```
# nhập thư viện
import numpy as np
import pandas as pd
# tạo hàm lấy dữ liệu
def loadCsv(filename) -> pd.DataFrame:
    '''Code here'''
   return pd.read csv(filename)
# tạo hàm biến đổi cột định tính, dùng phương pháp one hot
def transform(data, columns_trans): # data dang dataframe, data_trans là cột cần biến đổi --> dạng Series, nhiều cột cần biến đổi thì bỏ vào
    for i in columns trans:
       unique = data[i].unique() + '-' + i # trả lại mảng
        # tạo ma trận 0
       matrix_0 = np.zeros((len(data), len(unique)), dtype = int)
        frame 0 = pd.DataFrame(matrix_0, columns = unique)
        for index, value in enumerate(data[i]):
           frame_0.at[index, value + '-' + i] = 1
       data[unique] = frame_0
   return data # trả lại data truyền vào nhưng đã bị biến đổi
# tạo hàm scale dữ liệu về [0,1] (min max scaler)
def scale_data(data, columns_scale): # columns_scale là cột cần scale, nếu nhiều bỏ vào list ['a', 'b']
   for i in columns_scale:
       _max = data[i].max()
       _min = data[i].min()
        '''Code here'''
       min_max_scaler = lambda x: round((x - _min) / (_max - _min), 3) # Tính min-max
       data[i] = data[i].apply(min_max_scaler)
    return data # --> trả về frame
# hàm tính khoảng cách Cosine
def cosine_distance(train_X, test_X): # cả 2 đều dạng mảng
    dict_distance = dict()
   for index, value in enumerate(test_X, start = 1):
        for j in train X:
           result = np.sqrt(np.sum((j - value)**2))
           if index not in dict_distance:
               dict_distance[index] = [result]
           else:
               dict_distance[index].append(result)
   return dict_distance # {1: [6.0, 5.0], 2: [4.6, 3.1]}
# hàm gán kết quả theo k
def pred_test(k, train_X, test_X, train_y): # train_X, test_X là mảng, train_y là Series
   lst_predict = list()
   dict_distance = cosine_distance(train_X, test_X)
   train_y = train_y.to_frame(name = 'target').reset_index(drop = True) # train_y là frame
   frame_concat = pd.concat([pd.DataFrame(dict_distance), train_y], axis = 1)
   for i in range(1, len(dict_distance) + 1):
        sort_distance = frame_concat[[i, 'target']].sort_values(by = i, ascending = True)[:k] # sắp xếp và lấy k
        target_predict = sort_distance['target'].value_counts(ascending = False).index[0]
        lst_predict.append([i, target_predict])
   return lst_predict
Start coding or generate with AI.
## Demo qua drug200
data = loadCsv('drug200.csv')
data.head()
```

	Age	Sex	ВР	Cholesterol	Na_to_K	Drug
0	23	F	HIGH	HIGH	25.355	DrugY
1	47	M	LOW	HIGH	13.093	drugC
2	47	M	LOW	HIGH	10.114	drugC
3	28	F	NORMAL	HIGH	7.798	drugX
4	61	F	LOW	HIGH	18.043	DrugY

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		Age	Na_to_K	Drug	F-Sex	M-Sex	HIGH-BP	LOW-BP	NORMAL-BP	HIGH-Cholesterol	NORMAL-Cholesterol
-	0	23	25.355	DrugY	1	0	1	0	0	1	0
	1	47	13.093	drugC	0	1	0	1	0	1	0
	2	47	10.114	drugC	0	1	0	1	0	1	0
	3	28	7.798	drugX	1	0	0	0	1	1	0
	4	61	18.043	DrugY	1	0	0	1	0	1	0
	195	56	11.567	drugC	1	0	0	1	0	1	0
	196	16	12.006	drugC	0	1	0	1	0	1	0
	197	52	9.894	drugX	0	1	0	0	1	1	0
	198	23	14.020	drugX	0	1	0	0	1	0	1
	199	40	11.349	drugX	1	0	0	1	0	0	1
2	200 rows × 10 columns										

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 $scale_data(df, ['Age', 'Na_to_K']) # không cần gán biến vì nó trả về data đã truyền vào nhưng đã bị scale df$

	Age	Na_to_K	Drug	F-Sex	M-Sex	HIGH-BP	LOW-BP	NORMAL-BP	HIGH-Cholesterol	NORMAL-Cholesterol
0	0.136	0.597	DrugY	1	0	1	0	0	1	0
1	0.542	0.213	drugC	0	1	0	1	0	1	0
2	0.542	0.120	drugC	0	1	0	1	0	1	0
3	0.220	0.048	drugX	1	0	0	0	1	1	0
4	0.780	0.368	DrugY	1	0	0	1	0	1	0
195	0.695	0.166	drugC	1	0	0	1	0	1	0
196	0.017	0.179	drugC	0	1	0	1	0	1	0
197	0.627	0.113	drugX	0	1	0	0	1	1	0
198	0.136	0.242	drugX	0	1	0	0	1	0	1
199	0.424	0.159	drugX	1	0	0	1	0	0	1
196 197 198	0.017 0.627 0.136	0.179 0.113 0.242 0.159	drugC drugX drugX	0 0	1 1 1	0 0	1 0		1 1 1 0	

200 rows × 10 columns

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```
# tao data_X và target
data_X = df.drop(['Drug'], axis = 1).values
data_y = df['Drug']
```

```
print(data_X)
print(data_y)
 [[0.136 0.597 1.
                    ... 0. 1.
                    ... 0. 1.
     [0.542 0.213 0.
                                    0.
                                        ]
     [0.542 0.12 0.
                                    0.
     [0.627 0.113 0.
                               1.
                                    0.
                      ... 1.
     [0.136 0.242 0.
                      ... 1.
                              0.
                                    1.
     [0.424 0.159 1.
                     ... 0. 0.
          DrugY
          drugC
    1
    2
          drugC
          drugX
    4
          DrugY
    195
          drugC
    196
          drugC
    197
          drugX
    198
          drugX
          drugX
    Name: Drug, Length: 200, dtype: object
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from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(data_X, data_y, test_size = 0.2, random_state = 0)
print(X_train)
print(X_test)
print(y_train)
print(y_test)
print(len(X_train), len(X_test), len(y_train), len(y_test))
print(type(y_train))
                             1.
 [[0.458 0.462 1.
                    ... 0.
                                    0.
                    ... 0.
     [0.237 0.206 0.
                               1.
                                    0.
     [0.271 0.754 0.
     [0.424 0.12 1.
                      ... 1.
                              1.
                                    0.
                     ... 0.
     [0.898 0.126 0.
                    ... 1.
1. 0
     [0.407 0.343 1.
                               0.
                                    1.
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    [[0.136 0.032 0.
                          0.
                                    0.
     [0.22 0.207 1.
                    0. 0. 0. 1.
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1.
     [0.458 0.43 0.
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                                            1.
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                                    0.
     [0.085 0.918 0.
                           1.
                                                 0.
     [0.169 0.398 0.
                    1. 0.
                                0. 1.
                                           1.
     [0.085 0.169 1.
                                      0.
     [0.119 0.073 1.
                                0.
                                    1.
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                    0. 0.
     [0.373 0.179 1.
                               1.
                                      0.
                                           0.
     [0.475 0.285 0.
                     1.
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                                      0.
                    1.
     [0.678 0.031 0.
                           0. 0. 1.
                                            0.
                                0.
     [0.153 0.1 0.
                    1. 1.
0. 1.
                                      0.
                                            0.
     [0.068 0.616 1.
                                 0.
                                      0.
     [0.119 0.518 1. 0. 1.
                    1. 0.
1. 1.
                                1.
     [0.373 0.327 0.
                                      0.
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                                                 1.
     [0.763 0.24 0.
                                 0.
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                                           1.
     [0.441 0.149 0.
                    1.
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     [0.847 0.801 1.
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     0.932 0.445 1.
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     [0.407 0.24 0.
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     [0.271 0.338 0.
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     [0.356 0.288 1.
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                                 0.
                                      0.
                                            0.
                    1.
     [0.78 0.099 0.
                           0.
                                0.
                                     1.
                                            1.
     [0.864 0.057 1.
                                 0.
     [0.322 0.208 1.
                           0.
                                1.
                                      0.
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                    0.
                                1.
     [0.39 0.738 1.
                           0.
                                      0.
                                           0.
     [0.356 0.328 1.
                           0.
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                                      1.
                                            1.
     [0.881 0.301 1.
                                            1.
                    0.
1.
     [0.712 0.614 1.
                           0.
                                0.
                                            0.
     [0.322 0.506 0.
                           0.
                                 0.
                                      1.
     [0.593 0.188 1.
                                0. 1.
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     [0.915 0.162 0.
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                                                 1.
     [0.186 0.247 1.
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                                      0.
                                           1.
     [0.576 0.076 0.
                           1.
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                                      0.
     [0.542 0.013 1.
                     0.
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                                 0.
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                                            0.
                    1.
     [0.542 0.76 0.
                           0.
                                      0.
                                            0.
                                1.
                                                 1.
     [0.831 0.459 0.
                           1. 0.
                                      0.
                                            0.
     [0.288 0.038 1.
                                            1.
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1.

[0.78 0.368 1.

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0.
                                     0.
 [0.39 0.158 1.
                 0.
                      1. 0.
                                            1.
                                               ]
 [0.119 0.178 0.
                       0.
                                1.
                 1.
                            0.
                                      1.
      DrugY
      drugA
66
26
      DrugY
113
      drugX
168
      DrugY
67
      drugX
192
      DrugY
117
      drugX
47
      drugC
```

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```
test_pred = pred_test(6, X_train, X_test, y_train)
df_test_pred = pd.DataFrame(test_pred).drop([0], axis = 1)
df_test_pred.index = range(1, len(test_pred) + 1)
df_test_pred.columns = ['Predict']

df_actual = pd.DataFrame(y_test)
df_actual.index = range(1, len(y_test) + 1)
df_actual.columns = ['Actual']

pd.concat([df_test_pred, df_actual], axis = 1)
```

	Predict	Actual
1	drugC	drugC
2	drugX	drugX
3	DrugY	DrugY
4	DrugY	DrugY
5	DrugY	DrugY
6	DrugY	drugX
7	drugX	drugX
8	drugX	drugX
9	DrugY	DrugY
10	drugX	drugX

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13	DrugY	DrugY
14	drugX	DrugY
15	drugB	drugB
16	drugC	drugC
17	DrugY	DrugY
18	drugX	DrugY
19	drugX	drugX
	1 A	D 1/