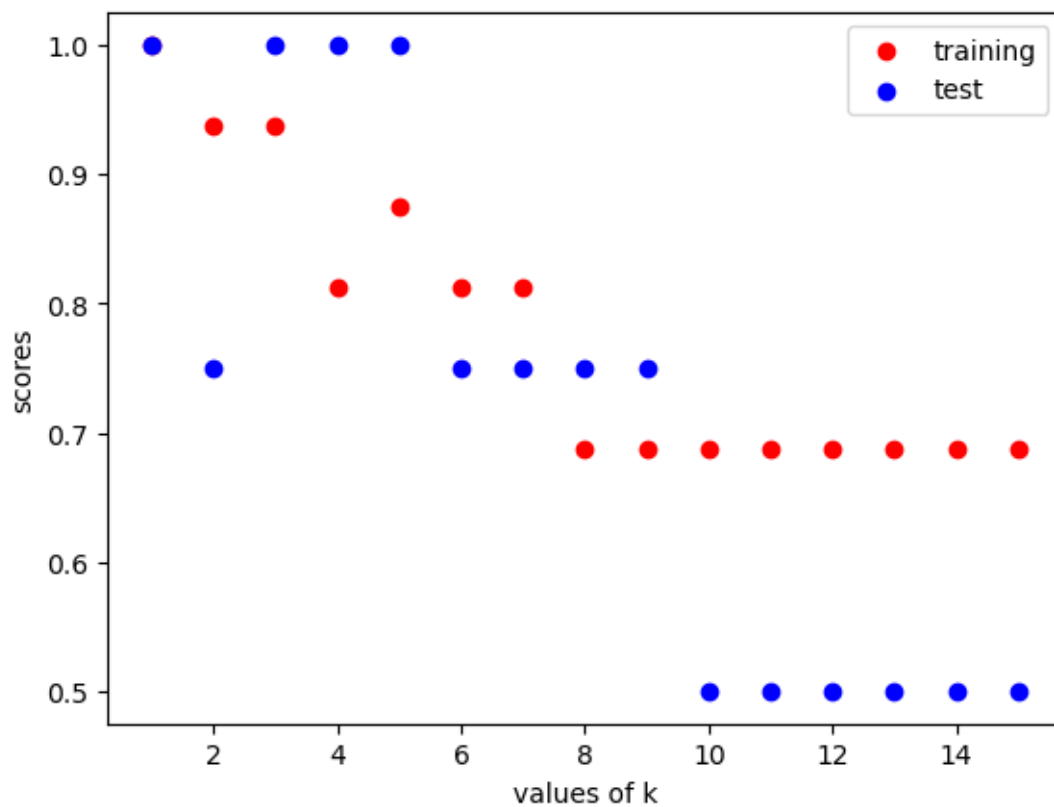
```
[42]: # plotting the strip plot for training values  
sns.stripplot(x=K, y=train)  
plt.xlabel('values of k')  
plt.ylabel('training scores')  
plt.show()
```



```
[43]: # plotting the strip plot for test values
sns.stripplot(x=K, y=test)
plt.xlabel('values of k')
plt.ylabel('test scores')
plt.show()
```

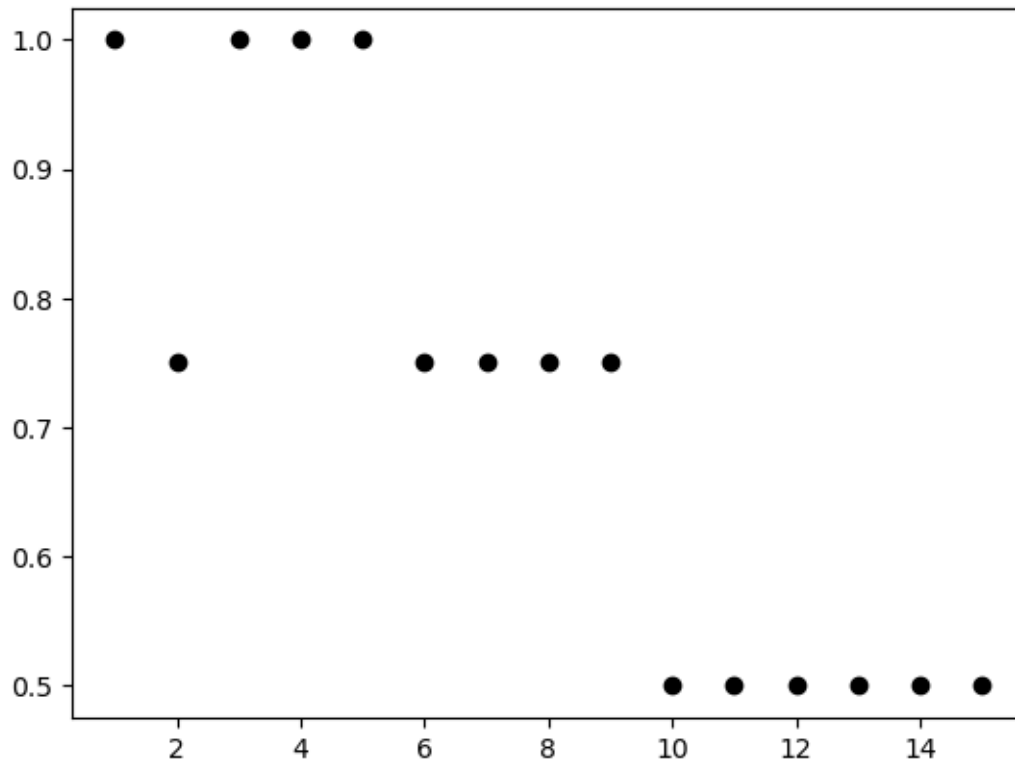


```
[44]: # overlapping plot of training and test values
plt.scatter(K, train, color='r')
plt.scatter(K, test, color='b')
plt.xlabel('values of k')
plt.ylabel('scores')
plt.legend(['training', 'test'])
plt.show()
```



```
[45]: plt.scatter(K,accuracy, color='k')  
plt.show()
```





2 from the graph above, the optimum k values are 1, 3, 4, and 5

```
[48]: new_k = [1,3,5]      # k values going to be used
x_new = [[5.20, 500]]
y_new = []
for k in new_k:
    clf_2 = KNeighborsClassifier(n_neighbors=k)
    clf_2.fit(X_train, y_train)
    pred_y = clf_2.predict(x_new)
    y_new.append(pred_y)
print(y_new)
```

```
[array([1], dtype=int64), array([1], dtype=int64), array([0], dtype=int64)]
```