

# Global powerplant database

Covering approximately 30,000 power plants from 164 countries and includes thermal plants (e.g. coal, gas, oil, nuclear, biomass, waste, geothermal) and renewables (e.g. hydro, wind, solar)

### Objective:



To predict gigawatt per hour output using:

https://datasets.wri.org/dataset/globalpowerplantdatabase

capacity\_mw

latitude

longitude

prior2

prior1

Current





### Data cleaning:

-Dataset was approx 29000 rows with 24 columns

-Removed all categorical columns to get to final 6 for exploration and 3

concatenated datasets.



	capacity_mw	latitude	longitude	prior2	prior1	current
0	33.00	32.3220	65.1190	0.0	0.0	0.0
1	66.00	34.5560	69.4787	0.0	0.0	0.0
2	100.00	34.6410	69.7170	0.0	0.0	0.0
3	11.55	34.4847	70.3633	0.0	0.0	0.0
4	42.00	34.5638	69.1134	0.0	0.0	0.0
	:::				888	
89725	50.00	-12.9667	28.6333	0.0	0.0	0.0
89726	20.00	-12.8167	28.2000	0.0	0.0	0.0
89727	108.00	-17.9167	25.8500	0.0	0.0	0.0
89728	920.00	-18.3835	26.4700	0.0	0.0	0.0
89729	750.00	-16.5222	28.7619	0.0	0.0	0.0

89730 rows × 6 columns

#### Concatenation:



-Applied Lambda Function to normalize data

Concatenated 3 seperate data sets into 1:

Giving new dataset of 89730 rows

-This was done to get output values for



2013 -2014-2015-2016-207 and convert to prior 2 prior 1 and current(2017)



### Data collection

#### Values to be tested:

	capacity_mw	latitude	longitude	prior2	prior1	current
count	89730.000000	89730.000000	89730.000000	89730.000000	89730.000000	89730.000000
mean	186.294810	32.497558	-12.458680	169.183678	173.764769	178.141796
std	525.697713	23.654208	76.977407	1188.848308	1193.766551	1198.127832
min	1.000000	-77.847000	-179.977700	-989.619000	-989.619000	-934.944000
25%	4.772240	28.861200	-79.211900	0.000000	0.000000	0.000000
50%	18.900000	40.071250	-3.746700	0.000000	0.000000	0.000000
75%	100.000000	47.134900	24.933200	0.000000	0.000000	0.225000
max	22500.000000	71.292000	179.388700	59546.865000	59546.865000	59546.865000



#### Model results



Linear Regression:

Root mean Squared error:

0.8028505579010543 0.8292692491000222 mse is: 213561.5927061941 rmse is: 462.12724730986605

Decision Tree Regressor:

Random Forest Regressor:

R squared is: 0.8065837936494216 mse is: 241938.09765189045 rmse is: 491.8720338176287

R squared is: 0.876757516996191 mse is: 154160.04920390082 rmse is: 392.63220601970596



### Results/findings



#### KNN Regressor:

R squared is: 0.8606902117234034

mse is: 174258.12343168532 rmse is: 417.442359412273

## Linear regression after adding clusters:

0.5828643982315611

R squared is: 0.5741420296834572

#### Decision tree after clusters:

R squared is: 0.7808436596373483 mse is: 274135.60154098034

rmse is: 523.5796038244617







#### RandomForestRegressor

R squared is: 0.8311492012368606

mse is: 211210.02117945813 rmse is: 459.5759144901505

#### KNeighborsRegressor

R squared is: 0.78081714636564

mse is: 274168.766137891 rmse is: 523.6112738834898



### Conclusion:



- -Dataset had many Nan values to be taken into consideration
- -Need more practice cleaning and preprocessing
- -Visualizations in improved model
- -Include OLS and Kelbow
- -Include secondary fuel sources into dataset

