

```
# Lab 8: Naive Bayes Classification Algorithm
# 0. Installation and Imports
```

```
# Uncomment if not done previously
!pip install ucimlrepo
```

```
from ucimlrepo import fetch_ucirepo
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
import matplotlib.pyplot as plt
```

```
Requirement already satisfied: ucimlrepo in /usr/local/lib/python3.12/dist-packages (0.0.7)
Requirement already satisfied: pandas>=1.0.0 in /usr/local/lib/python3.12/dist-packages (from ucimlrepo) (2.2.2)
Requirement already satisfied: certifi>=2020.12.5 in /usr/local/lib/python3.12/dist-packages (from ucimlrepo) (2025.11.12)
Requirement already satisfied: numpy>=1.26.0 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.0.2)
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.1)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.12/dist-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
```

```
# Lab 8: Naive Bayes Classification Algorithm
# 1. Load Dataset and Inspect
```

```
coverttype = fetch_ucirepo(id=31)
```

```
X = coverttype.data.features
y = coverttype.data.targets
```

```
print("Metadata:\n", coverttype.metadata)
```

```
Metadata:
{'uci_id': 31, 'name': 'Coverttype', 'repository_url': 'https://archive.ics.uci.edu/dataset/31/coverttype', 'data_url': 'https://archive.ics.uci.edu/dataset/31/coverttype'}
```

```
print("Sample features:\n", X.head())
```

```
Sample features:
   Elevation  Aspect  Slope  Horizontal_Distance_To_Hydrology  \
0      2596      51      3                      258
1      2590      56      2                      212
2      2804     139      9                      268
3      2785     155     18                      242
4      2595      45      2                      153

   Vertical_Distance_To_Hydrology  Horizontal_Distance_To_Roadways  \
0                      0                      510
1                      -6                      390
2                      65                     3180
3                     118                     3090
4                      -1                      391

   Hillshade_9am  Hillshade_Noon  Hillshade_3pm  \
0           221           232           148
1           220           235           151
2           234           238           135
3           238           238           122
4           220           234           150

   Horizontal_Distance_To_Fire_Points  ...  Soil_Type34  Soil_Type35  \
0                      6279  ...           0           0
1                      6225  ...           0           0
2                      6121  ...           0           0
3                      6211  ...           0           0
4                      6172  ...           0           0

   Soil_Type36  Soil_Type37  Soil_Type38  Soil_Type39  Soil_Type40  \
0           0           0           0           0           0
1           0           0           0           0           0
2           0           0           0           0           0
3           0           0           0           0           0
4           0           0           0           0           0

   Wilderness_Area2  Wilderness_Area3  Wilderness_Area4
0           0           0           0
1           0           0           0
2           0           0           0
3           0           0           0
4           0           0           0
```

```
[5 rows x 54 columns]
```

```
print("Sample targets:\n", y.head())
```

```
Sample targets:
```

```
  Cover_Type
0          5
1          5
2          2
3          2
4          5
```

```
print("\nVariables:\n", covertype.variables)
```

```
54      Wilderness_Area4  Feature  Integer      None
description units missing_values
0      None  None      no
1      None  None      no
2      None  None      no
3      None  None      no
4      None  None      no
5      None  None      no
6      None  None      no
7      None  None      no
8      None  None      no
9      None  None      no
10     None  None      no
11     None  None      no
12     None  None      no
13     None  None      no
14     None  None      no
15     None  None      no
16     None  None      no
17     None  None      no
18     None  None      no
19     None  None      no
20     None  None      no
21     None  None      no
22     None  None      no
23     None  None      no
24     None  None      no
25     None  None      no
26     None  None      no
27     None  None      no
28     None  None      no
29     None  None      no
30     None  None      no
31     None  None      no
32     None  None      no
33     None  None      no
34     None  None      no
35     None  None      no
36     None  None      no
37     None  None      no
38     None  None      no
39     None  None      no
40     None  None      no
41     None  None      no
42     None  None      no
43     None  None      no
44     None  None      no
45     None  None      no
46     None  None      no
47     None  None      no
48     None  None      no
49     None  None      no
50     None  None      no
51     None  None      no
52     None  None      no
53     None  None      no
54     None  None      no
```

```
# Lab 8: Naive Bayes Classification Algorithm
# 2. Train-Test Split
```

```
X_train, X_test, y_train, y_test = train_test_split(
    X, y.values.ravel(), test_size=0.2, random_state=42
)
```

```
print("Training set size:", X_train.shape)
print("Testing set size:", X_test.shape)
```

```
Training set size: (464809, 54)
Testing set size: (116203, 54)
```

```
# Lab 8: Naive Bayes Classification Algorithm
# 3. Training and Prediction
```

```
nb = GaussianNB()
nb.fit(X_train, y_train)
y_pred = nb.predict(X_test)
```

```
# Lab 8: Naive Bayes Classification Algorithm
# 4. Evaluation
```

```
print("Accuracy Score:", accuracy_score(y_test, y_pred))
print("Classification Report:\n", classification_report(y_test, y_pred))
print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

```
Accuracy Score: 0.4568814918719826
Classification Report:
              precision    recall  f1-score   support

     1         0.50         0.78         0.61       42557
     2         0.83         0.16         0.27       56500
     3         0.45         0.81         0.57        7121
     4         0.21         0.81         0.34         526
     5         0.08         0.61         0.14        1995
     6         0.36         0.08         0.14        3489
     7         0.36         0.81         0.50        4015

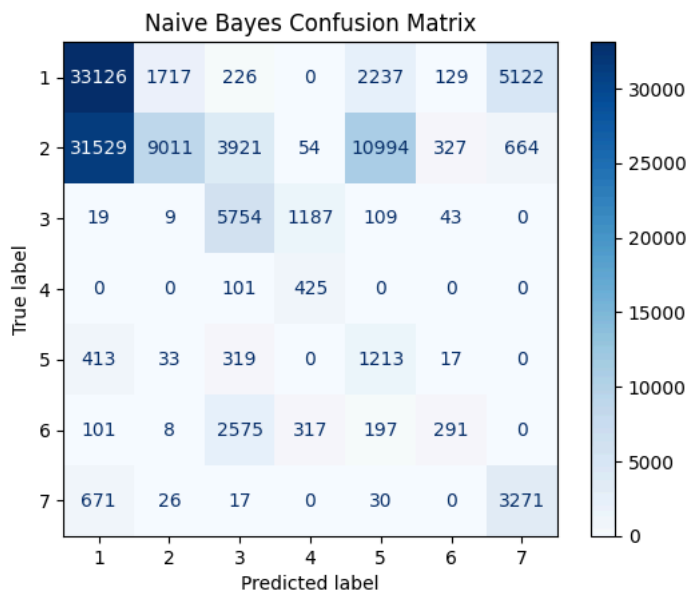
 accuracy                   0.46       116203
 macro avg              0.40         0.58         0.37       116203
 weighted avg           0.64         0.46         0.41       116203
```

```
Confusion Matrix:
[[33126  1717   226    0  2237   129  5122]
 [31529  9011  3921   54 10994   327   664]
 [   19     9  5754  1187   109    43     0]
 [    0     0   101   425     0     0     0]
 [  413    33   319     0  1213    17     0]
 [  101     8  2575   317   197   291     0]
 [  671    26    17     0    30     0  3271]]
```

```
# Lab 8: Naive Bayes Classification Algorithm
# 5. Confusion Matrix Visualization
```

```
from sklearn.metrics import ConfusionMatrixDisplay
```

```
cm = confusion_matrix(y_test, y_pred)
ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=nb.classes_).plot(cmap="Blues")
plt.title("Naive Bayes Confusion Matrix")
plt.show()
```



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