

## MODEL BUILDING -- LINEAR REGRESSION

SPLIT THE TRAIN AND TEST DATASET

```
In [10]: # !pip install pandas
import pandas as pd
# !pip install numpy
import numpy as np
# !pip install seaborn
import seaborn as sns
# !pip install matplotlib
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
# !pip install ipywidgets
import ipywidgets as widgets
```

READ THE CLEANED DATASET

```
In [11]: df = pd.read_csv('Cleaned_df.csv')
df = df.drop(['Unnamed: 0'],axis = 1)
```

SPLIT THE DATASET INTO X AND Y

```
In [12]: x = df.iloc[:,2:]
y = df.iloc[:,1]
```

SPLIT THE DATASET INTO TRAIN AND TEST OF X AND Y

```
In [13]: # !pip install sklearn
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)
```

CHECK FOR SHAPE OF TRAIN AND TEST SETS

```
In [14]: print('Shape of x_train: ',x_train.shape)
print('Shape of x_test: ',x_test.shape)
print('Shape of y_train: ',y_train.shape)
print('Shape of y_test: ',y_test.shape)
```

```
Shape of x_train: (196159, 19)
Shape of x_test: (49040, 19)
Shape of y_train: (196159,)
Shape of y_test: (49040,)
```

FIT THE MODELS

```
In [15]: from sklearn.linear_model import LinearRegression
lr = LinearRegression()
lr = lr.fit(x_train,y_train)
```

PREDICT THE MODEL

```
In [16]: y_pred = lr.predict(x_test)
```

MODEL-1

```
In [17]: # !pip install statmodels
import statsmodels.api as sm
x_train_sm = x_train

x_train_sm = sm.add_constant(x_train_sm)

mlm = sm.OLS(y_train,x_train_sm).fit()

mlm.params
print(mlm.summary())
```

```

=====
                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.938
Model:                  OLS                  Adj. R-squared:           0.938
Method:                 Least Squares        F-statistic:             1.552e+05
Date:                  Mon, 10 May 2021      Prob (F-statistic):       0.00
Time:                  00:40:34              Log-Likelihood:          -1.2446e+06
No. Observations:      196159               AIC:                    2.489e+06
Df Residuals:          196139               BIC:                    2.489e+06
Df Model:              19
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-3124.6801	9.735	-320.962	0.000	-3143.761	-3105.599
Avg_Ambient_Temp	-4.8097	0.095	-50.819	0.000	-4.995	-4.624
Avg_Generator_Speed	-0.0036	0.077	-0.047	0.963	-0.154	0.146
Avg_Nacelle_Pos	5.365e-05	0.003	0.016	0.987	-0.006	0.006
Avg_Pitch_Angle	12.3866	0.045	275.231	0.000	12.298	12.475
Avg_Rotor_Speed	116.5211	8.183	14.239	0.000	100.483	132.560
Avg_Wind_Speed	32.1814	0.325	99.074	0.000	31.545	32.818
Bearing_DE_Temp	-9.4580	0.153	-61.715	0.000	-9.758	-9.158
Bearing_NDE_Temp	-8.7528	0.158	-55.406	0.000	-9.062	-8.443
Gearbox_bearing_Temp	2.9521	0.256	11.512	0.000	2.449	3.455
Gearbox_oil_Temp	26.9376	0.341	78.954	0.000	26.269	27.606
Generator_wind_Temp_1	116.9005	2.156	54.210	0.000	112.674	121.127
Generator_wind_Temp_2	-144.2204	2.128	-67.784	0.000	-148.391	-140.050
Generator_wind_Temp_3	34.7784	0.289	120.193	0.000	34.211	35.345
Generators_sliprings_Temp	14.6944	0.154	95.160	0.000	14.392	14.997
Hidraulic_group_pressure	0.2188	0.014	15.428	0.000	0.191	0.247
Nacelle_Misalignment_Avg_Wind_Dir	0.0027	0.002	1.479	0.139	-0.001	0.006
Trafo_1_wind_Temp	-1.9114	0.084	-22.684	0.000	-2.077	-1.746
Trafo_2_wind_Temp	0.4971	0.071	7.003	0.000	0.358	0.636
Trafo_3_wind_Temp	3.8216	0.074	51.661	0.000	3.677	3.967

```

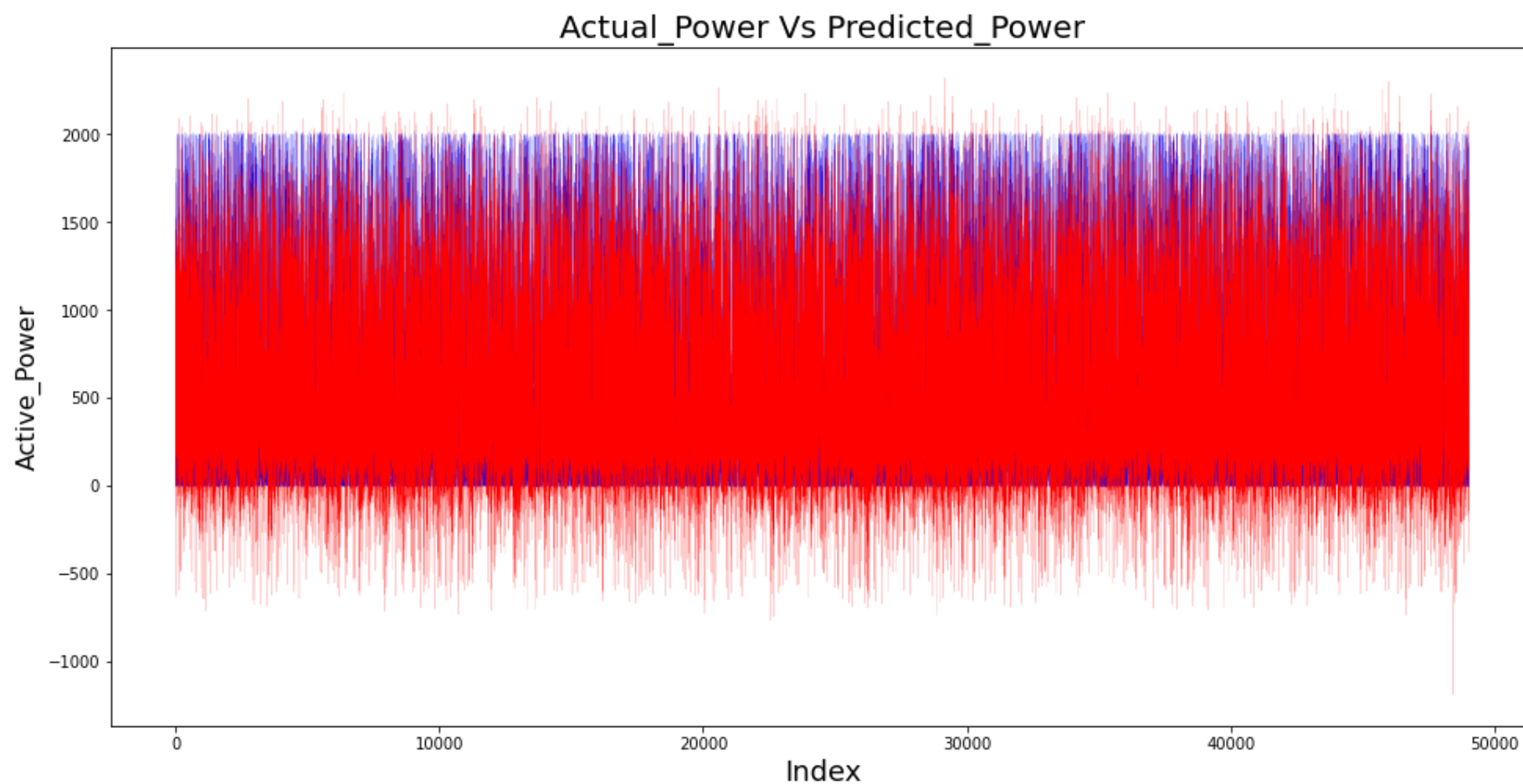
=====
Omnibus:                 34340.376      Durbin-Watson:              1.998
Prob(Omnibus):           0.000         Jarque-Bera (JB):          263120.746
Skew:                    0.636         Prob(JB):                  0.00
Kurtosis:                 8.530         Cond. No.                  3.86e+04
=====

```

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.  
 [2] The condition number is large, 3.86e+04. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [18]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



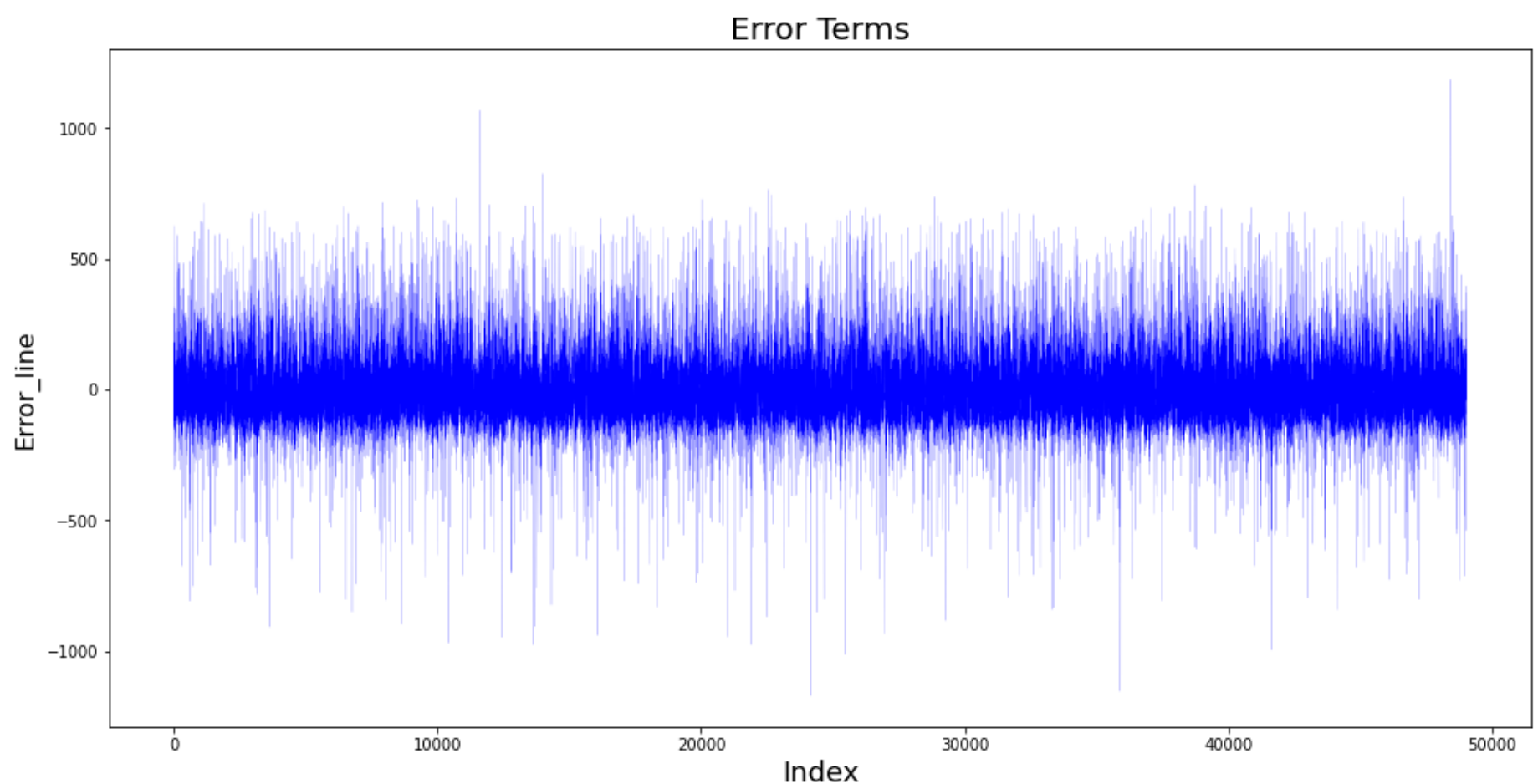
```
In [19]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x_train.values, i)
                    for i in range(len(x_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	70.956154
1	Avg_Generator_Speed	83644.989518
2	Avg_Nacelle_Pos	3.898009
3	Avg_Pitch_Angle	13.630874
4	Avg_Rotor_Speed	84271.377809
5	Avg_Wind_Speed	44.248870
6	Bearing_DE_Temp	615.214488
7	Bearing_NDE_Temp	669.374913
8	Gearbox_bearing_Temp	2756.496545
9	Gearbox_oil_Temp	2659.267301
10	Generator_wind_Temp_1	231529.737063
11	Generator_wind_Temp_2	232227.893434
12	Generator_wind_Temp_3	4572.275104
13	Generators_sliprings_Temp	450.132150
14	Hidraulic_group_pressure	81.460160
15	Nacelle_Misalignment_Avg_Wind_Dir	1.930151
16	Trafo_1_wind_Temp	311.304992
17	Trafo_2_wind_Temp	273.175199
18	Trafo_3_wind_Temp	285.184301

```
In [20]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [21]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred)
r_squared = r2_score(y_test, y_pred)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 18890.960947349187

r\_square\_value : 0.938 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-2

```
In [22]: x1_train = x_train[['Avg_Ambient_Temp', 'Avg_Generator_Speed', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed',
                             'Avg_Wind_Speed', 'Bearing_DE_Temp', 'Bearing_NDE_Temp', 'Gearbox_bearing_Temp', 'Gearbox_oil_Temp',
                             'Generator_wind_Temp_1', 'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                             'Hydraulic_group_pressure', 'Nacelle_Misalignment_Avg_Wind_Dir', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                             'Trafo_3_wind_Temp']]
x1_test = x_test[['Avg_Ambient_Temp', 'Avg_Generator_Speed', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed',
                   'Avg_Wind_Speed', 'Bearing_DE_Temp', 'Bearing_NDE_Temp', 'Gearbox_bearing_Temp', 'Gearbox_oil_Temp',
                   'Generator_wind_Temp_1', 'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                   'Hydraulic_group_pressure', 'Nacelle_Misalignment_Avg_Wind_Dir', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                   'Trafo_3_wind_Temp']]
```

Avg\_Nacelle\_Pos -- variable is removed because of its significant value is more than 0.05.

```
In [23]: lr1 = lr.fit(x1_train,y_train)
```

```
In [24]: y_pred1 = lr1.predict(x1_test)
```

```
In [25]: import statsmodels.api as sm
x1_train_sm = x1_train

x1_train_sm = sm.add_constant(x1_train_sm)

mlm1 = sm.OLS(y_train,x1_train_sm).fit()

mlm1.params
print(mlm1.summary())
```

```

=====
                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.938
Model:                  OLS                  Adj. R-squared:           0.938
Method:                 Least Squares         F-statistic:             1.639e+05
Date:                  Mon, 10 May 2021       Prob (F-statistic):       0.00
Time:                  00:40:49               Log-Likelihood:          -1.2446e+06
No. Observations:      196159                AIC:                    2.489e+06
Df Residuals:          196140                BIC:                    2.489e+06
Df Model:              18
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-3124.6778	9.734	-320.999	0.000	-3143.757	-3105.599
Avg_Ambient_Temp	-4.8097	0.095	-50.888	0.000	-4.995	-4.624
Avg_Generator_Speed	-0.0036	0.077	-0.047	0.963	-0.154	0.146
Avg_Pitch_Angle	12.3866	0.045	275.319	0.000	12.298	12.475
Avg_Rotor_Speed	116.5212	8.183	14.239	0.000	100.483	132.560
Avg_Wind_Speed	32.1813	0.325	99.090	0.000	31.545	32.818
Bearing_DE_Temp	-9.4580	0.153	-61.722	0.000	-9.758	-9.158
Bearing_NDE_Temp	-8.7529	0.158	-55.415	0.000	-9.062	-8.443
Gearbox_bearing_Temp	2.9522	0.256	11.512	0.000	2.450	3.455
Gearbox_oil_Temp	26.9375	0.341	78.975	0.000	26.269	27.606
Generator_wind_Temp_1	116.8993	2.155	54.242	0.000	112.675	121.123
Generator_wind_Temp_2	-144.2192	2.126	-67.825	0.000	-148.387	-140.052
Generator_wind_Temp_3	34.7784	0.289	120.207	0.000	34.211	35.345
Generators_sliprings_Temp	14.6946	0.154	95.511	0.000	14.393	14.996
Hidraulic_group_pressure	0.2188	0.014	15.428	0.000	0.191	0.247
Nacelle_Misalignment_Avg_Wind_Dir	0.0027	0.002	1.479	0.139	-0.001	0.006
Trafo_1_wind_Temp	-1.9114	0.084	-22.687	0.000	-2.077	-1.746
Trafo_2_wind_Temp	0.4971	0.071	7.003	0.000	0.358	0.636
Trafo_3_wind_Temp	3.8216	0.074	51.661	0.000	3.677	3.967

```

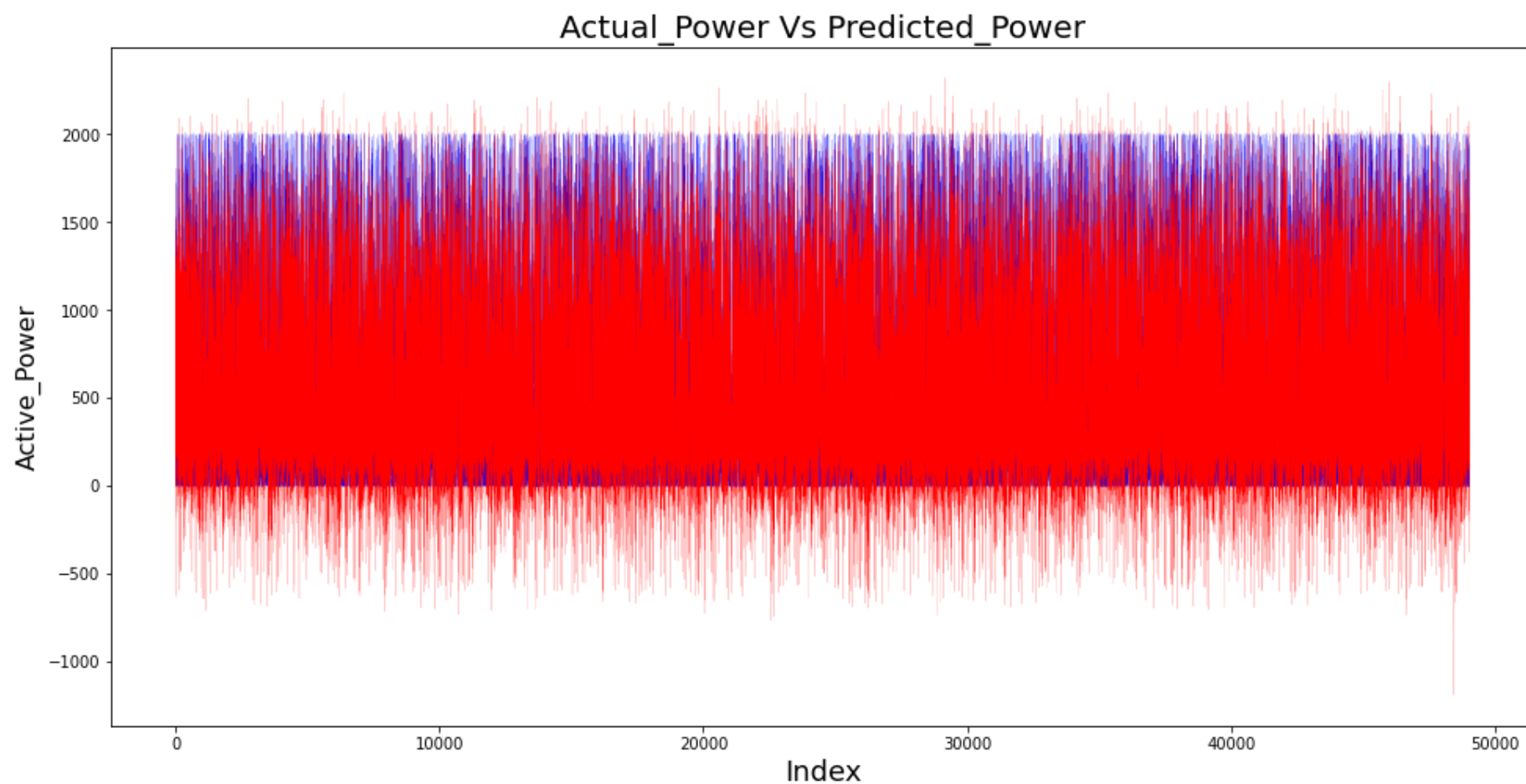
=====
Omnibus:                 34340.705      Durbin-Watson:           1.998
Prob(Omnibus):           0.000      Jarque-Bera (JB):        263122.547
Skew:                   0.636      Prob(JB):                0.00
Kurtosis:                8.530      Cond. No.                3.83e+04
=====

```

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.  
 [2] The condition number is large, 3.83e+04. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [26]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred1, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [27]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x1_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x1_train.values, i)
                    for i in range(len(x1_train.columns))]

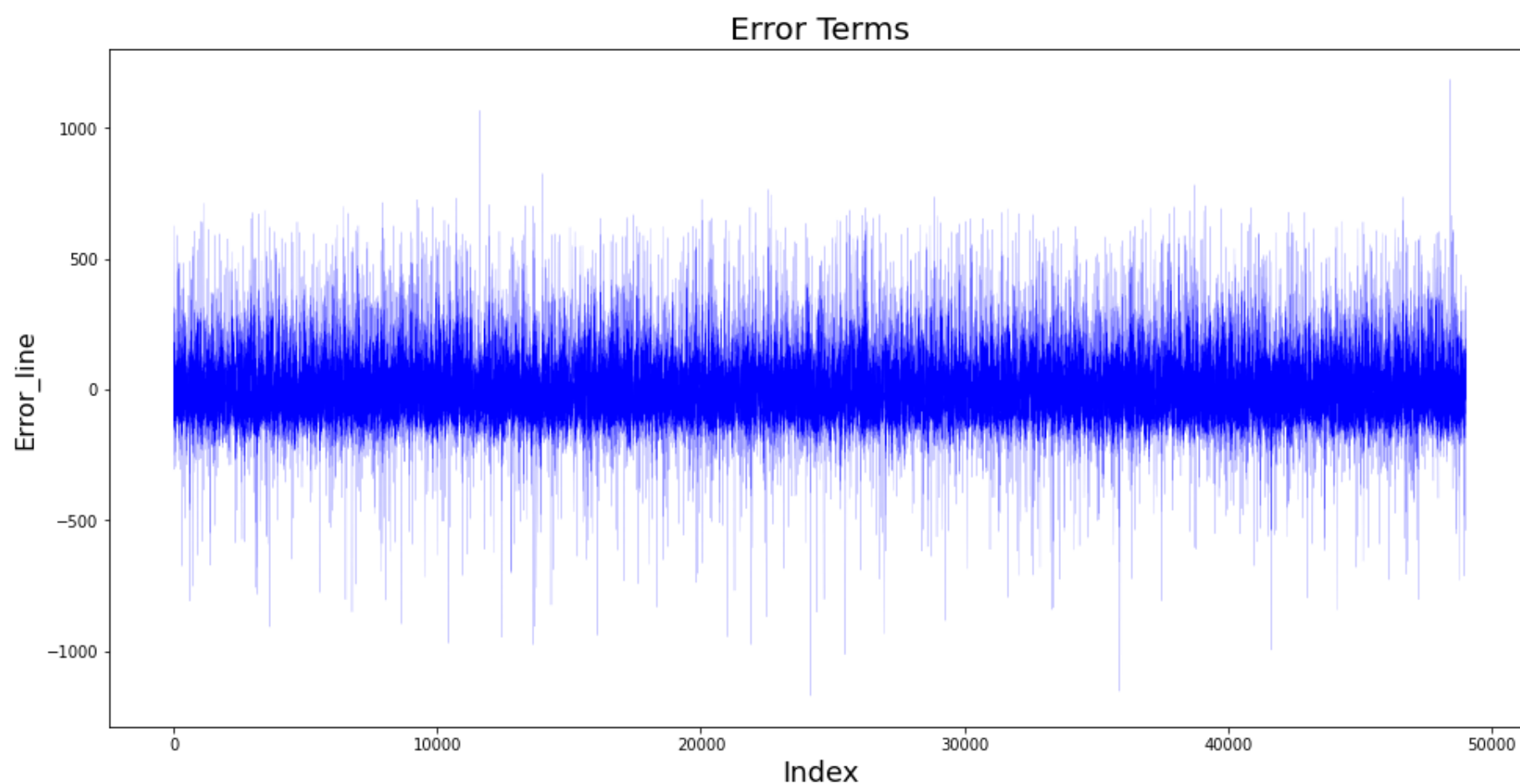
print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	70.755721
1	Avg_Generator_Speed	83644.968373
2	Avg_Pitch_Angle	13.620940
3	Avg_Rotor_Speed	84271.360161
4	Avg_Wind_Speed	44.235271
5	Bearing_DE_Temp	615.113078
6	Bearing_NDE_Temp	669.097227
7	Gearbox_bearing_Temp	2756.379755
8	Gearbox_oil_Temp	2658.394396
9	Generator_wind_Temp_1	231153.373715
10	Generator_wind_Temp_2	231860.090841
11	Generator_wind_Temp_3	4571.090312
12	Generators_sliprings_Temp	446.646861
13	Hidraulic_group_pressure	81.459161
14	Nacelle_Misalignment_Avg_Wind_Dir	1.930145
15	Trafo_1_wind_Temp	311.207540
16	Trafo_2_wind_Temp	273.126137
17	Trafo_3_wind_Temp	285.184025



```
In [28]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)               # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [29]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred1)
r_squared = r2_score(y_test, y_pred1)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 18890.96745365323

r\_square\_value : 0.938 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

### MODEL-3

```
In [30]: x2_train = x_train[['Avg_Ambient_Temp', 'Avg_Generator_Speed', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed',
                             'Avg_Wind_Speed', 'Bearing_DE_Temp', 'Bearing_NDE_Temp', 'Gearbox_bearing_Temp', 'Gearbox_oil_Temp',
                             'Generator_wind_Temp_1', 'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                             'Hidraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                             'Trafo_3_wind_Temp']]
x2_test = x_test[['Avg_Ambient_Temp', 'Avg_Generator_Speed', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed',
                   'Avg_Wind_Speed', 'Bearing_DE_Temp', 'Bearing_NDE_Temp', 'Gearbox_bearing_Temp', 'Gearbox_oil_Temp',
                   'Generator_wind_Temp_1', 'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                   'Hidraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                   'Trafo_3_wind_Temp']]
```

Nacelle\_Misalignment\_Avg\_Wind\_Dir variable is removed because of its significant value is more than 0.05.

```
In [31]: lr2 = lr.fit(x2_train,y_train)
```

```
In [32]: y_pred2 = lr2.predict(x2_test)
```

```
In [33]: import statsmodels.api as sm
x2_train_sm = x2_train

x2_train_sm = sm.add_constant(x2_train_sm)

mlm2 = sm.OLS(y_train,x2_train_sm).fit()

mlm2.params
print(mlm2.summary())
```

```

=====
                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.938
Model:                  OLS                  Adj. R-squared:            0.938
Method:                 Least Squares        F-statistic:              1.735e+05
Date:                  Mon, 10 May 2021      Prob (F-statistic):       0.00
Time:                  00:41:03              Log-Likelihood:           -1.2446e+06
No. Observations:      196159               AIC:                     2.489e+06
Df Residuals:          196141               BIC:                     2.489e+06
Df Model:              17
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-3124.4585	9.733	-321.013	0.000	-3143.535	-3105.382
Avg_Ambient_Temp	-4.7949	0.094	-51.016	0.000	-4.979	-4.611
Avg_Generator_Speed	0.0012	0.076	0.015	0.988	-0.149	0.151
Avg_Pitch_Angle	12.3884	0.045	275.462	0.000	12.300	12.477
Avg_Rotor_Speed	116.0305	8.176	14.191	0.000	100.005	132.056
Avg_Wind_Speed	32.1640	0.325	99.101	0.000	31.528	32.800
Bearing_DE_Temp	-9.4560	0.153	-61.711	0.000	-9.756	-9.156
Bearing_NDE_Temp	-8.7499	0.158	-55.401	0.000	-9.060	-8.440
Gearbox_bearing_Temp	2.9580	0.256	11.536	0.000	2.455	3.461
Gearbox_oil_Temp	26.9373	0.341	78.975	0.000	26.269	27.606
Generator_wind_Temp_1	116.9224	2.155	54.254	0.000	112.699	121.146
Generator_wind_Temp_2	-144.2520	2.126	-67.844	0.000	-148.419	-140.085
Generator_wind_Temp_3	34.7897	0.289	120.287	0.000	34.223	35.357
Generators_sliprings_Temp	14.6734	0.153	95.790	0.000	14.373	14.974
Hidraulic_group_pressure	0.2189	0.014	15.435	0.000	0.191	0.247
Trafo_1_wind_Temp	-1.9121	0.084	-22.696	0.000	-2.077	-1.747
Trafo_2_wind_Temp	0.4958	0.071	6.986	0.000	0.357	0.635
Trafo_3_wind_Temp	3.8223	0.074	51.673	0.000	3.677	3.967

```

=====
Omnibus:                 34340.512      Durbin-Watson:              1.998
Prob(Omnibus):            0.000      Jarque-Bera (JB):           263101.815
Skew:                     0.636      Prob(JB):                   0.00
Kurtosis:                 8.529      Cond. No.                   3.80e+04
=====

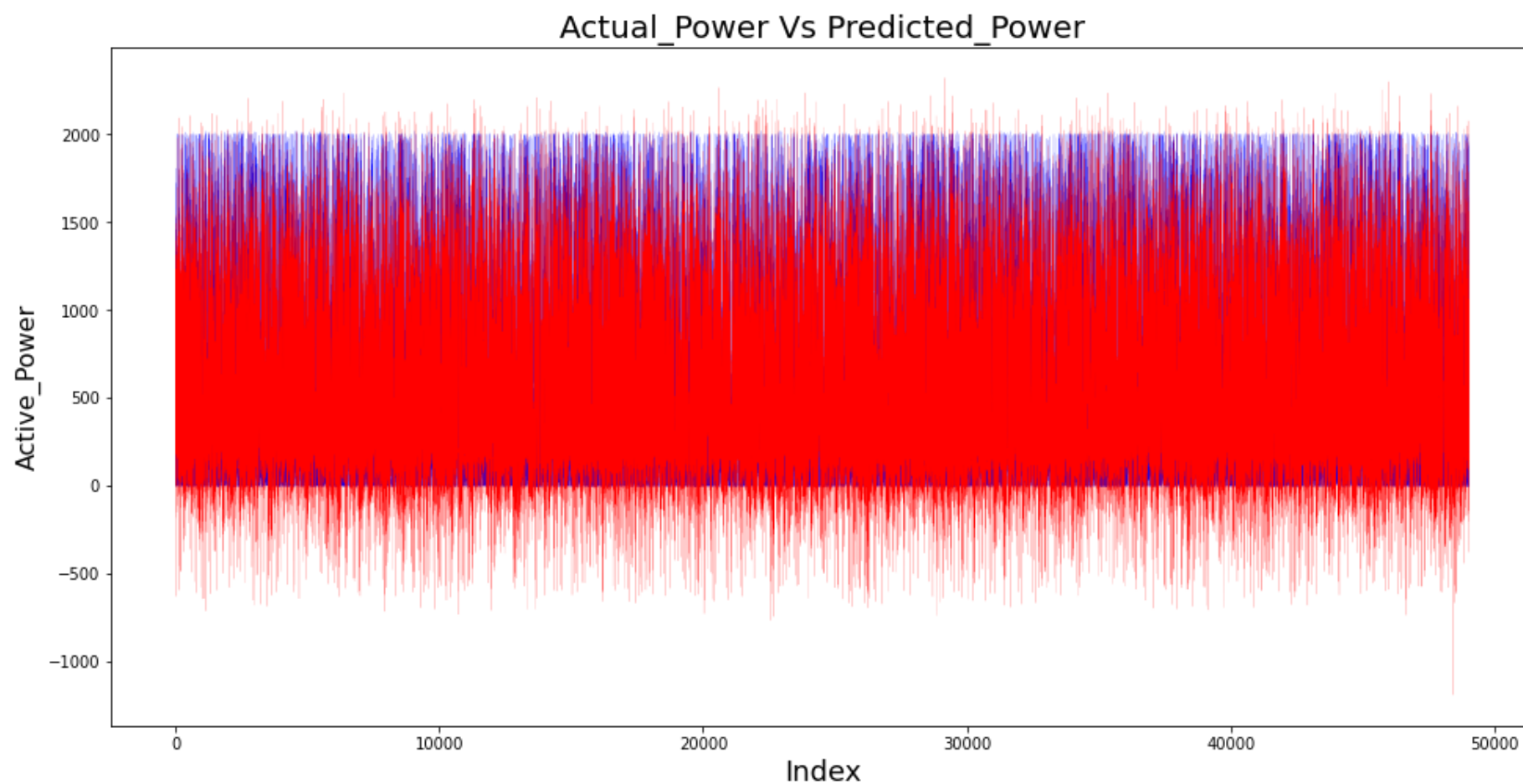
```

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 3.8e+04. This might indicate that there are strong multicollinearity or other numerical problems.



```
In [34]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred2, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



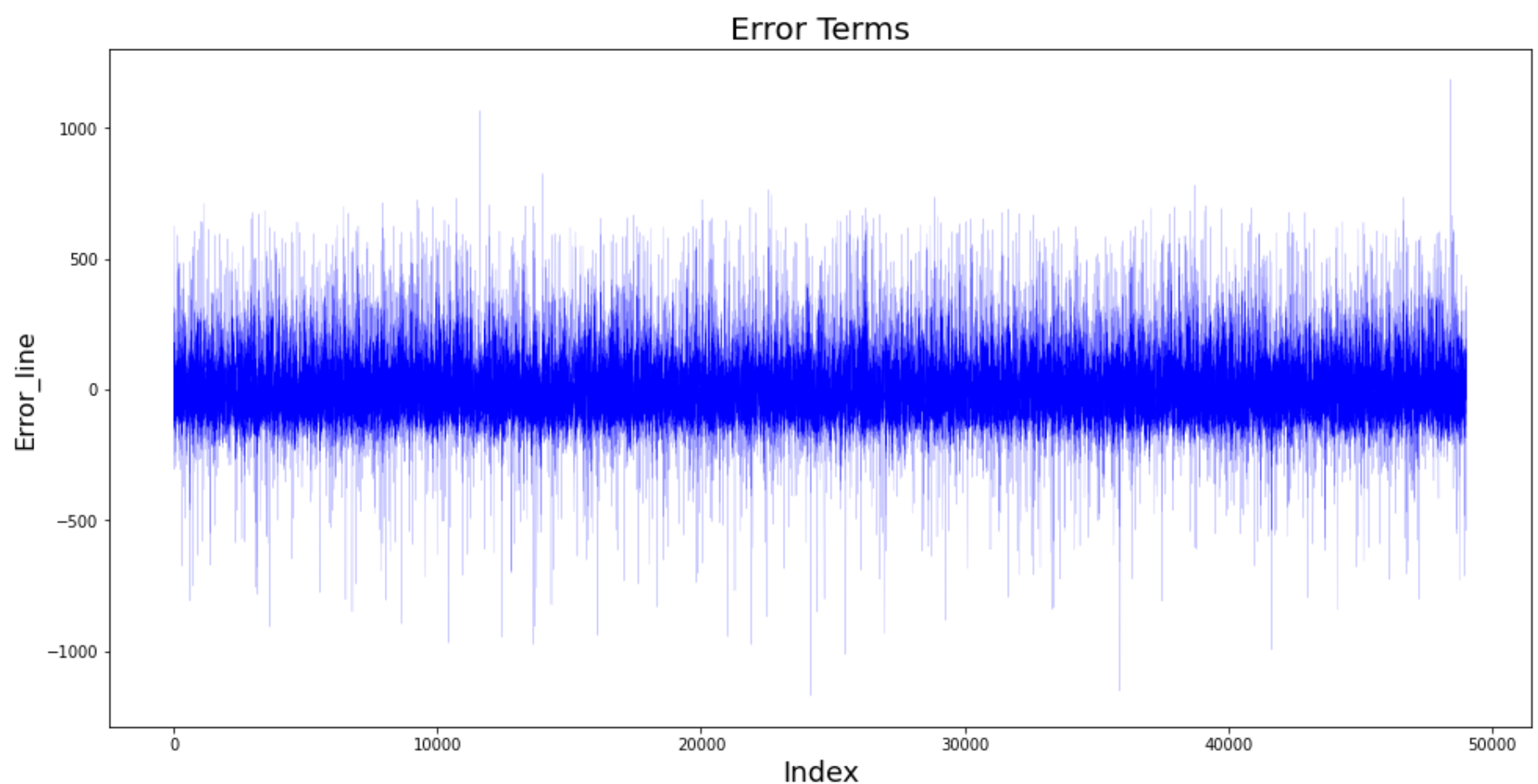
```
In [35]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x2_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x2_train.values, i)
                    for i in range(len(x2_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	69.958895
1	Avg_Generator_Speed	83497.589288
2	Avg_Pitch_Angle	13.609280
3	Avg_Rotor_Speed	84131.728964
4	Avg_Wind_Speed	44.178478
5	Bearing_DE_Temp	615.044433
6	Bearing_NDE_Temp	669.020778
7	Gearbox_bearing_Temp	2755.936095
8	Gearbox_oil_Temp	2657.976214
9	Generator_wind_Temp_1	231150.900292
10	Generator_wind_Temp_2	231849.033783
11	Generator_wind_Temp_3	4567.485073
12	Generators_sliprings_Temp	442.858267
13	Hidraulic_group_pressure	81.452678
14	Trafo_1_wind_Temp	311.202159
15	Trafo_2_wind_Temp	273.080192
16	Trafo_3_wind_Temp	285.172116

```
In [36]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred2, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)               # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [37]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred2)
r_squared = r2_score(y_test, y_pred2)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 18889.494881377803

r\_square\_value : 0.938 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-4

```
In [38]: x3_train = x_train[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed',
                             'Avg_Wind_Speed', 'Bearing_DE_Temp', 'Bearing_NDE_Temp', 'Gearbox_bearing_Temp', 'Gearbox_oil_Temp',
                             'Generator_wind_Temp_1', 'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                             'Hidraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                             'Trafo_3_wind_Temp']]
x3_test = x_test[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed',
                   'Avg_Wind_Speed', 'Bearing_DE_Temp', 'Bearing_NDE_Temp', 'Gearbox_bearing_Temp', 'Gearbox_oil_Temp',
                   'Generator_wind_Temp_1', 'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                   'Hidraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                   'Trafo_3_wind_Temp']]
```

'Avg\_Generator\_Speed' variable is removed because of its significant value is more than 0.05.

```
In [39]: lr3 = lr.fit(x3_train,y_train)
```

```
In [40]: y_pred3 = lr3.predict(x3_test)
```

```
In [41]: import statsmodels.api as sm
x3_train_sm = x3_train

x3_train_sm = sm.add_constant(x3_train_sm)

mlm3 = sm.OLS(y_train,x3_train_sm).fit()

mlm3.params
print(mlm3.summary())
```

#### OLS Regression Results

```
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.938
Model:                  OLS                  Adj. R-squared:           0.938
Method:                 Least Squares        F-statistic:             1.844e+05
Date:                  Mon, 10 May 2021      Prob (F-statistic):       0.00
Time:                  00:41:14              Log-Likelihood:          -1.2446e+06
No. Observations:      196159               AIC:                    2.489e+06
Df Residuals:          196142               BIC:                    2.489e+06
Df Model:              16
Covariance Type:       nonrobust
=====
```

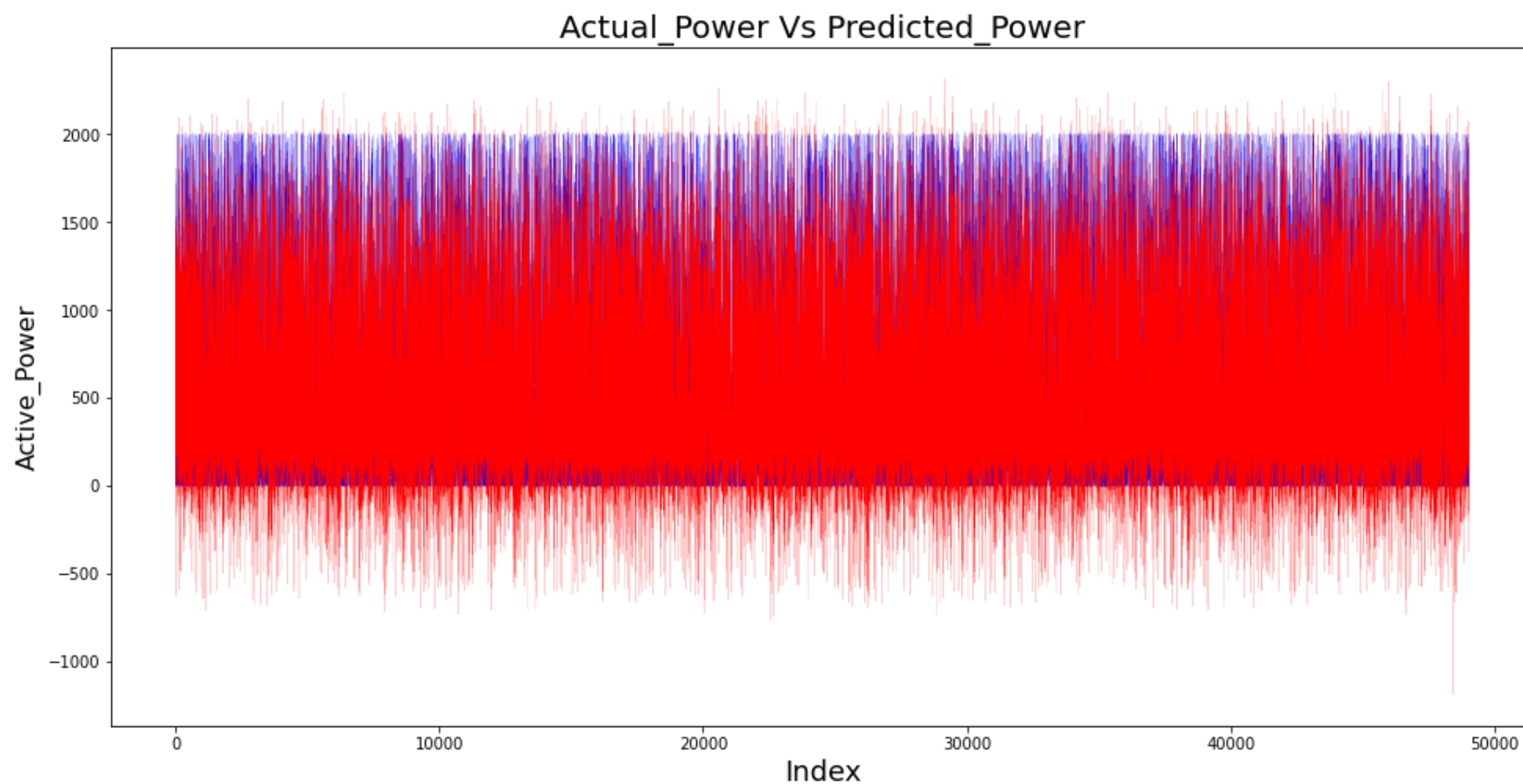
	coef	std err	t	P> t	[0.025	0.975]
const	-3124.4575	9.733	-321.021	0.000	-3143.534	-3105.381
Avg_Ambient_Temp	-4.7949	0.094	-51.025	0.000	-4.979	-4.611
Avg_Pitch_Angle	12.3884	0.045	275.466	0.000	12.300	12.477
Avg_Rotor_Speed	116.1561	0.332	349.570	0.000	115.505	116.807
Avg_Wind_Speed	32.1635	0.323	99.522	0.000	31.530	32.797
Bearing_DE_Temp	-9.4560	0.153	-61.714	0.000	-9.756	-9.156
Bearing_NDE_Temp	-8.7500	0.158	-55.406	0.000	-9.060	-8.440
Gearbox_bearing_Temp	2.9578	0.256	11.550	0.000	2.456	3.460
Gearbox_oil_Temp	26.9375	0.341	79.005	0.000	26.269	27.606
Generator_wind_Temp_1	116.9223	2.155	54.255	0.000	112.698	121.146
Generator_wind_Temp_2	-144.2517	2.126	-67.845	0.000	-148.419	-140.084
Generator_wind_Temp_3	34.7897	0.289	120.288	0.000	34.223	35.357
Generators_sliprings_Temp	14.6733	0.153	95.838	0.000	14.373	14.973
Hidraulic_group_pressure	0.2189	0.014	15.436	0.000	0.191	0.247
Trafo_1_wind_Temp	-1.9121	0.084	-22.696	0.000	-2.077	-1.747
Trafo_2_wind_Temp	0.4958	0.071	6.989	0.000	0.357	0.635
Trafo_3_wind_Temp	3.8223	0.074	51.687	0.000	3.677	3.967

```
=====
Omnibus:                34340.835      Durbin-Watson:           1.998
Prob(Omnibus):           0.000        Jarque-Bera (JB):        263102.847
Skew:                   0.636        Prob(JB):                0.00
Kurtosis:               8.529        Cond. No.                9.34e+03
=====
```

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.  
 [2] The condition number is large, 9.34e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [42]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred3, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



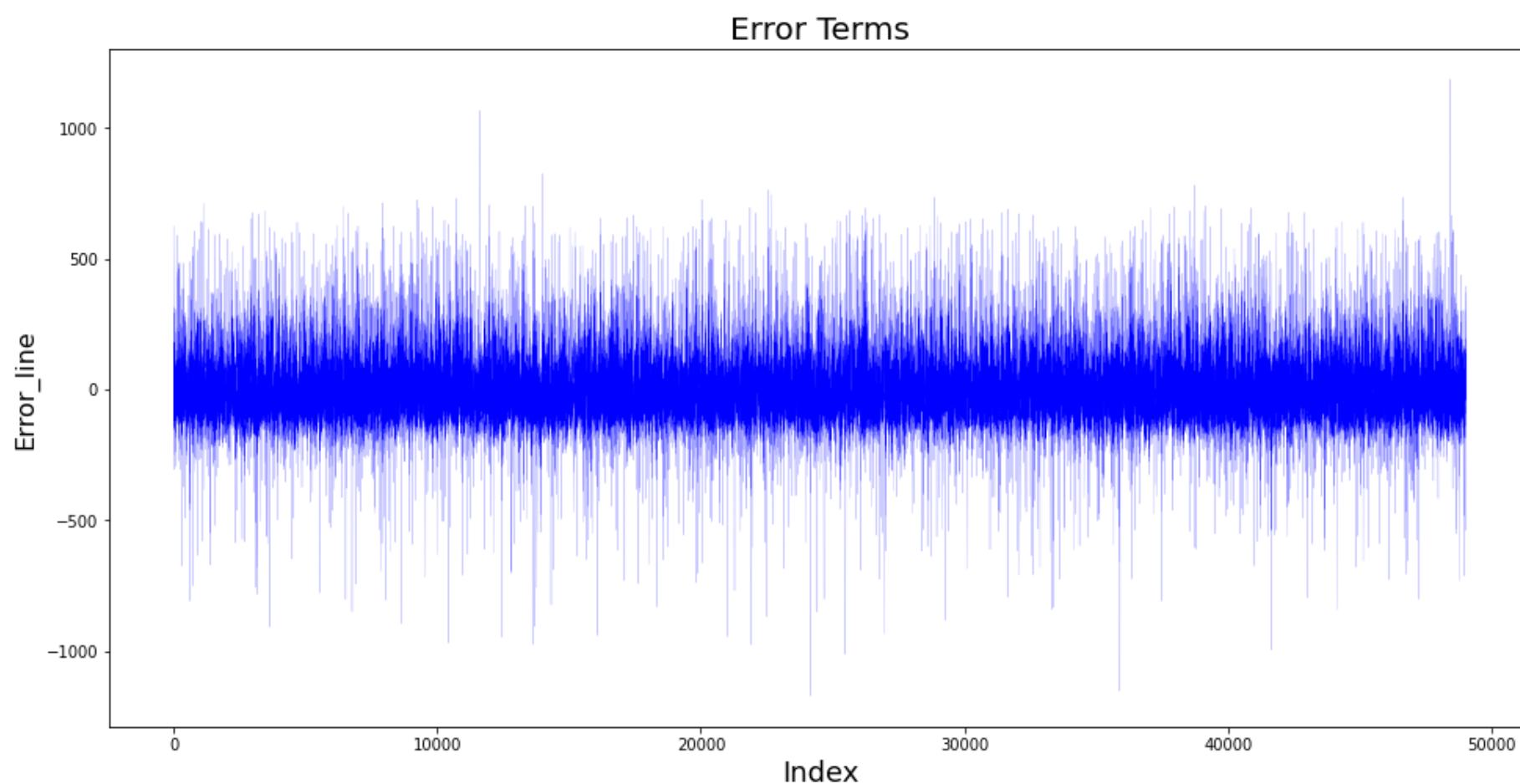
```
In [43]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x3_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x3_train.values, i)
                    for i in range(len(x3_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	69.934265
1	Avg_Pitch_Angle	13.608803
2	Avg_Rotor_Speed	137.958877
3	Avg_Wind_Speed	43.805196
4	Bearing_DE_Temp	614.971703
5	Bearing_NDE_Temp	668.889859
6	Gearbox_bearing_Temp	2748.459206
7	Gearbox_oil_Temp	2653.258020
8	Generator_wind_Temp_1	231139.296838
9	Generator_wind_Temp_2	231830.976611
10	Generator_wind_Temp_3	4567.470518
11	Generators_sliprings_Temp	442.432227
12	Hidraulic_group_pressure	81.429600
13	Trafo_1_wind_Temp	311.197941
14	Trafo_2_wind_Temp	272.863521
15	Trafo_3_wind_Temp	285.012433

```
In [44]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred3, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [45]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred3)
r_squared = r2_score(y_test, y_pred3)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 18889.474403181845

r\_square\_value : 0.938 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-5

```
In [46]: x4_train = x_train[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                             'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1',
                             'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                             'Hidraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                             'Trafo_3_wind_Temp']]
x4_test = x_test[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                   'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1',
                   'Generator_wind_Temp_2', 'Generator_wind_Temp_3', 'Generators_sliprings_Temp',
                   'Hidraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp',
                   'Trafo_3_wind_Temp']]
```

'Bearing\_NDE\_Temp' & 'Bearing\_DE\_Temp' These variables are Auto Correlated each other. So, i removed one variable ('Bearing\_NDE\_Temp')

```
In [47]: lr4 = lr.fit(x4_train,y_train)
```

```
In [48]: y_pred4 = lr4.predict(x4_test)
```

```
In [49]: import statsmodels.api as sm
x4_train_sm = x4_train

x4_train_sm = sm.add_constant(x4_train_sm)

mlm4 = sm.OLS(y_train,x4_train_sm).fit()

mlm4.params
print(mlm4.summary())
```

```

                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.937
Model:                  OLS                  Adj. R-squared:           0.937
Method:                 Least Squares        F-statistic:              1.934e+05
Date:                   Mon, 10 May 2021      Prob (F-statistic):       0.00
Time:                   00:41:23             Log-Likelihood:          -1.2461e+06
No. Observations:       196159              AIC:                     2.492e+06
Df Residuals:           196143              BIC:                     2.492e+06
Df Model:               15
Covariance Type:        nonrobust
=====
                        coef      std err          t      P>|t|      [0.025      0.975]
-----
const                -3052.2801      9.720    -314.006      0.000    -3071.332    -3033.228
Avg_Ambient_Temp        -4.7923      0.095    -50.603      0.000      -4.978      -4.607
Avg_Pitch_Angle         12.3403      0.045    272.326      0.000      12.252      12.429
Avg_Rotor_Speed        116.4816      0.335    347.893      0.000     115.825     117.138
Avg_Wind_Speed          32.7520      0.326    100.614      0.000      32.114      33.390
Bearing_DE_Temp        -14.3852      0.126   -114.421      0.000     -14.632     -14.139
Gearbox_bearing_Temp     2.7047      0.258     10.482      0.000       2.199       3.211
Gearbox_oil_Temp        25.0503      0.342     73.269      0.000      24.380      25.720
Generator_wind_Temp_1    117.0581      2.172     53.898      0.000     112.801     121.315
Generator_wind_Temp_2   -144.9050      2.143    -67.627      0.000    -149.105    -140.705
Generator_wind_Temp_3    34.3993      0.291    118.053      0.000      33.828      34.970
Generators_sliprings_Temp 12.7109      0.150     84.675      0.000      12.417      13.005
Hidraulic_group_pressure  0.2117      0.014     14.814      0.000       0.184       0.240
Trafo_1_wind_Temp       -1.3993      0.084    -16.582      0.000      -1.565      -1.234
Trafo_2_wind_Temp        0.5641      0.071      7.891      0.000       0.424       0.704
Trafo_3_wind_Temp        3.3687      0.074     45.481      0.000       3.224       3.514
=====
Omnibus:                32465.610    Durbin-Watson:            1.998
Prob(Omnibus):           0.000    Jarque-Bera (JB):         253064.039
Skew:                    0.582    Prob(JB):                  0.00
Kurtosis:                8.441    Cond. No.                  9.11e+03
=====

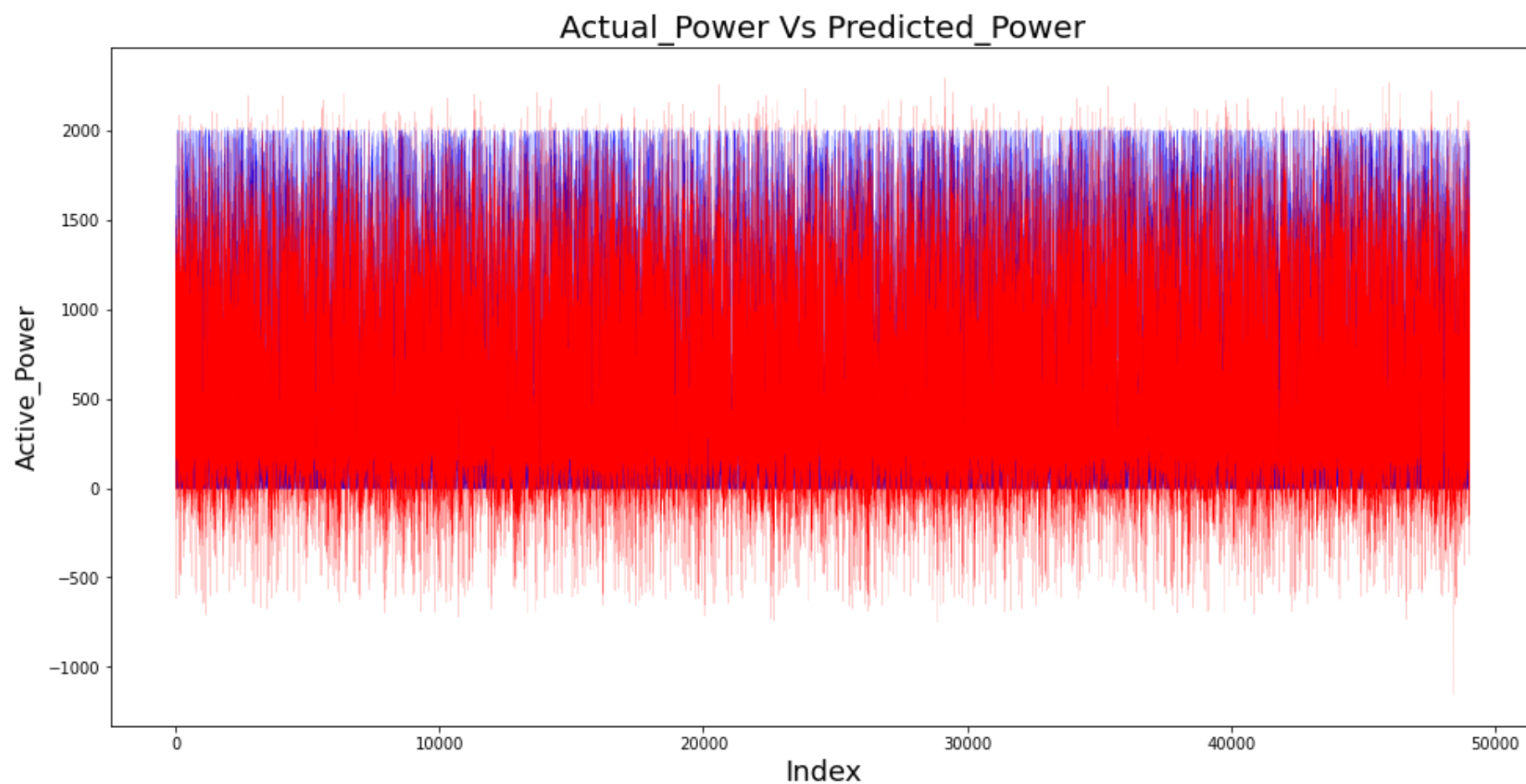
```

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 9.11e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```
In [50]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred4, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



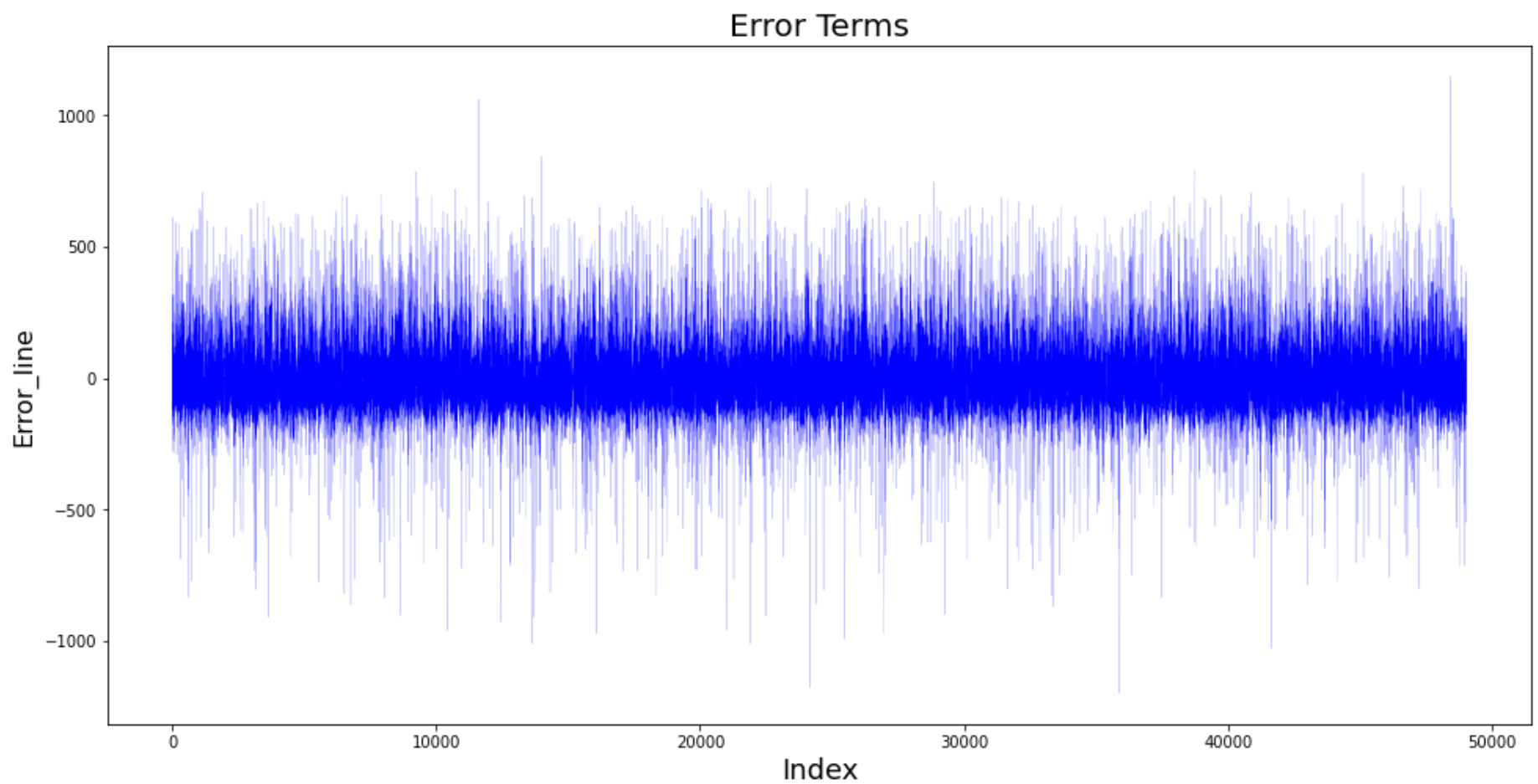
```
In [51]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x4_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x4_train.values, i)
                    for i in range(len(x4_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	69.931381
1	Avg_Pitch_Angle	13.608564
2	Avg_Rotor_Speed	137.953312
3	Avg_Wind_Speed	43.748536
4	Bearing_DE_Temp	411.876170
5	Gearbox_bearing_Temp	2742.710813
6	Gearbox_oil_Temp	2652.418117
7	Generator_wind_Temp_1	230826.581554
8	Generator_wind_Temp_2	231643.128321
9	Generator_wind_Temp_3	4566.958546
10	Generators_sliprings_Temp	421.338858
11	Hidraulic_group_pressure	81.347775
12	Trafo_1_wind_Temp	306.452596
13	Trafo_2_wind_Temp	272.843233
14	Trafo_3_wind_Temp	281.161026

```
In [52]: c = [i for i in range(1,49041,1)]
```

```
plt.plot(c,y_test-y_pred4, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [53]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred4)
r_squared = r2_score(y_test, y_pred4)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 19200.959868118574

r\_square\_value : 0.937 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-6

```
In [54]: x5_train = x_train[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                             'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1',
                             'Generator_wind_Temp_3', 'Generators_sliprings_Temp', 'Hidraulic_group_pressure', 'Trafo_1_wind_Temp',
                             'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
x5_test = x_test[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                   'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1',
                   'Generator_wind_Temp_3', 'Generators_sliprings_Temp', 'Hidraulic_group_pressure', 'Trafo_1_wind_Temp',
                   'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
```

'Generator\_wind\_Temp\_1', 'Generator\_wind\_Temp\_2' These variables are Auto Correlated each other. So, i removed one variable ('Generator\_wind\_Temp\_2')

```
In [55]: lr5 = lr.fit(x5_train,y_train)
```

```
In [56]: y_pred5 = lr5.predict(x5_test)
```

```
In [57]: import statsmodels.api as sm
x5_train_sm = x5_train

x5_train_sm = sm.add_constant(x5_train_sm)

mlm5 = sm.OLS(y_train,x5_train_sm).fit()

mlm5.params
print(mlm5.summary())
```

```

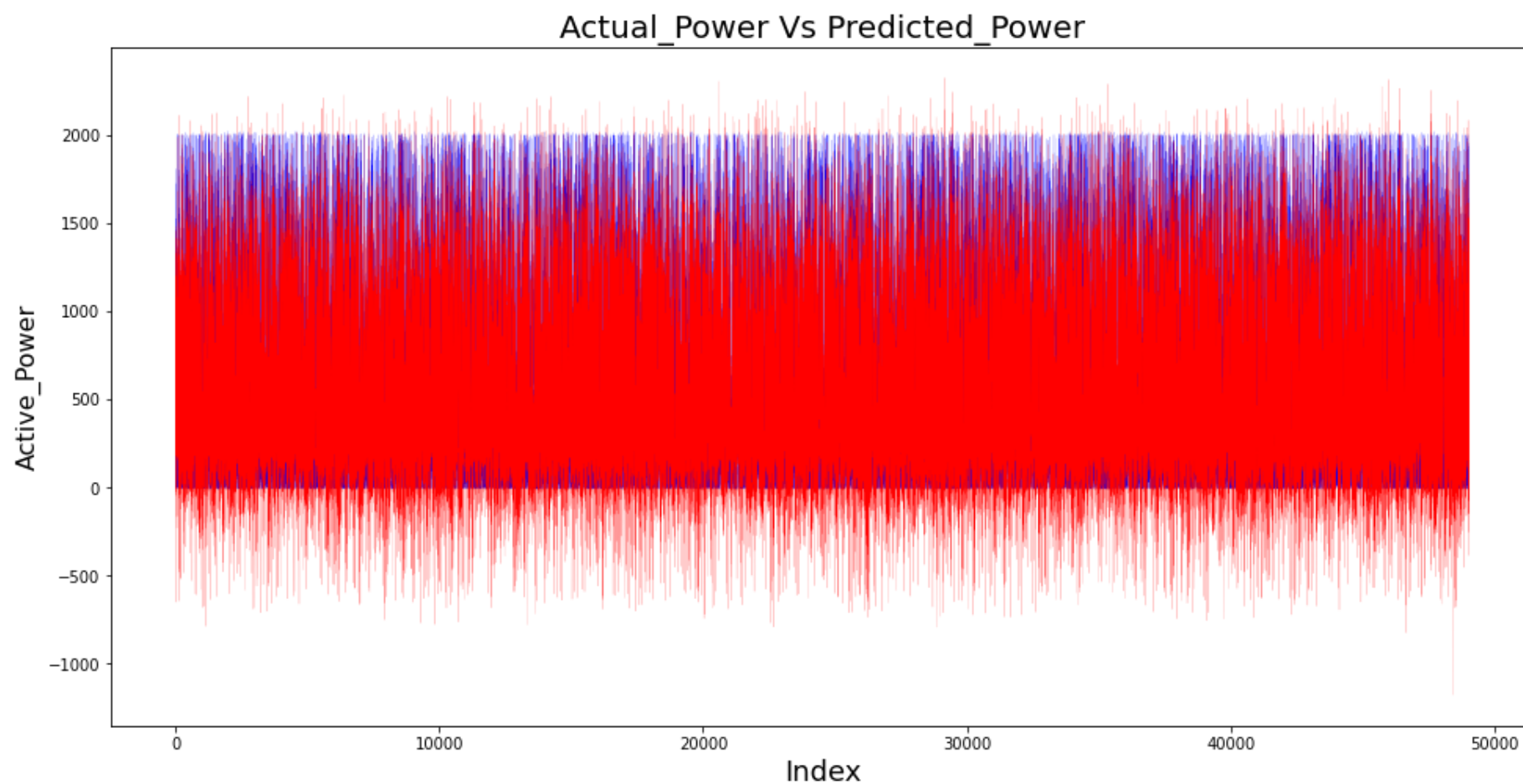
                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.935
Model:                  OLS                  Adj. R-squared:           0.935
Method:                 Least Squares        F-statistic:              2.022e+05
Date:                  Mon, 10 May 2021      Prob (F-statistic):       0.00
Time:                  00:41:31              Log-Likelihood:           -1.2484e+06
No. Observations:      196159               AIC:                     2.497e+06
Df Residuals:          196144               BIC:                     2.497e+06
Df Model:              14
Covariance Type:       nonrobust
=====
                        coef      std err      t      P>|t|      [0.025      0.975]
-----
const                -3214.7902      9.528    -337.409    0.000    -3233.465    -3196.116
Avg_Ambient_Temp      -5.2072      0.096    -54.469    0.000      -5.395      -5.020
Avg_Pitch_Angle       13.5097      0.042    318.841    0.000      13.427      13.593
Avg_Rotor_Speed      116.8875      0.339    345.162    0.000     116.224     117.551
Avg_Wind_Speed        38.0114      0.320    118.876    0.000      37.385      38.638
Bearing_DE_Temp      -13.9962      0.127   -110.167    0.000     -14.245     -13.747
Gearbox_bearing_Temp   1.1050      0.260     4.251    0.000       0.596       1.614
Gearbox_oil_Temp      26.3377      0.345    76.270    0.000      25.661      27.014
Generator_wind_Temp_1 -28.4262      0.302   -94.229    0.000     -29.017     -27.835
Generator_wind_Temp_3  34.8105      0.295   118.121    0.000      34.233      35.388
Generators_sliprings_Temp 13.4166      0.151    88.567    0.000      13.120      13.714
Hidraulic_group_pressure 0.2348      0.014    16.246    0.000       0.206       0.263
Trafo_1_wind_Temp     -2.0116      0.085   -23.700    0.000      -2.178      -1.845
Trafo_2_wind_Temp      0.9973      0.072    13.847    0.000       0.856       1.139
Trafo_3_wind_Temp      3.7910      0.075    50.777    0.000       3.645       3.937
=====
Omnibus:              38016.002    Durbin-Watson:            2.000
Prob(Omnibus):        0.000    Jarque-Bera (JB):        287712.537
Skew:                 0.732    Prob(JB):                 0.00
Kurtosis:             8.750    Cond. No.                 8.53e+03
=====

```

Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 8.53e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [58]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred5, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



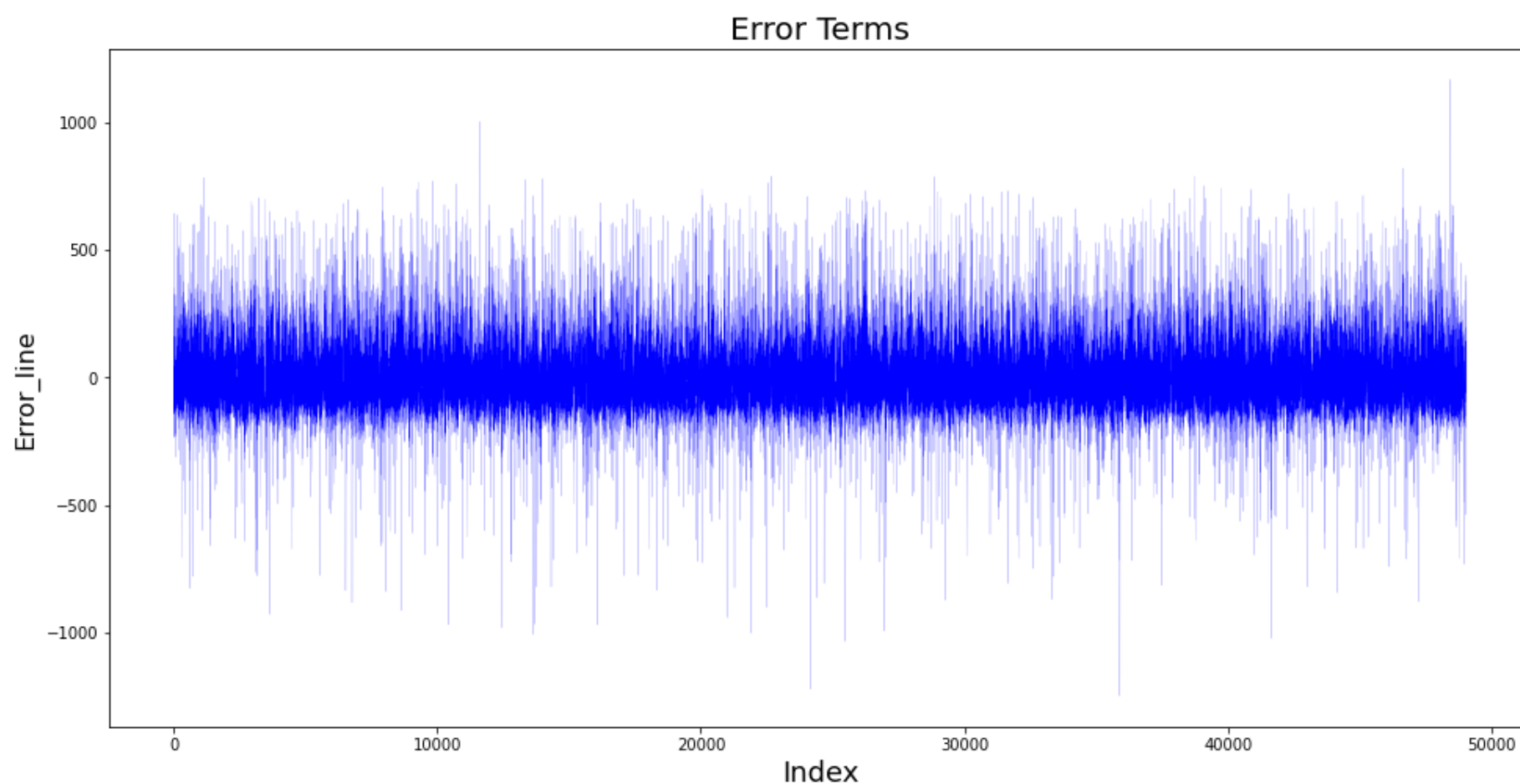
```
In [59]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x5_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x5_train.values, i)
                    for i in range(len(x5_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	69.503414
1	Avg_Pitch_Angle	11.769708
2	Avg_Rotor_Speed	137.732875
3	Avg_Wind_Speed	41.224769
4	Bearing_DE_Temp	411.291111
5	Gearbox_bearing_Temp	2737.457889
6	Gearbox_oil_Temp	2601.629653
7	Generator_wind_Temp_1	4492.552730
8	Generator_wind_Temp_3	4566.868338
9	Generators_sliprings_Temp	420.397298
10	Hidraulic_group_pressure	81.121386
11	Trafo_1_wind_Temp	300.566284
12	Trafo_2_wind_Temp	269.594255
13	Trafo_3_wind_Temp	278.373809

```
In [60]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred5, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [61]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred5)
r_squared = r2_score(y_test, y_pred5)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 19596.03455227658

r\_square\_value : 0.935 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-7

```
In [62]: x6_train = x_train[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                             'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1', 'Generators_sliprings_Temp',
                             'Hydraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
x6_test = x_test[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                  'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1', 'Generators_sliprings_Temp',
                  'Hydraulic_group_pressure', 'Trafo_1_wind_Temp', 'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
```

'Generator\_wind\_Temp\_1', 'Generator\_wind\_Temp\_3' These variables are Auto Correlated each other. So, i removed one variable ('Generator\_wind\_Temp\_3')

```
In [63]: lr6 = lr.fit(x6_train,y_train)
```

```
In [64]: y_pred6 = lr6.predict(x6_test)
```

```
In [65]: import statsmodels.api as sm
x6_train_sm = x6_train

x6_train_sm = sm.add_constant(x6_train_sm)

mlm6 = sm.OLS(y_train,x6_train_sm).fit()

mlm6.params
print(mlm6.summary())
```

#### OLS Regression Results

```
=====
Dep. Variable:      Avg_Active_Power      R-squared:      0.931
Model:              OLS                   Adj. R-squared:  0.931
Method:             Least Squares         F-statistic:     2.023e+05
Date:               Mon, 10 May 2021       Prob (F-statistic): 0.00
Time:               00:41:37              Log-Likelihood:  -1.2551e+06
No. Observations:   196159                AIC:             2.510e+06
Df Residuals:       196145                BIC:             2.510e+06
Df Model:           13
Covariance Type:    nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-3102.5963	9.812	-316.211	0.000	-3121.827	-3083.365
Avg_Ambient_Temp	-6.5052	0.098	-66.187	0.000	-6.698	-6.313
Avg_Pitch_Angle	14.1497	0.043	325.337	0.000	14.064	14.235
Avg_Rotor_Speed	120.3763	0.349	344.772	0.000	119.692	121.061
Avg_Wind_Speed	36.8071	0.331	111.278	0.000	36.159	37.455
Bearing_DE_Temp	-14.6500	0.131	-111.524	0.000	-14.907	-14.393
Gearbox_bearing_Temp	-2.0452	0.268	-7.643	0.000	-2.570	-1.521
Gearbox_oil_Temp	27.2033	0.357	76.133	0.000	26.503	27.904
Generator_wind_Temp_1	5.8439	0.086	68.316	0.000	5.676	6.012
Generators_sliprings_Temp	16.2227	0.155	104.770	0.000	15.919	16.526
Hidraulic_group_pressure	0.2097	0.015	14.018	0.000	0.180	0.239
Trafo_1_wind_Temp	-2.4433	0.088	-27.840	0.000	-2.615	-2.271
Trafo_2_wind_Temp	1.2977	0.074	17.420	0.000	1.152	1.444
Trafo_3_wind_Temp	4.3404	0.077	56.280	0.000	4.189	4.492

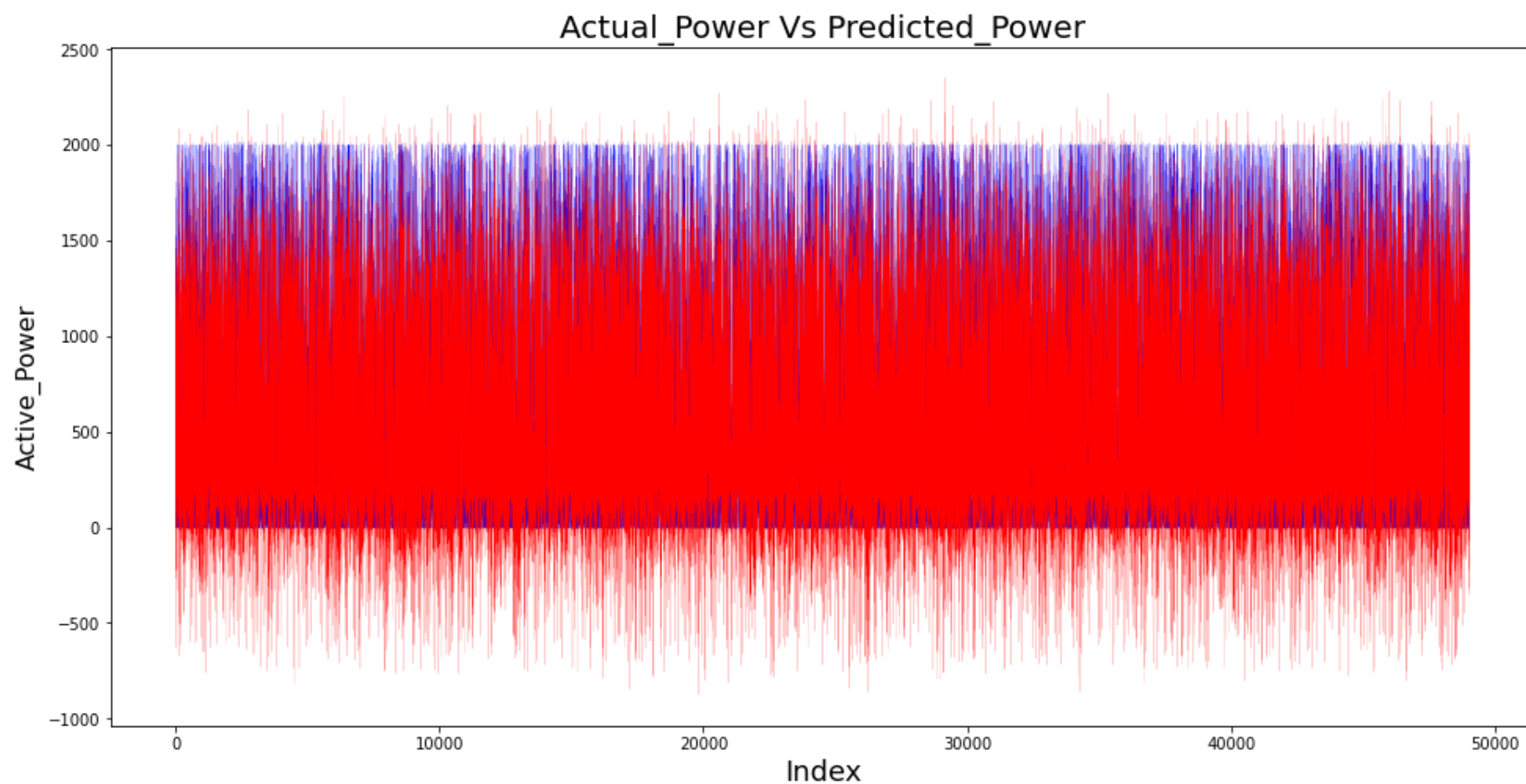
```
=====
Omnibus:      44311.715      Durbin-Watson:      2.000
Prob(Omnibus): 0.000      Jarque-Bera (JB):    302008.341
Skew:          0.917      Prob(JB):            0.00
Kurtosis:      8.795      Cond. No.             8.21e+03
=====
```

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 8.21e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```
In [66]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred6, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



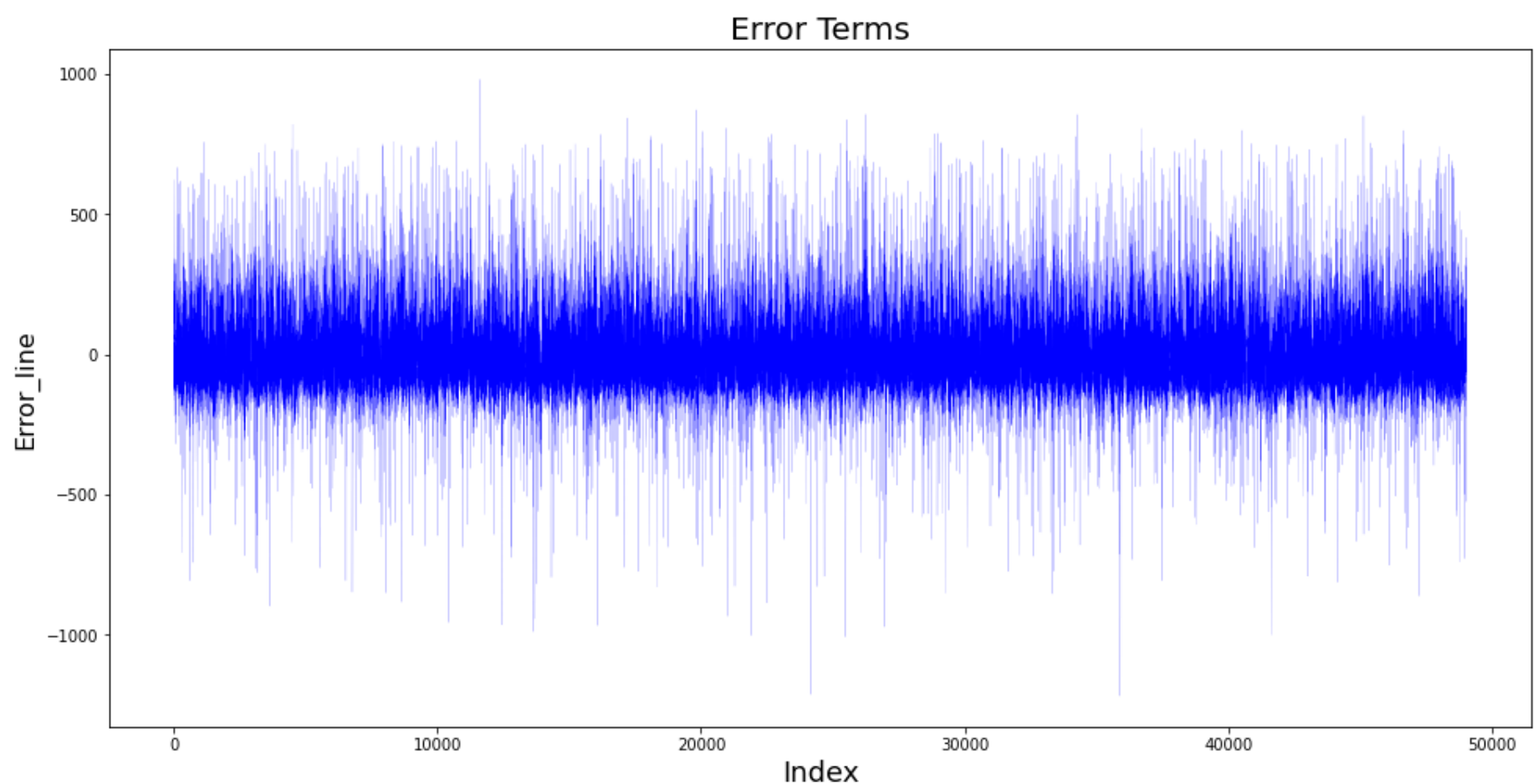
```
In [67]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x6_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x6_train.values, i)
                    for i in range(len(x6_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	68.670256
1	Avg_Pitch_Angle	11.569321
2	Avg_Rotor_Speed	136.876007
3	Avg_Wind_Speed	41.172345
4	Bearing_DE_Temp	410.596555
5	Gearbox_bearing_Temp	2695.365135
6	Gearbox_oil_Temp	2568.161972
7	Generator_wind_Temp_1	289.828345
8	Generators_sliprings_Temp	408.753682
9	Hidraulic_group_pressure	81.101311
10	Trafo_1_wind_Temp	300.300197
11	Trafo_2_wind_Temp	269.403294
12	Trafo_3_wind_Temp	277.515705

```
In [68]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred6, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [69]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred6)
r_squared = r2_score(y_test, y_pred6)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 20977.0031614993

r\_square\_value : 0.931 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-8

```
In [70]: x7_train = x_train[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                             'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1', 'Generators_sliprings_Temp',
                             'Hydraulic_group_pressure', 'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
x7_test = x_test[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                  'Gearbox_bearing_Temp', 'Gearbox_oil_Temp', 'Generator_wind_Temp_1', 'Generators_sliprings_Temp',
                  'Hydraulic_group_pressure', 'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
```

'Trafo\_1\_wind\_Temp', 'Trafo\_2\_wind\_Temp' These variables are Auto Correlated each other. So, i removed one variable ('Trafo\_1\_wind\_Temp')

```
In [71]: lr7 = lr.fit(x7_train,y_train)
```

```
In [72]: y_pred7 = lr7.predict(x7_test)
```

```
In [73]: import statsmodels.api as sm
x7_train_sm = x7_train

x7_train_sm = sm.add_constant(x7_train_sm)

mlm7 = sm.OLS(y_train,x7_train_sm).fit()

mlm7.params
print(mlm7.summary())
```

```

=====
                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.930
Model:                  OLS                  Adj. R-squared:           0.930
Method:                 Least Squares        F-statistic:             2.182e+05
Date:                  Mon, 10 May 2021      Prob (F-statistic):       0.00
Time:                  00:41:44              Log-Likelihood:          -1.2555e+06
No. Observations:      196159               AIC:                    2.511e+06
Df Residuals:          196146               BIC:                    2.511e+06
Df Model:              12
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-3139.2723	9.742	-322.237	0.000	-3158.367	-3120.178
Avg_Ambient_Temp	-6.8940	0.097	-70.721	0.000	-7.085	-6.703
Avg_Pitch_Angle	14.3159	0.043	331.652	0.000	14.231	14.400
Avg_Rotor_Speed	121.1415	0.349	347.360	0.000	120.458	121.825
Avg_Wind_Speed	36.6781	0.331	110.681	0.000	36.029	37.328
Bearing_DE_Temp	-14.3146	0.131	-109.217	0.000	-14.571	-14.058
Gearbox_bearing_Temp	-1.6877	0.268	-6.302	0.000	-2.213	-1.163
Gearbox_oil_Temp	27.1097	0.358	75.726	0.000	26.408	27.811
Generator_wind_Temp_1	6.0710	0.085	71.155	0.000	5.904	6.238
Generators_sliprings_Temp	16.4537	0.155	106.206	0.000	16.150	16.757
Hidraulic_group_pressure	0.2275	0.015	15.196	0.000	0.198	0.257
Trafo_2_wind_Temp	-0.2415	0.050	-4.828	0.000	-0.340	-0.143
Trafo_3_wind_Temp	3.3085	0.068	48.824	0.000	3.176	3.441

```

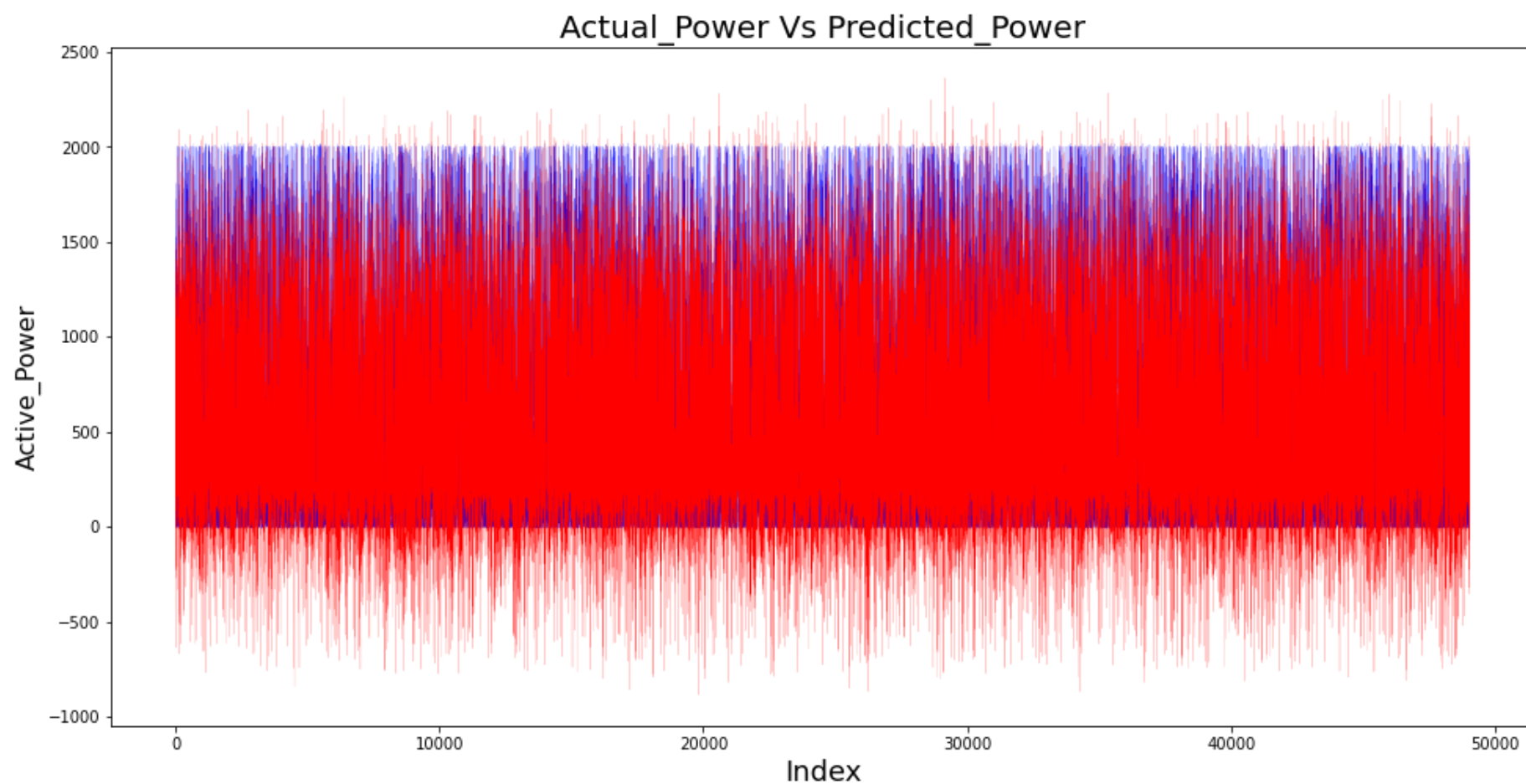
=====
Omnibus:                45147.168      Durbin-Watson:           1.999
Prob(Omnibus):           0.000        Jarque-Bera (JB):        312774.572
Skew:                   0.932         Prob(JB):                0.00
Kurtosis:               8.898         Cond. No.                7.91e+03
=====

```

Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 7.91e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [74]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred7, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



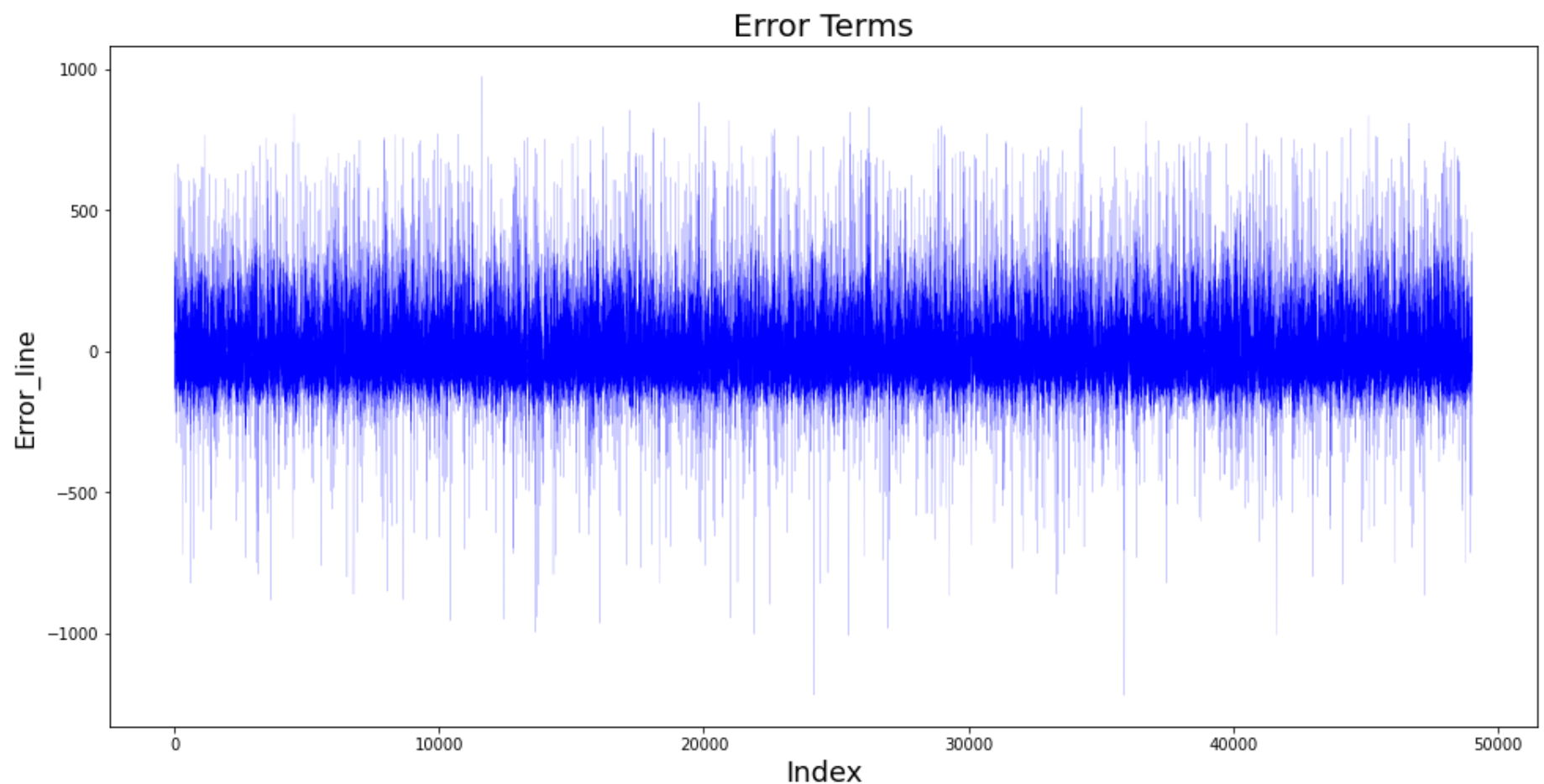
```
In [75]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x7_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x7_train.values, i)
                    for i in range(len(x7_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	67.113971
1	Avg_Pitch_Angle	11.360998
2	Avg_Rotor_Speed	135.742639
3	Avg_Wind_Speed	41.169139
4	Bearing_DE_Temp	407.308201
5	Gearbox_bearing_Temp	2679.419095
6	Gearbox_oil_Temp	2524.926123
7	Generator_wind_Temp_1	281.508267
8	Generators_sliprings_Temp	408.074206
9	Hidraulic_group_pressure	81.100535
10	Trafo_2_wind_Temp	121.867959
11	Trafo_3_wind_Temp	214.358740

```
In [76]: c = [i for i in range(1,49041,1)]
```

```
plt.plot(c,y_test-y_pred7, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [77]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred7)
r_squared = r2_score(y_test, y_pred7)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 21034.474288578378

r\_square\_value : 0.93 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-9

```
In [78]: x8_train = x_train[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                             'Gearbox_oil_Temp', 'Generator_wind_Temp_1', 'Generators_sliprings_Temp',
                             'Hydraulic_group_pressure', 'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
x8_test = x_test[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                   'Gearbox_oil_Temp', 'Generator_wind_Temp_1', 'Generators_sliprings_Temp',
                   'Hydraulic_group_pressure', 'Trafo_2_wind_Temp', 'Trafo_3_wind_Temp']]
```

'Gearbox\_bearing\_Temp', 'Gearbox\_oil\_Temp' These variables are Auto Correlated each other. So, i removed one variable ('Gearbox\_bearing\_Temp')

```
In [79]: lr8 = lr.fit(x8_train,y_train)
```

```
In [80]: y_pred8 = lr8.predict(x8_test)
```

```
In [81]: import statsmodels.api as sm
x8_train_sm = x8_train

x8_train_sm = sm.add_constant(x8_train_sm)

mlm8 = sm.OLS(y_train,x8_train_sm).fit()

mlm8.params
print(mlm8.summary())
```

#### OLS Regression Results

```
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.930
Model:                  OLS                   Adj. R-squared:           0.930
Method:                 Least Squares         F-statistic:              2.380e+05
Date:                   Mon, 10 May 2021       Prob (F-statistic):       0.00
Time:                   00:41:49              Log-Likelihood:           -1.2555e+06
No. Observations:       196159                AIC:                     2.511e+06
Df Residuals:           196147                BIC:                     2.511e+06
Df Model:               11
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-3126.5294	9.531	-328.042	0.000	-3145.210	-3107.849
Avg_Ambient_Temp	-6.8243	0.097	-70.454	0.000	-7.014	-6.634
Avg_Pitch_Angle	14.3342	0.043	332.803	0.000	14.250	14.419
Avg_Rotor_Speed	120.5063	0.334	360.910	0.000	119.852	121.161
Avg_Wind_Speed	36.9414	0.329	112.361	0.000	36.297	37.586
Bearing_DE_Temp	-14.3849	0.131	-110.143	0.000	-14.641	-14.129
Gearbox_oil_Temp	25.3162	0.217	116.567	0.000	24.890	25.742
Generator_wind_Temp_1	6.0969	0.085	71.535	0.000	5.930	6.264
Generators_sliprings_Temp	16.2095	0.150	108.055	0.000	15.915	16.504
Hidraulic_group_pressure	0.2131	0.015	14.399	0.000	0.184	0.242
Trafo_2_wind_Temp	-0.2131	0.050	-4.277	0.000	-0.311	-0.115
Trafo_3_wind_Temp	3.3676	0.067	50.175	0.000	3.236	3.499

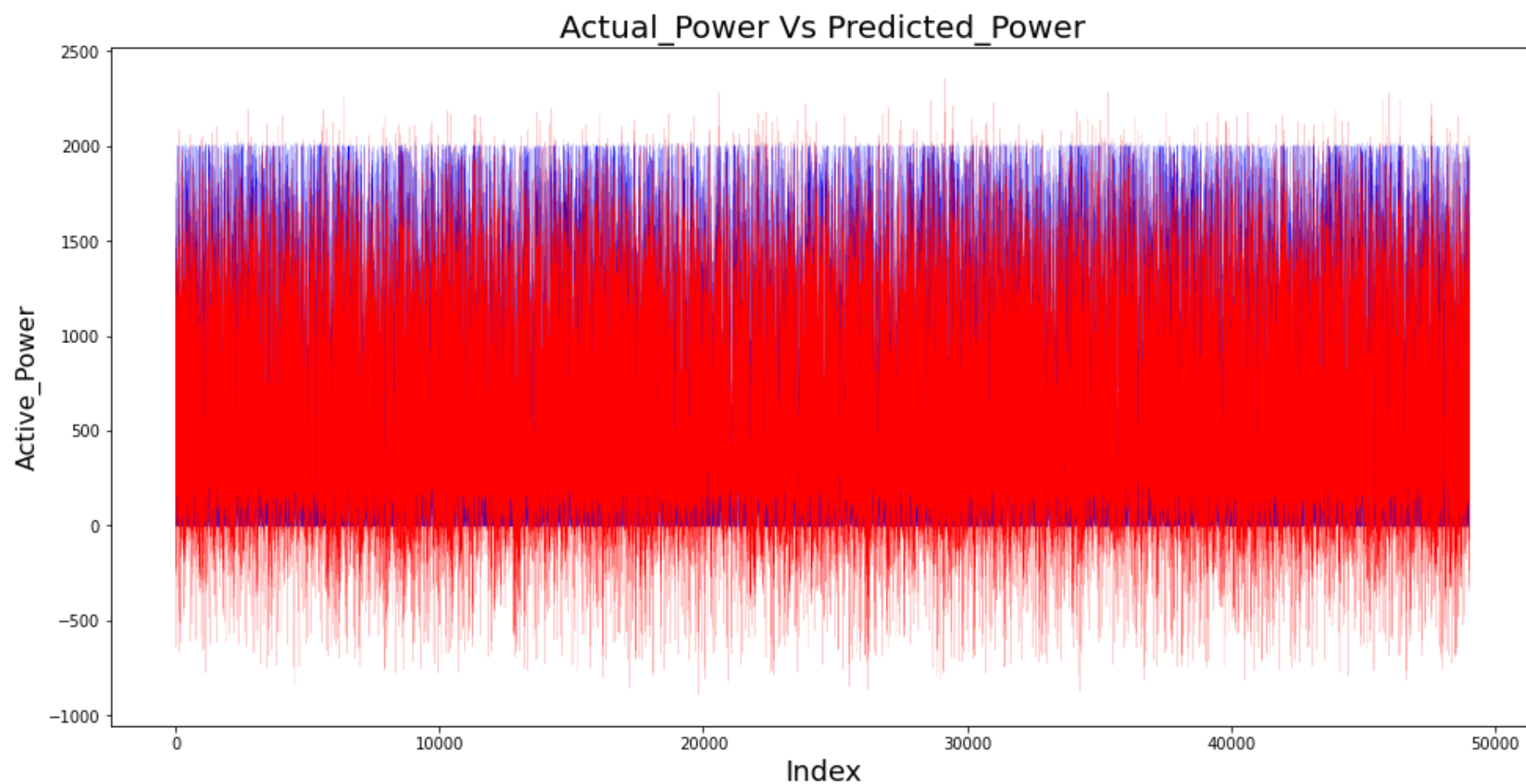
```
=====
Omnibus:                44892.103      Durbin-Watson:            1.999
Prob(Omnibus):           0.000        Jarque-Bera (JB):         317595.110
Skew:                    0.920        Prob(JB):                 0.00
Kurtosis:                8.956        Cond. No.                 7.50e+03
=====
```

#### Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 7.5e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```
In [82]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred8, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



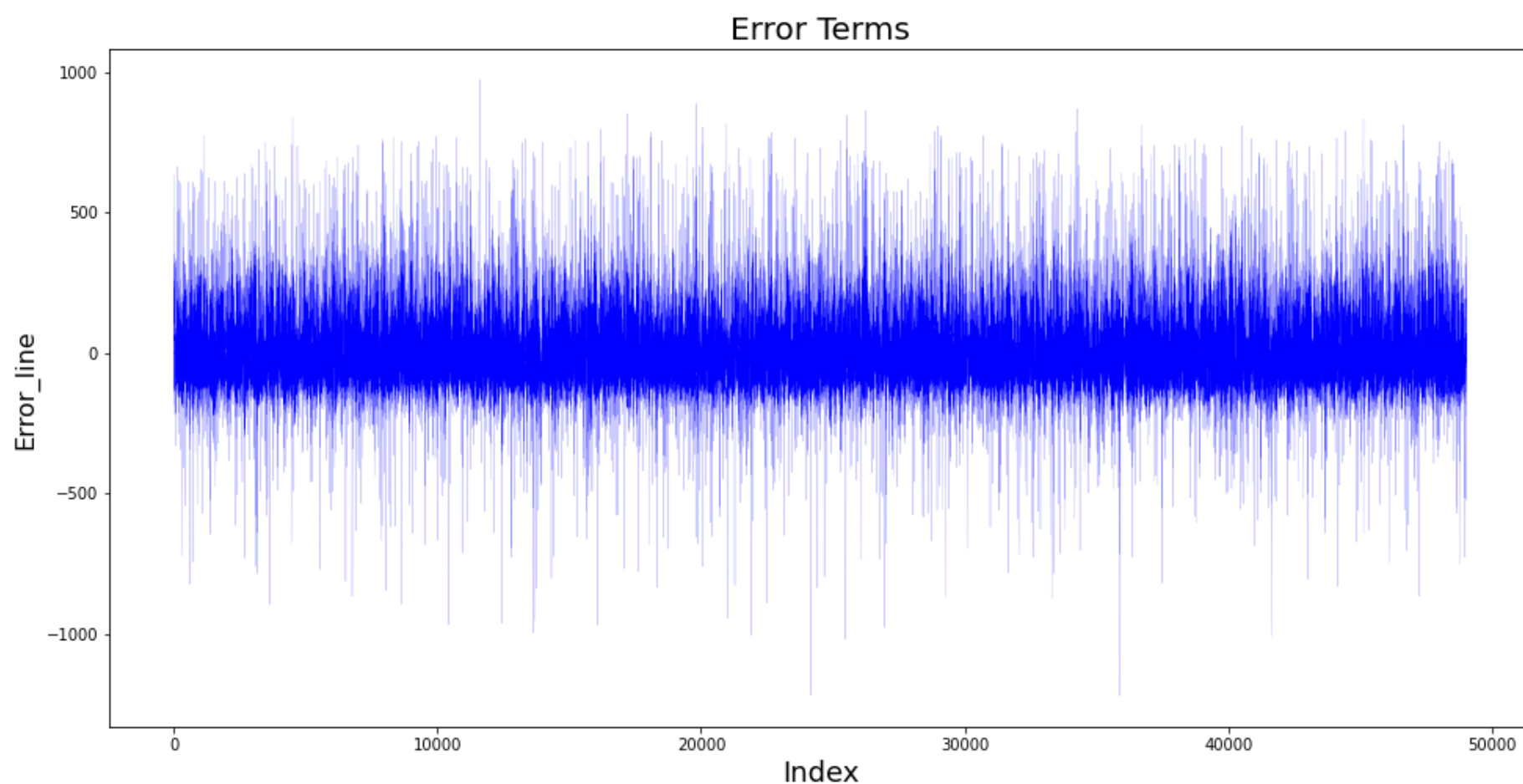
```
In [83]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x8_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x8_train.values, i)
                    for i in range(len(x8_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	65.952156
1	Avg_Pitch_Angle	11.302001
2	Avg_Rotor_Speed	122.102926
3	Avg_Wind_Speed	40.566746
4	Bearing_DE_Temp	404.377687
5	Gearbox_oil_Temp	495.460649
6	Generator_wind_Temp_1	280.816467
7	Generators_sliprings_Temp	385.166722
8	Hidraulic_group_pressure	80.314856
9	Trafo_2_wind_Temp	120.734064
10	Trafo_3_wind_Temp	210.089479

```
In [84]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred8, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)               # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [85]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred8)
r_squared = r2_score(y_test, y_pred8)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 21034.049515555536

r\_square\_value : 0.93 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-10

```
In [86]: x9_train = x_train[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                             'Generator_wind_Temp_1', 'Generators_sliprings_Temp', 'Gearbox_oil_Temp']]
x9_test = x_test[['Avg_Ambient_Temp', 'Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed', 'Bearing_DE_Temp',
                  'Generator_wind_Temp_1', 'Generators_sliprings_Temp', 'Gearbox_oil_Temp']]
```

'Hydraulic\_group\_pressure', 'Trafo\_2\_wind\_Temp', 'Trafo\_3\_wind\_Temp' These variables are Auto Correlated. So, i removed the variable

```
In [87]: lr9 = lr.fit(x9_train,y_train)
```

```
In [88]: y_pred9 = lr9.predict(x9_test)
```

```
In [89]: import statsmodels.api as sm
x9_train_sm = x9_train

x9_train_sm = sm.add_constant(x9_train_sm)

mlm9 = sm.OLS(y_train,x9_train_sm).fit()

mlm9.params
print(mlm9.summary())
```

```

=====
                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.929
Model:                  OLS                  Adj. R-squared:           0.929
Method:                 Least Squares        F-statistic:             3.206e+05
Date:                   Mon, 10 May 2021     Prob (F-statistic):      0.00
Time:                   00:41:54             Log-Likelihood:          -1.2574e+06
No. Observations:       196159              AIC:                    2.515e+06
Df Residuals:           196150              BIC:                    2.515e+06
Df Model:                8
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-3055.6816	9.234	-330.900	0.000	-3073.781	-3037.582
Avg_Ambient_Temp	-6.3651	0.096	-66.020	0.000	-6.554	-6.176
Avg_Pitch_Angle	14.3732	0.040	359.233	0.000	14.295	14.452
Avg_Rotor_Speed	120.9132	0.328	368.437	0.000	120.270	121.556
Avg_Wind_Speed	35.6474	0.326	109.257	0.000	35.008	36.287
Bearing_DE_Temp	-13.6919	0.131	-104.309	0.000	-13.949	-13.435
Generator_wind_Temp_1	6.7556	0.085	79.484	0.000	6.589	6.922
Generators_sliprings_Temp	16.0050	0.151	106.030	0.000	15.709	16.301
Gearbox_oil_Temp	27.1594	0.216	125.792	0.000	26.736	27.583

```

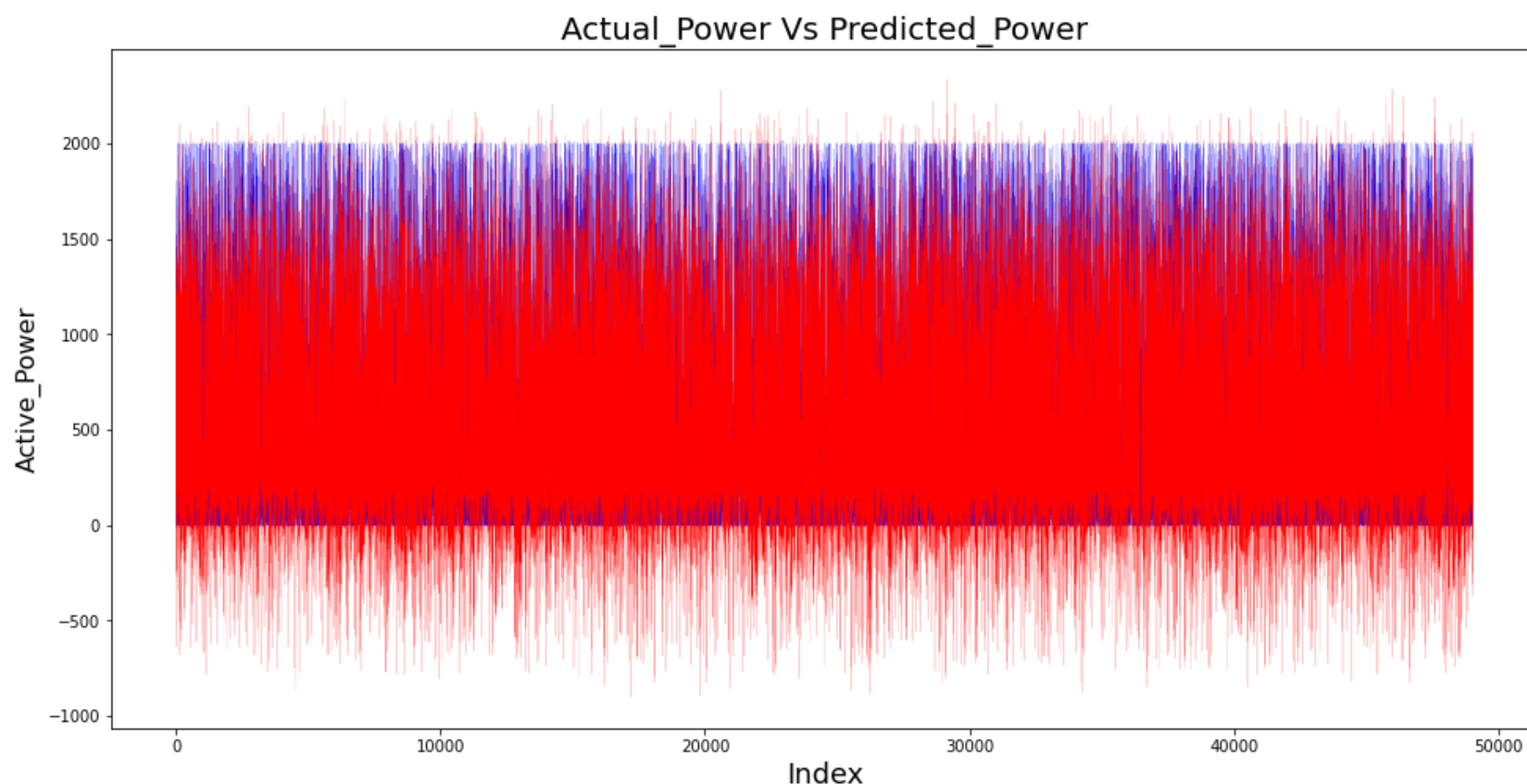
=====
Omnibus:                 46906.283      Durbin-Watson:           1.999
Prob(Omnibus):            0.000      Jarque-Bera (JB):        338443.212
Skew:                     0.963      Prob(JB):                0.00
Kurtosis:                 9.140      Cond. No.                3.33e+03
=====

```

Notes:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 3.33e+03. This might indicate that there are strong multicollinearity or other numerical problems.

```
In [90]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred9, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



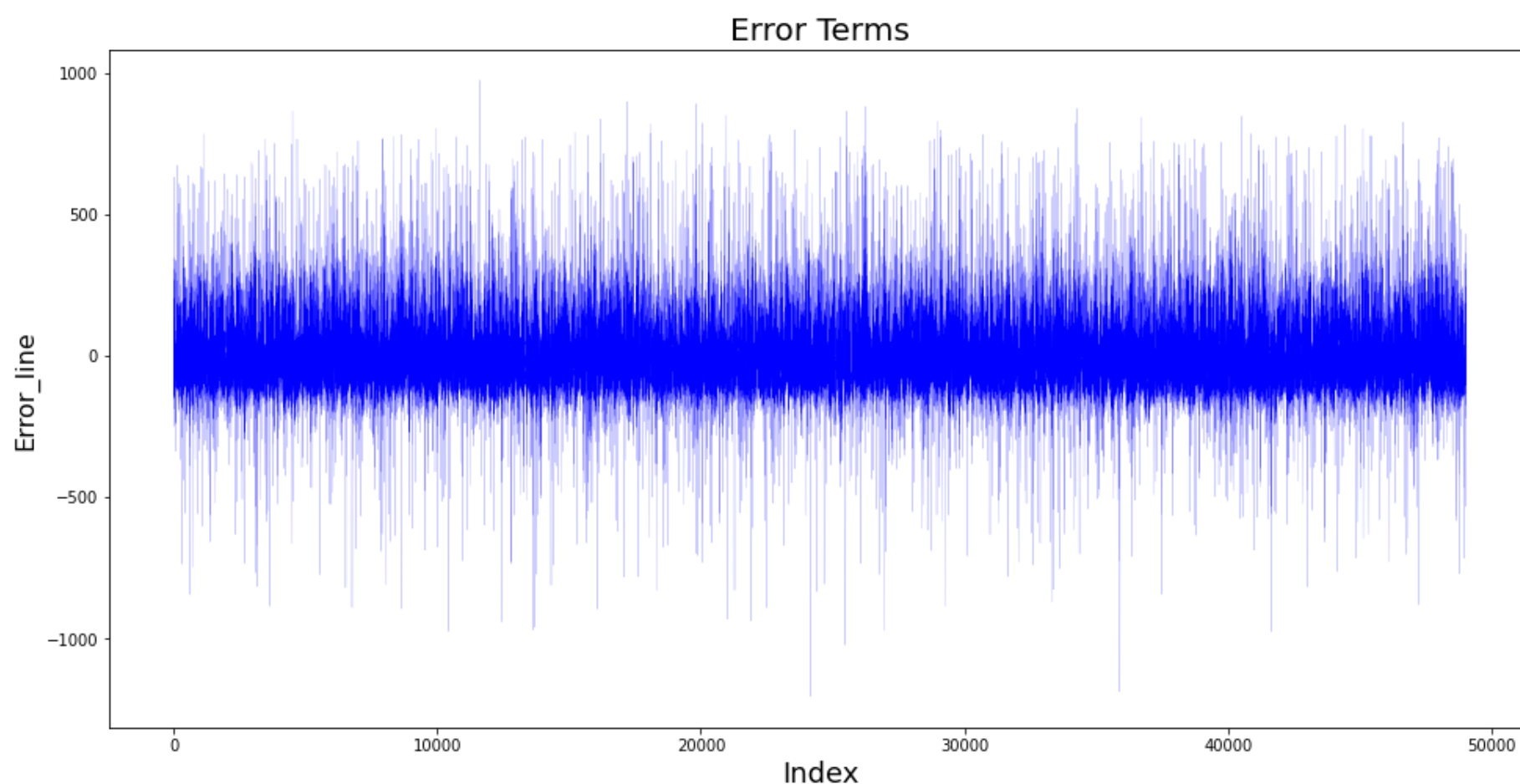
```
In [91]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x9_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x9_train.values, i)
                    for i in range(len(x9_train.columns))]

print(vif_data)
```

	feature	VIF
0	Avg_Ambient_Temp	63.729292
1	Avg_Pitch_Angle	9.520462
2	Avg_Rotor_Speed	112.348573
3	Avg_Wind_Speed	39.174582
4	Bearing_DE_Temp	400.707358
5	Generator_wind_Temp_1	264.268772
6	Generators_sliprings_Temp	383.082762
7	Gearbox_oil_Temp	230.627366

```
In [92]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred9, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-Label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [93]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred9)
r_squared = r2_score(y_test, y_pred9)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

Mean\_Squared\_Error : 21443.519337970352

r\_square\_value : 0.929 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed

#### MODEL-10

```
In [94]: x10_train = x_train[['Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed']]
x10_test = x_test[['Avg_Pitch_Angle', 'Avg_Rotor_Speed', 'Avg_Wind_Speed']]
```

Removed all temperature variables.

```
In [95]: lr10 = lr.fit(x10_train,y_train)
```

```
In [96]: y_pred10 = lr10.predict(x10_test)
```

```
In [97]: import statsmodels.api as sm
x10_train_sm = x10_train

x10_train_sm = sm.add_constant(x10_train_sm)

mlm10 = sm.OLS(y_train,x10_train_sm).fit()

mlm10.params
print(mlm10.summary())
```

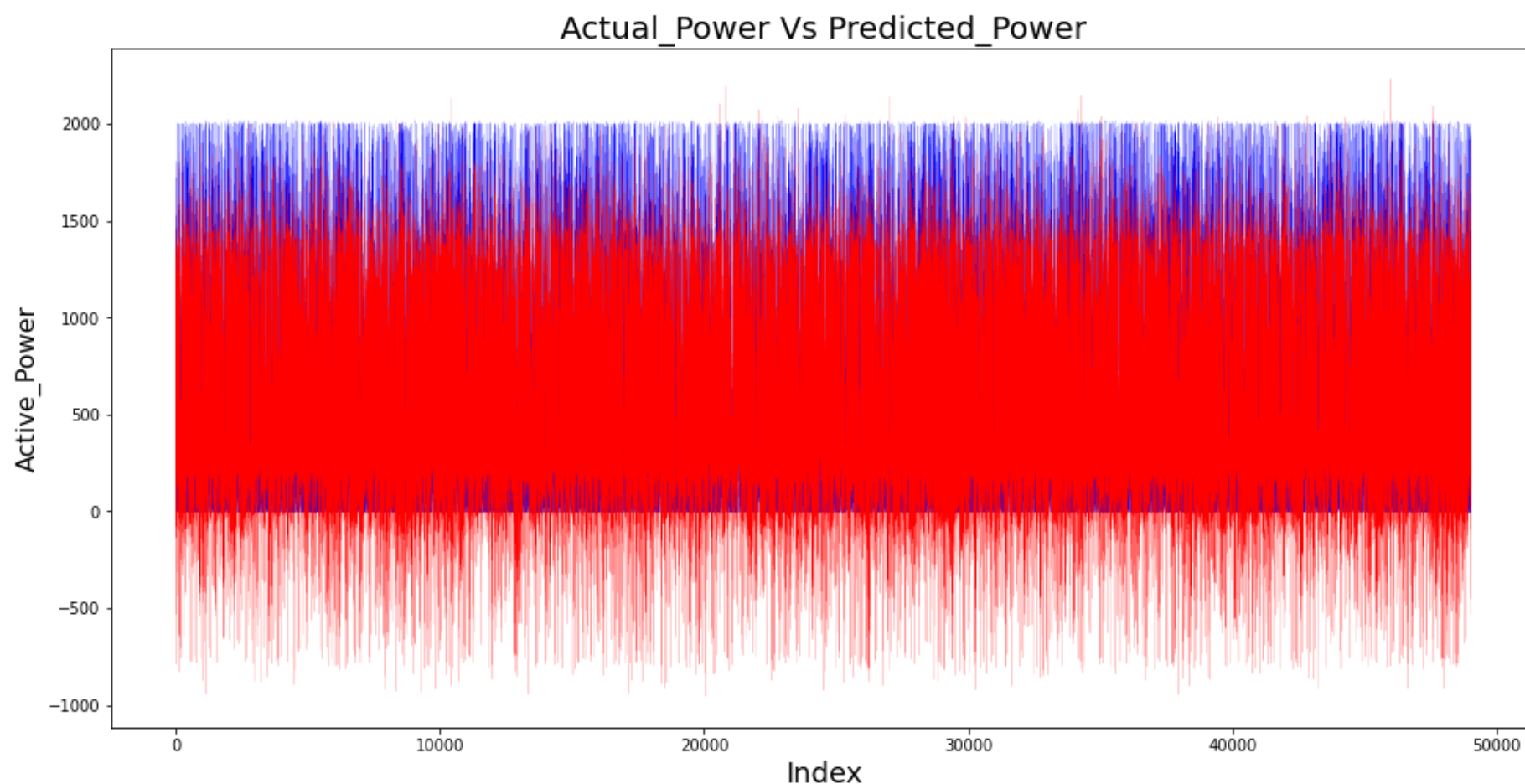
```

                        OLS Regression Results
=====
Dep. Variable:          Avg_Active_Power      R-squared:                0.901
Model:                  OLS                   Adj. R-squared:           0.901
Method:                 Least Squares         F-statistic:             5.930e+05
Date:                  Mon, 10 May 2021       Prob (F-statistic):      0.00
Time:                  00:41:58               Log-Likelihood:         -1.2903e+06
No. Observations:      196159                AIC:                    2.581e+06
Df Residuals:          196155                BIC:                    2.581e+06
Df Model:              3
Covariance Type:       nonrobust
=====
                        coef      std err          t      P>|t|      [0.025      0.975]
-----
const                -1667.8072      2.493    -668.920      0.000    -1672.694    -1662.920
Avg_Pitch_Angle        17.4895      0.045     391.660      0.000      17.402      17.577
Avg_Rotor_Speed       163.9809      0.332     494.522      0.000     163.331     164.631
Avg_Wind_Speed         54.4955      0.308     176.775      0.000      53.891      55.100
=====
Omnibus:              54262.901    Durbin-Watson:           1.997
Prob(Omnibus):         0.000    Jarque-Bera (JB):        227683.002
Skew:                  1.313    Prob(JB):                0.00
Kurtosis:              7.578    Cond. No.                165.
=====
```

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
In [98]: #Actual vs Predicted
c = [i for i in range(1,49041,1)]
plt.plot(c,y_test, color="blue", linewidth=0.1, linestyle="-")
plt.plot(c,y_pred10, color="red", linewidth=0.1, linestyle="-")
plt.title('Actual_Power Vs Predicted_Power', fontsize=20)
plt.xlabel('Index', fontsize=18)
plt.ylabel('Active_Power', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```





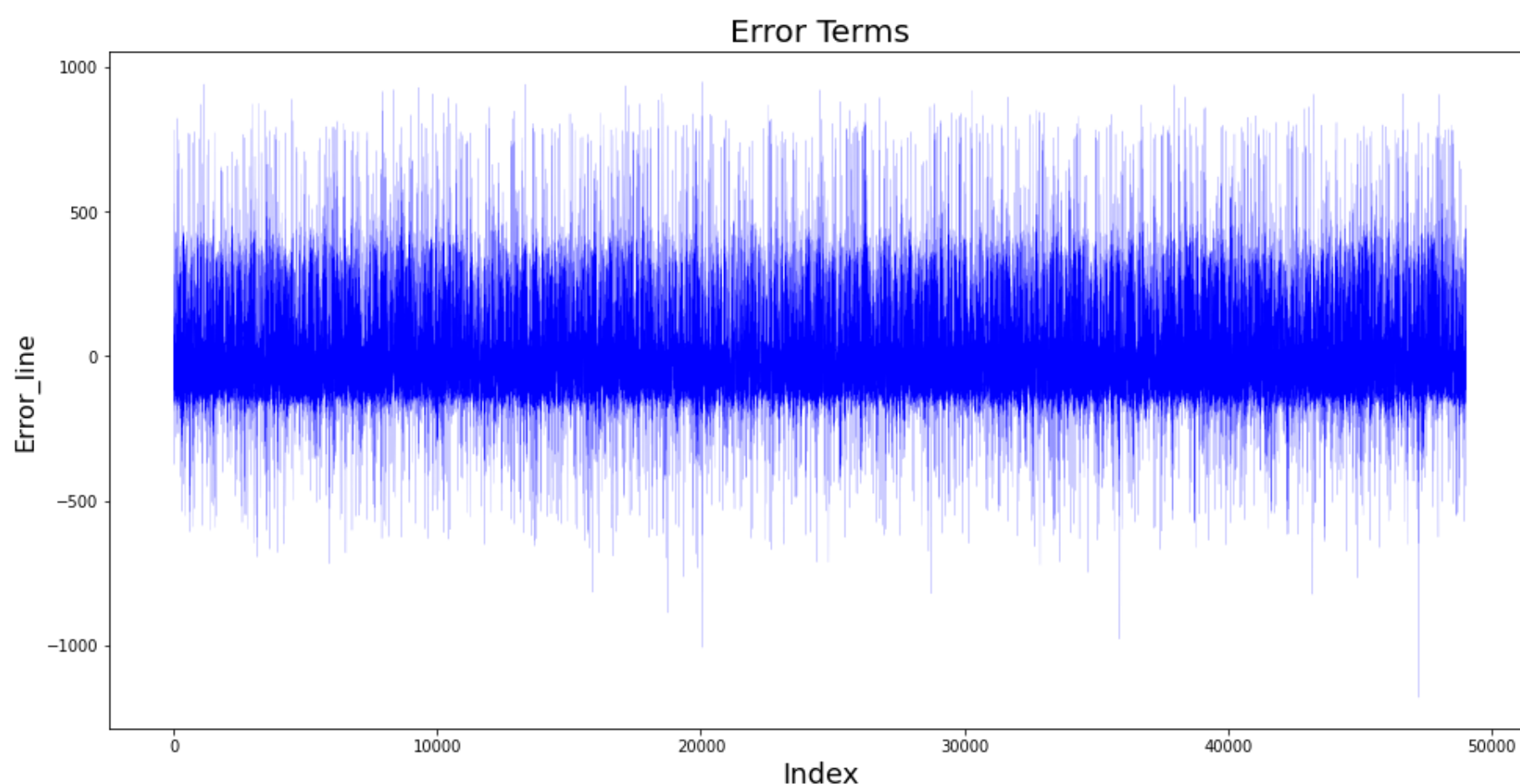
```
In [99]: from statsmodels.stats.outliers_influence import variance_inflation_factor
# VIF dataframe
vif_data = pd.DataFrame()
vif_data["feature"] = x10_train.columns
# calculating VIF for each feature
vif_data["VIF"] = [variance_inflation_factor(x10_train.values, i)
                    for i in range(len(x10_train.columns))]

print(vif_data)
```

```
      feature      VIF
0 Avg_Pitch_Angle  1.699851
1 Avg_Rotor_Speed 15.872537
2  Avg_Wind_Speed 16.956543
```

```
In [100]: c = [i for i in range(1,49041,1)]

plt.plot(c,y_test-y_pred10, color="blue", linewidth=0.1, linestyle="-")
plt.title('Error Terms', fontsize=20)           # Plot heading
plt.xlabel('Index', fontsize=18)                # X-label
plt.ylabel('Error_line', fontsize=16)
plt.subplots_adjust(left=0.4, bottom=0.1, right=2.5, top=1.6)
```



```
In [101]: from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred10)
r_squared = r2_score(y_test, y_pred10)
print('Mean_Squared_Error :',mse)
print('r_square_value :', round(r_squared,3),"% Variance of the Active Power is Explained by the Wind Speed, Pitch Angle
```

```
Mean_Squared_Error : 30233.20183664925
```

```
r_square_value : 0.9 % Variance of the Active Power is Explained by the Wind Speed, Pitch Angle and Rotor Speed
```

REGRESSION EQUATION

```
In [102]: @widgets.interact(Wind_Speed = range(1, 25, 1),
                             Pitch_Angle = range(1, 95, 1), Rotor_Speed = range(1, 20, 1))
def Regression_Equation(Wind_Speed, Pitch_Angle, Rotor_Speed):
    Intercept = mlm10.params[0]
    Coefficient = mlm10.params[1]*Pitch_Angle + mlm10.params[2]*Rotor_Speed + mlm10.params[3]*Wind_Speed

    y_Predicted = Intercept + Coefficient # y = a + bx Equation.
    if y_Predicted < 0:
        print("Please Choose Other Values: Active Power is '0' at these values")
    else:
        print('Active Power Predicted: ',y_Predicted) # Basic Starting Values are 3, 77, 1
```

```
interactive(children=(Dropdown(description='Wind_Speed', options=(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 1...
```

Thank You...