

Support Vector Machine on MNIST Data

1 vs 5 Classification

Import Libraries

```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn import svm
from sklearn.metrics import accuracy_score, classification_report
```

```
c:\Users\Jai Sharma\CondaFiles\envs\ml_env\lib\site-packages\scipy\__init__.py:146: UserWarning: A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy (detected version 1.23.1)
warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}")
```

Helper Functions

```
In [ ]: def symmetry(input):
        """
        compute average symmetry value score for each image
        """
        flipped = np.flip(input,axis=1)
        matching = np.abs(input-flipped)
        return np.mean(matching, axis=1)

def intensity(input):
    """
    compute average intensity value score for each image
    """
    return np.mean(input, axis=1)

def evaluate_model(clf, X_train, y_train, X_test, y_test, train=True):
    """
    function to compute Model Accuracy and Classification Report
    """

    if train:
        pred = clf.predict(X_train)
        clf_report = pd.DataFrame(classification_report(y_train, pred, output_dict=True))
        print("=====")
        print(f"Training Accuracy Score: {accuracy_score(y_train, pred) * 100:.2f}%")
        print("=====")
        print(f"CLASSIFICATION REPORT:\n{clf_report}")
        print("=====")

    elif train==False:
        pred = clf.predict(X_test)
        clf_report = pd.DataFrame(classification_report(y_test, pred, output_dict=True))
        print("=====")
        print(f"Testing Accuracy Score: {accuracy_score(y_test, pred) * 100:.2f}%")
        print("=====")
        print(f"CLASSIFICATION REPORT:\n{clf_report}")
        print("=====")
```

Import Training Data

```
In [ ]: path = "Data/train_pr4.csv"
data = pd.read_csv(path)
X = np.array(data)[:,:1]      # features
y = np.array(data)[:,:0]      # Labels
```

Convert Feature data to Symmetry and Intensity Scores

```
In [ ]: # convert default feature information to intensity and symmetry scores
x_int = intensity(X).reshape(X.shape[0],1)
x_sym = symmetry(X).reshape(X.shape[0],1)

# split data - +1 and -1
xp1, xp2 = x_int[np.where(y==1)],x_sym[np.where(y==1)]
xn1, xn2 = x_int[np.where(y==5)],x_sym[np.where(y==5)]

# training data - features and Labels
```

```
x_train = np.hstack((x_int, x_sym))
y_train = y
```

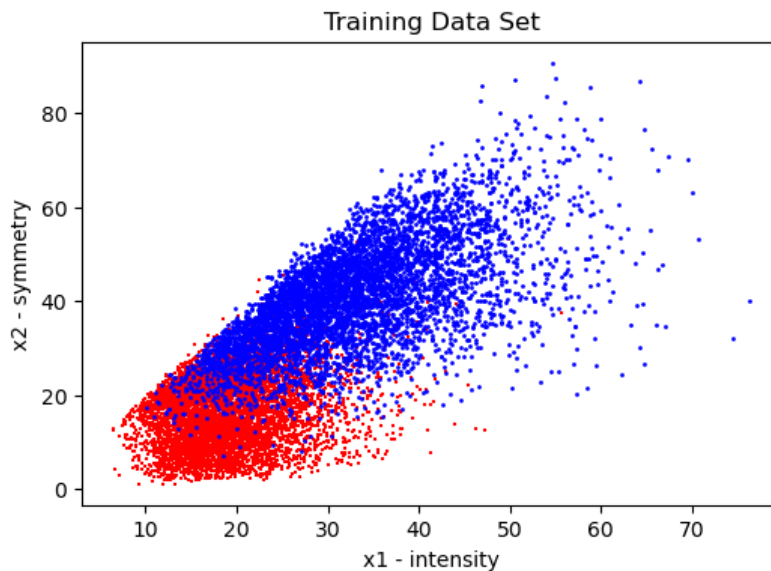
Visualize Training Data

```
In [ ]: # plot the line, the points, and the nearest vectors to the plane
f = plt.figure()
f.set_figwidth(6)
f.set_figheight(4)

# plot +1 and -1 labelled data
plt.scatter(xp1,xp2,c='r',label='1',marker='+', s=1)
plt.scatter(xn1,xn2,c='b',label='5',marker='o',s=1)

# general plot formatting
plt.xlabel("x1 - intensity")
plt.ylabel("x2 - symmetry")
plt.title("Training Data Set")
```

```
Out[ ]: Text(0.5, 1.0, 'Training Data Set')
```



SVM on Training Data

```
In [ ]: clf = svm.SVC(kernel="linear", C=10)
clf.fit(x_train,y_train)

# get parameters --> weights, biase and margins
w = clf.coef_[0]
a = -w[0] / w[1]
margin = 1 / np.sqrt(np.sum(clf.coef_**2))

print("The Weights are --> w1: ", np.round(w[0],3) , " | w2: ", np.round(w[1],3))
print("The Biased term is ---> b: ", np.round(a,3))
print("The Margin term is ---> margin: ", np.round(margin,3))

# Equation of Separator Hyperplane
xx = np.linspace(0,80)
yy = a * xx - (clf.intercept_[0]) / w[1]

# Equation of Decision Boundary
yy_down = yy - np.sqrt(1 + a**2) * margin
yy_up = yy + np.sqrt(1 + a**2) * margin

# get accuracy of training model
print("Linear SVM Model")
evaluate_model(clf, x_train, y_train, 1,1, train=True)
```

The Weights are --> w1: 0.062 | w2: 0.16
The Biased term is ---> b: -0.389
The Margin term is ---> margin: 5.829
Linear SVM Model

Training Accuracy Score: 90.71%

CLASSIFICATION REPORT:

	1	5	accuracy	macro avg	weighted avg
precision	0.902467	0.913312	0.907095	0.907889	0.907300
recall	0.933254	0.874562	0.907095	0.903908	0.907095
f1-score	0.917602	0.893517	0.907095	0.905560	0.906868
support	6742.000000	5421.000000	0.907095	12163.000000	12163.000000

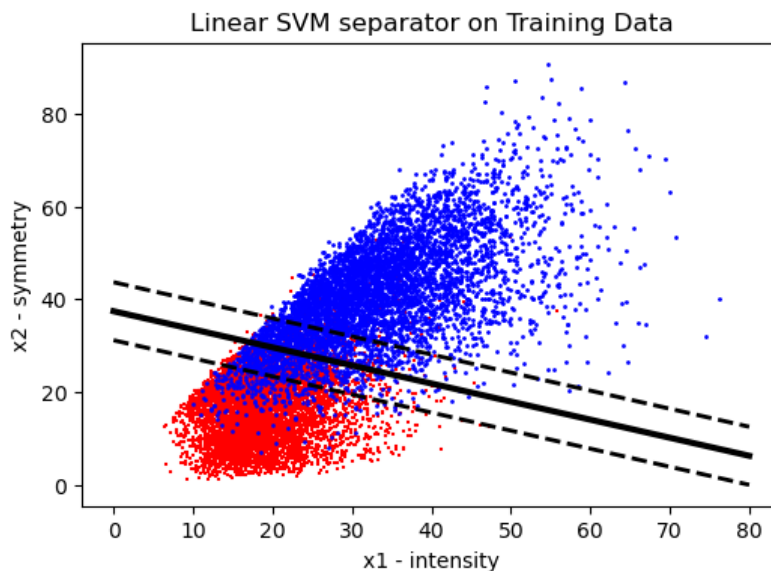
```
In [ ]: # plot the line, the points, and the nearest vectors to the plane
f = plt.figure()
f.set_figwidth(6)
f.set_figheight(4)

# plot +1 and -1 labelled data
plt.scatter(xp1,xp2,c='r',label='1',marker='+', s=1)
plt.scatter(xn1,xn2,c='b',label='5',marker='o',s=1)

# plot separator, margins and decision boundary
plt.plot(xx, yy, "k-", label='separator', linewidth = 3)
plt.plot(xx, yy_down, "k--",linewidth = 2)
plt.plot(xx, yy_up, "k--",linewidth = 2)

# general plot formatting
plt.xlabel("x1 - intensity")
plt.ylabel("x2 - symmetry")
plt.title("Linear SVM separator on Training Data")
```

Out[]: Text(0.5, 1.0, 'Linear SVM separator on Training Data')



Polynomial SVM on Training Data

```
In [ ]: model = svm.SVC(kernel='poly', degree=2)
model.fit(x_train, y_train)

# Create grid to evaluate model
xx_poly = yy_poly = np.linspace(0,80)
YY, XX = np.meshgrid(yy_poly, xx_poly)
xy_poly = np.vstack([XX.ravel(), YY.ravel()]).T

# compute separator and decision boundaries
Z = model.decision_function(xy_poly).reshape(XX.shape)

# get accuracy of testing model
print("Polynomial SVM Model [N = 2]")
evaluate_model(model, x_train, y_train, _, _, train=True)
```

Polynomial SVM Model [N = 2]

Training Accuracy Score: 90.83%

CLASSIFICATION REPORT:

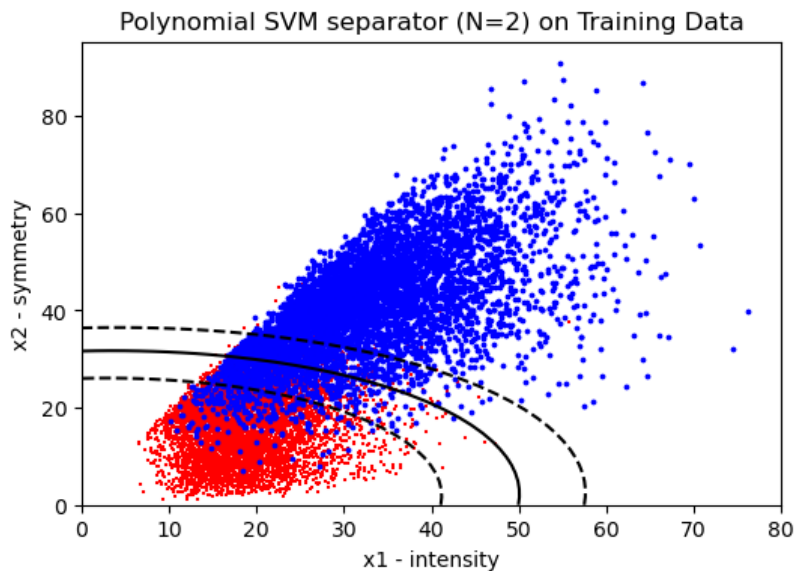
	1	5	accuracy	macro avg	weighted avg
precision	0.898456	0.921991	0.908329	0.910224	0.908946
recall	0.940967	0.867737	0.908329	0.904352	0.908329
f1-score	0.919220	0.894042	0.908329	0.906631	0.907998
support	6742.000000	5421.000000	0.908329	12163.000000	12163.000000

```
In [ ]: # plot trained polynomial SVM fit on training data
fig, ax = plt.subplots(figsize=(6,4))

# plot +1 and -1 labelled data
plt.scatter(xp1,xp2,c='r',label='1',marker='+', s=3)
plt.scatter(xn1,xn2,c='b',label='5',marker='o', s=3)

# plot separator, margins and decision boundary
ax.contour(XX, YY, Z, colors='k', levels=[-1, 0, 1], alpha=1, linestyles=['--', '-', '--'])

# general plot formatting
plt.xlabel("x1 - intensity")
plt.ylabel("x2 - symmetry")
plt.title("Polynomial SVM separator (N=2) on Training Data")
plt.show()
```



SVM on Testing Data

```
In [ ]: path = "Data/test_pr4.csv"
data = pd.read_csv(path)
X = np.array(data[:,1:]) # features
y = np.array(data[:,0]) # Labels

# convert default feature information to intensity and symmetry scores
x_int = intensity(X).reshape(X.shape[0],1)
x_sym = symmetry(X).reshape(X.shape[0],1)

# split data - +1 and -1
xp1, xp2 = x_int[np.where(y==1)], x_sym[np.where(y==1)]
xn1, xn2 = x_int[np.where(y==5)], x_sym[np.where(y==5)]

# training data - features and labels
x_test = np.hstack((x_int, x_sym))
y_test = y

# get accuracy of testing model
print("Linear SVM Model")
evaluate_model(clf, x_train, y_train, x_test, y_test, train=False)
print(" ")
print(" ")
print(" ")
```

```
print("Polynomial SVM Model [N = 2]")
# get accuracy of testing model
evaluate_model(model, x_train, y_train, x_test, y_test, train=False)
```

Linear SVM Model

Testing Accuracy Score: 91.51%

CLASSIFICATION REPORT:

	1	5	accuracy	macro avg	weighted avg
precision	0.916162	0.913793	0.915146	0.914978	0.915120
recall	0.933921	0.891256	0.915146	0.912588	0.915146
f1-score	0.924956	0.902384	0.915146	0.913670	0.915023
support	1135.000000	892.000000	0.915146	2027.000000	2027.000000

Polynomial SVM Model [N = 2]

Testing Accuracy Score: 91.66%

CLASSIFICATION REPORT:

	1	5	accuracy	macro avg	weighted avg
precision	0.912116	0.922807	0.916626	0.917462	0.916821
recall	0.941850	0.884529	0.916626	0.913190	0.916626
f1-score	0.926745	0.903263	0.916626	0.915004	0.916411
support	1135.000000	892.000000	0.916626	2027.000000	2027.000000

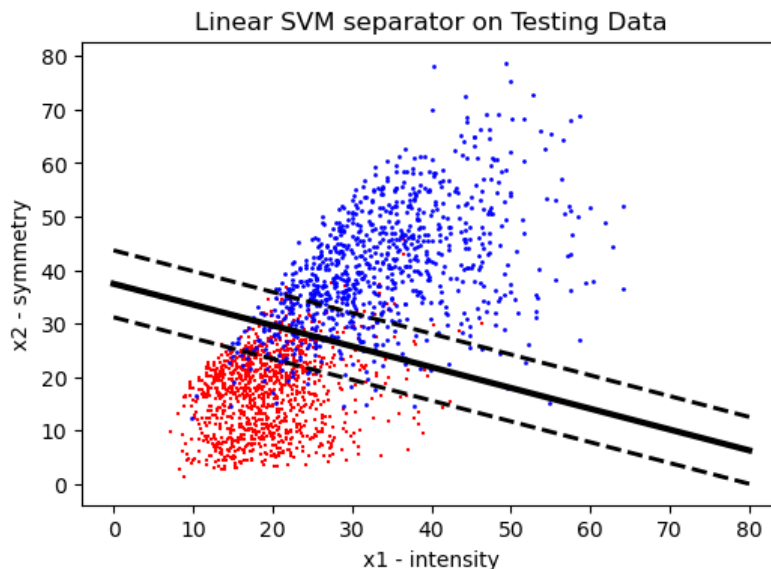
Visualize Test Data with Trained SVM hyperplane

```
In [ ]: # plot the line, the points, and the nearest vectors to the plane
f = plt.figure()
f.set_figwidth(6)
f.set_figheight(4)
# plot +1 and -1 labelled data
plt.scatter(xp1, xp2, c='r', label='1', marker='+', s=1)
plt.scatter(xn1, xn2, c='b', label='5', marker='o', s=1)

# plot separator, margins and decision boundary
plt.plot(xx, yy, "k-", label='separator', linewidth = 3)
plt.plot(xx, yy_down, "k--", linewidth = 2)
plt.plot(xx, yy_up, "k--", linewidth = 2)

# general plot formatting
plt.xlabel("x1 - intensity")
plt.ylabel("x2 - symmetry")
plt.title("Linear SVM separator on Testing Data")
```

Out[]: Text(0.5, 1.0, 'Linear SVM separator on Testing Data')



```
In [ ]: # plot the line, the points, and the nearest vectors to the plane
f = plt.figure()
f.set_figwidth(6)
```

```
f.set_figheight(4)

# plot +1 and -1 labelled data
plt.scatter(xp1,xp2,c='r',label='1',marker='+', s=2)
plt.scatter(xn1,xn2,c='b',label='5',marker='o',s=2)

# plot separator, margins and decision boundary
plt.contour(XX, YY, Z, colors='k', levels=[-1, 0, 1], alpha=1.0, linestyles=['--', '-', '--'])

# general plot formatting
plt.xlabel("x1 - intensity")
plt.ylabel("x2 - symmetry")
plt.title("Polynomial SVM separator (N=2) on Testing Data")
```

Out[]: Text(0.5, 1.0, 'Polynomial SVM separator (N=2) on Testing Data')

