

report-1

March 18, 2025

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

```
[18]: np.random.seed(42)
```

1. DATA LOADING & INITIAL OVERVIEW

```
[10]: df = pd.read_csv('/content/student_data.csv')
```

```
[11]: print("DATA INFO:")
print(df.info())
```

DATA INFO:

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

RangeIndex: 6607 entries, 0 to 6606

Data columns (total 20 columns):

#	Column	Non-Null Count	Dtype
0	Hours_Studied	6607 non-null	int64
1	Attendance	6607 non-null	int64
2	Parental_Involvement	6607 non-null	object
3	Access_to_Resources	6607 non-null	object
4	Extracurricular_Activities	6607 non-null	object
5	Sleep_Hours	6607 non-null	int64
6	Previous_Scores	6607 non-null	int64
7	Motivation_Level	6607 non-null	object
8	Internet_Access	6607 non-null	object
9	Tutoring_Sessions	6607 non-null	int64
10	Family_Income	6607 non-null	object
11	Teacher_Quality	6529 non-null	object
12	School_Type	6607 non-null	object
13	Peer_Influence	6607 non-null	object
14	Physical_Activity	6607 non-null	int64
15	Learning_Disabilities	6607 non-null	object
16	Parental_Education_Level	6517 non-null	object
17	Distance_from_Home	6540 non-null	object

```

18 Gender                6607 non-null    object
19 Exam_Score            6607 non-null    int64
dtypes: int64(7), object(13)
memory usage: 1.0+ MB
None

```

```

[12]: print("\nSTATISTICAL DESCRIPTION:")
      print(df.describe(include='all'))

```

STATISTICAL DESCRIPTION:

	Hours_Studied	Attendance	Parental_Involvement	Access_to_Resources	\
count	6607.000000	6607.000000	6607	6607	
unique	NaN	NaN	3	3	
top	NaN	NaN	Medium	Medium	
freq	NaN	NaN	3362	3319	
mean	19.975329	79.977448	NaN	NaN	
std	5.990594	11.547475	NaN	NaN	
min	1.000000	60.000000	NaN	NaN	
25%	16.000000	70.000000	NaN	NaN	
50%	20.000000	80.000000	NaN	NaN	
75%	24.000000	90.000000	NaN	NaN	
max	44.000000	100.000000	NaN	NaN	

	Extracurricular_Activities	Sleep_Hours	Previous_Scores	\
count	6607	6607.000000	6607.000000	
unique	2	NaN	NaN	
top	Yes	NaN	NaN	
freq	3938	NaN	NaN	
mean	NaN	7.02906	75.070531	
std	NaN	1.46812	14.399784	
min	NaN	4.00000	50.000000	
25%	NaN	6.00000	63.000000	
50%	NaN	7.00000	75.000000	
75%	NaN	8.00000	88.000000	
max	NaN	10.00000	100.000000	

	Motivation_Level	Internet_Access	Tutoring_Sessions	Family_Income	\
count	6607	6607	6607.000000	6607	
unique	3	2	NaN	3	
top	Medium	Yes	NaN	Low	
freq	3351	6108	NaN	2672	
mean	NaN	NaN	1.493719	NaN	
std	NaN	NaN	1.230570	NaN	
min	NaN	NaN	0.000000	NaN	
25%	NaN	NaN	1.000000	NaN	
50%	NaN	NaN	1.000000	NaN	
75%	NaN	NaN	2.000000	NaN	

max	NaN	NaN	8.000000	NaN
-----	-----	-----	----------	-----

	Teacher_Quality	School_Type	Peer_Influence	Physical_Activity \
count	6529	6607	6607	6607.000000
unique	3	2	3	NaN
top	Medium	Public	Positive	NaN
freq	3925	4598	2638	NaN
mean	NaN	NaN	NaN	2.967610
std	NaN	NaN	NaN	1.031231
min	NaN	NaN	NaN	0.000000
25%	NaN	NaN	NaN	2.000000
50%	NaN	NaN	NaN	3.000000
75%	NaN	NaN	NaN	4.000000
max	NaN	NaN	NaN	6.000000

	Learning_Disabilities	Parental_Education_Level	Distance_from_Home \
count	6607	6517	6540
unique	2	3	3
top	No	High School	Near
freq	5912	3223	3884
mean	NaN	NaN	NaN
std	NaN	NaN	NaN
min	NaN	NaN	NaN
25%	NaN	NaN	NaN
50%	NaN	NaN	NaN
75%	NaN	NaN	NaN
max	NaN	NaN	NaN

	Gender	Exam_Score
count	6607	6607.000000
unique	2	NaN
top	Male	NaN
freq	3814	NaN
mean	NaN	67.235659
std	NaN	3.890456
min	NaN	55.000000
25%	NaN	65.000000
50%	NaN	67.000000
75%	NaN	69.000000
max	NaN	101.000000

```
[17]: print("\nMISSING VALUES PER COLUMN:")
      print(df.isnull().sum())
```

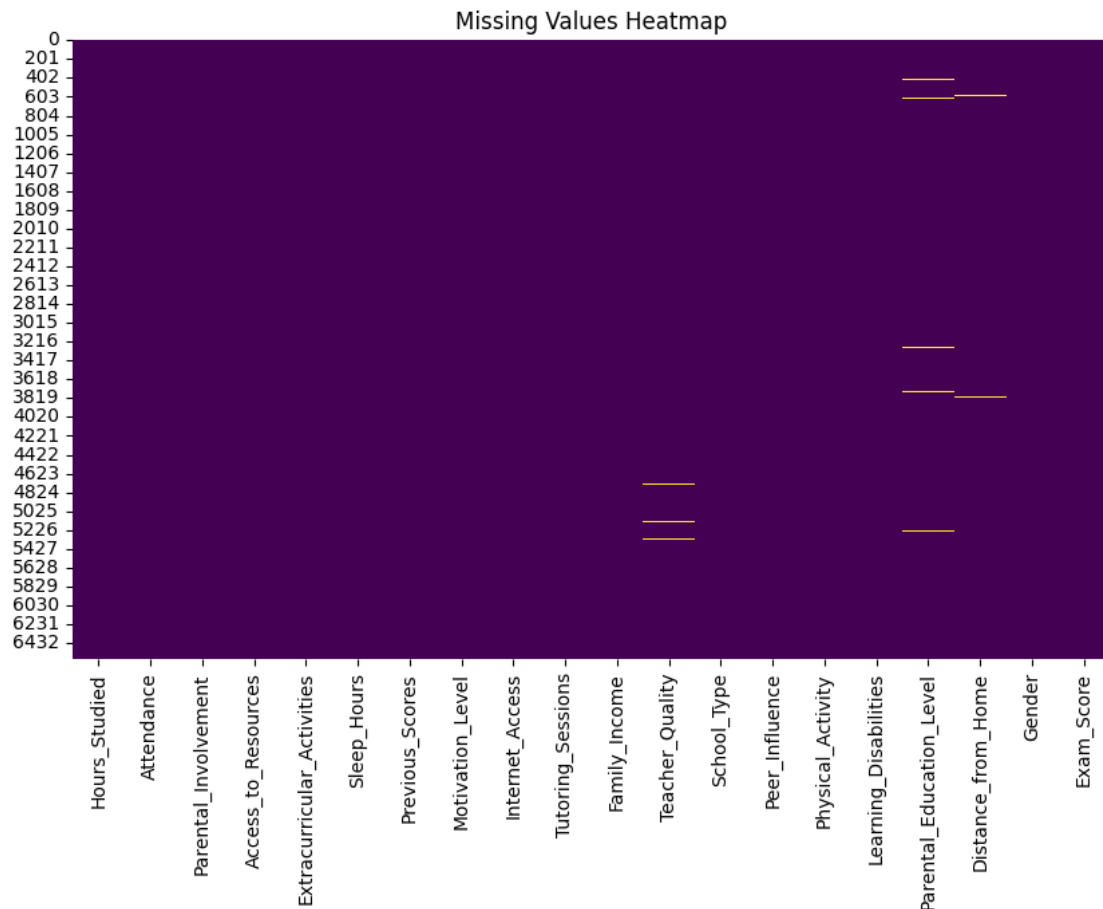
```
MISSING VALUES PER COLUMN:
Hours_Studied      0
Attendance         0
```

Parental_Involvement	0
Access_to_Resources	0
Extracurricular_Activities	0
Sleep_Hours	0
Previous_Scores	0
Motivation_Level	0
Internet_Access	0
Tutoring_Sessions	0
Family_Income	0
Teacher_Quality	78
School_Type	0
Peer_Influence	0
Physical_Activity	0
Learning_Disabilities	0
Parental_Education_Level	90
Distance_from_Home	67
Gender	0
Exam_Score	0

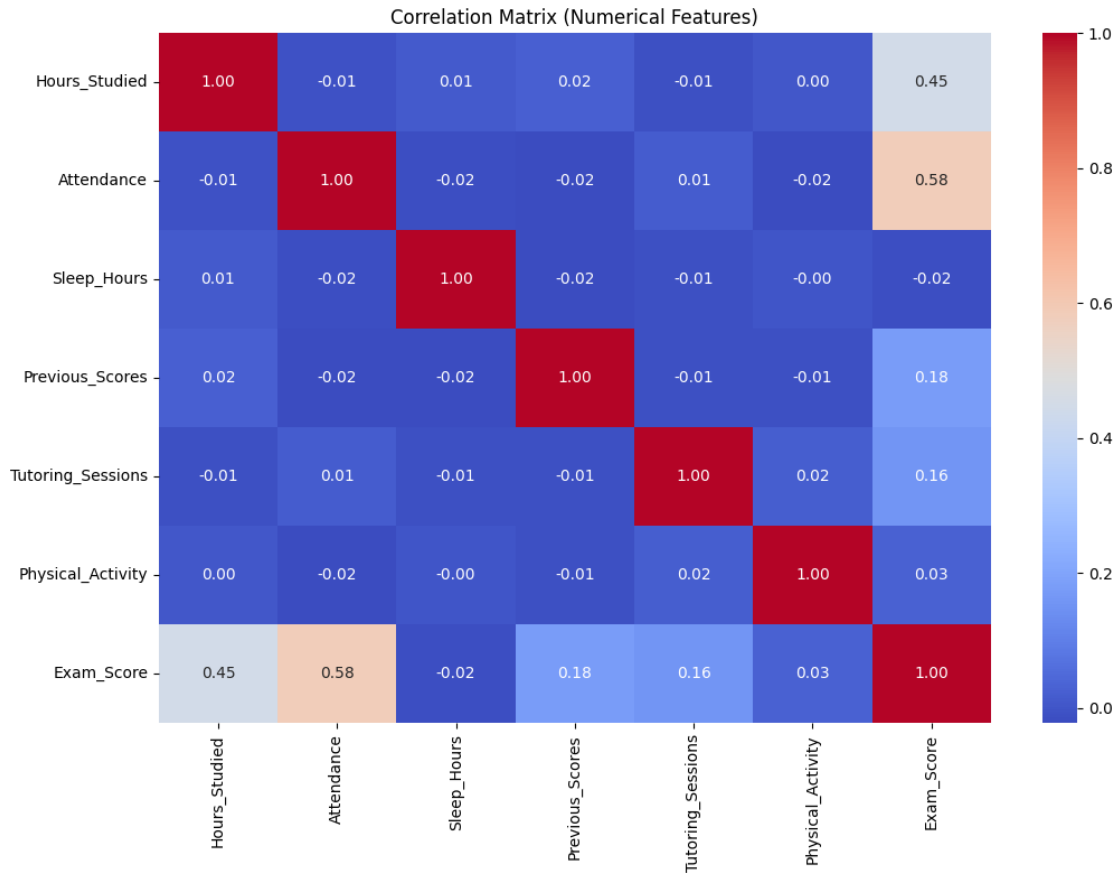
dtype: int64

2. EXPLORATORY DATA ANALYSIS (EDA) & VISUALIZATIONS

```
[19]: # 2.1 Missing Values Heatmap
plt.figure(figsize=(10, 6))
sns.heatmap(df.isnull(), cmap='viridis', cbar=False)
plt.title('Missing Values Heatmap')
plt.show()
```

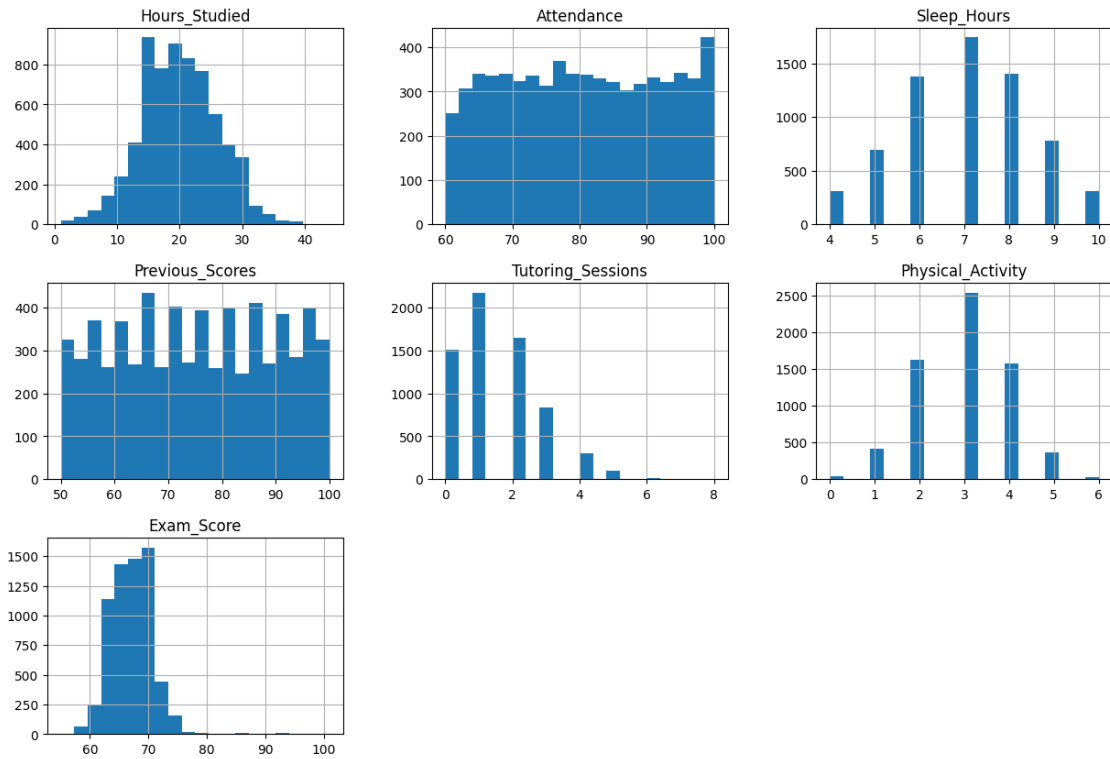


```
[20]: # 2.2 Correlation Matrix (for numerical features)
numerical_cols = ['Hours_Studied', 'Attendance', 'Sleep_Hours',
                  'Previous_Scores', 'Tutoring_Sessions', 'Physical_Activity',
                  'Exam_Score']
plt.figure(figsize=(12, 8))
sns.heatmap(df[numerical_cols].corr(), annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Correlation Matrix (Numerical Features)')
plt.show()
```

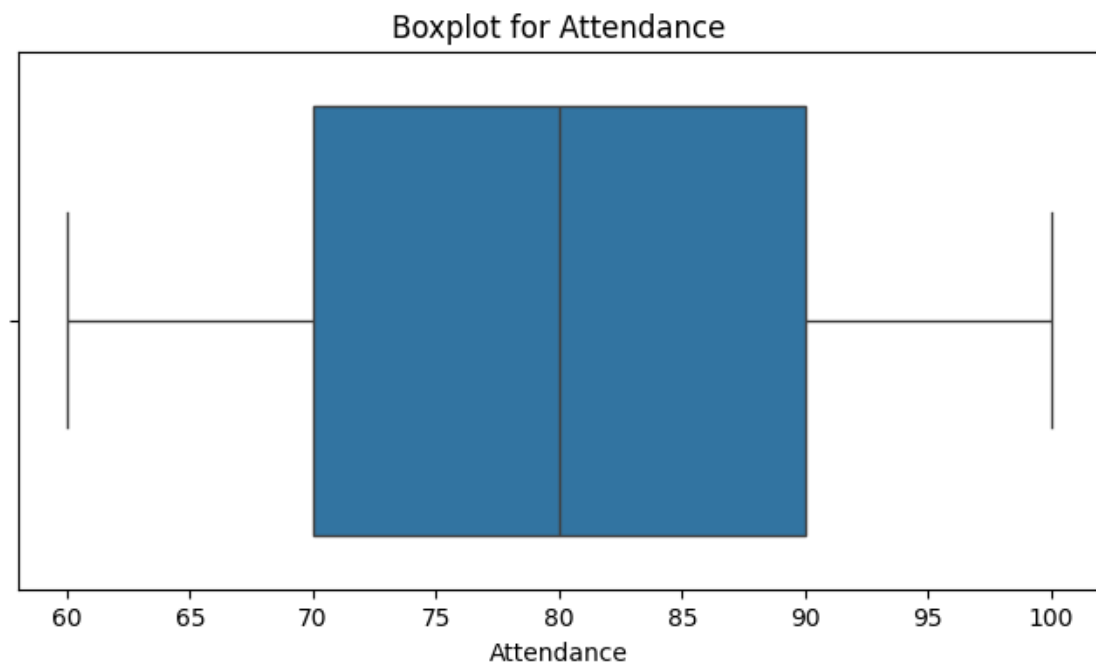
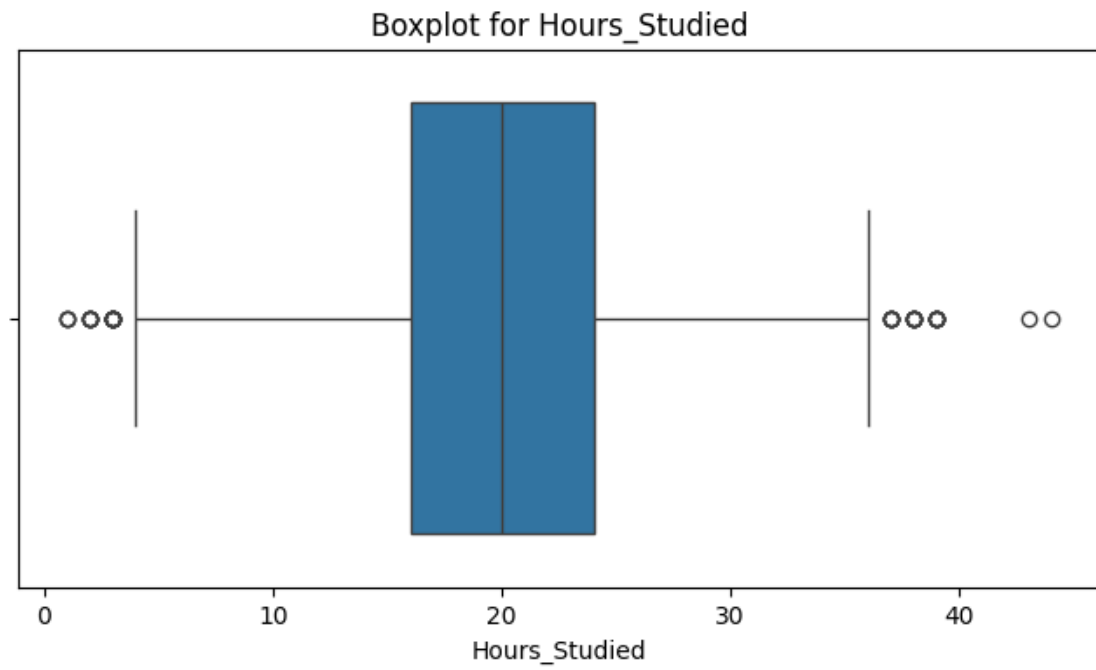


```
[21]: # 2.3 Histograms for Numerical Features
df[numerical_cols].hist(bins=20, figsize=(15, 10))
plt.suptitle('Histograms of Numerical Features')
plt.show()
```

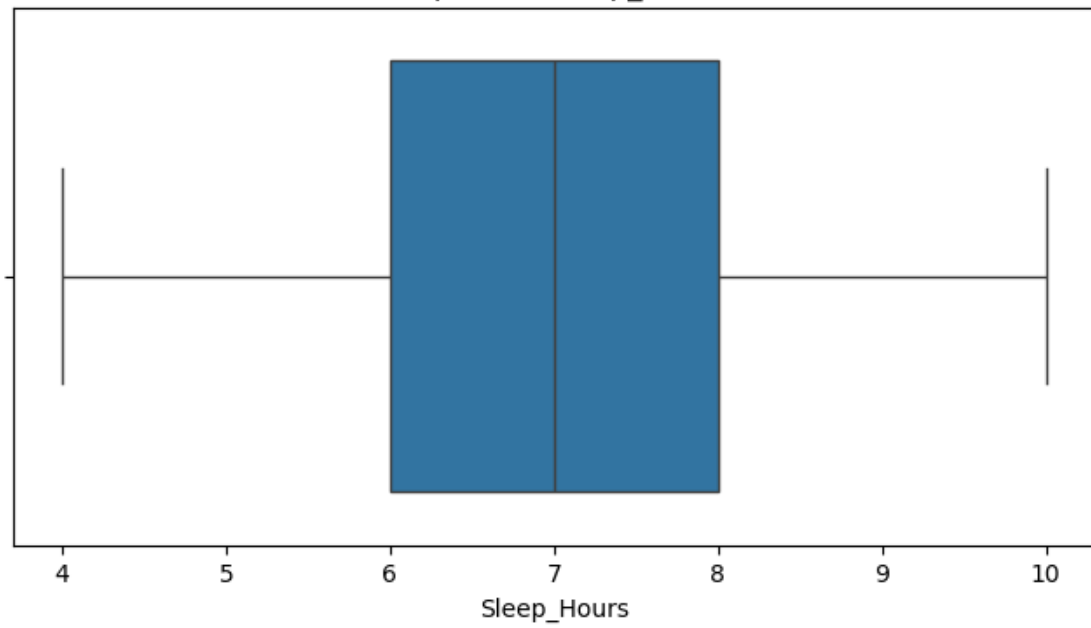
Histograms of Numerical Features



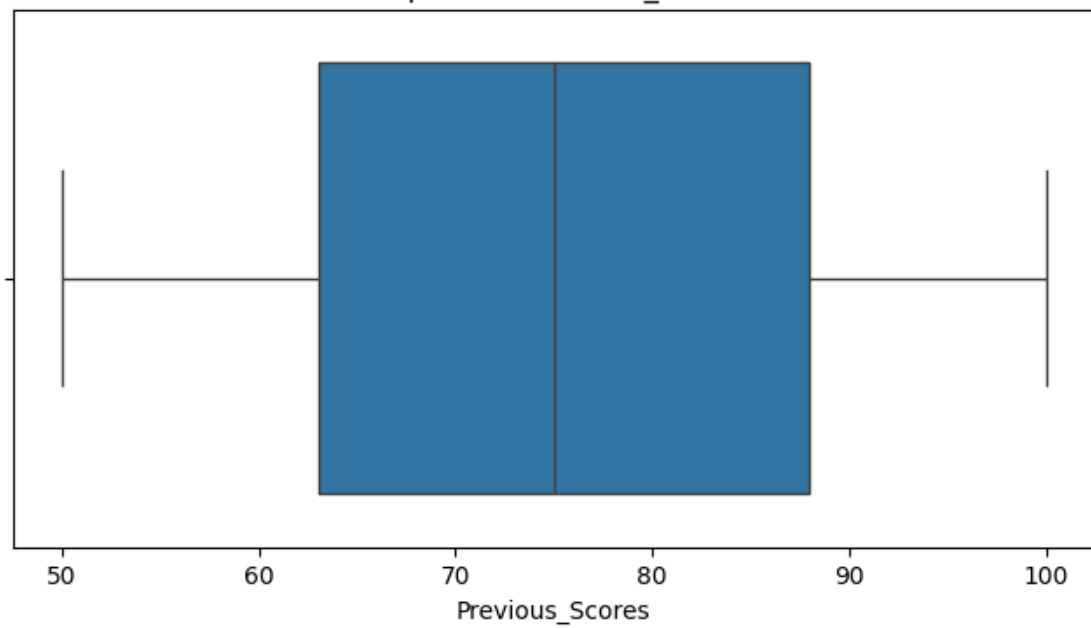
```
[22]: # 2.4 Boxplots for Numerical Features
for col in numerical_cols:
    plt.figure(figsize=(8, 4))
    sns.boxplot(x=df[col])
    plt.title(f'Boxplot for {col}')
    plt.show()
```

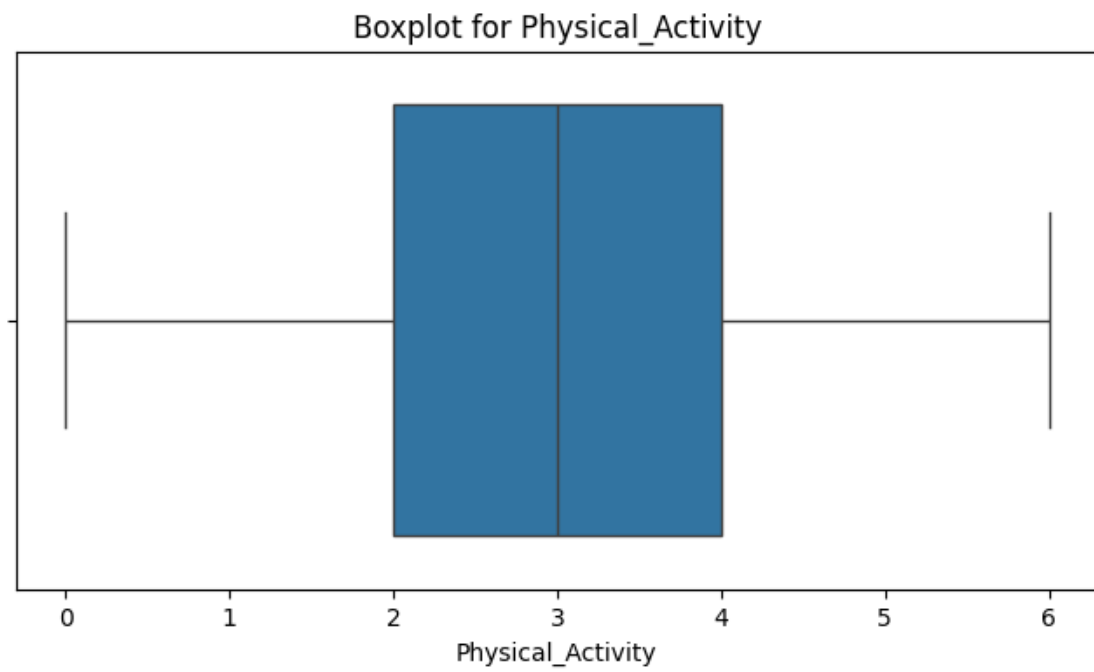
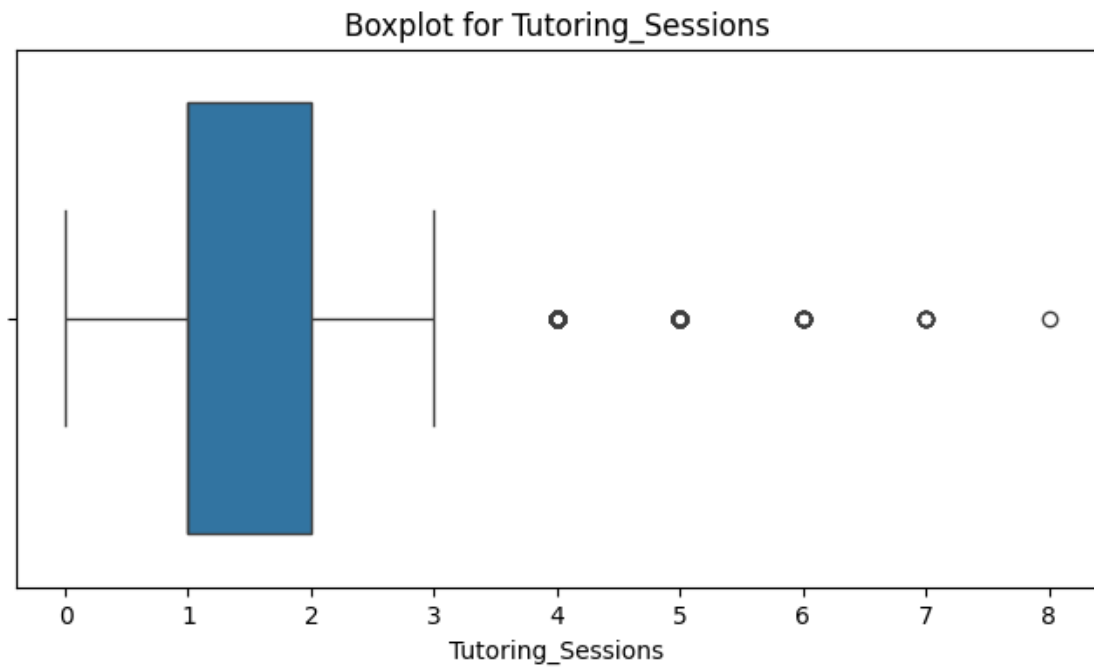


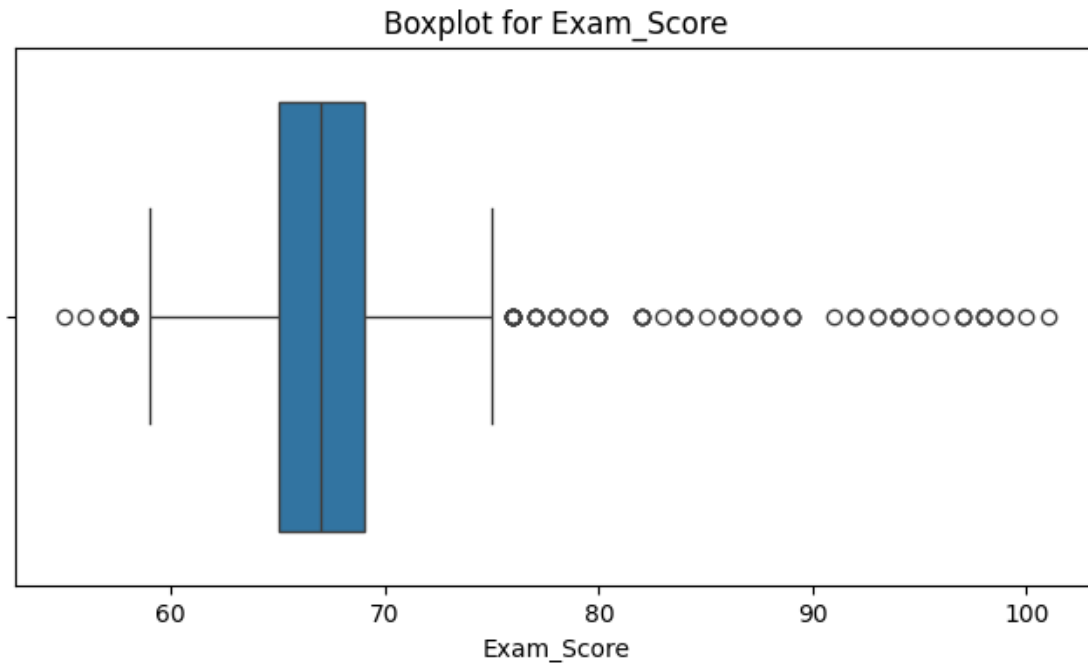
Boxplot for Sleep_Hours



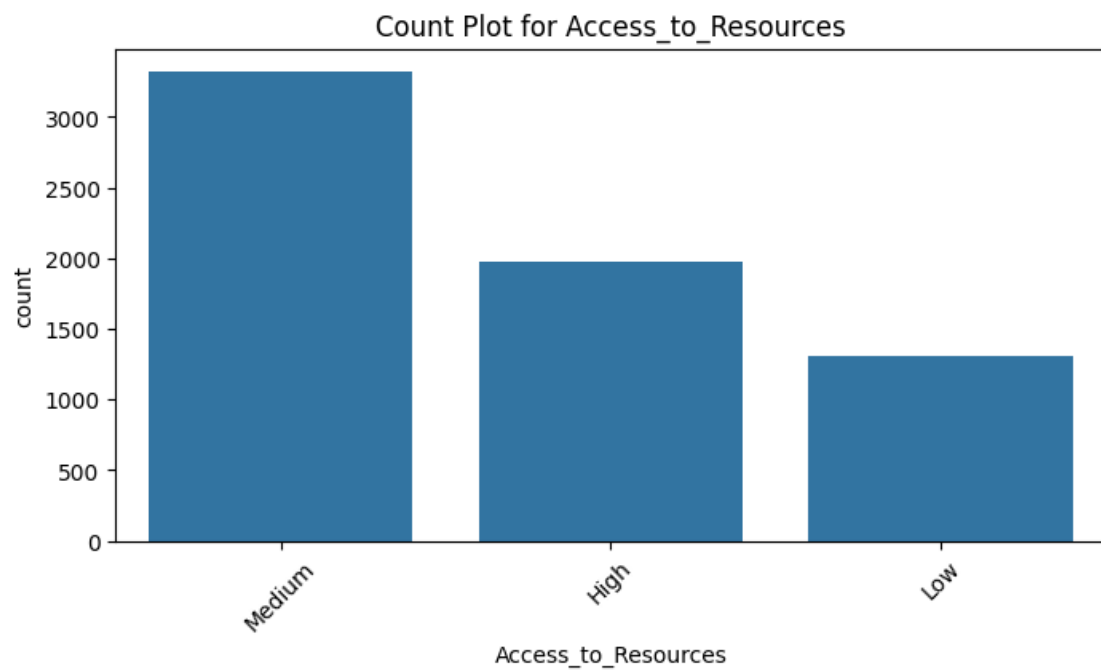
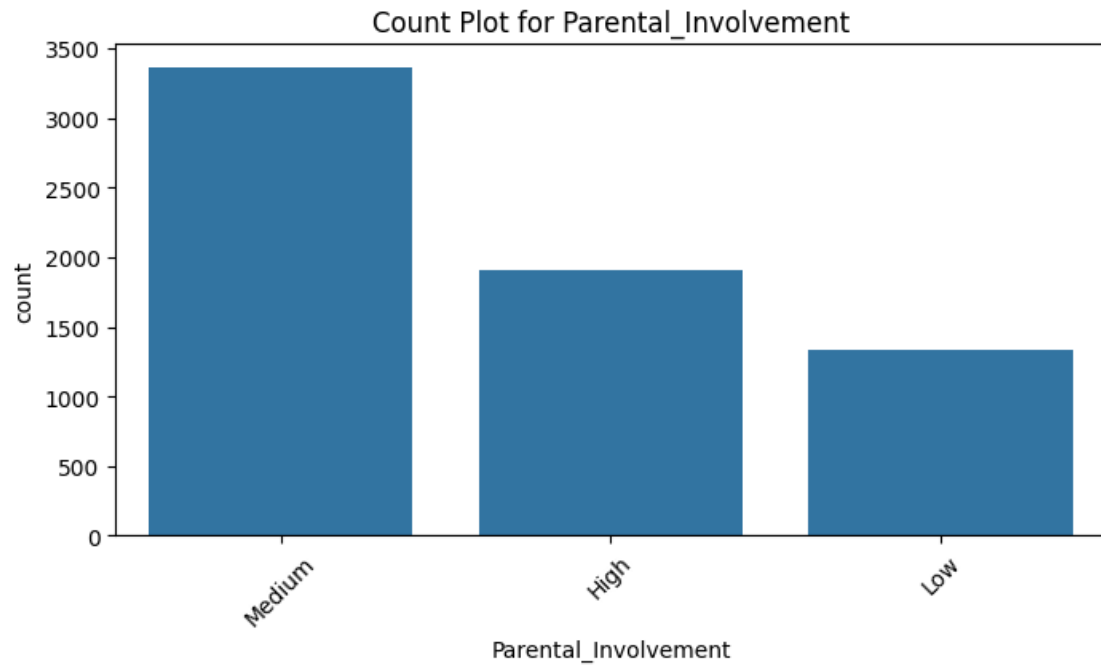
Boxplot for Previous_Scores

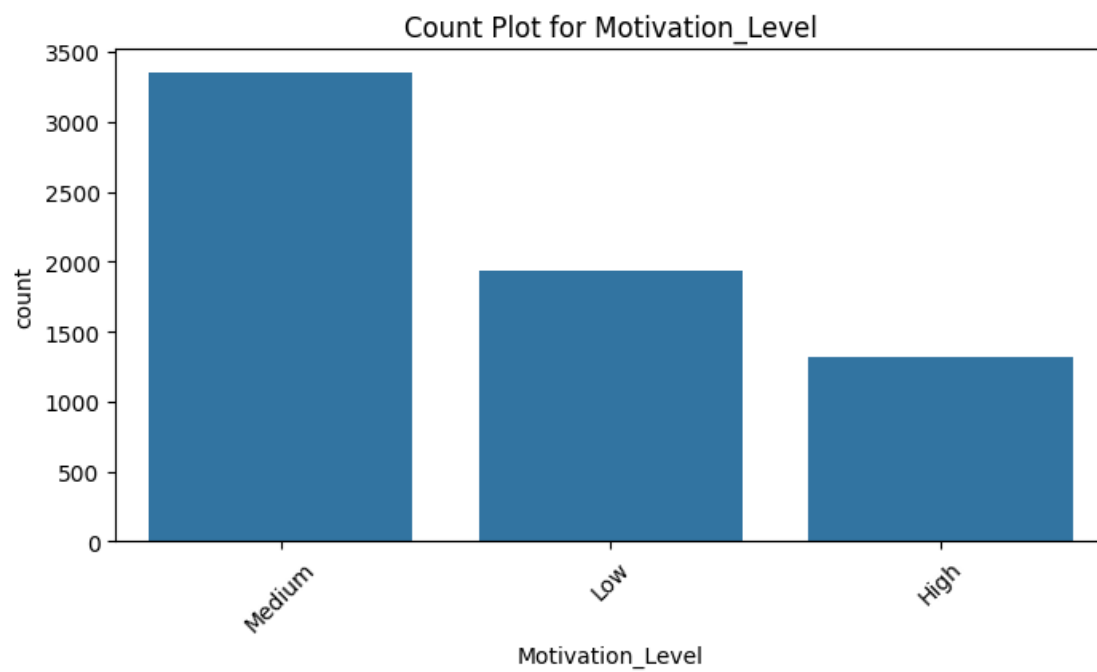
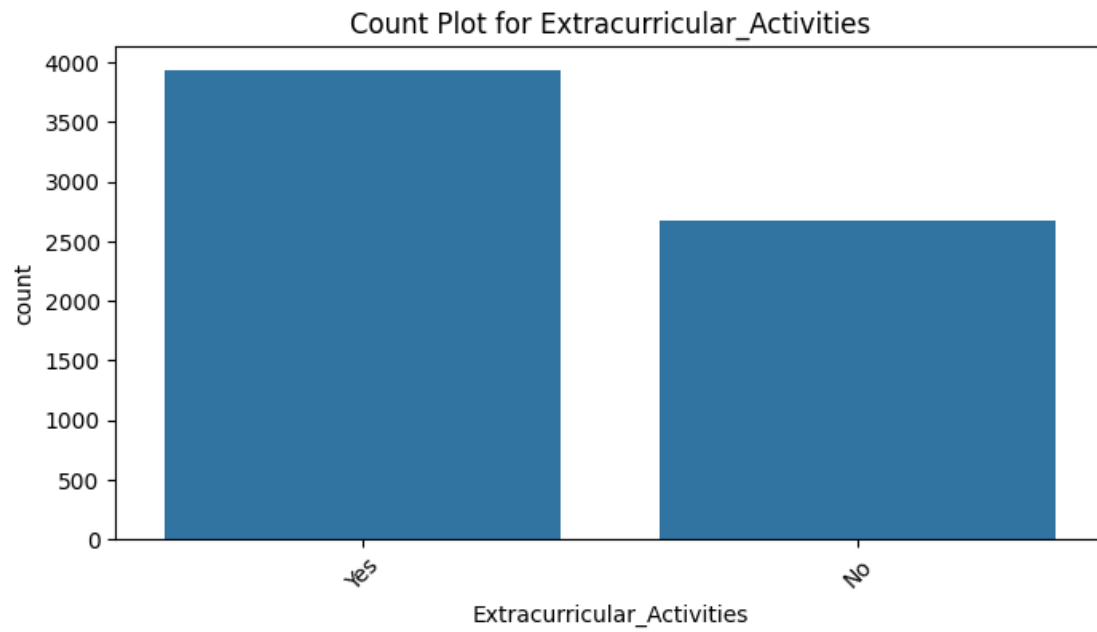


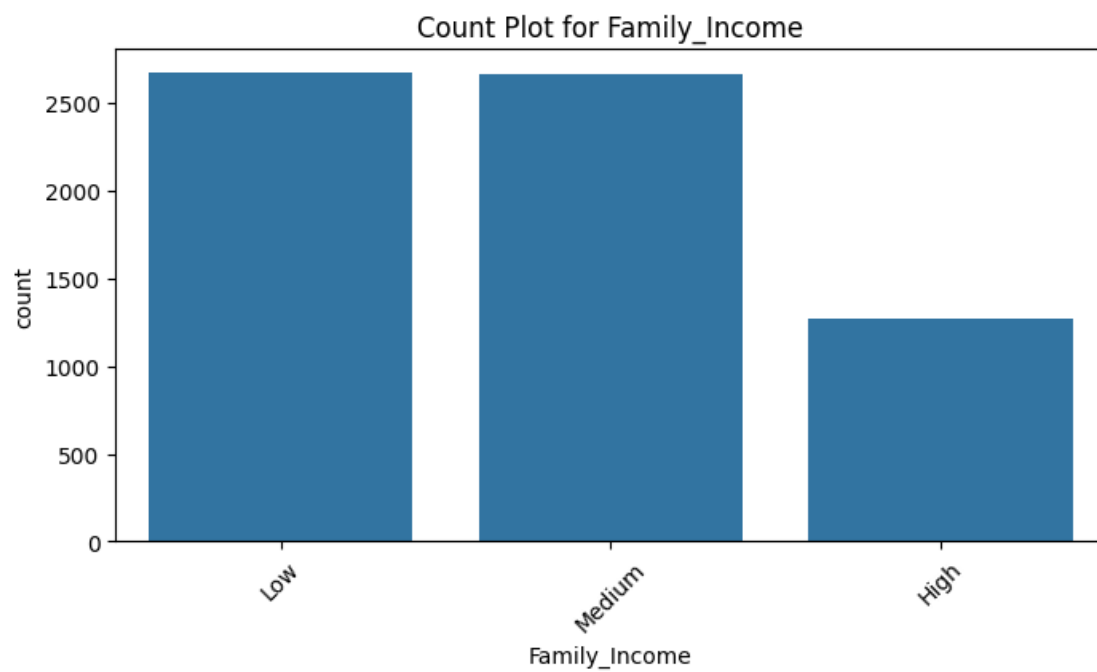
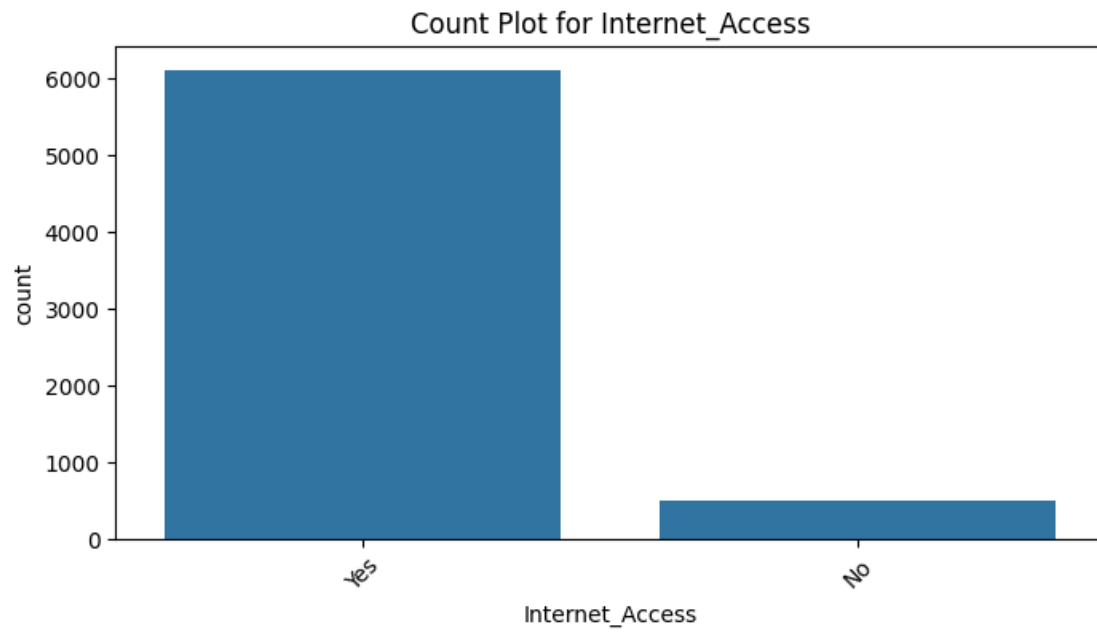


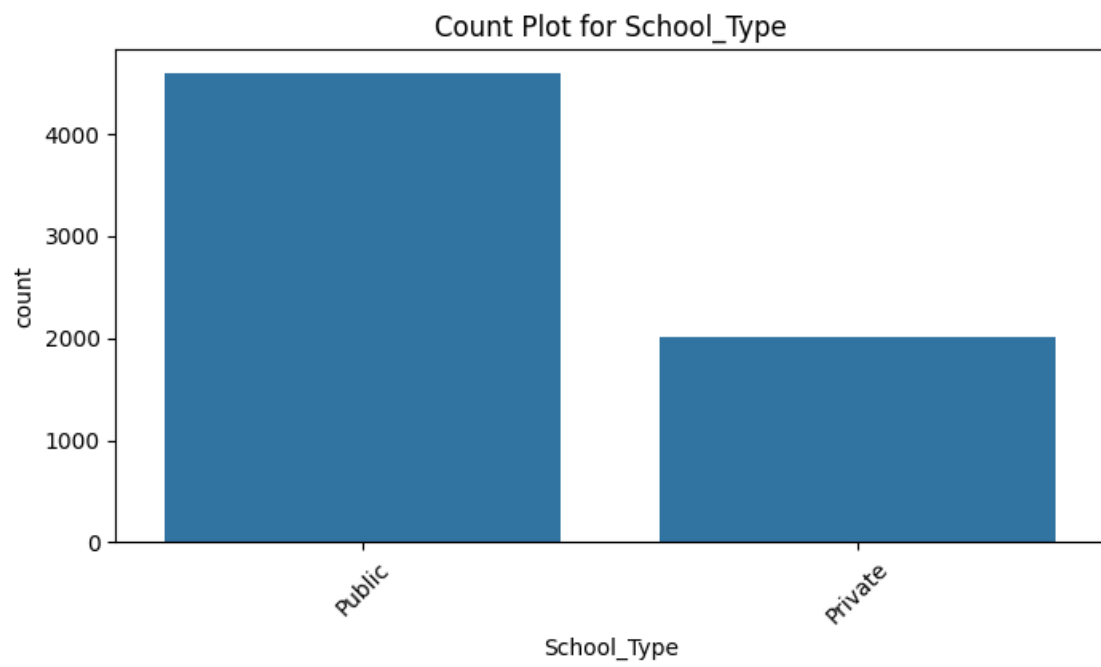
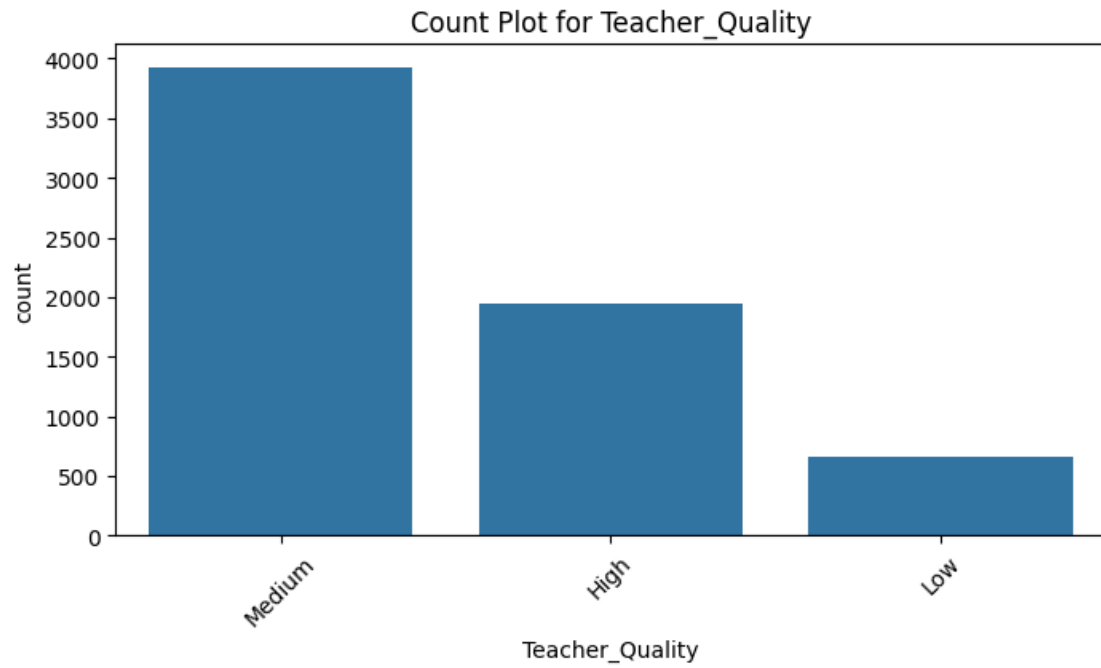


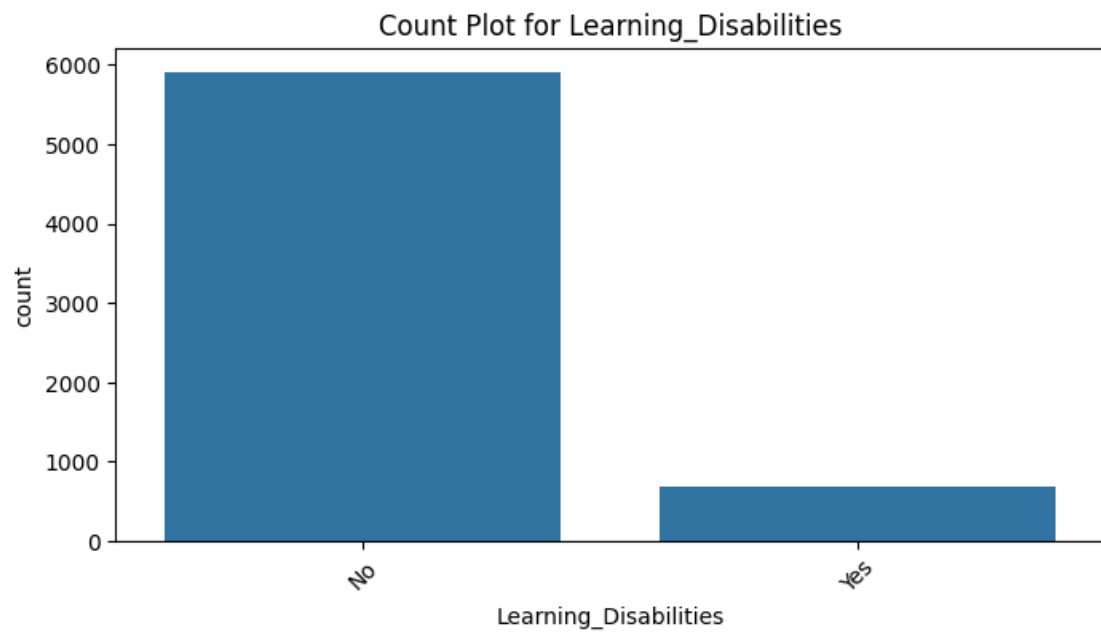
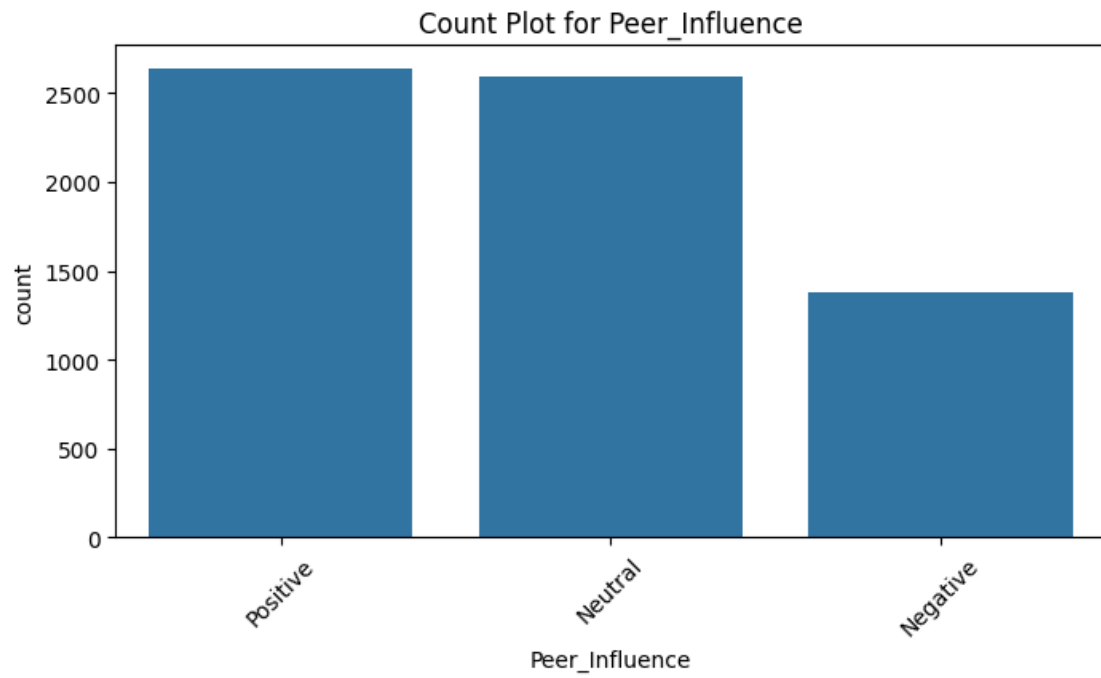
```
[23]: # 2.5 Count Plots for Categorical Features
categorical_cols = ['Parental_Involvement', 'Access_to_Resources',
    ↪ 'Extracurricular_Activities',
    ↪ 'Motivation_Level', 'Internet_Access', 'Family_Income',
    ↪ 'Teacher_Quality',
    ↪ 'School_Type', 'Peer_Influence', 'Learning_Disabilities',
    ↪ 'Parental_Education_Level', 'Distance_from_Home', 'Gender']
for col in categorical_cols:
    plt.figure(figsize=(8, 4))
    sns.countplot(data=df, x=col, order=df[col].value_counts().index)
    plt.title(f'Count Plot for {col}')
    plt.xticks(rotation=45)
    plt.show()
```

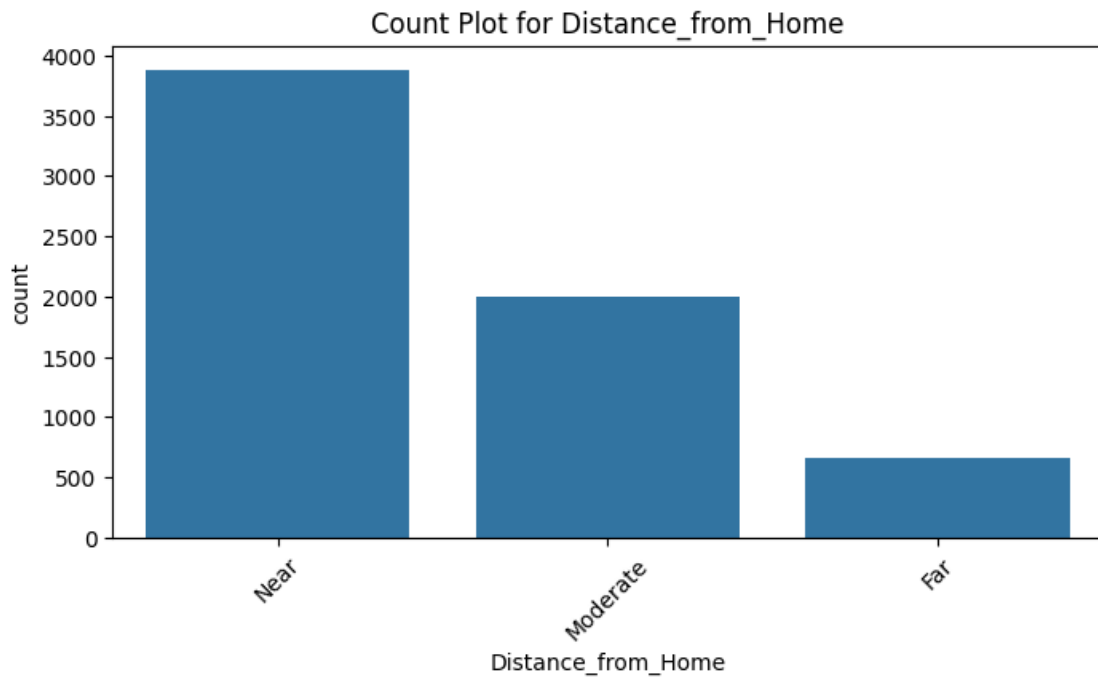
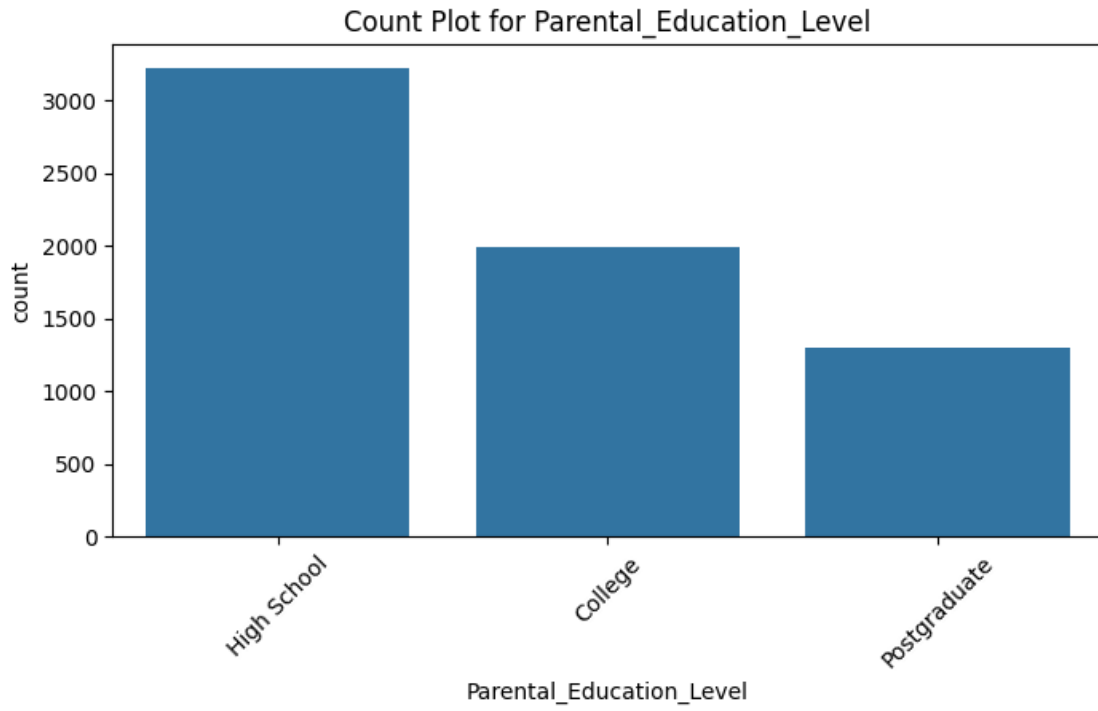


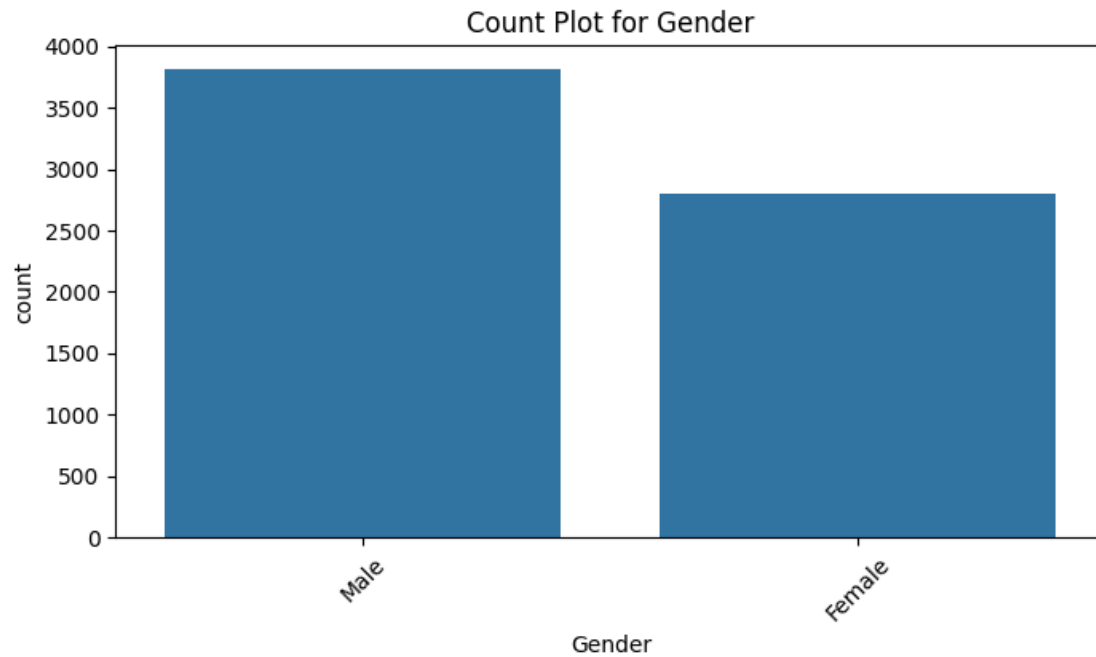




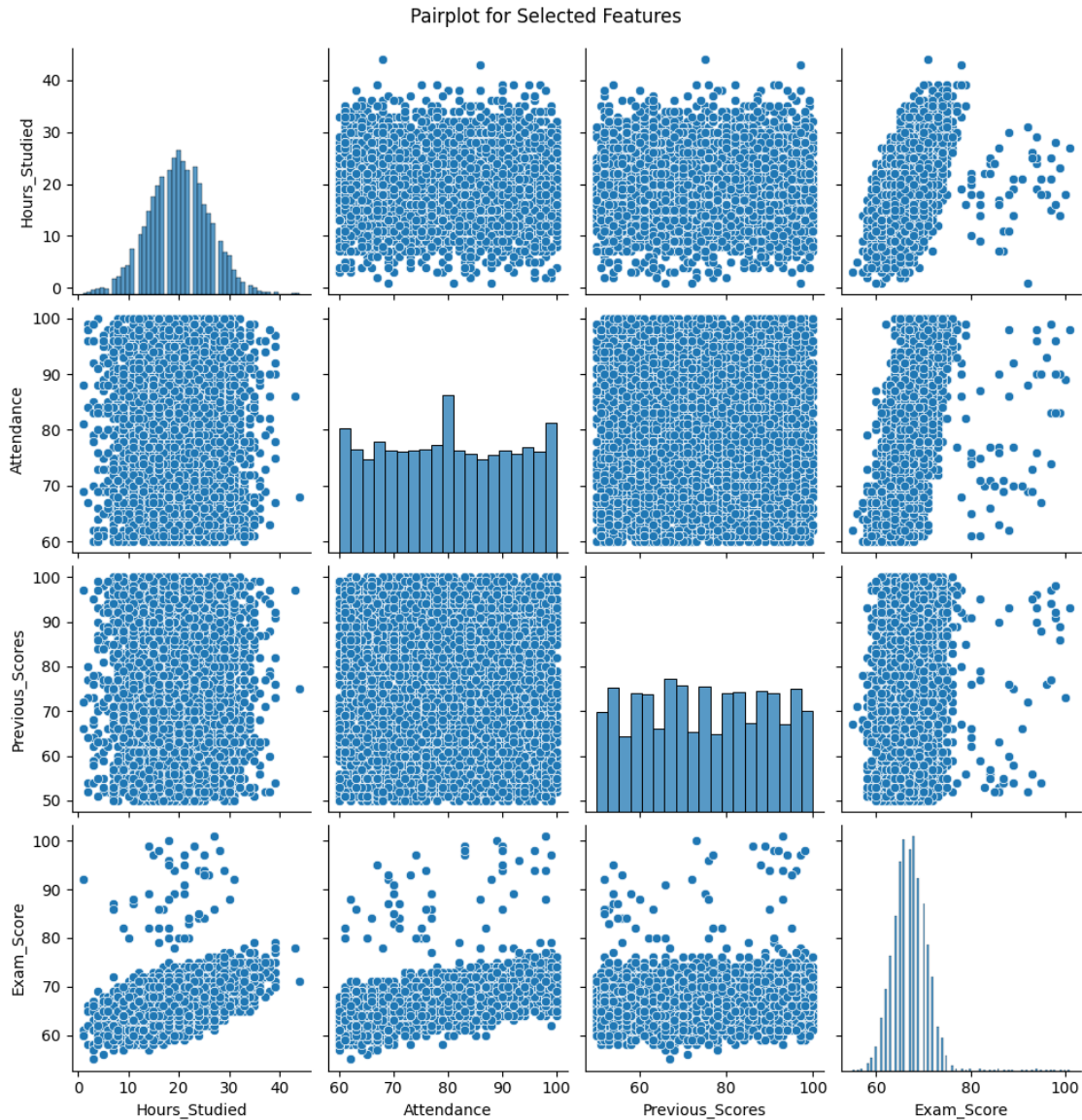








```
[24]: # 2.6 Pairplot (subset of features for clarity)
subset_cols = ['Hours_Studied', 'Attendance', 'Previous_Scores', 'Exam_Score']
sns.pairplot(df[subset_cols])
plt.suptitle('Pairplot for Selected Features', y=1.02)
plt.show()
```



0.0.1 3. DATA PREPROCESSING

```
[26]: # Fix missing values without inplace=True
for col in df.columns:
    if df[col].dtype == 'object':
        df[col] = df[col].fillna(df[col].mode()[0])
    else:
        df[col] = df[col].fillna(df[col].median())

# Define mapping dictionaries for ordinal features:
mapping_dicts = {
    'Parental_Involvement': {'Low': 0, 'Medium': 1, 'High': 2},
```

```

    'Access_to_Resources': {'Low': 0, 'Medium': 1, 'High': 2},
    'Motivation_Level': {'Low': 0, 'Medium': 1, 'High': 2},
    'Family_Income': {'Low': 0, 'Medium': 1, 'High': 2},
    'Teacher_Quality': {'Low': 0, 'Medium': 1, 'High': 2},
    'Parental_Education_Level': {'High School': 0, 'College': 1, 'Postgraduate':
↪ 2},
    'Distance_from_Home': {'Near': 0, 'Moderate': 1, 'Far': 2}
}
for col, mapping in mapping_dicts.items():
    if col in df.columns:
        df[col] = df[col].map(mapping)

# Map binary features
binary_mapping = {'Yes': 1, 'No': 0}
binary_cols = ['Extracurricular_Activities', 'Internet_Access',
↪ 'Learning_Disabilities']
for col in binary_cols:
    if col in df.columns:
        df[col] = df[col].map(binary_mapping)

# For Gender and School_Type (nominal), use LabelEncoder
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df['Gender'] = le.fit_transform(df['Gender'])
df['School_Type'] = le.fit_transform(df['School_Type'])

# Normalize numerical features using StandardScaler
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
df[numerical_cols] = scaler.fit_transform(df[numerical_cols])

```

```

[28]: # Convert non-numeric columns to numeric using Label Encoding or proper mapping
from sklearn.preprocessing import LabelEncoder

# Apply Label Encoding to non-numeric columns
le = LabelEncoder()
for col in X.columns:
    if X[col].dtype == 'object':
        X[col] = le.fit_transform(X[col])

# Train-test split
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)

# Train Random Forest Regressor
from sklearn.ensemble import RandomForestRegressor

```

```

rf_model = RandomForestRegressor(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)

# Make predictions
y_pred = rf_model.predict(X_test)

# Evaluate the model
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
print("Regression Model Evaluation:")
print("MAE:", mean_absolute_error(y_test, y_pred))
print("MSE:", mean_squared_error(y_test, y_pred))
print("R2 Score:", r2_score(y_test, y_pred))

# Visualize Feature Importance
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

importances = rf_model.feature_importances_
features = X.columns
sorted_indices = np.argsort(importances)[::-1]

plt.figure(figsize=(12, 6))
sns.barplot(x=importances[sorted_indices], y=features[sorted_indices],
            palette='coolwarm')
plt.title('Feature Importance for Exam Score Prediction')
plt.xlabel('Importance')
plt.ylabel('Feature')
plt.show()

```

Regression Model Evaluation:

MAE: 0.36610804440674277

MSE: 0.3992196714922818

R2 Score: 0.5725856264764716

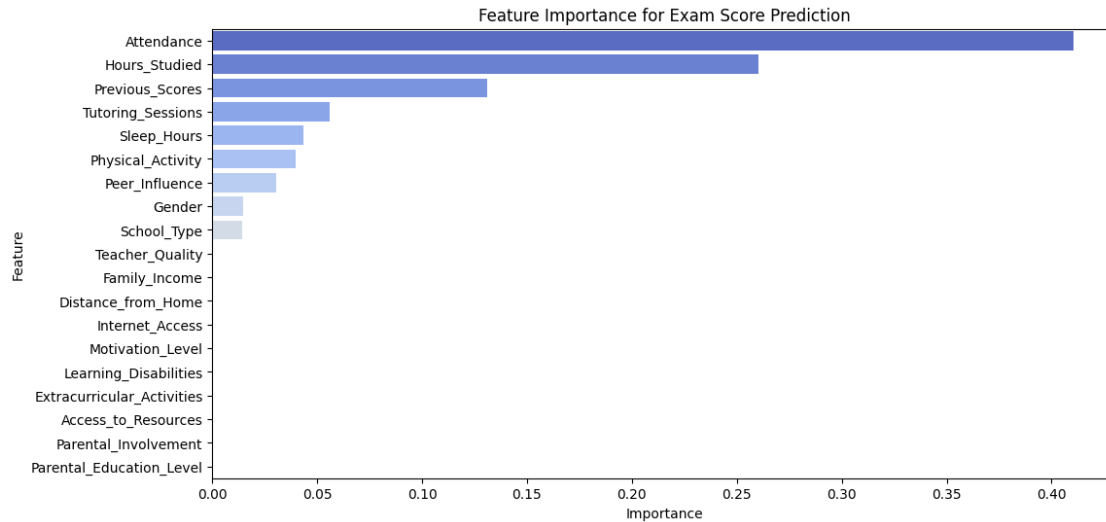
<ipython-input-28-8af86d708512>:39: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```

sns.barplot(x=importances[sorted_indices], y=features[sorted_indices],
            palette='coolwarm')

```



```
[40]: from sklearn.impute import SimpleImputer
from sklearn.cluster import KMeans

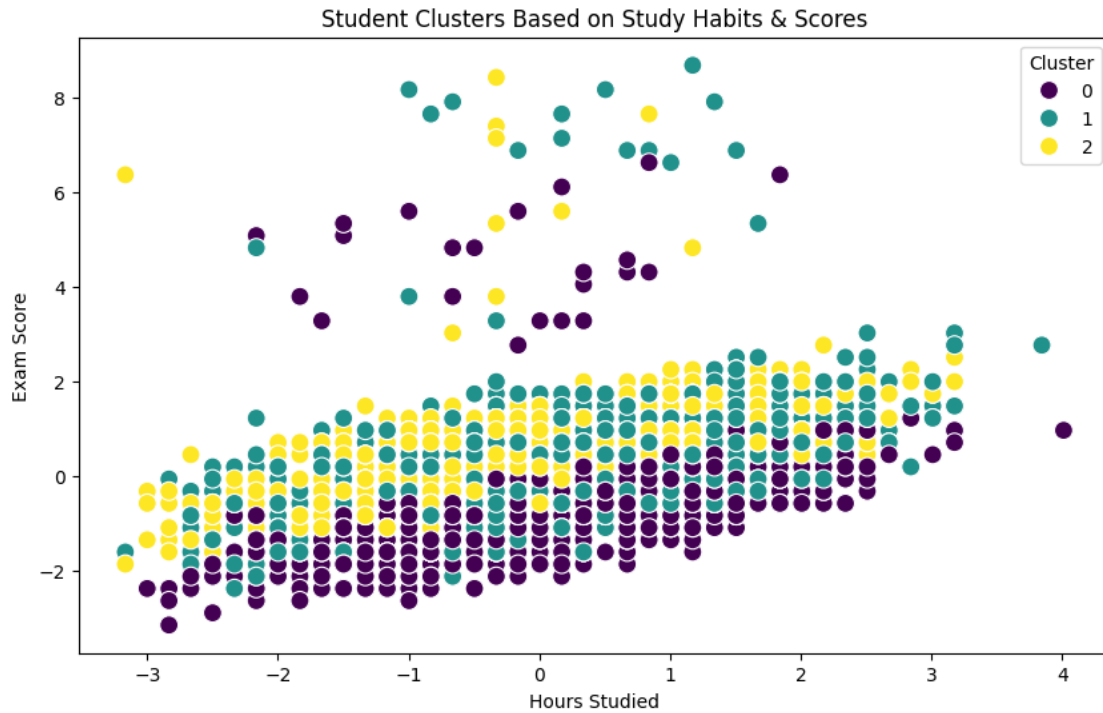
# Define clustering features
clustering_features = ['Hours_Studied', 'Motivation_Level', 'Attendance',
    ↪ 'Previous_Scores']

# Fill columns with only NaNs with 0 (or another default value)
df[clustering_features] = df[clustering_features].fillna(0)

# Handle missing values by filling with the mean of each column (optional)
imputer = SimpleImputer(strategy='mean')
df[clustering_features] = pd.DataFrame(imputer.
    ↪ fit_transform(df[clustering_features]), columns=clustering_features)

# Apply KMeans clustering
kmeans = KMeans(n_clusters=3, random_state=42, n_init=10)
df['Cluster'] = kmeans.fit_predict(df[clustering_features])

# Visualize clusters
plt.figure(figsize=(10, 6))
sns.scatterplot(x=df['Hours_Studied'], y=df['Exam_Score'], hue=df['Cluster'],
    ↪ palette='viridis', s=100)
plt.title('Student Clusters Based on Study Habits & Scores')
plt.xlabel('Hours Studied')
plt.ylabel('Exam Score')
plt.show()
```



```
[43]: def generate_recommendation(row, df):
    """
    Generates a detailed recommendation based on every feature in the student_
    ↪data.
    This function uses the original (non-normalized, non-encoded) values.
    """
    recs = []

    # Hours Studied
    if row['Hours_Studied'] < df['Hours_Studied'].median():
        recs.append("Increase study hours to build a stronger knowledge base.")
    else:
        recs.append("Study hours are satisfactory; continue your routine.")

    # Attendance
    if row['Attendance'] < df['Attendance'].median():
        recs.append("Improve attendance to avoid missing key lessons.")
    else:
        recs.append("Your attendance is commendable.")

    # Parental Involvement
    if row['Parental_Involvement'] == 'Low':
        recs.append("Enhance parental involvement for better academic support.")
    elif row['Parental_Involvement'] == 'Medium':
```

```

        recs.append("Consider discussing progress with parents to boost_
↪involvement.")
    else:
        recs.append("Parental involvement is excellent; keep it up.")

    # Access to Resources
    if row['Access_to_Resources'] == 'Low':
        recs.append("Work on improving access to study materials and resources.
↪")
    elif row['Access_to_Resources'] == 'Medium':
        recs.append("Better resource access may help improve learning outcomes.
↪")
    else:
        recs.append("Access to resources is strong; leverage them further.")

    # Extracurricular Activities
    if row['Extracurricular_Activities'] == 'No':
        recs.append("Consider joining extracurricular activities to boost_
↪overall development.")
    else:
        recs.append("Extracurricular engagement is a positive aspect of your_
↪profile.")

    # Sleep Hours
    if row['Sleep_Hours'] < 7:
        recs.append("Increase sleep hours for improved concentration and energy.
↪")
    else:
        recs.append("Your sleep schedule is on track.")

    # Previous Scores
    if row['Previous_Scores'] < df['Previous_Scores'].median():
        recs.append("Review previous scores and identify areas needing_
↪improvement.")
    else:
        recs.append("Maintain your strong performance from past scores.")

    # Motivation Level
    if row['Motivation_Level'] == 'Low':
        recs.append("Boost your motivation through goal-setting and counseling.
↪")
    elif row['Motivation_Level'] == 'Medium':
        recs.append("Explore techniques to further enhance your motivation.")
    else:
        recs.append("Your motivation is a strength; keep the momentum.")

```



```

# Internet Access
if row['Internet_Access'] == 'No':
    recs.append("Ensure reliable internet access to utilize online_
↪resources.")
else:
    recs.append("Internet access is sufficient for your learning needs.")

# Tutoring Sessions
if row['Tutoring_Sessions'] == 0:
    recs.append("Consider scheduling tutoring sessions to address_
↪challenging topics.")
else:
    recs.append("Tutoring sessions are being effectively utilized.")

# Family Income
if row['Family_Income'] == 'Low':
    recs.append("Explore free resources and scholarship opportunities to_
↪supplement your studies.")
else:
    recs.append("Your financial context supports access to additional_
↪learning materials.")

# Teacher Quality
if row['Teacher_Quality'] == 'Low':
    recs.append("Supplement classroom learning with external tutoring due_
↪to teacher quality concerns.")
elif row['Teacher_Quality'] == 'Medium':
    recs.append("Consider extra help to reinforce lessons taught in class.")
else:
    recs.append("Teacher quality is strong; maximize the benefits of_
↪classroom instruction.")

# School Type
recs.append(f"Being in a {row['School_Type']} school can offer unique_
↪advantages-leverage them accordingly.")

# Peer Influence
if row['Peer_Influence'] == 'Negative':
    recs.append("Work on improving peer influence by joining positive study_
↪groups.")
elif row['Peer_Influence'] == 'Neutral':
    recs.append("Engage with peers to build a supportive academic network.")
else:
    recs.append("Positive peer influence is an asset; maintain your network.
↪")

```

```

    # Physical Activity
    if row['Physical_Activity'] < 3: # assuming a scale where <3 indicates low
↪activity
        recs.append("Increase physical activity to boost concentration and
↪health.")
    else:
        recs.append("Your level of physical activity supports overall
↪well-being.")

    # Learning Disabilities
    if row['Learning_Disabilities'] == 'Yes':
        recs.append("Utilize special education resources and strategies to
↪address learning disabilities.")
    else:
        recs.append("No learning disabilities detected.")

    # Parental Education Level
    recs.append(f"Your parental education level
↪({row['Parental_Education_Level']}) may influence home support-discuss
↪strategies with your family.")

    # Distance from Home
    if row['Distance_from_Home'] in ['Moderate', 'Far']:
        recs.append("Consider time management techniques to overcome challenges
↪due to longer commute.")
    else:
        recs.append("Proximity to school is an advantage for regular attendance.
↪")

    # Gender
    recs.append(f"Your gender ({row['Gender']}) can be factored into tailored
↪guidance if needed.")

    return "\n".join(recs)

```

```

[48]: def main_pipeline():
    # --- EDA and Preprocessing already executed above ---
    # (See sections 1-4)

    # Regression & Clustering evaluations have been printed and visualized.

    # Generate personalized recommendations for the first 10 students using the
↪original (non-processed) data.
    original_df = pd.read_csv('student_data.csv')

    # Impute missing values for original_df using updated Pandas syntax

```

```

for col in original_df.columns:
    if original_df[col].dtype == 'object':
        original_df[col] = original_df[col].fillna(original_df[col].
mode()[0])
    else:
        original_df[col] = original_df[col].fillna(original_df[col].
median())

print("\nPERSONALIZED LEARNING RECOMMENDATIONS (First 10 Students):")
for i in range(10):
    print(f"\n--- Student {i+1} ---")
    print(generate_recommendation(original_df.iloc[i], original_df))

# --- Example: Predict Exam Score and Generate Recommendation for a New
Student ---
def predict_and_recommend(new_student):
    """
    Accepts a dictionary 'new_student' with the original (non-processed)
feature values.

    Returns the predicted exam score (from the regression model), the
cluster-based recommendation,
    and a detailed personalized recommendation based on every feature.
    """
    # Generate detailed recommendation based on original values
    recommendation = generate_recommendation(new_student, original_df)

    # Preprocess new_student for prediction:
    new_df = pd.DataFrame([new_student])

    # Fill missing values
    for col in new_df.columns:
        if new_df[col].dtype == 'object':
            new_df[col] = new_df[col].fillna(original_df[col].mode()[0])
        else:
            new_df[col] = new_df[col].fillna(original_df[col].median())

    # Map ordinal features
    for col, mapping in mapping_dicts.items():
        if col in new_df.columns:
            new_df[col] = new_df[col].map(mapping)

    # Map binary features
    for col in binary_cols:
        if col in new_df.columns:
            new_df[col] = new_df[col].map(binary_mapping)

# --- Handle unseen labels for 'Gender' and 'School_Type' ---

```

```

# Convert to lowercase to ensure consistency
new_df['Gender'] = new_df['Gender'].str.lower()
new_df['School_Type'] = new_df['School_Type'].str.lower()

# Handle unseen labels: add new labels dynamically
for col in ['Gender', 'School_Type', 'Peer_Influence']:
    if not set(new_df[col]).issubset(le.classes_):
        le.classes_ = np.append(le.classes_, new_df[col].unique())
    new_df[col] = le.transform(new_df[col])

# --- Exclude non-numeric columns from scaling ---
categorical_cols = ['Gender', 'School_Type', 'Peer_Influence'] # Don't
↳scale these
cols_to_scale = [col for col in numerical_cols if col not in
↳categorical_cols]

# Scale only numerical features (using the same scaler as before)
new_df[cols_to_scale] = scaler.transform(new_df[cols_to_scale])

# Predict exam score using the regression model
predicted_score = rf_model.predict(new_df.
↳drop(columns=['Exam_Score']))[0]

# Predict cluster using selected clustering features
predicted_cluster = kmeans.predict(new_df[clustering_features])[0]
if predicted_cluster == 0:
    cluster_recommendation = "High Performer: Consider advanced
↳coursework and mentorship."
elif predicted_cluster == 1:
    cluster_recommendation = "Moderate Performer: Focus on structured
↳study plans and additional tutoring."
else:
    cluster_recommendation = "Low Performer: Increase parental
↳involvement, improve sleep habits, and seek structured coaching."

return (f"Predicted Exam Score: {predicted_score:.2f}\n"
        f"Cluster Recommendation: {cluster_recommendation}\n\n"
        f"Detailed Personalized Recommendation:\n{recommendation}")

# Example new student data (original values)
new_student = {
    'Hours_Studied': 5,
    'Attendance': 90,
    'Parental_Involvement': 'Low',
    'Access_to_Resources': 'High',
    'Extracurricular_Activities': 'No',

```

```

        'Sleep_Hours': 6,
        'Previous_Scores': 65,
        'Motivation_Level': 'Low',
        'Internet_Access': 'Yes',
        'Tutoring_Sessions': 0,
        'Family_Income': 'Low',
        'Teacher_Quality': 'Low',
        'School_Type': 'Public',
        'Peer_Influence': 'Negative',
        'Physical_Activity': 2,
        'Learning_Disabilities': 'No',
        'Parental_Education_Level': 'High School',
        'Distance_from_Home': 'Near',
        'Gender': 'Male',
        'Exam_Score': 0 # Placeholder (target is not used in prediction)
    }

    print("\nNEW STUDENT PREDICTION & RECOMMENDATION:")
    print(predict_and_recommend(new_student))

if __name__ == '__main__':
    main_pipeline()

```

PERSONALIZED LEARNING RECOMMENDATIONS (First 10 Students):

```

--- Student 1 ---
Study hours are satisfactory; continue your routine.
Your attendance is commendable.
Enhance parental involvement for better academic support.
Access to resources is strong; leverage them further.
Consider joining extracurricular activities to boost overall development.
Your sleep schedule is on track.
Review previous scores and identify areas needing improvement.
Boost your motivation through goal-setting and counseling.
Internet access is sufficient for your learning needs.
Consider scheduling tutoring sessions to address challenging topics.
Explore free resources and scholarship opportunities to supplement your studies.
Consider extra help to reinforce lessons taught in class.
Being in a Public school can offer unique advantages-leverage them accordingly.
Positive peer influence is an asset; maintain your network.
Your level of physical activity supports overall well-being.
No learning disabilities detected.
Your parental education level (High School) may influence home support-discuss
strategies with your family.
Proximity to school is an advantage for regular attendance.
Your gender (Male) can be factored into tailored guidance if needed.

```

--- Student 2 ---

Increase study hours to build a stronger knowledge base.
Improve attendance to avoid missing key lessons.
Enhance parental involvement for better academic support.
Better resource access may help improve learning outcomes.
Consider joining extracurricular activities to boost overall development.
Your sleep schedule is on track.
Review previous scores and identify areas needing improvement.
Boost your motivation through goal-setting and counseling.
Internet access is sufficient for your learning needs.
Tutoring sessions are being effectively utilized.
Your financial context supports access to additional learning materials.
Consider extra help to reinforce lessons taught in class.
Being in a Public school can offer unique advantages-leverage them accordingly.
Work on improving peer influence by joining positive study groups.
Your level of physical activity supports overall well-being.
No learning disabilities detected.
Your parental education level (College) may influence home support-discuss strategies with your family.
Consider time management techniques to overcome challenges due to longer commute.
Your gender (Female) can be factored into tailored guidance if needed.

--- Student 3 ---

Study hours are satisfactory; continue your routine.
Your attendance is commendable.
Consider discussing progress with parents to boost involvement.
Better resource access may help improve learning outcomes.
Extracurricular engagement is a positive aspect of your profile.
Your sleep schedule is on track.
Maintain your strong performance from past scores.
Explore techniques to further enhance your motivation.
Internet access is sufficient for your learning needs.
Tutoring sessions are being effectively utilized.
Your financial context supports access to additional learning materials.
Consider extra help to reinforce lessons taught in class.
Being in a Public school can offer unique advantages-leverage them accordingly.
Engage with peers to build a supportive academic network.
Your level of physical activity supports overall well-being.
No learning disabilities detected.
Your parental education level (Postgraduate) may influence home support-discuss strategies with your family.
Proximity to school is an advantage for regular attendance.
Your gender (Male) can be factored into tailored guidance if needed.

--- Student 4 ---

Study hours are satisfactory; continue your routine.

Your attendance is commendable.
Enhance parental involvement for better academic support.
Better resource access may help improve learning outcomes.
Extracurricular engagement is a positive aspect of your profile.
Your sleep schedule is on track.
Maintain your strong performance from past scores.
Explore techniques to further enhance your motivation.
Internet access is sufficient for your learning needs.
Tutoring sessions are being effectively utilized.
Your financial context supports access to additional learning materials.
Consider extra help to reinforce lessons taught in class.
Being in a Public school can offer unique advantages-leverage them accordingly.
Work on improving peer influence by joining positive study groups.
Your level of physical activity supports overall well-being.
No learning disabilities detected.
Your parental education level (High School) may influence home support-discuss strategies with your family.
Consider time management techniques to overcome challenges due to longer commute.
Your gender (Male) can be factored into tailored guidance if needed.

--- Student 5 ---

Increase study hours to build a stronger knowledge base.
Your attendance is commendable.
Consider discussing progress with parents to boost involvement.
Better resource access may help improve learning outcomes.
Extracurricular engagement is a positive aspect of your profile.
Increase sleep hours for improved concentration and energy.
Review previous scores and identify areas needing improvement.
Explore techniques to further enhance your motivation.
Internet access is sufficient for your learning needs.
Tutoring sessions are being effectively utilized.
Your financial context supports access to additional learning materials.
Teacher quality is strong; maximize the benefits of classroom instruction.
Being in a Public school can offer unique advantages-leverage them accordingly.
Engage with peers to build a supportive academic network.
Your level of physical activity supports overall well-being.
No learning disabilities detected.
Your parental education level (College) may influence home support-discuss strategies with your family.
Proximity to school is an advantage for regular attendance.
Your gender (Female) can be factored into tailored guidance if needed.

--- Student 6 ---

Increase study hours to build a stronger knowledge base.
Your attendance is commendable.
Consider discussing progress with parents to boost involvement.
Better resource access may help improve learning outcomes.

Extracurricular engagement is a positive aspect of your profile.
Your sleep schedule is on track.
Maintain your strong performance from past scores.
Explore techniques to further enhance your motivation.
Internet access is sufficient for your learning needs.
Tutoring sessions are being effectively utilized.
Your financial context supports access to additional learning materials.
Consider extra help to reinforce lessons taught in class.
Being in a Public school can offer unique advantages-leverage them accordingly.
Positive peer influence is an asset; maintain your network.
Your level of physical activity supports overall well-being.
No learning disabilities detected.
Your parental education level (Postgraduate) may influence home support-discuss strategies with your family.
Proximity to school is an advantage for regular attendance.
Your gender (Male) can be factored into tailored guidance if needed.

--- Student 7 ---

Study hours are satisfactory; continue your routine.
Your attendance is commendable.
Consider discussing progress with parents to boost involvement.
Work on improving access to study materials and resources.
Extracurricular engagement is a positive aspect of your profile.
Your sleep schedule is on track.
Review previous scores and identify areas needing improvement.
Boost your motivation through goal-setting and counseling.
Internet access is sufficient for your learning needs.
Tutoring sessions are being effectively utilized.
Explore free resources and scholarship opportunities to supplement your studies.
Consider extra help to reinforce lessons taught in class.
Being in a Private school can offer unique advantages-leverage them accordingly.
Engage with peers to build a supportive academic network.
Increase physical activity to boost concentration and health.
No learning disabilities detected.
Your parental education level (High School) may influence home support-discuss strategies with your family.
Consider time management techniques to overcome challenges due to longer commute.
Your gender (Male) can be factored into tailored guidance if needed.

--- Student 8 ---

Study hours are satisfactory; continue your routine.
Improve attendance to avoid missing key lessons.
Enhance parental involvement for better academic support.
Access to resources is strong; leverage them further.
Extracurricular engagement is a positive aspect of your profile.
Increase sleep hours for improved concentration and energy.
Review previous scores and identify areas needing improvement.

Explore techniques to further enhance your motivation.
Internet access is sufficient for your learning needs.
Tutoring sessions are being effectively utilized.
Your financial context supports access to additional learning materials.
Teacher quality is strong; maximize the benefits of classroom instruction.
Being in a Public school can offer unique advantages-leverage them accordingly.
Work on improving peer influence by joining positive study groups.
Increase physical activity to boost concentration and health.
No learning disabilities detected.
Your parental education level (High School) may influence home support-discuss strategies with your family.
Consider time management techniques to overcome challenges due to longer commute.
Your gender (Male) can be factored into tailored guidance if needed.

--- Student 9 ---

Increase study hours to build a stronger knowledge base.
Your attendance is commendable.
Consider discussing progress with parents to boost involvement.
Access to resources is strong; leverage them further.
Consider joining extracurricular activities to boost overall development.
Increase sleep hours for improved concentration and energy.
Maintain your strong performance from past scores.
Your motivation is a strength; keep the momentum.
Internet access is sufficient for your learning needs.
Consider scheduling tutoring sessions to address challenging topics.
Your financial context supports access to additional learning materials.
Supplement classroom learning with external tutoring due to teacher quality concerns.
Being in a Private school can offer unique advantages-leverage them accordingly.
Engage with peers to build a supportive academic network.
Increase physical activity to boost concentration and health.
No learning disabilities detected.
Your parental education level (College) may influence home support-discuss strategies with your family.
Proximity to school is an advantage for regular attendance.
Your gender (Male) can be factored into tailored guidance if needed.

--- Student 10 ---

Study hours are satisfactory; continue your routine.
Your attendance is commendable.
Consider discussing progress with parents to boost involvement.
Better resource access may help improve learning outcomes.
Extracurricular engagement is a positive aspect of your profile.
Your sleep schedule is on track.
Review previous scores and identify areas needing improvement.
Explore techniques to further enhance your motivation.
Internet access is sufficient for your learning needs.

Consider scheduling tutoring sessions to address challenging topics.
Your financial context supports access to additional learning materials.
Teacher quality is strong; maximize the benefits of classroom instruction.
Being in a Public school can offer unique advantages-leverage them accordingly.
Positive peer influence is an asset; maintain your network.
Your level of physical activity supports overall well-being.
No learning disabilities detected.
Your parental education level (High School) may influence home support-discuss strategies with your family.
Consider time management techniques to overcome challenges due to longer commute.
Your gender (Male) can be factored into tailored guidance if needed.

NEW STUDENT PREDICTION & RECOMMENDATION:

Predicted Exam Score: 2.77

Cluster Recommendation: Moderate Performer: Focus on structured study plans and additional tutoring.

Detailed Personalized Recommendation:

Increase study hours to build a stronger knowledge base.
Your attendance is commendable.
Enhance parental involvement for better academic support.
Access to resources is strong; leverage them further.
Consider joining extracurricular activities to boost overall development.
Increase sleep hours for improved concentration and energy.
Review previous scores and identify areas needing improvement.
Boost your motivation through goal-setting and counseling.
Internet access is sufficient for your learning needs.
Consider scheduling tutoring sessions to address challenging topics.
Explore free resources and scholarship opportunities to supplement your studies.
Supplement classroom learning with external tutoring due to teacher quality concerns.
Being in a Public school can offer unique advantages-leverage them accordingly.
Work on improving peer influence by joining positive study groups.
Increase physical activity to boost concentration and health.
No learning disabilities detected.
Your parental education level (High School) may influence home support-discuss strategies with your family.
Proximity to school is an advantage for regular attendance.
Your gender (Male) can be factored into tailored guidance if needed.