

```
In [1]: # importing libraries
import numpy as nm
import matplotlib.pyplot as mtp
import pandas as pd
from sklearn import metrics

#importing datasets
data_set= pd.read_csv('user_data.csv')

#Extracting Independent and dependent Variable
x= data_set.iloc[:, [2,3]].values
y= data_set.iloc[:, 4].values

# Splitting the dataset into training and test set.
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test= train_test_split(x, y, test_size= 0.25, random_state=0)

#feature Scaling
from sklearn.preprocessing import StandardScaler
st_x= StandardScaler()
x_train= st_x.fit_transform(x_train)
x_test= st_x.transform(x_test)
```

```
In [2]: data_set
```

```
Out[2]:
```

|     | User ID  | Gender | Age | EstimatedSalary | Purchased |
|-----|----------|--------|-----|-----------------|-----------|
| 0   | 15624510 | Male   | 19  | 19000           | 0         |
| 1   | 15810944 | Male   | 35  | 20000           | 0         |
| 2   | 15668575 | Female | 26  | 43000           | 0         |
| 3   | 15603246 | Female | 27  | 57000           | 0         |
| 4   | 15804002 | Male   | 19  | 76000           | 0         |
| ... | ...      | ...    | ... | ...             | ...       |
| 395 | 15691863 | Female | 46  | 41000           | 1         |
| 396 | 15706071 | Male   | 51  | 23000           | 1         |
| 397 | 15654296 | Female | 50  | 20000           | 1         |
| 398 | 15755018 | Male   | 36  | 33000           | 0         |
| 399 | 15594041 | Female | 49  | 36000           | 1         |

400 rows × 5 columns

```
In [3]: x_test
```

```
Out[3]: array([[ -0.80480212,  0.50496393],
               [ -0.01254409, -0.5677824 ],
               [ -0.30964085,  0.1570462 ],
               [ -0.80480212,  0.27301877],
               [ -0.30964085, -0.5677824 ],
               [ -1.10189888, -1.43757673],
               [ -0.70576986, -1.58254245],
               [ -0.21060859,  2.15757314],
               [ -1.99318916, -0.04590581],
               [  0.8787462 , -0.77073441],
               [ -0.80480212, -0.59677555],
```

[-1.00286662, -0.42281668],  
 [-0.11157634, -0.42281668],  
 [ 0.08648817, 0.21503249],  
 [-1.79512465, 0.47597078],  
 [-0.60673761, 1.37475825],  
 [-0.11157634, 0.21503249],  
 [-1.89415691, 0.44697764],  
 [ 1.67100423, 1.75166912],  
 [-0.30964085, -1.37959044],  
 [-0.30964085, -0.65476184],  
 [ 0.8787462 , 2.15757314],  
 [ 0.28455268, -0.53878926],  
 [ 0.8787462 , 1.02684052],  
 [-1.49802789, -1.20563157],  
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 [ 1.86906873, 1.51972397],  
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 [-1.20093113, 0.30201192],  
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 [-1.49802789, -1.43757673],  
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 [ 2.1661655 , -0.79972756],  
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 [-0.21060859, 0.85288166],  
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 [ 2.1661655 , 0.38899135],  
 [-1.39899564, 0.56295021],  
 [-1.10189888, -0.33583725],  
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 [ 0.38358493, 0.27301877],  
 [ 0.18552042, -0.27785096],  
 [ 1.47293972, -1.03167271],  
 [ 0.8787462 , 1.08482681],  
 [ 10099, 2.15757314],

```
[ 2.06713324,  0.38899135],
[-1.39899564, -0.42281668],
[-1.20093113, -1.00267957],
[ 1.96810099, -0.91570013],
[ 0.38358493,  0.30201192],
[ 0.18552042,  0.1570462 ],
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[-1.49802789, -0.62576869],
[-1.29996338, -1.06066585],
[-1.39899564,  0.41798449],
[-1.10189888,  0.76590222],
[-1.49802789, -0.19087153],
[ 0.97777845, -1.06066585],
[ 0.97777845,  0.59194336],
[ 0.38358493,  0.99784738]])
```

In [4]: `y_test`

Out[4]: `array([0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1,
0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 0, 1,
1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 1], dtype=int64)`

In [5]: `#Fitting K-NN classifier to the training set
from sklearn.neighbors import KNeighborsClassifier
classifier= KNeighborsClassifier(n_neighbors=5, metric='minkowski', p=2 )
classifier.fit(x_train, y_train)`

Out[5]: `KNeighborsClassifier()`

In [6]: `#Predicting the test set result
y_pred= classifier.predict(x_test)`

In [7]: `y_pred`

Out[7]: `array([0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1,
0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1,
0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 1,
1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1], dtype=int64)`

In [8]: `#Now we will create the Confusion Matrix for our K-NN model to see the accuracy of the cla
#Creating the Confusion matrix
from sklearn.metrics import confusion_matrix
cm= confusion_matrix(y_test, y_pred)`

In [9]: `cm`

Out[9]: `array([[64, 4],
[ 3, 29]], dtype=int64)`

In [10]: `#Visulaizing the trianing set result
from matplotlib.colors import ListedColormap
x_set, y_set = x_train, y_train`

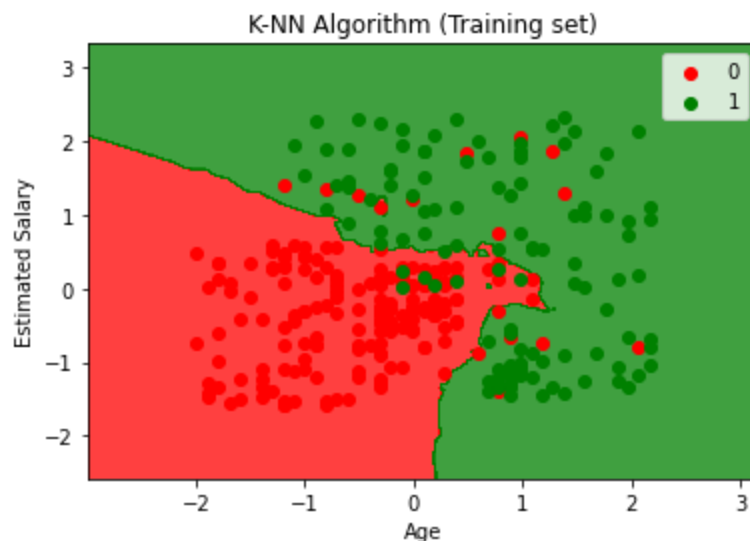
```

x1, x2 = nm.meshgrid(nm.arange(start = x_set[:, 0].min() - 1, stop = x_set[:, 0].max() + 1, step = 0.01),
nm.arange(start = x_set[:, 1].min() - 1, stop = x_set[:, 1].max() + 1, step = 0.01))
mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(), x2.ravel()]).T).reshape(x1.shape),
alpha = 0.75, cmap = ListedColormap(('red', 'green' )))
mtp.xlim(x1.min(), x1.max())
mtp.ylim(x2.min(), x2.max())
for i, j in enumerate(nm.unique(y_set)):
    mtp.scatter(x_set[y_set == j, 0], x_set[y_set == j, 1],
               c = ListedColormap(('red', 'green'))(i), label = j)
mtp.title('K-NN Algorithm (Training set)')
mtp.xlabel('Age')
mtp.ylabel('Estimated Salary')
mtp.legend()
mtp.show()

```

\*c\* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with \*x\* & \*y\*. Please use the \*color\* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.

\*c\* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with \*x\* & \*y\*. Please use the \*color\* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.



In [11]:

```

"""
As we can see the graph is showing the red point and green points.
The green points are for Purchased(1) and Red Points for not Purchased(0) variable.
The graph is showing an irregular boundary instead of showing any straight line or any curve.
"""

```

Out[11]: \nAs we can see the graph is showing the red point and green points. \nThe green points are for Purchased(1) and Red Points for not Purchased(0) variable.\nThe graph is showing an irregular boundary instead of showing any straight line or any curve because it is a K-NN algorithm, \n'

In [14]:

```

#Visualizing the test set result
from matplotlib.colors import ListedColormap
x_set, y_set = x_test, y_test
x1, x2 = nm.meshgrid(nm.arange(start = x_set[:, 0].min() - 1, stop = x_set[:, 0].max() + 1, step = 0.01),
nm.arange(start = x_set[:, 1].min() - 1, stop = x_set[:, 1].max() + 1, step = 0.01))
mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(), x2.ravel()]).T).reshape(x1.shape),
alpha = 0.75, cmap = ListedColormap(('red', 'green' )))
mtp.xlim(x1.min(), x1.max())
mtp.ylim(x2.min(), x2.max())
for i, j in enumerate(nm.unique(y_set)):
    mtp.scatter(x_set[y_set == j, 0], x_set[y_set == j, 1],

```

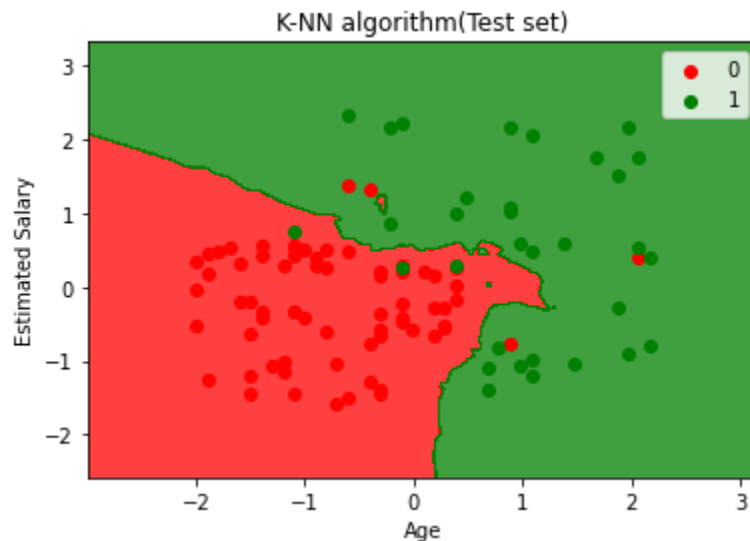
```

c = ListedColormap(('red', 'green'))(i), label = j)
mtp.title('K-NN algorithm(Test set)')
mtp.xlabel('Age')
mtp.ylabel('Estimated Salary')
mtp.legend()
mtp.show()

```

\*c\* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with \*x\* & \*y\*. Please use the \*color\* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.

\*c\* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with \*x\* & \*y\*. Please use the \*color\* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.



In [15]:

```

accuracy = metrics.accuracy_score(y_test,y_pred)
report = metrics.classification_report(y_test,y_pred)
cm = metrics.confusion_matrix(y_test,y_pred)

print("Classification report:")
print("Accuracy: ", accuracy)
print(report)
print("Confusion matrix:")
print(cm)

```

Classification report:

Accuracy: 0.93

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0            | 0.96      | 0.94   | 0.95     | 68      |
| 1            | 0.88      | 0.91   | 0.89     | 32      |
| accuracy     |           |        | 0.93     | 100     |
| macro avg    | 0.92      | 0.92   | 0.92     | 100     |
| weighted avg | 0.93      | 0.93   | 0.93     | 100     |

Confusion matrix:

```

[[64  4]
 [ 3 29]]

```

In [ ]: