

## Support Vector Classifier Implementation

```
In [1]: import pandas as pd  
import numpy as np  
import matplotlib.p  
import seaborn as s
```

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

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

```
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
```

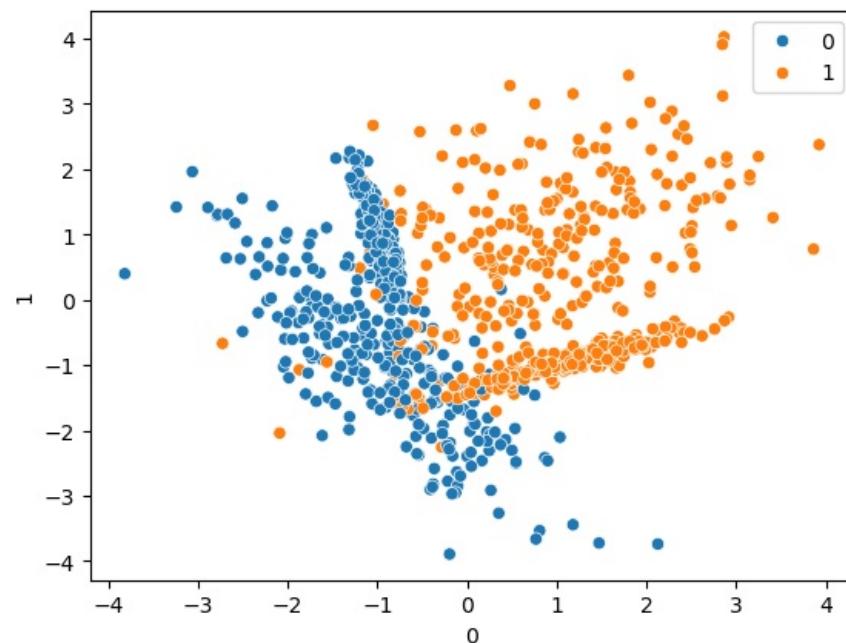
```
In [7]: # sns.scatterplot(pd.DataFrame(X)[0],pd.DataFrame(X)[1],hue=y)
import seaborn as sns
import pandas as pd

# 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      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=0.507 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]:

- ▶ **GridSearchCV** (1) (?)
- ▶ **best\_estimator\_**: SVC

```
In [27]: 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 [ ]:
```