Assignment:

Try SVM classifier on MNIST dataset, compare the preformance of linear, polynomial and RBF kernels.

```
import sys, os
import matplotlib.pyplot as plt
from sklearn import svm
from sklearn.model_selection import train_test_split
import numpy as np
from sklearn.svm import SVC
import pandas as pd

train_data=pd.read_csv('/content/mnist_train.csv')
test_data=pd.read_csv('/content/mnist_test.csv')

train_data.shape

_> (60000, 785)

test_data.shape

_> (10000, 785)
```

train_data.head()

₽		label	1x1	1x2	1x3	1x4	1x5	1x6	1x7	1x8	1x9	1x10	1x11	1x12	1x13	1x14	1x1
	0	5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	2	4	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	3	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	4	9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	

5 rows × 785 columns

```
test_data.head()
```

C→

	label	1x1	1x2	1x3	1x4	1x5	1x6	1x7	1x8	1x9	1x10	1x11	1x12	1x13	1x14	1x1
0	7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
2	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
3	Ω	Λ	Λ	Λ	Λ	Λ	Λ	Λ	Λ	Λ	Ω	Ω	Λ	Λ	Λ	

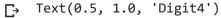
```
x = train_data.iloc[2, 1:]
```

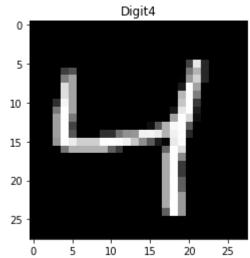
x.shape

x = x.values.reshape(28,28)

plt.imshow(x, cmap='gray')

plt.title("Digit"+str(train_data.iloc[2, 0]))





```
# data types
print(train_data.info())
print(test_data.info())
```

C < class 'pandas.core.frame.DataFrame'>
RangeIndex: 60000 entries, 0 to 59999
Columns: 785 entries, label to 28x28

dtypes: int64(785) memory usage: 359.3 MB

None

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Columns: 785 entries, label to 28x28

dtypes: int64(785) memory usage: 59.9 MB

None

```
## Separating the X and Y variable
y_train = train_data['label']
y_test = test_data['label']
## Dropping the variable 'label' from X variable
```

```
X_train = train_data.drop(columns = 'label')
X test = test data.drop(columns = 'label')
X_{train} = X_{train}/255.0
X_test=X_test/255.0
print("X_train:", X_train.shape)
print("X_test:", X_test.shape)
 C→ X_train: (60000, 784)
     X test: (10000, 784)
# linear model
model linear = SVC(kernel='linear')
model_linear.fit(X_train, y_train)
# predict
y_pred = model_linear.predict(X_test)
# confusion matrix and accuracy
from sklearn import metrics
from sklearn.metrics import confusion matrix
# accuracy
print("accuracy:", metrics.accuracy_score(y_test,y_pred), "\n")
print("Precision:",metrics.precision_score(y_test, y_pred,average='macro'), "\n")
print("Recall:",metrics.recall_score(y_test, y_pred,average='macro'), "\n")
# cm
print(metrics.confusion_matrix(y_true=y_test, y_pred=y_pred))
     accuracy: 0.9404
     Precision: 0.9399179178125057
     Recall: 0.9394223313176161
     [[ 957
                    4
                         1
                                   6
                                                        1]
          0 1122
                    3
                         2
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                   16
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          8
                        25
                             7
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                                        6
                                             7
                                                877
                  6
                                                        8]
          7
               7
                    2
                        11
                             33
                                            18
                                                   5 922]]
                                   4
```

polynomial model

```
model_poly = SVC(kernel='poly')
model_poly.fit(X_train, y_train)
# predict
y_pred_poly = model_poly.predict(X_test)
# confusion matrix and accuracy
# accuracy
print("accuracy:", metrics.accuracy_score(y_test,y_pred_poly), "\n")
print("Precision:",metrics.precision_score(y_test, y_pred_poly,average='macro'), "\n")
print("Recall:",metrics.recall_score(y_test, y_pred_poly,average='macro'), "\n")
# cm
print(metrics.confusion_matrix(y_true=y_test, y_pred=y_pred_poly))
     accuracy: 0.9771
     Precision: 0.9771145861721802
     Recall: 0.9769380249358539
                    2
     [[ 969
                                                        01
                                                   2
          0 1127
                    2
                         1
                              0
                                                        0]
               2 1006
                         0
                              2
                                                        0]
          6
                                   1
               2
                    3
                       984
                              0
                                                   5
                                                        4]
          0
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                         0
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                    0
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      2
                       0
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                                   6 937
                                                        0]
                                            995
          0
              15
                    8
                        1
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                                        0
                                                        81
                    2
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                                        0
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                                                 949
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          1
               1
          3
               6
                             14
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                                              6
                                                   1 969]]
# rbf model
model_rbf = SVC(kernel='rbf')
model_rbf.fit(X_train, y_train)
# predict
y_pred_rbf = model_rbf.predict(X_test)
# confusion matrix and accuracy
# accuracy
print("accuracy:", metrics.accuracy_score(y_test,y_pred_rbf), "\n")
print("Precision:",metrics.precision_score(y_test, y_pred_rbf,average='macro'), "\n")
print("Recall:",metrics.recall_score(y_test, y_pred_rbf,average='macro'), "\n")
# cm
print(metrics.confusion_matrix(y_true=y_test, y_pred=y_pred_rbf))
```

☐ accuracy: 0.9792

Precision: 0.9791973966593345

Recall: 0.9790919842945065

]]	973	0	1	0	0	2	1	1	2	0]
[0	1126	3	1	0	1	1	1	2	0]
[6	1	1006	2	1	0	2	7	6	1]
[0	0	2	995	0	2	0	5	5	1]
[0	0	5	0	961	0	3	0	2	11]
[2	0	0	9	0	871	4	1	4	1]
[6	2	0	0	2	3	944	0	1	0]
[0	6	11	1	1	0	0	996	2	11]
[3	0	2	6	3	2	2	3	950	3]
[3	4	1	7	10	2	1	7	4	970]]