

Time series analysis on the stock price of Tesla Inc.

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1 Abstract

2 Introduction

3 Data Description

4 Exploratory Data Analysis

To obtain a comprehensive understanding of the data, we conduct explanatory data analysis (EDA) first. Figure 1(a) is the time series plot of all the given time points. We observe that the stock prices of Tesla before 2020 are averagely and considerably lower than those after 2020. The significantly different scales of different parts of the time series make it hard to visually examine the trend and seasonality pattern of the time series. Moreover, since we are majorly interested in the recent activities of Tesla, we do not have to analyze all the available data. Therefore, for the sake of interest and convenience, we decide only to analyze the last 300 time points, which cover the period from 2020-08-26 to 2021-11-02 excluding weekends. Thus, whenever we use the word “data” in the following analysis, we implicitly mean the time series of the last three hundred time points.

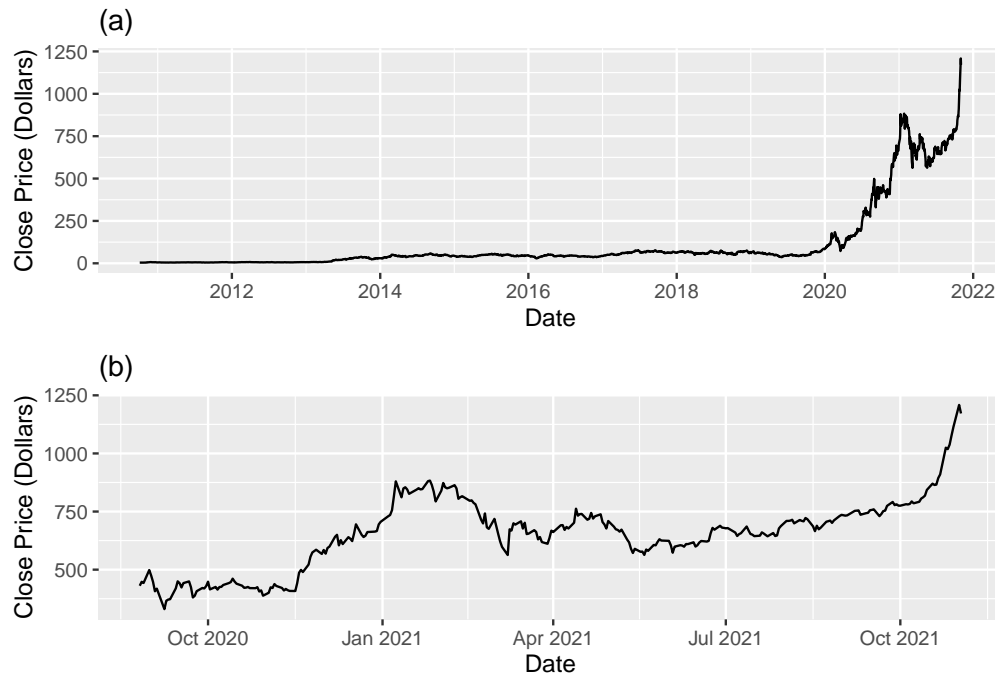


Figure 1: (a) Time series plot of all available trading days. (b) Time series plot of last 300 trading days

Figure 1(b) is the time series plot of the close prices of Tesla in the last three hundred trading days before and including 2021-11-02. We first observe that our data is roughly homoscedastic based on Figure 1(b). To verify our observation, we try the square root and natural log transformations and see whether they effectively stabilize the variance of the time series. Their plots are below in Figure 2.

We can see that both transformations unnecessarily increase the variance of the time series before mid-November in 2020 and do not change the variance of other time series data. Although both transformations shorten the vertical distance between the maximum and minimum of the time series after Oct. 2021, the spike after Oct. 2021 is more like an increasing trend than a considerable fluctuation. In short, both transformations are redundant, and we do not need to use any variance stabilizing transformation.

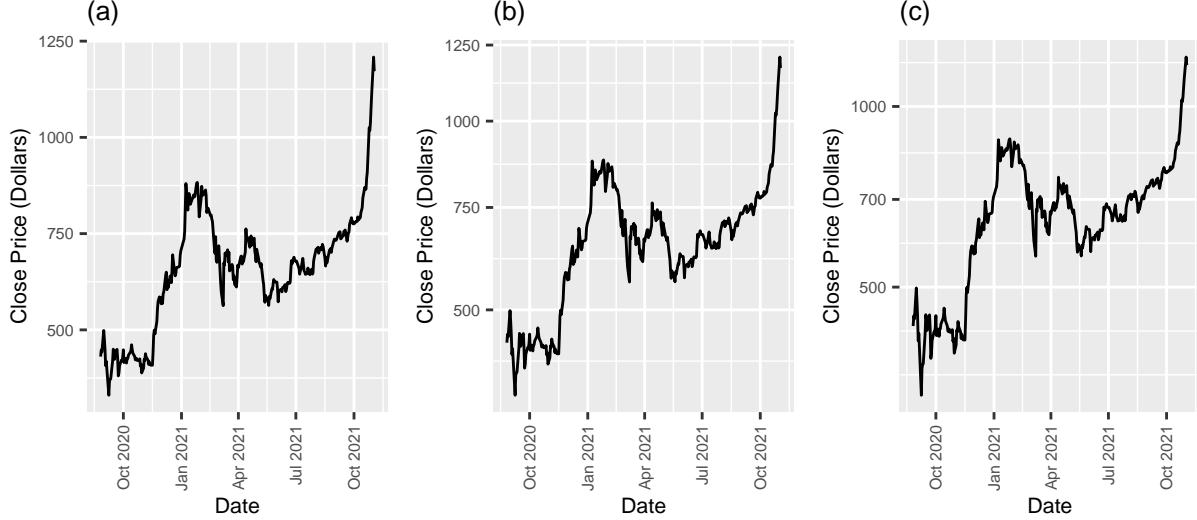


Figure 2: (a): Original time series. (b): Square root transformed time series. (c): Natural log transformed time series.

Back to Figure 1(b), intuitively, the data is not stationary because of a nonlinear and generally increasing trend. The trend first increases until around Feb. 2021 and then decreases until around Mid-May. 2021. Finally, the trend increases again until the end of the time series. Nonetheless, we do not observe an obvious or significant seasonality pattern. It matches the intuition since the granularity of our data is day, and the structure of stock price data is too complicated to have a seasonality pattern.

In conclusion, based on all the previous discussions in EDA, we decide to construct possible models on the original time series data, including only the last three hundred time points.

5 Model Construction

With a comprehensive understanding of our data, we start to experiment and construct different time series model. We choose and build two non-parametric signal models of the trend and seasonality in our data. We aim to make the residuals approximately weekly stationary. We do not consider any parametric trend model because we think the trend of the stock price data is too complicated to be modeled by a parametric model, such as a high-order polynomial. Certainly, we could use a 15 or 20 order polynomial, but it may overfit the training data and produce imprecise predictions. We do not consider a parametric seasonality model either because we do not find a clear seasonality pattern in our data by the EDA. Finally, based on each signal model, we provide two ARMA models or its extension, such as SARMA or ARIMA, to whiten the residuals of the signal model. Thus, we have four candidate models, and we will explain how we select a final model among them in the next section.

5.1 Non-parametric Signal Model: exponential smoothing

In this signal model, we choose exponential smoothing with weight $\alpha = 0.8$ and lag $k = 10$ and a seasonal differencing with period $d = 5$.

We experiment with different combinations of α and k with a careful consideration of overfitting issue. we choose $k = 10$ as the final value because we want to only use past two weeks, which are ten days in our data, to forecast. We choose $\alpha = 0.8$ as the final value because we think it best balances the smoothing effect and the capture of trend pattern among $(0, 1)$. Indeed, the smoothing line in Figure 4(a) fits the data in the way

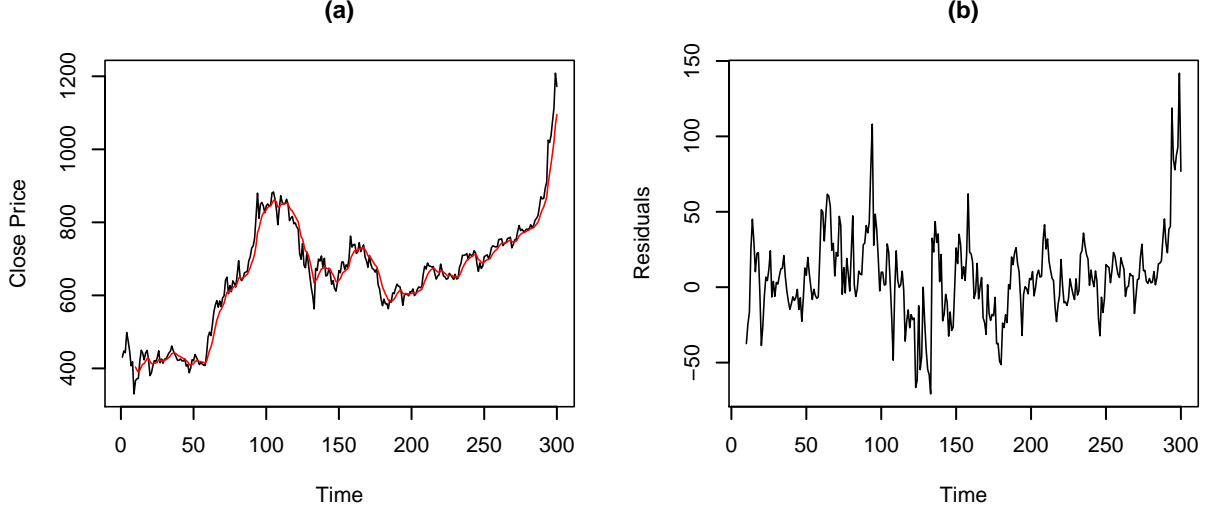


Figure 3: (a): Time series plot of the original data and fitted values. (b): The residual plot of exponential smoothing.

that we want. Note that we lose the first nine time points due to the computation process of the exponential smoothing.

However, the residual plot Figure 4(b) is fairly non-stationary, as it has cycling fluctuation pattern and still slightly nonlinear trend. It might be due to that we intentionally let exponential smoothing not fit the data perfectly. Next, We use the seasonal differencing with period $d = 5$, which is one week in our data, to further make the residuals more stationary.

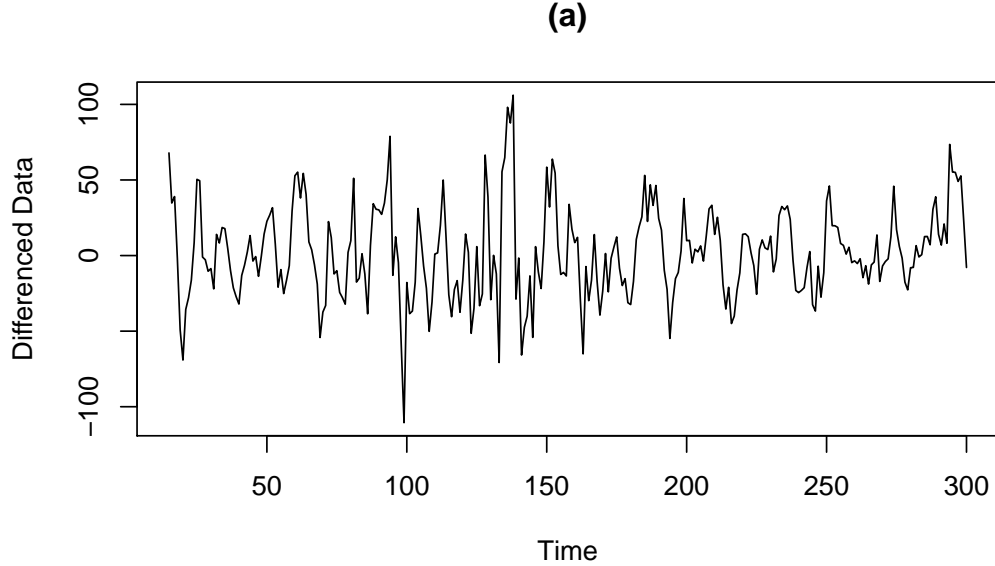


Figure 4: (a): Time series plot of the seasonal differenced ($d = 5$) residuals from the previous smoothing.

Indeed, now the differenced residuals become more stationary. There seems to be a contradiction that recalling in EDA, we claim that there is not a clear seasonality in our data. However, the effect of the seasonal differencing here implies a possible seasonality with period $d = 5$. We think it is because the seasonal differencing is actually removing the remaining trend left by the exponential smoothing instead of the seasonality. Nevertheless, We believe that the time series of the differenced residuals shown in Figure 5(a) is stationary enough for us to build ARMA models on it.

5.2 Non-parametric Signal Model: second-order differencing

In this model, we choose the second-order differencing to remove the trend. We observe that after the first-order differencing, there is still some trend pattern, such as the increasing one between 270 and 300, as shown by Figure TODO. This matches our previous analysis that the trend of our data is nonlinear in EDA. Thus, we take another differencing and acquire the second-order differencing data shown in Figure TODO.

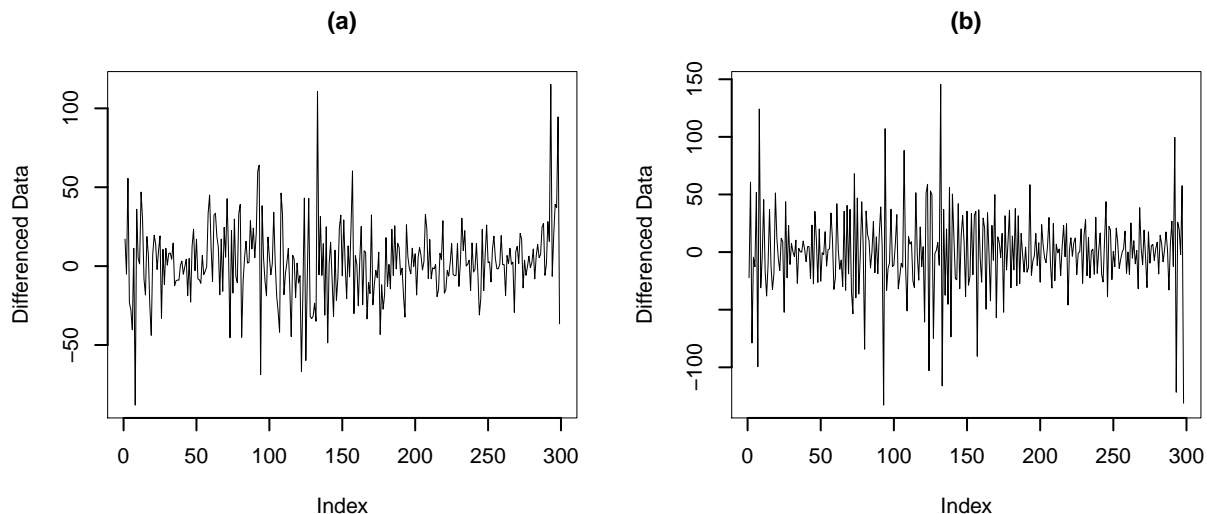


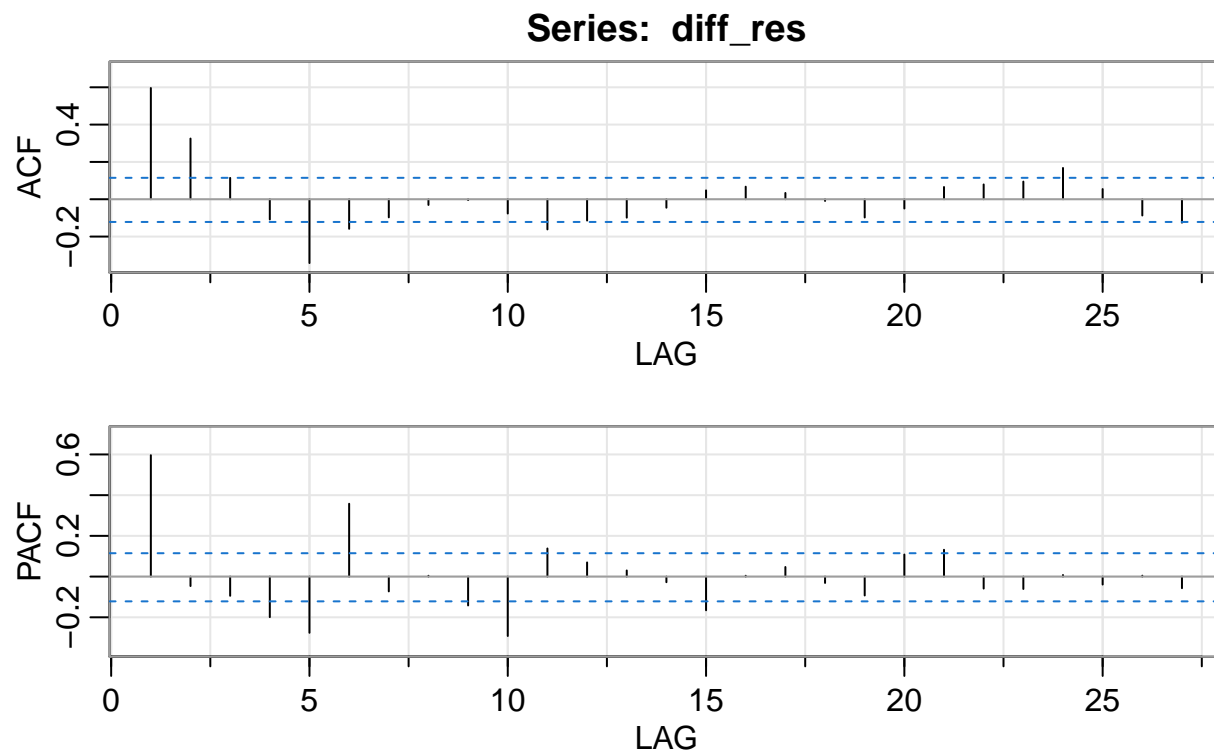
Figure 5: (a): The first-order differenced data. (b): The second-order differenced data.

The second-order differenced time series is more stationary than the first-order differenced time series. We can keep trying more higher-order differencings, but they may overfit our data. Therefore, we think the second-order differenced time series is already stationary enough for us to build ARMA model on it.

6 Model Comparison and Selection

6.1 [Exponential Smoothing + Seasonal Differencing] Model 2 (Ruoja Zhang)

Here we fit seasonal ARMA model of the seasonal differenced residuals after reducing the trend using exponential smoothing.



```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12]
## ACF   0.6  0.33  0.11 -0.11 -0.34 -0.16 -0.10 -0.03  0.00 -0.08 -0.16 -0.11
## PACF  0.6 -0.05 -0.09 -0.20 -0.28  0.36 -0.07  0.00 -0.14 -0.29  0.14  0.07
##      [,13] [,14] [,15] [,16] [,17] [,18] [,19] [,20] [,21] [,22] [,23] [,24]
## ACF  -0.10 -0.05  0.05  0.07  0.03 -0.01 -0.10 -0.05  0.07  0.08  0.09  0.17
## PACF  0.03 -0.03 -0.17  0.00  0.05 -0.03 -0.09  0.11  0.13 -0.06 -0.06  0.01
##      [,25] [,26] [,27]
## ACF   0.05 -0.09 -0.13
## PACF -0.04  0.00 -0.06
```

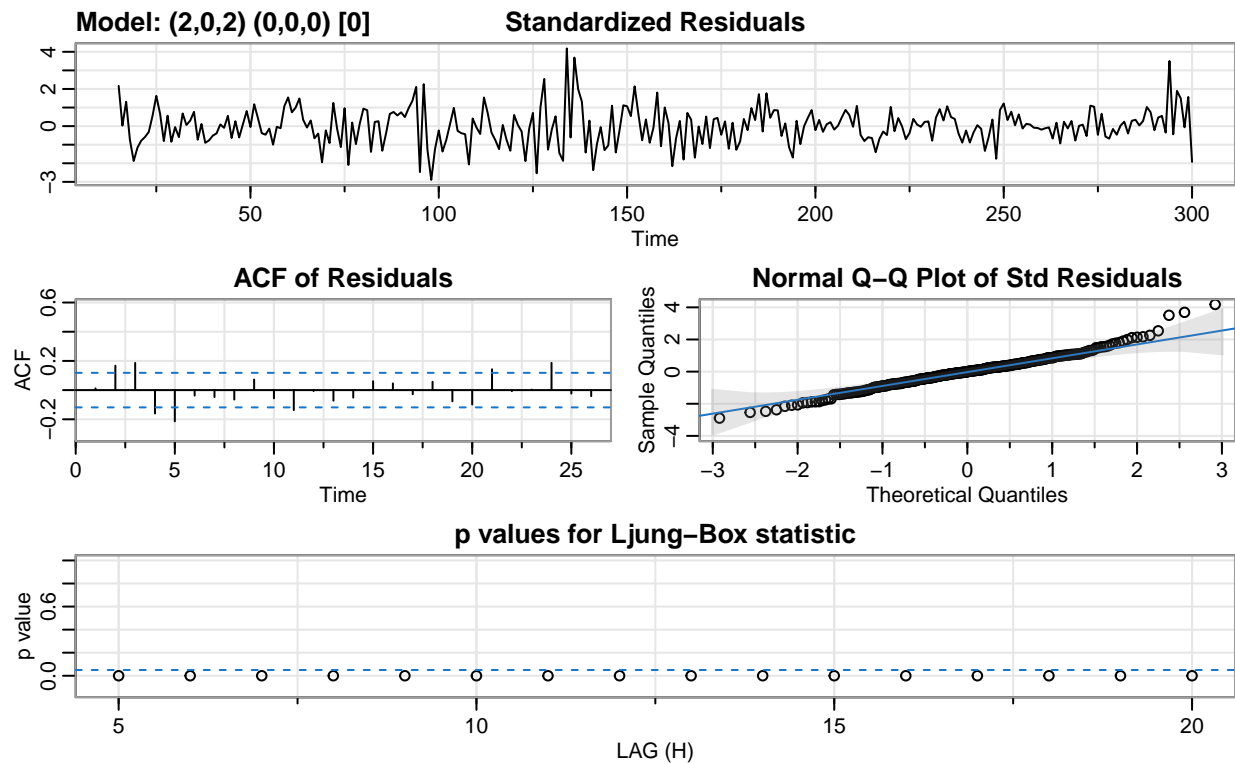
From the acf and pacf plots of the differenced residuals above we can observe negative spike at lag = 5 for both plots. Neither the acf or the pacf plot shows reasonable cutoff. So here we need to choose a model where $p > 0$ and $q > 0$.

First, using `auto.arima()` to see the model suggested by R:

```
## Series: diff_res
## ARIMA(2,0,2) with non-zero mean
##
## Coefficients:
##          ar1      ar2      ma1      ma2      mean
##       -0.7582 -0.0918  1.5843  0.9232  1.7195
## s.e.   0.0653  0.0714  0.0272  0.0304  2.5360
##
## sigma^2 estimated as 522.5:  log likelihood=-1299.87
## AIC=2611.75  AICc=2612.05  BIC=2633.68
##
## initial value 3.407572
## iter 2 value 3.403434
```

```
## iter    3 value 3.346164
## iter    4 value 3.258636
## iter    5 value 3.191177
## iter    6 value 3.183256
## iter    7 value 3.183046
## iter    8 value 3.182992
## iter    9 value 3.182857
## iter   10 value 3.182711
## iter   11 value 3.181575
## iter   12 value 3.181352
## iter   13 value 3.179559
## iter   14 value 3.178037
## iter   15 value 3.172427
## iter   16 value 3.171845
## iter   17 value 3.165885
## iter   18 value 3.164774
## iter   19 value 3.163647
## iter   20 value 3.163294
## iter   21 value 3.162700
## iter   22 value 3.160894
## iter   23 value 3.154028
## iter   24 value 3.153765
## iter   25 value 3.151051
## iter   26 value 3.150299
## iter   27 value 3.149226
## iter   28 value 3.149077
## iter   29 value 3.148956
## iter   30 value 3.148868
## iter   31 value 3.148832
## iter   32 value 3.148827
## iter   33 value 3.148827
## iter   33 value 3.148827
## iter   33 value 3.148827
## final   value 3.148827
## converged
## initial value 3.140662
## iter    2 value 3.139733
## iter    3 value 3.132547
## iter    4 value 3.129600
## iter    5 value 3.128885
## iter    6 value 3.127033
## iter    7 value 3.126798
## iter    8 value 3.126665
## iter    9 value 3.126656
## iter   10 value 3.126635
## iter   11 value 3.126344
## iter   12 value 3.126119
## iter   13 value 3.126091
## iter   14 value 3.126077
## iter   15 value 3.126074
## iter   16 value 3.126073
## iter   17 value 3.126073
## iter   17 value 3.126073
## iter   17 value 3.126073
```

```
## final value 3.126073
## converged
```



```
## [1] 9.131981
```

```
## [1] 9.13273
```

```
## [1] 9.20868
```

From the performance of the model recommended by R presented above, we can see that the model does not pass the Ljung-Box test (all the p-values are below the blue band). Combined with my observations from the acf and pacf plots, I adjusted the model parameters and tried the lower the AIC, AICc, and BIC values of the model at the same time. In the end, keeping other parameters unchanged, I reduced the values for p and q from 2 to 1, and chose $S = 5$ and $Q = 2$, and the model presented p-values lying far beyond the blue band in the Ljung-Box statistic plot with relatively lower AIC, AICc, and BIC values than before:

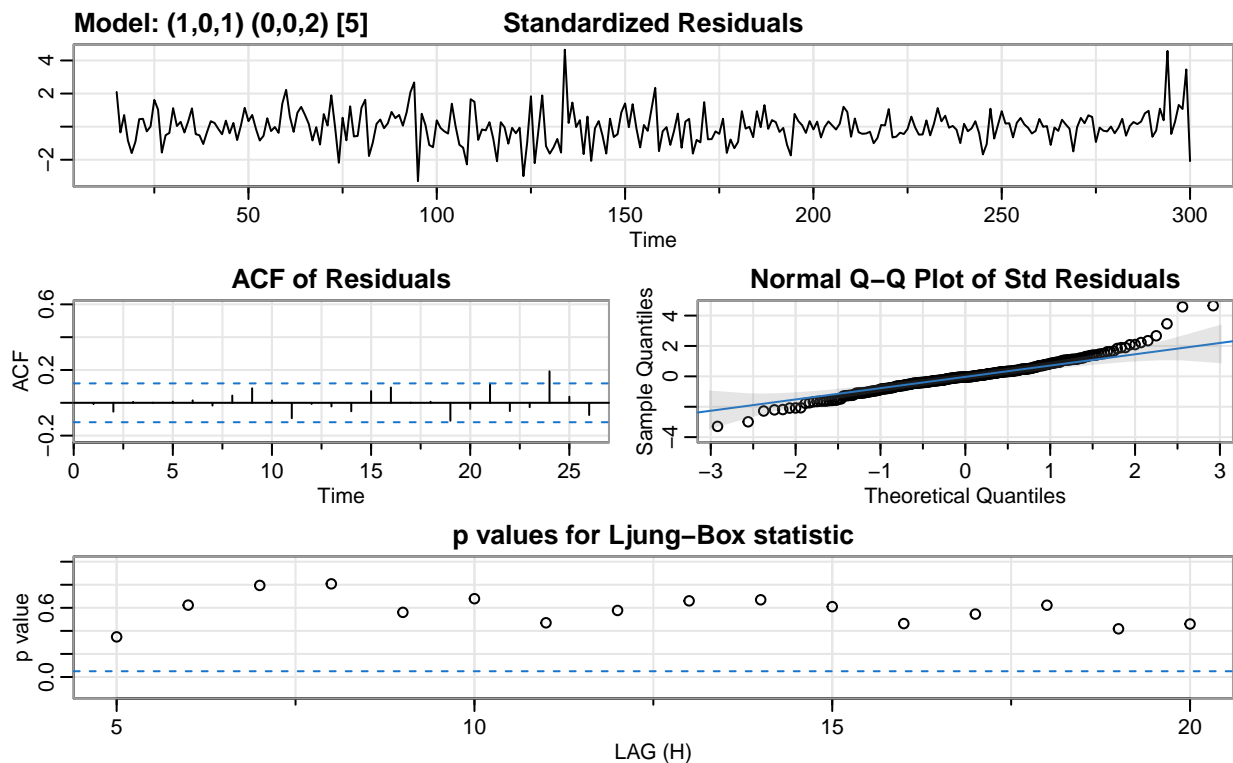
```
## initial value 3.407921
## iter 2 value 3.268990
## iter 3 value 2.974443
## iter 4 value 2.966190
## iter 5 value 2.957344
## iter 6 value 2.954149
## iter 7 value 2.948214
## iter 8 value 2.946849
## iter 9 value 2.946355
## iter 10 value 2.946259
## iter 11 value 2.946247
```



```

## iter 12 value 2.946246
## iter 12 value 2.946246
## final value 2.946246
## converged
## initial value 2.938108
## iter 2 value 2.934170
## iter 3 value 2.929293
## iter 4 value 2.927175
## iter 5 value 2.926083
## iter 6 value 2.925194
## iter 7 value 2.923708
## iter 8 value 2.922816
## iter 9 value 2.922277
## iter 10 value 2.922094
## iter 11 value 2.922050
## iter 12 value 2.922045
## iter 13 value 2.922045
## iter 14 value 2.922045
## iter 15 value 2.922045
## iter 16 value 2.922045
## iter 16 value 2.922045
## iter 16 value 2.922045
## final value 2.922045
## converged

```



From the acf plot of the residuals, we can see that most of the sample residuals autocorrelations fit nicely inside the confidence band. From the normal q-q plot of the stanard residuals we can see that normal distribution is a convincing assumption. We also have large p-values that suggests weak groupwise residual autocorrelations in the plot for Ljung-Box statistic. The model seems like a good fit.

Then we report the AIC, AICc, BIC values for the model:

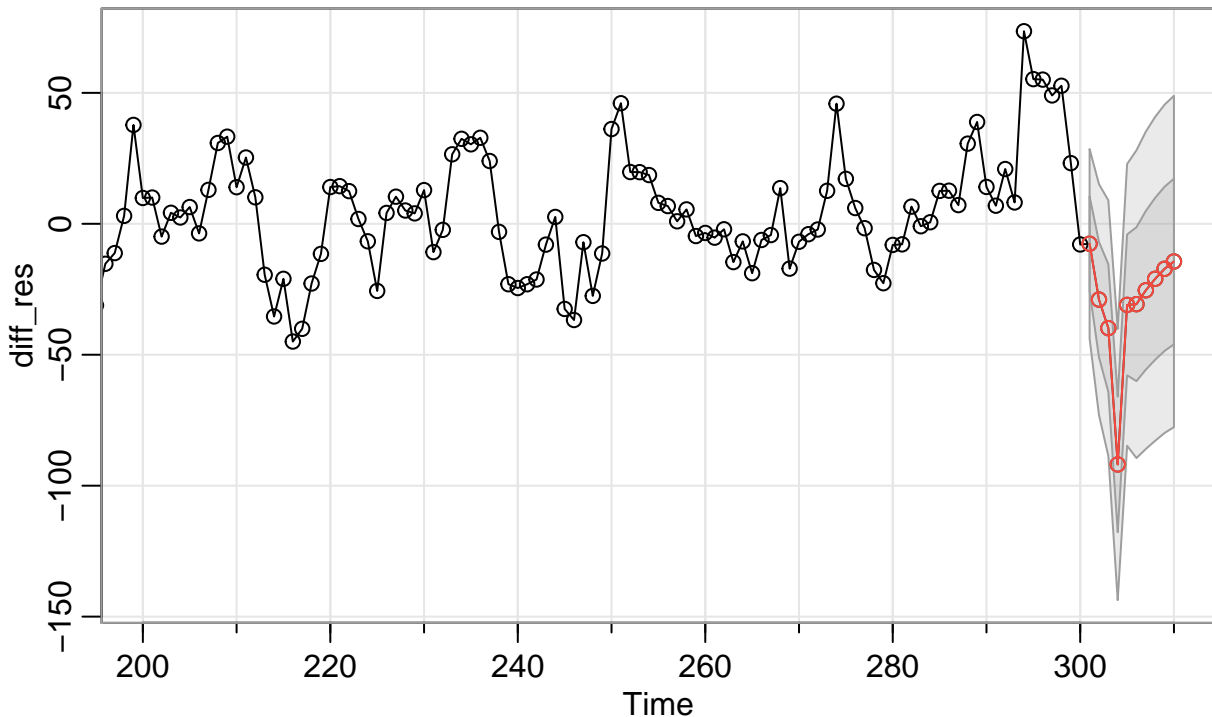
```
## [1] 8.723925
```

```
## [1] 8.724674
```

```
## [1] 8.800624
```

We then perform time-series cross validation. For this model, we generate forecasts for every 10 days (2 weeks) and compute the sum of squares of errors of the forecasts. Then average the sum of squares of errors of forecasts over the days considered, which generates the mse of this model.

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 250	679.7153	649.2001	710.2304	633.0464	726.3842
## 251	679.7153	638.6613	720.7692	616.9287	742.5019
## 252	679.7153	630.3222	729.1084	604.1751	755.2555
## 253	679.7153	623.2004	736.2301	593.2833	766.1473
## 254	679.7153	616.8807	742.5498	583.6181	775.8124
## 255	679.7153	611.1410	748.2896	574.8400	784.5906
## 256	679.7153	605.8459	753.5847	566.7418	792.6887
## 257	679.7153	600.9058	758.5248	559.1866	800.2440
## 258	679.7153	596.2576	763.1730	552.0777	807.3528
## 259	679.7153	591.8549	767.5756	545.3445	814.0861



##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 251	703.6415	673.1112	734.1718	656.9494	750.3336
## 252	703.6415	662.5672	744.7159	640.8237	766.4593
## 253	703.6415	654.2239	753.0592	628.0638	779.2193
## 254	703.6415	647.0986	760.1845	617.1665	790.1165

## 255	703.6415	640.7757	766.5073	607.4966	799.7864
## 256	703.6415	635.0332	772.2499	598.7141	808.5690
## 257	703.6415	629.7354	777.5476	590.6119	816.6711
## 258	703.6415	624.7929	782.4902	583.0529	824.2301
## 259	703.6415	620.1423	787.1407	575.9406	831.3425
## 260	703.6415	615.7375	791.5455	569.2039	838.0791

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 252	708.0051	677.5336	738.4766	661.4030	754.6073
## 253	708.0051	667.0099	749.0004	645.3084	770.7019
## 254	708.0051	658.6827	757.3276	632.5730	783.4373
## 255	708.0051	651.5711	764.4391	621.6968	794.3135
## 256	708.0051	645.2605	770.7498	612.0455	803.9648
## 257	708.0051	639.5290	776.4813	603.2799	812.7304
## 258	708.0051	634.2415	781.7688	595.1933	820.8170
## 259	708.0051	629.3084	786.7019	587.6489	828.3614
## 260	708.0051	624.6668	791.3434	580.5502	835.4601
## 261	708.0051	620.2705	795.7398	573.8266	842.1837

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 253	710.8805	680.4689	741.2921	664.3700	757.3910
## 254	710.8805	669.9659	751.7951	648.3070	773.4540
## 255	710.8805	661.6551	760.1060	635.5967	786.1644
## 256	710.8805	654.5575	767.2036	624.7418	797.0192
## 257	710.8805	648.2592	773.5018	615.1095	806.6516
## 258	710.8805	642.5390	779.2221	606.3611	815.3999
## 259	710.8805	637.2618	784.4992	598.2905	823.4706
## 260	710.8805	632.3385	789.4226	590.7608	831.0002
## 261	710.8805	627.7061	794.0550	583.6762	838.0849
## 262	710.8805	623.3184	798.4427	576.9657	844.7953

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 254	702.132	671.7709	732.4932	655.6987	748.5654
## 255	702.132	661.2853	742.9788	639.6623	764.6017
## 256	702.132	652.9882	751.2758	626.9731	777.2910
## 257	702.132	645.9024	758.3616	616.1363	788.1278
## 258	702.132	639.6146	764.6494	606.5199	797.7442
## 259	702.132	633.9039	770.3602	597.7860	806.4780
## 260	702.132	628.6355	775.6286	589.7288	814.5353
## 261	702.132	623.7203	780.5437	582.2117	822.0524
## 262	702.132	619.0956	785.1685	575.1387	829.1253
## 263	702.132	614.7151	789.5489	568.4394	835.8246

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 255	710.9412	680.6300	741.2523	664.5843	757.2981
## 256	710.9412	670.1617	751.7207	648.5744	773.3080
## 257	710.9412	661.8783	760.0041	635.9060	785.9764
## 258	710.9412	654.8042	767.0782	625.0870	796.7953
## 259	710.9412	648.5267	773.3556	615.4865	806.3959
## 260	710.9412	642.8254	779.0570	606.7670	815.1153
## 261	710.9412	637.5657	784.3167	598.7230	823.1593
## 262	710.9412	632.6586	789.2238	591.2183	830.6641
## 263	710.9412	628.0415	793.8409	584.1570	837.7254
## 264	710.9412	623.6683	798.2141	577.4688	844.4136

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 256	728.9131	698.6192	759.2070	682.5825	775.2437
## 257	728.9131	688.1568	769.6694	666.5817	791.2445
## 258	728.9131	679.8781	777.9481	653.9205	803.9057
## 259	728.9131	672.8079	785.0182	643.1077	814.7185
## 260	728.9131	666.5341	791.2921	633.5126	824.3136
## 261	728.9131	660.8359	796.9902	624.7981	833.0281
## 262	728.9131	655.5793	802.2469	616.7587	841.0675
## 263	728.9131	650.6749	807.1512	609.2582	848.5680
## 264	728.9131	646.0604	811.7658	602.2009	855.6253
## 265	728.9131	641.6897	816.1365	595.5164	862.3097

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 257	735.0393	704.8001	765.2785	688.7924	781.2861
## 258	735.0393	694.3566	775.7220	672.8205	797.2581
## 259	735.0393	686.0929	783.9857	660.1822	809.8964
## 260	735.0393	679.0355	791.0431	649.3889	820.6897
## 261	735.0393	672.7730	797.3056	639.8112	830.2674
## 262	735.0393	667.0851	802.9934	631.1124	838.9662
## 263	735.0393	661.8379	808.2406	623.0875	846.9911
## 264	735.0393	656.9425	813.1361	615.6006	854.4780
## 265	735.0393	652.3363	817.7422	608.5560	861.5225
## 266	735.0393	647.9735	822.1051	601.8837	868.1949

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 258	734.185	704.0050	764.3649	688.0287	780.3412
## 259	734.185	693.5820	774.7879	672.0881	796.2818
## 260	734.185	685.3345	783.0355	659.4746	808.8953
## 261	734.185	678.2909	790.0790	648.7024	819.6675
## 262	734.185	672.0407	796.3292	639.1435	829.2265
## 263	734.185	666.3640	802.0059	630.4617	837.9082
## 264	734.185	661.1271	807.2428	622.4525	845.9174
## 265	734.185	656.2412	812.1287	614.9803	853.3896
## 266	734.185	651.6441	816.7258	607.9496	860.4204
## 267	734.185	647.2898	821.0801	601.2903	867.0796

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 259	732.5695	702.4482	762.6908	686.5030	778.6360
## 260	732.5695	692.0455	773.0935	670.5933	794.5457
## 261	732.5695	683.8140	781.3251	658.0043	807.1347
## 262	732.5695	676.7841	788.3549	647.2531	817.8859
## 263	732.5695	670.5460	794.5930	637.7127	827.4263
## 264	732.5695	664.8804	800.2587	629.0479	836.0911
## 265	732.5695	659.6536	805.4854	621.0543	844.0847
## 266	732.5695	654.7773	810.3618	613.5965	851.5425
## 267	732.5695	650.1890	814.9500	606.5795	858.5595
## 268	732.5695	645.8432	819.2958	599.9331	865.2059

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 260	733.47	703.4072	763.5327	687.4930	779.4469
## 261	733.47	693.0247	773.9152	671.6143	795.3256
## 262	733.47	684.8092	782.1307	659.0497	807.8902
## 263	733.47	677.7930	789.1469	648.3194	818.6205

## 264	733.47	671.5670	795.3729	638.7976	828.1423
## 265	733.47	665.9124	801.0275	630.1496	836.7904
## 266	733.47	660.6958	806.2441	622.1715	844.7684
## 267	733.47	655.8289	811.1110	614.7283	852.2117
## 268	733.47	651.2496	815.6903	607.7248	859.2151
## 269	733.47	646.9123	820.0276	601.0914	865.8485

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 261	750.975	720.9305	781.0195	705.0259	796.9241
## 262	750.975	710.5542	791.3957	689.1568	812.7932
## 263	750.975	702.3437	799.6063	676.5998	825.3501
## 264	750.975	695.3318	806.6182	665.8760	836.0739
## 265	750.975	689.1095	812.8404	656.3600	845.5900
## 266	750.975	683.4583	818.4916	647.7172	854.2328
## 267	750.975	678.2449	823.7050	639.7439	862.2060
## 268	750.975	673.3810	828.5690	632.3052	869.6447
## 269	750.975	668.8045	833.1455	625.3060	876.6439
## 270	750.975	664.4697	837.4802	618.6766	883.2733

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 262	753.5805	723.5932	783.5678	707.7188	799.4422
## 263	753.5805	713.2367	793.9243	691.8799	815.2811
## 264	753.5805	705.0417	802.1192	679.3469	827.8141
## 265	753.5805	698.0432	809.1178	668.6435	838.5175
## 266	753.5805	691.8328	815.3282	659.1455	848.0155
## 267	753.5805	686.1923	820.9687	650.5192	856.6418
## 268	753.5805	680.9888	826.1722	642.5611	864.5999
## 269	753.5805	676.1342	831.0268	635.1365	872.0244
## 270	753.5805	671.5663	835.5946	628.1507	879.0103
## 271	753.5805	667.2399	839.9211	621.5339	885.6271

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 263	754.732	724.8022	784.6618	708.9584	800.5057
## 264	754.732	714.4656	794.9985	693.1499	816.3142
## 265	754.732	706.2864	803.1776	680.6409	828.8232
## 266	754.732	699.3013	810.1628	669.9580	839.5061
## 267	754.732	693.1028	816.3612	660.4783	848.9858
## 268	754.732	687.4732	821.9909	651.8685	857.5956
## 269	754.732	682.2797	827.1844	643.9257	865.5383
## 270	754.732	677.4343	832.0297	636.5154	872.9487
## 271	754.732	672.8753	836.5888	629.5429	879.9211
## 272	754.732	668.5571	840.9070	622.9388	886.5252

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 264	738.1162	708.2079	768.0245	692.3755	783.8570
## 265	738.1162	697.8787	778.3537	676.5783	799.6541
## 266	738.1162	689.7054	786.5270	664.0783	812.1541
## 267	738.1162	682.7253	793.5071	653.4031	822.8293
## 268	738.1162	676.5313	799.7011	643.9302	832.3022
## 269	738.1162	670.9057	805.3267	635.3266	840.9058
## 270	738.1162	665.7159	810.5165	627.3896	848.8429
## 271	738.1162	660.8741	815.3584	619.9846	856.2479
## 272	738.1162	656.3183	819.9141	613.0171	863.2153
## 273	738.1162	652.0032	824.2292	606.4178	869.8147

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 265	742.5116	712.6580	772.3653	696.8544	788.1688
## 266	742.5116	702.3476	782.6756	681.0861	803.9371
## 267	742.5116	694.1893	790.8340	668.6090	816.4143
## 268	742.5116	687.2219	797.8014	657.9533	827.0700
## 269	742.5116	681.0392	803.9841	648.4977	836.5256
## 270	742.5116	675.4239	809.5994	639.9098	845.1135
## 271	742.5116	670.2436	814.7797	631.9872	853.0360
## 272	742.5116	665.4106	819.6127	624.5957	860.4275
## 273	742.5116	660.8631	824.1601	617.6410	867.3822
## 274	742.5116	656.5559	828.4673	611.0537	873.9695

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 266	744.2922	714.4949	774.0894	698.7212	789.8631
## 267	744.2922	704.2041	784.3803	682.9827	805.6016
## 268	744.2922	696.0611	792.5232	670.5291	818.0552
## 269	744.2922	689.1069	799.4774	659.8936	828.6908
## 270	744.2922	682.9359	805.6484	650.4558	838.1285
## 271	744.2922	677.3312	811.2531	641.8842	846.7002
## 272	744.2922	672.1606	816.4237	633.9765	854.6078
## 273	744.2922	667.3368	821.2476	626.5990	861.9853
## 274	744.2922	662.7979	825.7864	619.6574	868.9269
## 275	744.2922	658.4988	830.0855	613.0826	875.5017

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 267	754.6762	724.9215	784.4309	709.1704	800.1821
## 268	754.6762	714.6454	794.7071	693.4544	815.8981
## 269	754.6762	706.5141	802.8384	681.0186	828.3339
## 270	754.6762	699.5698	809.7827	670.3982	838.9543
## 271	754.6762	693.4076	815.9449	660.9739	848.3785
## 272	754.6762	687.8109	821.5416	652.4145	856.9379
## 273	754.6762	682.6478	826.7047	644.5182	864.8343
## 274	754.6762	677.8308	831.5217	637.1512	872.2012
## 275	754.6762	673.2984	836.0541	630.2196	879.1329
## 276	754.6762	669.0055	840.3470	623.6541	885.6983

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 268	756.7586	727.0596	786.4577	711.3378	802.1794
## 269	756.7586	716.8026	796.7146	695.6512	817.8660
## 270	756.7586	708.6865	804.8307	683.2387	830.2786
## 271	756.7586	701.7552	811.7620	672.6382	840.8791
## 272	756.7586	695.6045	817.9127	663.2315	850.2857
## 273	756.7586	690.0183	823.4989	654.6881	858.8291
## 274	756.7586	684.8648	828.6524	646.8066	866.7107
## 275	756.7586	680.0568	833.4604	639.4534	874.0639
## 276	756.7586	675.5329	837.9843	632.5347	880.9826
## 277	756.7586	671.2480	842.2692	625.9815	887.5357

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 269	759.2169	729.5729	788.8608	713.8803	804.5534
## 270	759.2169	719.3350	799.0987	698.2228	820.2109
## 271	759.2169	711.2339	807.1998	685.8333	832.6004
## 272	759.2169	704.3155	814.1182	675.2525	843.1812

## 273	759.2169	698.1762	820.2575	665.8633	852.5704
## 274	759.2169	692.6004	825.8333	657.3357	861.0980
## 275	759.2169	687.4565	830.9772	649.4688	868.9649
## 276	759.2169	682.6574	835.7763	642.1292	876.3045
## 277	759.2169	678.1419	840.2918	635.2234	883.2103
## 278	759.2169	673.8649	844.5688	628.6824	889.7513

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 270	733.0747	703.3987	762.7506	687.6892	778.4601
## 271	733.0747	693.1498	772.9996	672.0148	794.1346
## 272	733.0747	685.0399	781.1094	659.6119	806.5374
## 273	733.0747	678.1140	788.0353	649.0196	817.1297
## 274	733.0747	671.9681	794.1812	639.6203	826.5290
## 275	733.0747	666.3863	799.7631	631.0835	835.0658
## 276	733.0747	661.2368	804.9125	623.2081	842.9412
## 277	733.0747	656.4325	809.7168	615.8606	850.2887
## 278	733.0747	651.9122	814.2372	608.9473	857.2020
## 279	733.0747	647.6306	818.5187	602.3992	863.7501

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 271	738.7495	709.1248	768.3741	693.4425	784.0565
## 272	738.7495	698.8936	778.6054	677.7952	799.7038
## 273	738.7495	690.7978	786.7012	665.4137	812.0852
## 274	738.7495	683.8839	793.6151	654.8398	822.6592
## 275	738.7495	677.7486	799.7504	645.4567	832.0423
## 276	738.7495	672.1764	805.3226	636.9347	840.5643
## 277	738.7495	667.0358	810.4632	629.0729	848.4261
## 278	738.7495	662.2398	815.2591	621.7381	855.7609
## 279	738.7495	657.7273	819.7717	614.8367	862.6622
## 280	738.7495	653.4531	824.0458	608.3000	869.1990

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 272	750.6209	721.0335	780.2084	705.3708	795.8711
## 273	750.6209	710.8151	790.4268	689.7431	811.4988
## 274	750.6209	702.7294	798.5125	677.3772	823.8647
## 275	750.6209	695.8242	805.4177	666.8165	834.4254
## 276	750.6209	689.6966	811.5453	657.4452	843.7967
## 277	750.6209	684.1313	817.1106	648.9339	852.3080
## 278	750.6209	678.9972	822.2447	641.0819	860.1600
## 279	750.6209	674.2073	827.0346	633.7563	867.4856
## 280	750.6209	669.7004	831.5415	626.8636	874.3783
## 281	750.6209	665.4316	835.8103	620.3351	880.9068

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 273	753.3381	723.8045	782.8717	708.1704	798.5058
## 274	753.3381	713.6047	793.0715	692.5712	814.1051
## 275	753.3381	705.5338	801.1424	680.2278	826.4485
## 276	753.3381	698.6411	808.0351	669.6863	836.9899
## 277	753.3381	692.5247	814.1515	660.3321	846.3441
## 278	753.3381	686.9696	819.7066	651.8363	854.8399
## 279	753.3381	681.8449	824.8314	643.9986	862.6776
## 280	753.3381	677.0637	829.6126	636.6864	869.9898
## 281	753.3381	672.5650	834.1113	629.8062	876.8700
## 282	753.3381	668.3039	838.3723	623.2896	883.3866

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 274	772.2848	742.7603	801.8094	727.1309	817.4388
## 275	772.2848	732.5636	812.0061	711.5364	833.0332
## 276	772.2848	724.4951	820.0745	699.1968	845.3728
## 277	772.2848	717.6046	826.9651	688.6586	855.9111
## 278	772.2848	711.4900	833.0796	679.3072	865.2625
## 279	772.2848	705.9366	838.6330	670.8140	873.7557
## 280	772.2848	700.8134	843.7562	662.9787	881.5909
## 281	772.2848	696.0337	848.5360	655.6687	888.9009
## 282	772.2848	691.5364	853.0333	648.7907	895.7790
## 283	772.2848	687.2766	857.2930	642.2760	902.2936

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 275	789.4525	759.9450	818.9600	744.3246	834.5803
## 276	789.4525	749.7542	829.1507	728.7392	850.1657
## 277	789.4525	741.6904	837.2145	716.4067	862.4982
## 278	789.4525	734.8038	844.1011	705.8746	873.0304
## 279	789.4525	728.6928	850.2121	696.5286	882.3763
## 280	789.4525	723.1426	855.7623	688.0403	890.8646
## 281	789.4525	718.0224	860.8826	680.2096	898.6954
## 282	789.4525	713.2454	865.6595	672.9038	906.0011
## 283	789.4525	708.7507	870.1543	666.0297	912.8752
## 284	789.4525	704.4934	874.4115	659.5188	919.3861

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 276	778.7492	749.2814	808.2171	733.6821	823.8164
## 277	778.7492	739.1043	818.3942	718.1176	839.3809
## 278	778.7492	731.0514	826.4471	705.8016	851.6968
## 279	778.7492	724.1740	833.3245	695.2837	862.2148
## 280	778.7492	718.0712	839.4272	685.9502	871.5482
## 281	778.7492	712.5285	844.9700	677.4734	880.0251
## 282	778.7492	707.4151	850.0833	669.6532	887.8453
## 283	778.7492	702.6446	854.8539	662.3572	895.1413
## 284	778.7492	698.1559	859.3426	655.4924	902.0061
## 285	778.7492	693.9044	863.5941	648.9902	908.5083

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 277	781.0539	751.6392	810.4686	736.0680	826.0398
## 278	781.0539	741.4805	820.6273	720.5316	841.5762
## 279	781.0539	733.4421	828.6658	708.2379	853.8699
## 280	781.0539	726.5772	835.5307	697.7389	864.3689
## 281	781.0539	720.4854	841.6225	688.4223	873.6855
## 282	781.0539	714.9526	847.1552	679.9607	882.1471
## 283	781.0539	709.8485	852.2593	672.1546	889.9532
## 284	781.0539	705.0865	857.0213	664.8718	897.2360
## 285	781.0539	700.6060	861.5019	658.0194	904.0885
## 286	781.0539	696.3621	865.7458	651.5289	910.5789

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 278	776.0374	746.6731	805.4017	731.1285	820.9462
## 279	776.0374	736.5317	815.5430	715.6187	836.4561
## 280	776.0374	728.5071	823.5677	703.3461	848.7287
## 281	776.0374	721.6539	830.4209	692.8650	859.2097

## 282	776.0374	715.5725	836.5022	683.5644	868.5104
## 283	776.0374	710.0493	842.0255	675.1173	876.9575
## 284	776.0374	704.9539	847.1209	667.3246	884.7502
## 285	776.0374	700.2001	851.8747	660.0542	892.0205
## 286	776.0374	695.7272	856.3476	653.2135	898.8612
## 287	776.0374	691.4906	860.5842	646.7342	905.3406

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 279	775.3017	745.9906	804.6129	730.4742	820.1292
## 280	775.3017	735.8676	814.7358	714.9924	835.6110
## 281	775.3017	727.8575	822.7459	702.7420	847.8614
## 282	775.3017	721.0167	829.5867	692.2800	858.3234
## 283	775.3017	714.9464	835.6570	682.9962	867.6072
## 284	775.3017	709.4331	841.1703	674.5644	876.0390
## 285	775.3017	704.3469	846.2565	666.7858	883.8177
## 286	775.3017	699.6018	851.0017	659.5286	891.0748
## 287	775.3017	695.1369	855.4665	652.7003	897.9031
## 288	775.3017	690.9080	859.6954	646.2327	904.3707

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 280	780.9072	751.6451	810.1693	736.1547	825.6597
## 281	780.9072	741.5390	820.2753	720.6988	841.1156
## 282	780.9072	733.5423	828.2721	708.4689	853.3455
## 283	780.9072	726.7130	835.1014	698.0243	863.7901
## 284	780.9072	720.6528	841.1616	688.7561	873.0583
## 285	780.9072	715.1488	846.6656	680.3384	881.4760
## 286	780.9072	710.0711	851.7433	672.5728	889.2416
## 287	780.9072	705.3339	856.4805	665.3277	896.4867
## 288	780.9072	700.8765	860.9379	658.5108	903.3036
## 289	780.9072	696.6547	865.1597	652.0541	909.7603

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 281	780.6217	751.4123	809.8312	735.9497	825.2937
## 282	780.6217	741.3244	819.9190	720.5217	840.7218
## 283	780.6217	733.3421	827.9014	708.3138	852.9297
## 284	780.6217	726.5251	834.7184	697.8880	863.3554
## 285	780.6217	720.4758	840.7677	688.6365	872.6070
## 286	780.6217	714.9817	846.2618	680.2339	881.0096
## 287	780.6217	709.9132	851.3303	672.4823	888.7612
## 288	780.6217	705.1844	856.0591	665.2503	895.9932
## 289	780.6217	700.7351	860.5084	658.4457	902.7978
## 290	780.6217	696.5208	864.7226	652.0005	909.2430

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 282	782.5372	753.3797	811.6947	737.9446	827.1297
## 283	782.5372	743.3098	821.7646	722.5440	842.5303
## 284	782.5372	735.3416	829.7327	710.3578	854.7166
## 285	782.5372	728.5367	836.5376	699.9506	865.1238
## 286	782.5372	722.4982	842.5762	690.7155	874.3589
## 287	782.5372	717.0138	848.0605	682.3278	882.7465
## 288	782.5372	711.9543	853.1200	674.5900	890.4844
## 289	782.5372	707.2340	857.8404	667.3709	897.7035
## 290	782.5372	702.7926	862.2818	660.5783	904.4960
## 291	782.5372	698.5858	866.4885	654.1446	910.9297

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 283	792.5027	763.3850	821.6205	747.9709	837.0345
## 284	792.5027	753.3288	831.6766	732.5914	852.4141
## 285	792.5027	745.3715	839.6339	720.4218	864.5837
## 286	792.5027	738.5759	846.4295	710.0287	874.9767
## 287	792.5027	732.5456	852.4598	700.8062	884.1992
## 288	792.5027	727.0687	857.9367	692.4300	892.5754
## 289	792.5027	722.0161	862.9893	684.7027	900.3027
## 290	792.5027	717.3022	867.7032	677.4935	907.5119
## 291	792.5027	712.8669	872.1386	670.7102	914.2952
## 292	792.5027	708.6658	876.3396	664.2853	920.7201

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 284	786.1913	757.1204	815.2621	741.7313	830.6513
## 285	786.1913	747.0804	825.3021	726.3764	846.0061
## 286	786.1913	739.1360	833.2465	714.2264	858.1561
## 287	786.1913	732.3513	840.0312	703.8502	868.5324
## 288	786.1913	726.3307	846.0518	694.6425	877.7400
## 289	786.1913	720.8627	851.5198	686.2798	886.1027
## 290	786.1913	715.8182	856.5643	678.5650	893.8175
## 291	786.1913	711.1119	861.2706	671.3673	901.0152
## 292	786.1913	706.6837	865.6988	664.5950	907.7876
## 293	786.1913	702.4895	869.8931	658.1804	914.2021

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 285	791.3651	762.3426	820.3877	746.9789	835.7513
## 286	791.3651	752.3193	830.4110	731.6496	851.0806
## 287	791.3651	744.3880	838.3423	719.5198	863.2104
## 288	791.3651	737.6146	845.1157	709.1608	873.5695
## 289	791.3651	731.6040	851.1262	699.9684	882.7619
## 290	791.3651	726.1450	856.5852	691.6196	891.1107
## 291	791.3651	721.1089	861.6213	683.9176	898.8127
## 292	791.3651	716.4105	866.3198	676.7319	905.9984
## 293	791.3651	711.9896	870.7406	669.9707	912.7595
## 294	791.3651	707.8023	874.9279	663.5668	919.1634

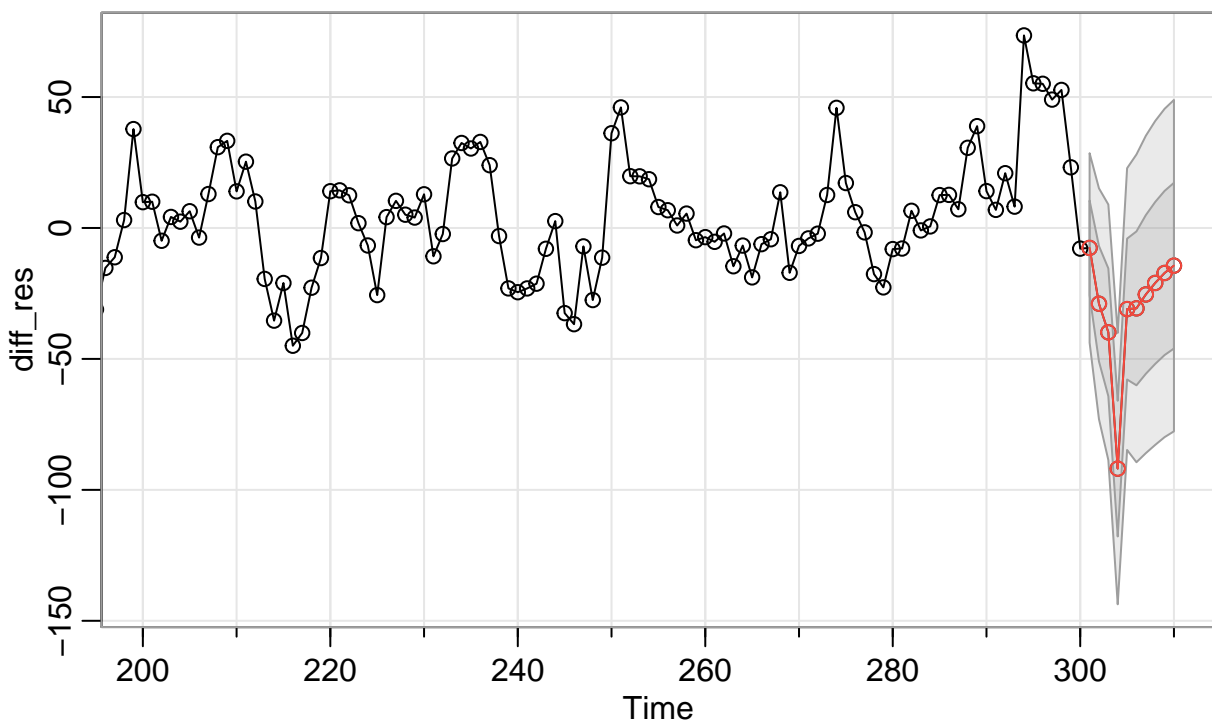
##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 286	804.2845	775.2926	833.2764	759.9452	848.6237
## 287	804.2845	765.2799	843.2891	744.6321	863.9368
## 288	804.2845	757.3570	851.2119	732.5151	876.0538
## 289	804.2845	750.5908	857.9782	722.1671	886.4019
## 290	804.2845	744.5866	863.9824	712.9844	895.5846
## 291	804.2845	739.1334	869.4356	704.6444	903.9245
## 292	804.2845	734.1026	874.4664	696.9505	911.6184
## 293	804.2845	729.4091	879.1599	689.7724	918.7965
## 294	804.2845	724.9929	883.5761	683.0185	925.5505
## 295	804.2845	720.8100	887.7589	676.6213	931.9476

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 287	810.4005	781.4551	839.3459	766.1323	854.6686
## 288	810.4005	771.4584	849.3425	750.8437	869.9572
## 289	810.4005	763.5482	857.2527	738.7462	882.0548
## 290	810.4005	756.7928	864.0081	728.4147	892.3863

## 291	810.4005	750.7982	870.0027	719.2467	901.5542
## 292	810.4005	745.3538	875.4472	710.9201	909.8808
## 293	810.4005	740.3311	880.4699	703.2386	917.5624
## 294	810.4005	735.6451	885.1558	696.0720	924.7290
## 295	810.4005	731.2360	889.5649	689.3288	931.4721
## 296	810.4005	727.0598	893.7411	682.9420	937.8590

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 288	817.5281	788.6272	846.4289	773.3281	861.7281
## 289	817.5281	778.6460	856.4101	758.0630	876.9931
## 290	817.5281	770.7480	864.3081	745.9841	889.0720
## 291	817.5281	764.0030	871.0531	735.6685	899.3876
## 292	817.5281	758.0176	877.0385	726.5147	908.5414
## 293	817.5281	752.5815	882.4746	718.2009	916.8552
## 294	817.5281	747.5666	887.4895	710.5312	924.5249
## 295	817.5281	742.8878	892.1683	703.3756	931.6805
## 296	817.5281	738.4855	896.5706	696.6429	938.4132
## 297	817.5281	734.3157	900.7404	690.2658	944.7903

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 289	840.4798	811.5649	869.3947	796.2583	884.7014
## 290	840.4798	801.5788	879.3809	780.9858	899.9738
## 291	840.4798	793.6770	887.2827	768.9010	912.0586
## 292	840.4798	786.9287	894.0310	758.5804	922.3793
## 293	840.4798	780.9404	900.0193	749.4221	931.5376
## 294	840.4798	775.5017	905.4580	741.1043	939.8554
## 295	840.4798	770.4842	910.4754	733.4308	947.5288
## 296	840.4798	765.8032	915.1565	726.2718	954.6879
## 297	840.4798	761.3987	919.5609	719.5357	961.4239
## 298	840.4798	757.2270	923.7327	713.1556	967.8041



##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 290	867.147	838.1956	896.0984	822.8696	911.4243
## 291	867.147	828.1969	906.0971	807.5779	926.7160
## 292	867.147	820.2851	914.0089	795.4778	938.8161
## 293	867.147	813.5282	920.7657	785.1442	949.1497
## 294	867.147	807.5324	926.7615	775.9744	958.3196
## 295	867.147	802.0868	932.2071	767.6460	966.6479
## 296	867.147	797.0631	937.2309	759.9629	974.3310
## 297	867.147	792.3761	941.9178	752.7948	981.4991
## 298	867.147	787.9661	946.3278	746.0503	988.2436
## 299	867.147	783.7891	950.5049	739.6621	994.6319

cross validation error 69985112

7 Final Model

7.1 Model interpretation

7.2 Prediction

8 Conclusion