

lab3

January 20, 2025

```
[2]: import seaborn as sns  
import matplotlib.pyplot as plt
```

```
[1]: import pandas as pd  
  
df = pd.read_csv("./data.csv")
```

```
[33]: from sklearn.preprocessing import StandardScaler  
from sklearn.model_selection import train_test_split  
scaler = StandardScaler()
```

```
[35]: train, test = train_test_split(df, test_size=0.2)
```

```
[37]: X_train = train[['Socioeconomic Score',  
    'Study Hours',  
    'Sleep Hours',  
    'Attendance (%)']]  
  
X_test = test[['Socioeconomic Score',  
    'Study Hours',  
    'Sleep Hours',  
    'Attendance (%)']]
```

```
[38]: y_train = train[['Grades']]  
y_test = test[['Grades']]
```

```
[39]: from sklearn.linear_model import LinearRegression
```

```
[40]: regressor = LinearRegression(n_jobs=5)
```

```
[41]: regressor.fit(X_train, y_train)
```

```
[41]: LinearRegression(n_jobs=5)
```

```
[42]: y_pred = regressor.predict(X_test)
```

```
[43]: from sklearn.metrics import mean_absolute_error
      from sklearn.metrics import mean_squared_error
      from sklearn.metrics import r2_score
```

```
[44]: mae = mean_absolute_error(y_pred=y_pred, y_true=y_test)
      mse = mean_squared_error(y_pred=y_pred, y_true=y_test)
      r2 = r2_score(y_pred=y_pred, y_true=y_test)
```

```
[45]: print(f"R2 Score: {r2}")
      print(f"MSE: {mse}")
      print(f"MAE: {mae}")
```

```
R2 Score: 0.7928316268914254
MSE: 17.24643198486114
MAE: 3.2081125696592454
```

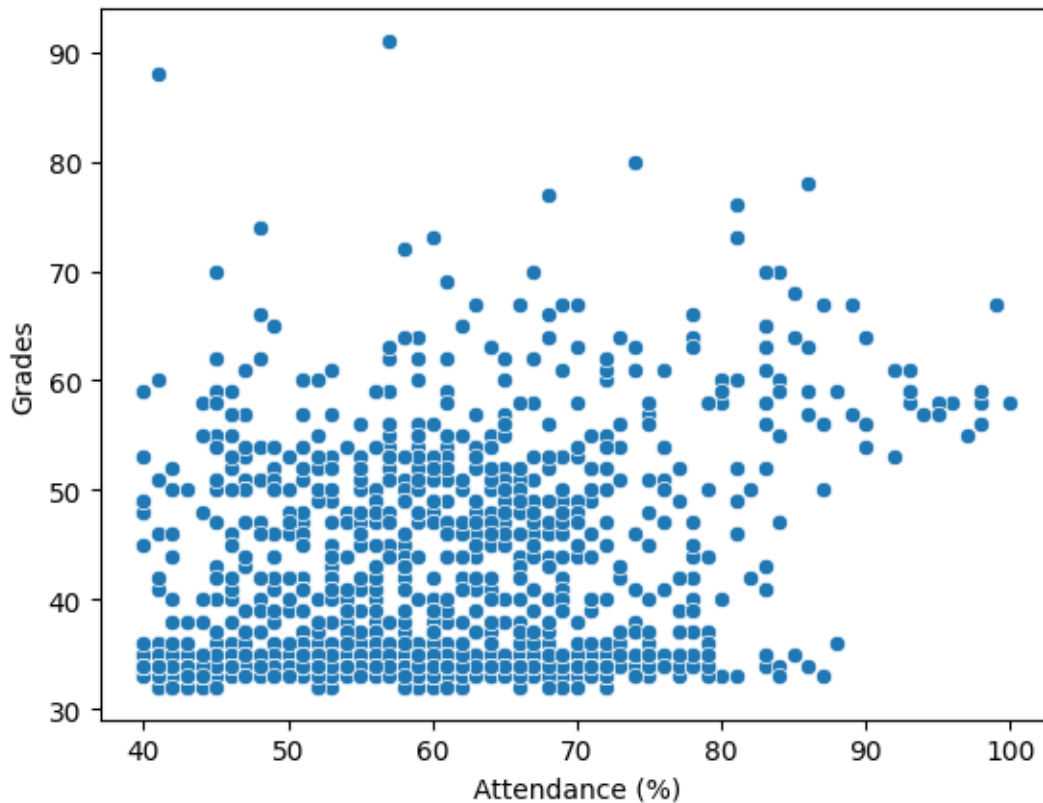
```
[46]: y_test[:5], y_pred[:5]
```

```
[46]: (      Grades
      602      43.0
      1196     34.0
      1226     35.0
      474      36.0
      1252     34.0,
      array([[40.57640149],
             [24.00012658],
             [34.21593047],
             [29.70645367],
             [31.41475296]]))
```

```
[47]: print (df.columns)
```

```
Index(['Socioeconomic Score', 'Study Hours', 'Sleep Hours', 'Attendance (%)',
      'Grades'],
      dtype='object')
```

```
[48]: sns.scatterplot(x=df['Attendance (%)'], y=df['Grades'])
      plt.show()
```



Logistic Regression

```
[3]: df = pd.read_csv("./diabetes.csv")
```

```
[51]: train1, test1 = train_test_split(df, test_size=0.2)
```

```
[53]: X_train1 = train1[['Pregnancies',  
    'Glucose',  
    'BloodPressure',  
    'SkinThickness',  
    'Insulin',  
    'BMI',  
    'DiabetesPedigreeFunction',  
    'Age']]
```

```
X_test1 = test1[['Pregnancies',  
    'Glucose',  
    'BloodPressure',  
    'SkinThickness',  
    'Insulin',  
    'BMI',
```

```
'DiabetesPedigreeFunction',  
'Age']]
```

```
[54]: print(train1.columns)
```

```
Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',  
      'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],  
      dtype='object')
```

```
[55]: y_train1 = train1[['Outcome']]  
      y_test1 = test1[['Outcome']]
```

```
[56]: y_train1 = y_train1.values.ravel()  
      y_test1 = y_test1.values.ravel()
```

```
[57]: from sklearn.linear_model import LogisticRegression  
      from sklearn.metrics import accuracy_score,  
      ↪confusion_matrix, classification_report
```

```
[58]: model = LogisticRegression(n_jobs=5, max_iter=500)  
      model.fit(X_train1, y_train1)
```

```
[58]: LogisticRegression(max_iter=500, n_jobs=5)
```

```
[59]: y_pred1 = model.predict(X_test1)  
      accuracy = accuracy_score(y_test1, y_pred1)  
      print(f"Accuracy: {accuracy:.2f}")
```

Accuracy: 0.75

```
[60]: cm = confusion_matrix(y_test1, y_pred1)  
      print("Confusion Matrix:")  
      print(cm)
```

Confusion Matrix:

```
[[87 12]  
 [26 29]]
```

```
[61]: report = classification_report(y_test1, y_pred1)  
      print("Classification Report:")  
      print(report)
```

Classification Report:

	precision	recall	f1-score	support
0	0.77	0.88	0.82	99
1	0.71	0.53	0.60	55

accuracy			0.75	154
macro avg	0.74	0.70	0.71	154
weighted avg	0.75	0.75	0.74	154

```
[62]: y_test1[:10], y_pred1[:10]
```

```
[62]: (array([0, 0, 1, 0, 1, 0, 1, 0, 1, 0]), array([0, 0, 1, 0, 0, 0, 0, 0, 0, 0]))
```

```
[63]: sns.scatterplot(x=df['Glucose'], y=df['BloodPressure'])
plt.show()
```

