

SVM Kernels Indepth Intuition And Practical Explanation

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

```
In [2]: x = np.linspace(-5.0,5.0, 100)
y = np.sqrt(10**2 - x**2)
```

```
In [3]: x.shape
```

```
Out[3]: (100,)
```

```
In [4]: y.shape
```

```
Out[4]: (100,)
```

```
In [5]: y = np.hstack([y,-y])
x = np.hstack([x,-x])
```

```
In [6]: x.shape
```

```
Out[6]: (200,)
```

```
In [7]: y.shape
```

```
Out[7]: (200,)
```

```
In [8]: x1 = np.linspace(-5.0, 5.0, 100)
y1 = np.sqrt(5**2 - x1**2)
y1=np.hstack([y1,-y1])
x1=np.hstack([x1,-x1])
```

```
In [9]: y1.shape
```

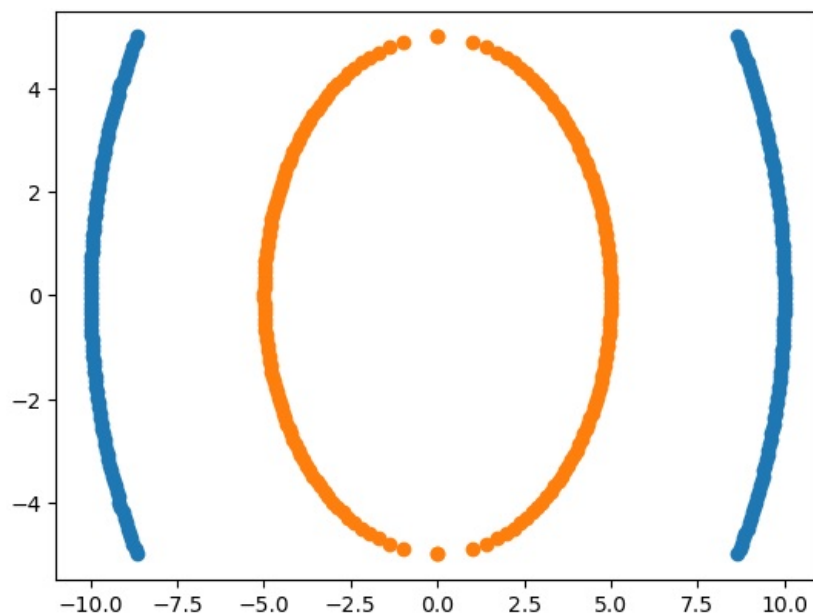
```
Out[9]: (200,)
```

```
In [10]: x1.shape
```

```
Out[10]: (200,)
```

```
In [11]: plt.scatter(y,x)
plt.scatter(y1,x1)
```

```
Out[11]: <matplotlib.collections.PathCollection at 0x25935ba63d0>
```



```
In [ ]:
```

```
In [12]: # Change a DataFrame (x,y)
df1 = pd.DataFrame(np.vstack([y,x]).T,columns = ['X1','X2'])
df1['Y'] = 0
```

```
# Change a DataFrame (x1,y1)
df2 = pd.DataFrame(np.vstack([y1,x1]).T,columns = ['X1','X2'])
df2['Y'] = 1
```

```
In [13]: #append the two dataframe(df1,df2)
df = df1.append(df2)
df.head()
```

C:\Users\dines\AppData\Local\Temp\ipykernel_13024\1397825995.py:2: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.

```
df = df1.append(df2)
```

```
Out[13]:
```

	X1	X2	Y
0	8.660254	-5.000000	0
1	8.717792	-4.898999	0
2	8.773790	-4.79798	0
3	8.828277	-4.69697	0
4	8.881281	-4.59596	0

```
In [14]: x = df.iloc[:, :2]
y = df['Y']
```

```
In [15]: x.head()
```

```
Out[15]:
```

	X1	X2
0	8.660254	-5.000000
1	8.717792	-4.898999
2	8.773790	-4.79798
3	8.828277	-4.69697
4	8.881281	-4.59596

```
In [16]: y
```

```
Out[16]:
```

0	0
1	0
2	0
3	0
4	0
...	...
195	1
196	1
197	1
198	1
199	1

Name: Y, Length: 400, dtype: int64

```
In [17]: #Split the dataset into train test
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 = 0)
```

```
In [18]: x_train.shape
```

```
Out[18]: (300, 2)
```

```
In [19]: y_train.shape
```

```
Out[19]: (300,)
```

```
In [20]: x_test.shape
```

```
Out[20]: (100, 2)
```

```
In [21]: x_test.shape
```

```
Out[21]: (100, 2)
```

```
In [ ]:
```

Base_Model building

```
In [22]: from sklearn.svm import SVC
```

```
In [23]: cls = SVC(kernel = "rbf")
cls.fit(x_train,y_train)
```

Out[23]: SVC

SVC()

In [24]: *#Evaluation Metrics*
from sklearn.metrics **import** accuracy_score
y_pred = cls.predict(x_test)
accuracy_score(y_test,y_pred)

Out[24]: 1.0

In []:

In [25]: df.head()

Out[25]:

	X1	X2	Y
0	8.660254	-5.00000	0
1	8.717792	-4.89899	0
2	8.773790	-4.79798	0
3	8.828277	-4.69697	0
4	8.881281	-4.59596	0

In []:

Polynomial Kernel

In [26]: *# We need to find components for the Polynomial Kernel*
*#X1,X2,X1_square,X2_square,X1*X2*
df['X1_Square']= df['X1']**2
df['X2_Square']= df['X2']**2
df['X1*X2'] = (df['X1'] *df['X2'])
df.head()

Out[26]:

	X1	X2	Y	X1_Square	X2_Square	X1*X2
0	8.660254	-5.00000	0	75.000000	25.000000	-43.301270
1	8.717792	-4.89899	0	75.999898	24.000102	-42.708375
2	8.773790	-4.79798	0	76.979390	23.020610	-42.096467
3	8.828277	-4.69697	0	77.938476	22.061524	-41.466150
4	8.881281	-4.59596	0	78.877155	21.122845	-40.818009

In [27]: df['X1'].shape

Out[27]: (400,)

In [28]: *# Independent and Dependent features*
x = df[['X1', 'X2', 'X1_Square', 'X2_Square', 'X1*X2']]
y = df['Y']

In [29]: y

Out[29]:

0	0
1	0
2	0
3	0
4	0
...	
195	1
196	1
197	1
198	1
199	1

Name: Y, Length: 400, dtype: int64

In [30]: *#train_test_split*
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = 0.25,random_state = 0)

In [31]: x_train.shape

Out[31]: (300, 5)

In [32]: y_train.shape

Out[32]: (300,)

```
In [33]: #import poltly
import plotly.express as px

fig = px.scatter_3d(df, x="X1", y="X2", z="X1*X2", color='Y')
fig.show()
```

[more info](#)



```
In [34]: fig = px.scatter_3d(df, x="X1_Square", y="X1_Square", z="X1*X2", color='Y')
fig.show()
```

```
In [35]: classifier = SVC(kernel="linear")
classifier.fit(x_train, y_train)
y_pred = classifier.predict(x_test)
accuracy_score(y_test, y_pred)
```

Out[35]: 1.0

In []:

