

✓ CUSTOMER CHURN DATASET:-

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score
```

```
df = pd.read_csv("/content/customer_churn_dataset-training-master[1].csv")
print(df.head(10))
```

| | CustomerID | Age | Gender | Tenure | Usage Frequency | Support Calls | \ |
|---|------------|------|--------|--------|-----------------|---------------|---|
| 0 | 2.0 | 30.0 | Female | 39.0 | 14.0 | 5.0 | |
| 1 | 3.0 | 65.0 | Female | 49.0 | 1.0 | 10.0 | |
| 2 | 4.0 | 55.0 | Female | 14.0 | 4.0 | 6.0 | |
| 3 | 5.0 | 58.0 | Male | 38.0 | 21.0 | 7.0 | |
| 4 | 6.0 | 23.0 | Male | 32.0 | 20.0 | 5.0 | |
| 5 | 8.0 | 51.0 | Male | 33.0 | 25.0 | 9.0 | |
| 6 | 9.0 | 58.0 | Female | 49.0 | 12.0 | 3.0 | |
| 7 | 10.0 | 55.0 | Female | 37.0 | 8.0 | 4.0 | |
| 8 | 11.0 | 39.0 | Male | 12.0 | 5.0 | 7.0 | |
| 9 | 12.0 | 64.0 | Female | 3.0 | 25.0 | 2.0 | |

| | Payment Delay | Subscription Type | Contract Length | Total Spend | \ |
|---|---------------|-------------------|-----------------|-------------|---|
| 0 | 18.0 | Standard | Annual | 932.0 | |
| 1 | 8.0 | Basic | Monthly | 557.0 | |
| 2 | 18.0 | Basic | Quarterly | 185.0 | |
| 3 | 7.0 | Standard | Monthly | 396.0 | |
| 4 | 8.0 | Basic | Monthly | 617.0 | |
| 5 | 26.0 | Premium | Annual | 129.0 | |
| 6 | 16.0 | Standard | Quarterly | 821.0 | |
| 7 | 15.0 | Premium | Annual | 445.0 | |
| 8 | 4.0 | Standard | Quarterly | 969.0 | |
| 9 | 11.0 | Standard | Quarterly | 415.0 | |

| | Last Interaction | Churn |
|---|------------------|-------|
| 0 | 17.0 | 1.0 |
| 1 | 6.0 | 1.0 |
| 2 | 3.0 | 1.0 |
| 3 | 29.0 | 1.0 |
| 4 | 20.0 | 1.0 |
| 5 | 8.0 | 1.0 |
| 6 | 24.0 | 1.0 |
| 7 | 30.0 | 1.0 |
| 8 | 13.0 | 1.0 |
| 9 | 29.0 | 1.0 |

```
print(df.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 440833 entries, 0 to 440832
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   CustomerID            440832 non-null float64
1   Age                   440832 non-null float64
2   Gender                440832 non-null object
3   Tenure                440832 non-null float64
4   Usage Frequency       440832 non-null float64
5   Support Calls         440832 non-null float64
6   Payment Delay         440832 non-null float64
7   Subscription Type     440832 non-null object
8   Contract Length       440832 non-null object
9   Total Spend           440832 non-null float64
10  Last Interaction       440832 non-null float64
11  Churn                 440832 non-null float64
dtypes: float64(9), object(3)
memory usage: 40.4+ MB
None
```

```
print(df.isnull().sum())
```

```
CustomerID      1
Age             1
Gender          1
Tenure         1
Usage Frequency 1
Support Calls   1
Payment Delay   1
Subscription Type 1
Contract Length 1
Total Spend     1
Last Interaction 1
Churn           1
dtype: int64
```

```
print(df.describe())
```

```
count    CustomerID      Age      Tenure  Usage Frequency  \
count  440832.000000  440832.000000  440832.000000  440832.000000
mean    225398.667955    39.373153    31.256336    15.807494
std     129531.918550    12.442369    17.255727     8.586242
min       2.000000     18.000000     1.000000     1.000000
25%     113621.750000    29.000000    16.000000     9.000000
50%     226125.500000    39.000000    32.000000    16.000000
75%     337739.250000    48.000000    46.000000    23.000000
max     449999.000000    65.000000    60.000000    30.000000

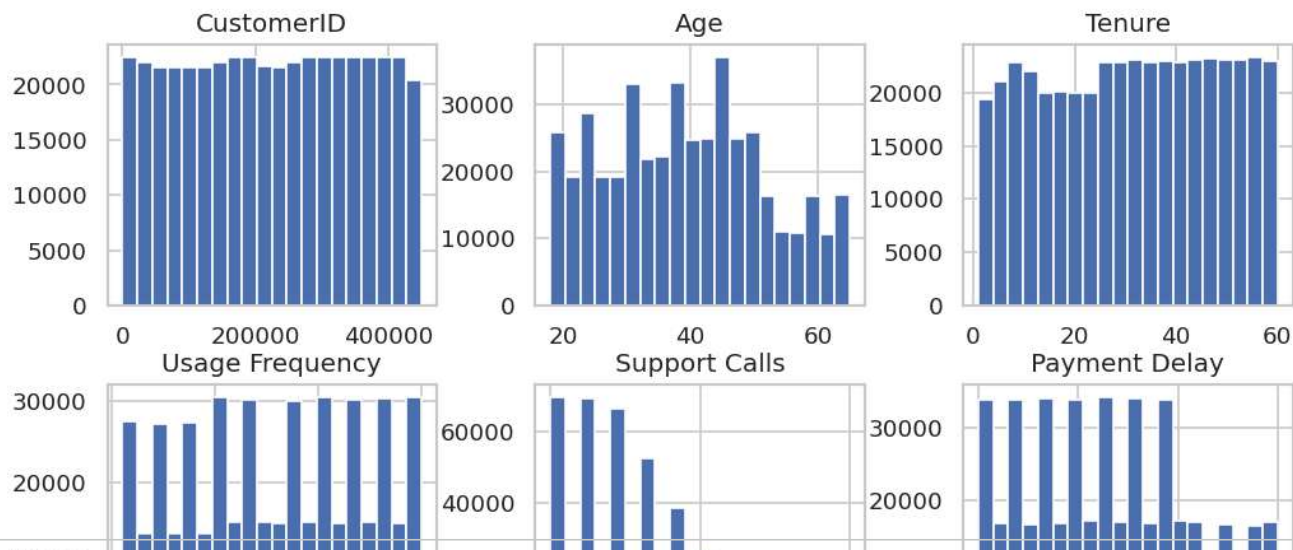
count    Support Calls  Payment Delay  Total Spend  Last Interaction  \
count  440832.000000  440832.000000  440832.000000  440832.000000
mean       3.604437    12.965722    631.616223    14.480868
std       3.070218     8.258063    240.803001     8.596208
min       0.000000     0.000000    100.000000     1.000000
25%       1.000000     6.000000    480.000000     7.000000
50%       3.000000    12.000000    661.000000    14.000000
75%       6.000000    19.000000    830.000000    22.000000
max      10.000000    30.000000   1000.000000    30.000000

count    Churn
count  440832.000000
mean     0.567107
std     0.495477
min     0.000000
25%     0.000000
50%     1.000000
75%     1.000000
max     1.000000
```

EDA:-

```
df.hist(figsize=(10,8), bins=20)
plt.suptitle("Histograms of Numeric Features")
plt.show()
```

Histograms of Numeric Features

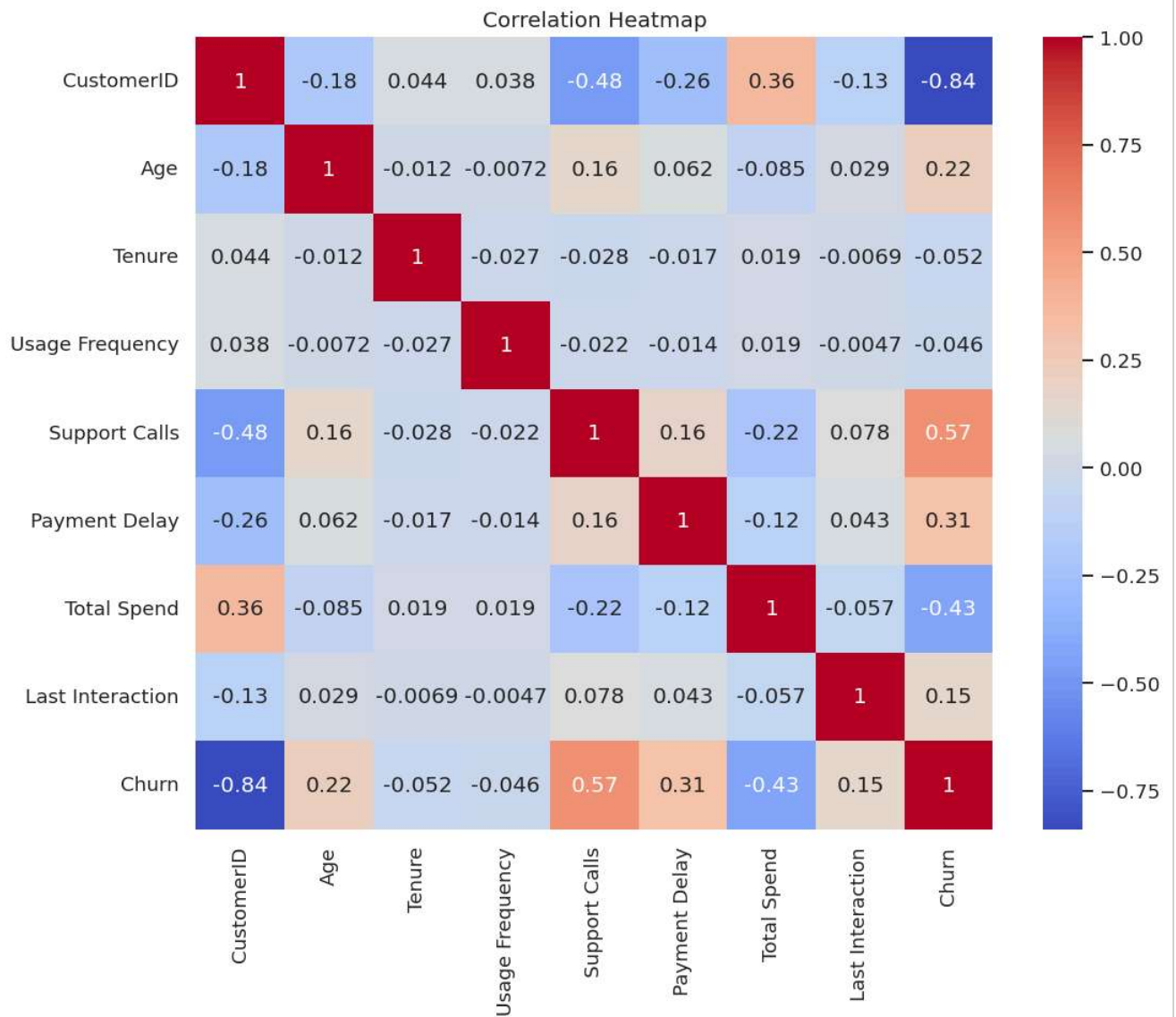


```
plt.figure(figsize=(10,6))
sns.boxplot(data=df.select_dtypes(include=[np.number]))
plt.title("Boxplots of Numeric Columns")
plt.show()
```

Boxplots of Numeric Columns



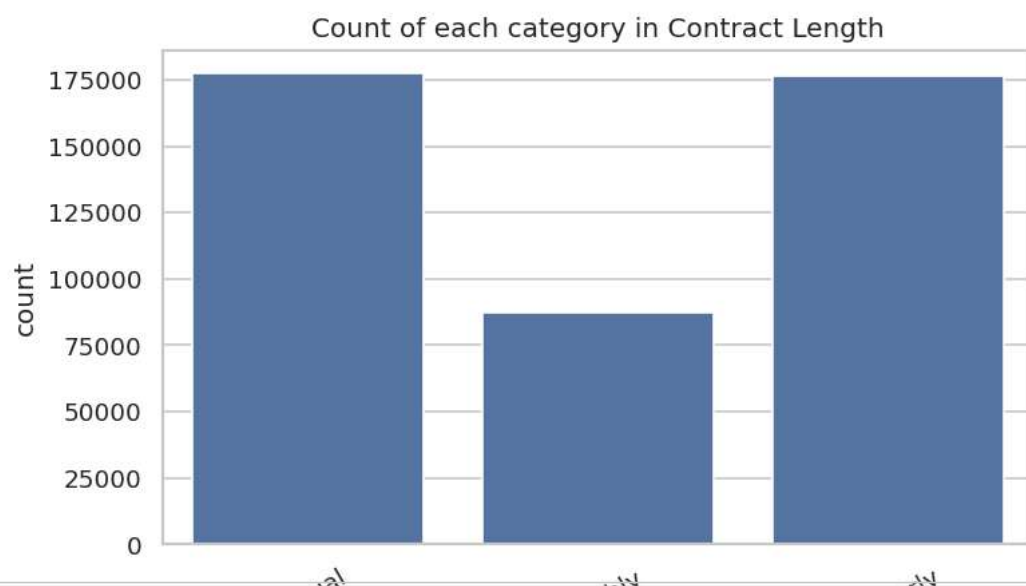
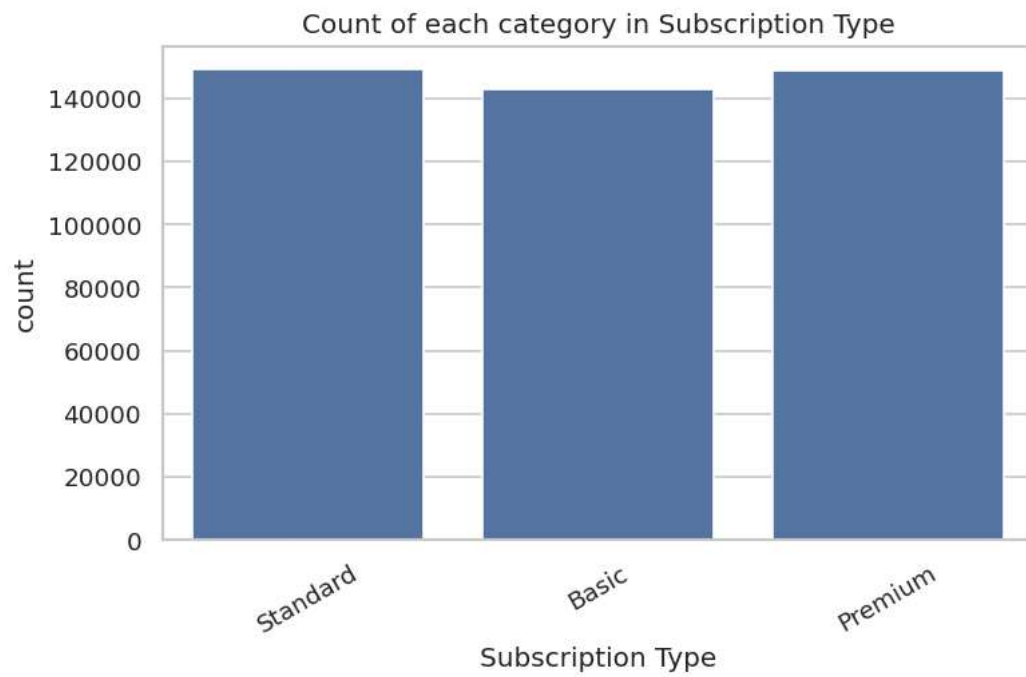
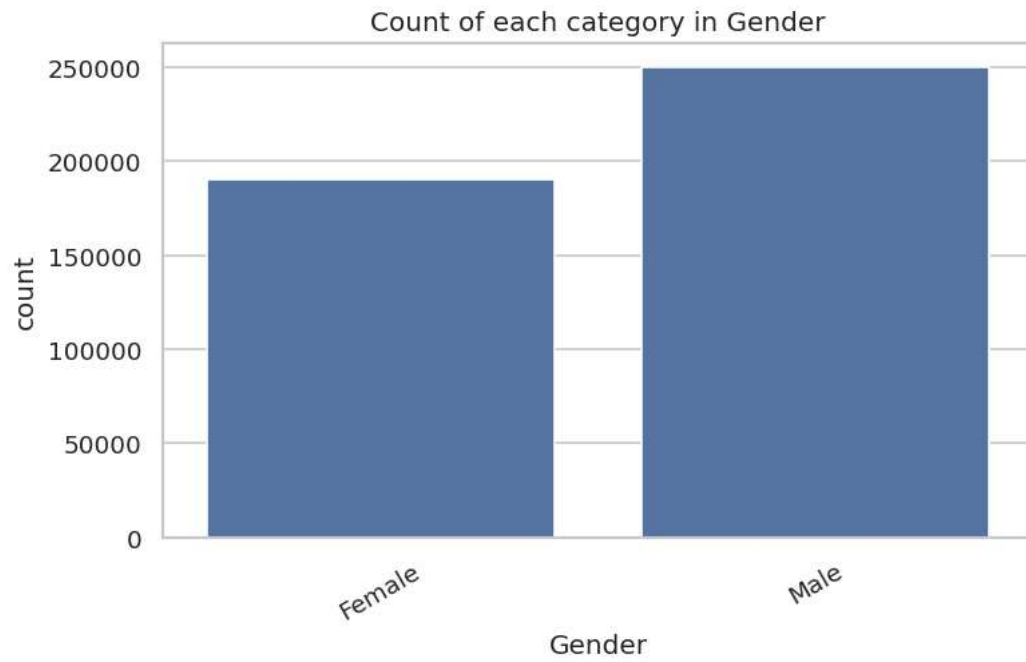
```
plt.figure(figsize=(10,8))
sns.heatmap(df.select_dtypes(include=[np.number]).corr(), annot=True, cmap='coolwarm')
plt.title("Correlation Heatmap")
plt.show()
```



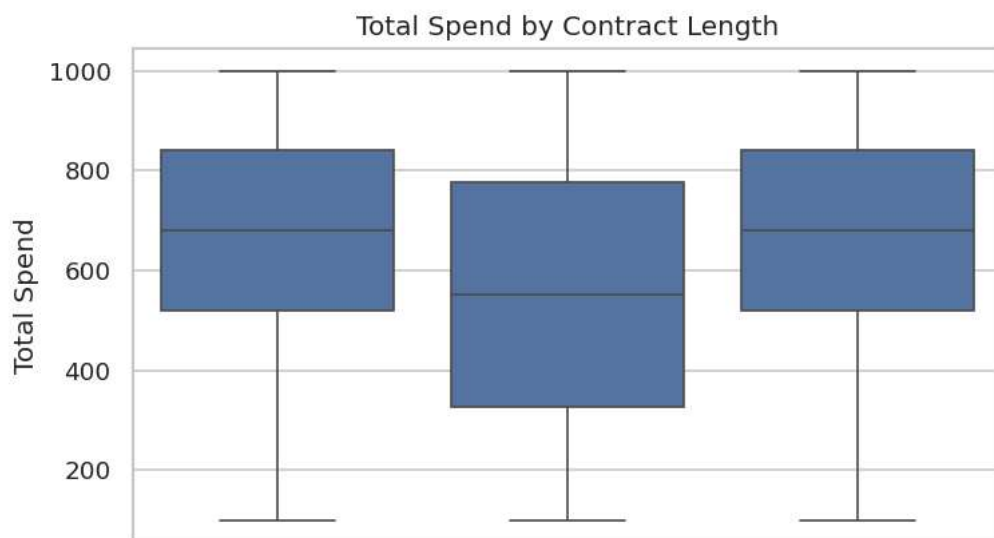
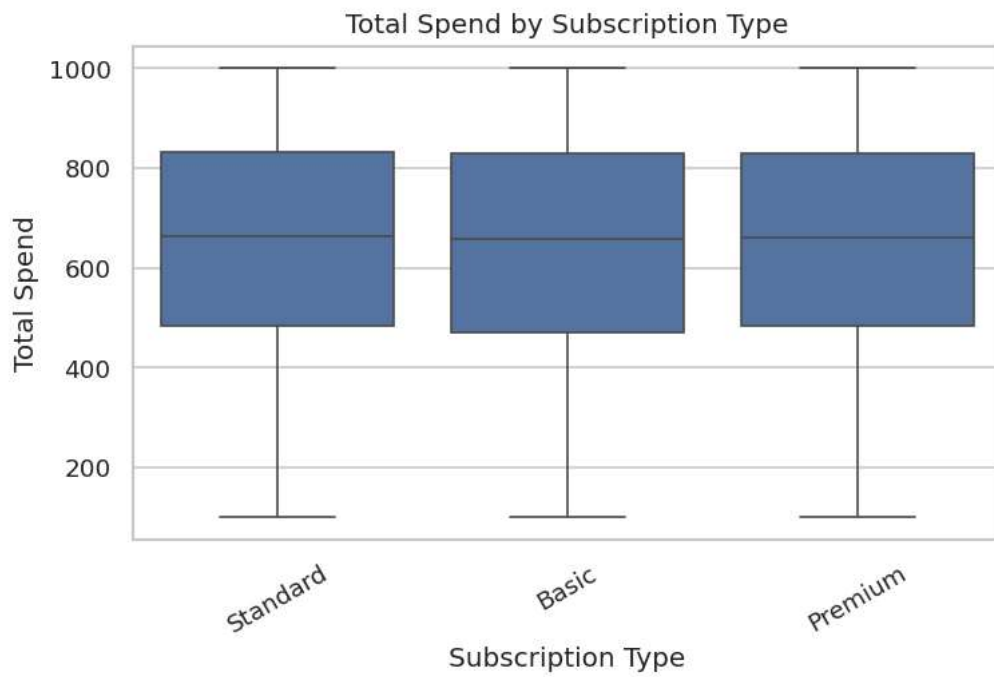
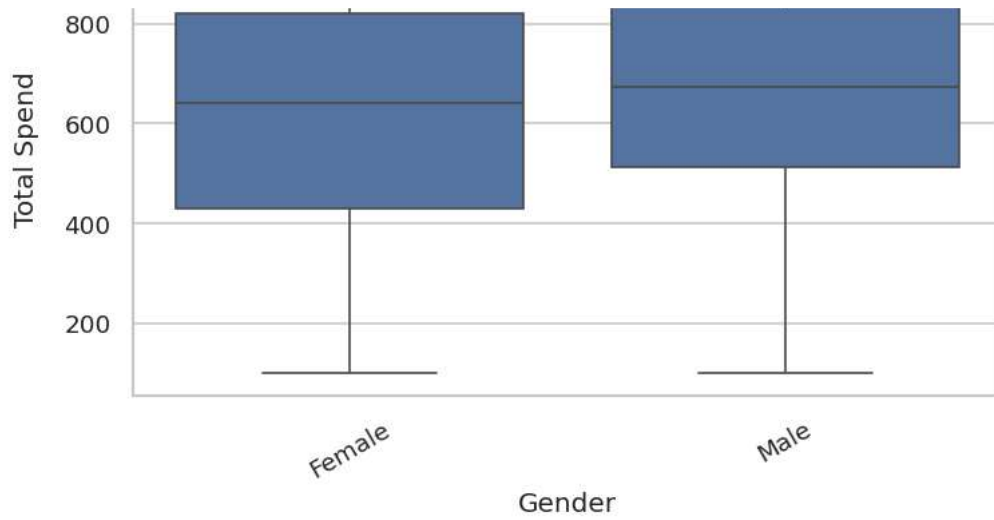
```
cat_cols = df.select_dtypes(include=['object']).columns
```

```
for col in cat_cols:
    plt.figure(figsize=(7,4))
    sns.countplot(data=df, x=col)
    plt.title(f"Count of each category in {col}")
    plt.xticks(rotation=30)
    plt.show()
```

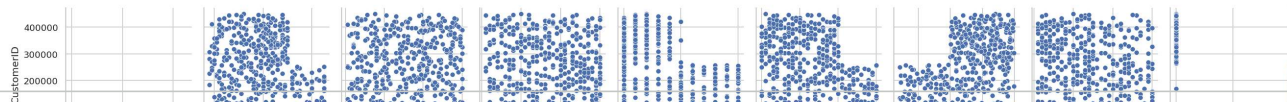
```
for col in cat_cols:
    plt.figure(figsize=(7,4))
    sns.boxplot(data=df, x=col, y='Total Spend')
    plt.title(f"Total Spend by {col}")
    plt.xticks(rotation=30)
    plt.show()
```



```
sns.pairplot(df.select_dtypes(include=[np.number]).sample(500))  
plt.suptitle("Pairplot of Numeric Features", y=1.02)  
plt.show()
```

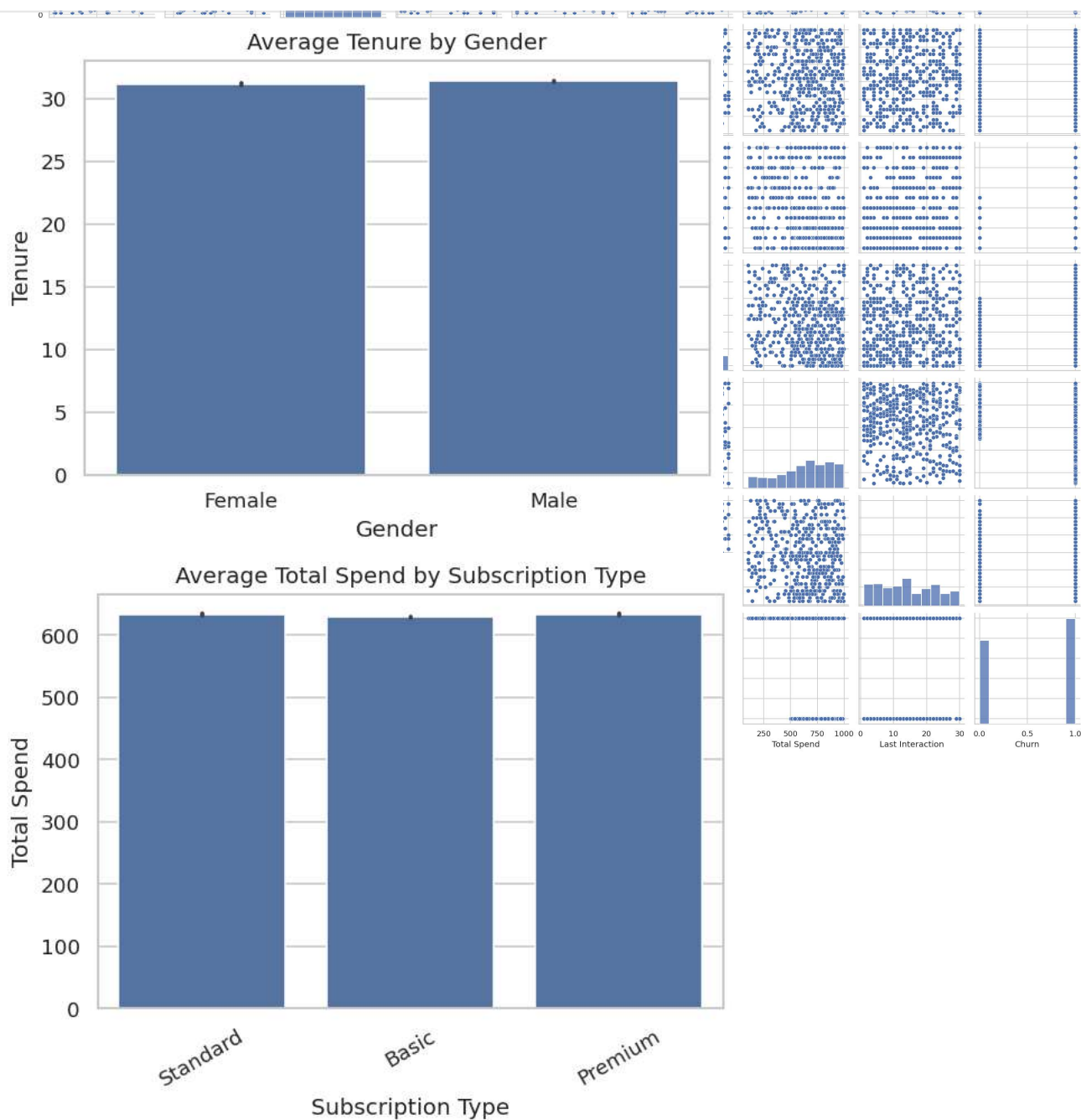


Pairplot of Numeric Features



```
plt.figure(figsize=(6,4))
sns.barplot(x='Gender', y='Tenure', data=df)
plt.title("Average Tenure by Gender")
plt.show()

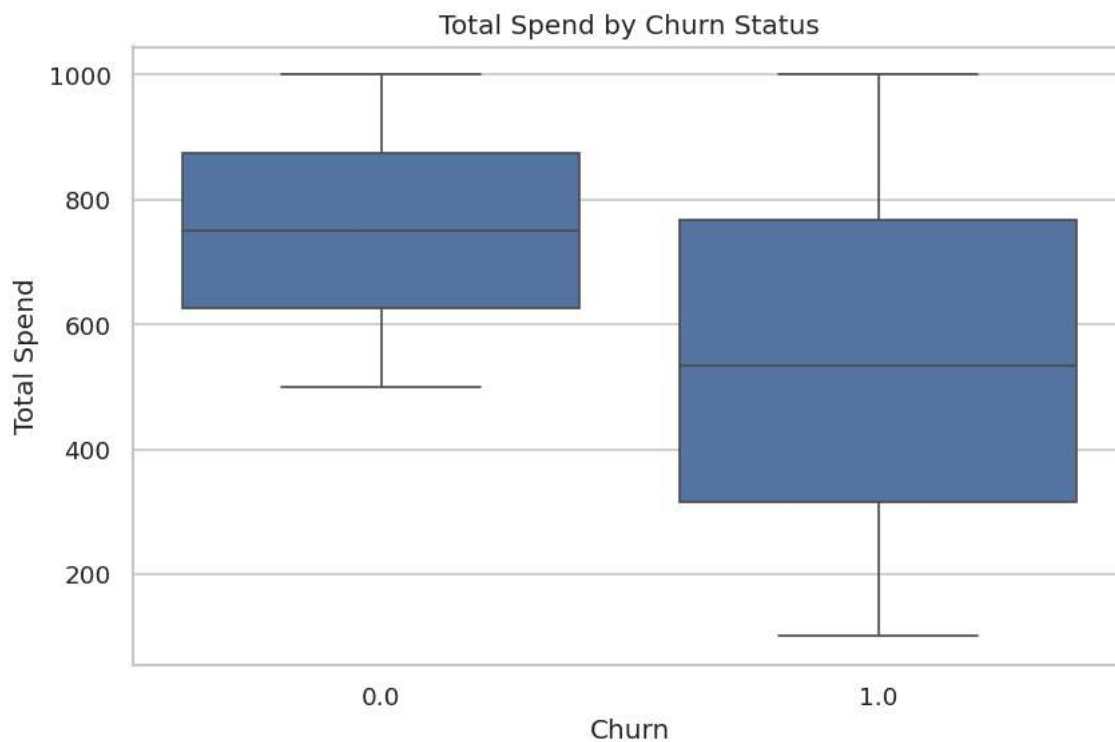
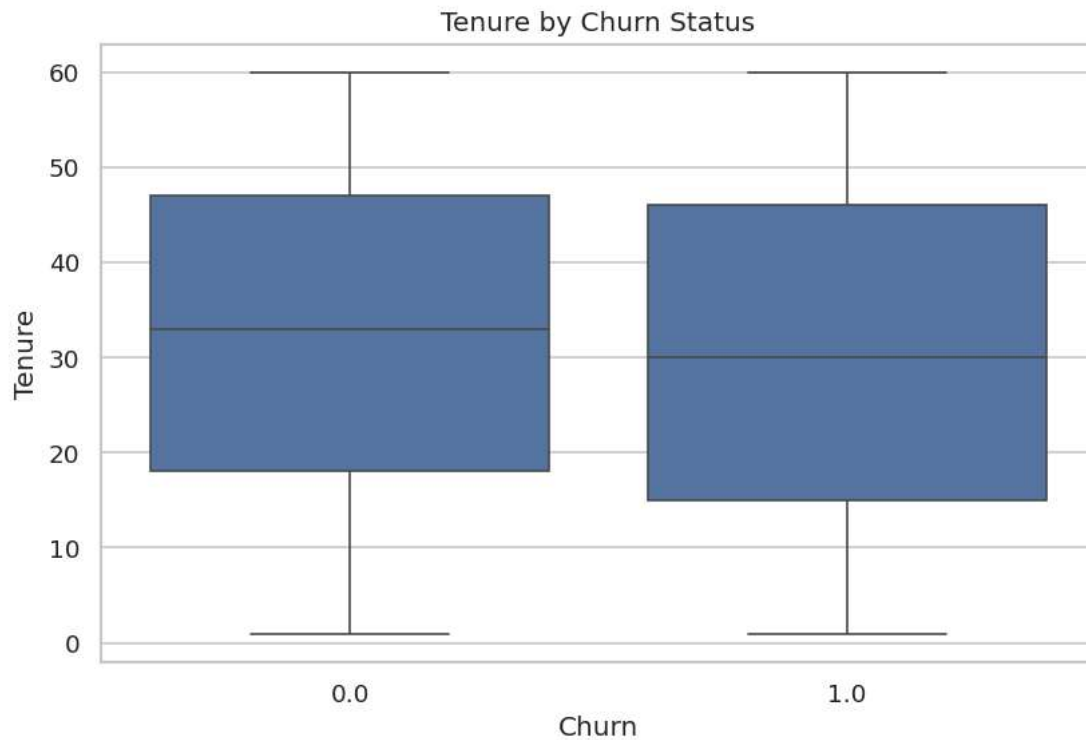
plt.figure(figsize=(6,4))
sns.barplot(x='Subscription Type', y='Total Spend', data=df)
plt.title("Average Total Spend by Subscription Type")
plt.xticks(rotation=30)
plt.show()
```



```
sns.boxplot(x='Churn', y='Tenure', data=df)
plt.title("Tenure by Churn Status")
plt.show()
```



```
sns.boxplot(x='Churn', y='Total Spend', data=df)
plt.title("Total Spend by Churn Status")
plt.show()
```



✓ Prepare Data for Simple Linear Regression:-

```
X = df[['Total Spend']]
y = df['Tenure']
```

```
print("Intercept:", model.intercept_)  
print("Coefficient:", model.coef_)
```

```
Intercept: 30.404650371404227  
Coefficient: [0.00136611]
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
```

```
model = LinearRegression()  
model.fit(X_train, y_train)
```

```
▼ LinearRegression ⓘ ?  
LinearRegression()
```

```
y_pred = model.predict(X_test)
```

```
mse = mean_squared_error(y_test, y_pred)  
mae = mean_absolute_error(y_test, y_pred)  
r2 = r2_score(y_test, y_pred)  
  
print("Model Evaluation Results:")  
print("Mean Squared Error:", round(mse, 2))  
print("Mean Absolute Error:", round(mae, 2))  
print("R2 Score:", round(r2, 4))
```

```
Model Evaluation Results:  
Mean Squared Error: 298.62  
Mean Absolute Error: 14.93  
R2 Score: 0.0003
```

```
plt.scatter(X_test, y_test, color='blue', alpha=0.5, label='Actual')  
plt.plot(X_test, y_pred, color='red', label='Predicted Line')  
plt.xlabel('Total Spend')  
plt.ylabel('Tenure')  
plt.title('Simple Linear Regression')  
plt.legend()  
plt.show()
```