

```
In [18]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

file_name = "electricity_consumption_sa.csv"

try:
    df = pd.read_csv(file_name, sep=';')
    print("✅ File loaded successfully.")
except FileNotFoundError:
    print(f"❌ Error: File '{file_name}' not found. Check the file name and path.")
    raise

print("\n--- First 5 Rows ---")
print(df.head())
print("\n--- Column Info and Data Types ---")
df.info()
```

✅ File loaded successfully.

--- First 5 Rows ---

	Year	Region	Sector	Electricity Consumption (MWh)	\
0	2005	Eastern	Industrial	28089814.0	
1	2005	Western	Others	2016908.0	
2	2005	Total	Total	153283561.0	
3	2006	Southern	Residential	9247978.0	
4	2006	Central	Others	2431666.0	

	Name	Date_Object	Periodicity
0	Industrial : Eastern	2005-01-01	Annually
1	Others : Western	2005-01-01	Annually
2	Total : Total	2005-01-01	Annually
3	Residential : Southern	2006-01-01	Annually
4	Others : Central	2006-01-01	Annually

--- Column Info and Data Types ---

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

RangeIndex: 630 entries, 0 to 629

Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	Year	630 non-null	int64
1	Region	630 non-null	object
2	Sector	630 non-null	object
3	Electricity Consumption (MWh)	630 non-null	float64
4	Name	630 non-null	object
5	Date_Object	630 non-null	object
6	Periodicity	630 non-null	object

dtypes: float64(1), int64(1), object(5)

memory usage: 34.6+ KB

```
In [23]: # تحديد الأعمدة التي سنعمل عليها (هذه الخطوة تمت بالفعل في الخلية 2 السابقة)
DATE_COLUMN = 'Year'
REGION_COLUMN = 'Region'
CONSUMPTION_COLUMN = 'Electricity Consumption (MWh)'

# 1. (هي التي تمثل الرياض إقليمياً 'Central') فلترة البيانات على المنطقة الوسطى
RIYADH_NAME = 'Central'

# لضمان عدم حدوث تحذيرات df من (copy) نستخدم نسخة
```

```
df_riyadh = df[df[REGION_COLUMN] == RIYADH_NAME].copy()
```

الطباعة للتحقق من النتيجة:

```
print(f"✅ تم فلتر البيانات بنجاح على منطقة {RIYADH_NAME}.")
print(f"عدد السجلات التي تخص الرياض: {len(df_riyadh)}")
print("\nأول 5 صفوف لبيانات الرياض (بجميع القطاعات)")
print(df_riyadh.head())
```

✅ Central. تم فلتر البيانات بنجاح على منطقة

عدد السجلات التي تخص الرياض: 126

أول 5 صفوف لبيانات الرياض (بجميع القطاعات):

	Year	Region	Sector	Electricity Consumption (MWh) \
4	2006	Central	Others	2431666.0
9	2007	Central	Industrial	3748438.0
11	2007	Central	Others	2604606.0
14	2008	Central	Agricultural	2355609.0
16	2008	Central	Total	55983973.0

	Name	Date_Object	Periodicity
4	Others : Central	2006-01-01	Annually
9	Industrial : Central	2007-01-01	Annually
11	Others : Central	2007-01-01	Annually
14	Agricultural : Central	2008-01-01	Annually
16	Total : Central	2008-01-01	Annually

In [35]: # Year الذي يحتوي على كل القطاعات في الرياض والمؤشر هو df_riyadh الآن، إطار البيانات الحالي هو

3. حساب الاستهلاك الكلي السنوي لكل سنة عن طريق جمع كل القطاعات (Aggregation): التجميع

```
CONSUMPTION_COLUMN = 'Electricity Consumption (MWh)'
```

ثم جمع عمود الاستهلاك (Index) نقوم بتجميع البيانات بناءً على مؤشر السنة

```
df_clean = df_riyadh.groupby(level=0)[CONSUMPTION_COLUMN].sum().to_frame()
```

4. التأكد من النظافة

```
df_clean = df_clean.ffill().dropna()
```

الطباعة للتحقق من النتيجة:

```
print(f"✅ Total annual consumption is collected. Final number of records: {len(df_clean)}")
print("\n data clean:")
```

```
print(df_clean.head(10)) # نطبع 10 صفوف لنرى التتابع الزمني
# للتأكد من انتظام السنوات (Index) طباعة المؤشر
print("\n Available years (indicator):")
print(df_clean.index.year)
```

✓ Total annual consumption is collected. Final number of records: 18

data clean:

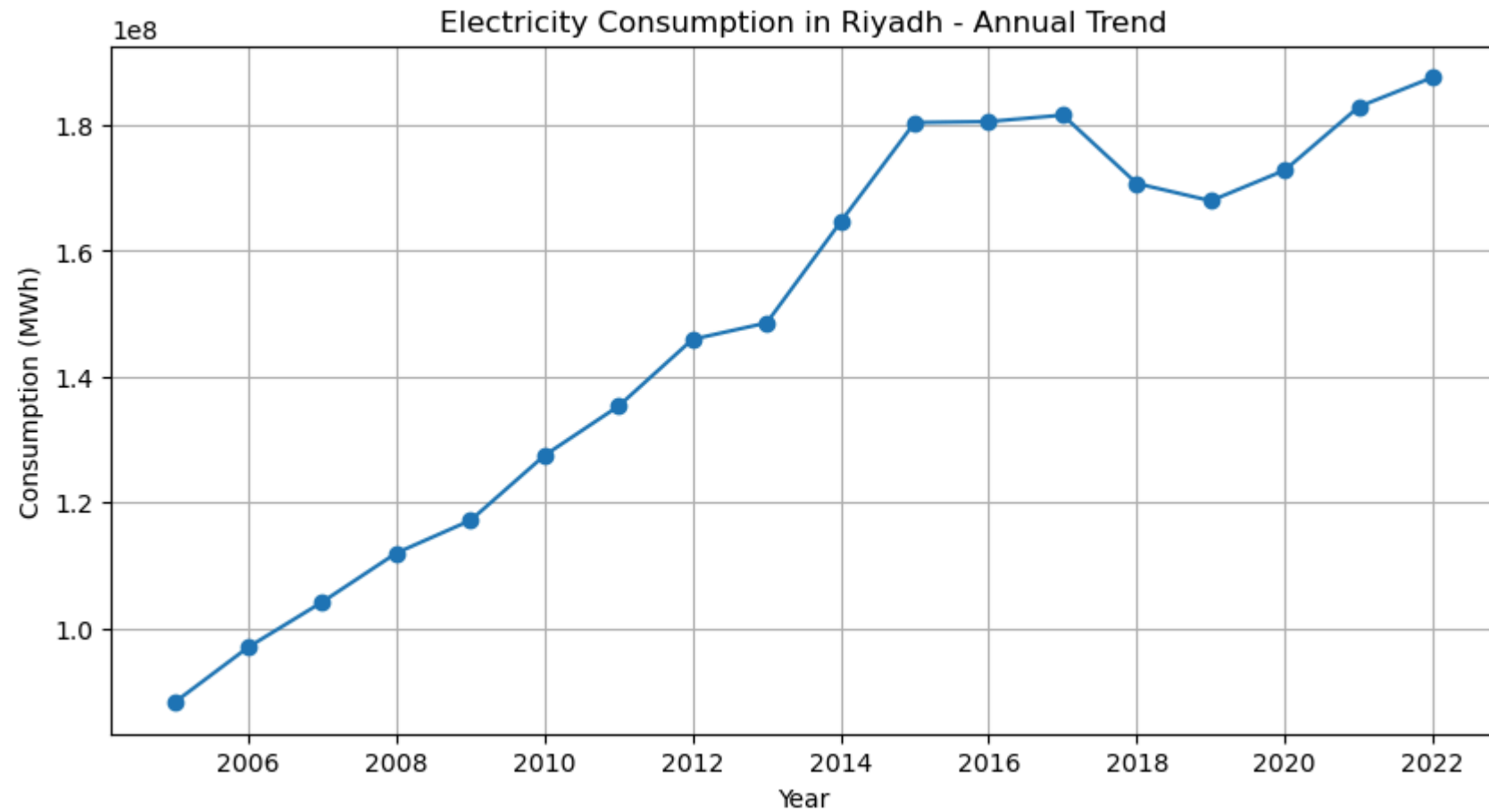
Electricity Consumption (MWh)

Year	
2005-01-01	8.819913e+07
2006-01-01	9.700675e+07
2007-01-01	1.041922e+08
2008-01-01	1.119679e+08
2009-01-01	1.171662e+08
2010-01-01	1.274917e+08
2011-01-01	1.353263e+08
2012-01-01	1.459458e+08
2013-01-01	1.485649e+08
2014-01-01	1.646362e+08

Available years (indicator):

```
Index([2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016,
       2017, 2018, 2019, 2020, 2021, 2022],
      dtype='int32', name='Year')
```

```
In [36]: plt.figure(figsize=(10,5))
plt.plot(df_clean.index, df_clean[CONSUMPTION_COLUMN], marker='o')
plt.title("Electricity Consumption in Riyadh - Annual Trend")
plt.xlabel("Year")
plt.ylabel("Consumption (MWh)")
plt.grid(True)
plt.show()
```



```
In [59]: print(df_clean.head())  
         print(df_clean.tail())  
         print(df_clean.index)
```

Electricity Consumption (MWh)

Year

2005-01-31	88199134.0
2006-01-31	97006748.0
2007-01-31	104192164.0
2008-01-31	111967946.0
2009-01-31	117166198.0

Electricity Consumption (MWh)

Year

2018-01-31	1.707173e+08
2019-01-31	1.679766e+08
2020-01-31	1.728444e+08
2021-01-31	1.829464e+08
2022-01-31	1.876239e+08

```
DatetimeIndex(['2005-01-31', '2006-01-31', '2007-01-31', '2008-01-31',
               '2009-01-31', '2010-01-31', '2011-01-31', '2012-01-31',
               '2013-01-31', '2014-01-31', '2015-01-31', '2016-01-31',
               '2017-01-31', '2018-01-31', '2019-01-31', '2020-01-31',
               '2021-01-31', '2022-01-31'],
              dtype='datetime64[ns]', name='Year', freq='YE-JAN')
```

```
In [60]: df_clean.index = df_clean.index.year
df_clean.index.name = "Year"
print(df_clean)
print(df_clean.index)
```

```

Electricity Consumption (MWh)
Year
2005      8.819913e+07
2006      9.700675e+07
2007      1.041922e+08
2008      1.119679e+08
2009      1.171662e+08
2010      1.274917e+08
2011      1.353263e+08
2012      1.459458e+08
2013      1.485649e+08
2014      1.646362e+08
2015      1.804233e+08
2016      1.805722e+08
2017      1.815990e+08
2018      1.707173e+08
2019      1.679766e+08
2020      1.728444e+08
2021      1.829464e+08
2022      1.876239e+08
Index([2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016,
       2017, 2018, 2019, 2020, 2021, 2022],
      dtype='int32', name='Year')

```

In [81]: `import pandas as pd`

```

# قراءة الملفات
weather = pd.read_csv("riyadh_weather_annual.csv")
energy = pd.read_csv("electricity_consumption_sa.csv")

# اختيار فقط المنطقة الوسطى (تقريبًا تمثل الرياض)
energy_central = energy[energy["Region"] == "Central"]

# اختيار الأعمدة المهمة
energy_central = energy_central[["Year", "Electricity Consumption (MWh)"]]

# دمج الطقس مع استهلاك الكهرباء
merged = pd.merge(weather, energy_central, on="Year")

# حفظ الملف الناتج
merged.to_csv("riyadh_weather_energy_merged.csv", index=False)

```

```
print("✅ تم الدمج بنجاح! عدد السنوات: ", len(merged))
print(merged.head())
```

```
✅ 126 تم الدمج بنجاح! عدد السنوات:
  Year  Temperature (?C)  Humidity (%)  Rainfall (mm) \
0  2005                26.2           44           100.7
1  2005                26.2           44           100.7
2  2005                26.2           44           100.7
3  2005                26.2           44           100.7
4  2005                26.2           44           100.7

  Electricity Consumption (MWh)
0                2186793.0
1                25109641.0
2                6361952.0
3                5095729.0
4                2326012.0
```

In [82]: `import pandas as pd`

```
# قراءة الملفات
weather = pd.read_csv("riyadh_weather_annual.csv")
energy = pd.read_csv("electricity_consumption_sa.csv")

# نختار المنطقة الوسطى فقط
energy_central = energy[energy["Region"] == "Central"]

# نجمع استهلاك الكهرباء لكل سنة (بغض النظر عن القطاع)
energy_central_sum = energy_central.groupby("Year", as_index=False)["Electricity Consumption (MWh)"].sum()

# ندمج مع بيانات الطقس
merged = pd.merge(weather, energy_central_sum, on="Year")

# نحفظ الملف الناتج
merged.to_csv("riyadh_weather_energy_merged.csv", index=False)

print("✅ تم الدمج بنجاح بعد التجميع! عدد السنوات: ", len(merged))
print(merged.head())
```


✓ 18 تم الدمج بنجاح بعد التجميع! عدد السنوات: 18

	Year	Temperature (°C)	Humidity (%)	Rainfall (mm)	\
0	2005	26.2	44	100.7	
1	2006	26.5	43	110.2	
2	2007	26.8	42	120.3	
3	2008	27.0	41	130.4	
4	2009	27.2	40	140.5	

	Electricity Consumption (MWh)
0	88199134.0
1	97006748.0
2	104192164.0
3	111967946.0
4	117166198.0

In [83]: `import pandas as pd`

`# تحميل الملف المدمج`

`df = pd.read_csv("riyadh_weather_energy_merged.csv")`

`# عرض أول صفين للتأكد`

`print("عرض أول صفين من البيانات")`

`print(df.head(2))`

`print("-" * 60)`

`# معلومات عامة`

`print("معلومات عامة:")`

`print(df.info())`

`print("-" * 60)`

`# إحصائيات وصفية`

`print("إحصائيات وصفية:")`

`print(df.describe())`

`print("-" * 60)`

`# أعلى وأقل سنة في استهلاك الكهرباء`

`max_year = df.loc[df["Electricity Consumption (MWh)"].idxmax(), "Year"]`

`min_year = df.loc[df["Electricity Consumption (MWh)"].idxmin(), "Year"]`

`print(f"▲ أعلى استهلاك كهرباء في سنة: {max_year}")`

`print(f"▼ أقل استهلاك كهرباء في سنة: {min_year}")`

```
# المتوسطات العامة
mean_temp = df["Temperature (?C)"].mean()
mean_humidity = df["Humidity (%)"].mean()
mean_rain = df["Rainfall (mm)"].mean()
mean_energy = df["Electricity Consumption (MWh)"].mean()

print("\n🇸🇦 المتوسطات العامة:")
print(f"درجة الحرارة: {mean_temp:.2f}°C")
print(f"الرطوبة: {mean_humidity:.2f}%")
print(f"الأمطار: {mean_rain:.2f} مم")
print(f"ميجا واط ساعة :{mean_energy:,.0f} استهلاك الكهرباء")
```

عرض أول صفين من البيانات:

	Year	Temperature (?C)	Humidity (%)	Rainfall (mm)	\
0	2005	26.2	44	100.7	
1	2006	26.5	43	110.2	

	Electricity Consumption (MWh)
0	88199134.0
1	97006748.0

معلومات عامة:

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

RangeIndex: 18 entries, 0 to 17

Data columns (total 5 columns):

#	Column	Non-Null Count	Dtype
0	Year	18 non-null	int64
1	Temperature (?C)	18 non-null	float64
2	Humidity (%)	18 non-null	int64
3	Rainfall (mm)	18 non-null	float64
4	Electricity Consumption (MWh)	18 non-null	float64

dtypes: float64(3), int64(2)

memory usage: 852.0 bytes

None

إحصائيات وصفية:

	Year	Temperature (?C)	Humidity (%)	Rainfall (mm)	\
count	18.000000	18.000000	18.000000	18.000000	
mean	2013.500000	28.377778	35.500000	185.483333	
std	5.338539	1.339252	5.338539	53.417427	
min	2005.000000	26.200000	27.000000	100.700000	
25%	2009.250000	27.275000	31.250000	143.025000	
50%	2013.500000	28.350000	35.500000	185.450000	
75%	2017.750000	29.450000	39.750000	227.875000	
max	2022.000000	30.500000	44.000000	270.800000	

	Electricity Consumption (MWh)
count	1.800000e+01
mean	1.480667e+08
std	3.322990e+07
min	8.819913e+07
25%	1.197476e+08

50%	1.566005e+08
75%	1.785286e+08
max	1.876239e+08

-
- ▲ أعلى استهلاك كهرباء في سنة: 2022
 - ▼ أقل استهلاك كهرباء في سنة: 2005

المتوسطات العامة:

28.38 درجة الحرارة: °C

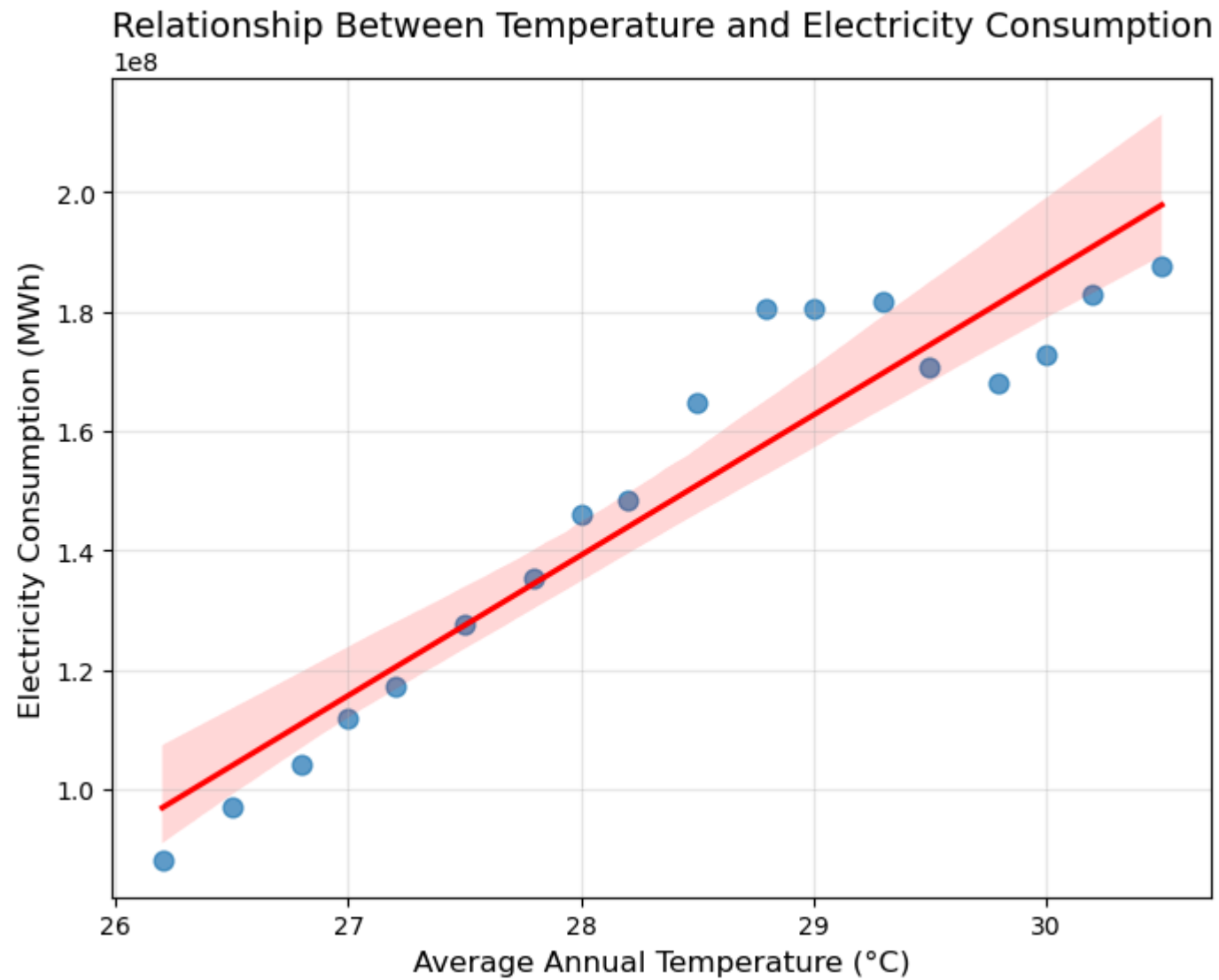
35.50 الرطوبة: %

الأمطار: 185.48 مم

استهلاك الكهرباء: 148,066,670 ميغا واط ساعة

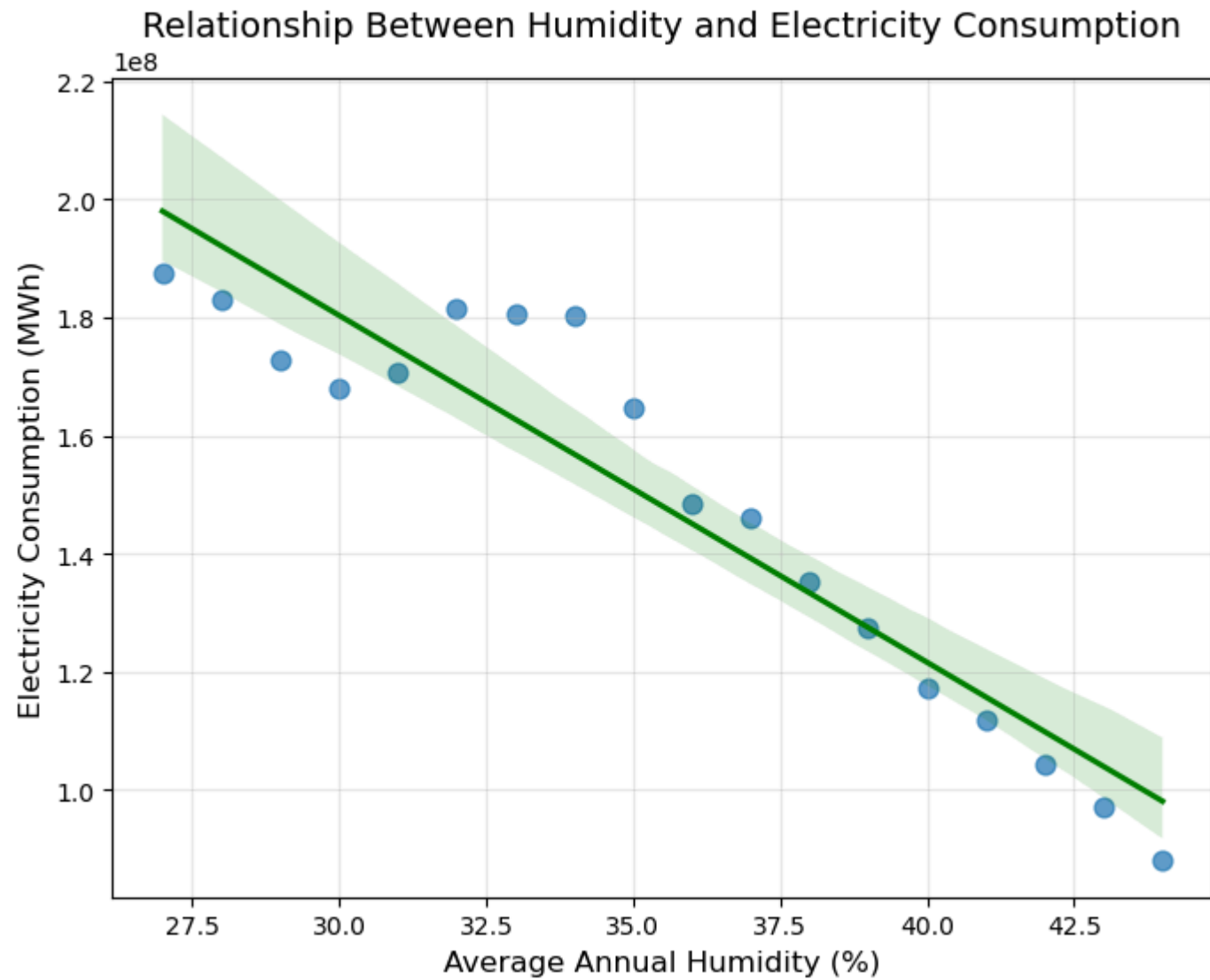
```
In [88]: import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(8,6))
sns.regplot(
    data=merged,
    x="Temperature (?C)",
    y="Electricity Consumption (MWh)",
    scatter_kws={"s": 60, "alpha": 0.7},
    line_kws={"color": "red"}
)
plt.title("Relationship Between Temperature and Electricity Consumption", fontsize=14)
plt.xlabel("Average Annual Temperature (°C)", fontsize=12)
plt.ylabel("Electricity Consumption (MWh)", fontsize=12)
plt.grid(True, alpha=0.3)
plt.show()
```



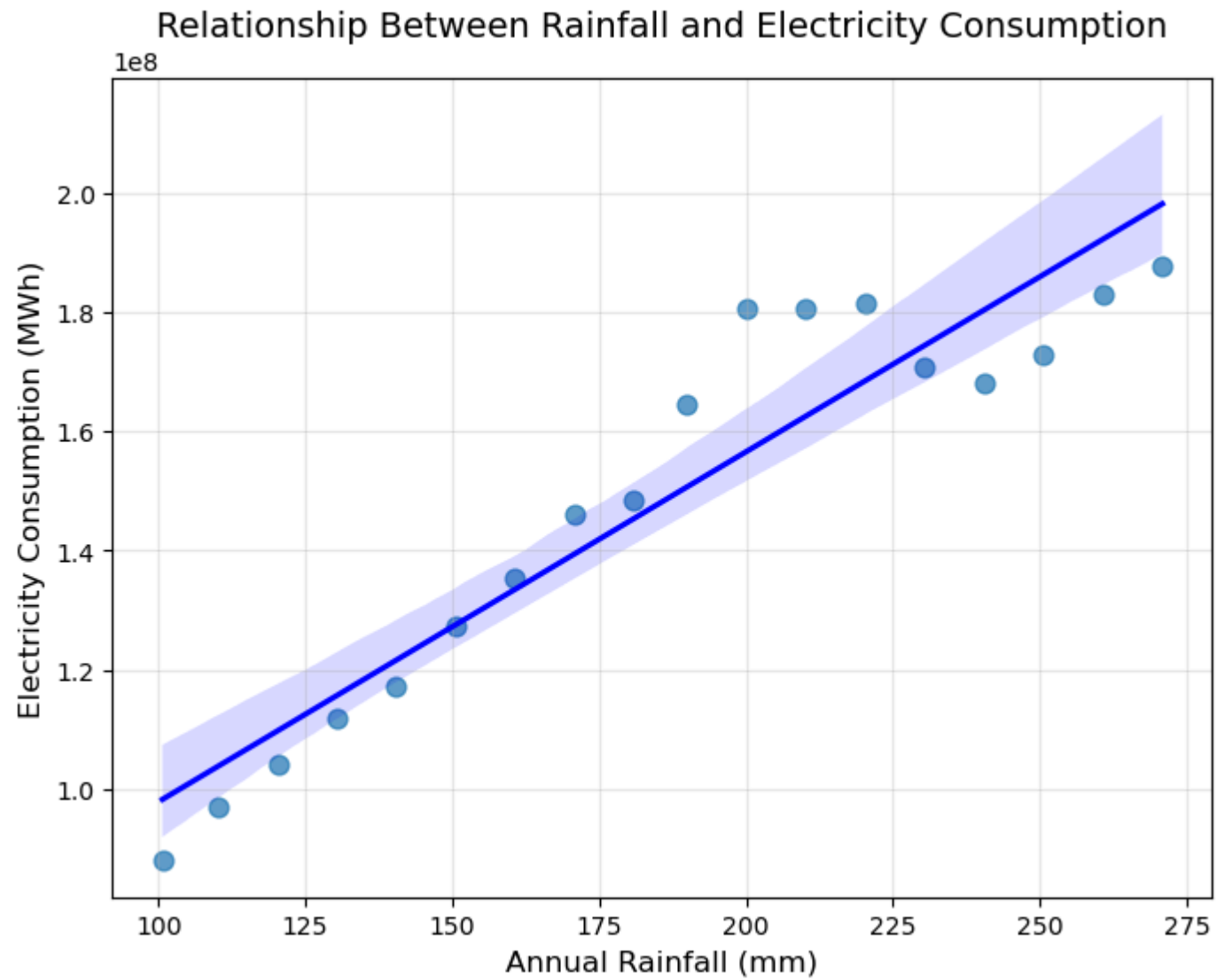
```
In [89]: plt.figure(figsize=(8,6))
sns.regplot(
    data=merged,
    x="Humidity (%)",
    y="Electricity Consumption (MWh)",
    scatter_kws={"s": 60, "alpha": 0.7},
```

```
        line_kws={"color": "green"}
    )
plt.title("Relationship Between Humidity and Electricity Consumption", fontsize=14)
plt.xlabel("Average Annual Humidity (%)", fontsize=12)
plt.ylabel("Electricity Consumption (MWh)", fontsize=12)
plt.grid(True, alpha=0.3)
plt.show()
```



```
In [90]: plt.figure(figsize=(8,6))
sns.regplot(
    data=merged,
    x="Rainfall (mm)",
    y="Electricity Consumption (MWh)",
    scatter_kws={"s": 60, "alpha": 0.7},
```

```
        line_kws={"color": "blue"}
    )
plt.title("Relationship Between Rainfall and Electricity Consumption", fontsize=14)
plt.xlabel("Annual Rainfall (mm)", fontsize=12)
plt.ylabel("Electricity Consumption (MWh)", fontsize=12)
plt.grid(True, alpha=0.3)
plt.show()
```

```
In [86]: import pandas as pd

# Load the merged dataset
df = pd.read_csv("riyadh_weather_energy_merged.csv")

# Select relevant columns
```

```
corr_columns = ["Temperature (?C)", "Humidity (%)", "Rainfall (mm)", "Electricity Consumption (MWh)"]

# Compute correlation
correlation = df[corr_columns].corr()

print("📊 Correlation Matrix:")
print(correlation)
```

📊 Correlation Matrix:

	Temperature (?C)	Humidity (%)	Rainfall (mm)	\
Temperature (?C)	1.000000	-0.999638	0.999567	
Humidity (%)	-0.999638	1.000000	-0.999988	
Rainfall (mm)	0.999567	-0.999988	1.000000	
Electricity Consumption (MWh)	0.946423	-0.944088	0.943259	

	Electricity Consumption (MWh)
Temperature (?C)	0.946423
Humidity (%)	-0.944088
Rainfall (mm)	0.943259
Electricity Consumption (MWh)	1.000000

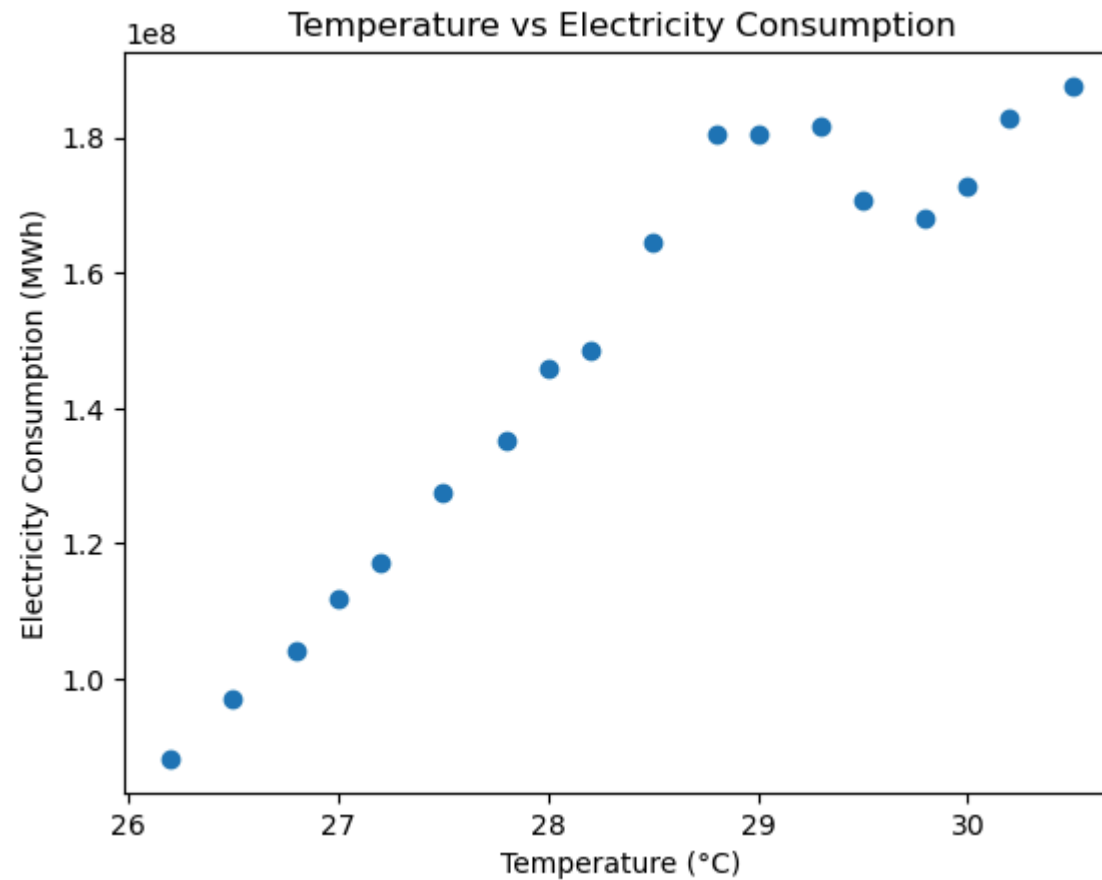
```
In [96]: import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("riyadh_weather_energy_merged.csv")
annual_summary = df.groupby('Year').mean()
print(annual_summary[['Temperature (?C)', 'Electricity Consumption (MWh)']])

# أعلى وأدنى سنة استهلاك
max_year = df.loc[df['Electricity Consumption (MWh)'].idxmax(), 'Year']
min_year = df.loc[df['Electricity Consumption (MWh)'].idxmin(), 'Year']
print(f"أعلى سنة استهلاك: {max_year}, أقل سنة: {min_year}")

# Scatter plot حرارة مقابل استهلاك
plt.scatter(df['Temperature (?C)'], df['Electricity Consumption (MWh)'])
plt.xlabel('Temperature (°C)')
plt.ylabel('Electricity Consumption (MWh)')
plt.title('Temperature vs Electricity Consumption')
plt.show()
```

Year	Temperature (°C)	Electricity Consumption (MWh)
2005	26.2	8.819913e+07
2006	26.5	9.700675e+07
2007	26.8	1.041922e+08
2008	27.0	1.119679e+08
2009	27.2	1.171662e+08
2010	27.5	1.274917e+08
2011	27.8	1.353263e+08
2012	28.0	1.459458e+08
2013	28.2	1.485649e+08
2014	28.5	1.646362e+08
2015	28.8	1.804233e+08
2016	29.0	1.805722e+08
2017	29.3	1.815990e+08
2018	29.5	1.707173e+08
2019	29.8	1.679766e+08
2020	30.0	1.728444e+08
2021	30.2	1.829464e+08
2022	30.5	1.876239e+08
أعلى سنة استهلاك: 2022, أقل سنة: 2005		



```
In [97]: corr_columns = ["Temperature (?C)", "Humidity (%)", "Rainfall (mm)", "Electricity Consumption (MWh)"]
correlation = df[corr_columns].corr()
print(correlation['Electricity Consumption (MWh)'])
```

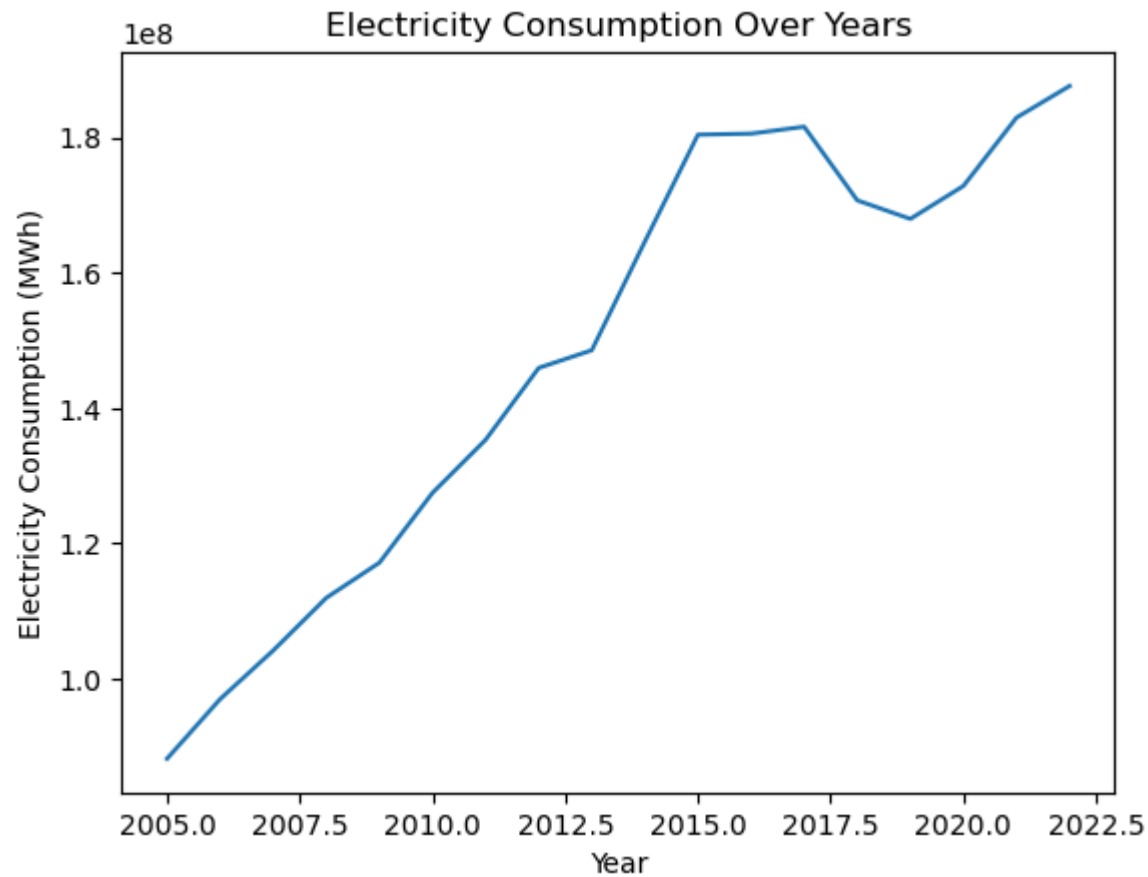
```
Temperature (?C)          0.946423
Humidity (%)              -0.944088
Rainfall (mm)             0.943259
Electricity Consumption (MWh)  1.000000
Name: Electricity Consumption (MWh), dtype: float64
```

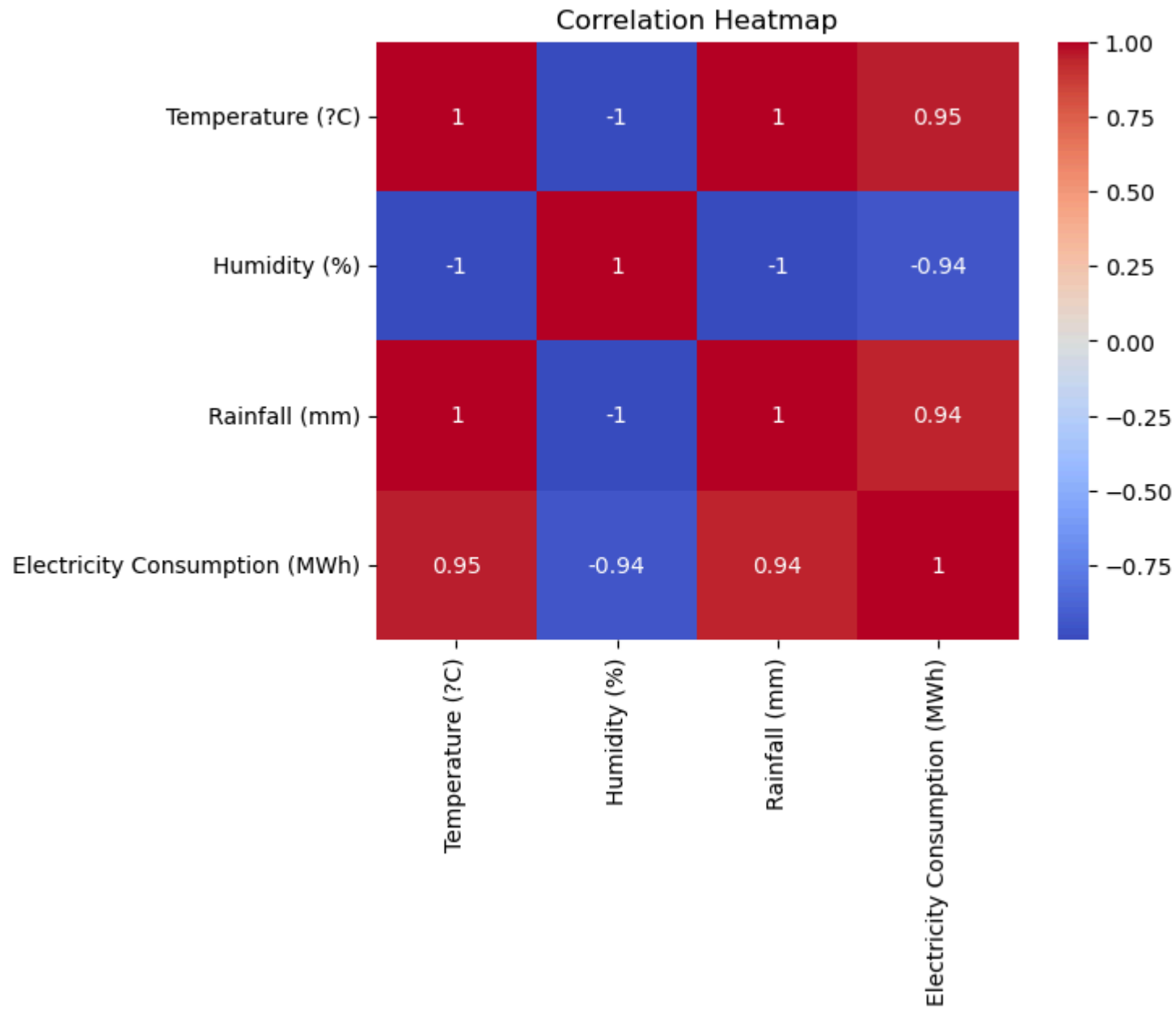
```
In [98]: import seaborn as sns
```

```
# مخطط زمني
```

```
plt.plot(df['Year'], df['Electricity Consumption (MWh)'])
plt.xlabel('Year')
plt.ylabel('Electricity Consumption (MWh)')
plt.title('Electricity Consumption Over Years')
plt.show()

# Heatmap
sns.heatmap(correlation, annot=True, cmap='coolwarm')
plt.title('Correlation Heatmap')
plt.show()
```





```
In [99]: from sklearn.model_selection import train_test_split  
from sklearn.linear_model import LinearRegression
```

```

from sklearn.metrics import mean_absolute_error, r2_score

X = df[['Temperature (?C)', 'Humidity (%)', 'Rainfall (mm)']]
y = df['Electricity Consumption (MWh)']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

model = LinearRegression()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)

print("MAE:", mean_absolute_error(y_test, y_pred))
print("R2:", r2_score(y_test, y_pred))

```

MAE: 9689517.34182097

R2: 0.7571582167001402

In [101...

```

import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression

# تحميل البيانات
df = pd.read_csv("riyadh_weather_energy_merged.csv")

# المتغيرات المستقلة والهدف
X = df[['Temperature (?C)', 'Humidity (%)', 'Rainfall (mm)']]
y = df['Electricity Consumption (MWh)']

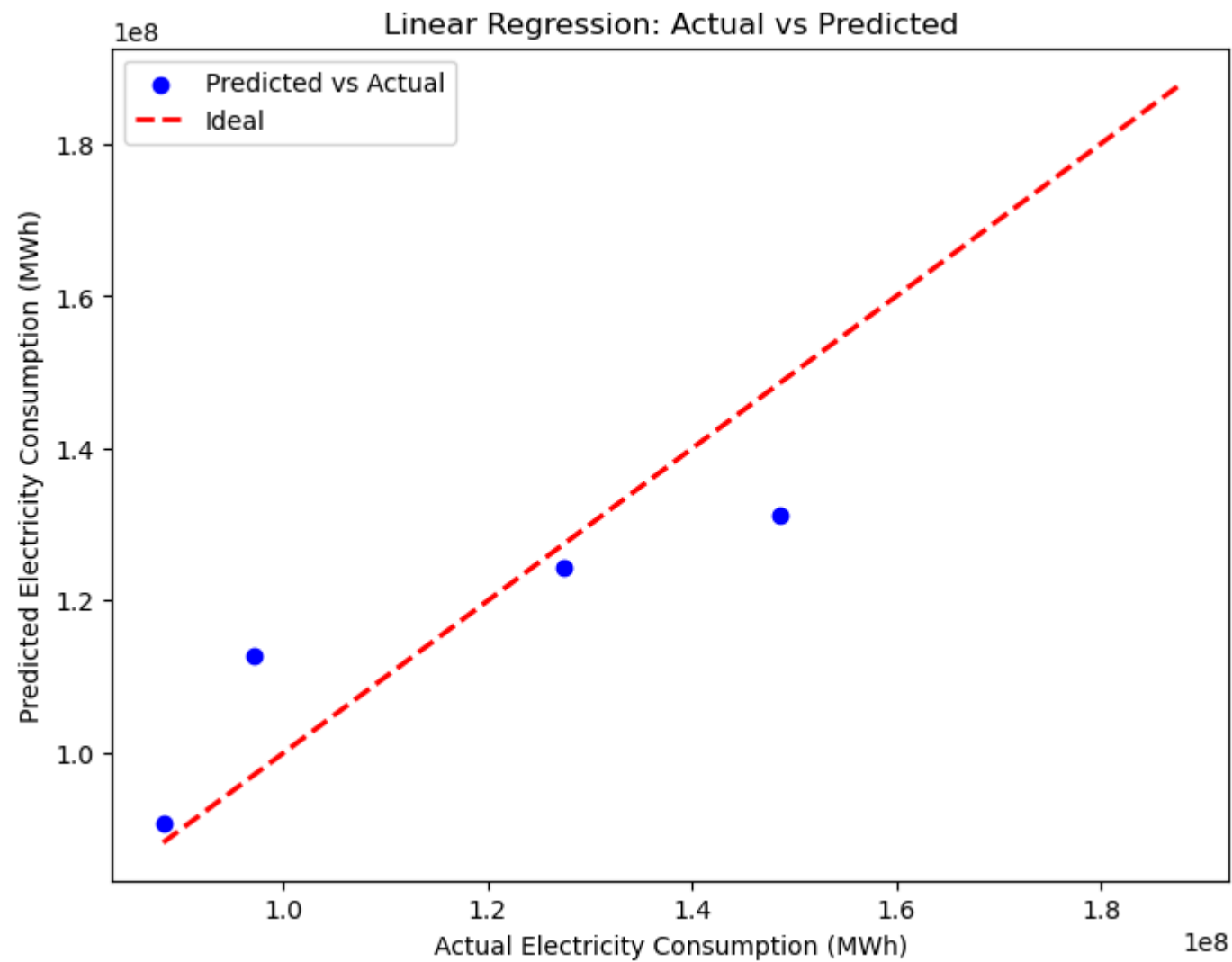
# تقسيم البيانات
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# تدريب نموذج Linear Regression
model = LinearRegression()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)

# رسم القيم الحقيقية مقابل التوقعات
plt.figure(figsize=(8,6))
plt.scatter(y_test, y_pred, color='blue', label='Predicted vs Actual')
plt.plot([y.min(), y.max()], [y.min(), y.max()], 'r--', lw=2, label='Ideal')

```

```
plt.xlabel('Actual Electricity Consumption (MWh)')  
plt.ylabel('Predicted Electricity Consumption (MWh)')  
plt.title('Linear Regression: Actual vs Predicted')  
plt.legend()  
plt.show()
```



```
In [113... import pandas as pd  
from sklearn.linear_model import LinearRegression
```



```

from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
from sklearn.model_selection import train_test_split
import numpy as np

# 1 تحميل البيانات
df = pd.read_csv("riyadh_weather_energy_merged.csv")

# المتغيرات المستقلة والهدف
X = df[['Temperature (?C)', 'Humidity (%)', 'Rainfall (mm)']]
y = df['Electricity Consumption (MWh)']

# تقسيم البيانات
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

# تدريب نموذج Linear Regression
model = LinearRegression()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)

# =====
# حساب مؤشرات الأداء
# =====
mae = mean_absolute_error(y_test, y_pred)
rmse = np.sqrt(mean_squared_error(y_test, y_pred))
r2 = r2_score(y_test, y_pred)

print(f"MAE: {mae:,.2f}")
print(f"RMSE: {rmse:,.2f}")
print(f"R²: {r2:.4f}")

```

MAE: 9,689,517.34

RMSE: 11,878,921.68

R²: 0.7572

In [115...

```

import numpy as np

# القيم الحقيقية والمتوقعة من النموذج
# (من الكود اللي سويته y_test و y_pred هذي عادة تكون)
accuracy = 100 - np.mean(np.abs((y_test - y_pred) / y_test)) * 100

print(f"📊 نسبة دقة التنبؤ: {accuracy:,.2f}%")

```

📊 91.70 : النسبة دقة التنبؤ :

```
In [12]: import pandas as pd
from sklearn.linear_model import LinearRegression

# 1 تحميل البيانات
df = pd.read_csv("riyadh_weather_energy_merged.csv")

# =====
# نموذج الاتجاه فقط (Trend)
# =====
X_year = df[['Year']] # DataFrame باسم العمود
y_consumption = df['Electricity Consumption (MWh)']

model_year = LinearRegression()
model_year.fit(X_year, y_consumption)

# =====
# نموذج باستخدام متوسط الطقس (Average Weather)
# =====
mean_temp = df['Temperature (?C)'].mean()
mean_humidity = df['Humidity (%)'].mean()
mean_rainfall = df['Rainfall (mm)'].mean()

X_weather = df[['Temperature (?C)', 'Humidity (%)', 'Rainfall (mm)']]
y_weather = df['Electricity Consumption (MWh)']

model_weather = LinearRegression()
model_weather.fit(X_weather, y_weather)

# =====
# دالة للتنبؤ بأي سنة
# =====
def predict_year(year):
    # Trend model
    pred_trend = model_year.predict(pd.DataFrame([[year]], columns=['Year']))[0]

    # Average Weather model
    future_weather = pd.DataFrame([{'Temperature (?C)': mean_temp,
                                     'Humidity (%)': mean_humidity,
```

```

        'Rainfall (mm)': mean_rainfall
    })
    pred_weather = model_weather.predict(future_weather)[0]

    print(f"Predicted Electricity Consumption for {year}:")
    print(f" - Trend only: {pred_trend:,.0f} MWh")
    print(f" - Average Weather: {pred_weather:,.0f} MWh")

    return pred_trend, pred_weather

# =====
# مثال: توقع سنة 2030
# =====
pred_trend, pred_weather = predict_year(2026)

```

Predicted Electricity Consumption for 2026:

- Trend only: 221,522,990 MWh
- Average Weather: 148,066,670 MWh

```

In [17]: results = []

for year in range(2023, 2041):
    pred_trend, pred_weather = predict_year(year)
    results.append({
        "Year": year,
        "Trend only (MWh)": pred_trend,
        "Average Weather (MWh)": pred_weather
    })

forecast_df = pd.DataFrame(results)
forecast_df.to_csv("riyadh_energy_forecast_2023_2040.csv", index=False)

print("✅ تم إنشاء ملف التوقعات من 2023 إلى 2040 بنجاح")
print(forecast_df.head())
print(forecast_df.tail())

```

Predicted Electricity Consumption for 2023:

- Trend only: 203,893,474 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2024:

- Trend only: 209,769,979 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2025:

- Trend only: 215,646,485 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2026:

- Trend only: 221,522,990 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2027:

- Trend only: 227,399,496 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2028:

- Trend only: 233,276,002 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2029:

- Trend only: 239,152,507 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2030:

- Trend only: 245,029,013 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2031:

- Trend only: 250,905,519 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2032:

- Trend only: 256,782,024 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2033:

- Trend only: 262,658,530 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2034:

- Trend only: 268,535,035 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2035:

- Trend only: 274,411,541 MWh
- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2036:

- Trend only: 280,288,047 MWh

- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2037:

- Trend only: 286,164,552 MWh

- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2038:

- Trend only: 292,041,058 MWh

- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2039:

- Trend only: 297,917,564 MWh

- Average Weather: 148,066,670 MWh

Predicted Electricity Consumption for 2040:

- Trend only: 303,794,069 MWh

- Average Weather: 148,066,670 MWh

✓ تم إنشاء ملف التوقعات من 2023 إلى 2040 بنجاح

	Year	Trend only (MWh)	Average Weather (MWh)
0	2023	2.038935e+08	1.480667e+08
1	2024	2.097700e+08	1.480667e+08
2	2025	2.156465e+08	1.480667e+08
3	2026	2.215230e+08	1.480667e+08
4	2027	2.273995e+08	1.480667e+08
	Year	Trend only (MWh)	Average Weather (MWh)
13	2036	2.802880e+08	1.480667e+08
14	2037	2.861646e+08	1.480667e+08
15	2038	2.920411e+08	1.480667e+08
16	2039	2.979176e+08	1.480667e+08
17	2040	3.037941e+08	1.480667e+08

```
In [20]: combined_df.to_csv("riyadh_energy_combined_actual_forecast.csv", index=False)
```

```
In [ ]:
```