LAB # 04

SUPERVISED LEARNING (NAÏVE BAYES ALGORITHM)

OBJECTIVE

Implementing supervised learning, Naïve Bayes algorithm for training, testing and classification

Lab Tasks:

1. Implement Naïve Bayes Algorithm on the given dataset in Fig 1 to predict whether the players can play or not when the weather is overcast and the temperature is mild.

```
# Step 2: Encode each column separately
weather_encoder = LabelEncoder()
temperature_encoder = LabelEncoder()
play_encoder = LabelEncoder()

df['Weather'] = weather_encoder.fit_transform(df['Weather'])
df['Temperature'] = temperature_encoder.fit_transform(df['Temperature'])
df['Play'] = play_encoder.fit_transform(df['Play'])

# Separate features and target
X = df[['Weather', 'Temperature']]
y = df['Play']
```

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```
# Step 3: Train the Naïve Bayes model
model = CategoricalNB()
model.fit(X, y)
  CategoricalNB
CategoricalNB()
# Step 4: Make a prediction for "overcast" and "mild"
# Use the encoders to transform 'Overcast' and 'Mild'
overcast = weather_encoder.transform(['Overcast'])[0]
mild = temperature_encoder.transform(['Mild'])[0]
# Create a DataFrame for the prediction input with the correct feature names
predict_df = pd.DataFrame([[overcast, mild]], columns=['Weather', 'Temperature'])
# Predict if players can play
prediction = model.predict(predict_df)
# Convert prediction back to original label
play prediction = play encoder.inverse transform(prediction)
print("Prediction (Play):", play_prediction[0])
Prediction (Play): Yes
```

2. Consider the given dataset. Implement Naïve Bayes Algorithm to classify youth/medium/yes/fair.

```
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 #Muhammad Saud Hassan
 import pandas as pd
 from sklearn.preprocessing import LabelEncoder
 from sklearn.naive_bayes import CategoricalNB
 # Step 1: Create the dataset
 data = {
    'age': ['youth', 'youth', 'middle_aged', 'senior', 'senior', 'senior', 'middle_aged', 'youth',
             'youth', 'senior', 'youth', 'middle_aged', 'middle_aged', 'senior'],
    'income': ['high', 'high', 'high', 'medium', 'low', 'low', 'low', 'medium',
               'low', 'medium', 'medium', 'medium', 'high', 'medium'],
     'student': ['no', 'no', 'no', 'yes', 'yes', 'yes', 'no',
                'yes', 'yes', 'no', 'yes', 'no'],
     'credit_rating': ['fair', 'excellent', 'fair', 'fair', 'fair', 'excellent', 'excellent', 'fair',
                       'fair', 'fair', 'excellent', 'excellent', 'fair', 'excellent'],
     'buys_computer': ['no', 'no', 'yes', 'yes', 'yes', 'no', 'yes', 'no',
                      'yes', 'yes', 'yes', 'yes', 'no']
 # Convert to DataFrame
 df = pd.DataFrame(data)
```

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```
# Step 2: Encode each column separately
age_encoder = LabelEncoder()
income_encoder = LabelEncoder()
student_encoder = LabelEncoder()
credit_rating_encoder = LabelEncoder()
buys_computer_encoder = LabelEncoder()

df['age'] = age_encoder.fit_transform(df['age'])
df['income'] = income_encoder.fit_transform(df['income'])
df['student'] = student_encoder.fit_transform(df['student'])
df['credit_rating'] = credit_rating_encoder.fit_transform(df['credit_rating'])
df['buys_computer'] = buys_computer_encoder.fit_transform(df['buys_computer'])

# Separate features and target
X = df[['age', 'income', 'student', 'credit_rating']]
y = df['buys_computer']
```

```
# Step 3: Train the Naïve Bayes model
model = CategoricalNB()
model.fit(X, y)
```

```
# Step 4: Make a prediction for "youth", "medium", "yes", "fair"
# Use the encoders to transform each input feature
youth = age_encoder.transform(['youth'])[0]
medium = income_encoder.transform(['medium'])[0]
yes = student_encoder.transform(['yes'])[0]
fair = credit_rating_encoder.transform(['fair'])[0]

# Create a DataFrame for the prediction input with the correct feature names
predict_df = pd.DataFrame([[youth, medium, yes, fair]], columns=['age', 'income', 'student', 'credit_rating'])

# Predict if the customer will buy a computer
prediction = model.predict(predict_df)

# Convert prediction back to original label
buys_computer_prediction = buys_computer_encoder.inverse_transform(prediction)
print("Prediction (buys_computer):", buys_computer_prediction[0])

Prediction (buys_computer): yes
```

Home Tasks:

Using Naïve Bayes Algorithm, predict whether a student will pass or fail based on their study hours, attendance, percentage and participation in extra circular activities.

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```
#22F-BSE-138
 #Muhammad Saud Hassan
 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
 # Step 1: Create sample data
 data = {
     'Study_Hours': [5, 6, 7, 2, 3, 8, 4, 1],
      'Attendance': [90, 85, 88, 60, 70, 95, 80, 50],
     'Percentage': [75, 78, 82, 65, 68, 85, 72, 55],
     'Extracurricular': [1, 1, 0, 0, 1, 1, 0, 0], # 1 = participates, 0 = does not participate
     'Pass': ['Yes', 'Yes', 'Yes', 'No', 'No', 'Yes', 'No', 'No']
 df = pd.DataFrame(data)
 # Step 2: Convert categorical target to numerical values
 df['Pass'] = df['Pass'].map({'No': 0, 'Yes': 1})
 # Step 3: Define features and target variable
 X = df[['Study_Hours', 'Attendance', 'Percentage', 'Extracurricular']]
 y = df['Pass']
# Step 4: Split the data into training and testing sets
 \textbf{X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.3, random\_state=42) } 
# Step 5: Train the Naïve Bayes model
model = GaussianNB()
model.fit(X_train, y_train)
# Step 6: Make predictions on the test set
y_pred = model.predict(X_test)
# Step 7: Evaluate the model
print("Accuracy:", accuracy_score(y_test, y_pred))
print("Classification Report:\n", classification_report(y_test, y_pred, zero_division=1))
print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
# Step 8: Predict for a new student
new_student = pd.DataFrame([[6, 85, 80, 1]], columns=['Study_Hours', 'Attendance', 'Percentage', 'Extracurricular'])
prediction = model.predict(new_student)
result = "Pass" if prediction[0] == 1 else "Fail"
print("Prediction for new student:", result)
 Accuracy: 0.0
  Classification Report:
                   precision recall f1-score support
               a
                        0.00
                                  1.00
                                                0.00
                                                             0.0
               1
                        1.00
                                   0.00
                                                0.00
                                                            3.0
                                                0.00
                                                            3.0
      accuracy
                       0.50
                                  0.50
                                              0.00
                                                            3.0
     macro avg
                                                0.00
  weighted avg
                        1.00
                                   0.00
                                                             3.0
  Confusion Matrix:
  [[0 0]
  [3 0]]
  Prediction for new student: Fail
```