

# practical-exam-08

May 23, 2023

```
[1]: from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force\_remount=True).

## 1 Problem Statement 8

Implement Simple Naïve Bayes classification algorithm using Python/R on iris.csv dataset. Compute Confusion matrix to find TP, FP, TN, FN, Accuracy, Error rate, Precision, Recall on the given dataset.

```
[3]: import pandas as pd
```

```
[106]: data = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/exam_datasets/
↳5-8-13-14.iris.csv')
```

<IPython.core.display.HTML object>

```
[109]: data.info()
```

<IPython.core.display.HTML object>

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 150 entries, 0 to 149

Data columns (total 5 columns):

| # | Column       | Non-Null Count | Dtype   |
|---|--------------|----------------|---------|
| 0 | sepal.length | 150 non-null   | float64 |
| 1 | sepal.width  | 150 non-null   | float64 |
| 2 | petal.length | 150 non-null   | float64 |
| 3 | petal.width  | 150 non-null   | float64 |
| 4 | variety      | 150 non-null   | object  |

dtypes: float64(4), object(1)

memory usage: 6.0+ KB

```
[120]: x = data.iloc[:, :4].values
y = data['variety'].values
```

x

<IPython.core.display.HTML object>

```
[120]: array([[5.1, 3.5, 1.4, 0.2],
              [4.9, 3. , 1.4, 0.2],
              [4.7, 3.2, 1.3, 0.2],
              [4.6, 3.1, 1.5, 0.2],
              [5. , 3.6, 1.4, 0.2],
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              [4.5, 2.3, 1.3, 0.3],
```

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```
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[6.5, 3. , 5.2, 2. ],
[6.2, 3.4, 5.4, 2.3],
[5.9, 3. , 5.1, 1.8]])
```

```
[113]: from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2,
↳random_state=42)
```

<IPython.core.display.HTML object>

```
[114]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
```

<IPython.core.display.HTML object>

```
[116]: from sklearn.naive_bayes import GaussianNB
classifier = GaussianNB()
classifier.fit(x_train, y_train)
```

<IPython.core.display.HTML object>

```
[116]: GaussianNB()
```

```
[117]: y_pred = classifier.predict(x_test)
y_pred
```

<IPython.core.display.HTML object>

```
[117]: array(['Versicolor', 'Setosa', 'Virginica', 'Versicolor', 'Versicolor',
'Setosa', 'Versicolor', 'Virginica', 'Versicolor', 'Versicolor',
'Virginica', 'Setosa', 'Setosa', 'Setosa', 'Setosa', 'Versicolor',
'Virginica', 'Versicolor', 'Versicolor', 'Virginica', 'Setosa',
'Virginica', 'Setosa', 'Virginica', 'Virginica', 'Virginica',
'Virginica', 'Virginica', 'Setosa', 'Setosa'], dtype='<U10')
```

```
[132]: from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
from sklearn.metrics import accuracy_score, precision_score, recall_score

accuracy = accuracy_score(y_test, y_pred)
error_rate = 1 - accuracy
precision = precision_score(y_test, y_pred, average='weighted')
recall = recall_score(y_test, y_pred, average='weighted')

print(f"Accuracy: {accuracy}")
print(f"Error Rate: {error_rate}")
print(f"Precision: {precision}")
print(f"Recall: {recall}")
```

<IPython.core.display.HTML object>

```
Accuracy: 1.0
Error Rate: 0.0
Precision: 1.0
Recall: 1.0
Accuracy : 1.0
```

```
[122]: df = pd.DataFrame({'Real Values':y_test, 'Predicted Values':y_pred})
df
```

<IPython.core.display.HTML object>

```
[122]:
```

|    | Real Values | Predicted Values |
|----|-------------|------------------|
| 0  | Versicolor  | Versicolor       |
| 1  | Setosa      | Setosa           |
| 2  | Virginica   | Virginica        |
| 3  | Versicolor  | Versicolor       |
| 4  | Versicolor  | Versicolor       |
| 5  | Setosa      | Setosa           |
| 6  | Versicolor  | Versicolor       |
| 7  | Virginica   | Virginica        |
| 8  | Versicolor  | Versicolor       |
| 9  | Versicolor  | Versicolor       |
| 10 | Virginica   | Virginica        |
| 11 | Setosa      | Setosa           |
| 12 | Setosa      | Setosa           |
| 13 | Setosa      | Setosa           |
| 14 | Setosa      | Setosa           |
| 15 | Versicolor  | Versicolor       |
| 16 | Virginica   | Virginica        |
| 17 | Versicolor  | Versicolor       |
| 18 | Versicolor  | Versicolor       |
| 19 | Virginica   | Virginica        |

|    |           |           |
|----|-----------|-----------|
| 20 | Setosa    | Setosa    |
| 21 | Virginica | Virginica |
| 22 | Setosa    | Setosa    |
| 23 | Virginica | Virginica |
| 24 | Virginica | Virginica |
| 25 | Virginica | Virginica |
| 26 | Virginica | Virginica |
| 27 | Virginica | Virginica |
| 28 | Setosa    | Setosa    |
| 29 | Setosa    | Setosa    |