

A python Program to implement SVM Classifier Model

Aim :

To implement a SVM classifier model using python and determine its accuracy.

Algorithm :

Step 1:- Import Necessary libraries

1. Import numpy as np
2. Import pandas as pd
3. Import svm from sklearn
4. Import matplotlib.pyplot as plt.
5. Import seaborn as sns
6. Set the font-scale attribute to 1.2 in seaborn

Step 2:- Load and Display Dataset

1. Read the dataset (moffin.csv) using pd.read_csv().
2. ~~Display the first five instances using the head() function.~~

Step 3:- Plot Initial Data.

1. Use the sns.heatmap() function.
2. Set the x and y axes to "Sugar" and "Flour".

3. Assign "Type" to the hue parameter
4. Set the palette to "Set 1".
- 5) Set fit_intercept to false
- 6) Plot the Graph

Step 4: Prepare Data for SVM

1. Extract "Sugar" and "Butter" columns from the recipe dataset and assign to variables "sugar_butter".
2. Create a new variable 'type_label'.
3. For each value in the 'Type' column assign 0 if it's a "Muffin" and 1 otherwise

Step 5: Train SVM Model

1. Import the SVC model from the svm library
2. Create the SVC model with Kernel type set to linear
3. Fit the model using 'sugar_butter' and 'type_label' as the parameters

Step 6: Calculate Support Vector Boundaries.

1. Assign the first support vector

Step 11: Split Dataset

Step 12: Train new model

Step 13: Make prediction

Step 14: Evaluate Model

Program

```
import numpy as np
```

```
import pandas as pd
```

```
from sklearn import svm
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns; sns.set(font_scale=10)
```

```
recipes = pd.read_csv('.../CSV')
```

```
recipes.head()
```

```
recipes.shape
```

```
# Import 'Saggy'. You 'data=recipes'
```

hue = 'Type'

```
palette = 'Set1'. fill.legend = False, scatter
```

kws = {'s': 7000}

```
sugars_butter = recipes[['Sugar', 'Flour']]
```

```
type_label = np.where(recipes['Type'] == 'Muffin')
```

```
model = svm.SVC(kernel='linear')
```

```
model.fit(sugars_butter, type_label)
```

~~def c(kernel='linear')~~

~~w = model.coef_[0]~~

~~hyperplane_a = -w[0] / w[1]~~

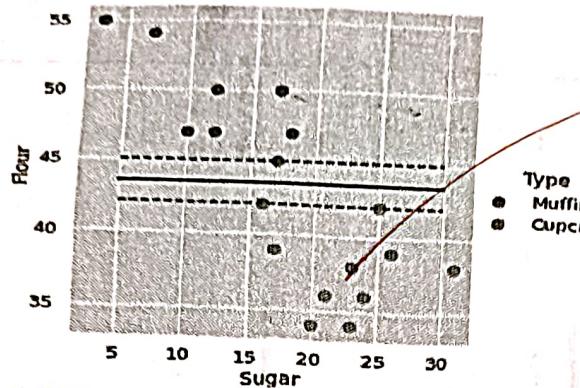
~~x2 = np.linspace(5, 30)~~

~~y2 = a + x2 * (-model.intercept_[1] / model.coef_[1])~~

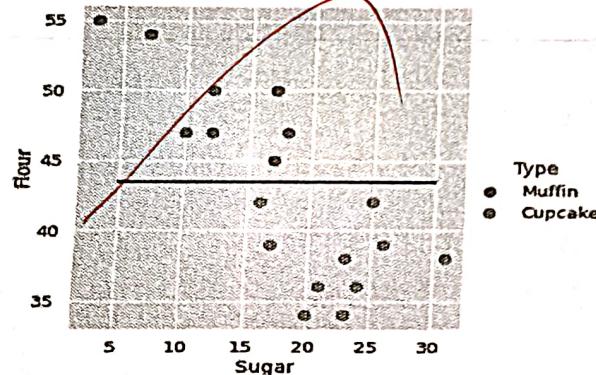
~~b = model.support_vectors_[0][1]~~

~~yy_down = a + x2 * (b[1] - a * b[0])~~

~~b = model.support_vectors_[1][1]~~



[<matplotlib.lines.Line2D at 0x7fca4a98ba50>]



$yy - \text{clown} = a * xx + b * yy - c$

Ins. Impot ('Sugar', 'Flour'; palette = Decipal
hue = 'Type', palette = 'Set1', fit_deg = False,

Scatter = kws = { "s": 100 })

plt.plot(xx, yy, linewidth = 2, color =
'black')

Scatter kws = { "s": 70 })

plt.plot(xx, yy, linewidth = 2, color = 'black')

Scatter w = { "s": 70 })

plt.plot(xx, yy, linewidth = 2, color = 'black')

plt.plot(xx, yy - clown, 'k--')

plt.plot(xx, yy - clown, 'k--')

plt.Scatter (model.support_vector_[:, 0], model

support_vector_[:, -1], s = 80, facecolor = 'none')

from sklearn.metrics import confusion_matrix

from sklearn.model_selection import train_test_
split

from sklearn.metrics import classification_
report

x_train, x_test, y_train, y_test

model.fit (x_train, y_train)

pred = model.predict (x_test)

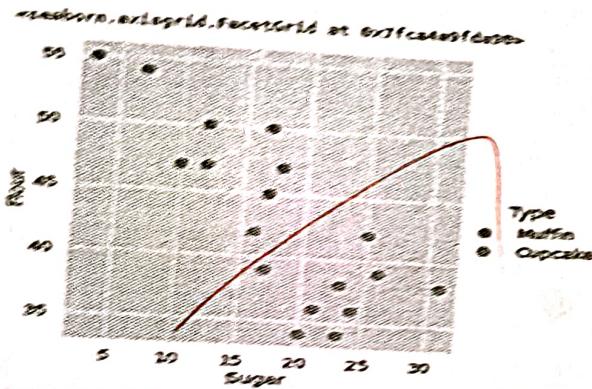
print (pred)

print (confusion_matrix (y_test, pred))

print (classification_report (y_test, pred)).

Result:-

Thus the python program to implement
SVM has been executed successfully.



(continued)

$\text{M}_1 = \text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10}$

$$(\text{M}_1)^2 = \text{m}_1^2 + \text{m}_2^2 + \text{m}_3^2 + \text{m}_4^2 + \text{m}_5^2 + \text{m}_6^2 + \text{m}_7^2 + \text{m}_8^2 + \text{m}_9^2 + \text{m}_{10}^2$$

$$\text{M}_1^2 = 22.5^2 + 22.5^2 + 22.5^2 + 22.5^2 + 22.5^2 + 22.5^2 + 22.5^2 + 22.5^2 + 22.5^2 + 22.5^2 = 450$$

$$\sqrt{\text{M}_1^2} = \sqrt{450} = 21.2$$

$\text{M}_2 = \text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10}$

$\text{M}_2^2 = (\text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10})^2 = 450$

$(\text{M}_2)^2 = (\text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10})^2 = 450$

$\sqrt{(\text{M}_2)^2} = \sqrt{450} = 21.2$

$\text{M}_3 = \text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10}$

$\text{M}_3^2 = (\text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10})^2 = 450$

$\sqrt{\text{M}_3^2} = \sqrt{450} = 21.2$

$\text{M}_4 = \text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10}$

$\text{M}_4^2 = (\text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10})^2 = 450$

$(\text{M}_4)^2 = (\text{m}_1 + \text{m}_2 + \text{m}_3 + \text{m}_4 + \text{m}_5 + \text{m}_6 + \text{m}_7 + \text{m}_8 + \text{m}_9 + \text{m}_{10})^2 = 450$

$\sqrt{(\text{M}_4)^2} = \sqrt{450} = 21.2$