



Tsukiji Tuna Quantities

Tokyo Fish Market ("Tsukiji" to Toyosu)

Background:

- Established in 1935
 - Relocated following the Great Kanto Earthquake in 1923
 - Relocated again in 2018 for the Olympics
- The largest wholesale fish and seafood market in the world
- Hub for Culinary wonders, Marine Industry & Tourism



Tsukiji Tuna Dataset

January 2003 - December 2016

Quantity

Cumulative Monthly Quantity in Metric Tonnes
(1000 kgs)

1

Species

Bluefin Tuna, Bigeye Tuna, Southern Bluefin Tuna

2

State

Fresh, Frozen

3

Fleet

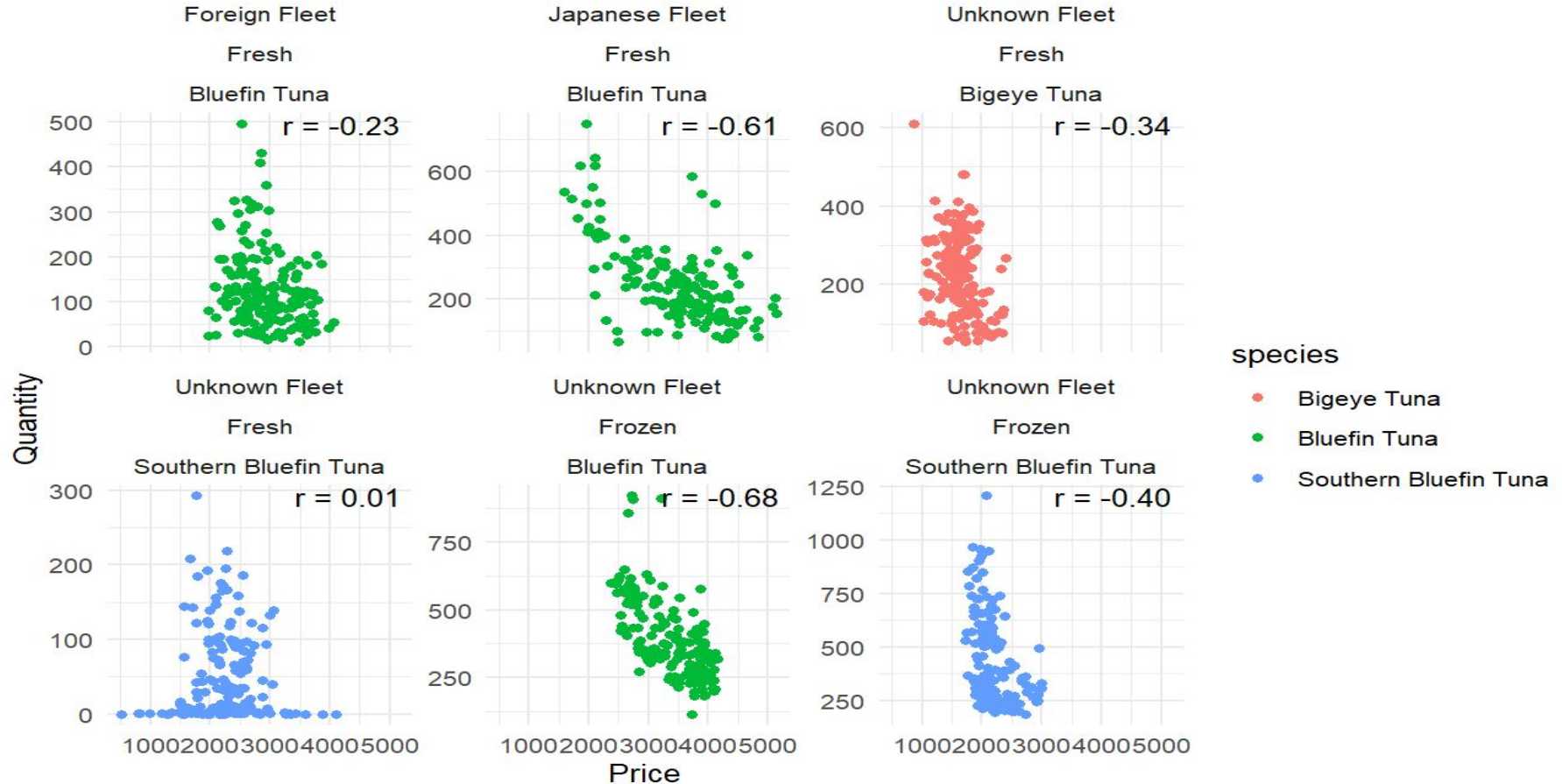
Japanese Fleet, Foreign Fleet, Unknown Fleet

4



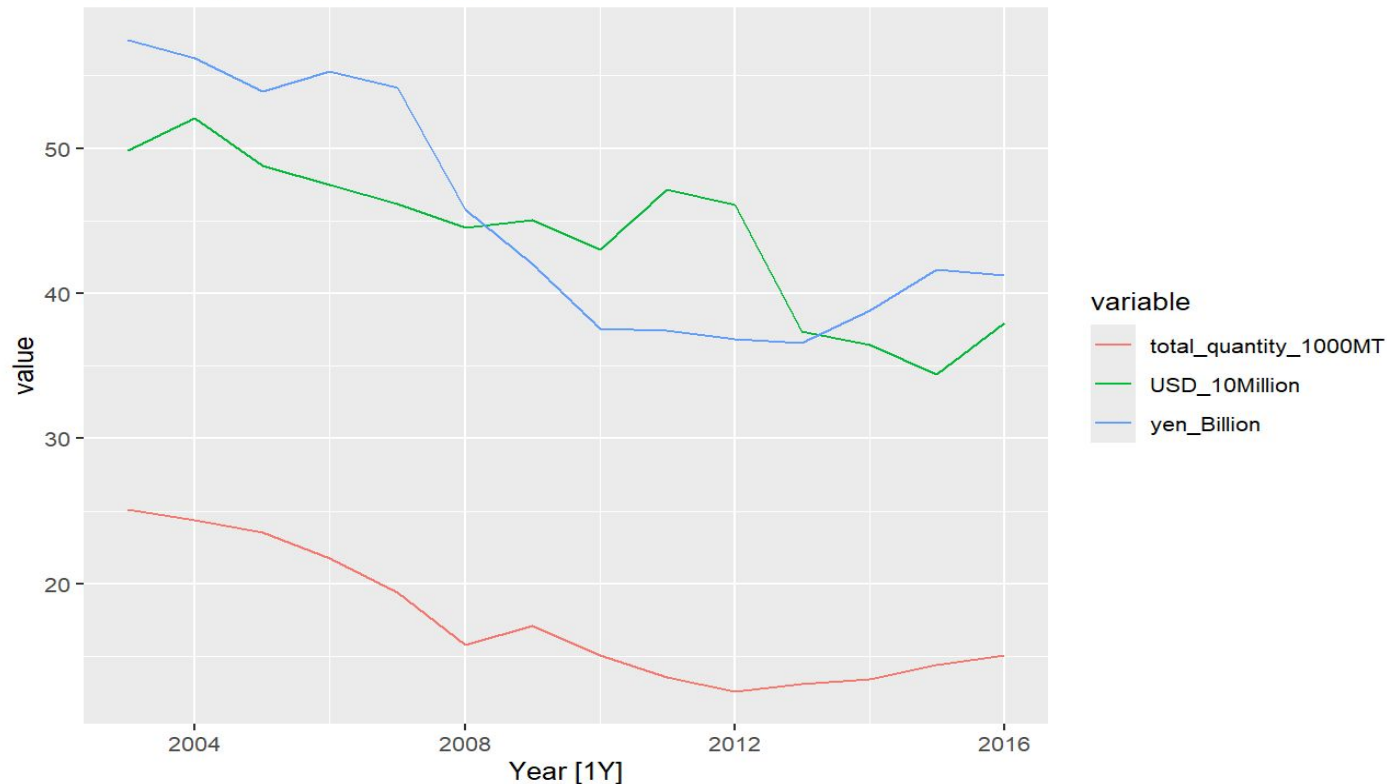
Tsujiki Tuna - Price v Quantity

Price vs Quantity-Correlation by Fleet, State, and Species

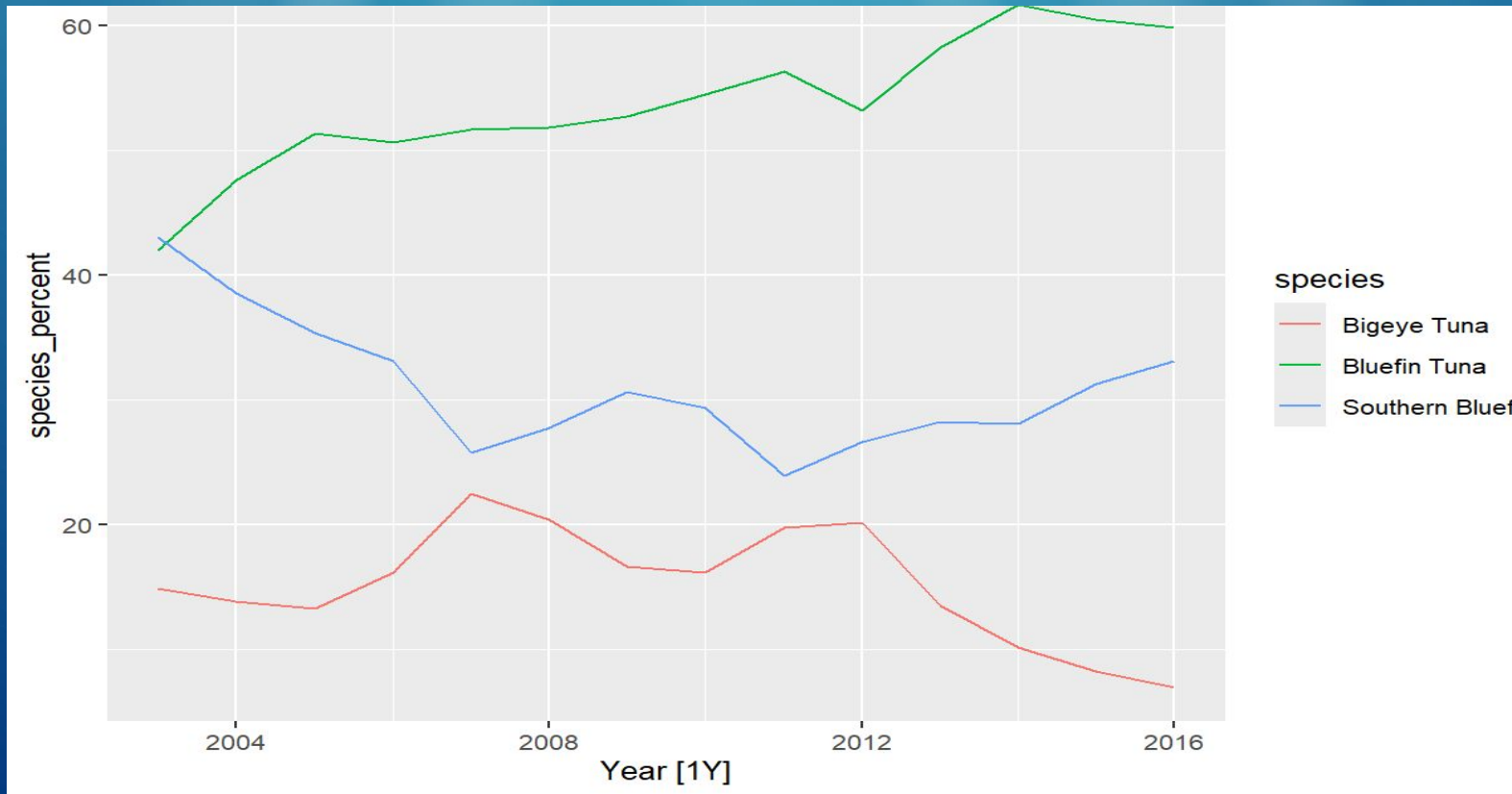


Tsujiki Tuna -Price Trends

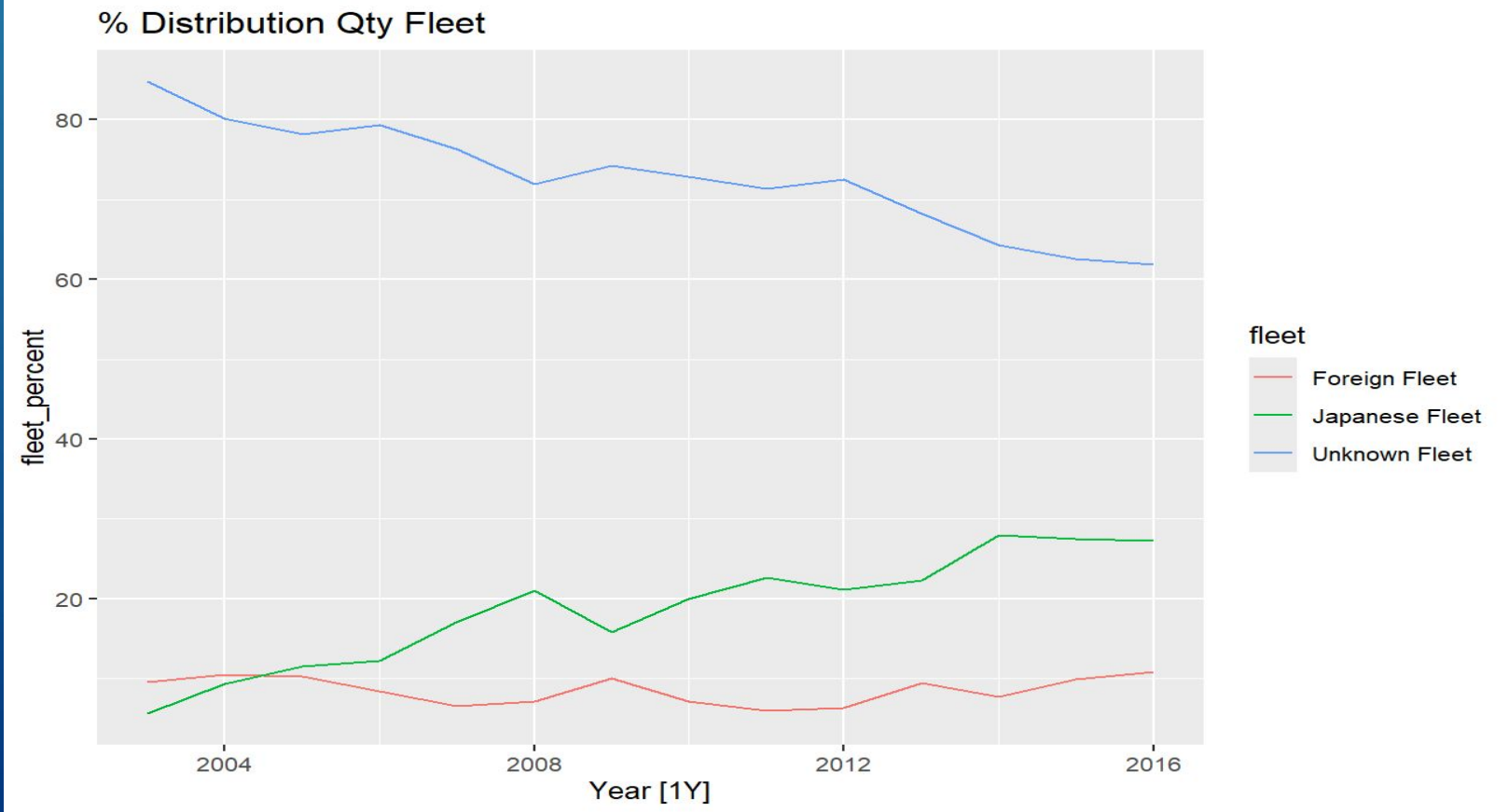
Yearly Qty and Price Trends



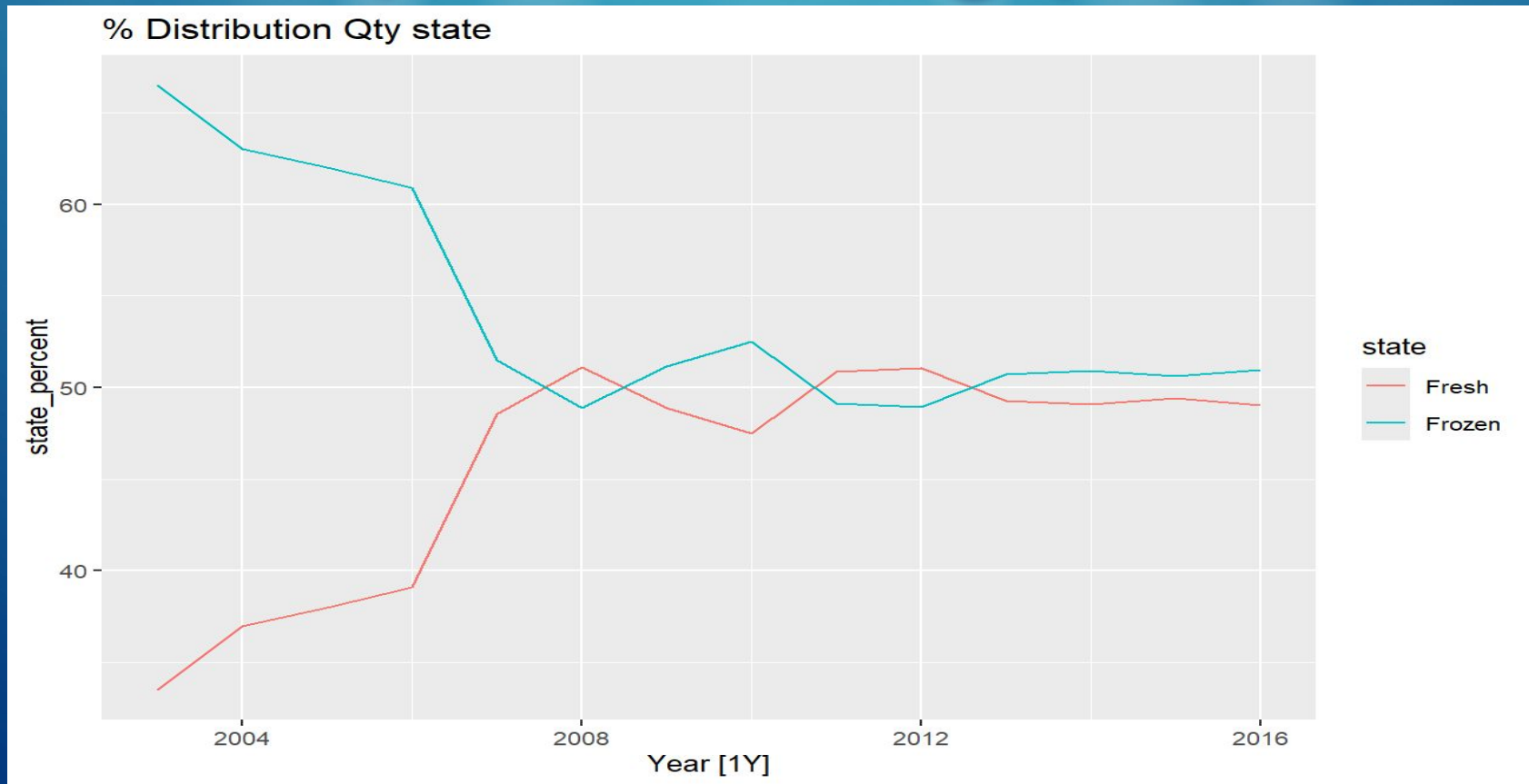
Species Trends



Fleet Trends



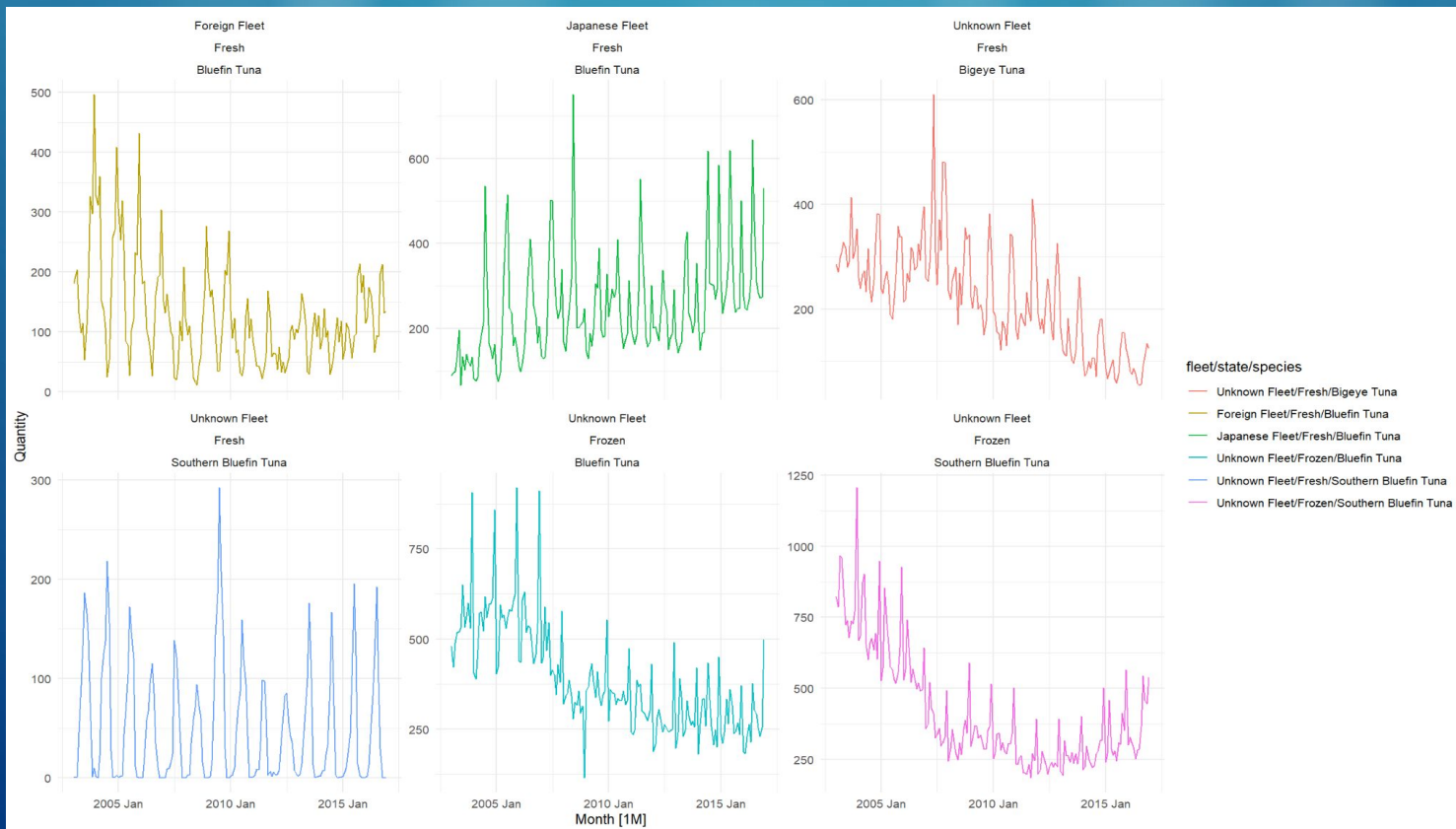
State (of fish storage) Trends





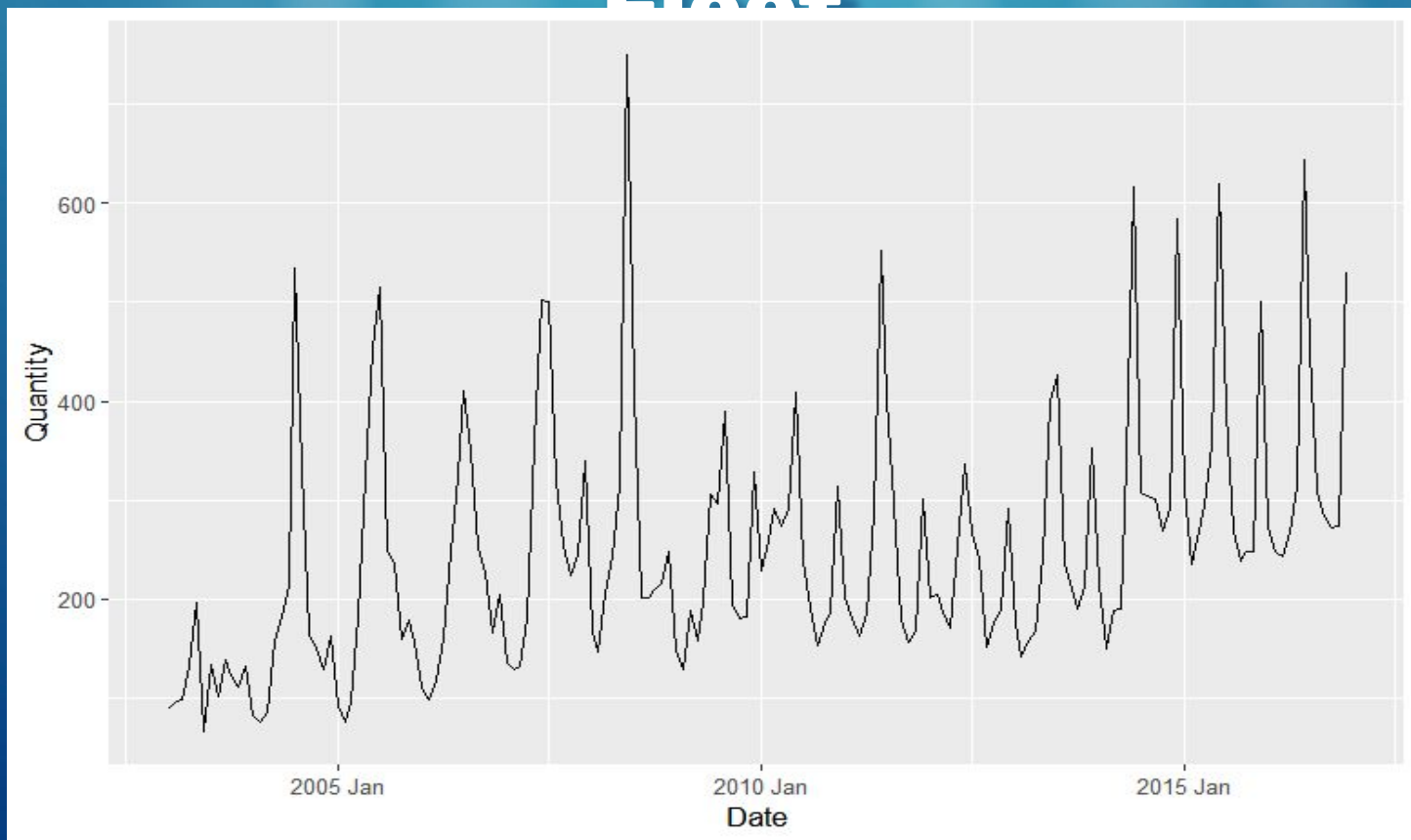
Tsukiji Tuna Dataset

January 2003 - December 2016



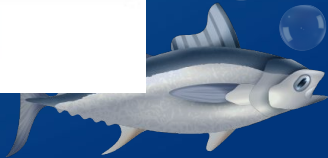
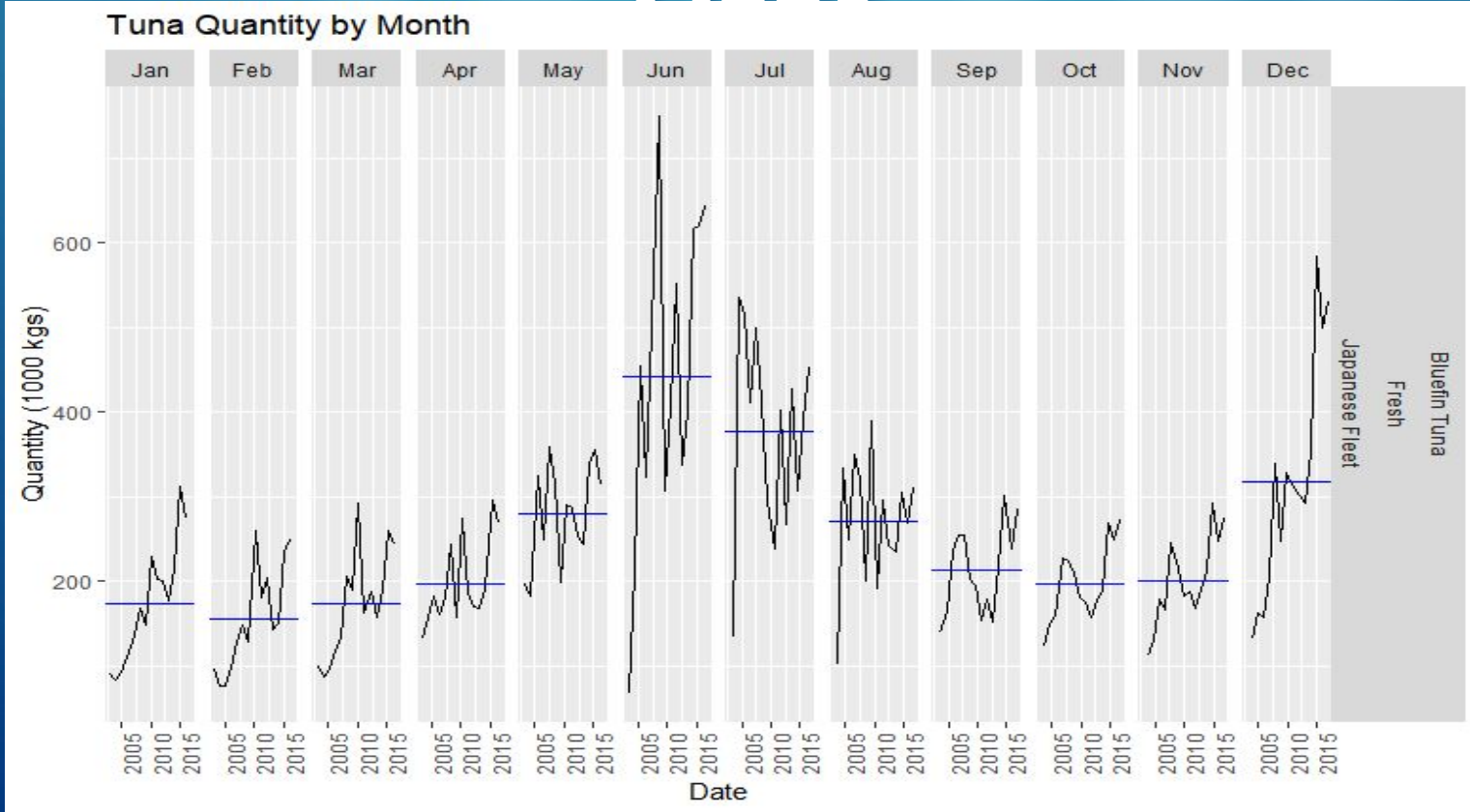


Bluefin Tuna, Fresh, Japanese Fleet



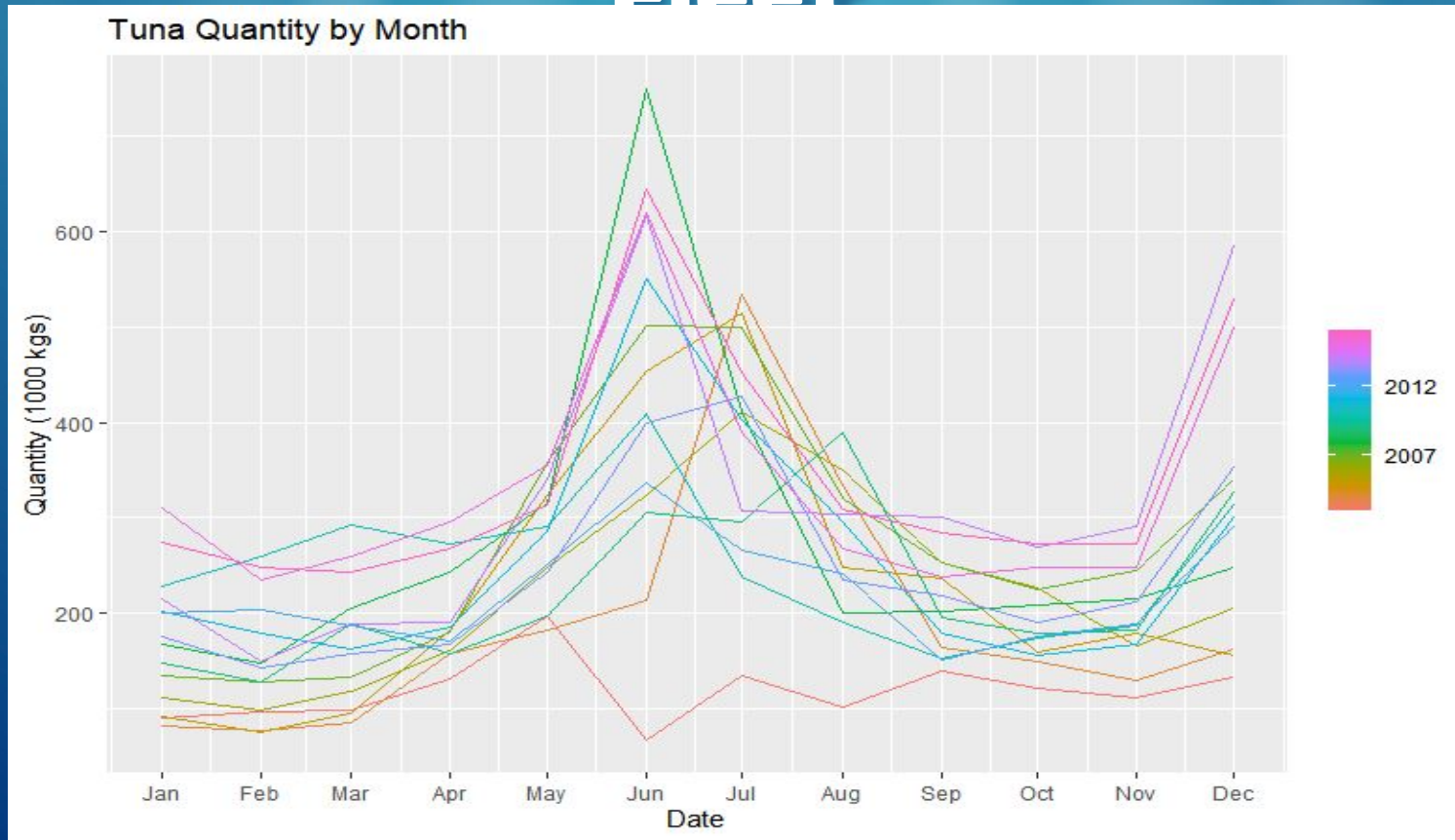


Bluefin Tuna, Fresh, Japanese Fleet



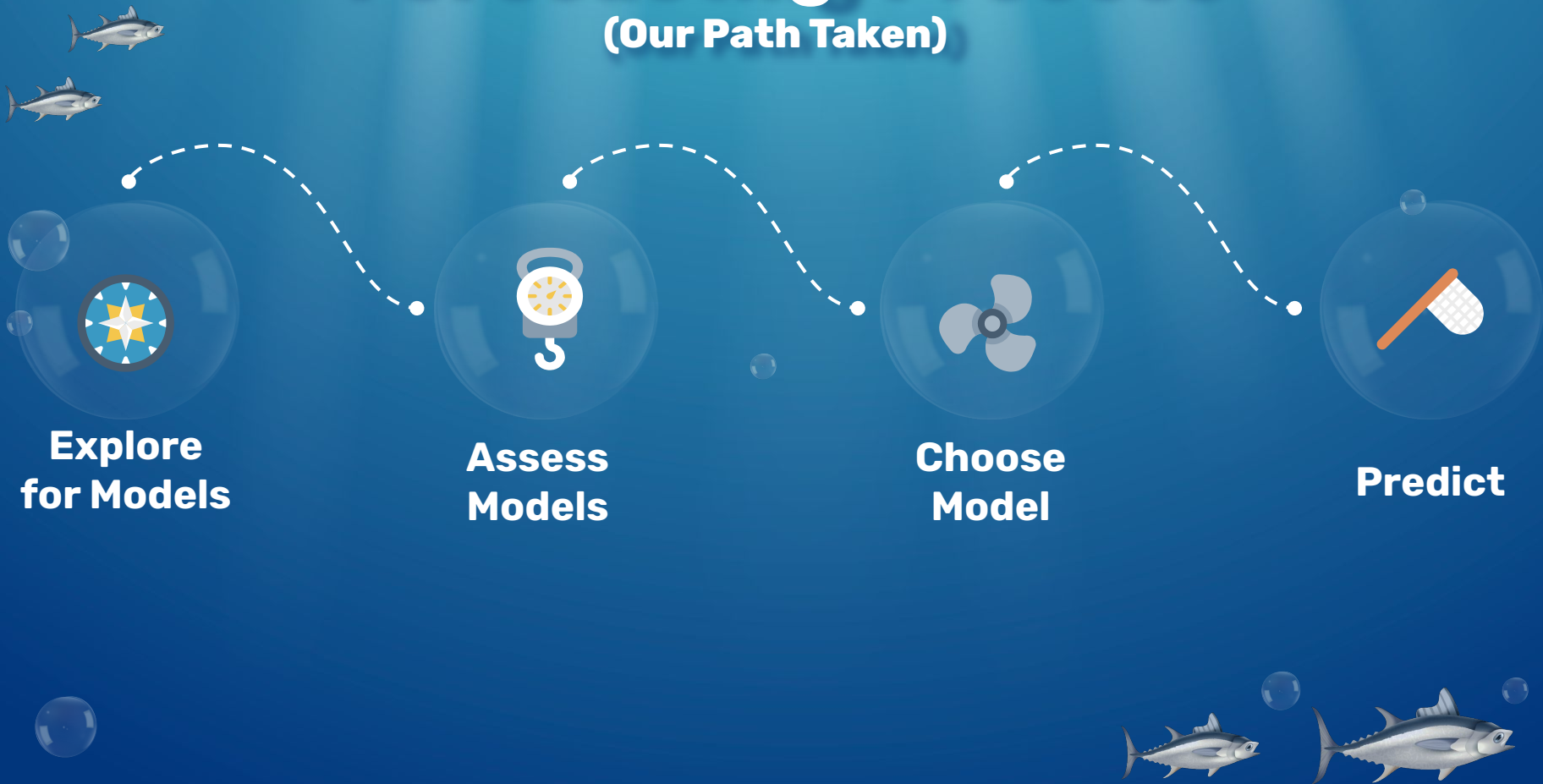


Bluefin Tuna, Fresh, Japanese Fleet



Forecasting Process

(Our Path Taken)



1 Explore for Models

Viewing the STL decomposition can help understand the trend and seasonal components of the data to assist with model selection.

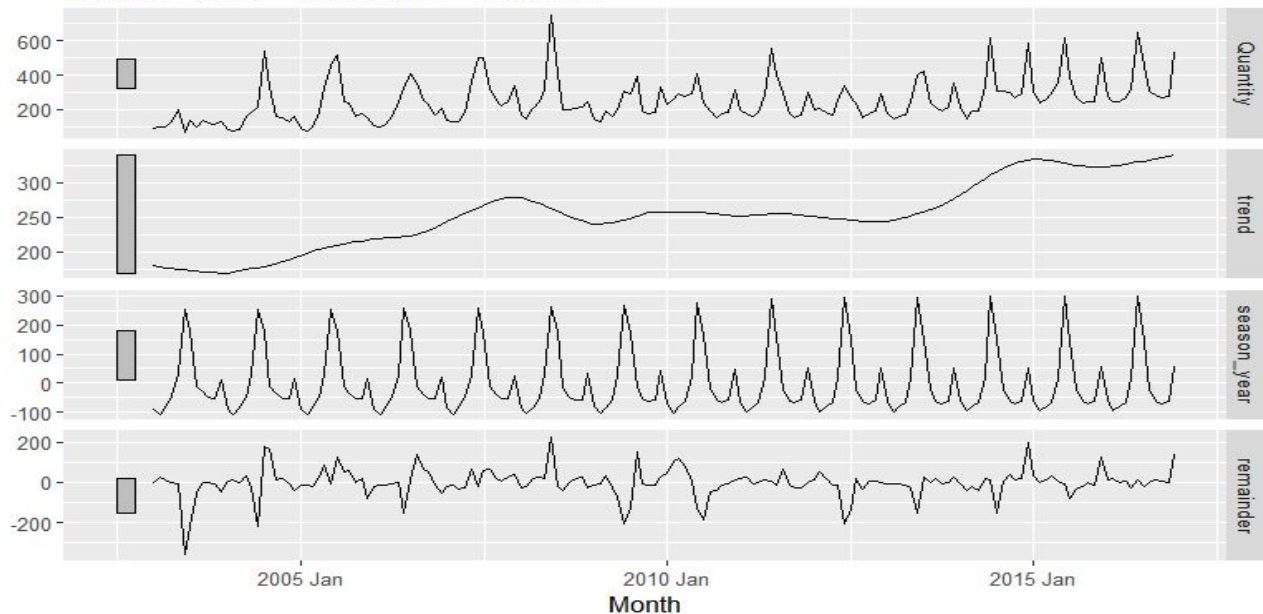
Because it appears that the trend is somewhat **linear** and there is apparent **seasonality**, we will move forward with the following models:

- auto ETS model
- the auto ARIMA model
- a time series linear regression model
- seasonal naive model.

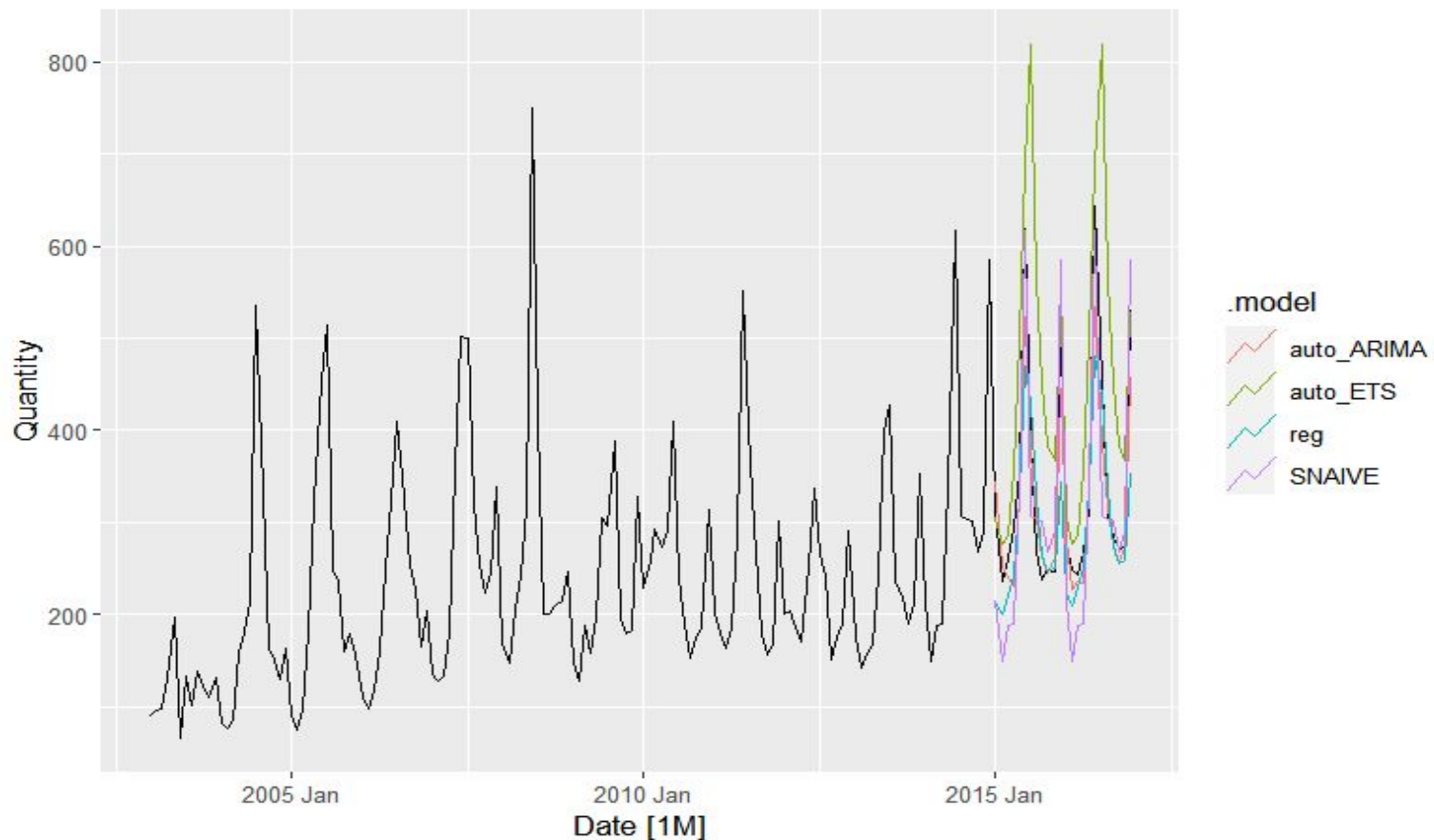
```
tuna %>%  
  model(STL(Quantity ~ trend(window=24)+season(window=12),  
          robust = TRUE)) %>%  
  components() %>%  
  autoplot()
```

STL decomposition

Quantity = trend + season_year + remainder



2 Assess Models



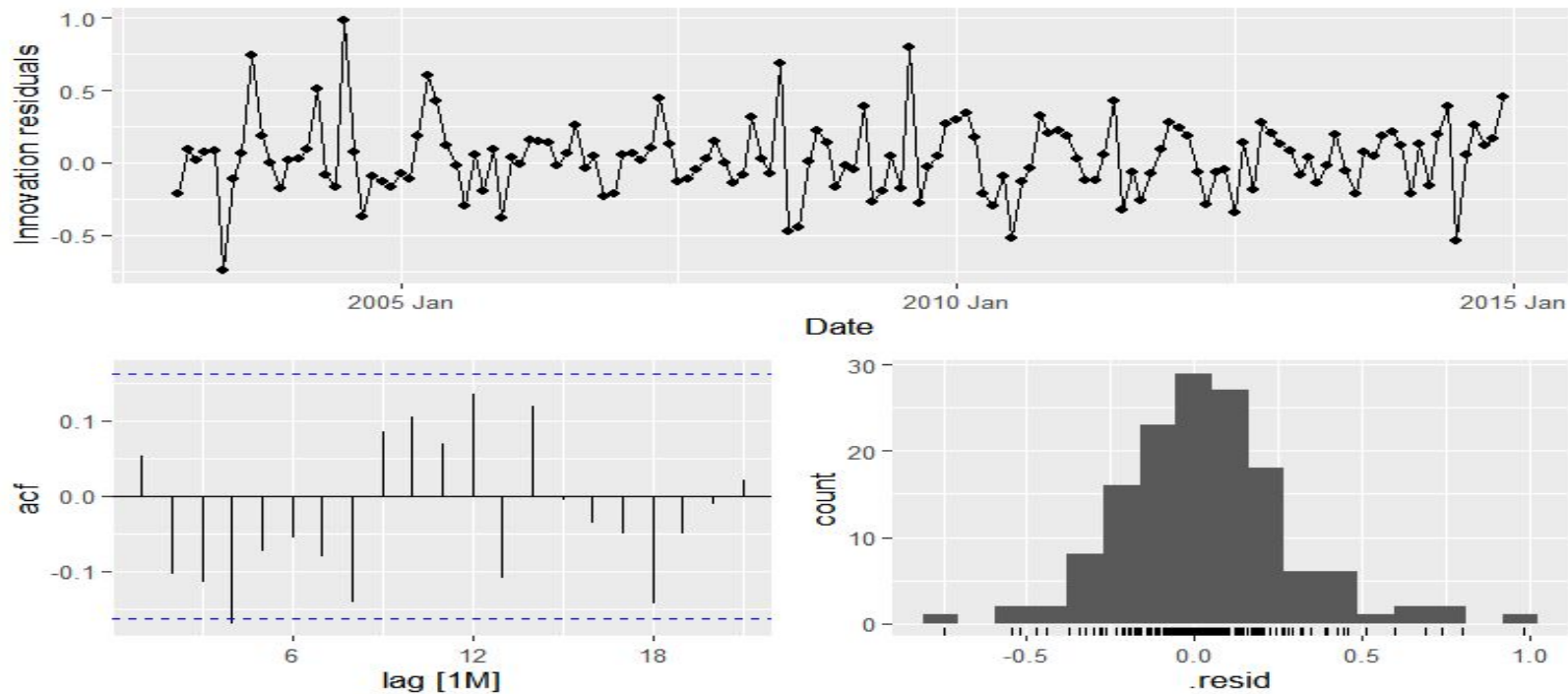
2 Assess Models

| Model | RMSE | MAE | MPE | MAPE |
|---------------------------------------|------|------|-------|------|
| ETS(M,N,M) | 168 | 124 | -39.6 | 39.8 |
| SNAIVE | 65.6 | 54 | 8.12 | 17.6 |
| REG | 74.3 | 52.2 | 8.56 | 13.6 |
| ARIMA(1,0,0)(0,1,1) [12] w/ drift> | 43.5 | 32.3 | 3.64 | 8.58 |



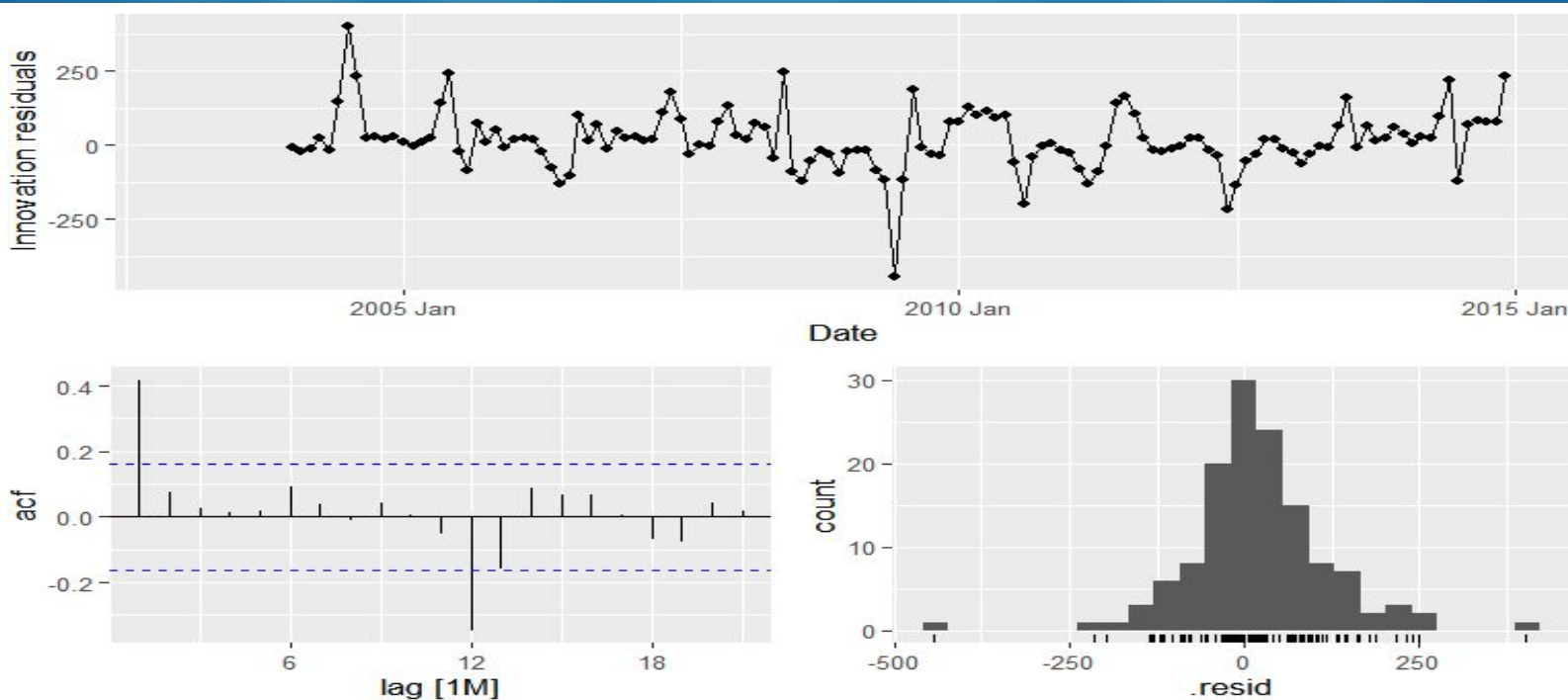
2 Assess Models - ETS(M,N,M)

| Model | RMSE | MAE | MPE | MAPE |
|------------|------|-----|-------|------|
| ETS(M,N,M) | 168 | 124 | -39.6 | 39.8 |



2 Assess Models - Seasonal Naive

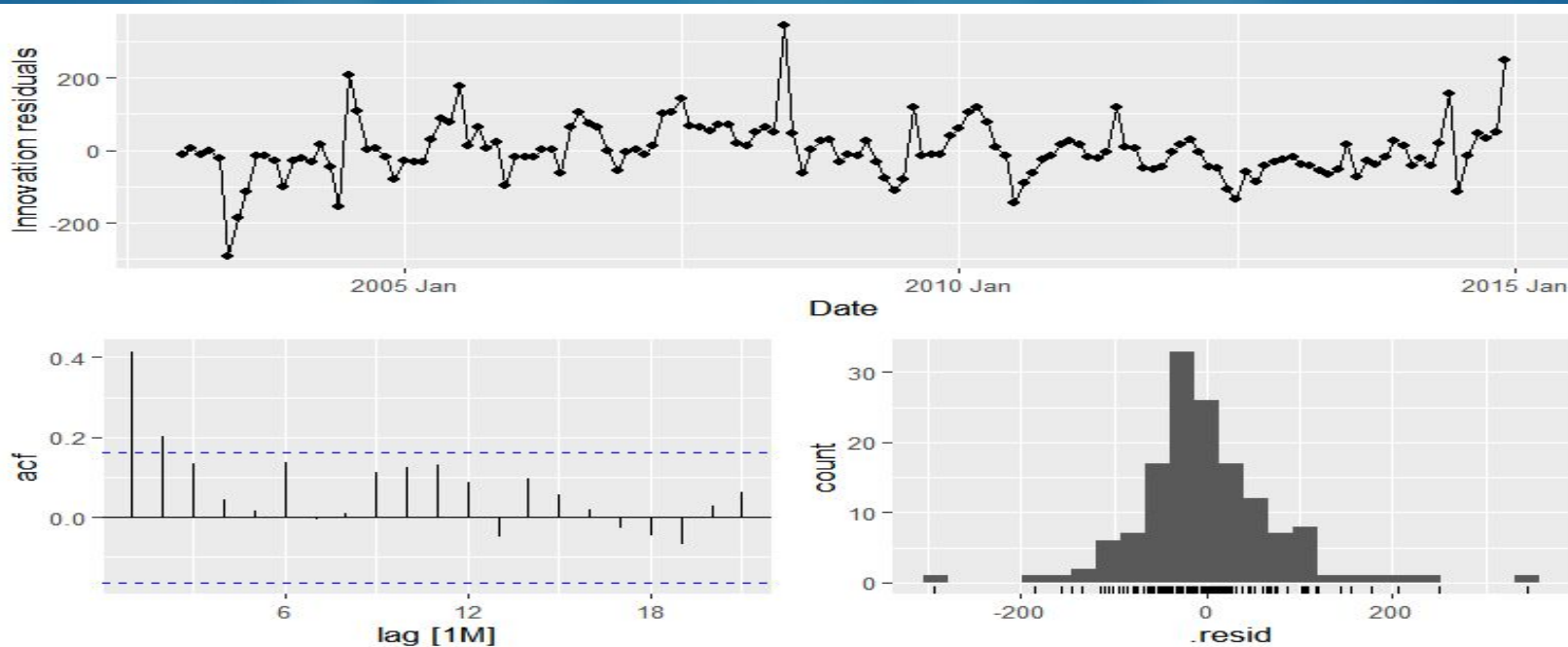
| Model | RMSE | MAE | MPE | MAPE |
|--------|------|-----|------|------|
| SNAIVE | 65.6 | 54 | 8.12 | 17.6 |



2

Assess Models -Time Series Linear Regression

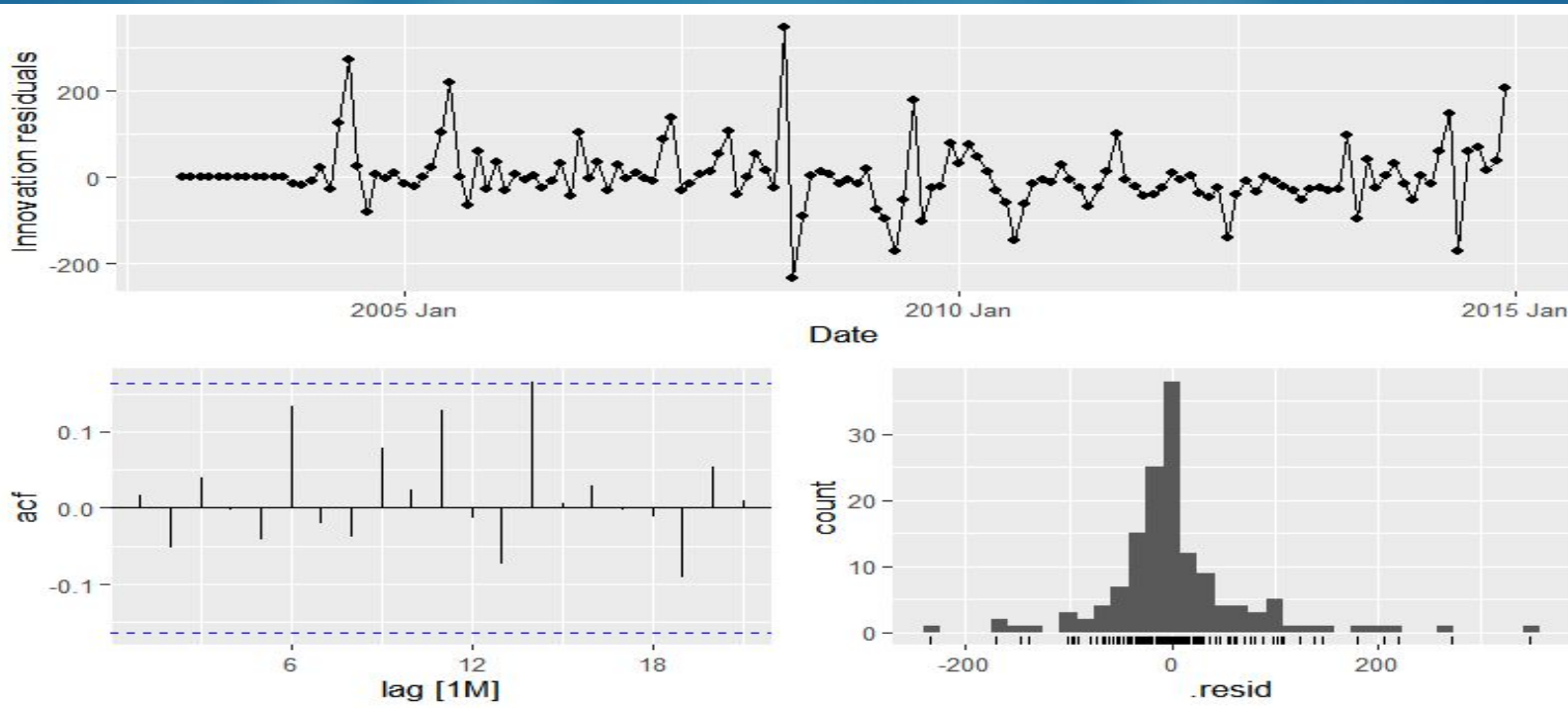
| Model | RMSE | MAE | MPE | MAPE |
|-------|------|------|------|------|
| REG | 74.3 | 52.2 | 8.56 | 13.6 |



2

Assess Models - ARIMA

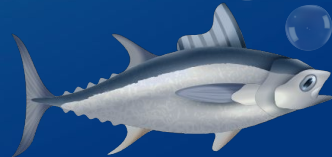
| Model | RMSE | MAE | MPE | MAPE |
|-------|------|------|------|------|
| ARIMA | 43.5 | 32.3 | 3.64 | 8.58 |



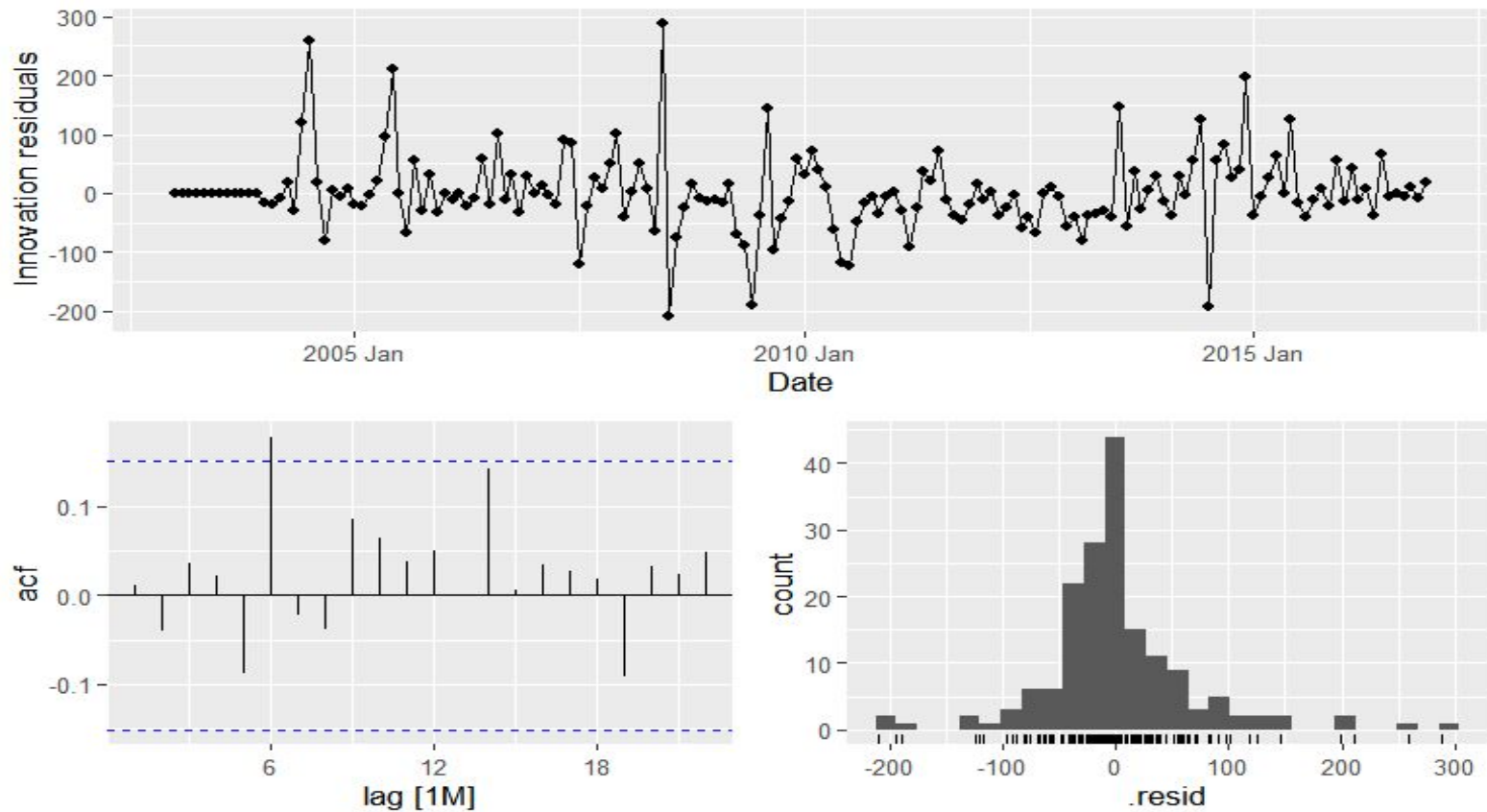
2

Assess Models - Cross Validation

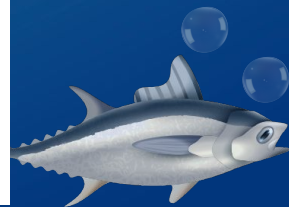
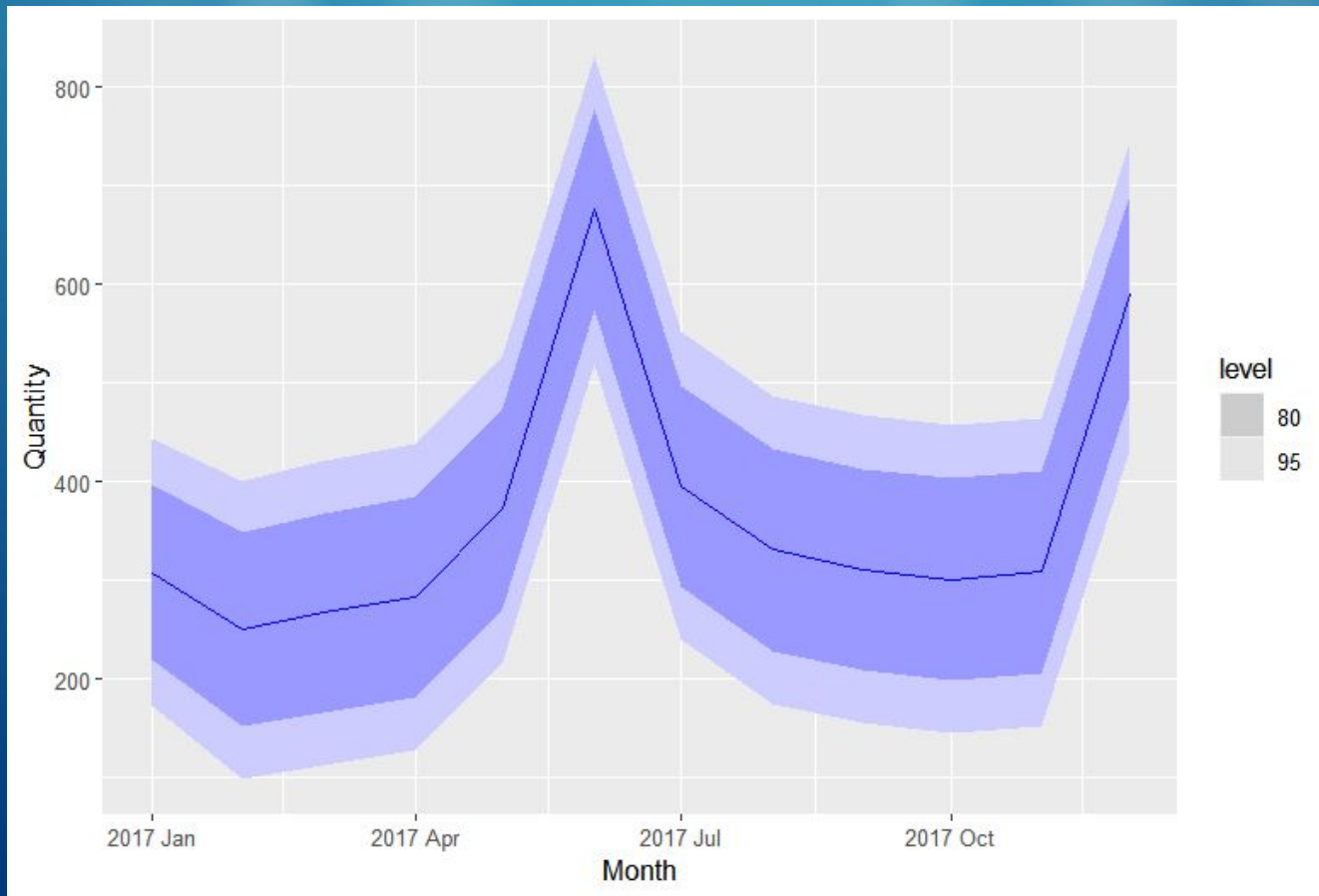
| Model | RMSE | MAE | MPE | MAPE |
|------------|------|------|-------|------|
| Auto_ARIMA | 88 | 55.6 | -12.9 | 19 |
| TSLR | 93.9 | 67.6 | -20.9 | 27.7 |



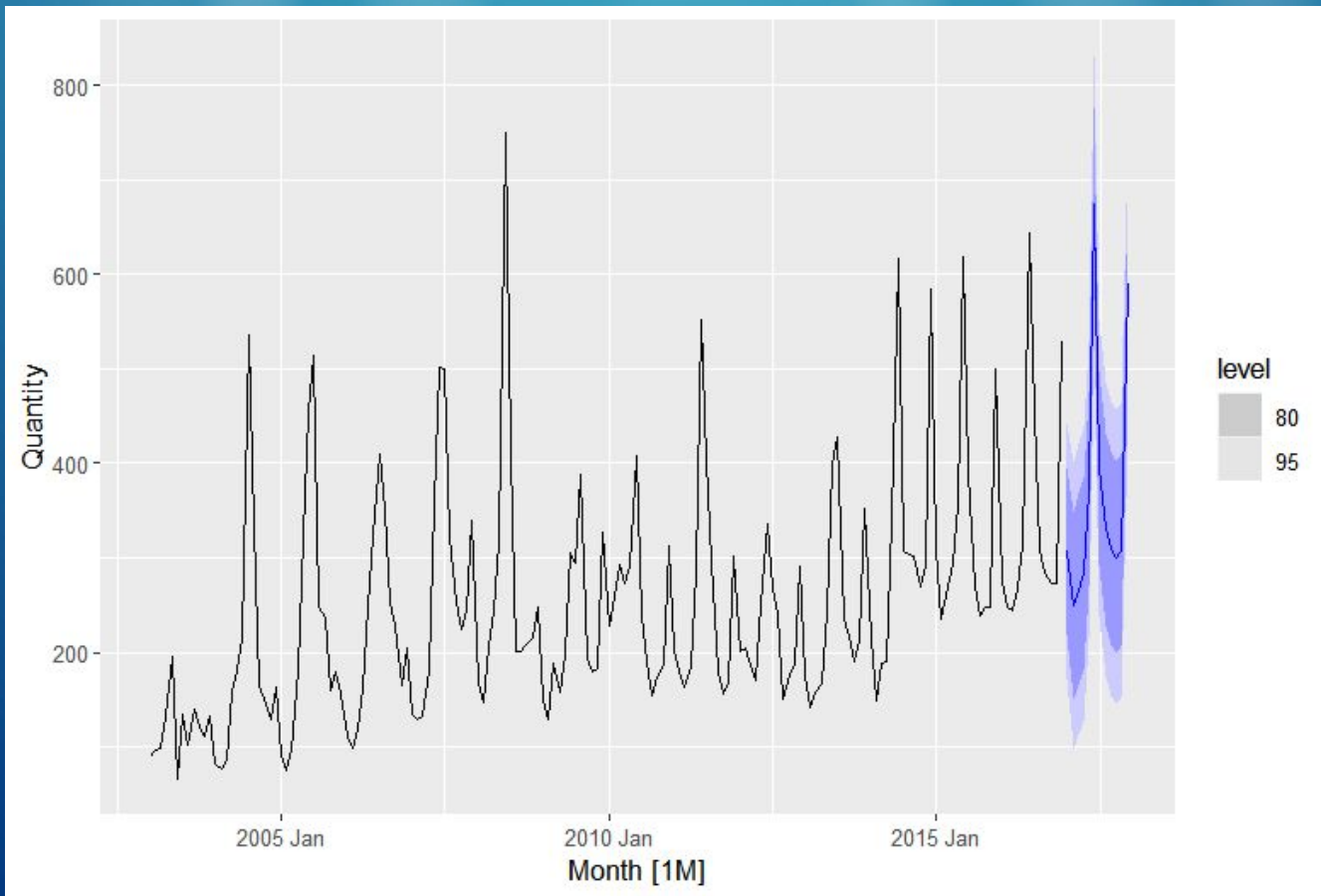
3 Choose Final Model - ARIMA



4 Predict



4 Predict



4 Predict

| Date | .mean | 95%_lower | 95%_upper |
|----------|----------|-----------|-----------|
| 2017 Jan | 307.7826 | 173.0202 | 442.5451 |
| 2017 Feb | 250.0182 | 99.19982 | 400.8367 |
| 2017 Mar | 266.6017 | 111.9929 | 421.2105 |
| 2017 Apr | 283.249 | 127.6978 | 438.8003 |
| 2017 May | 372.0632 | 216.275 | 527.8515 |
| 2017 Jun | 675.2784 | 519.4304 | 831.1265 |
| 2017 Jul | 395.5412 | 239.678 | 551.4043 |
| 2017 Aug | 330.3864 | 174.5195 | 486.2534 |
| 2017 Sep | 310.6118 | 154.7439 | 466.4797 |
| 2017 Oct | 300.8849 | 145.0168 | 456.7531 |
| 2017 Nov | 307.8657 | 151.9975 | 463.7339 |
| 2017 Dec | 589.3663 | 433.4981 | 745.2346 |



Questions and or comments?