```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import classification_report, confusion_matrix
from sklearn.manifold import TSNE
import time
import matplotlib.pyplot as plt
plt.style.use('seaborn-whitegrid')
%matplotlib inline
def accuracy_score(matrix):
   tp_sum = 0
   score = 0
   np.array(matrix)
   for i in range(len(matrix)):
       tp_sum += matrix[i][i]
   score = tp_sum / np.sum(matrix)
   return score
def knn(train_data, test_data, train_label, test_label):
   errors = []
   scores = []
   for i in range (5,51,5):
       start_time = time.time()
       knn_e = KNeighborsClassifier(n_neighbors=i)
       knn_e.fit(train_data, train_label)
       pred_label = knn_e.predict(test_data)
       end_time = time.time()
       errors.append(np.mean(pred_label != test_label)) # storing mean-e
       print("===========")
       print("\nReport using k=", i, ":\n")
       print(classification_report(test_label, pred_label))
       matrix = confusion_matrix(test_label, pred_label)
       accuracy = accuracy_score(matrix)*100
       print("\nAccuracy score: ", "%.2f" % accuracy, "%"," Execution
       print("=========="")
       scores.append("%.2f" % accuracy)
   plt.plot(range(5,51,5), scores, color='green', linestyle='dashed', ma
   plt.xticks(np.arange(0, 56, step=5))
   plt.yticks(np.arange(90, 95.2, step=0.2))
   plt.xlabel("value of k")
   plt.ylabel("accuracy")
   plt.show()
```

```
data = pd.read_csv("mnist_test.csv")
```

[4] data.head(5)

	label	1x1	1x2	1x3	1x4	1x5	1x6	1x7	1x8	1x9
0	7	0	0	0	0	0	0	0	0	0
1	2	0	0	0	0	0	0	0	0	0
2	1	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0
4	4	0	0	0	0	0	0	0	0	0

5 rows × 785 columns

```
features = data.drop("label", axis=1)
labels = data.iloc[:,0]
# Converting data into int64 to float64
features = features.astype('float64') #
# Objects for data optimization
m_scale = MinMaxScaler()
s_scale = StandardScaler()
e_data = TSNE(n_components=2, perplexity=30, random_state=100)
# Optimizing data
m_scale_optimized_data = m_scale.fit_transform(features)
s_scale_optimized_data = s_scale.fit_transform(features)
embeded_data = e_data.fit_transform(features)
# Optimizing embeded data
m_scale_embeded_data = m_scale.fit_transform(embeded_data)
s_scale_embeded_data = s_scale.fit_transform(embeded_data)
# Spliting data into training and test data
r_train_data, r_test_data, r_train_label, r_test_label = train_test_split
m_train_data, m_test_data, m_train_label, m_test_label = train_test_split
s_train_data, s_test_data, s_train_label, s_test_label = train_test_split
```

e\_train\_data, e\_test\_data, e\_train\_label, e\_test\_label = train\_test\_split

```
# Spliting optimized embeded data into training and test data
me_train_data, me_test_data, me_train_label, me_test_label = train_test_s
se_train_data, se_test_data, se_train_label, se_test_label = train_test_s
```

### kNN with raw data

[24] knn<u>(</u>r\_train\_data, r\_test\_data, r\_train\_label, r\_test\_label<u>)</u>

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Report using k= 5:

	precision	recall	f1-score	support
0	0.95	0.99	0.97	302
1	0.90	1.00	0.94	355
2	0.98	0.93	0.95	284
3	0.95	0.96	0.96	297
4	0.97	0.90	0.93	302
5	0.95	0.92	0.93	262
6	0.98	0.97	0.98	302
7	0.95	0.94	0.95	323
8	0.98	0.91	0.94	278
9	0.90	0.93	0.91	295
micro avg	0.95	0.95	0.95	3000
macro avg	0.95	0.95	0.95	3000
weighted avg	0.95	0.95	0.95	3000

Accuracy score: 94.77 % Execution time: 50.491 s

Report using k= 10:

	precision	recall	f1-score	support
0	0.93	0.99	0.96	302
1	0.87	1.00	0.93	355
2	0.98	0.90	0.94	284

## kNN with min-max optimized raw data

knn<u>(</u>m\_train\_data, m\_test\_data, m\_train\_label, m\_test\_label<u>)</u>

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#### Report using k= 5:

		precision	recall	f1-score	support
	0	0.96	0.99	0.97	273
	1	0.89	0.99	0.94	344
	2	0.98	0.89	0.93	307
	3	0.94	0.95	0.95	310
	4	0.94	0.92	0.93	297
	5	0.95	0.92	0.94	262
	6	0.95	0.97	0.96	269
	7	0.93	0.94	0.94	308
	8	0.97	0.92	0.95	302
	9	0.94	0.93	0.93	328
micro	avg	0.94	0.94	0.94	3000
macro	avg	0.95	0.94	0.94	3000
weighted	avg	0.94	0.94	0.94	3000

Accuracy score: 94.30 % Execution time: 38.923 s

Report using k= 10:

	precision	recall	f1-score	support
0	0.95	0.99	0.97	273
1	0.87	0.99	0.93	344
2	0.98	0.88	0.93	307

## kNN with standard scaler optimized raw data

knn(s\_train\_data, s\_test\_data, s\_train\_label, s\_test\_label)

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#### Report using k= 5:

	precision	recall	f1-score	support
0	0.93	0.98	0.95	316
1	0.90	0.99	0.94	339
2	0.93	0.88	0.91	304
3	0.88	0.96	0.91	294
4	0.89	0.91	0.90	294
5	0.92	0.88	0.90	284

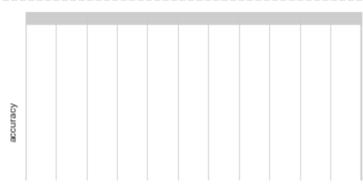
	6	0.95	0.94	0.94	279	
	7	0.90	0.92	0.91	277	
	8	0.96	0.81	0.88	297	
	9	0.90	0.86	0.88	316	
micro a	vg	0.91	0.91	0.91	3000	
macro a	vg	0.92	0.91	0.91	3000	
weighted a	vg	0.91	0.91	0.91	3000	
Accuracy s	core: 9	1.37 %	Execution	time:	47.251 s	
Report using k= 10 :						
	prec	ision r	ecall f1-	score	support	
	0	0.92	0.98	0.95	316	
	1	0.86	0.99	0.92	339	

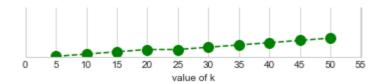
### kNN with embeded data

knn(e\_train\_data, e\_test\_data, e\_train\_label, e\_test\_label)

	0	0.98	0.97	0.97	293
	1	0.95	1.00	0.97	339
	2	0.98	0.94	0.96	296
	3	0.94	0.95	0.95	317
	4	0.95	0.91	0.93	301
	5	0.91	0.93	0.92	265
	6	0.95	0.98	0.97	289
	7	0.93	0.94	0.94	298
	8	0.94	0.90	0.92	295
	9	0.92	0.91	0.92	307
micro	avg	0.94	0.94	0.94	3000
macro	avg	0.94	0.94	0.94	3000
${\tt weighted}$	avg	0.94	0.94	0.94	3000

Accuracy score: 94.47 % Execution time: 0.113 s



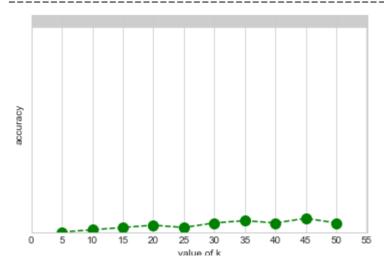


## kNN with min-max optimized embeded data

[28] knn(me\_train\_data, me\_test\_data, me\_train\_label, me\_test\_label)

		precision	recall	tl-score	support
	0	0.97	0.98	0.97	312
	1	0.92	1.00	0.96	346
	2	0.98	0.90	0.94	294
	3	0.95	0.93	0.94	329
	4	0.93	0.92	0.93	295
	5	0.93	0.93	0.93	274
	6	0.96	0.97	0.97	274
	7	0.93	0.94	0.94	300
	8	0.95	0.93	0.94	296
	9	0.93	0.94	0.93	280
micro	avg	0.94	0.94	0.94	3000
macro	avg	0.95	0.94	0.94	3000
weighted	avg	0.95	0.94	0.94	3000

Accuracy score: 94.47 % Execution time: 0.098 s



[29]

# kNN with standard scaler optimized embeded data

	1	0.92	1.00	0.96	362
	2	0.99	0.93	0.96	297
	3	0.93	0.94	0.93	305
	4	0.97	0.92	0.94	314
	5	0.92	0.90	0.91	263
	6	0.95	1.00	0.98	275
	7	0.94	0.93	0.94	311
	8	0.94	0.90	0.92	301
	9	0.92	0.93	0.92	284
micro	avg	0.94	0.94	0.94	3000
macro	avg	0.94	0.94	0.94	3000
weighted	avg	0.94	0.94	0.94	3000

Accuracy score: 94.40 % Execution time: 0.092 s

#### Report using k= 50:

	precision	recall	f1-score	support
0	0.97	0.98	0.97	288
1	0.92	1.00	0.96	362
2	0.99	0.93	0.96	297
3	0.93	0.94	0.93	305
4	0.97	0.92	0.94	314
5	0.92	0.90	0.91	263
6	0.95	1.00	0.98	275
7	0.94	0.93	0.94	311
8	0.94	0.90	0.92	301
9	0.92	0.93	0.92	284