

EnergyConsumption

June 21, 2024

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

```
[43]: df = pd.read_csv("Energy_consumption.csv")
```

```
[44]: df.head()
```

```
[44]:
```

	Timestamp	Temperature	Humidity	SquareFootage	Occupancy	\
0	2022-01-01 00:00:00	25.139433	43.431581	1565.693999	5	
1	2022-01-01 01:00:00	27.731651	54.225919	1411.064918	1	
2	2022-01-01 02:00:00	28.704277	58.907658	1755.715009	2	
3	2022-01-01 03:00:00	20.080469	50.371637	1452.316318	1	
4	2022-01-01 04:00:00	23.097359	51.401421	1094.130359	9	

	HVACUsage	LightingUsage	RenewableEnergy	DayOfWeek	Holiday	\
0	On	Off	2.774699	Monday	No	
1	On	On	21.831384	Saturday	No	
2	Off	Off	6.764672	Sunday	No	
3	Off	On	8.623447	Wednesday	No	
4	On	Off	3.071969	Friday	No	

	EnergyConsumption
0	75.364373
1	83.401855
2	78.270888
3	56.519850
4	70.811732

```
[45]: import pandas as pd

df['Timestamp'] = pd.to_datetime(df['Timestamp'])
```

```
[46]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
```

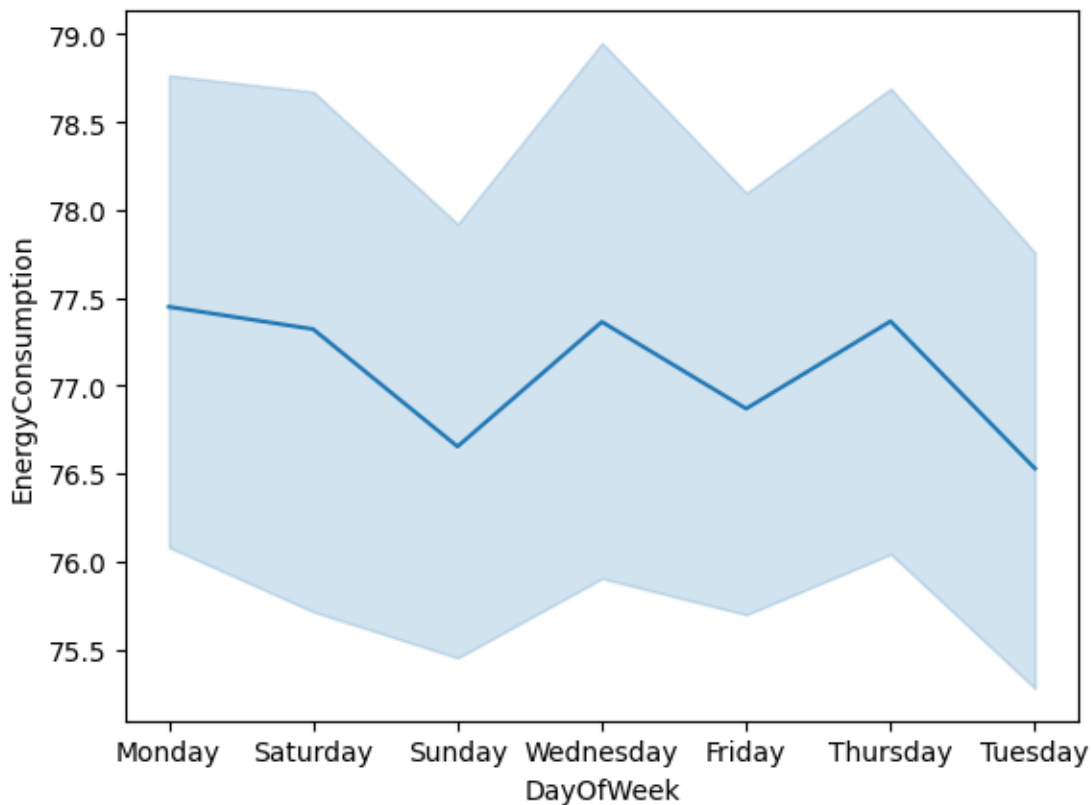
Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
0	Timestamp	1000 non-null	datetime64[ns]
1	Temperature	1000 non-null	float64
2	Humidity	1000 non-null	float64
3	SquareFootage	1000 non-null	float64
4	Occupancy	1000 non-null	int64
5	HVACUsage	1000 non-null	object
6	LightingUsage	1000 non-null	object
7	RenewableEnergy	1000 non-null	float64
8	DayOfWeek	1000 non-null	object
9	Holiday	1000 non-null	object
10	EnergyConsumption	1000 non-null	float64

dtypes: datetime64[ns](1), float64(5), int64(1), object(4)
memory usage: 86.1+ KB

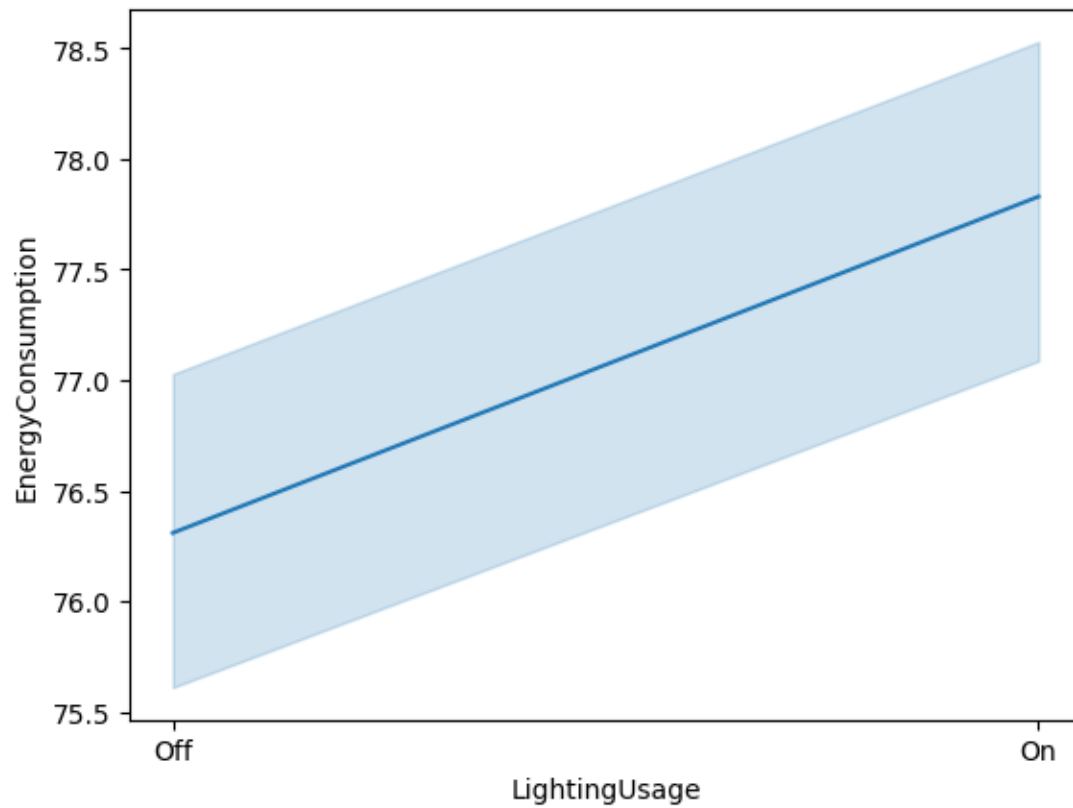
```
[47]: sns.lineplot(x="DayOfWeek",y="EnergyConsumption",data=df)
```

```
[47]: <Axes: xlabel='DayOfWeek', ylabel='EnergyConsumption'>
```



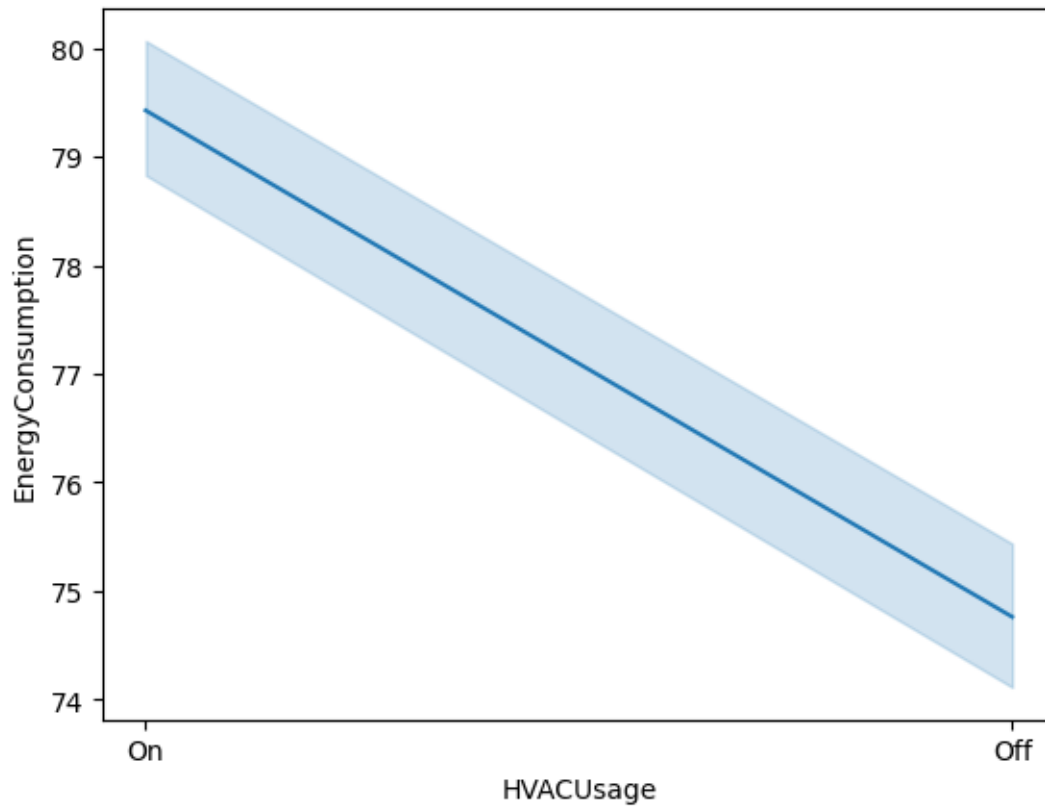
```
[48]: sns.lineplot(y="EnergyConsumption",x="LightingUsage",data=df)
```

```
[48]: <Axes: xlabel='LightingUsage', ylabel='EnergyConsumption'>
```



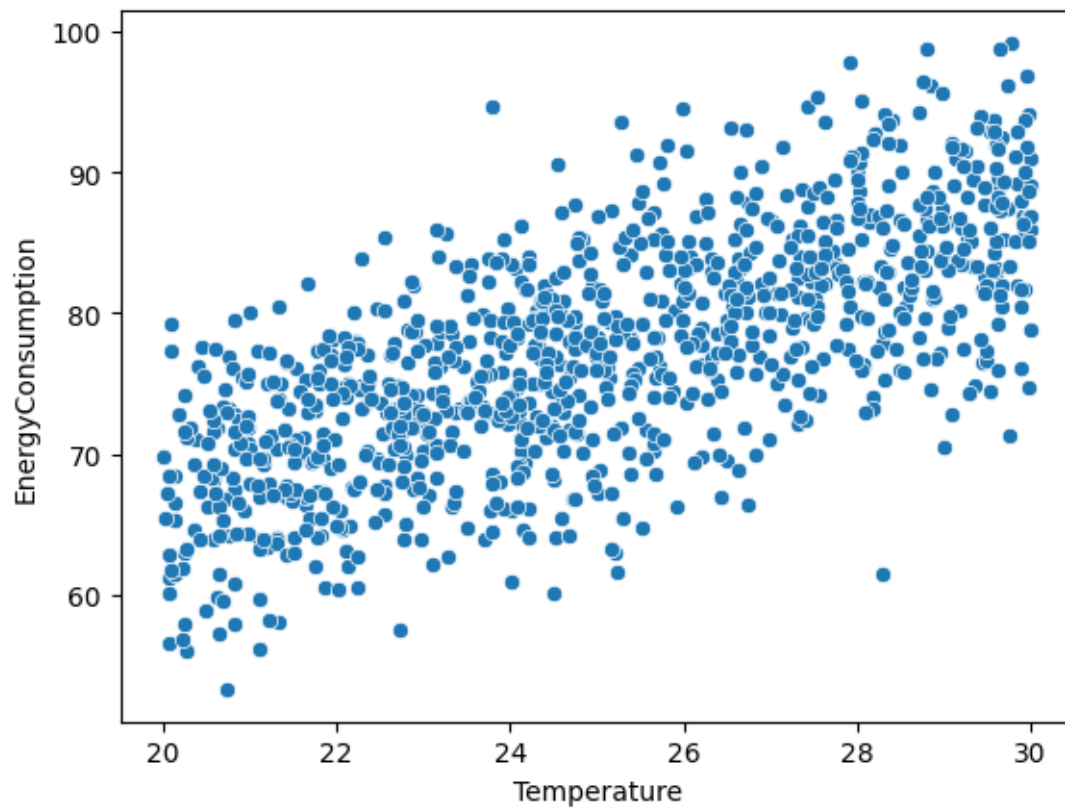
```
[49]: sns.lineplot(y="EnergyConsumption",x="HVACUsage",data=df)
```

```
[49]: <Axes: xlabel='HVACUsage', ylabel='EnergyConsumption'>
```



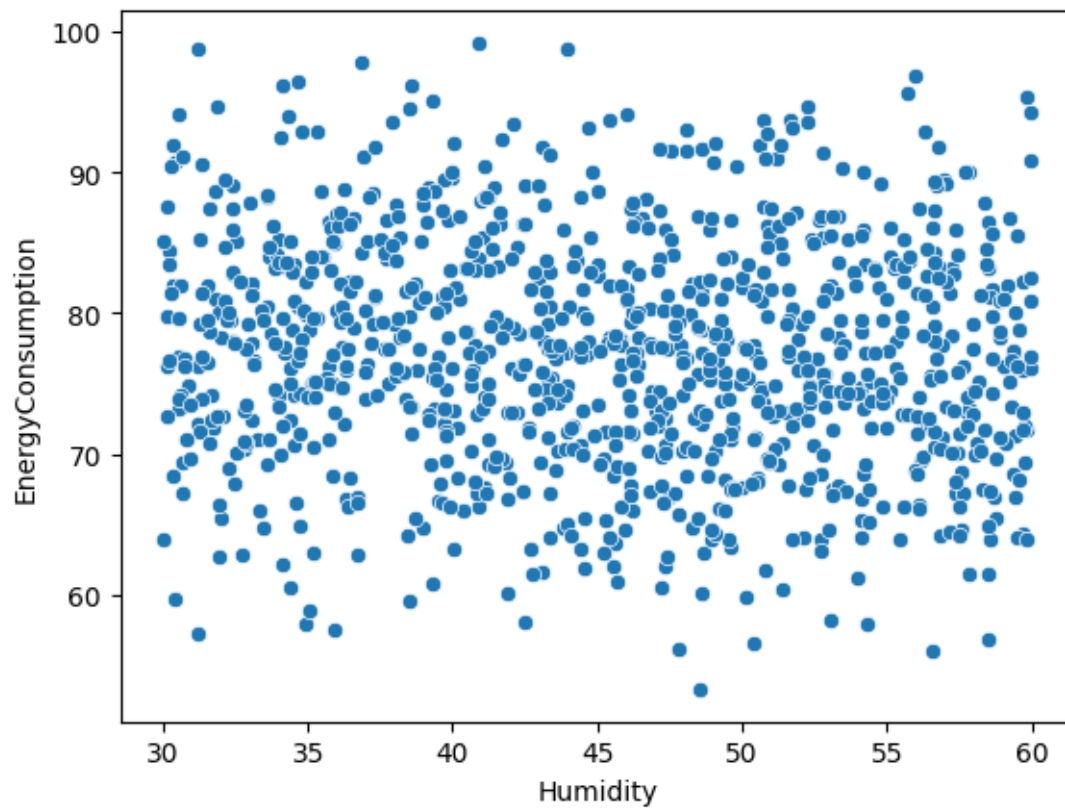
```
[50]: sns.scatterplot(y="EnergyConsumption",x="Temperature",data=df)
```

```
[50]: <Axes: xlabel='Temperature', ylabel='EnergyConsumption'>
```



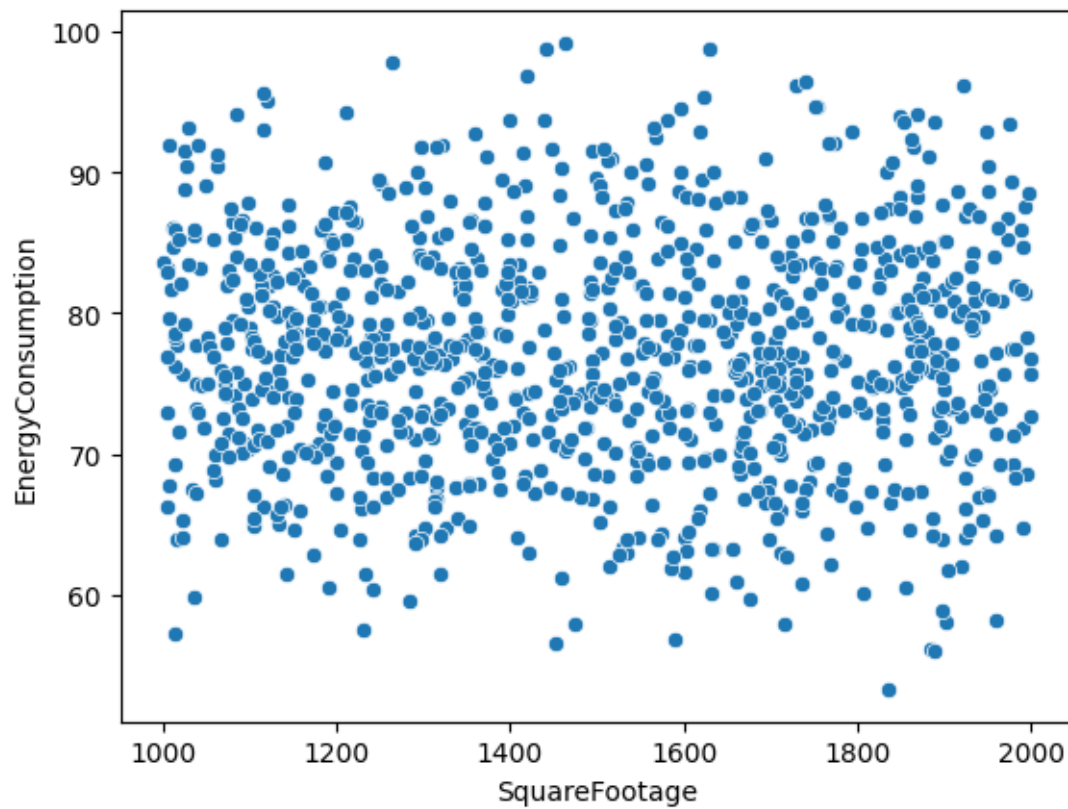
```
[51]: sns.scatterplot(y="EnergyConsumption",x="Humidity",data=df)
```

```
[51]: <Axes: xlabel='Humidity', ylabel='EnergyConsumption'>
```



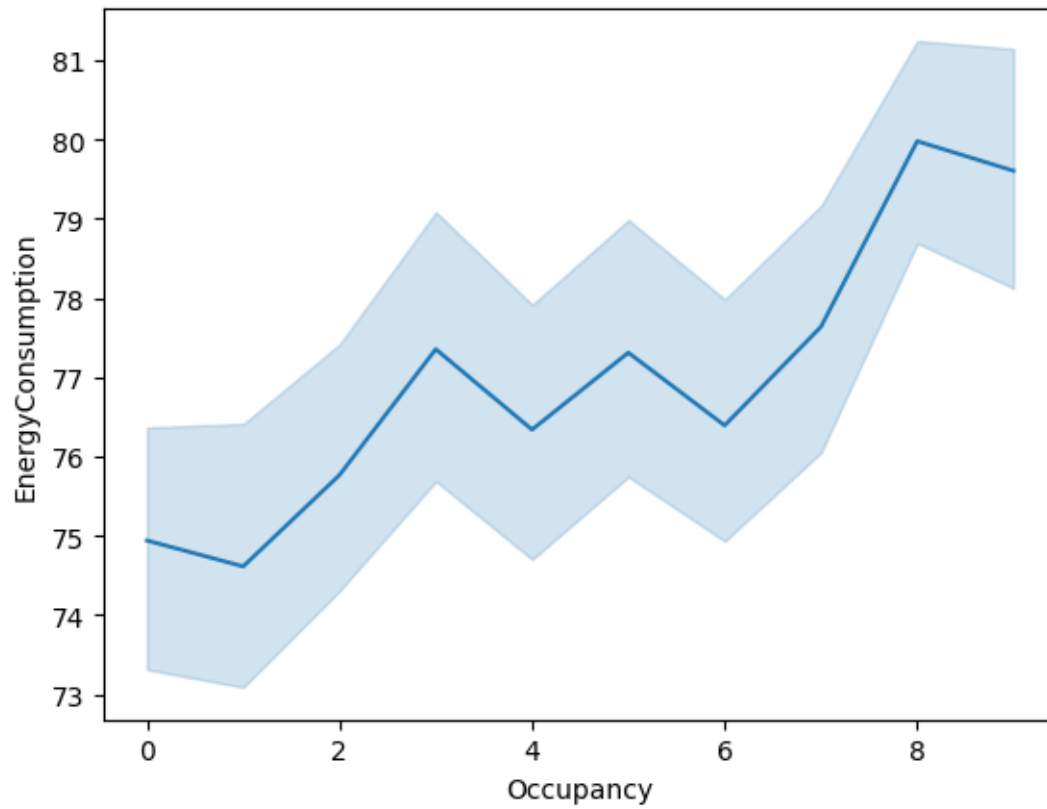
```
[52]: sns.scatterplot(y="EnergyConsumption",x="SquareFootage",data=df)
```

```
[52]: <Axes: xlabel='SquareFootage', ylabel='EnergyConsumption'>
```



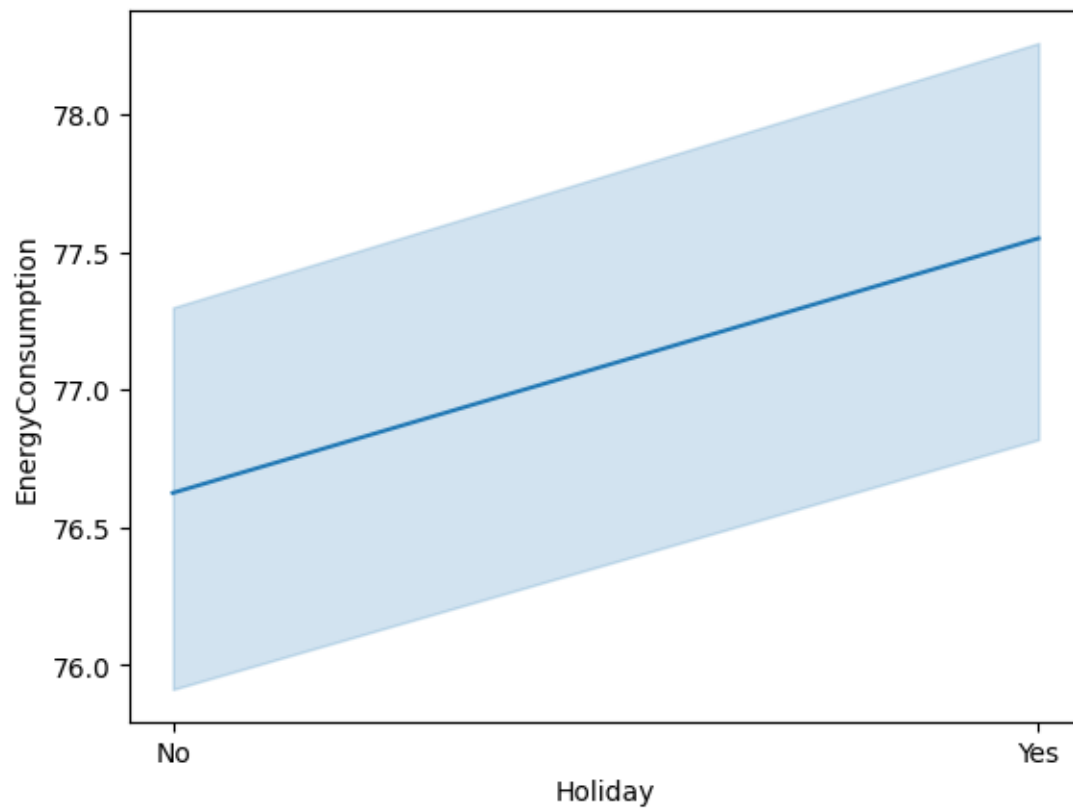
```
[53]: sns.lineplot(y="EnergyConsumption",x="Occupancy",data=df)
```

```
[53]: <Axes: xlabel='Occupancy', ylabel='EnergyConsumption'>
```



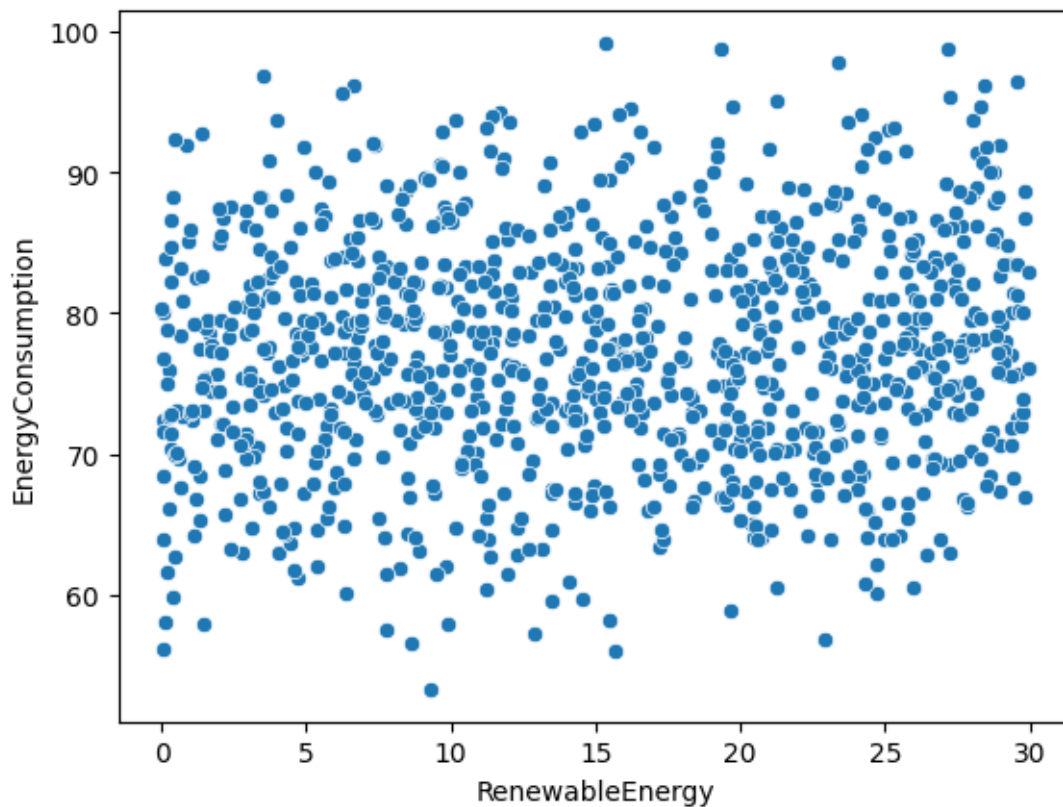
```
[54]: sns.lineplot(x="Holiday",y="EnergyConsumption",data=df)
```

```
[54]: <Axes: xlabel='Holiday', ylabel='EnergyConsumption'>
```

```
[55]: sns.scatterplot(x="RenewableEnergy",y="EnergyConsumption",data=df)
```

```
[55]: <Axes: xlabel='RenewableEnergy', ylabel='EnergyConsumption'>
```



```
[56]: from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
for i in ['HVACUsage', 'LightingUsage', 'Holiday', "DayOfWeek"]:
    df[i] = le.fit_transform(df[i])
```

```
[57]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Timestamp              1000 non-null   datetime64[ns]
1   Temperature            1000 non-null   float64
2   Humidity               1000 non-null   float64
3   SquareFootage          1000 non-null   float64
4   Occupancy             1000 non-null   int64
5   HVACUsage             1000 non-null   int32
6   LightingUsage          1000 non-null   int32
7   RenewableEnergy        1000 non-null   float64
8   DayOfWeek              1000 non-null   int32
```

```

9    Holiday          1000 non-null    int32
10   EnergyConsumption 1000 non-null    float64
dtypes: datetime64[ns](1), float64(5), int32(4), int64(1)
memory usage: 70.4 KB

```

```
[58]: df.head()
```

```

[58]:
      Timestamp  Temperature  Humidity  SquareFootage  Occupancy \
0 2022-01-01 00:00:00    25.139433  43.431581    1565.693999      5
1 2022-01-01 01:00:00    27.731651  54.225919    1411.064918      1
2 2022-01-01 02:00:00    28.704277  58.907658    1755.715009      2
3 2022-01-01 03:00:00    20.080469  50.371637    1452.316318      1
4 2022-01-01 04:00:00    23.097359  51.401421    1094.130359      9

      HVACUsage  LightingUsage  RenewableEnergy  DayOfWeek  Holiday \
0             1              0           2.774699         1         0
1             1              1          21.831384         2         0
2             0              0           6.764672         3         0
3             0              1           8.623447         6         0
4             1              0           3.071969         0         0

      EnergyConsumption
0           75.364373
1           83.401855
2           78.270888
3           56.519850
4           70.811732

```

```

[59]: X = df.drop(columns=['EnergyConsumption', 'Timestamp'])
      y = df['EnergyConsumption']

```

```

[60]: from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test = train_test_split(X, y, train_size=0.8,
      random_state=42)

```

0.1 Selecting the model and choosing the parameters

```

[65]: from sklearn.linear_model import LinearRegression

      model_params = {
          'linear_regression': {
              'model': LinearRegression(),
              'params': {
                  'fit_intercept': [True, False],
                  'copy_X': [True, False],
                  'n_jobs': [None, 1, -1],
                  'positive': [True, False]
              }
          }
      }

```

```

    }
}
}

model = LinearRegression()

```

```

[69]: from sklearn.model_selection import GridSearchCV
scores = []

for model_name, mp in model_params.items():
    clf = GridSearchCV(mp['model'], mp['params'], cv=5,
    ↪return_train_score=False)
    clf.fit(X_train, y_train)
    scores.append({
        'model': model_name,
        'best_score': clf.best_score_,
        'best_params': clf.best_params_
    })

best = pd.DataFrame(scores, columns=['model', 'best_score', 'best_params'])
best

```

```

[69]:          model  best_score \
0  linear_regression    0.612601

          best_params
0  {'copy_X': True, 'fit_intercept': True, 'n_job...

```

```

[75]: # Fit the final model with the best parameters
best_model_params = best.loc[best['model'] == 'linear_regression',
    ↪'best_params'].values[0]
final_model = LinearRegression(**best_model_params)
final_model.fit(X_train, y_train)

```

```

[75]: LinearRegression(positive=True)

```

```

[78]: y_pred = final_model.predict(X_test)

```

```

[79]: from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
import numpy as np

mae = mean_absolute_error(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)

```

```
r2 = r2_score(y_test, y_pred)

print("Mean Absolute Error:", mae)
print("Mean Squared Error:", mse)
print("Root Mean Squared Error:", rmse)
print("R2 Score:", r2)
```

Mean Absolute Error: 4.105494668584672
Mean Squared Error: 26.478930032168474
Root Mean Squared Error: 5.145768167355431
R² Score: 0.5957407580531275