

An aerial photograph of a multi-lane highway in Singapore, showing heavy traffic. The road is flanked by green trees and a concrete median. A white rectangular box is superimposed over the center of the image, containing the text "CAN COE PREMIUMS BE PREDICTED?".

CAN COE PREMIUMS BE  
PREDICTED?

## BACKGROUND

- **Certificate of Entitlement (COE)**
  - Right to own a vehicle for 10 years.
  - Separated into 5 Categories (A - E)
  - Quotas allocated by government.
- **Bidding Exercises**
  - Semi-Monthly (1<sup>st</sup> and 3<sup>rd</sup> week of each month)
  - Quota per category to be announced before each bidding exercise.
- **Purchasing Cars**
  - Before buying a car, there are various COEs packages to be bought.



## PROBLEM STATEMENT

- As a car dealer in Singapore, I want to accurately predict COE premiums. This will allow the company to maximise profits and set prices that are competitive in the market.
- Successfully bid for COE as soon as possible.

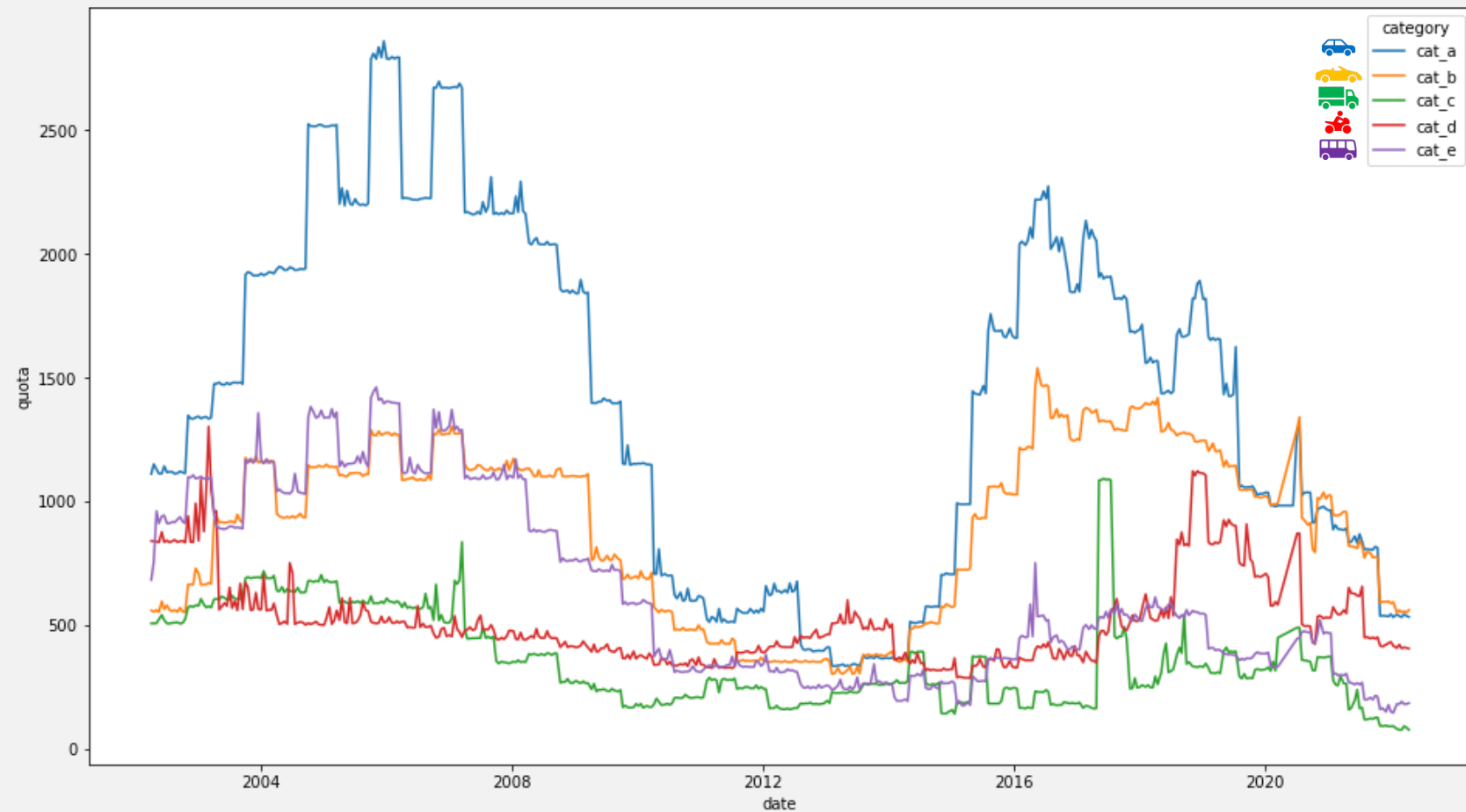
## GOAL

- In order to be able to accurately bid for a COE, I will be focusing on reducing MSE of the predictions.
- High success rate in bidding for COE is also important.
  - Success: Predicted price  $\geq$  premium price

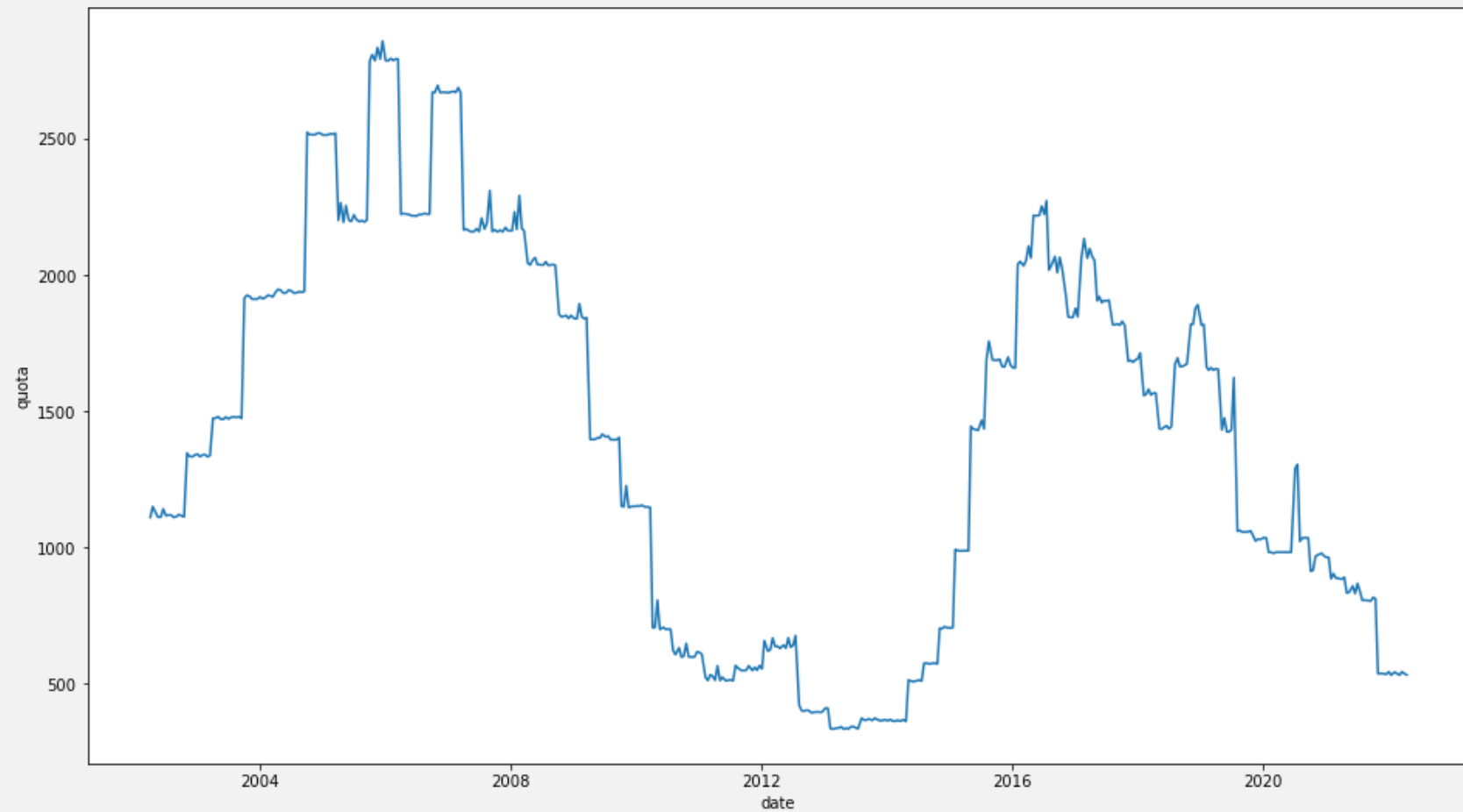


# EXPLORATORY DATA ANALYSIS

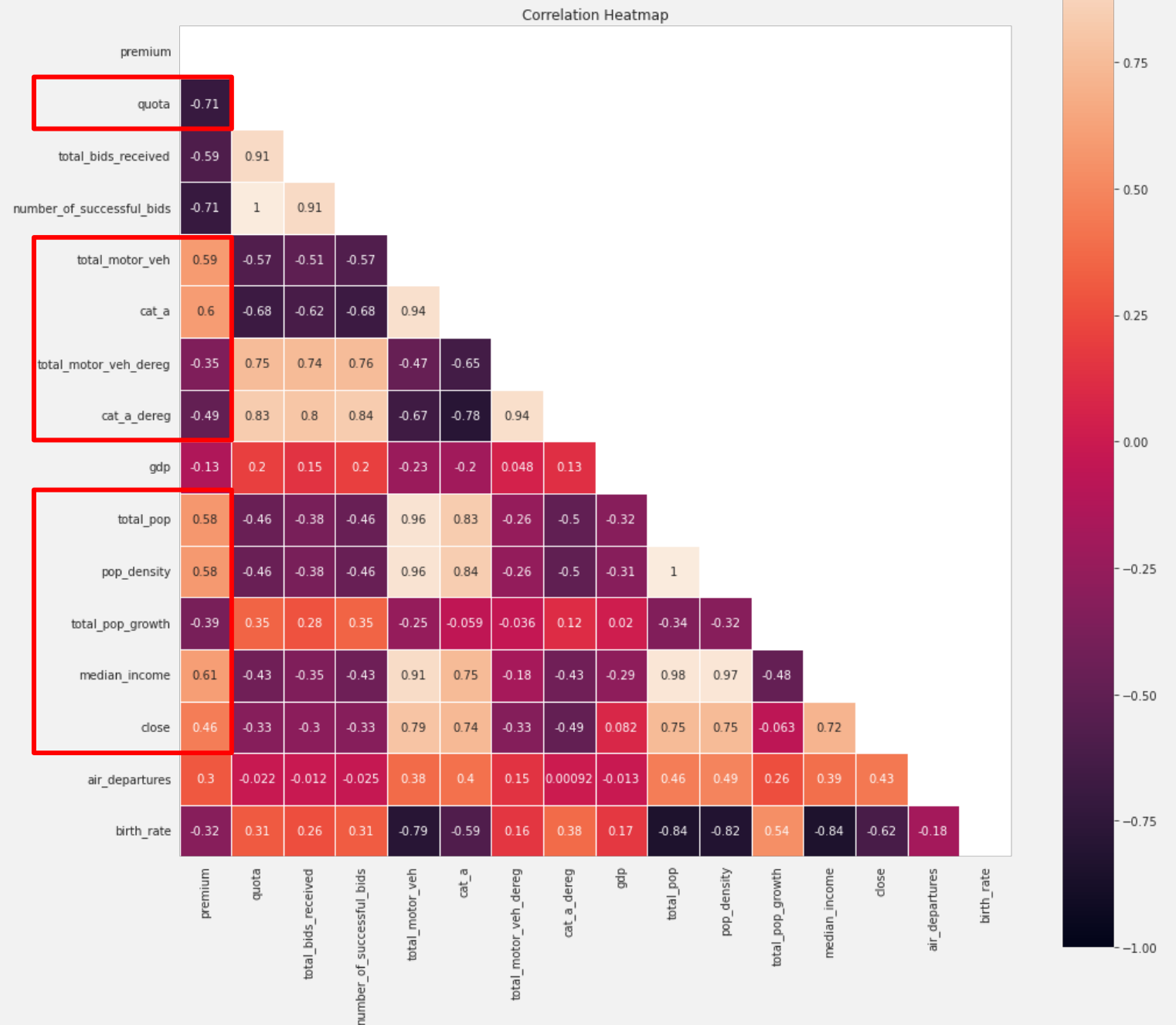
## COE – QUOTAS BY CATEGORY



## FOCUS – CATEGORY A PREMIUMS

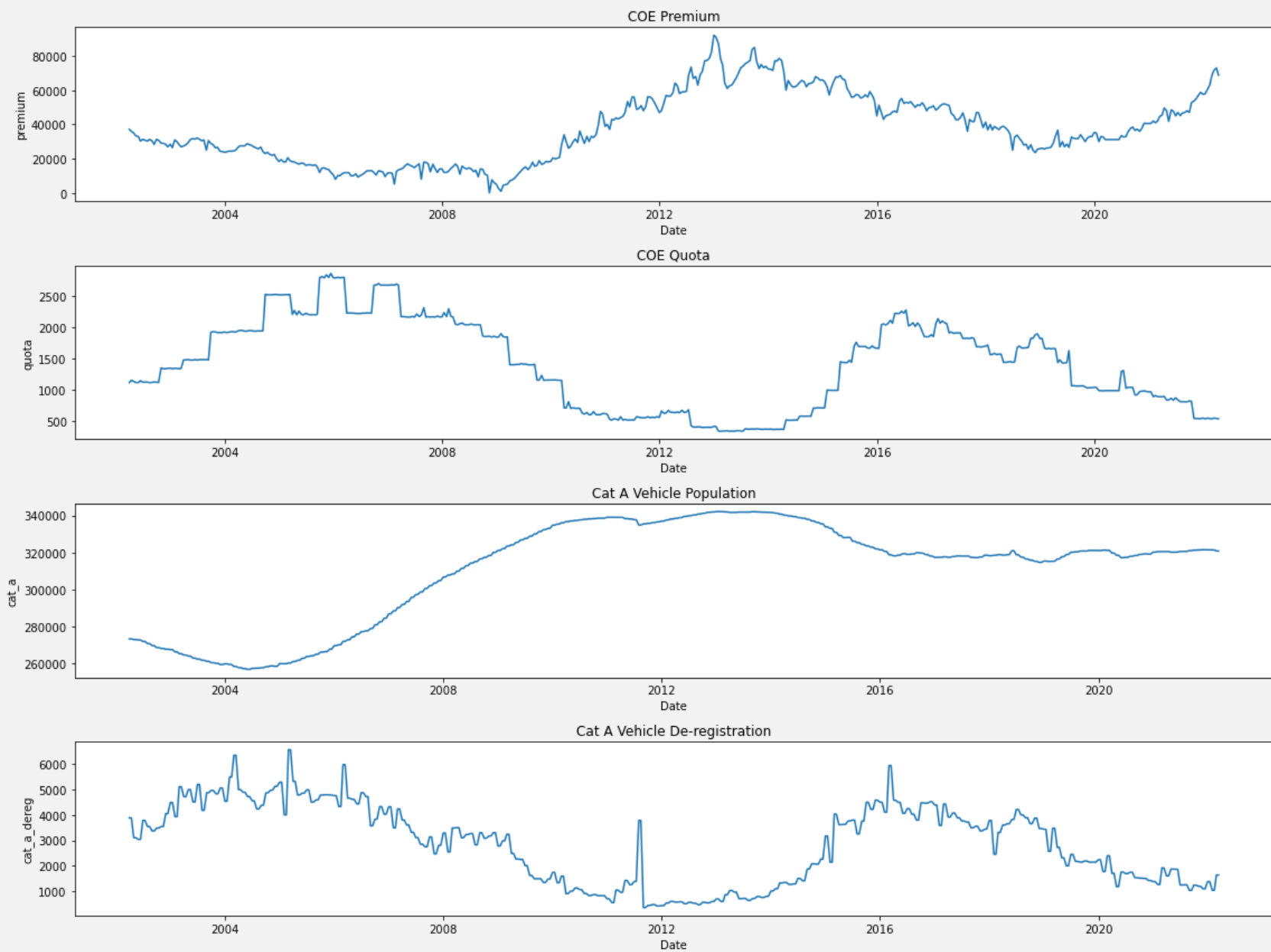


# EDA - HEATMAP

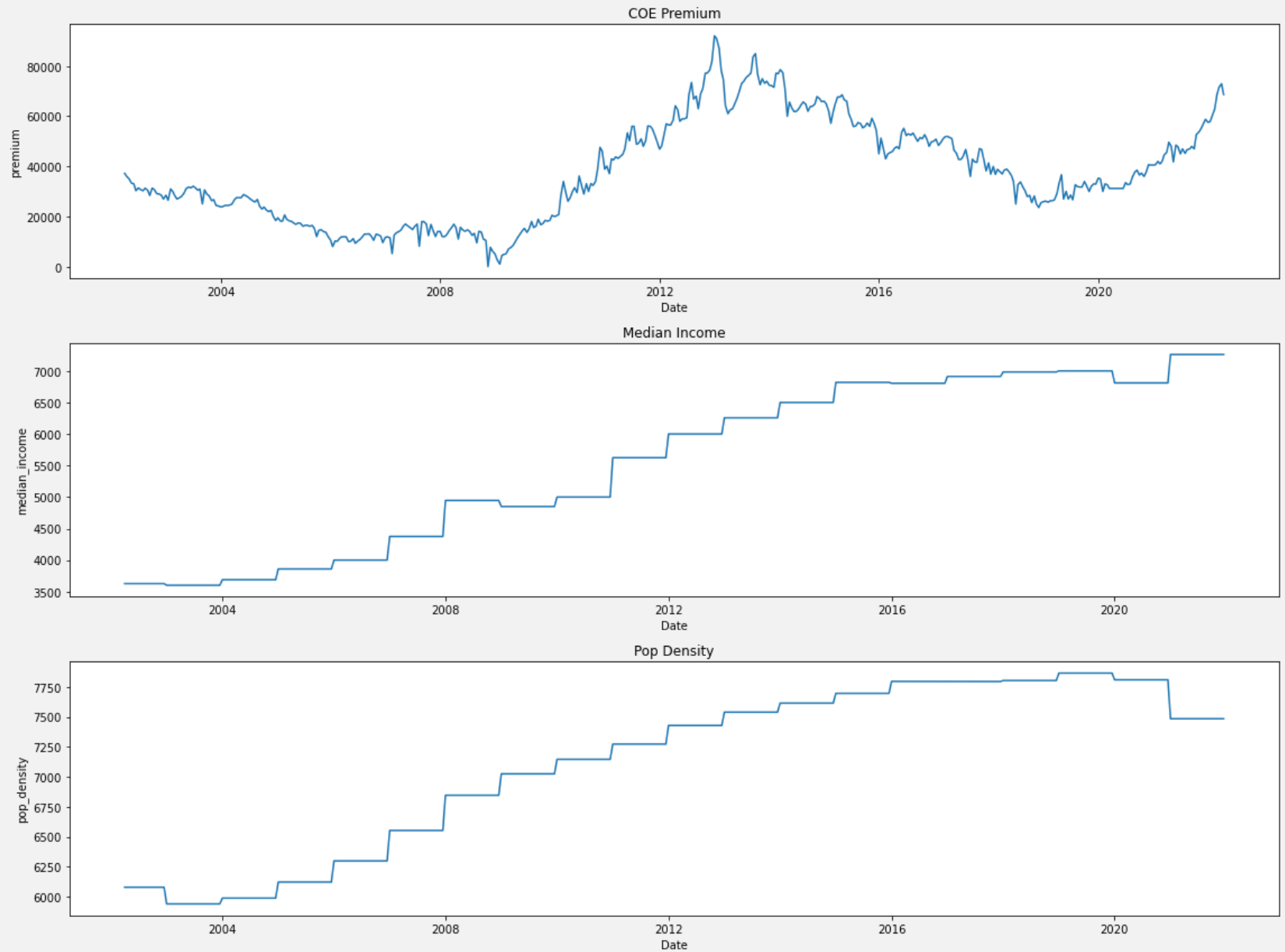




# EDA – CAT A VEHICLE ATTRIBUTES



## EDA – OTHER ECONOMIC VARIABLES



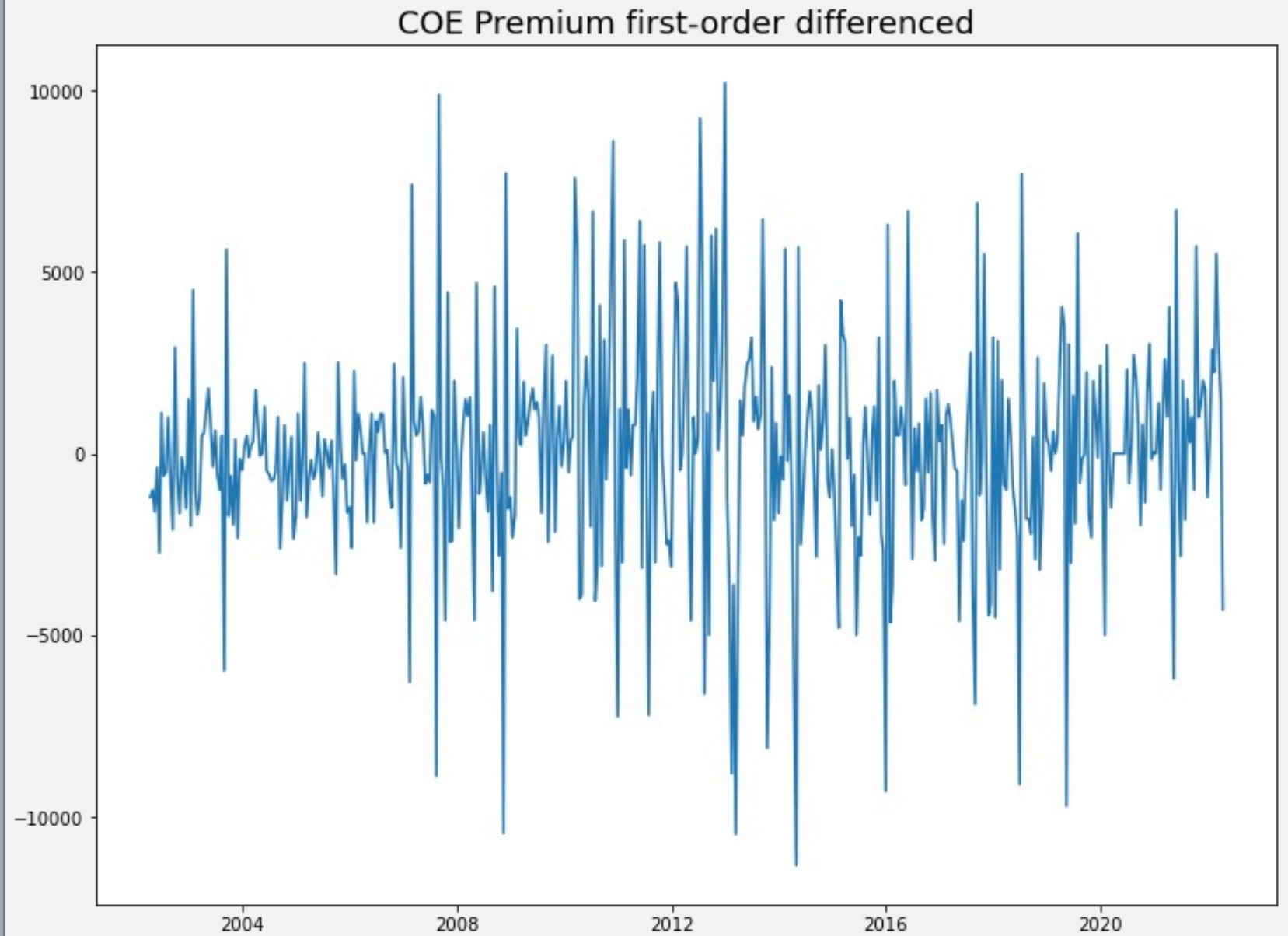
## KEY FINDINGS FROM EDA

- COE premium has a general upward trend.
- COE's 10 years seasonality is not apparent.
- Overall population of vehicle remains the same but median income is increasing.
- Variables have a strong correlation to premium prices but poor estimator.
  - Resolution of variable information is too low vs premium price.

# FEATURE ENGINEERING

- Scarcity – simulate behavioral demand
  - Quota / total bids received
- Lags for other variables
  - Yearly, Quarterly, Monthly data used to predict for the next cycle

CHECK FOR  
STATIONARITY



# MODELING

Model	Univariate / Multivariate	MSE	Success* %
<u>ARIMA</u>	Univariate	170,766,416.86	43.3
<u>Linear Regression</u>	Univariate	86,105,841.98	83.5
Linear Regression	Multivariate	627,776,111.23	100.0
Linear Regression w Lasso	Multivariate	629,132,479.33	100.0
Linear Regression w Ridge	Multivariate	629,122,991.38	100.0
<u>SARIMA</u>	Univariate	151,149,668.92	50.5

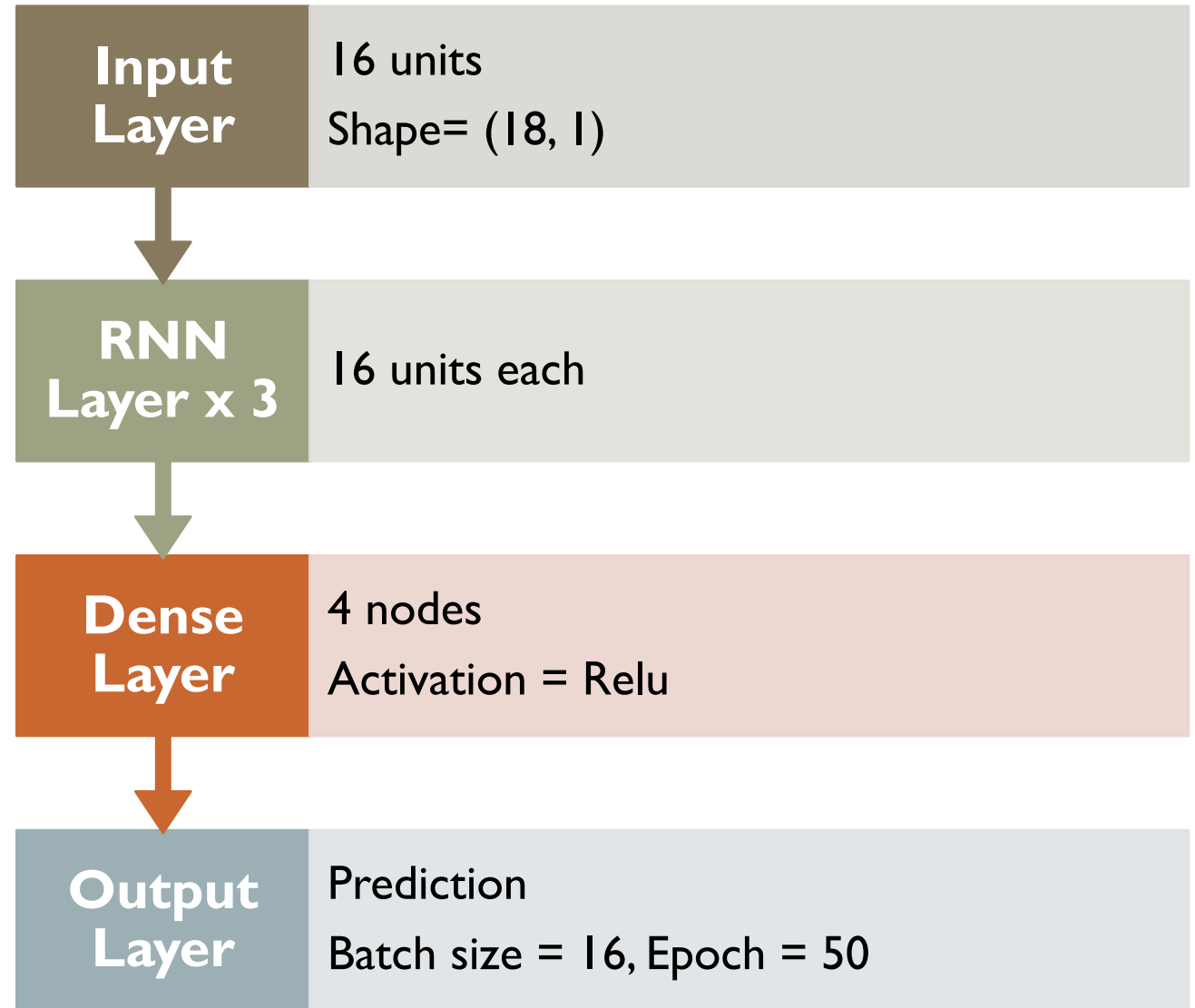
Variables used:

Linear Regression Model: Quota

Linear Regression Model 2: Quota, Deregistration of Cat A vehicle, Population Growth, STI close price

\* Success: Predicted price  $\geq$  premium price

STRUCTURE OF  
RECURRENT NEURAL  
NETWORK MODEL



# MODELING – RECURRENT NEURAL NETWORK

Model	Univariate / Multivariate	MSE	Success %
RNN - LSTM	Univariate	15,133,745.46	40.5
<a href="#">RNN - GRU</a>	Univariate	7,453,959.03	75.9
<a href="#">RNN - GRU 1</a>	Multivariate	8,823,055.69	20.3
RNN - GRU 2	Multivariate	19,145,021.12	5.1
RNN - GRU 3	Multivariate	10,139,610.50	19.0

Variables for Multivariate models:

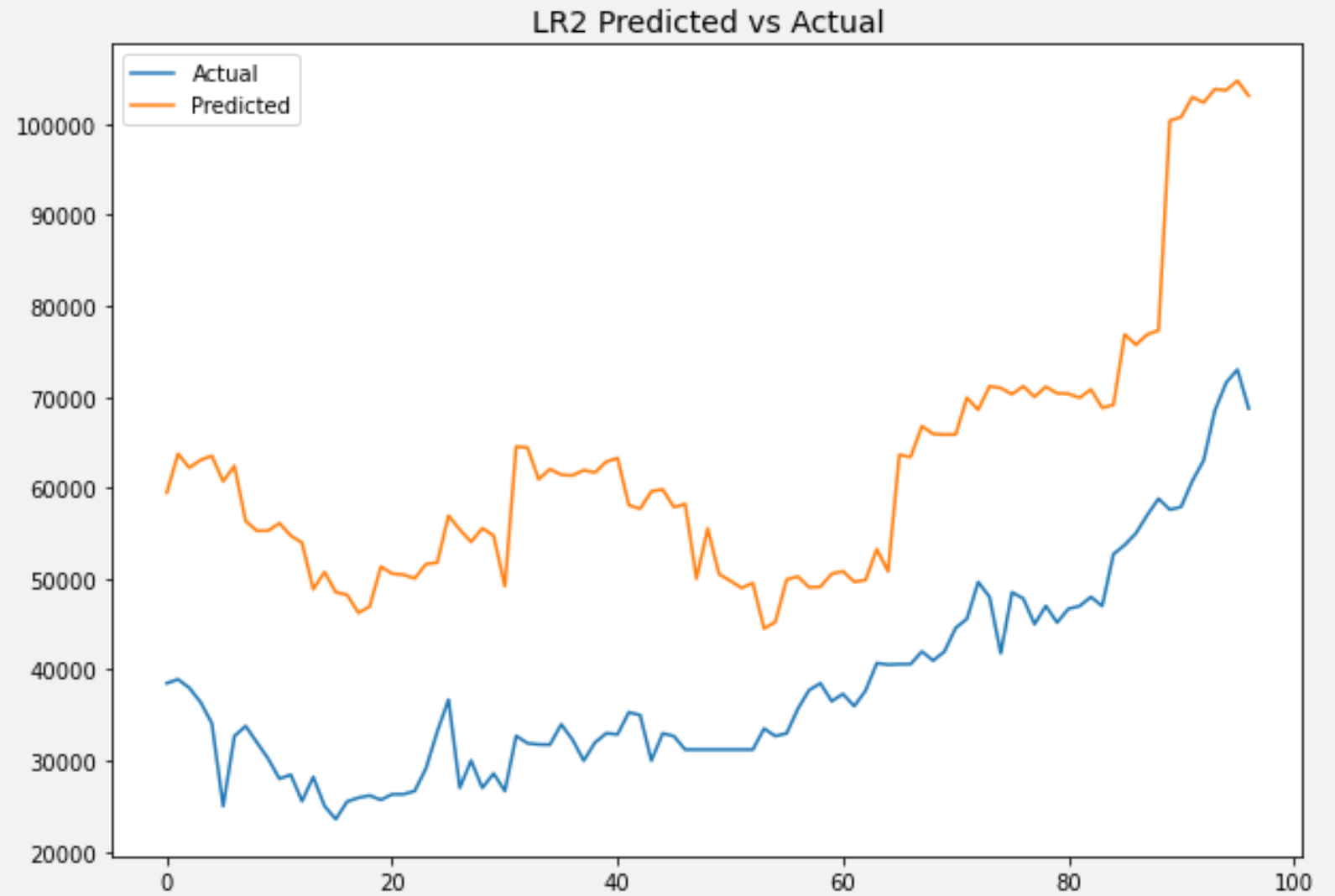
Model 1: Premium, Scarcity

Model 2: Premium, Quota, Deregistration of Cat A vehicle, Population Growth, STI close price

Model 3: Premium, Scarcity, Deregistration of Cat A vehicle, STI close price



## ERROR ANALYSIS – SUCCESS RATE



**MSE**

**Success %**

627,776,111.23

100

## ERROR ANALYSIS – SUCCESS RATE

- Not accurate to compare across different success rates.
- Using a function, I added a different increment each model so that all models have an 90% success rate.

# ERROR ANALYSIS – SUCCESS RATE

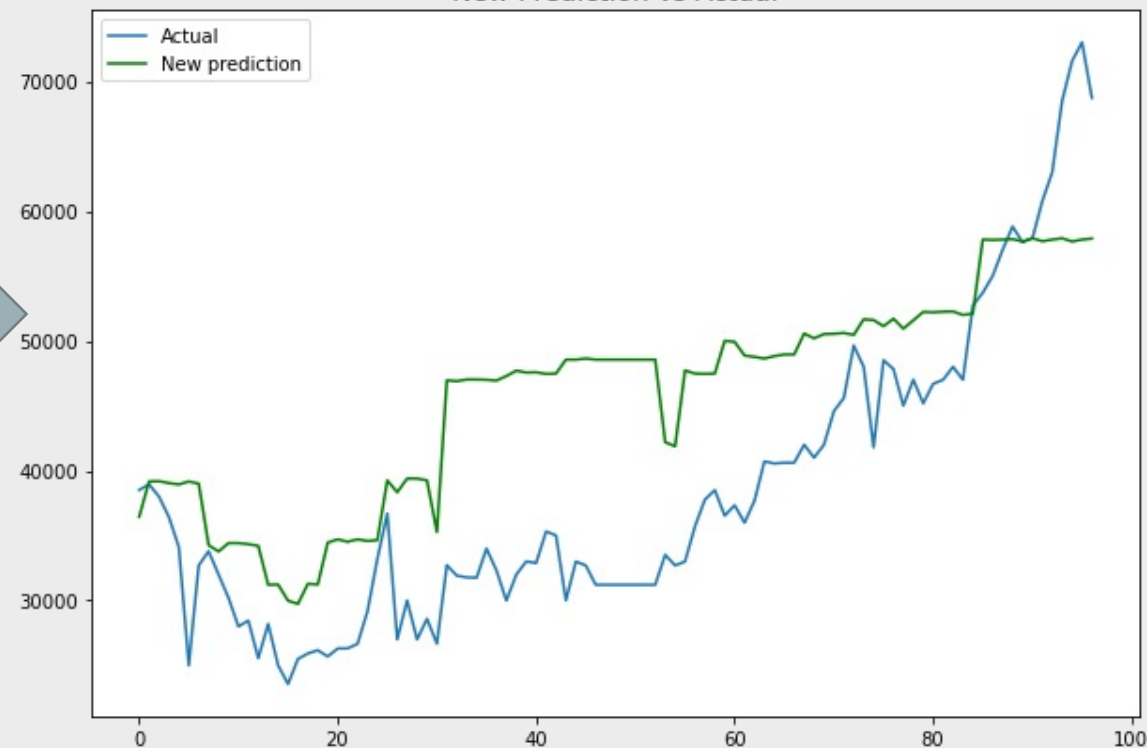
LR Predicted vs Actual



83.5%

+ \$1190

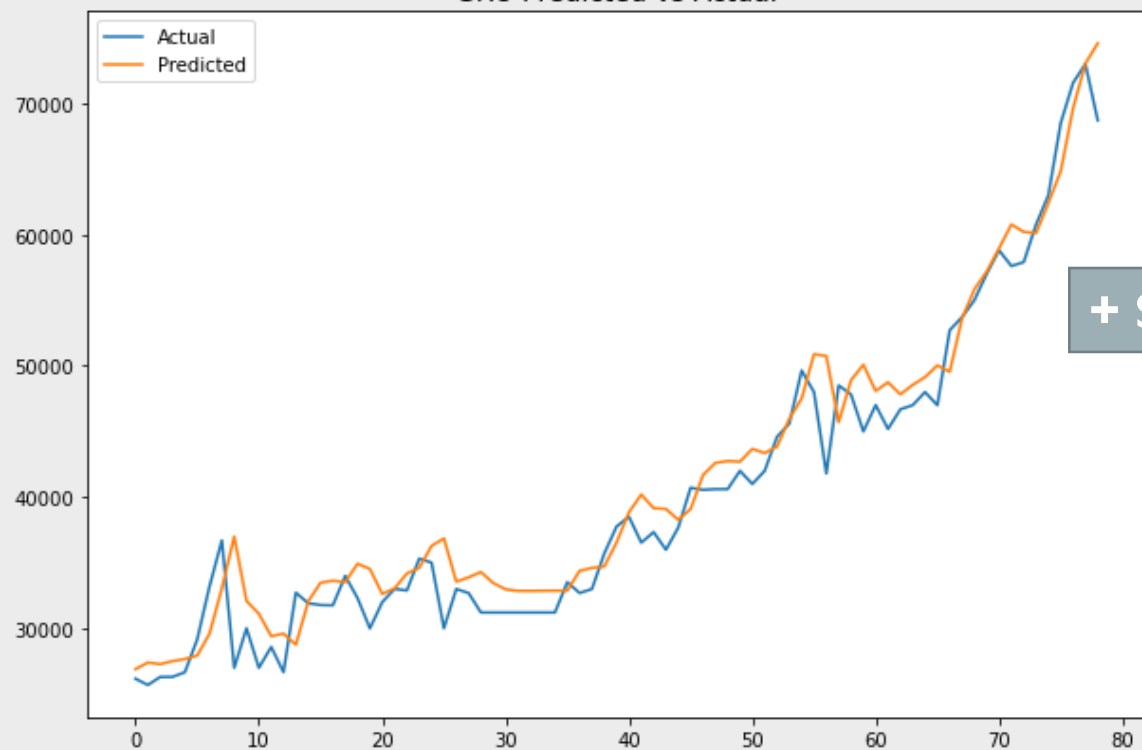
New Prediction vs Actual



90.7%

# ERROR ANALYSIS – SUCCESS RATE

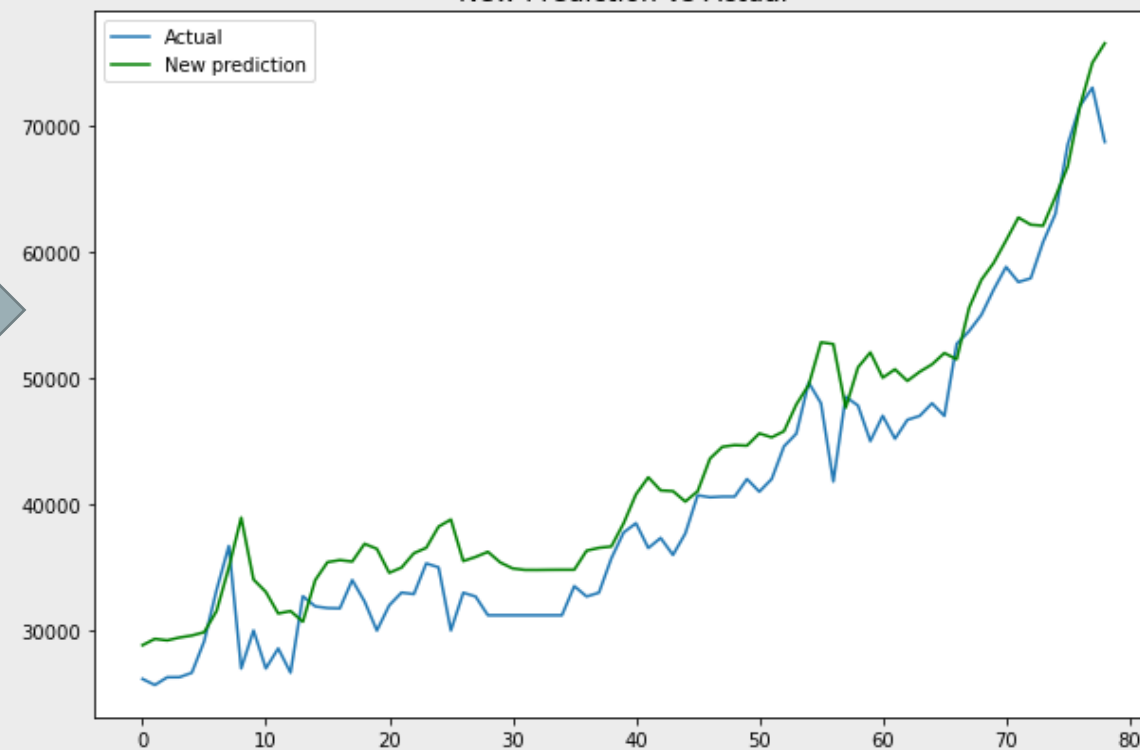
GRU Predicted vs Actual



75.9%

+ \$1943

New Prediction vs Actual



91.1%

## ERROR ANALYSIS – SUCCESS RATE

<b>Model</b>	<b>Average difference (Predicted vs Actual)</b>	<b>Standard Deviation</b>
Linear Regression	8830.4	5205.9
Linear Regression 2	11499.5	5687.4
GRU - Univariate	3558.0	2133.1
GRU - Multivariate 1	3849.1	2134.4
GRU - Multivariate 2	4738.3	2216.6
GRU - Multivariate 3	3892.5	2142.2

# MODEL SELECTION

- Lowest MSE
- High Success Rate with:
  - Low average difference to actual
  - Low standard deviation
- Eventual model selected:
  - Univariate GRU model



## PREDICTION FOR MAY'S BIDDING

- Using the model, the predicted premium: \$69,825.63
- To achieve 90% success rate: \$1943 is added to the prediction
- Final prediction = \$71,768.63
- Actual Premium = \$70,901



## FUTURE AREAS OF WORK

- Due to time constraints, other categories naively uses the same model for CAT A.
- COE's 10 years seasonality might have an effect but there are too little data points to show.
- Prediction for more than just the next cycle
  - Able to price COE packages according to predictions.





THANK YOU



ARIMA

COE premium with ARIMA(4, 1, 4) Predictions



**MSE**

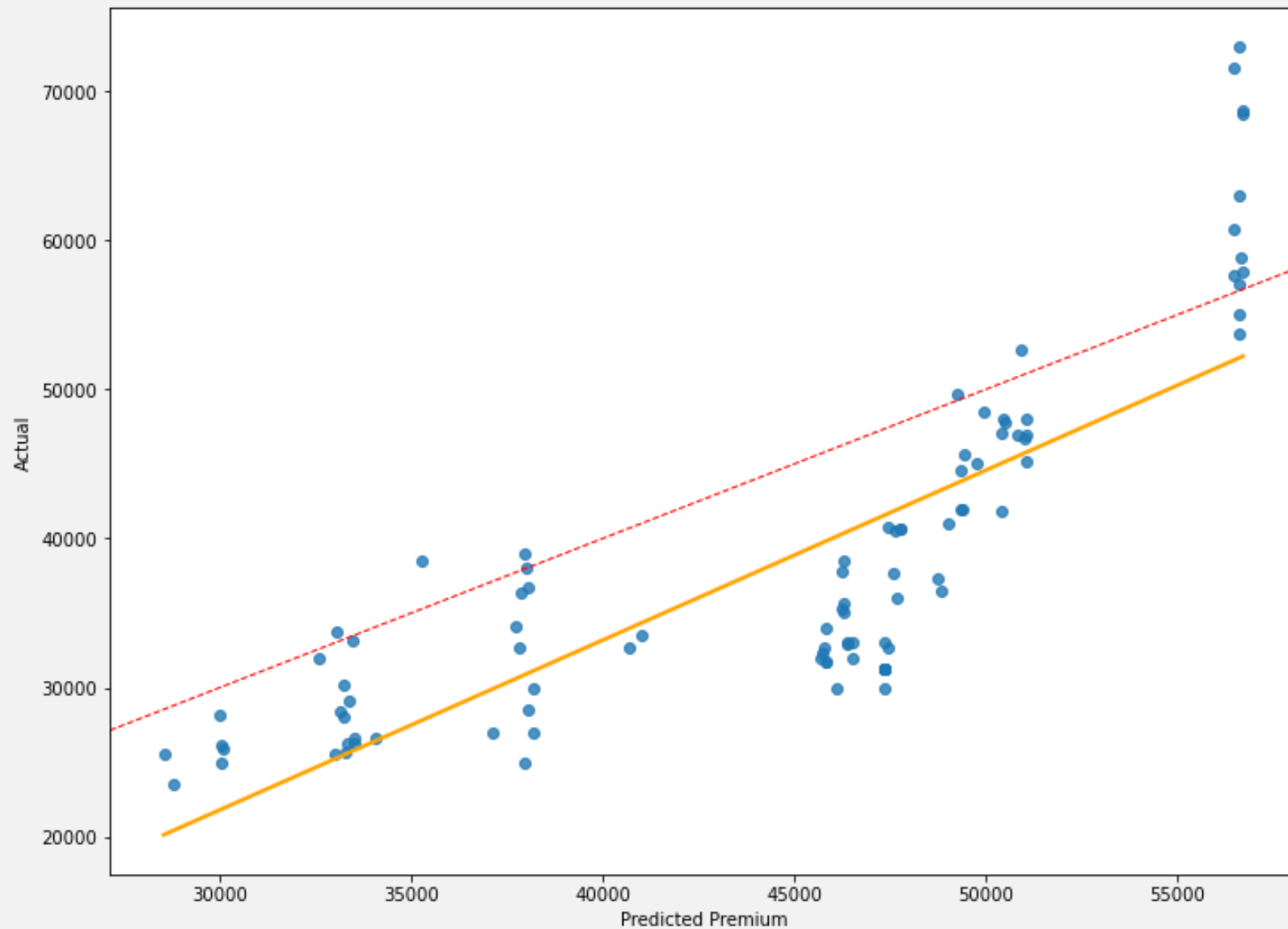
170,766,416.86

**Success %**

43.3



# LINEAR REGRESSION



MSE	Success %
86,105,841.98	83.5



SARIMA

COE premium with SARIMA(4, 1, 4) x (0, 1, 4, 18) Predictions



**MSE**

**Success %**

151,149,668.92

50.5



# UNIVARIATE GRU



MSE
7,453,959.03

Success %
75.9





# MULTIVARIATE GRU



MSE	Success %
8,823,055.69	20.3

