

# Kernel-Support-Vector-Machine

January 14, 2025

```
[ ]: '''  
    Kernel SVM -->  
  
    A Kernel Support Vector Machine (SVM) is an extension of the traditional  
    ↪Support Vector Machine  
    that allows it to handle data that is not linearly separable. By using  
    ↪kernel functions,  
    the SVM maps the input data into a higher-dimensional space where a linear  
    ↪separator (hyperplane) can be found.  
  
    Kernel -->  
  
    In Support Vector Machines (SVM), the kernel is a mathematical function  
    ↪that transforms the input data  
    into a higher-dimensional space, making it easier to find a hyperplane that  
    ↪separates the data into different classes.  
  
    Why is a kernel needed ?  
  
    SVMs aim to find the optimal hyperplane that separates data points of  
    ↪different classes.  
    However, not all datasets are linearly separable in their original feature  
    ↪space.  
    The kernel trick helps SVMs tackle such datasets by projecting them into a  
    ↪higher-dimensional space  
    where they can be separated linearly.  
    '''
```

```
[ ]: '''  
    Key Concepts -->  
  
    Support Vectors :  
    Data points that lie closest to the decision boundary (margin) and  
    ↪influence its position and orientation.  
  
    Kernel Trick :
```

Instead of explicitly transforming data to a higher-dimensional space, the  
→SVM uses a kernel  
function to compute the dot product of data points in that space, saving  
→computational resources.

*Objective :*

Maximize the margin (distance between the hyperplane and the nearest data  
→points) while correctly  
classifying the data.

*Hyperparameters :*

*C (Regularization Parameter):* Controls the trade-off between maximizing the  
→margin and minimizing  
classification error.

*Kernel Type :*

Defines the transformation applied to the input data.

'''

[ ]: '''

*Advantages of Kernel SVM -->*

*Handles both linearly and non-linearly separable data.  
Versatile with various kernels for different datasets.  
High performance in small to medium-sized datasets.*

*Disadvantages of Kernel SVM -->*

*Computationally intensive for large datasets.  
Requires careful selection of kernel and hyperparameters (C, ).  
Sensitive to noise and outliers.*

'''

[1]: # Import Libraries -->

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report,
→confusion_matrix
```

[2]: # Importing Dataset -->

```
data = pd.read_csv('Data/Social_Network_Ads.csv')
```

```
data.head(10)
```

```
[2]:   Age  EstimatedSalary  Purchased
     0   19             19000           0
     1   35             20000           0
     2   26             43000           0
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     5   27             58000           0
     6   27             84000           0
     7   32            150000           1
     8   25             33000           0
     9   35             65000           0
```

```
[3]: x_data = data.iloc[:, :-1].values
     y_data = data.iloc[:, -1].values
```

```
[4]: #   Splitting The Dataset -->

     x_train, x_test, y_train, y_test = train_test_split(x_data, y_data, test_size=0.
     ↪2, random_state=42)
```

```
[5]: x_train
```

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```

```
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```

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```

```
[ ]: # Applying Feature Scaling -->
```

```
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
```

```
[8]: x_train
```

```
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 [-0.4780302 , -0.85177573],



```

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[-1.16487283, -1.02631249],
[ 0.50317355,  1.82445454],
[ 0.11069205,  0.19544481],
[-0.57615058,  0.45724994]])

```

```
[9]: x_test
```

```

[9]: array([[ 0.79753468, -1.40447546],
 [ 2.07309956,  0.51542886],
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 [ 0.99377543,  0.74814454],
 [-0.87051171, -1.22993871],
 [-0.77239133, -0.24089709],
 [ 0.89565505,  1.06812859],
 [-0.87051171,  0.36998156],
 [ 0.20881242,  0.13726589],
 [ 0.40505317, -0.15362871],
 [-0.28178945, -0.15362871],

```

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 [ 0.01257167, -0.4445233 ],  
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 [ 2.17121993, 0.92268129],  
 [-1.16487283, 1.38811264],

```

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[ 1.38625693,  0.57360778],
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[ 0.11069205, -0.32816546],
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[-1.85171546,  0.42816048],
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[ 1.58249768, -1.28811763],
[-0.28178945, -0.67723898],
[-0.0855487 ,  0.22453427]])

```

```
[10]: # Building The Model -->
```

```

model = SVC(kernel='rbf', random_state=42)
model.fit(x_train, y_train)

```

```
[10]: SVC(random_state=42)
```

```
[12]: y_pred = model.predict(x_test)
y_pred
```

```
[12]: array([1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0,
          1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1,
          0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1,
          1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0], dtype=int64)
```

```
[13]: # Accuracy -->
```

```

acc_score = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
class_report = classification_report(y_test, y_pred)

```

```
[14]: print("Accuracy Score --> ", acc_score)
```

```
Accuracy Score --> 0.925
```

```
[15]: print("Confusion Matrix -->\n\n", conf_matrix)
```

Confusion Matrix -->

```
[[47  5]
 [ 1 27]]
```

```
[16]: print("Classification Report -->\n\n", class_report)
```

Classification Report -->

	precision	recall	f1-score	support
0	0.98	0.90	0.94	52
1	0.84	0.96	0.90	28
accuracy			0.93	80
macro avg	0.91	0.93	0.92	80
weighted avg	0.93	0.93	0.93	80