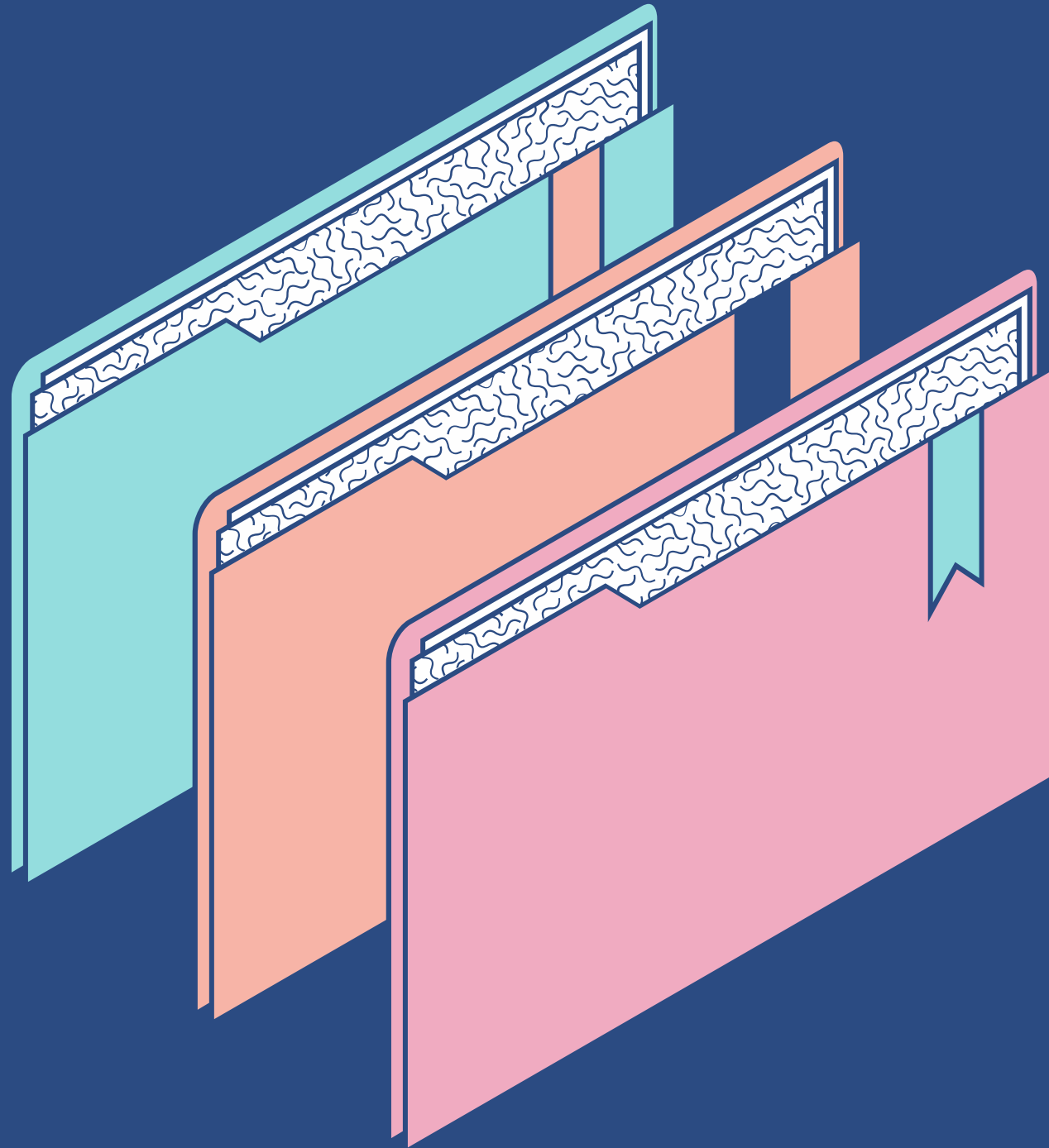


PHM08 Prognostics Data Challenge

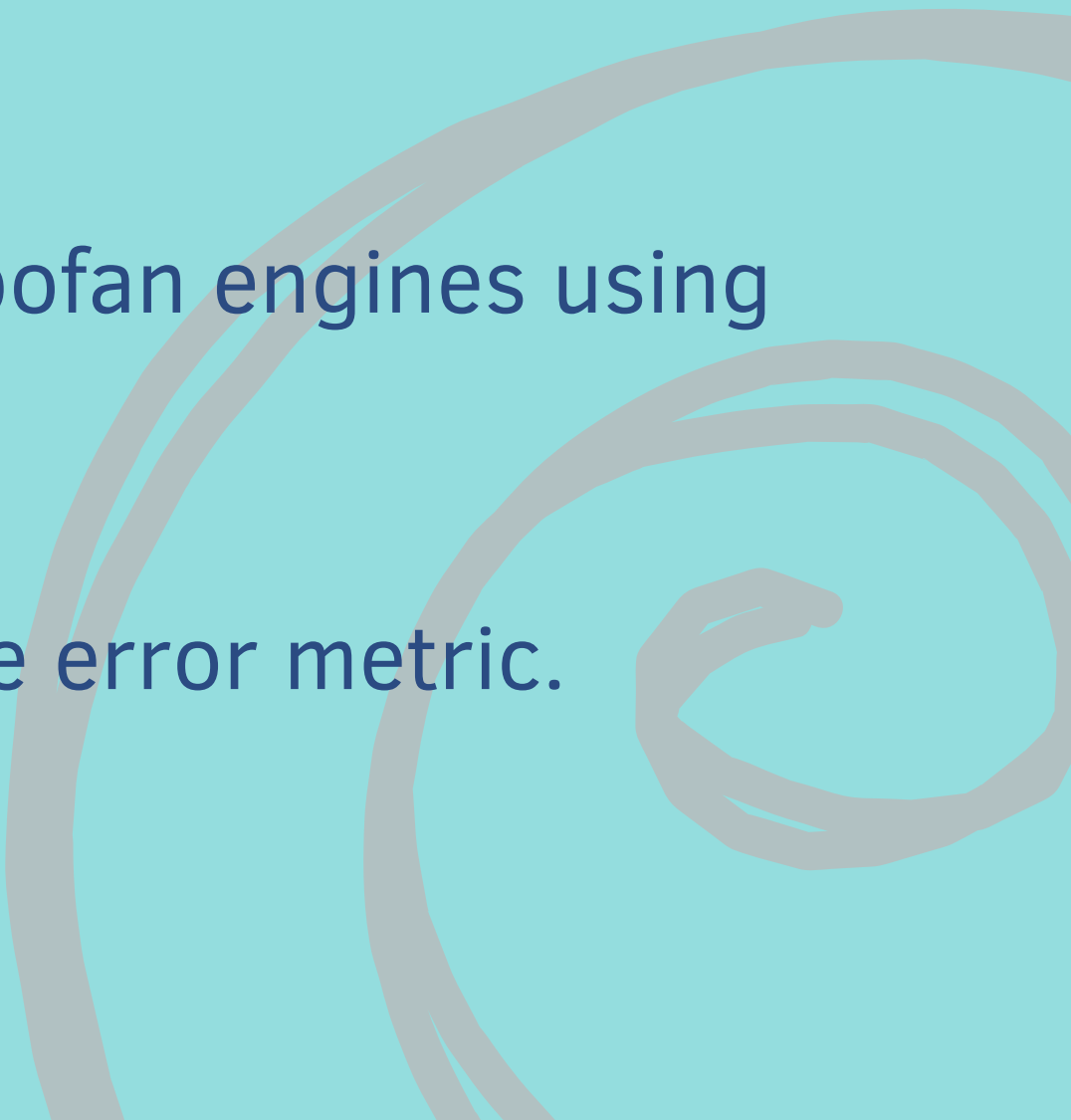
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Introduction & Problem Statement

- Predicting aircraft component life by using data to predict remaining operational cycles until failure is the problem statement addressed by the PHM 2008 Challenge.
 - Our project aims to predict the remaining useful life of turbofan engines using machine learning.
 - Our performance will be evaluated using the mean absolute error metric.
- 
- A decorative graphic consisting of several concentric, overlapping swirls in a light gray color, located in the bottom right corner of the slide.

Data

- We will use the dataset provided by the PHM 2008 Challenge, which includes sensor measurements and failure times of 218 turbofan engines.
- The datasets are:
 - train.txt
 - test.txt
 - final_test.txt
- Each row is a snapshot of data taken during a single operational cycle.
- The columns correspond to:
 - unit number
 - time, in cycles
 - operational setting 1-3
 - sensor measurement 1-21

Imported Libraries

```
library(TIDYVERSE)
```

```
library(CARET)
```




```
library(GGplot2)
```

```
library(RANDOMFOREST)
```



Loading the Data

- `train_data <- read.table("train.txt", header = TRUE)`
- `test_data <- read.table("test.txt", header = TRUE)`
- `final_test_data <- read.table("final_test.txt", header = TRUE)`

Data		
▶ <code>final_test_data</code>	55156 obs. of 26 variables	
▶ <code>test_data</code>	29820 obs. of 26 variables	
▶ <code>train_data</code>	45918 obs. of 26 variables	

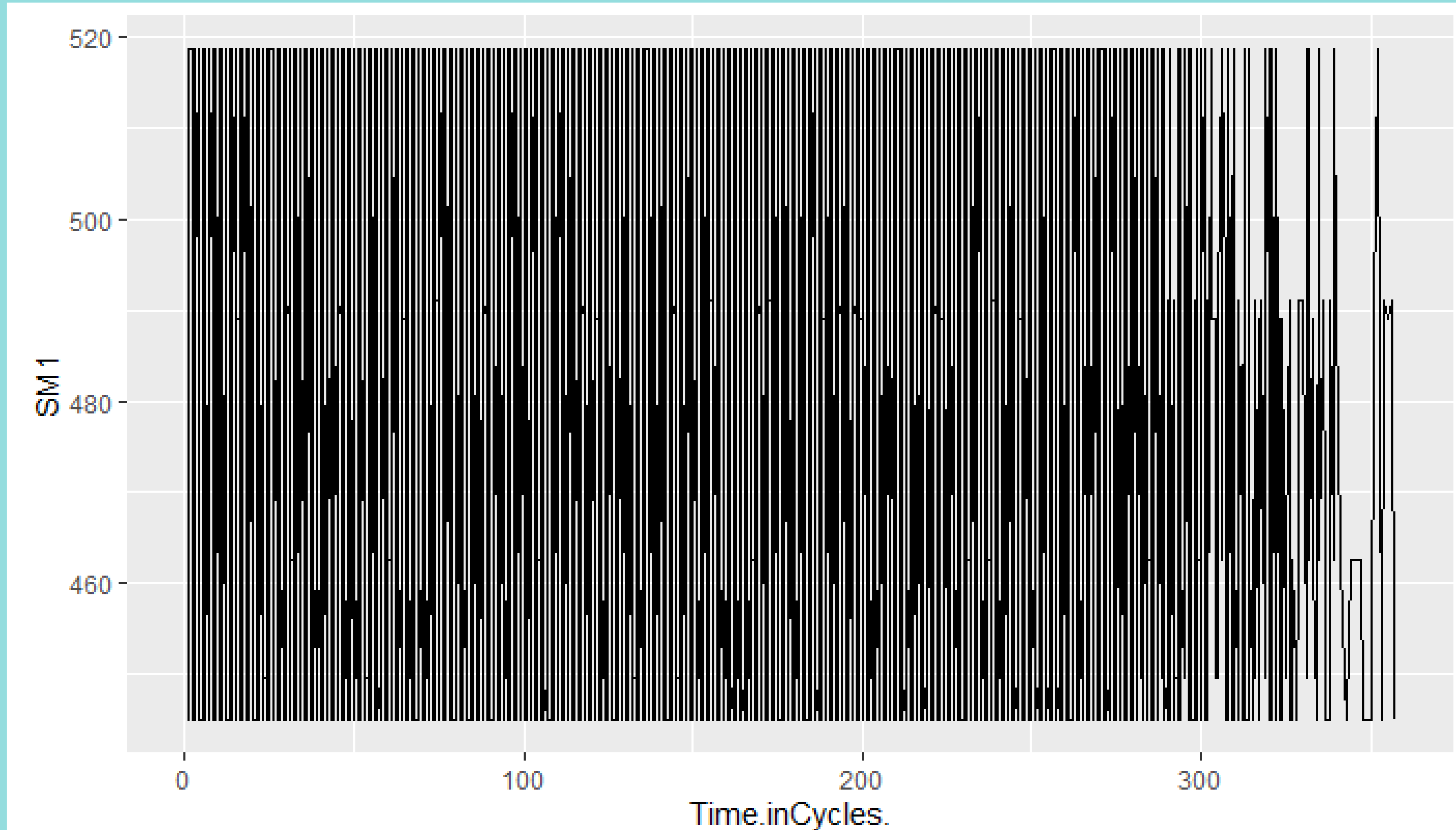
Cleaning the Data

Checked for missing values using the `sum(is.na())` function. If there are any missing values, they were removed.

- `sum(is.na(train_data))`

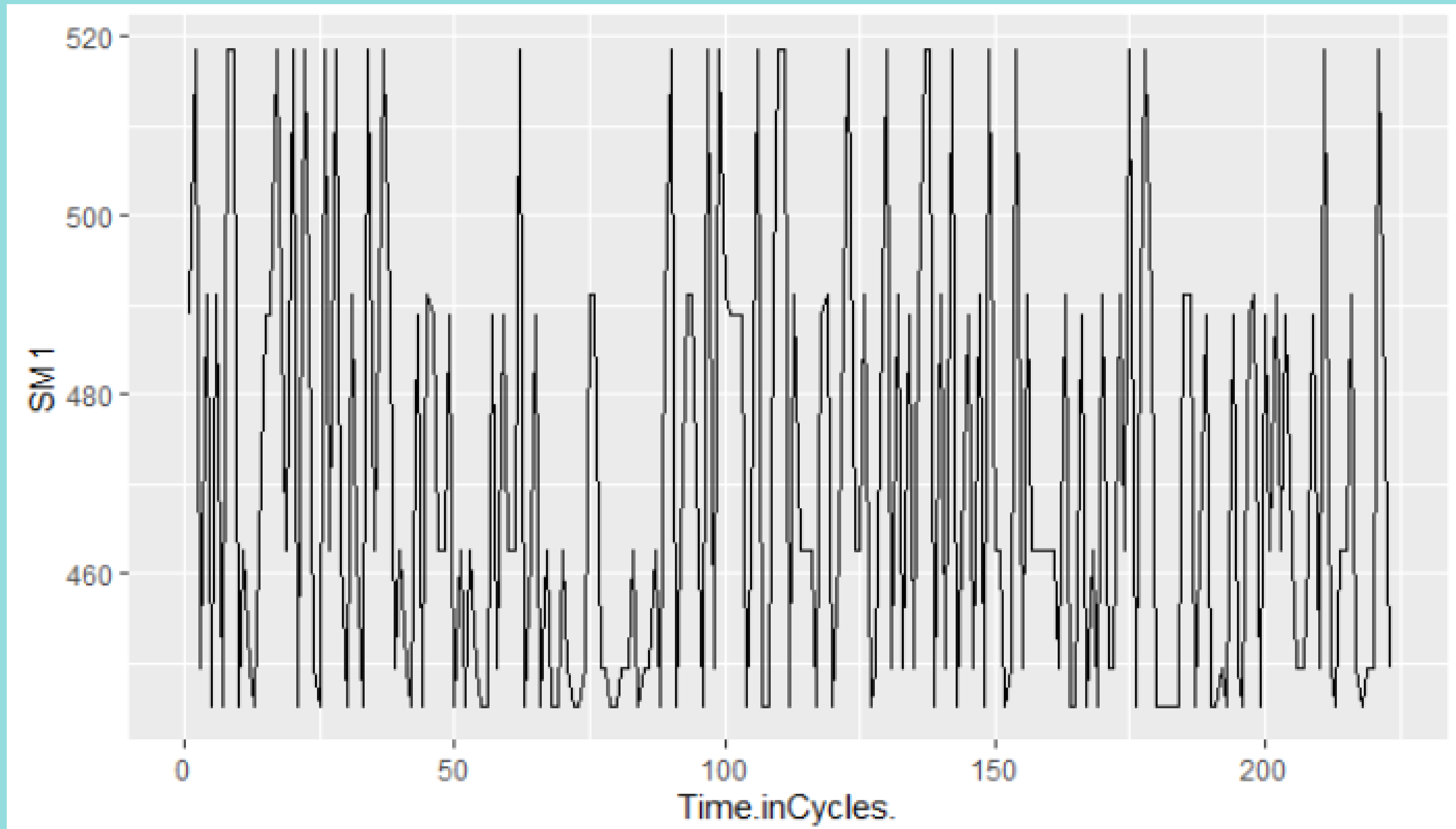
Visualization of Data

Sensor Measurement 1 vs Time Cycles



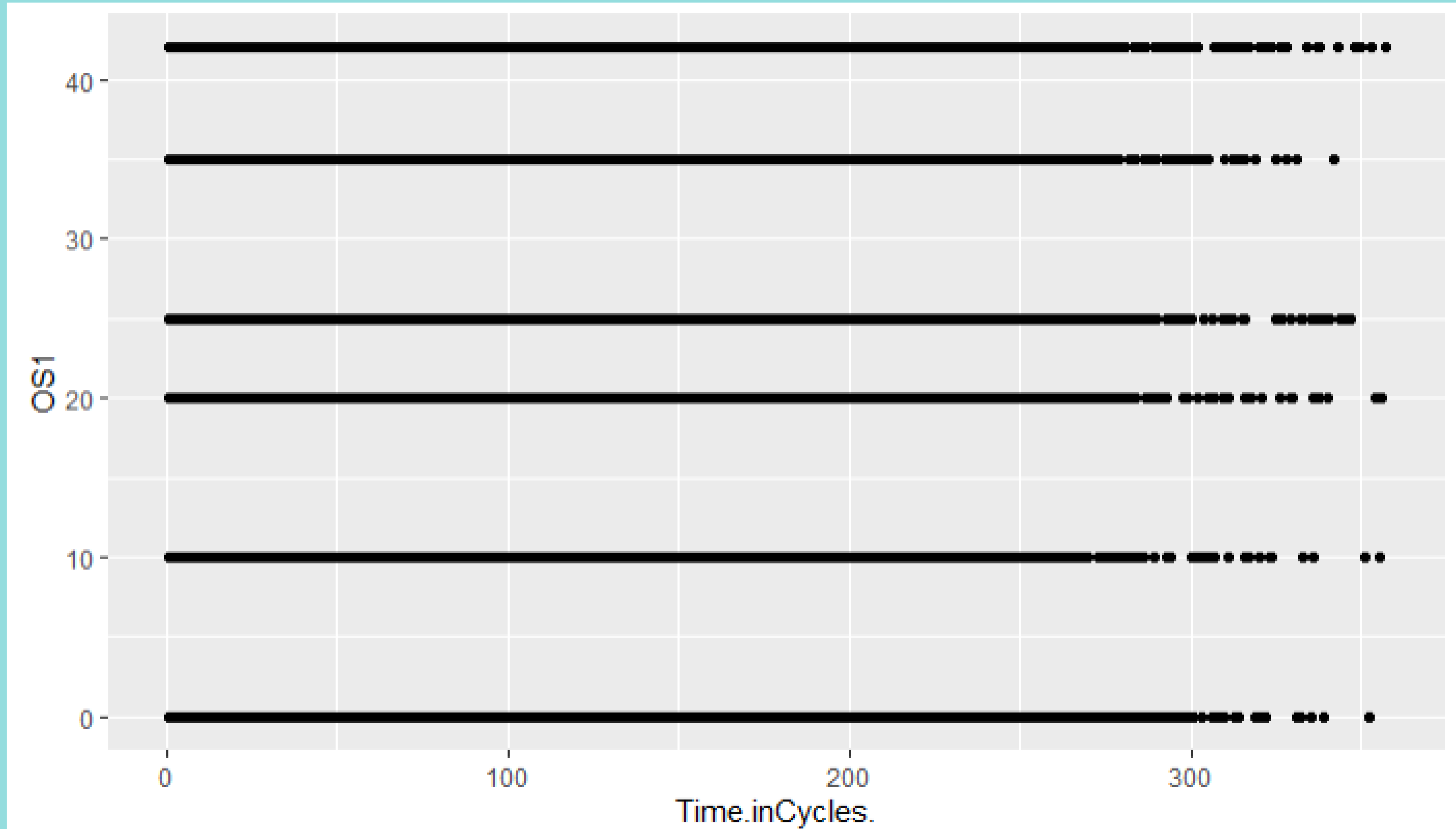
Visualization of Data

Sensor Measurement 1 vs Time Cycles for Engine 1



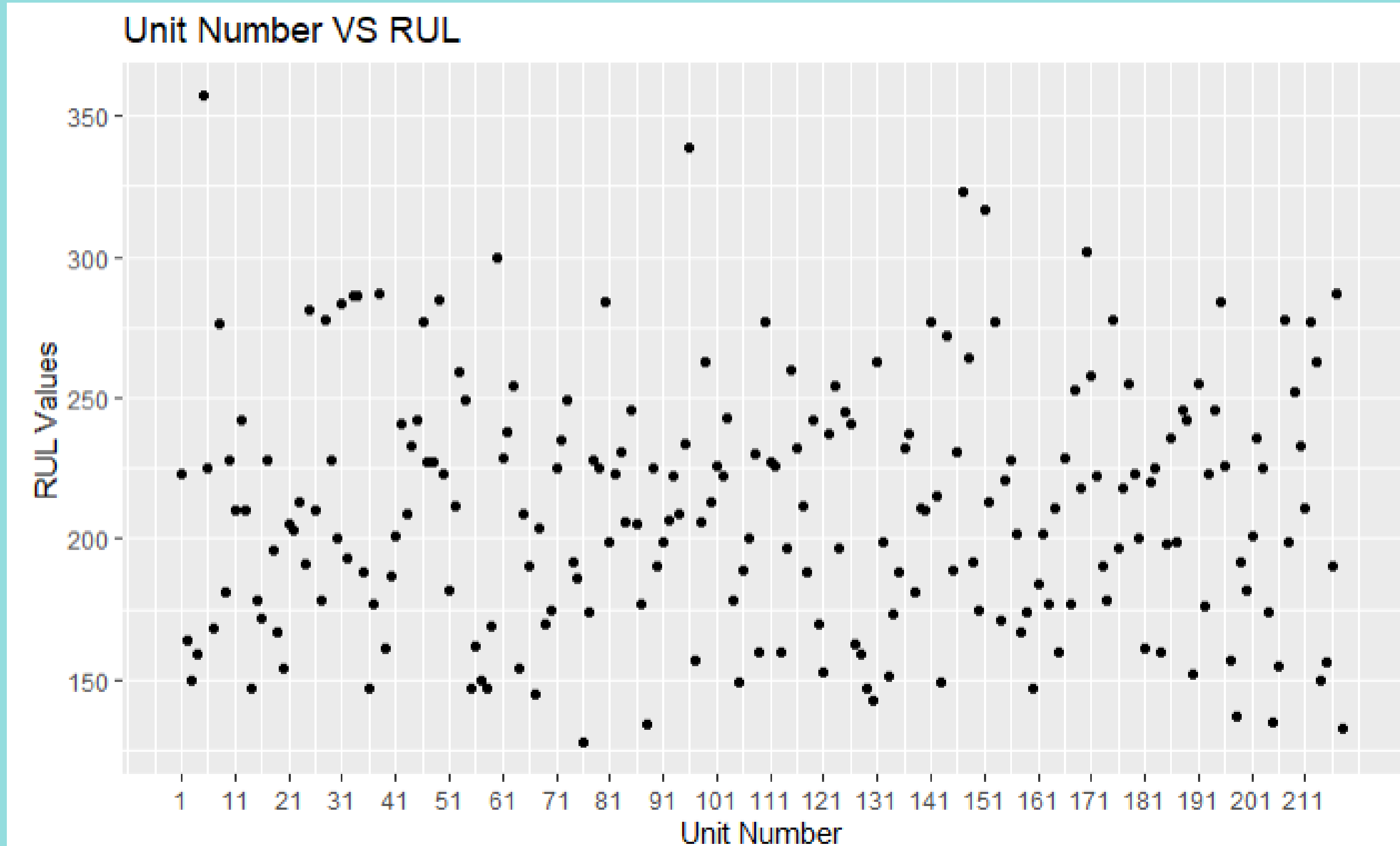
Visualization of Data

Operational Setting 1 vs Time Cycles for All Engines



Calculating the RUL (Remaining Useful Life) for the train_data

Calculated the RUL for each engine and visualized it as a scatterplot



Training

Training the data using the Random Forest Machine Learning Algorithm

```
model <- randomForest(RUL ~ ., data = train_data)
```

```
```{r}
set.seed(123)
Train a random forest model
library(randomForest)
model <- randomForest(RUL ~ ., data = train_data)
```
```

Prediction

Trying to predict the RULs of the test_data

```
# Make predictions on the testing data  
predictions <- predict(model, newdata = test_data)
```

| | UnitNumber | predicted_RUL |
|---|------------|---------------|
| 1 | 1 | 198.5743 |
| 2 | 2 | 197.8591 |
| 3 | 3 | 193.1419 |
| 4 | 4 | 177.8723 |
| 5 | 5 | 188.6867 |
| 6 | 6 | 197.3574 |
| 7 | 7 | 206.4155 |
| 8 | 8 | 195.1573 |
| 9 | 9 | 185.6197 |

Evaluation

After prediction of the RULs, we calculated the actual values and evaluated the accuracy of the prediction.

```
# Evaluate the model performance  
MAE <- mean(abs(predictions - test_data$predicted_RUL))
```

| | |
|-----|-----------------|
| MAE | 46.546851332746 |
|-----|-----------------|

Conclusion

- Our model predicts the remaining useful life of engines.
- The Random Forest method was used.
- We cleaned the data, trained our model, predicted, and evaluated our results.
- Key machine learning concepts were demonstrated.
- The code provides a framework for predictive model development.