Support-Vector-Machine

January 14, 2025

[ ]:
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'''
   Support Vector Machine -->

   Support Vector Machine (SVM) is a supervised machine learning algorithm␣
↪used for
   classification, regression, and outlier detection tasks. Its primary goal␣
↪is to
   find the optimal hyperplane that best separates the classes in the feature␣
↪space.
   For datasets that are not linearly separable, SVM employs kernel functions␣
↪to project
   the data into a higher-dimensional space, making it easier to find a␣
↪decision boundary.
'''
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[ ]:
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'''
   How SVM Works -->

   Hyperplane :

   A hyperplane is a decision boundary that separates the data points of␣
↪different classes.
   In 2D, it's a line; in 3D, it's a plane, and in higher dimensions, it's a␣
↪hyperplane.

   Support Vectors :

   The data points closest to the hyperplane are called support vectors.
   These points are critical as they influence the position and orientation of␣
↪the hyperplane.

   Maximizing the Margin :

   SVM tries to find the hyperplane that maximizes the margin (distance)␣
↪between the nearest points of the classes.

   Kernels for Nonlinear Data :
```

```
    If data is not linearly separable, SVM uses a kernel function to transform␣
 ↪the data into
    a higher-dimensional space where a linear hyperplane can be applied.

    Common kernel types :

    Linear Kernel
    Polynomial Kernel
    Radial Basis Function (RBF) or Gaussian Kernel
    Sigmoid Kernel
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'''
    Advantages of SVM -->

    Works well in high-dimensional spaces.
    Effective for linearly separable and non-linearly separable data.
    Robust to overfitting, especially for smaller datasets.

    Disadvantages of SVM -->

    Computationally expensive for large datasets.
    Choice of kernel and hyperparameters can significantly affect performance.
    Does not work well with noisy data or overlapping classes.

    Applications of SVM -->

    Image classification
    Text classification (e.g., spam detection)
    Bioinformatics (e.g., cancer detection)
    Handwriting recognition
'''
```

[11]:
```
#   Importing 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
3	27	57000	0
4	19	76000	0
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.
↪25, random_state=42)
```

[5]:
```
x_train
```

[5]:
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```

[6]:
```
y_train
```

[6]:
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```

[7]:
```
#   Feature Scaling -->

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

[12]:
```
x_train
```

[12]:
```
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```

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[ 0.812419  , -0.3184001 ],
```

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[-0.169554  ,  1.40504997],
[ 2.1871812 , -0.8149874 ],
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[ 1.3034055 ,  1.87242626],
[ 1.1070109 , -1.22394166],
```

```
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       [-1.74071081,  0.11976517],
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       [-1.15152701, -1.01946453],
       [ 0.5178271 ,  1.84321524],
       [ 0.1250379 ,  0.20739823],
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```

[13]:
```
x_test
```

[13]:
```
array([[ 0.812419  , -1.39920777],
       [ 2.0889839 ,  0.52871943],
       [-0.95513241, -0.75656537],
```

```
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[-0.85693511, -0.25997806],
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[-0.36594861, -0.78577639],
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```
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```

```
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```

[8]:
```
#   Building The Model -->

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

[8]:
```
SVC(kernel='linear', random_state=42)
```

[ ]:
```
#   Predicting Results -->

y_pred = model.predict(x_test)
y_pred
```

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

[14]:
```
#   Accuracy Score, Confusion Matrix and Classification Report -->

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

[16]:
```
print("Accuracy Score --> ",acc_score)
```

```
Accuracy Score -->  0.86
```

[18]:
```
print("Confusion Matrix -->\n\n", conf_matrix)
```

```
Confusion Matrix -->

 [[61  2]
 [12 25]]
```

[19]:
```
print("Classification Report -->\n\n", class_report)
```

```
Classification Report -->

              precision    recall  f1-score   support

           0       0.84      0.97      0.90        63
           1       0.93      0.68      0.78        37

    accuracy                           0.86       100
```

```
   macro avg       0.88      0.82      0.84       100
weighted avg       0.87      0.86      0.85       100
```