

# Baker Hughes México Data Science Hackathon 2022

Predicting modeling Challenge

Predictive team A

March 2022

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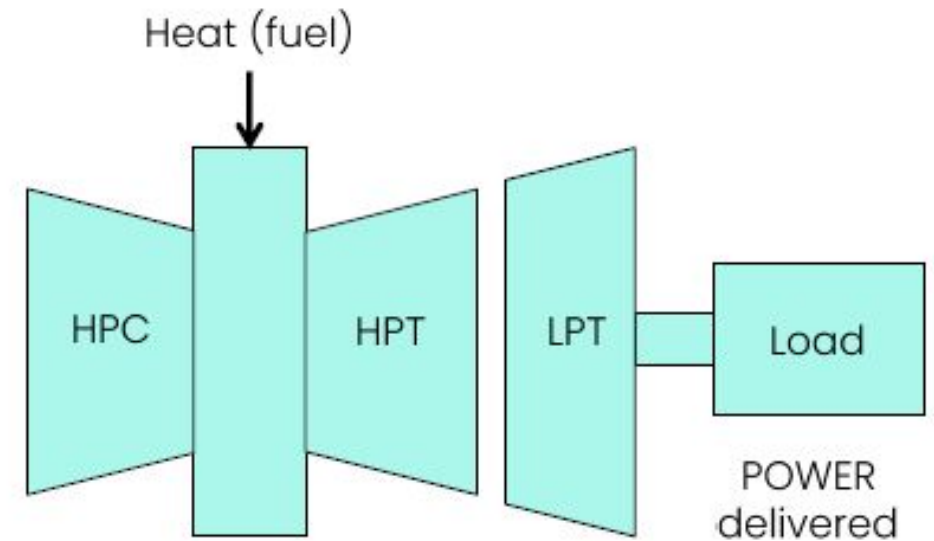


# Executive summary

- Exploratory data analysis
- Model selection by Root Mean Square Mean
- Linear regression model
- Model behaviour as expected
- Assumptions of variables made with the linear regression model.
- Code & Results

# Problem description

- In this competition we had to be able to develop a model for an Aeroderivative Gas Turbine to predict the POWER (kW) output from the Low Pressure Turbine (LPT).
- For this porpoise we handled synthetic data that simulates the behavior of Gas Turbine engines based at different locations worldwide.



# Exploratory data analysis

- **General data overview**

- Unusual 0 values
- NaN values

```
In [70]: .
Out[70]:
```

	T_AMB	P_AMB	CMP_SPEED	...	RH	WAR	POWER
0	1.450440	0.843522	0.000000	...	81.237441	0.000041	NaN
1	2.761142	0.843856	7870.729713	...	74.311313	0.000041	13332.692409
2	9.270325	0.843413	9898.625866	...	47.897182	0.000041	13026.684965
3	14.293265	0.844249	9850.791469	...	34.400729	0.000041	12773.507042
4	12.875213	0.843663	9828.508458	...	37.537882	0.000041	12768.092781

[5 rows x 12 columns]

- First, we guest it was a sensor's problem.
- Then we realize that the zeros correspond to the variable CMP\_SPEED and the NaN values where allocated to the turbine's variables.
- Showing that in those days the turbine was put-off

# Data preprocessing and feature engineering techniques

- **Handling missing values**

- Delete the rows with missing values
- Replace the missing values with the statistics mean
- Replace the missing values with zeros

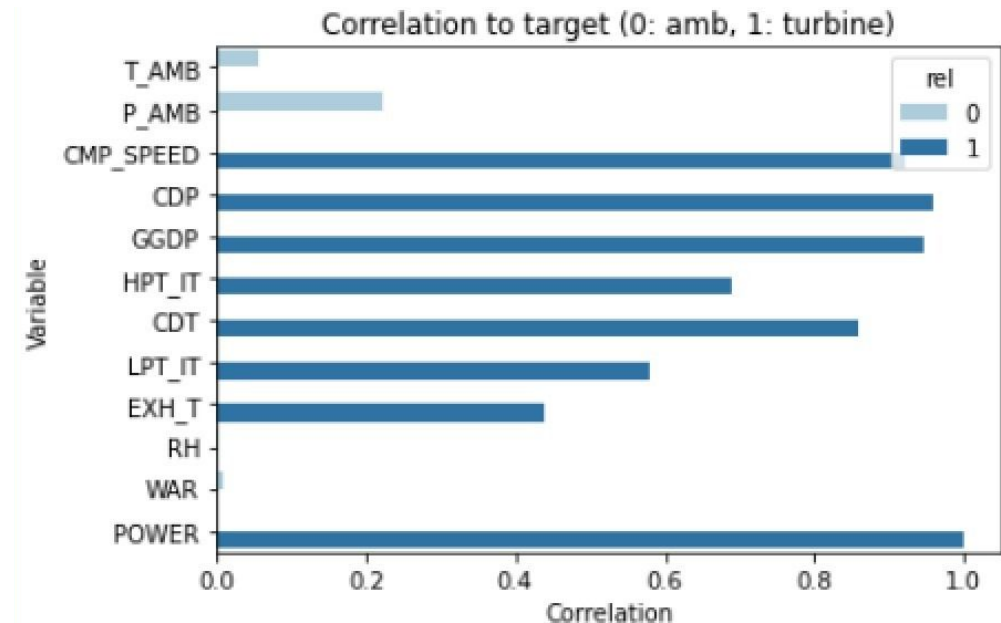
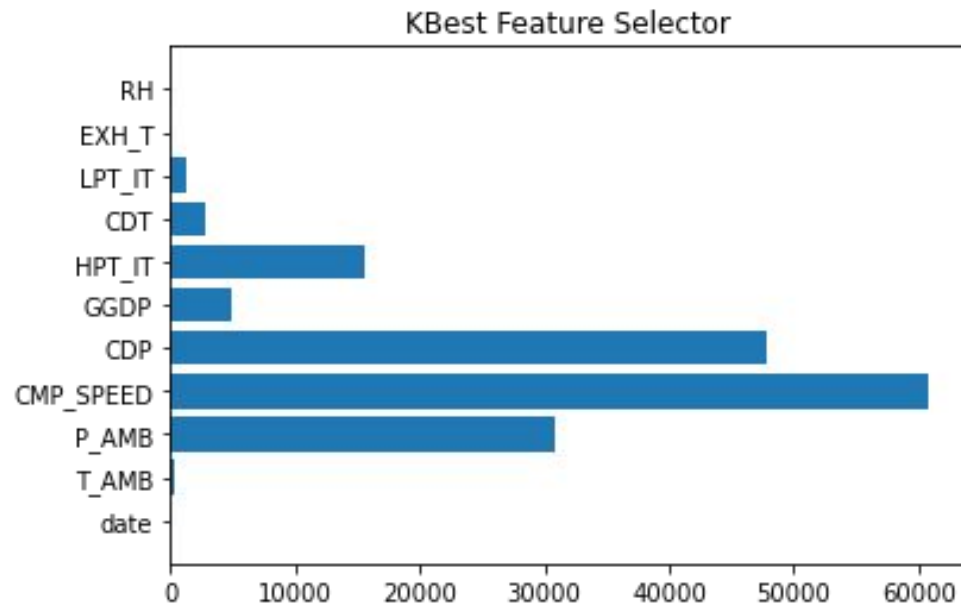
```
# Remove all nan values
df.dropna(inplace = True)
df_test.dropna(inplace = True)

# Replace with mean
for col in df.columns:
    if df[col].dtype == 'float':
        df[col].fillna(df[col].mean(), inplace = True)
for col in df_test.columns:
    if df_test[col].dtype == 'float':
        df_test[col].fillna(df_test[col].mean(), inplace = True)

# Handling missing values with zero
df.fillna(0, inplace = True)
df_test.fillna(0, inplace = True)
```

# Data preprocessing and feature engineering techniques

- **Determinate hyperparameters.**



CMP\_SPEED, CDP, GGDP, HPT\_IT, CDT, LPT\_IT, EXH\_T.

# Selection of feature variables

- CMP\_SPEED, CDP, GGDP, HPT\_IT, CDT, LPT\_IT, EXH\_T.
- CMP\_SPEED - compressor speed in RPM.
- CDP - compressor discharge pressure, barA.
- CDT - compressor discharge temperature, degC. —
- GGDP - gas generator discharge pressure, barA.
- HPT\_IT - High Pressure Turbine (HPT) inlet temperature, degC.
- LPT\_IT - Low Pressure Turbine (LPT) inlet temperature, degC.
- EXH\_T - exhaust temperature, degC.

# Model comparison and model selection techniques

- Based in Root Mean Square Error parameters applied to the test dataset (subset).
- Logistic regression
- Linear regression
- Ridge regression



# Model Selection - Linear Regression

**Actual vs Predicted data on  
training dataset**

date	Actual	Predicted
2021-07-18	0.000000	-51.594492
2021-10-02	5283.494881	5249.715082
2021-03-09	10749.132285	11160.228757
2021-10-04	17166.465997	16250.215063
2021-05-29	3103.702759	3376.599212

**Actual vs Predicted data on  
test splitted data**

date	actual	predicted
2021-10-28	10757.528311	10579.359239
2021-02-03	11769.494848	12475.357093
2021-01-09	12818.381578	13487.561463
2021-04-10	7040.412363	7128.064488
2021-02-20	5584.987785	5620.507935

\*Method 1

# Model Selection - Linear Regression

**Actual vs Predicted data on  
training dataset**

date	Actual	Predicted
2021-07-18	0.000000	-31.763825
2021-10-02	5283.494881	5308.745030
2021-03-09	10749.132285	11233.155887
2021-10-04	17166.465997	16976.987363
2021-05-29	3103.702759	2194.332666

**Actual vs Predicted data on  
test splitted data**

date	actual	predicted
2021-10-28	10757.528311	10733.030062
2021-02-03	11769.494848	12433.462689
2021-01-09	12818.381578	13489.899869
2021-04-10	7040.412363	7393.157403
2021-02-20	5584.987785	5647.803598

\*Method 2

# RMSE For Model Evaluation

$$RMSE = \sqrt{\frac{\sum_{i=1}^N (Predicted_i - Actual_i)^2}{N}}$$



```
rmse_train = (np.sqrt(mean_squared_error(target_train, target_train_predict)))  
rmse_test = (np.sqrt(mean_squared_error(target_test, target_test_predict)))  
  
print("RMSE for training data: {:.4f}".format(rmse_train))  
print("RMSE for test splitted data: {:.4f}".format(rmse_test))
```

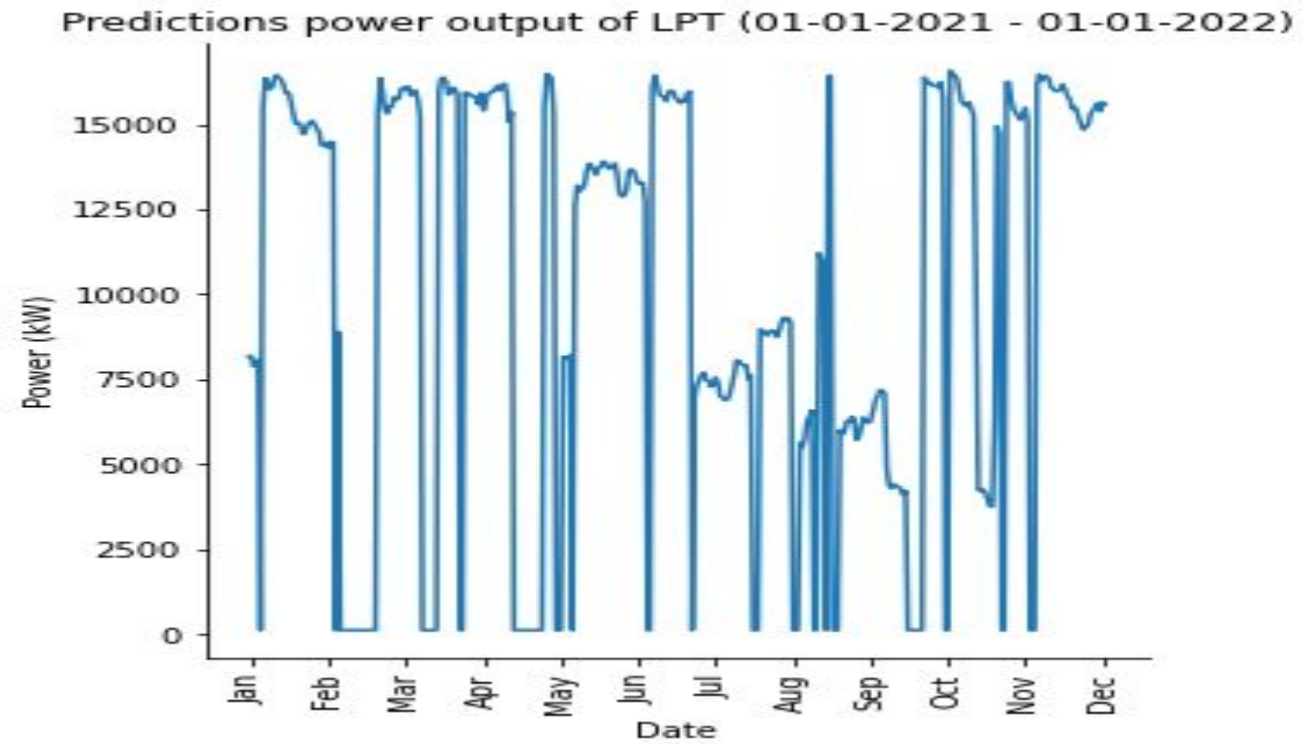
RMSE for training data: 354.4378  
RMSE for test splitted data: 363.4620

**Variables Selected by Method 1**

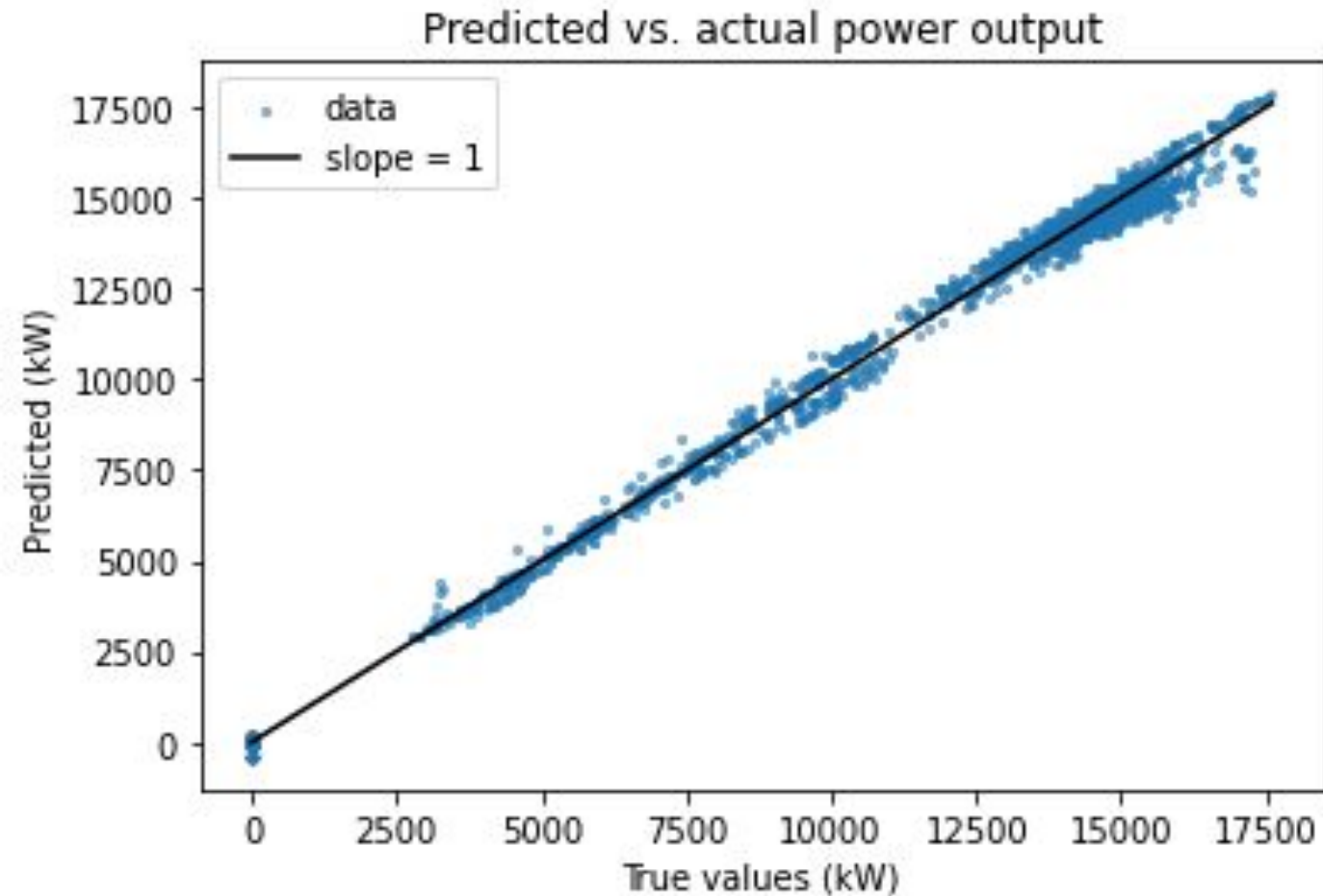
RMSE for training data: 457.8888  
RMSE for test splitted data: 482.0964

**Variables Selected by Method 2**

# Results



# Expected vs. predictions



# Model coefficients and model eval.

- Interception  $\rightarrow -3080.509774297314$
- Coefficients  $\rightarrow -0.78558, 189.443, 3008, 260.384, -562.218, 306.611$
- Independent variables  $\rightarrow \text{CMP\_SPEED}, \text{CDP}, \text{GGDP}, \text{HPT\_IT}, \text{LPT\_IT}, \text{EXH\_T}$
- $$\text{POWER} = -3080.509774297314 - 0.78558\text{CMP\_SPEED} + 189.443\text{CDP} + 3008\text{GGDP} + 260.384\text{HPT\_IT} - 562.218\text{LPT\_IT} + 306.611\text{EXH\_T}$$

R2: 0.9965

# References

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# Q&A SESSION



# Personal experience