



Experiment No- 2.2

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Branch: CSE

Semester: 5

Subject Name: Machine Learning

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Subject Code: 20CSP -317

- 1. Aim/Overview of the practical:** Implementing Naive Bayes' Classifier on any dataset and analyse the accuracy.
- 2. Task to be done/ Which logistics used:** Analysing accuracy by implementing the Naive Bayes' Classifier on any dataset.
- 3. Steps of experiment/Code:**

- 1. Importing libraries such as matplotlib, pandas and seaborn and reading the dataset.**

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
```

```
In [2]: dataset = pd.read_csv('NaiveBayes.csv')

# split the data into inputs and outputs
X = dataset.iloc[:, [0,1]].values
y = dataset.iloc[:, 2].values
```

2. Preprocessing the data for training the model and defining the model:

```
from sklearn.model_selection import train_test_split

# assign test data size 25%
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size= 0.25, random_state=0)
```

```
from sklearn.preprocessing import StandardScaler

# scaling the input data
sc_X = StandardScaler()
X_train = sc_X.fit_transform(X_train)
X_test = sc_X.fit_transform(X_test)
```

3. Implementing Naïve Bayes' Classifier using Bernoulli and calculating the accuracy.

```
from sklearn.naive_bayes import BernoulliNB

# initializing the NB
classifier = BernoulliNB()

# training the model
classifier.fit(X_train, y_train)

# testing the model
y_pred = classifier.predict(X_test)
```

```
from sklearn.metrics import accuracy_score

# printing the accuracy of the model
print(accuracy_score(y_pred, y_test))
```

0.8

4. Implementing Naïve Bayes' Classifier using Gaussian and calculating the accuracy

```
# import Gaussian Naive Bayes classifier
from sklearn.naive_bayes import GaussianNB

# create a Gaussian Classifier
classifier1 = GaussianNB()

# training the model
classifier1.fit(X_train, y_train)

# testing the model
y_pred1 = classifier1.predict(X_test)
```

```
from sklearn.metrics import accuracy_score

# printing the accuracy of the model
print(accuracy_score(y_test, y_pred1))
```

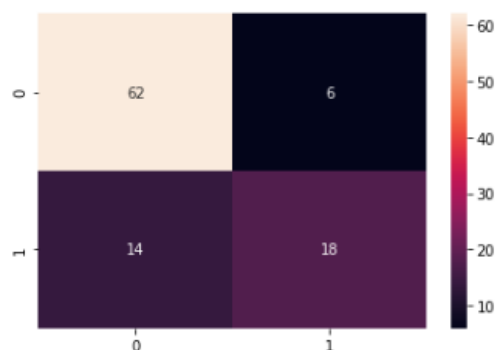
0.91

5. Displaying Confusion Matrices and Classification Reports:

```
# importing the required modules
import seaborn as sns
from sklearn.metrics import confusion_matrix

# passing actual and predicted values
cm = confusion_matrix(y_test, y_pred)

# true write data values in each cell of the matrix
sns.heatmap(cm, annot=True)
plt.savefig('confusion.png')
```



```
from sklearn.metrics import classification_report

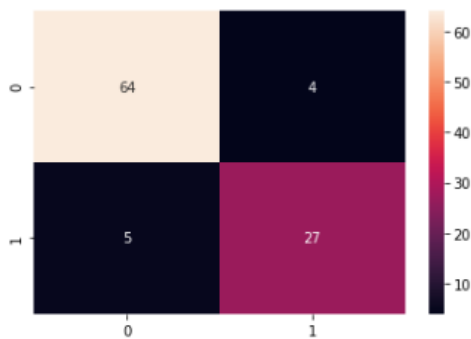
# printing the report
print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
0	0.82	0.91	0.86	68
1	0.75	0.56	0.64	32
accuracy			0.80	100
macro avg	0.78	0.74	0.75	100
weighted avg	0.79	0.80	0.79	100

```
import seaborn as sns
from sklearn.metrics import confusion_matrix

# passing actual and predicted values
cm = confusion_matrix(y_test, y_pred1)

# true write data values in each cell of the matrix
sns.heatmap(cm,annot=True)
plt.savefig('confusion.png')
```



```
from sklearn.metrics import classification_report

# printing the report
print(classification_report(y_test, y_pred1))
```

	precision	recall	f1-score	support
0	0.93	0.94	0.93	68
1	0.87	0.84	0.86	32
accuracy			0.91	100
macro avg	0.90	0.89	0.90	100
weighted avg	0.91	0.91	0.91	100



6. Implementing Naive Bayes' Classifier for label encoding:

```
from sklearn.naive_bayes import GaussianNB

# create a Gaussian Classifier
model = GaussianNB()

# train the model using the training sets
model.fit(weather_encoded, label)
```

7. Getting Error as we need to reshape the given data:

```
from sklearn.naive_bayes import GaussianNB

# create a Gaussian Classifier
model = GaussianNB()

# train the model using the training sets
model.fit(weather_encoded, label)

-----
ValueError                                Traceback (most recent call last)
<ipython-input-21-d1010b7d5489> in <module>
      5
      6 # train the model using the training sets
----> 7 model.fit(weather_encoded, label)

D:\Anaconda3\lib\site-packages\sklearn\naive_bayes.py in fit(self, X, y, sample_weight)
    208     self : object
    209     """
--> 210     X, y = self._validate_data(X, y)
    211     y = column_or_1d(y, warn=True)
    212     return self._partial_fit(X, y, np.unique(y), _refit=True,

D:\Anaconda3\lib\site-packages\sklearn\base.py in _validate_data(self, X, y, reset, validate)
    430         y = check_array(y, **check_y_params)
    431     else:
--> 432         X, y = check_X_y(X, y, **check_params)
    433         out = X, y
    434

D:\Anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)
    70         FutureWarning)
    71         kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
--> 72         return f(**kwargs)
    73     return inner_f
    74

D:\Anaconda3\lib\site-packages\sklearn\utils\validation.py in check_X_y(X, y, accept_sparse, force_all_finite, ensure_2d, allow_nd, multi_output, ensure_min_samples, ensure_min_rows)
    793         raise ValueError("y cannot be None")
    794
```

8. Implementing Naive Bayes' Classifier and predicting the output:

```
import numpy as np

# converting 1D array to 2D
weather_2d = np.reshape(weather_encoded, (-1, 1))
```

```
from sklearn.naive_bayes import GaussianNB

# create a Gaussian Classifier
model = GaussianNB()

# train the model using the training sets
model.fit(weather_2d, label)
```

```
GaussianNB()
```

```
predicted = model.predict([[0]]) # 0:Overcast

# printing predicted value
print(predicted)
```

```
[1]
```



Learning Outcomes (What I have learnt):

1. I have learnt about implementing Naïve Bayes' Classifier for Bernoulli and Gaussian.
2. I have learnt about constructing confusion matrix.
3. I have learnt about various libraries which are supported by python such as sklearn, numpy, matplotlib.
4. I have learnt about the various functions provided by various libraries.
5. I have understood the experiment very well.



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Evaluation Grid:

	Parameters	Marks Obtained	Maximum Marks
1.	Student Performance (Conduct of experiment) objectives/Outcomes.		12
2.	Viva Voce		10
3.	Submission of Work Sheet (Record)		8
	Total		30