

## Support Vector Classifier Implementation

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: ## Lets create synthetic data points
from sklearn.datasets import make_classification
```

[illegible]

In [4]: X

```
Out[4]: array([[ -1.04543182,   1.43917109],
 [  2.02614712,  -0.96344906],
 [  0.52477663,  -0.9804843 ],
 ...,
 [  1.74491441,   0.63550927],
 [-0.79715152,  -1.09210497],
 [  0.37480977,   1.16665105]])
```

```
In [5]: y
```

[illegible]

```
In [6]: pd.DataFrame(X)[0]
```

```
Out[6]: 0    -1.045432
        1     2.026147
        2     0.524777
        3    -0.433085
        4     0.836881
        ...
        995   -1.087434
        996    1.666602
        997    1.744914
        998   -0.797152
        999    0.374810
        Name: 0, Length: 1000, dtype: float64
```

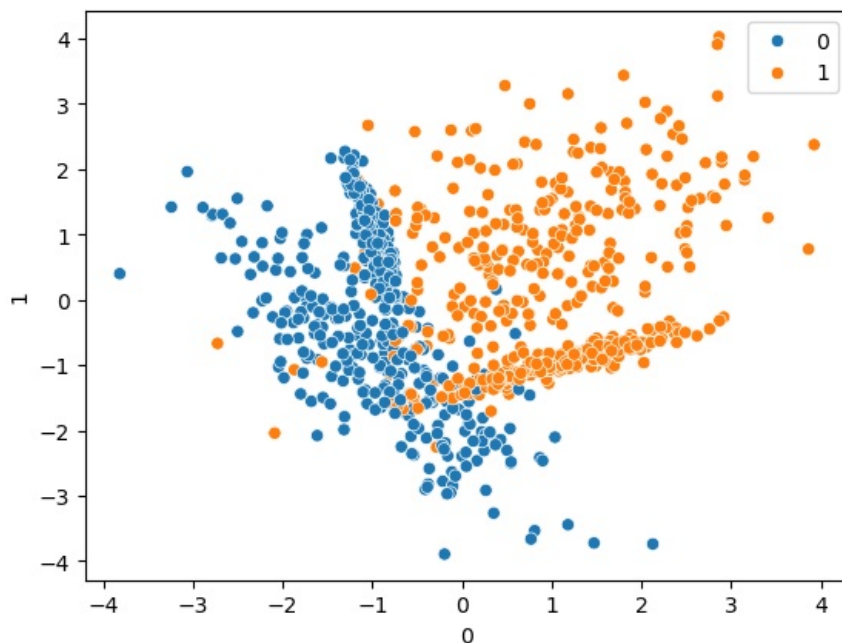
```
In [7]: # sns.scatterplot(pd.DataFrame(X)[0],pd.DataFrame(X)[1],hue=y)
import seaborn as sns
import pandas as pd

# Assuming X is your data and y is the variable for coloring
df = pd.DataFrame(X) # Convert X to a DataFrame if not already
print(df)
sns.scatterplot(x=df[0], y=df[1], hue=y)
```

```
      0      1
0  -1.045432  1.439171
1   2.026147 -0.963449
2   0.524777 -0.980484
3  -0.433085 -1.073359
4   0.836881 -0.906425
..      ...
995 -1.087434  1.214817
996  1.666602 -0.783846
997  1.744914  0.635509
998 -0.797152 -1.092105
999  0.374810  1.166651
```

```
[1000 rows x 2 columns]
```

```
Out[7]: <Axes: xlabel='0', ylabel='1'>
```



```
In [8]: from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.25,random_state=10)
```

```
In [10]: from sklearn.svm import SVC
```

```
In [11]: svc=SVC(kernel='linear')
```

```
In [12]: svc.fit(X_train,y_train)
```

```
Out[12]: SVC
SVC(kernel='linear')
```

```
In [13]: svc.coef_
```

```
Out[13]: array([[2.15359963, 0.48784076]])
```

```
In [14]: ## Prediction
y_pred=svc.predict(X_test)
```

```
In [15]: from sklearn.metrics import classification_report,confusion_matrix
```

```
In [16]: print(classification_report(y_test,y_pred))
print(confusion_matrix(y_test,y_pred))
```

	precision	recall	f1-score	support
0	0.88	0.93	0.91	122
1	0.93	0.88	0.91	128
accuracy			0.91	250
macro avg	0.91	0.91	0.91	250
weighted avg	0.91	0.91	0.91	250

[[114 8]
[ 15 113]]

```
In [17]: rbf=SVC(kernel='rbf')
```

```
In [18]: rbf.fit(X_train,y_train)
```

```
Out[18]: SVC
```

SVC()

```
In [19]: ## Prediction
y_pred1=rbf.predict(X_test)
```

```
In [20]: print(classification_report(y_test,y_pred1))
print(confusion_matrix(y_test,y_pred1))
```

	precision	recall	f1-score	support
0	0.91	0.98	0.94	122
1	0.97	0.91	0.94	128
accuracy			0.94	250
macro avg	0.94	0.94	0.94	250
weighted avg	0.94	0.94	0.94	250

[[119 3]
[ 12 116]]

```
In [21]: polynomial=SVC(kernel='poly')
polynomial.fit(X_train,y_train)
## Prediction
y_pred2=polynomial.predict(X_test)
print(classification_report(y_test,y_pred2))
print(confusion_matrix(y_test,y_pred2))
```

	precision	recall	f1-score	support
0	0.91	0.96	0.94	122
1	0.96	0.91	0.94	128
accuracy			0.94	250
macro avg	0.94	0.94	0.94	250
weighted avg	0.94	0.94	0.94	250

[[117 5]
[ 11 117]]

```
In [22]: sigmoid=SVC(kernel='sigmoid')
sigmoid.fit(X_train,y_train)
## Prediction
y_pred3=sigmoid.predict(X_test)
print(classification_report(y_test,y_pred3))
print(confusion_matrix(y_test,y_pred3))
```

	precision	recall	f1-score	support
0	0.78	0.84	0.81	122
1	0.83	0.78	0.81	128
accuracy			0.81	250
macro avg	0.81	0.81	0.81	250
weighted avg	0.81	0.81	0.81	250

```
[[102 20]
 [ 28 100]]
```

```
In [ ]: sigmoid.intercept_
```

## Hyperparameter Tuning With SVC

```
In [24]: from sklearn.model_selection import GridSearchCV
```

```
# defining parameter range
param_grid = {'C': [0.1, 1, 10, 100, 1000],
              'gamma': [1, 0.1, 0.01, 0.001, 0.0001],
              'kernel': ['rbf']}
```

```
In [25]: grid=GridSearchCV(SVC(),param_grid=param_grid,refit=True,cv=5,verbose=3)
```

```
In [26]: grid.fit(X_train,y_train)
```

```
Fitting 5 folds for each of 25 candidates, totalling 125 fits
[CV 1/5] END .....C=0.1, gamma=1, kernel=rbf;, score=0.920 total time= 0.0s
[CV 2/5] END .....C=0.1, gamma=1, kernel=rbf;, score=0.940 total time= 0.0s
[CV 3/5] END .....C=0.1, gamma=1, kernel=rbf;, score=0.927 total time= 0.0s
[CV 4/5] END .....C=0.1, gamma=1, kernel=rbf;, score=0.933 total time= 0.0s
[CV 5/5] END .....C=0.1, gamma=1, kernel=rbf;, score=0.933 total time= 0.0s
[CV 1/5] END .....C=0.1, gamma=0.1, kernel=rbf;, score=0.913 total time= 0.0s
[CV 2/5] END .....C=0.1, gamma=0.1, kernel=rbf;, score=0.933 total time= 0.0s
[CV 3/5] END .....C=0.1, gamma=0.1, kernel=rbf;, score=0.900 total time= 0.0s
[CV 4/5] END .....C=0.1, gamma=0.1, kernel=rbf;, score=0.907 total time= 0.0s
[CV 5/5] END .....C=0.1, gamma=0.1, kernel=rbf;, score=0.913 total time= 0.0s
[CV 1/5] END .....C=0.1, gamma=0.01, kernel=rbf;, score=0.853 total time= 0.0s
[CV 2/5] END .....C=0.1, gamma=0.01, kernel=rbf;, score=0.847 total time= 0.0s
[CV 3/5] END .....C=0.1, gamma=0.01, kernel=rbf;, score=0.880 total time= 0.0s
[CV 4/5] END .....C=0.1, gamma=0.01, kernel=rbf;, score=0.827 total time= 0.0s
[CV 5/5] END .....C=0.1, gamma=0.01, kernel=rbf;, score=0.880 total time= 0.0s
[CV 1/5] END .....C=0.1, gamma=0.001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 2/5] END .....C=0.1, gamma=0.001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 3/5] END .....C=0.1, gamma=0.001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 4/5] END .....C=0.1, gamma=0.001, kernel=rbf;, score=0.507 total time= 0.0s
[CV 5/5] END .....C=0.1, gamma=0.001, kernel=rbf;, score=0.507 total time= 0.0s
[CV 1/5] END .....C=0.1, gamma=0.0001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 2/5] END .....C=0.1, gamma=0.0001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 3/5] END .....C=0.1, gamma=0.0001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 4/5] END .....C=0.1, gamma=0.0001, kernel=rbf;, score=0.507 total time= 0.0s
[CV 5/5] END .....C=0.1, gamma=0.0001, kernel=rbf;, score=0.507 total time= 0.0s
[CV 1/5] END .....C=1, gamma=1, kernel=rbf;, score=0.927 total time= 0.0s
[CV 2/5] END .....C=1, gamma=1, kernel=rbf;, score=0.933 total time= 0.0s
[CV 3/5] END .....C=1, gamma=1, kernel=rbf;, score=0.933 total time= 0.0s
[CV 4/5] END .....C=1, gamma=1, kernel=rbf;, score=0.947 total time= 0.0s
[CV 5/5] END .....C=1, gamma=1, kernel=rbf;, score=0.947 total time= 0.0s
[CV 1/5] END .....C=1, gamma=0.1, kernel=rbf;, score=0.920 total time= 0.0s
[CV 2/5] END .....C=1, gamma=0.1, kernel=rbf;, score=0.947 total time= 0.0s
[CV 3/5] END .....C=1, gamma=0.1, kernel=rbf;, score=0.920 total time= 0.0s
[CV 4/5] END .....C=1, gamma=0.1, kernel=rbf;, score=0.940 total time= 0.0s
[CV 5/5] END .....C=1, gamma=0.1, kernel=rbf;, score=0.933 total time= 0.0s
[CV 1/5] END .....C=1, gamma=0.01, kernel=rbf;, score=0.900 total time= 0.0s
[CV 2/5] END .....C=1, gamma=0.01, kernel=rbf;, score=0.920 total time= 0.0s
[CV 3/5] END .....C=1, gamma=0.01, kernel=rbf;, score=0.893 total time= 0.0s
[CV 4/5] END .....C=1, gamma=0.01, kernel=rbf;, score=0.900 total time= 0.0s
[CV 5/5] END .....C=1, gamma=0.01, kernel=rbf;, score=0.913 total time= 0.0s
[CV 1/5] END .....C=1, gamma=0.001, kernel=rbf;, score=0.853 total time= 0.0s
[CV 2/5] END .....C=1, gamma=0.001, kernel=rbf;, score=0.847 total time= 0.0s
[CV 3/5] END .....C=1, gamma=0.001, kernel=rbf;, score=0.880 total time= 0.0s
[CV 4/5] END .....C=1, gamma=0.001, kernel=rbf;, score=0.827 total time= 0.0s
[CV 5/5] END .....C=1, gamma=0.001, kernel=rbf;, score=0.880 total time= 0.0s
[CV 1/5] END .....C=1, gamma=0.0001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 2/5] END .....C=1, gamma=0.0001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 3/5] END .....C=1, gamma=0.0001, kernel=rbf;, score=0.500 total time= 0.0s
[CV 4/5] END .....C=1, gamma=0.0001, kernel=rbf;, score=0.507 total time= 0.0s
[CV 5/5] END .....C=1, gamma=0.0001, kernel=rbf;, score=0.507 total time= 0.0s
[CV 1/5] END .....C=10, gamma=1, kernel=rbf;, score=0.927 total time= 0.0s
[CV 2/5] END .....C=10, gamma=1, kernel=rbf;, score=0.933 total time= 0.0s
[CV 3/5] END .....C=10, gamma=1, kernel=rbf;, score=0.927 total time= 0.0s
```

Out[26]:



```
grid.best_params_
```

```
Out[27]: {'C': 100, 'gamma': 0.1, 'kernel': 'rbf'}
```

```
In [28]: ## Prediction
y_pred4=grid.predict(X_test)
print(classification_report(y_test,y_pred4))
print(confusion_matrix(y_test,y_pred4))
```

	precision	recall	f1-score	support
0	0.92	0.97	0.94	122
1	0.97	0.92	0.94	128
accuracy			0.94	250
macro avg	0.94	0.94	0.94	250
weighted avg	0.95	0.94	0.94	250

```
[[118  4]
 [ 10 118]]
```

```
In [ ]:
```