

```
In [1]: # here, we implement knn by removing categorical variables (Type, LifeStyle)
import numpy as np
import pandas as pd
import seaborn as sns
import random
import math
import matplotlib.pyplot as plt
from sklearn import datasets, linear_model
from sklearn.model_selection import train_test_split
import operator
```

```
In [2]: 1 data1 = pd.read_csv('train_final.csv')
2 data2 = pd.read_csv('test_final.csv')
3 # X_train, X_crossVal, X_test = np.split(data, [600, 800])
4 # X_test
5 # xtrain, xtest = train_test_split(data, test_size=0.25)
6 # train_data = data[:50]
7 # test_data = data[50:]
8 # train_data[2]
9 data1.head()
10
```

Out[2]:

	id	Type	LifeStyle	Vacation	eCredit	salary	property	label
0	1	student	spend>saving	6	40	13.62	3.2804	C1
1	2	student	spend>saving	11	21	15.32	2.0232	C1
2	3	student	spend>saving	7	64	16.55	3.1202	C1
3	4	student	spend>saving	3	47	15.71	3.4022	C1
4	5	student	spend>saving	15	10	16.96	2.2825	C1

```
In [3]: 1 data2.head()
```

Out[3]:

	id	Type	LifeStyle	Vacation	eCredit	salary	property	label
0	1	student	spend<saving	12	19	14.7900	3.7697	C1
1	2	student	spend>>saving	29	10	16.1900	2.4839	C1
2	3	student	spend<<saving	28	60	15.4600	1.1885	C1
3	4	engineer	spend>saving	15	41	21.2600	1.4379	C1
4	5	librarian	spend<saving	2	9	19.7207	0.6913	C1

```
In [4]: 1 data1_label=data1[['label']]
        2 data1_label.head()
```

Out[4]:

	label
0	C1
1	C1
2	C1
3	C1
4	C1

```
In [5]: 1 data2_label=data2[['label']]
        2 data2_label.head()
```

Out[5]:

	label
0	C1
1	C1
2	C1
3	C1
4	C1

```
In [6]: 1 data1=data1.drop(['id','Type','LifeStyle','label'],axis=1)
        2 # data1.head()
        3 data1.head()
```

Out[6]:

	Vacation	eCredit	salary	property
0	6	40	13.62	3.2804
1	11	21	15.32	2.0232
2	7	64	16.55	3.1202
3	3	47	15.71	3.4022
4	15	10	16.96	2.2825

```
In [7]: 1 data2=data2.drop(['id','Type','LifeStyle','label'],axis=1)
        2 data2.head()
```

Out[7]:

	Vacation	eCredit	salary	property
0	12	19	14.7900	3.7697
1	29	10	16.1900	2.4839
2	28	60	15.4600	1.1885
3	15	41	21.2600	1.4379
4	2	9	19.7207	0.6913

```
In [8]: 1 data1_norm = (data1-data1.min()) / (data1.max()-data1.min())  
2 data1_norm.head()
```

Out[8]:

	Vacation	eCredit	salary	property
0	0.079365	0.107558	0.219960	0.183167
1	0.158730	0.052326	0.293102	0.112797
2	0.095238	0.177326	0.346023	0.174200
3	0.031746	0.127907	0.309882	0.189984
4	0.222222	0.020349	0.363663	0.127311

```
In [9]: 1 data2_norm = (data2-data2.min()) / (data2.max()-data2.min())  
2 data2_norm.head()
```

Out[9]:

	Vacation	eCredit	salary	property
0	0.20	0.058824	0.104637	0.398926
1	0.54	0.021008	0.175059	0.243041
2	0.52	0.231092	0.138339	0.085992
3	0.26	0.151261	0.430086	0.116229
4	0.00	0.016807	0.352657	0.025714

```
In [10]: 1 data1_norm['label']=data1_label['label']  
2 data1_norm.head()  
3
```

Out[10]:

	Vacation	eCredit	salary	property	label
0	0.079365	0.107558	0.219960	0.183167	C1
1	0.158730	0.052326	0.293102	0.112797	C1
2	0.095238	0.177326	0.346023	0.174200	C1
3	0.031746	0.127907	0.309882	0.189984	C1
4	0.222222	0.020349	0.363663	0.127311	C1

```
In [11]: 1 data2_norm['label']=data2_label['label']  
2 data2_norm.head()  
3
```

Out[11]:

	Vacation	eCredit	salary	property	label
0	0.20	0.058824	0.104637	0.398926	C1
1	0.54	0.021008	0.175059	0.243041	C1
2	0.52	0.231092	0.138339	0.085992	C1
3	0.26	0.151261	0.430086	0.116229	C1
4	0.00	0.016807	0.352657	0.025714	C1

```
In [12]: 1 xtrain=data1_norm.copy()  
2 xtest=data2_norm.copy()
```

```

In [13]: 1 k_list=[]
2 acc_list=[]
3 for k in range(1,26,2):
4     k_list.append(k)
5     predict=[]
6     def euc_distance(testrow,trainrow,length):
7         distance=0
8         #     for i in range(2):
9         #         if(testrow[i]==trainrow[i]):
10        #             distance+=1
11    #     for i in range(2):
12    #         if(testrow[i]==trainrow[i]):
13    #             distance+=1
14        for i in range(length):
15            distance+=pow((testrow[i]-trainrow[i]),2)
16        return math.sqrt(distance)
17    def getNeighbours(traindata,testRow,k):
18        distance_with_train=[]
19        length=len(testRow)-1
20        for x in range(len(traindata)):
21            dist=euc_distance(testRow,traindata[x],length)
22            distance_with_train.append((traindata[x],dist))
23        distance_with_train.sort(key=operator.itemgetter(1))
24        neighbors = []
25        for x in range(k):
26            neighbors.append(distance_with_train[x][0])
27        return neighbors
28    def getResponse(neighbors):
29        votes = {}
30        for x in range(len(neighbors)):
31            response = neighbors[x][-1]
32            if response in votes:
33                votes[response] += 1
34            else:
35                votes[response] = 1
36        sortedVotes = sorted(votes.items(), key=operator.itemgetter(1))
37        return sortedVotes[0][0]
38    def getAccuracy(xtest, predict):
39        correct = 0
40        for x in range(len(xtest)):
41            if xtest[x][-1] == predict[x]:
42                correct += 1
43        return (correct/float(len(xtest))) * 100.0
44    for i in range(len(xtest)):
45        neighbour=getNeighbours(xtrain.values,xtest.values[i],k)
46        #     print(neighbour)
47        result = getResponse(neighbour)
48        predict.append(result)
49    #     print('> predicted=' + repr(result) + ', actual=' + repr(xtest.values[i]))
50    accuracy = getAccuracy(xtest.values, predict)
51    acc_list.append(accuracy)
52    print('Accuracy: ' + repr(accuracy) + '%', 'with k=',k)

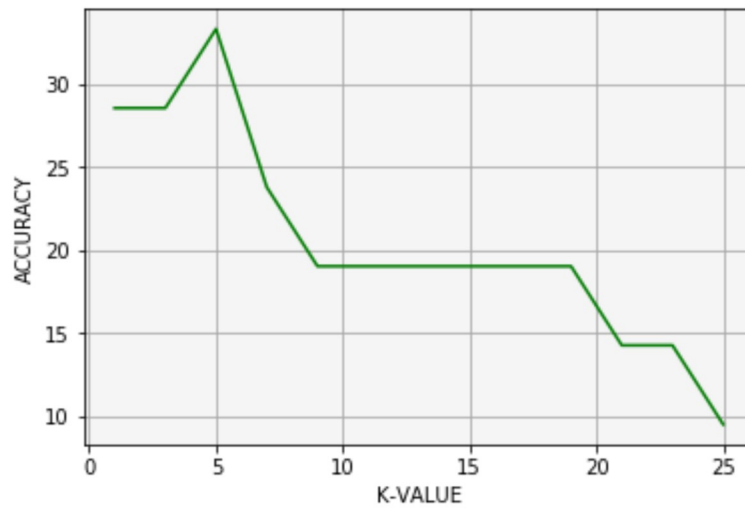
```

Accuracy: 28.57142857142857% with k= 1

Accuracy: 28.57142857142857% with k= 3

Accuracy: 33.33333333333333% with k= 5

```
In [14]: 1 plt.plot(k_list, acc_list, color='green')
          2 plt.xlabel('K-VALUE')
          3 plt.ylabel('ACCURACY')
          4 plt.grid(True)
          5 plt.show()
```



```
In [16]: 1 bigdata = pd.concat([xtrain, xtest], ignore_index=True)
          2 bigdata.head()
```

Out[16]:

	Vacation	eCredit	salary	property	label
0	0.079365	0.107558	0.219960	0.183167	C1
1	0.158730	0.052326	0.293102	0.112797	C1
2	0.095238	0.177326	0.346023	0.174200	C1
3	0.031746	0.127907	0.309882	0.189984	C1
4	0.222222	0.020349	0.363663	0.127311	C1

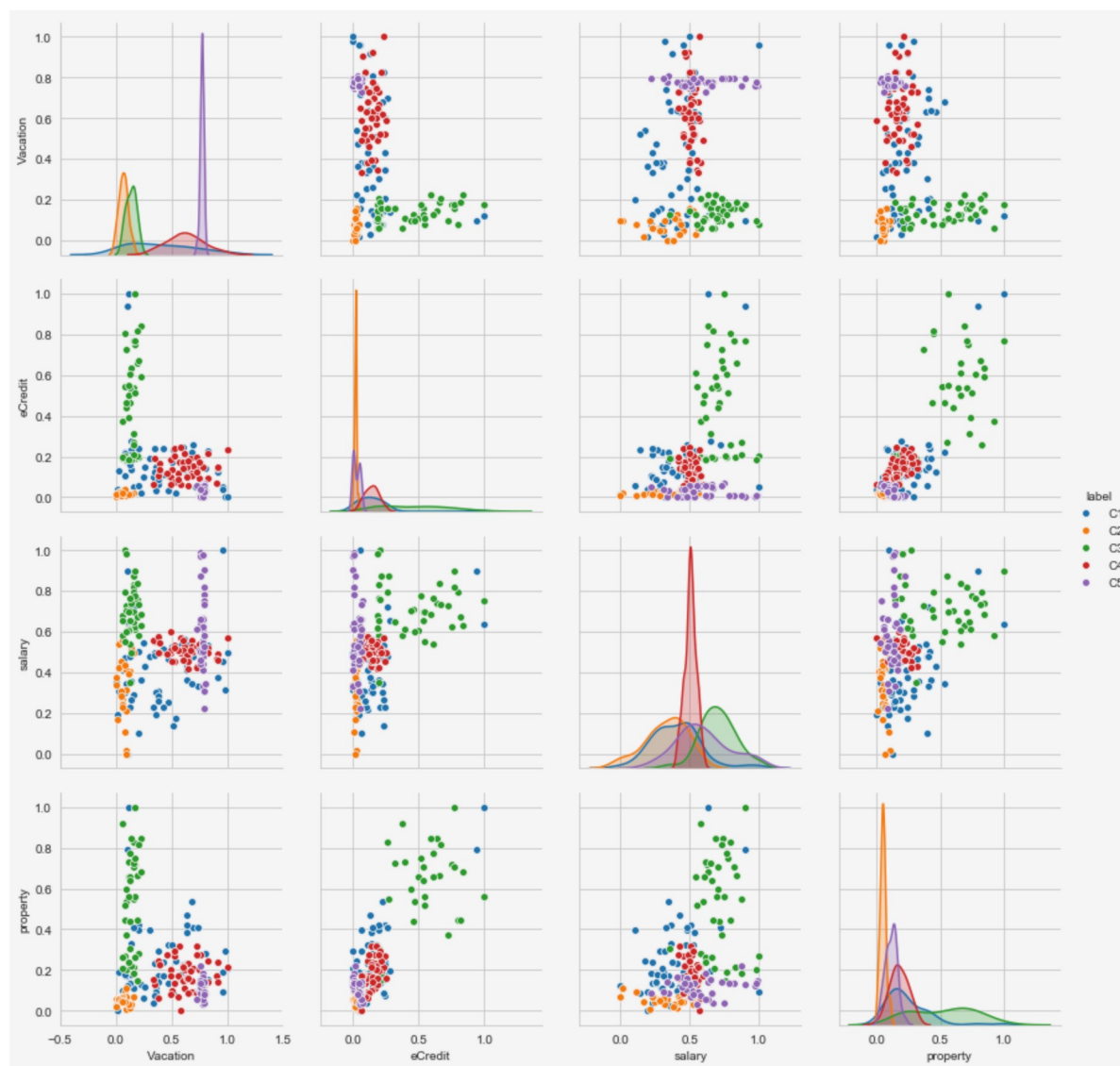
```
In [17]: 1 import seaborn as sns
2 plt.close();
3 sns.set_style("whitegrid");
4 sns.pairplot(bigdata, hue="label", size=3);
5 plt.show()
```

C:\Users\prem\Anaconda3\lib\site-packages\seaborn\axisgrid.py:2065: UserWarning: The `size` parameter has been renamed to `height`; please update your code.

warnings.warn(msg, UserWarning)

C:\Users\prem\Anaconda3\lib\site-packages\scipy\stats\stats.py:1713: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result either in an error or a different result.

return np.add.reduce(sorted[indexer] \* weights, axis=axis) / sumval



In [ ]: 1

In [ ]: 1

