

5. Write python programme to implement logistic Regression (StudieshoursVsFail/pass) .

```
# Step 1: Import necessary libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import (
    accuracy_score,
    confusion_matrix,
    classification_report,
    roc_curve,
    auc
)
```

```
# Step 2: Set random seed for reproducibility
np.random.seed(42)

# Step 3: Generate synthetic data
n_samples = 1000
hours_studied = np.random.uniform(0, 10, n_samples)

# Define the logistic function
def sigmoid(x):
    return 1 / (1 + np.exp(-x))

# Calculate the probability of passing
# Adjust the coefficients to control the steepness and midpoint of the curve
prob_pass = sigmoid(1.5 * (hours_studied - 5))

# Generate pass/fail outcomes based on the probabilities
pass_fail = np.random.binomial(1, prob_pass)
```

```
# Create a DataFrame
df = pd.DataFrame({
    'Hours_Studied': hours_studied,
    'Pass': pass_fail
})
print(df)
```

	Hours_Studied	Pass
0	3.745401	0
1	9.507143	1
2	7.319939	1
3	5.986585	1
4	1.560186	0
..

```
995      0.915821      0
996      9.173136      1
997      1.368186      0
998      9.502374      1
999      4.460058      0
```

```
[1000 rows x 2 columns]
```

```
# Step 4: Visualize the data
plt.figure(figsize=(8, 6))
sns.scatterplot(x='Hours_Studied', y='Pass', data=df, alpha=0.5)
plt.title('Study Hours vs. Pass/Fail')
plt.xlabel('Hours Studied')
plt.ylabel('Pass (1) / Fail (0)')
plt.grid(True)
plt.savefig('Study Hours vs. Pass/Fail')
plt.show()
```

```
# Step 5: Prepare features and target
X = df[['Hours_Studied']]
y = df['Pass']
```

```
# Step 6: Split into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42, stratify=y
)
```

```
# Step 7: Train the logistic regression model
model = LogisticRegression()
model.fit(X_train, y_train)
```

```
# Step 8: Make predictions
y_pred = model.predict(X_test)
y_pred_proba = model.predict_proba(X_test)[:, 1]
```

```
# Step 9: Evaluate the model
print("Accuracy:", accuracy_score(y_test, y_pred))
print("\nConfusion Matrix:\n", confusion_matrix(y_test, y_pred))
print("\nClassification Report:\n", classification_report(y_test, y_pred))
```

```
Accuracy: 0.9
```

```
Confusion Matrix:
```

```
[[94  9]
 [11 86]]
```

```
Classification Report:
```

	precision	recall	f1-score	support
0	0.90	0.91	0.90	103
1	0.91	0.89	0.90	97

accuracy			0.90	200
macro avg	0.90	0.90	0.90	200
weighted avg	0.90	0.90	0.90	200

```
# Step 10: Plot ROC curve
fpr, tpr, thresholds = roc_curve(y_test, y_pred_proba)
roc_auc = auc(fpr, tpr)
```

