

Let's first import the packages

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

from sklearn.impute import SimpleImputer
from sklearn.preprocessing import OneHotEncoder, StandardScaler
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
```

Import and display dataset's head

```
In [2]: df = pd.read_csv('Exam_Score_Prediction.csv')
print(df.head())
```

	student_id	age	gender	course	study_hours	class_attendance	\
0	200.99	17	male	diploma	2.78	92.9	
1	200.99	23	other	bca	3.37	64.8	
2	200.99	22	male	b.sc	7.88	76.8	
3	200.99	20	other	diploma	0.67	48.4	
4	200.99	20	female	diploma	0.89	71.6	

	internet_access	sleep_hours	sleep_quality	study_method	facility_rating
0	yes	7.4	poor	coaching	low
1	yes	4.6	average	online videos	medium
2	yes	8.5	poor	coaching	high
3	yes	5.8	average	online videos	low
4	yes	9.8	poor	coaching	low

	exam_difficulty	exam_score
0	hard	58.9
1	moderate	54.8
2	moderate	90.3
3	moderate	29.7
4	moderate	43.7

Display Dataset information

```
In [3]: print(df.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20000 entries, 0 to 19999
Data columns (total 13 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0  student_id            20000 non-null  float64
 1  age                   20000 non-null  int64
 2  gender                20000 non-null  object
 3  course                20000 non-null  object
 4  study_hours           20000 non-null  float64
 5  class_attendance      20000 non-null  float64
 6  internet_access       20000 non-null  object
 7  sleep_hours           20000 non-null  float64
 8  sleep_quality         20000 non-null  object
 9  study_method          20000 non-null  object
10  facility_rating       20000 non-null  object
11  exam_difficulty       20000 non-null  object
12  exam_score            20000 non-null  float64
dtypes: float64(5), int64(1), object(7)
memory usage: 2.0+ MB
None
```

Check if there is any missing values

```
In [4]: print("\nMissing Values:\n", df.isna().sum())
df.describe(include='all')
```

```
Missing Values:
 student_id      0
 age             0
 gender          0
 course          0
 study_hours     0
 class_attendance 0
 internet_access 0
 sleep_hours     0
 sleep_quality   0
 study_method    0
 facility_rating 0
 exam_difficulty 0
 exam_score      0
dtype: int64
```

Out[4]:

	student_id	age	gender	course	study_hours	class_attendance
count	20000.000000	20000.000000	20000	20000	20000.000000	20000.000000
unique	NaN	NaN	3	7	NaN	NaN
top	NaN	NaN	other	bca	NaN	NaN
freq	NaN	NaN	6726	2902	NaN	NaN
mean	10000.500000	20.473300	NaN	NaN	4.007604	70.017365
std	5770.211372	2.284458	NaN	NaN	2.308313	17.282262
min	200.990000	17.000000	NaN	NaN	0.080000	40.600000
25%	5000.750000	18.000000	NaN	NaN	2.000000	55.100000
50%	10000.500000	20.000000	NaN	NaN	4.040000	69.900000
75%	15000.250000	22.000000	NaN	NaN	6.000000	85.000000
max	19800.010000	24.000000	NaN	NaN	7.910000	99.400000

There is no any missing values, so we move further

Divide into categorial columns

```
In [5]: numeric_cols = ["age", "study_hours", "class_attendance", "sleep_hours",
categorical_cols = [
    "gender", "course", "internet_access",
    "sleep_quality", "study_method",
    "facility_rating", "exam_difficulty"
]
```

If there is any missing values then fill with mean value


```
In [6]: # Create imputers
num_imputer = SimpleImputer(strategy='mean')
cat_imputer = SimpleImputer(strategy='most_frequent')

df[numeric_cols] = num_imputer.fit_transform(df[numeric_cols])
df[categorical_cols] = cat_imputer.fit_transform(df[categorical_cols])

df.head()
```

Out[6]:

	student_id	age	gender	course	study_hours	class_attendance	internet_access
0	200.99	17.0	male	diploma	2.78	92.9	yes
1	200.99	23.0	other	bca	3.37	64.8	yes
2	200.99	22.0	male	b.sc	7.88	76.8	yes
3	200.99	20.0	other	diploma	0.67	48.4	yes
4	200.99	20.0	female	diploma	0.89	71.6	yes




Encode categorical features (One-Hot Encoding)

In [7]: `df_encoded = pd.get_dummies(df, columns=categorical_cols, drop_first=True)`
`df_encoded.head()`

Out[7]:

	student_id	age	study_hours	class_attendance	sleep_hours	exam_score	gender_r
0	200.99	17.0	2.78	92.9	7.4	58.9	
1	200.99	23.0	3.37	64.8	4.6	54.8	F
2	200.99	22.0	7.88	76.8	8.5	90.3	
3	200.99	20.0	0.67	48.4	5.8	29.7	F
4	200.99	20.0	0.89	71.6	9.8	43.7	F

5 rows × 25 columns



Feature Scaling (StandardScaler)

In [8]: `scaler = StandardScaler()`
`df_scaled = df_encoded.copy()`
`df_scaled[numeric_cols] = scaler.fit_transform(df_scaled[numeric_cols])`
`df_scaled.head()`

Out[8]:

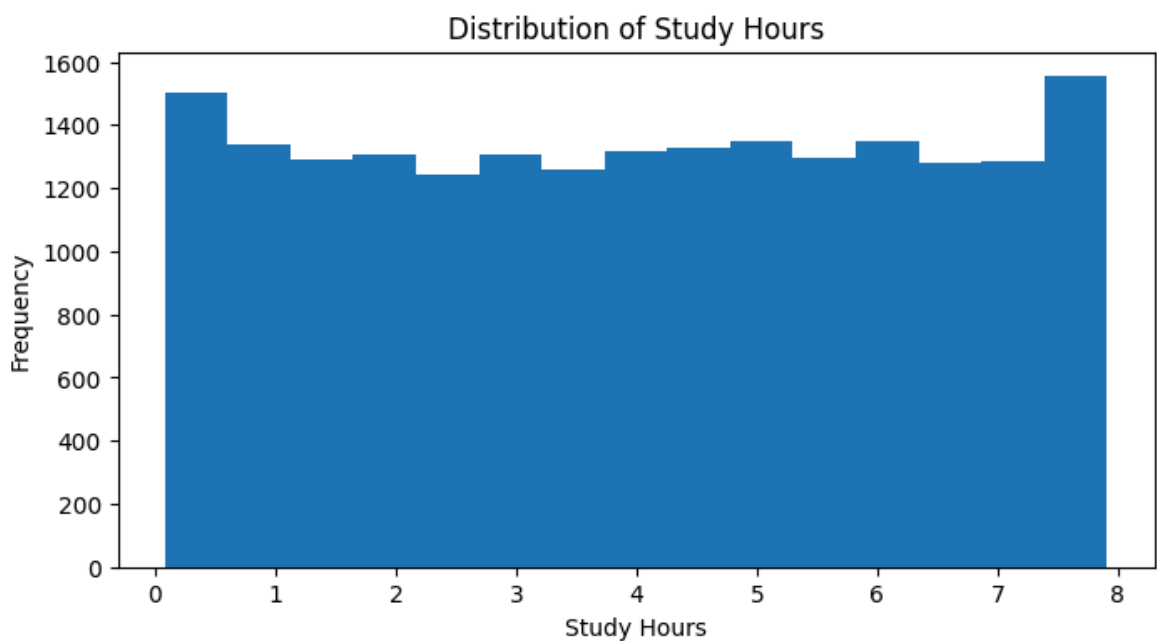
	student_id	age	study_hours	class_attendance	sleep_hours	exam_score	gen
0	200.99	-1.520442	-0.531832	1.324086	0.225999	-0.191095	
1	200.99	1.106067	-0.276227	-0.301899	-1.390586	-0.407934	
2	200.99	0.668315	1.677629	0.392472	0.861085	1.469576	
3	200.99	-0.207188	-1.445942	-1.250872	-0.697764	-1.735413	
4	200.99	-0.207188	-1.350632	0.091578	1.611643	-0.994987	

5 rows × 25 columns



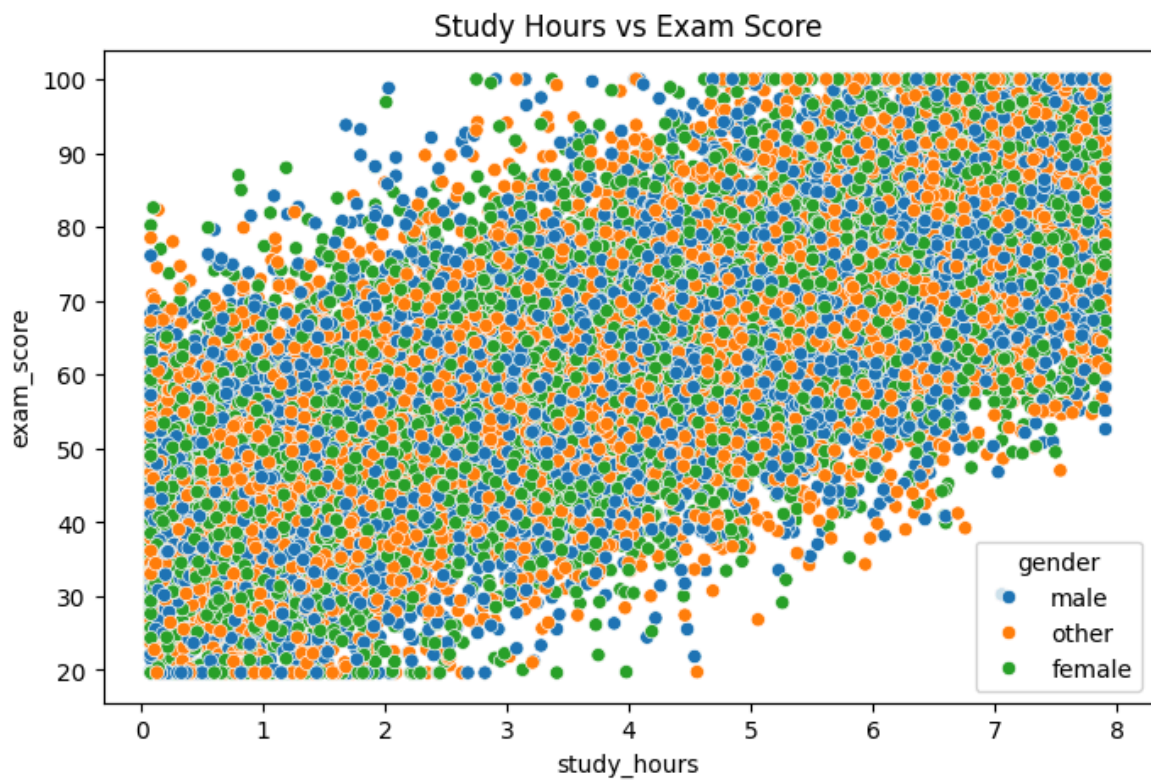
Histogram of a feature

```
In [9]: plt.figure(figsize=(8,4))
plt.hist(df['study_hours'], bins=15)
plt.title("Distribution of Study Hours")
plt.xlabel("Study Hours")
plt.ylabel("Frequency")
plt.show()
```

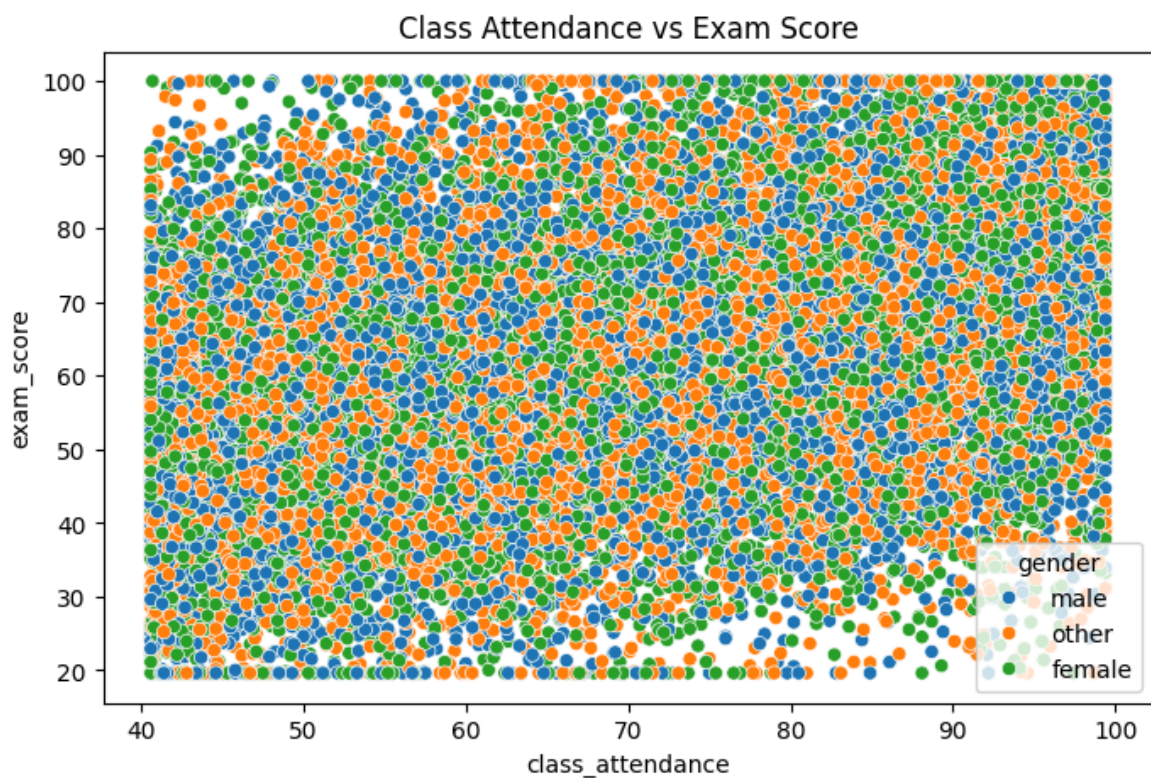


Scatter plot to see relationships

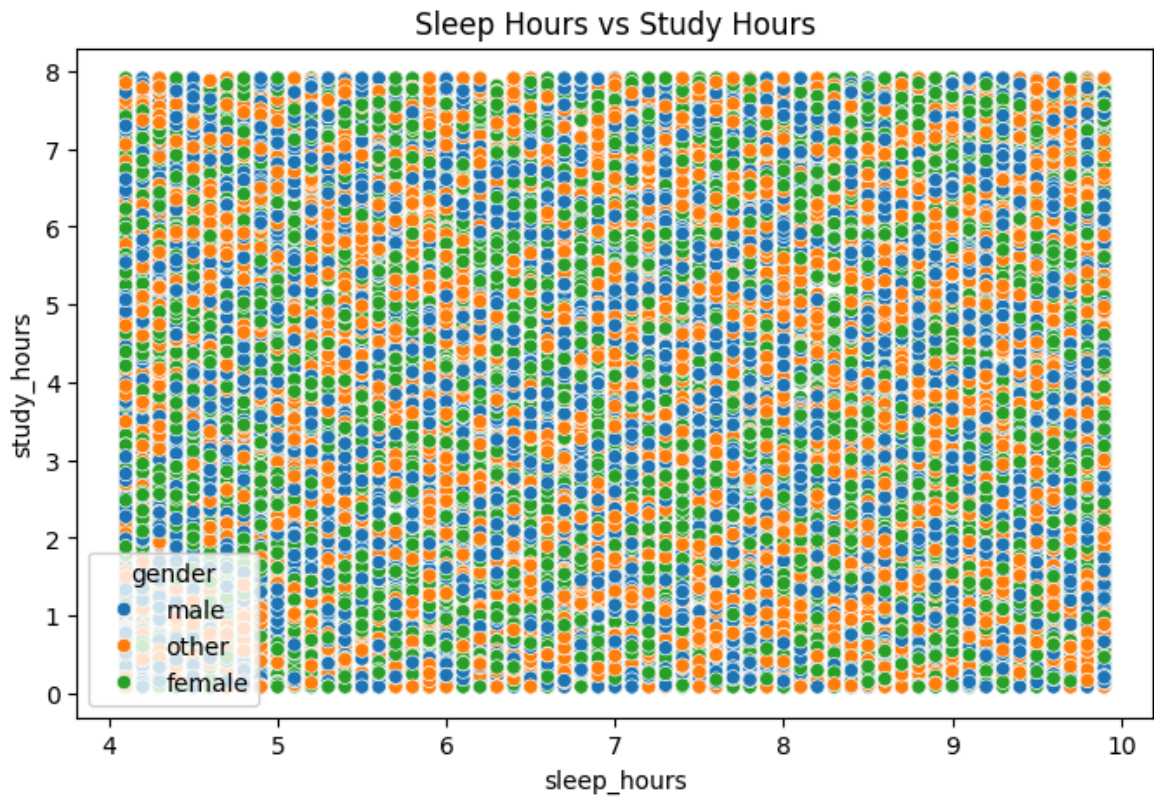
```
In [10]: plt.figure(figsize=(8,5))
sns.scatterplot(x="study_hours", y="exam_score", hue="gender", data=df)
plt.title("Study Hours vs Exam Score")
plt.show()
```



```
In [11]: plt.figure(figsize=(8,5))
sns.scatterplot(x="class_attendance", y="exam_score", hue="gender", data=
plt.title("Class Attendance vs Exam Score")
plt.show()
```

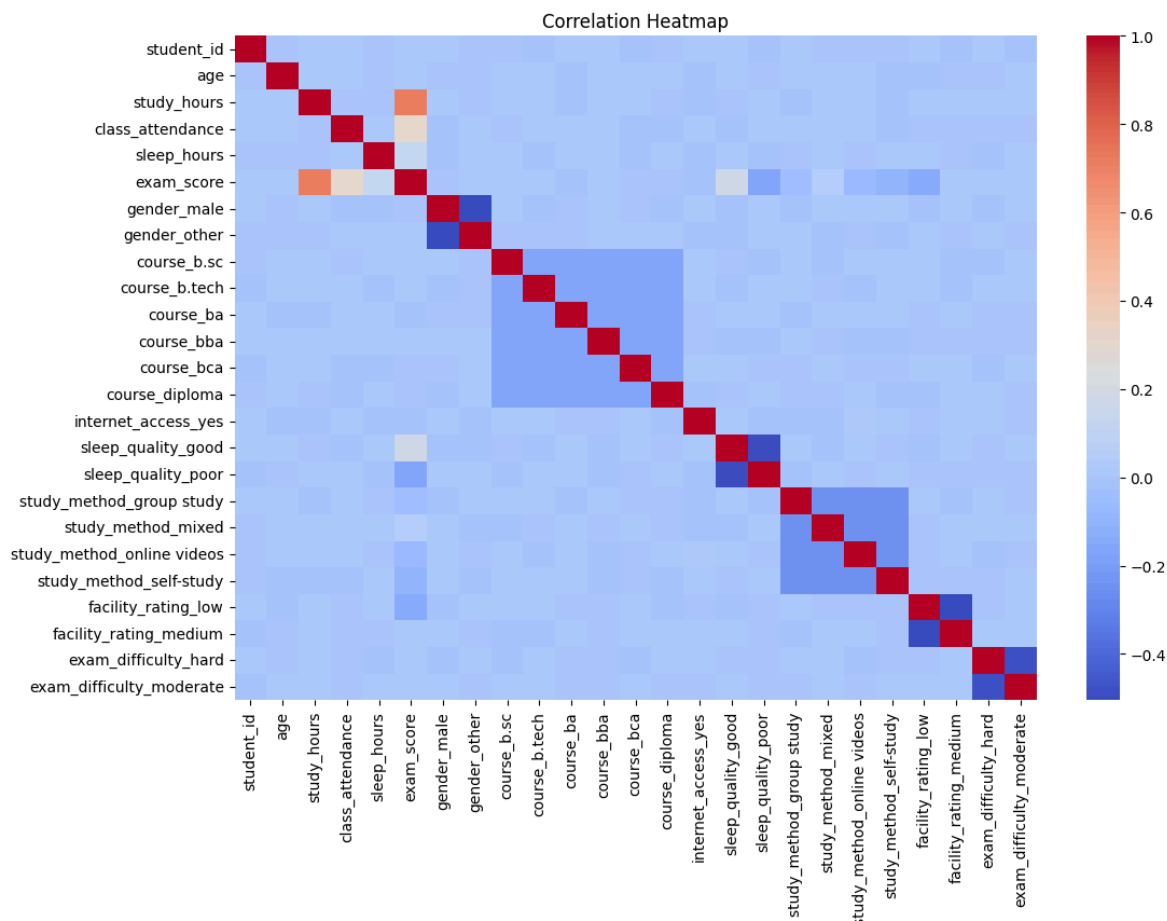


```
In [12]: plt.figure(figsize=(8,5))
sns.scatterplot(x="sleep_hours", y="study_hours", hue="gender", data=df)
plt.title("Sleep Hours vs Study Hours")
plt.show()
```

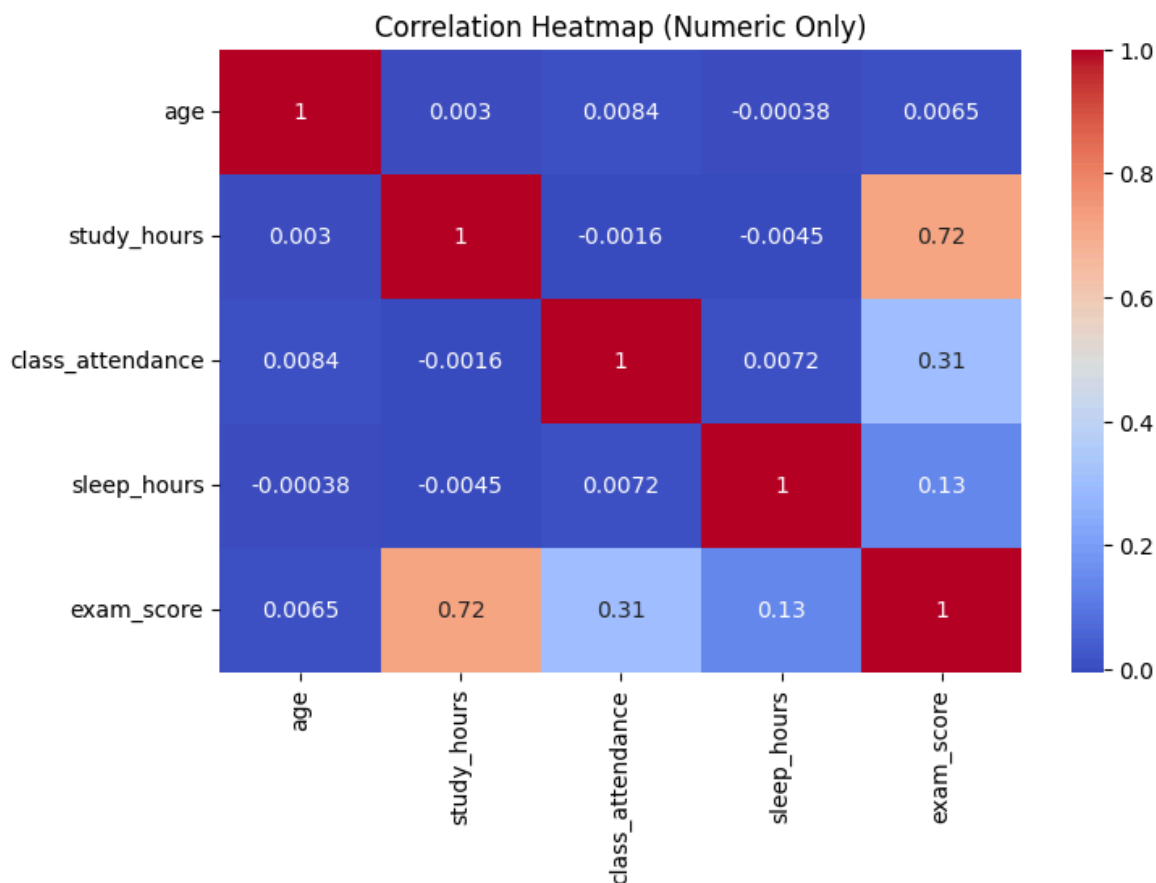



Correlation Heatmap

```
In [13]: plt.figure(figsize=(12,8))
sns.heatmap(df_encoded.corr(), annot=False, cmap='coolwarm')
plt.title("Correlation Heatmap")
plt.show()
```



```
In [14]: plt.figure(figsize=(8,5))
sns.heatmap(df[numeric_cols].corr(), annot=True, cmap="coolwarm")
plt.title("Correlation Heatmap (Numeric Only)")
plt.show()
```



In []: