

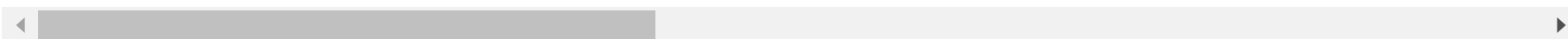
```
In [1]: import pandas as pd
import numpy as np
```

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

Out[2]:

	Year_Factor	State_Factor	building_class	facility_type	floor_area	year_built	energy_star_rating	ELEVATION	janua
0	1	State_1	Commercial	Grocery_store_or_food_market	61242.0	1942.0	11.0	2.4	
1	1	State_1	Commercial	Warehouse_Distribution_or_Shipping_center	274000.0	1955.0	45.0	1.8	
2	1	State_1	Commercial	Retail_Enclosed_mall	280025.0	1951.0	97.0	1.8	
3	1	State_1	Commercial	Education_Other_classroom	55325.0	1980.0	46.0	1.8	
4	1	State_1	Commercial	Warehouse_Nonrefrigerated	66000.0	1985.0	100.0	2.4	

5 rows × 64 columns



```
In [3]: df_null=df.isna().sum().values
df_null
```

```
Out[3]: array([ 0,  0,  0,  0,  0, 1837, 26709,  0,  0,
                0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
                0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
                0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
                0,  0,  0,  0, 41082, 41811, 41082, 45796,  0,
                0], dtype=int64)
```

```
In [4]: len(df)
```

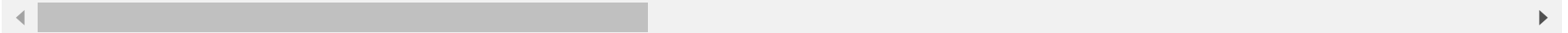
Out[4]: 75757

```
In [5]: df=df.dropna()
df.head()
```

Out[5]:

	Year_Factor	State_Factor	building_class	facility_type	floor_area	year_built	energy_star_rating	ELEVATION	january
404	2	State_1	Residential	Mixed_Use_Predominantly_Commercial	34173.0	1913.0	100.0	2.4	
405	2	State_1	Commercial	Lodging_Hotel	46800.0	1914.0	61.0	2.4	
406	2	State_1	Commercial	Lodging_Hotel	162214.0	1924.0	35.0	2.4	
407	2	State_1	Commercial	Lodging_Hotel	168000.0	1927.0	74.0	2.4	
411	2	State_1	Commercial	Lodging_Hotel	99000.0	1929.0	98.0	2.4	

5 rows × 64 columns



```
In [6]: df.isna().sum()
```

```
Out[6]: Year_Factor      0
State_Factor      0
building_class      0
facility_type      0
floor_area      0
..
direction_peak_wind_speed  0
max_wind_speed      0
days_with_fog      0
site_eui      0
id      0
Length: 64, dtype: int64
```

In [7]: `df.info()`

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

```
Int64Index: 11309 entries, 404 to 73148
```

```
Data columns (total 64 columns):
```

#	Column	Non-Null Count	Dtype
0	Year_Factor	11309 non-null	int64
1	State_Factor	11309 non-null	object
2	building_class	11309 non-null	object
3	facility_type	11309 non-null	object
4	floor_area	11309 non-null	float64
5	year_built	11309 non-null	float64
6	energy_star_rating	11309 non-null	float64
7	ELEVATION	11309 non-null	float64
8	january_min_temp	11309 non-null	int64
9	january_avg_temp	11309 non-null	float64
10	january_max_temp	11309 non-null	int64
11	february_min_temp	11309 non-null	int64
12	february_avg_temp	11309 non-null	float64
13	february_max_temp	11309 non-null	int64
14	march_min_temp	11309 non-null	int64
15	march_avg_temp	11309 non-null	float64
16	march_max_temp	11309 non-null	int64
17	april_min_temp	11309 non-null	int64
18	april_avg_temp	11309 non-null	float64
19	april_max_temp	11309 non-null	int64
20	may_min_temp	11309 non-null	int64
21	may_avg_temp	11309 non-null	float64
22	may_max_temp	11309 non-null	int64
23	june_min_temp	11309 non-null	int64
24	june_avg_temp	11309 non-null	float64
25	june_max_temp	11309 non-null	int64
26	july_min_temp	11309 non-null	int64
27	july_avg_temp	11309 non-null	float64
28	july_max_temp	11309 non-null	int64
29	august_min_temp	11309 non-null	int64
30	august_avg_temp	11309 non-null	float64
31	august_max_temp	11309 non-null	int64
32	september_min_temp	11309 non-null	int64
33	september_avg_temp	11309 non-null	float64
34	september_max_temp	11309 non-null	int64
35	october_min_temp	11309 non-null	int64

36	october_avg_temp	11309	non-null	float64
37	october_max_temp	11309	non-null	int64
38	november_min_temp	11309	non-null	int64
39	november_avg_temp	11309	non-null	float64
40	november_max_temp	11309	non-null	int64
41	december_min_temp	11309	non-null	int64
42	december_avg_temp	11309	non-null	float64
43	december_max_temp	11309	non-null	int64
44	cooling_degree_days	11309	non-null	int64
45	heating_degree_days	11309	non-null	int64
46	precipitation_inches	11309	non-null	float64
47	snowfall_inches	11309	non-null	float64
48	snowdepth_inches	11309	non-null	int64
49	avg_temp	11309	non-null	float64
50	days_below_30F	11309	non-null	int64
51	days_below_20F	11309	non-null	int64
52	days_below_10F	11309	non-null	int64
53	days_below_0F	11309	non-null	int64
54	days_above_80F	11309	non-null	int64
55	days_above_90F	11309	non-null	int64
56	days_above_100F	11309	non-null	int64
57	days_above_110F	11309	non-null	int64
58	direction_max_wind_speed	11309	non-null	float64
59	direction_peak_wind_speed	11309	non-null	float64
60	max_wind_speed	11309	non-null	float64
61	days_with_fog	11309	non-null	float64
62	site_eui	11309	non-null	float64
63	id	11309	non-null	int64

dtypes: float64(24), int64(37), object(3)

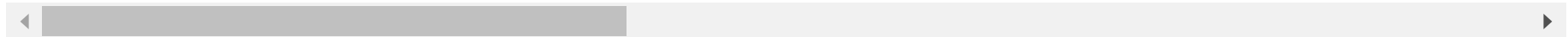
memory usage: 5.6+ MB

```
In [8]: df.describe()
```

```
Out[8]:
```

	Year_Factor	floor_area	year_built	energy_star_rating	ELEVATION	january_min_temp	january_avg_temp	january_max_temp	febr
count	11309.000000	1.130900e+04	11309.000000	11309.000000	11309.000000	11309.000000	11309.000000	11309.000000	
mean	4.239986	1.739833e+05	1950.422495	60.266160	25.995260	9.040676	32.028101	57.736847	
std	1.611790	2.394279e+05	36.709971	28.611072	35.606503	6.500818	5.151088	4.211625	
min	1.000000	8.019000e+03	0.000000	0.000000	1.800000	-9.000000	20.403226	44.000000	
25%	3.000000	6.683200e+04	1927.000000	39.000000	3.400000	6.000000	29.677419	56.000000	
50%	5.000000	9.835200e+04	1951.000000	66.000000	25.600000	8.000000	29.854839	56.000000	
75%	5.000000	1.725900e+05	1971.000000	84.000000	42.700000	11.000000	34.451613	59.000000	
max	6.000000	3.636683e+06	2015.000000	100.000000	201.800000	41.000000	55.096774	77.000000	

8 rows × 61 columns



```
In [9]: df_cat=df.select_dtypes(exclude=np.number)
df_cat
```

Out[9]:

	State_Factor	building_class	facility_type
404	State_1	Residential	Mixed_Use_Predominantly_Commercial
405	State_1	Commercial	Lodging_Hotel
406	State_1	Commercial	Lodging_Hotel
407	State_1	Commercial	Lodging_Hotel
411	State_1	Commercial	Lodging_Hotel
...
73140	State_11	Commercial	Warehouse_Uncategorized
73143	State_11	Commercial	Warehouse_Uncategorized
73146	State_11	Commercial	Office_Uncategorized
73147	State_11	Commercial	Warehouse_Uncategorized
73148	State_11	Commercial	Warehouse_Uncategorized

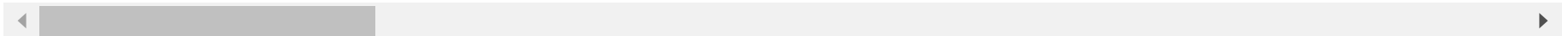
11309 rows × 3 columns

```
In [10]: df_en=pd.get_dummies(df_cat)
df_en
```

Out[10]:

	State_Factor_State_1	State_Factor_State_10	State_Factor_State_11	State_Factor_State_2	State_Factor_State_4	State_Factor_State_6	State_
404	1	0	0	0	0	0	
405	1	0	0	0	0	0	
406	1	0	0	0	0	0	
407	1	0	0	0	0	0	
411	1	0	0	0	0	0	
...
73140	0	0	1	0	0	0	
73143	0	0	1	0	0	0	
73146	0	0	1	0	0	0	
73147	0	0	1	0	0	0	
73148	0	0	1	0	0	0	

11309 rows × 45 columns

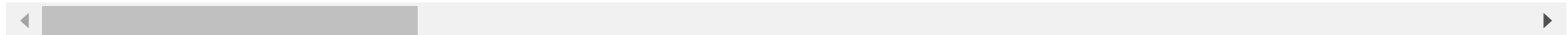



```
In [11]: df_new=pd.concat([df,df_en],axis=1)
df_new
```

Out[11]:

	Year_Factor	State_Factor	building_class	facility_type	floor_area	year_built	energy_star_rating	ELEVATION	janua
404	2	State_1	Residential	Mixed_Use_Predominantly_Commercial	34173.0	1913.0	100.0	2.4	
405	2	State_1	Commercial	Lodging_Hotel	46800.0	1914.0	61.0	2.4	
406	2	State_1	Commercial	Lodging_Hotel	162214.0	1924.0	35.0	2.4	
407	2	State_1	Commercial	Lodging_Hotel	168000.0	1927.0	74.0	2.4	
411	2	State_1	Commercial	Lodging_Hotel	99000.0	1929.0	98.0	2.4	
...
73140	6	State_11	Commercial	Warehouse_Uncategorized	39984.0	1956.0	75.0	57.3	
73143	6	State_11	Commercial	Warehouse_Uncategorized	22324.0	1941.0	53.0	57.3	
73146	6	State_11	Commercial	Office_Uncategorized	26510.0	1970.0	91.0	57.3	
73147	6	State_11	Commercial	Warehouse_Uncategorized	22058.0	1967.0	1.0	57.3	
73148	6	State_11	Commercial	Warehouse_Uncategorized	37400.0	1922.0	73.0	57.3	

11309 rows × 109 columns

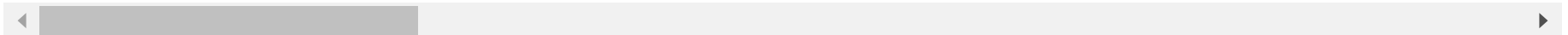


```
In [12]: df_new.drop(df_cat,axis=1,inplace=True)
df_new
```

Out[12]:

	Year_Factor	floor_area	year_built	energy_star_rating	ELEVATION	january_min_temp	january_avg_temp	january_max_temp	february_min_
404	2	34173.0	1913.0	100.0	2.4	27	48.951613		69
405	2	46800.0	1914.0	61.0	2.4	27	48.951613		69
406	2	162214.0	1924.0	35.0	2.4	27	48.951613		69
407	2	168000.0	1927.0	74.0	2.4	27	48.951613		69
411	2	99000.0	1929.0	98.0	2.4	27	48.951613		69
...
73140	6	39984.0	1956.0	75.0	57.3	28	43.451613		56
73143	6	22324.0	1941.0	53.0	57.3	28	43.451613		56
73146	6	26510.0	1970.0	91.0	57.3	28	43.451613		56
73147	6	22058.0	1967.0	1.0	57.3	28	43.451613		56
73148	6	37400.0	1922.0	73.0	57.3	28	43.451613		56

11309 rows × 106 columns



```
In [13]: df_new['cooling_degree_days']
```

```
Out[13]: 404      791
         405      791
         406      791
         407      791
         411      791
         ...
        73140     260
        73143     260
        73146     260
        73147     260
        73148     260
        Name: cooling_degree_days, Length: 11309, dtype: int64
```

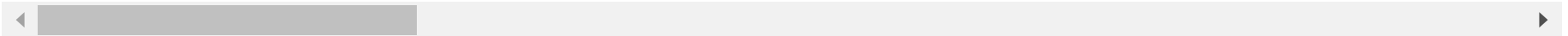
```
In [14]: from sklearn.impute import KNNImputer
         from category_encoders import TargetEncoder
```

```
In [15]: imputer = KNNImputer(n_neighbors=7)
df1 = pd.DataFrame(imputer.fit_transform(df_new), columns = df_new.columns)
df1
```

Out[15]:

	Year_Factor	floor_area	year_built	energy_star_rating	ELEVATION	january_min_temp	january_avg_temp	january_max_temp	february_min_
0	2.0	34173.0	1913.0	100.0	2.4	27.0	48.951613	69.0	
1	2.0	46800.0	1914.0	61.0	2.4	27.0	48.951613	69.0	
2	2.0	162214.0	1924.0	35.0	2.4	27.0	48.951613	69.0	
3	2.0	168000.0	1927.0	74.0	2.4	27.0	48.951613	69.0	
4	2.0	99000.0	1929.0	98.0	2.4	27.0	48.951613	69.0	
...
11304	6.0	39984.0	1956.0	75.0	57.3	28.0	43.451613	56.0	
11305	6.0	22324.0	1941.0	53.0	57.3	28.0	43.451613	56.0	
11306	6.0	26510.0	1970.0	91.0	57.3	28.0	43.451613	56.0	
11307	6.0	22058.0	1967.0	1.0	57.3	28.0	43.451613	56.0	
11308	6.0	37400.0	1922.0	73.0	57.3	28.0	43.451613	56.0	

11309 rows × 106 columns



```
In [16]: df1.isna().sum()
```

```
Out[16]: Year_Factor          0
         floor_area          0
         year_built          0
         energy_star_rating  0
         ELEVATION           0
         ..
         facility_type_Warehouse_Distribution_or_Shipping_center 0
         facility_type_Warehouse_Nonrefrigerated                 0
         facility_type_Warehouse_Refrigerated                   0
         facility_type_Warehouse_Selfstorage                    0
         facility_type_Warehouse_Uncategorized                  0
         Length: 106, dtype: int64
```



```
In [17]: df_all_te = df1.copy()
temp = [col for col in df_all_te.columns if 'temp' in col]

df_all_te['min_temp'] = df_all_te[temp].min(axis=1)
df_all_te['max_temp'] = df_all_te[temp].max(axis=1)
df_all_te['avg_temp'] = df_all_te[temp].mean(axis=1)
df_all_te['std_temp'] = df_all_te[temp].std(axis=1)
df_all_te['skew_temp'] = df_all_te[temp].skew(axis=1)

# by seasons
temp = pd.Series([col for col in df_all_te.columns if 'temp' in col])

winter_temp = temp[temp.apply(lambda x: ('january' in x or 'february' in x or 'december' in x))].values
spring_temp = temp[temp.apply(lambda x: ('march' in x or 'april' in x or 'may' in x))].values
summer_temp = temp[temp.apply(lambda x: ('june' in x or 'july' in x or 'august' in x))].values
autumn_temp = temp[temp.apply(lambda x: ('september' in x or 'october' in x or 'november' in x))].values

### winter
df_all_te['min_winter_temp'] = df_all_te[winter_temp].min(axis=1)
df_all_te['max_winter_temp'] = df_all_te[winter_temp].max(axis=1)
df_all_te['avg_winter_temp'] = df_all_te[winter_temp].mean(axis=1)
df_all_te['std_winter_temp'] = df_all_te[winter_temp].std(axis=1)
df_all_te['skew_winter_temp'] = df_all_te[winter_temp].skew(axis=1)
### spring
df_all_te['min_spring_temp'] = df_all_te[spring_temp].min(axis=1)
df_all_te['max_spring_temp'] = df_all_te[spring_temp].max(axis=1)
df_all_te['avg_spring_temp'] = df_all_te[spring_temp].mean(axis=1)
df_all_te['std_spring_temp'] = df_all_te[spring_temp].std(axis=1)
df_all_te['skew_spring_temp'] = df_all_te[spring_temp].skew(axis=1)
### summer
df_all_te['min_summer_temp'] = df_all_te[summer_temp].min(axis=1)
df_all_te['max_summer_temp'] = df_all_te[summer_temp].max(axis=1)
df_all_te['avg_summer_temp'] = df_all_te[summer_temp].mean(axis=1)
df_all_te['std_summer_temp'] = df_all_te[summer_temp].max(axis=1)
df_all_te['skew_summer_temp'] = df_all_te[summer_temp].max(axis=1)
## autumn
df_all_te['min_autumn_temp'] = df_all_te[autumn_temp].min(axis=1)
df_all_te['max_autumn_temp'] = df_all_te[autumn_temp].max(axis=1)
df_all_te['avg_autumn_temp'] = df_all_te[autumn_temp].mean(axis=1)
```

```
df_all_te['std_autumn_temp'] = df_all_te[autumn_temp].std(axis=1)  
df_all_te['skew_autumn_temp'] = df_all_te[autumn_temp].skew(axis=1)
```

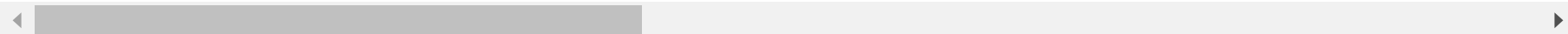
```
In [18]: df_all_te['month_cooling_degree_days'] = df_all_te['cooling_degree_days']/12  
df_all_te['month_heating_degree_days'] = df_all_te['heating_degree_days']/12
```

```
In [19]: df_all_te[temp]
```

Out[19]:

	january_min_temp	january_avg_temp	january_max_temp	february_min_temp	february_avg_temp	february_max_temp	march_min_temp	march_max_temp
0	27.0	48.951613	69.0	33.0	51.655172	78.0	34.0	84.0
1	27.0	48.951613	69.0	33.0	51.655172	78.0	34.0	84.0
2	27.0	48.951613	69.0	33.0	51.655172	78.0	34.0	84.0
3	27.0	48.951613	69.0	33.0	51.655172	78.0	34.0	84.0
4	27.0	48.951613	69.0	33.0	51.655172	78.0	34.0	84.0
...
11304	28.0	43.451613	56.0	34.0	47.672414	64.0	35.0	84.0
11305	28.0	43.451613	56.0	34.0	47.672414	64.0	35.0	84.0
11306	28.0	43.451613	56.0	34.0	47.672414	64.0	35.0	84.0
11307	28.0	43.451613	56.0	34.0	47.672414	64.0	35.0	84.0
11308	28.0	43.451613	56.0	34.0	47.672414	64.0	35.0	84.0

11309 rows × 41 columns




```
In [20]: df_all_te.cooling_degree_days
```

```
Out[20]: 0      791.0
         1      791.0
         2      791.0
         3      791.0
         4      791.0
         ...
        11304    260.0
        11305    260.0
        11306    260.0
        11307    260.0
        11308    260.0
        Name: cooling_degree_days, Length: 11309, dtype: float64
```

```
In [21]: tmp = df[['State_Factor', 'building_class', 'facility_type', 'site_eui']]
         df_all = df.drop(tmp.columns, axis=1)
```

```
In [22]: df_all_te = df.copy()

         cats = ['State_Factor', 'building_class', 'facility_type']
         for col in cats:
             encoder = TargetEncoder()
             df_all_te[f'te_{col}'] = encoder.fit_transform(df_all_te[col], df_all_te['site_eui'])
```

```
In [23]: # total area
         df_all_te['building_area'] = df_all_te['floor_area'] * df_all_te['ELEVATION']
         # rating energy by floor
         df_all_te['floor_energy_star_rating'] = df_all_te['energy_star_rating']/df_all_te['ELEVATION']
```

```
In [24]: df_all_te[["floor_area", "ELEVATION", "energy_star_rating", "floor_energy_star_rating", "building_area"]]
```

Out[24]:

	floor_area	ELEVATION	energy_star_rating	floor_energy_star_rating	building_area
404	34173.0	2.4	100.0	41.666667	82015.2
405	46800.0	2.4	61.0	25.416667	112320.0
406	162214.0	2.4	35.0	14.583333	389313.6
407	168000.0	2.4	74.0	30.833333	403200.0
411	99000.0	2.4	98.0	40.833333	237600.0
...
73140	39984.0	57.3	75.0	1.308901	2291083.2
73143	22324.0	57.3	53.0	0.924956	1279165.2
73146	26510.0	57.3	91.0	1.588133	1519023.0
73147	22058.0	57.3	1.0	0.017452	1263923.4
73148	37400.0	57.3	73.0	1.273997	2143020.0

11309 rows × 5 columns

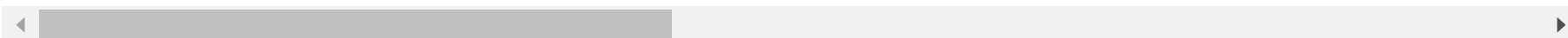
```
In [25]: df2=df_all_te.copy()
```

```
In [26]: df2.head()
```

```
Out[26]:
```

	Year_Factor	State_Factor	building_class	facility_type	floor_area	year_built	energy_star_rating	ELEVATION	january
404	2	State_1	Residential	Mixed_Use_Predominantly_Commercial	34173.0	1913.0	100.0	2.4	
405	2	State_1	Commercial	Lodging_Hotel	46800.0	1914.0	61.0	2.4	
406	2	State_1	Commercial	Lodging_Hotel	162214.0	1924.0	35.0	2.4	
407	2	State_1	Commercial	Lodging_Hotel	168000.0	1927.0	74.0	2.4	
411	2	State_1	Commercial	Lodging_Hotel	99000.0	1929.0	98.0	2.4	

5 rows × 69 columns



```
In [27]: df2.drop(['Year_Factor', 'State_Factor', 'building_class', 'facility_type'], axis=1, inplace=True)
```

```
In [28]: len(df2)
```

```
Out[28]: 11309
```

```
In [29]: from sklearn.model_selection import train_test_split
```

```
In [30]: x=df2
y=df['site_eui']
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2)
```

```
In [31]: x_train.shape,x_test.shape,y_train.shape,y_test.shape
```

```
Out[31]: ((9047, 65), (2262, 65), (9047,), (2262,))
```

```
In [32]: from catboost import CatBoostRegressor  
from sklearn.ensemble import RandomForestRegressor  
from xgboost import XGBRegressor  
from sklearn.linear_model import LinearRegression
```

```
In [33]: xg=XGBRegressor()  
model=xg.fit(x_train,y_train)
```

```
In [34]: pred=model.predict(x_test)
```

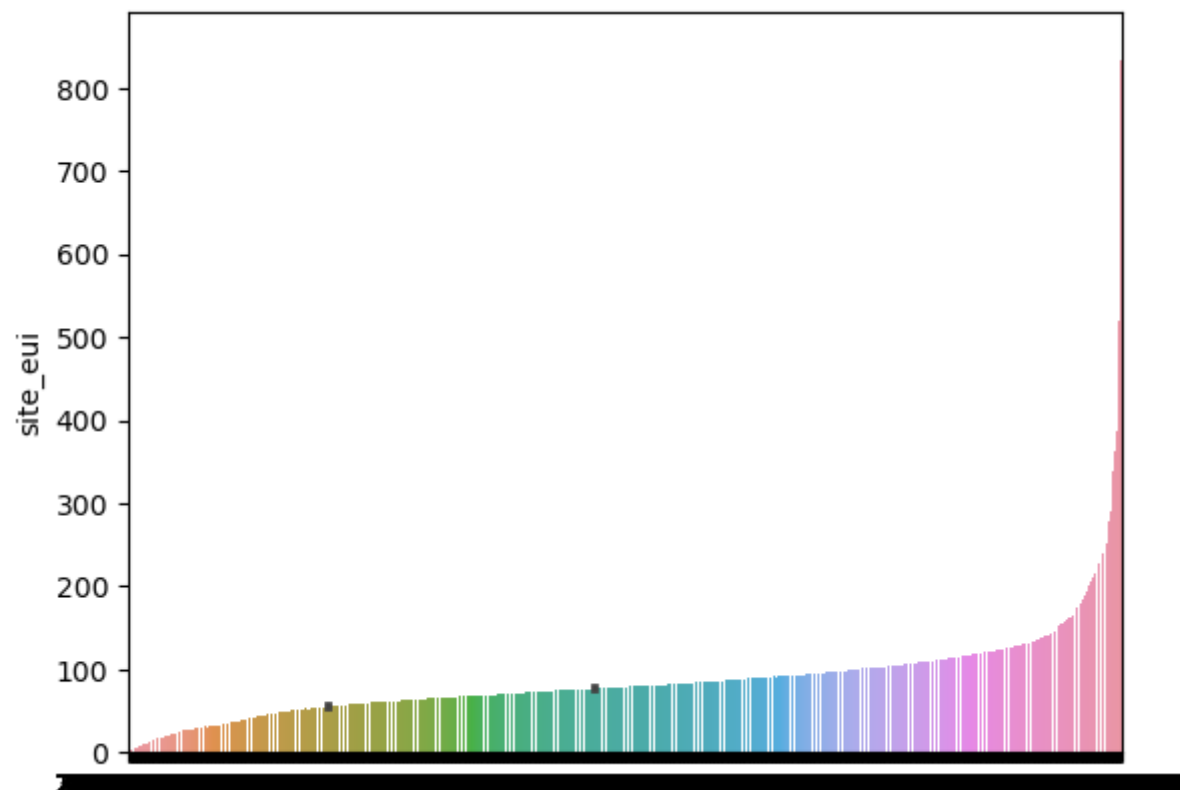
```
In [35]: model.score(x_train,y_train)
```

```
Out[35]: 0.9999923515570881
```

```
In [36]: import seaborn as sns
```

```
In [37]: sns.barplot(x=pred,y=y_test)
```

```
Out[37]: <Axes: ylabel='site_eui'>
```



```
In [38]: from sklearn.model_selection import GridSearchCV
```

```
In [39]: model_params={
    'catboostregresor':{
        'model':CatBoostRegressor(),
        'params':{
            'loss_function':['RMSE','MAE','Quantile:alpha=value','LogLinQuantile:alpha=value','Poisson']
        }
    },
    'xgboost':{
        'model':XGBRegressor(),
        'params':{
            'n_estimators':[100,200,300]
        }
    },
    'Linearregresion':{
        'model':LinearRegression(),
        'params':{}
    },
    'randomforest':{
        'model':RandomForestRegressor(),
        'params':{
            'n_estimators':[100,200,300]
        }
    }
}
```

```
In [41]: score=[]
for models,model_f in model_params.items():
    gs=GridSearchCV(model_f['model'],model_f['params'],cv=5)
    clf=gs.fit(x_train,y_train)
    score.append({'model':models,'params':clf.best_params_, 'accuracy':clf.best_score_})
```

```
432:   learn: 1.4286676      total: 1.99s   remaining: 2.6s
433:   learn: 1.4279253      total: 1.99s   remaining: 2.59s
434:   learn: 1.4236668      total: 1.99s   remaining: 2.59s
435:   learn: 1.4193439      total: 2s      remaining: 2.58s
436:   learn: 1.4178617      total: 2s      remaining: 2.58s
437:   learn: 1.4159913      total: 2s      remaining: 2.57s
438:   learn: 1.4120873      total: 2.01s   remaining: 2.57s
439:   learn: 1.4109190      total: 2.01s   remaining: 2.56s
440:   learn: 1.4085113      total: 2.02s   remaining: 2.56s
441:   learn: 1.4073307      total: 2.02s   remaining: 2.55s
442:   learn: 1.4047709      total: 2.03s   remaining: 2.55s
443:   learn: 1.4033696      total: 2.03s   remaining: 2.54s
444:   learn: 1.3992966      total: 2.03s   remaining: 2.54s
445:   learn: 1.3977175      total: 2.04s   remaining: 2.53s
446:   learn: 1.3934507      total: 2.06s   remaining: 2.54s
447:   learn: 1.3894592      total: 2.06s   remaining: 2.54s
448:   learn: 1.3878642      total: 2.07s   remaining: 2.54s
449:   learn: 1.3865175      total: 2.07s   remaining: 2.53s
450:   learn: 1.3828825      total: 2.07s   remaining: 2.52s
451:   learn: 1.3812622      total: 2.08s   remaining: 2.52s
```

```
In [42]: score
```

```
Out[42]: [{'model': 'catboostregresor',
  'params': {'loss_function': 'RMSE'},
  'accuracy': 0.9731014235094964},
 {'model': 'xgboost',
  'params': {'n_estimators': 300},
  'accuracy': 0.9991644338232544},
 {'model': 'Linearregresion', 'params': {}, 'accuracy': 1.0},
 {'model': 'randomforest',
  'params': {'n_estimators': 300},
  'accuracy': 0.9988719951589589}]
```

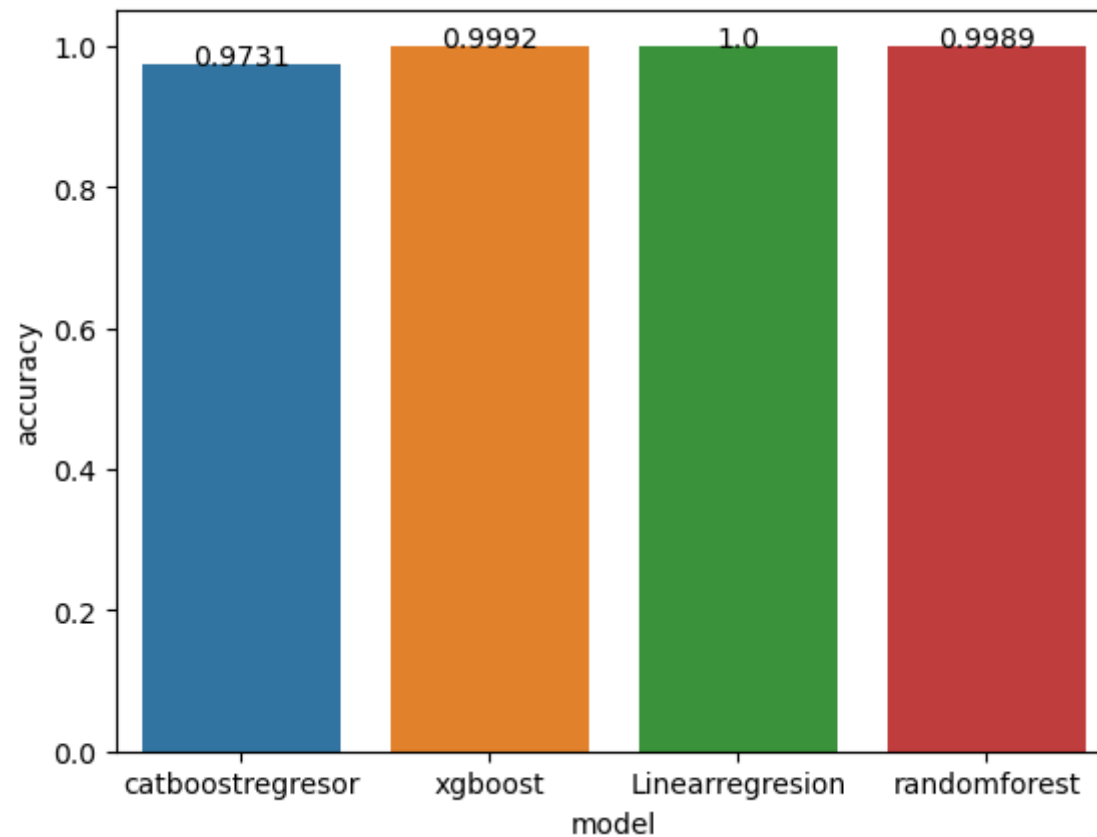
```
In [43]: model=pd.DataFrame(score)
```

```
In [63]: o=round(model['accuracy'],4)
o
```

```
Out[63]: 0    0.9731
1    0.9992
2    1.0000
3    0.9989
Name: accuracy, dtype: float64
```



```
In [64]: box=sns.barplot(y=model['accuracy'],x=model['model'])  
for i in range(4):  
    box.annotate(str(o[i]),xy=(i,o[i]),horizontalalignment='center')
```



```
In [ ]:
```