

# Forecasting Public Transport Usage

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## Import data

```
td <- read_csv("publicTransport_part.csv")

## Rows: 1323 Columns: 3

## -- Column specification -----
## Delimiter: ","
## chr   (1): DATE
## dbl   (1): DEMAND
## time  (1): TIME

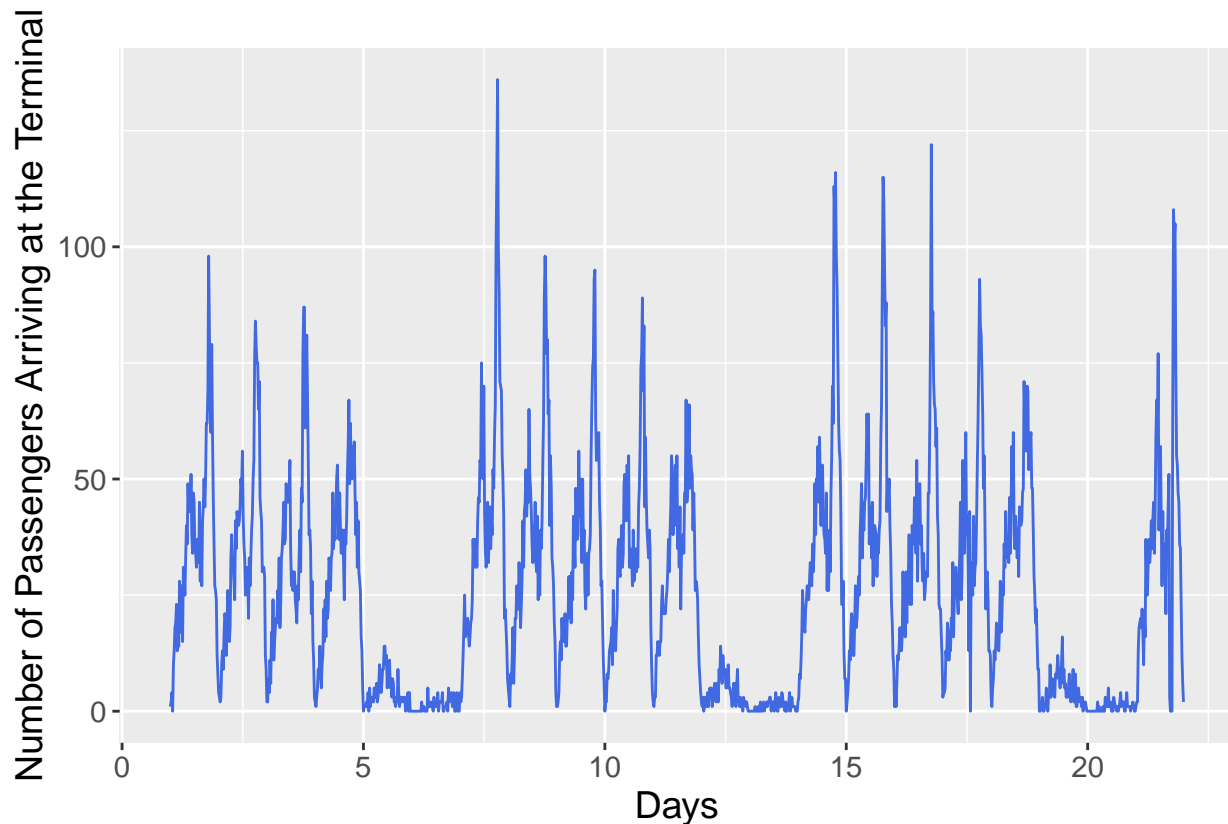
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
#View(td)
```

The time series runs from 1st March 2005 at 06:30 to 21st March at 22:00. There are 63 time periods per day and 1,323 time periods across the total time series to give us a total of 21 days of data. There is no data for the period between 22:00 and 06:30 as the bus terminal is probably closed during this period.

---

## Convert to ts and plot

```
td.ts <- ts(td$DEMAND, frequency = 63) # create a ts based on observations
autoplot(td.ts,
  xlab="Days",
  ylab="Number of Passengers Arriving at the Terminal",
  color="royalblue") +
  theme(text = element_text(size = 14))
```



```
## Overview of the data
```

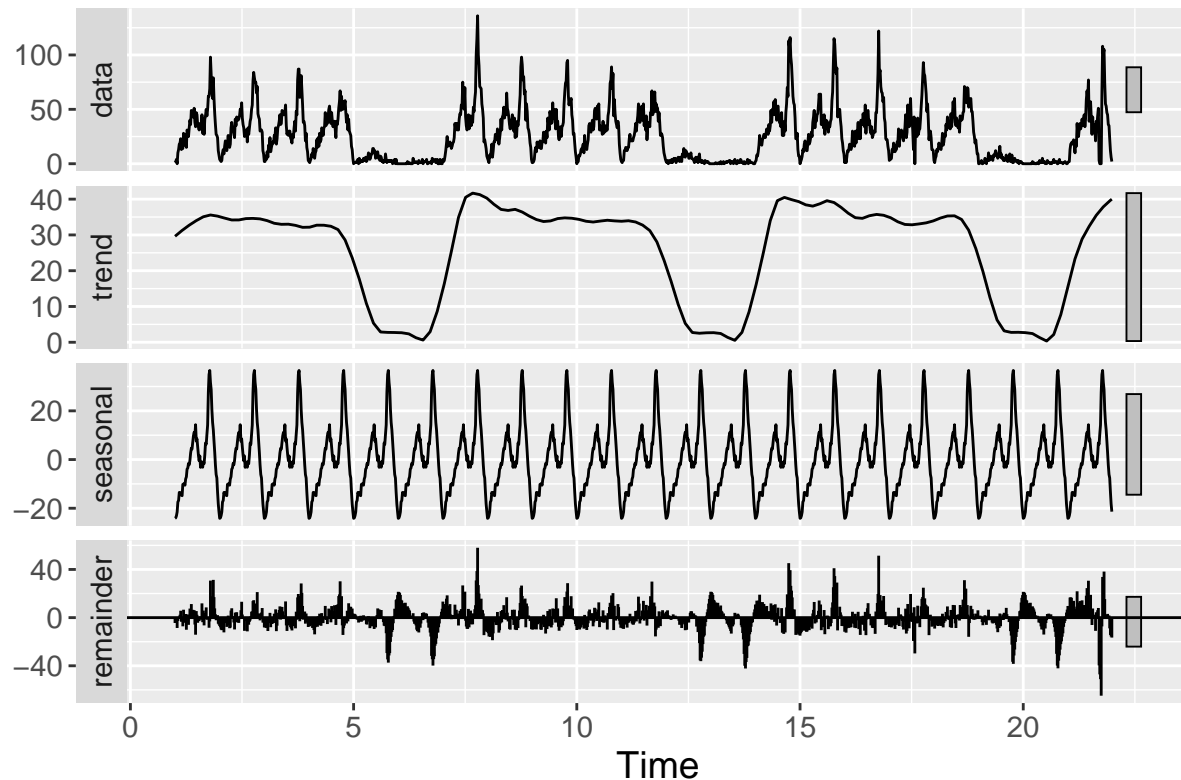
Looking at the data we can see that it has a clear seasonality to it, with the data looking like it starts on a Tuesday, where there is lots of usage for weekdays and usage then drops off for two weekend days. Each day seems to have two peaks, morning usage and evening usage, with the latter being much higher. Taking a Monday to Friday view there seems to be a drop in numbers as the week progresses, with Mondays having highest usage and Fridays least usage, but this isn't completely consistent. There doesn't appear to be a trend to the data, although we only have three weeks of data available. Given we have three weeks of data, it would be best divided into three parts with the first two used for training and the final part used for validation. With a frequency of 63 for the number of data points per day, this doesn't give us many data points to help with forecasting but as there doesn't seem to be a trend and the patterns seems to hold fairly consistently a reasonable forecast is still possible.

## Decompose the data into three components

```
# Components: Seasonal, trend, remainder

td.ts.dec <- stl(td.ts, s.window = "periodic")
autoplot(td.ts.dec,
  main = "Seasonal component, the trend and the remainder in the data") +
  theme(text = element_text(size = 14))
```

## Seasonal component, the trend and the remainder in 1



## Msts for object class

From the data presented it seems like multi-seasonal time series (msts) would be the best class of object for this dataset as there is seasonality on a daily and weekly basis. With a multi-seasonal time series we can account for a frequency of 63 (a daily seasonality) and a frequency of 441 (a weekly seasonality). This msts object allows for the multi-seasonal periods that the normal ts object can't take into consideration.

## ARIMA model

```
td.msts <- msts(td$DEMAND[1:1323], seasonal.periods = c(63, 441)) # daily and weekly seasonality
train.msts <- msts(td$DEMAND[1:882], seasonal.periods = c(63, 441)) # two weeks of training
valid.msts <- msts(td$DEMAND[883:1323], seasonal.periods = c(63, 441)) # one week of validation

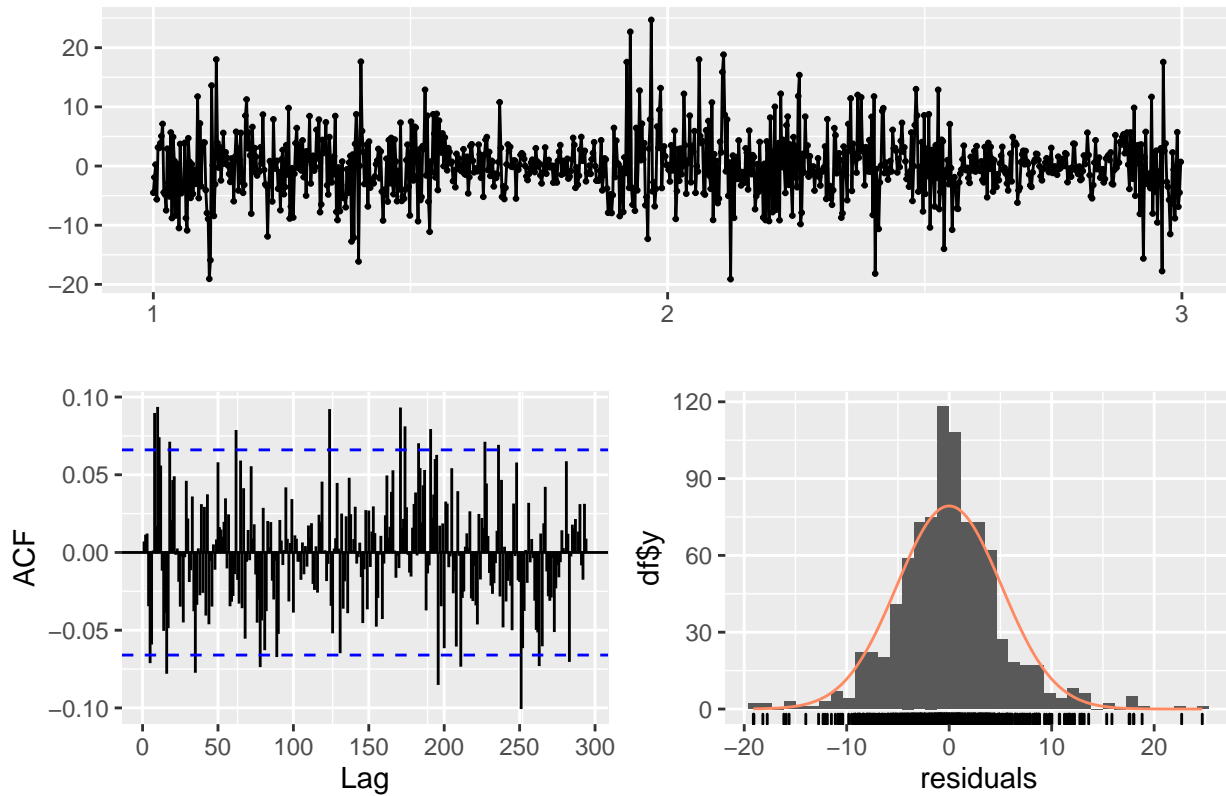
arima.mod <- auto.arima(train.msts,
                        xreg = fourier(train.msts, K = c(20, 40)))

arima.pred <- forecast(arima.mod, xreg = fourier(valid.msts, K = c(20, 40)))
accuracy(arima.pred, td.msts)
```

```
##              ME      RMSE      MAE MPE MAPE      MASE      ACF1
## Training set -0.001670793  5.111612 3.706197 NaN  Inf  0.7257695 0.007079521
## Test set      0.123312144 10.112429 5.356592 NaN  Inf  1.0489596 0.536574105
##              Theil's U
## Training set      NA
## Test set          NaN
```

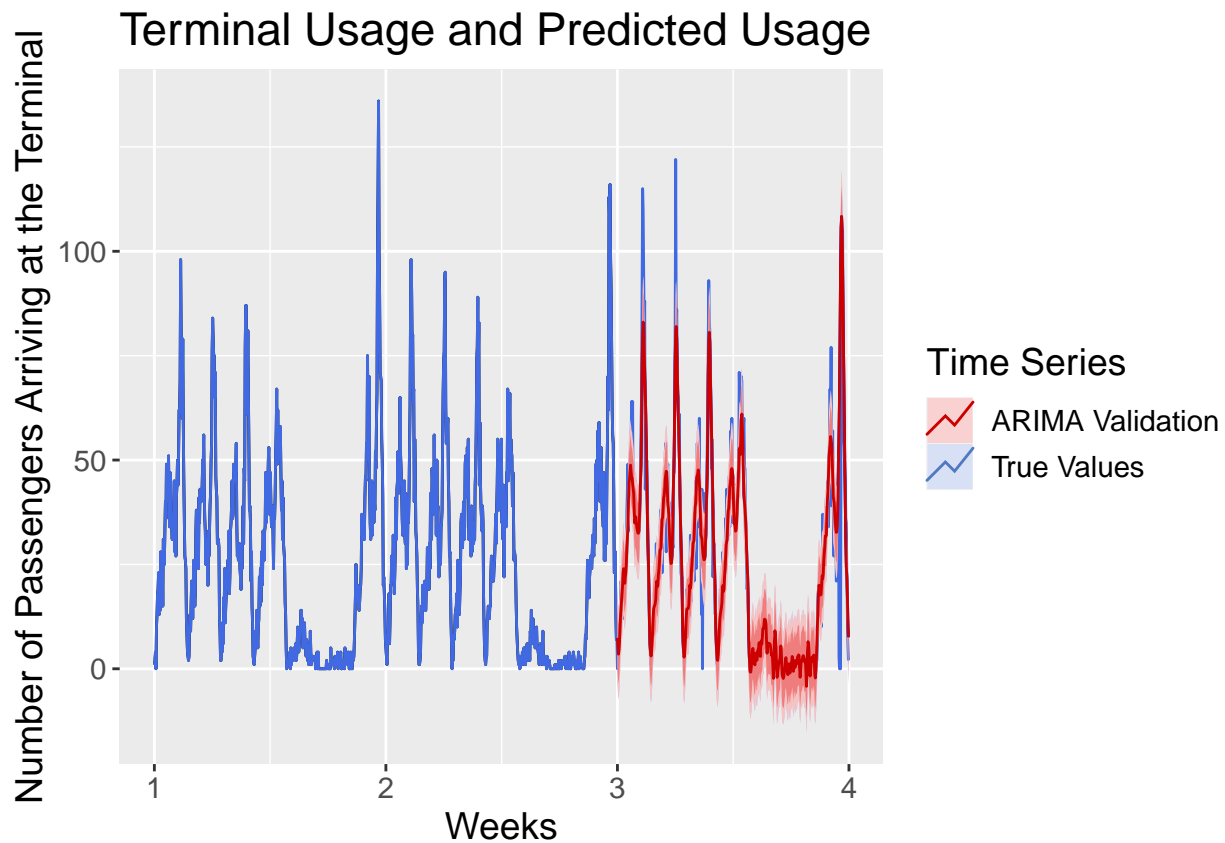
```
checkresiduals(arima.pred)
```

### Residuals from Regression with ARIMA(0,0,3) errors



```
##
##  Ljung-Box test
##
## data:  Residuals from Regression with ARIMA(0,0,3) errors
## Q* = 211.87, df = 62, p-value < 2.2e-16
##
## Model df: 114.    Total lags used: 176

clrs <- c("red", "royalblue")
autoplot(arima.pred, series = "ARIMA") +
  ggtitle("Terminal Usage and Predicted Usage") +
  xlab("Weeks") +
  ylab("Number of Passengers Arriving at the Terminal") +
  autolayer(td.msts, series = "True Values") +
  autolayer(arima.pred, series = "ARIMA Validation") +
  guides(colour=guide_legend(title="Time Series")) +
  scale_color_manual(values=clrs) +
  theme(text = element_text(size = 14))
```



This does reasonably well at forecasting the both daily highs, and also approximates the weekend drop in demand quite well, although Sunday's prediction seems a little off. K of 20, 40 took about 90 seconds to complete. Higher K values can produce better results but place additional demands on computer resources (K should not exceed half of the frequency).

As an alternative I'll try using a linear model with trend and seasonality as predictors.

```
lmts.mod <- tslm(train.msts ~ trend + season, lambda = "auto")
```

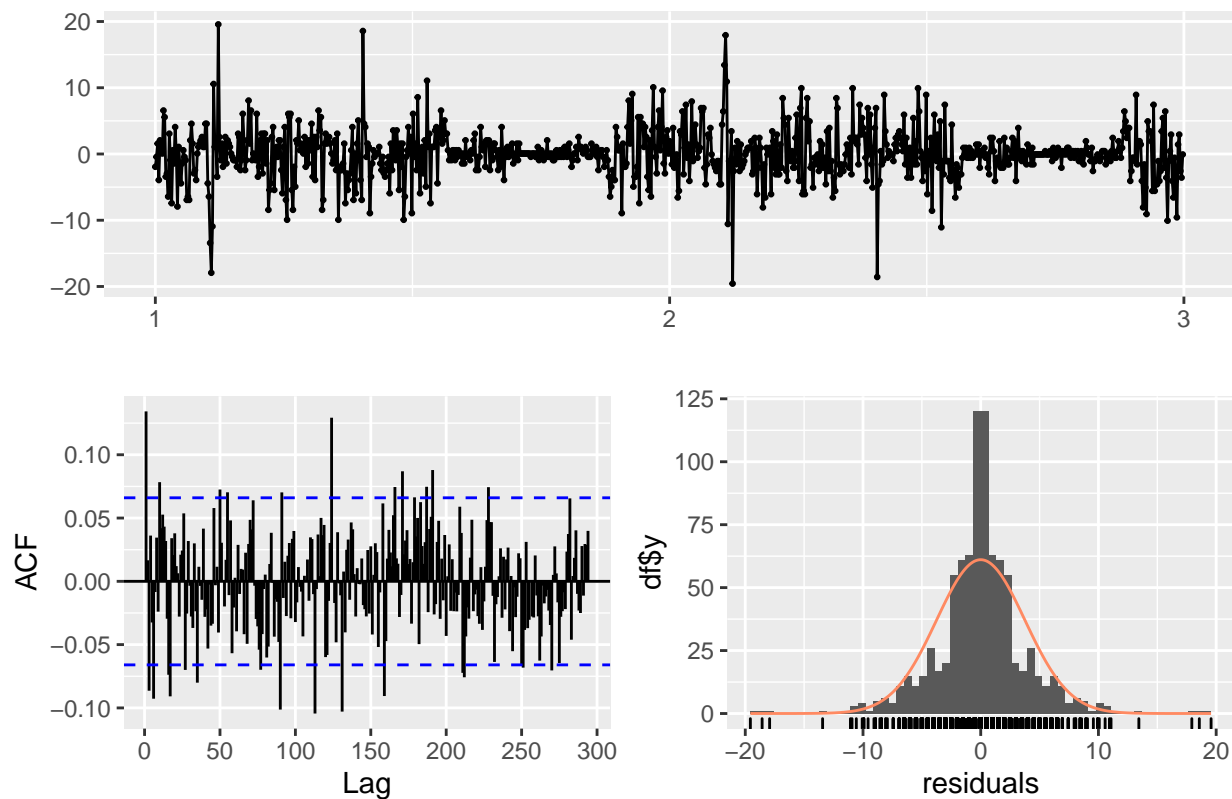
```
lmts.pred <- forecast(lmts.mod, h = 441)
```

```
accuracy(lmts.pred, td.msts)
```

```
##                      ME      RMSE      MAE      MPE      MAPE      MASE      ACF1
## Training set  1.388723e-16  3.754652  2.562389   NaN    Inf  0.5017822  0.1342548
## Test set     -7.029478e-02 10.396808  5.560137  -Inf    Inf  1.0888190  0.4373040
##           Theil's U
## Training set      NA
## Test set          0
```

```
checkresiduals(lmts.pred)
```

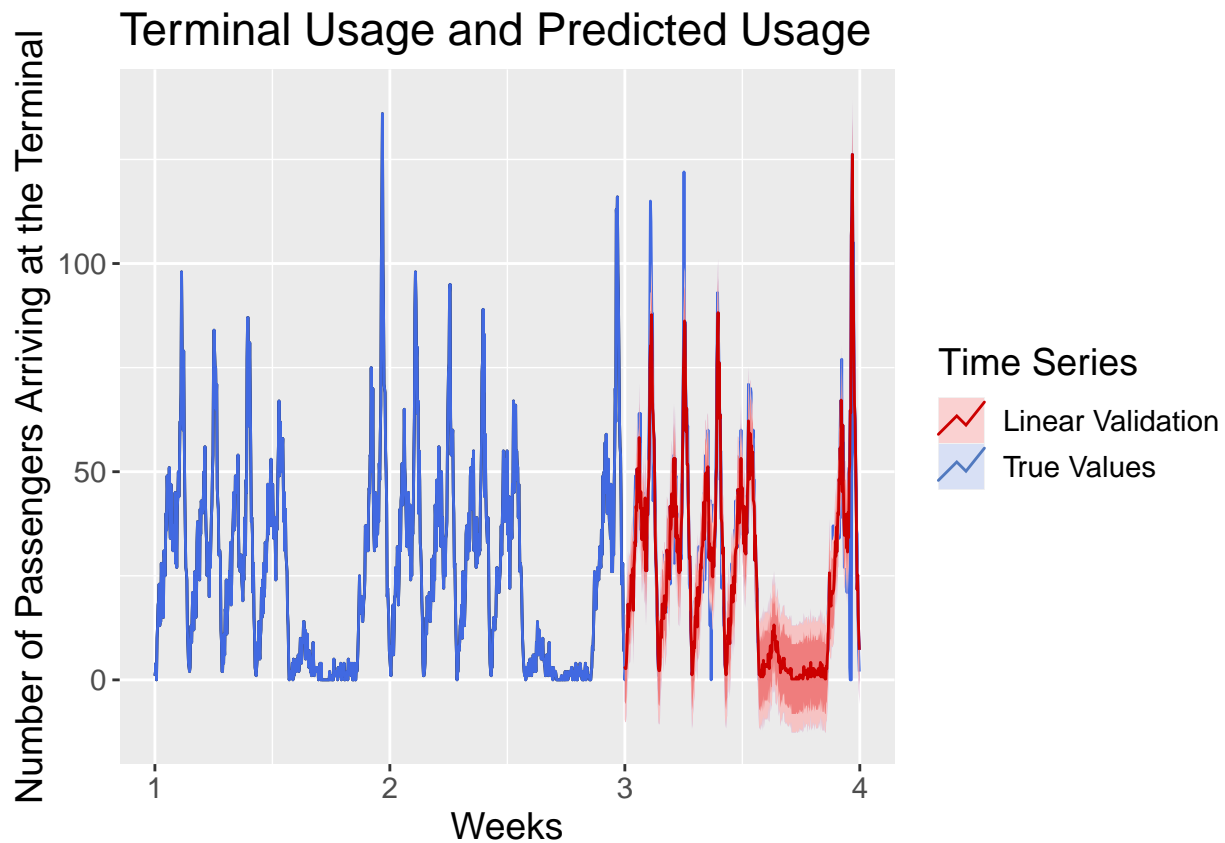
Residuals from Linear regression model



```
##
##  Ljung-Box test
##
## data:  Residuals from Linear regression model
## Q* = 1145.8, df = 3, p-value < 2.2e-16
##
## Model df: 442.    Total lags used: 445

# plot(lmts.pred)
# lines(td.msts)

autoplot(lmts.pred, x=td$DATE, colour=TRUE,
          main="Terminal Usage and Predicted Usage",
          xlab="Weeks",
          ylab="Number of Passengers Arriving at the Terminal",
          color="royalblue") +
  autolayer(td.msts, series = "True Values") +
  autolayer(lmts.pred, series = "Linear Validation") +
  guides(colour=guide_legend(title="Time Series")) +
  scale_color_manual(values=clrs) +
  theme(text = element_text(size = 14))
```



This model performs well and is computationally less intensive as it only took seconds to run although we do have coefficients for each 15-minute period and an intercept, which amounts to 442 in total. Although this has produced a good results I must be concerned with overfitting.

Next let's try a seasonal naive method

```
summary(lmts.mod)
```

```
##
## Call:
## tslm(formula = train.msts ~ trend + season, lambda = "auto")
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -19.568  -1.568   0.000   1.568  19.568
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.932e+00  3.763e+00  0.513  0.607998
## trend        3.085e-04  8.118e-04  0.380  0.704093
## season2     -5.003e-01  5.316e+00 -0.094  0.925060
## season3     -5.006e-01  5.316e+00 -0.094  0.925014
## season4      9.991e-01  5.316e+00  0.188  0.851010
## season5      4.999e+00  5.316e+00  0.940  0.347559
## season6      8.998e+00  5.316e+00  1.693  0.091213 .
## season7     1.500e+01  5.316e+00  2.821  0.004998 **
## season8      9.498e+00  5.316e+00  1.787  0.074677 .
## season9     1.450e+01  5.316e+00  2.727  0.006643 **
## season10    1.350e+01  5.316e+00  2.539  0.011460 *
```

## season11	1.400e+01	5.316e+00	2.633	0.008761	**
## season12	1.750e+01	5.316e+00	3.291	0.001077	**
## season13	2.200e+01	5.316e+00	4.138	4.20e-05	***
## season14	1.800e+01	5.316e+00	3.385	0.000775	***
## season15	2.150e+01	5.316e+00	4.044	6.21e-05	***
## season16	2.350e+01	5.316e+00	4.420	1.25e-05	***
## season17	1.350e+01	5.316e+00	2.539	0.011473	*
## season18	2.399e+01	5.316e+00	4.514	8.19e-06	***
## season19	2.599e+01	5.316e+00	4.890	1.42e-06	***
## season20	2.999e+01	5.316e+00	5.642	3.01e-08	***
## season21	2.899e+01	5.316e+00	5.454	8.22e-08	***
## season22	3.849e+01	5.316e+00	7.241	2.00e-12	***
## season23	3.749e+01	5.316e+00	7.053	6.83e-12	***
## season24	4.749e+01	5.316e+00	8.934	< 2e-16	***
## season25	3.949e+01	5.316e+00	7.429	5.73e-13	***
## season26	4.449e+01	5.316e+00	8.370	7.76e-16	***
## season27	4.099e+01	5.316e+00	7.711	8.40e-14	***
## season28	5.499e+01	5.316e+00	10.345	< 2e-16	***
## season29	4.149e+01	5.316e+00	7.805	4.38e-14	***
## season30	3.799e+01	5.316e+00	7.147	3.72e-12	***
## season31	4.199e+01	5.316e+00	7.899	2.27e-14	***
## season32	3.749e+01	5.316e+00	7.052	6.86e-12	***
## season33	3.049e+01	5.316e+00	5.736	1.81e-08	***
## season34	3.399e+01	5.316e+00	6.394	4.13e-10	***
## season35	3.049e+01	5.316e+00	5.735	1.81e-08	***
## season36	3.799e+01	5.316e+00	7.146	3.73e-12	***
## season37	2.899e+01	5.316e+00	5.453	8.27e-08	***
## season38	2.799e+01	5.316e+00	5.265	2.20e-07	***
## season39	4.049e+01	5.316e+00	7.616	1.61e-13	***
## season40	2.299e+01	5.316e+00	4.324	1.89e-05	***
## season41	3.449e+01	5.316e+00	6.488	2.35e-10	***
## season42	2.299e+01	5.316e+00	4.324	1.90e-05	***
## season43	3.699e+01	5.316e+00	6.958	1.26e-11	***
## season44	3.649e+01	5.316e+00	6.864	2.29e-11	***
## season45	4.249e+01	5.316e+00	7.992	1.18e-14	***
## season46	4.549e+01	5.316e+00	8.556	< 2e-16	***
## season47	5.349e+01	5.316e+00	10.061	< 2e-16	***
## season48	7.249e+01	5.316e+00	13.635	< 2e-16	***
## season49	7.699e+01	5.316e+00	14.482	< 2e-16	***
## season50	7.698e+01	5.316e+00	14.482	< 2e-16	***
## season51	8.448e+01	5.316e+00	15.892	< 2e-16	***
## season52	7.798e+01	5.316e+00	14.670	< 2e-16	***
## season53	6.198e+01	5.316e+00	11.660	< 2e-16	***
## season54	6.048e+01	5.316e+00	11.377	< 2e-16	***
## season55	5.648e+01	5.316e+00	10.625	< 2e-16	***
## season56	5.448e+01	5.316e+00	10.249	< 2e-16	***
## season57	4.398e+01	5.316e+00	8.273	1.56e-15	***
## season58	3.148e+01	5.316e+00	5.922	6.41e-09	***
## season59	2.598e+01	5.316e+00	4.887	1.43e-06	***
## season60	2.348e+01	5.316e+00	4.417	1.26e-05	***
## season61	1.848e+01	5.316e+00	3.476	0.000558	***
## season62	1.048e+01	5.316e+00	1.972	0.049284	*
## season63	2.981e+00	5.316e+00	0.561	0.575274	
## season64	-5.194e-01	5.316e+00	-0.098	0.922208	



```

## season65    -1.020e+00  5.316e+00  -0.192  0.847972
## season66    -1.020e+00  5.316e+00  -0.192  0.847927
## season67     4.796e-01  5.316e+00   0.090  0.928152
## season68     7.979e+00  5.316e+00   1.501  0.134086
## season69     1.098e+01  5.316e+00   2.065  0.039489 *
## season70     6.979e+00  5.316e+00   1.313  0.189960
## season71     1.498e+01  5.316e+00   2.817  0.005058 **
## season72     1.498e+01  5.316e+00   2.817  0.005059 **
## season73     1.448e+01  5.316e+00   2.723  0.006720 **
## season74     1.148e+01  5.316e+00   2.159  0.031394 *
## season75     1.698e+01  5.316e+00   3.193  0.001507 **
## season76     1.698e+01  5.316e+00   3.193  0.001507 **
## season77     1.298e+01  5.316e+00   2.441  0.015043 *
## season78     1.448e+01  5.316e+00   2.723  0.006726 **
## season79     1.598e+01  5.316e+00   3.005  0.002807 **
## season80     2.798e+01  5.316e+00   5.262  2.23e-07 ***
## season81     2.698e+01  5.316e+00   5.074  5.75e-07 ***
## season82     2.748e+01  5.316e+00   5.168  3.60e-07 ***
## season83     2.247e+01  5.316e+00   4.227  2.88e-05 ***
## season84     2.597e+01  5.316e+00   4.886  1.45e-06 ***
## season85     2.097e+01  5.316e+00   3.945  9.28e-05 ***
## season86     3.847e+01  5.316e+00   7.237  2.06e-12 ***
## season87     3.297e+01  5.316e+00   6.202  1.28e-09 ***
## season88     3.397e+01  5.316e+00   6.390  4.22e-10 ***
## season89     4.147e+01  5.316e+00   7.801  4.51e-14 ***
## season90     3.897e+01  5.316e+00   7.331  1.11e-12 ***
## season91     3.597e+01  5.316e+00   6.766  4.23e-11 ***
## season92     3.797e+01  5.316e+00   7.142  3.82e-12 ***
## season93     4.997e+01  5.316e+00   9.399  < 2e-16 ***
## season94     4.197e+01  5.316e+00   7.895  2.34e-14 ***
## season95     4.997e+01  5.316e+00   9.399  < 2e-16 ***
## season96     4.047e+01  5.316e+00   7.612  1.66e-13 ***
## season97     3.097e+01  5.316e+00   5.825  1.10e-08 ***
## season98     3.847e+01  5.316e+00   7.236  2.07e-12 ***
## season99     2.747e+01  5.317e+00   5.167  3.62e-07 ***
## season100    3.147e+01  5.317e+00   5.919  6.52e-09 ***
## season101    2.547e+01  5.317e+00   4.791  2.28e-06 ***
## season102    2.297e+01  5.317e+00   4.320  1.93e-05 ***
## season103    2.247e+01  5.317e+00   4.226  2.89e-05 ***
## season104    3.047e+01  5.317e+00   5.731  1.86e-08 ***
## season105    2.297e+01  5.317e+00   4.320  1.93e-05 ***
## season106    3.047e+01  5.317e+00   5.731  1.86e-08 ***
## season107    3.397e+01  5.317e+00   6.389  4.26e-10 ***
## season108    3.697e+01  5.317e+00   6.953  1.30e-11 ***
## season109    4.947e+01  5.317e+00   9.304  < 2e-16 ***
## season110    5.697e+01  5.317e+00  10.715  < 2e-16 ***
## season111    7.197e+01  5.317e+00  13.536  < 2e-16 ***
## season112    7.697e+01  5.317e+00  14.476  < 2e-16 ***
## season113    8.297e+01  5.317e+00  15.605  < 2e-16 ***
## season114    8.197e+01  5.317e+00  15.417  < 2e-16 ***
## season115    6.596e+01  5.317e+00  12.407  < 2e-16 ***
## season116    5.646e+01  5.317e+00  10.620  < 2e-16 ***
## season117    6.196e+01  5.317e+00  11.655  < 2e-16 ***
## season118    4.846e+01  5.317e+00   9.115  < 2e-16 ***

```

## season119	4.846e+01	5.317e+00	9.115	< 2e-16	***
## season120	3.996e+01	5.317e+00	7.516	3.18e-13	***
## season121	3.196e+01	5.317e+00	6.012	3.86e-09	***
## season122	2.596e+01	5.317e+00	4.883	1.46e-06	***
## season123	2.646e+01	5.317e+00	4.977	9.28e-07	***
## season124	2.096e+01	5.317e+00	3.943	9.38e-05	***
## season125	8.462e+00	5.317e+00	1.591	0.112217	
## season126	4.461e+00	5.317e+00	0.839	0.401864	
## season127	-2.039e+00	5.317e+00	-0.383	0.701556	
## season128	-1.039e+00	5.317e+00	-0.195	0.845132	
## season129	1.461e+00	5.317e+00	0.275	0.783683	
## season130	2.960e+00	5.317e+00	0.557	0.577982	
## season131	5.960e+00	5.317e+00	1.121	0.262933	
## season132	4.460e+00	5.317e+00	0.839	0.402068	
## season133	1.196e+01	5.317e+00	2.249	0.024990	*
## season134	1.146e+01	5.317e+00	2.155	0.031692	*
## season135	1.646e+01	5.317e+00	3.095	0.002091	**
## season136	1.146e+01	5.317e+00	2.155	0.031702	*
## season137	9.458e+00	5.317e+00	1.779	0.075961	.
## season138	1.896e+01	5.317e+00	3.565	0.000403	***
## season139	1.446e+01	5.317e+00	2.719	0.006806	**
## season140	1.496e+01	5.317e+00	2.813	0.005128	**
## season141	1.646e+01	5.317e+00	3.095	0.002093	**
## season142	1.846e+01	5.317e+00	3.471	0.000569	***
## season143	2.446e+01	5.317e+00	4.599	5.55e-06	***
## season144	2.346e+01	5.317e+00	4.411	1.29e-05	***
## season145	2.696e+01	5.317e+00	5.070	5.89e-07	***
## season146	2.846e+01	5.317e+00	5.352	1.41e-07	***
## season147	2.895e+01	5.317e+00	5.445	8.61e-08	***
## season148	3.445e+01	5.317e+00	6.480	2.47e-10	***
## season149	3.945e+01	5.317e+00	7.420	6.09e-13	***
## season150	3.045e+01	5.317e+00	5.727	1.89e-08	***
## season151	3.245e+01	5.317e+00	6.103	2.28e-09	***
## season152	4.595e+01	5.317e+00	8.642	< 2e-16	***
## season153	4.645e+01	5.317e+00	8.736	< 2e-16	***
## season154	4.295e+01	5.317e+00	8.078	6.40e-15	***
## season155	3.995e+01	5.317e+00	7.514	3.24e-13	***
## season156	4.745e+01	5.317e+00	8.924	< 2e-16	***
## season157	4.795e+01	5.317e+00	9.018	< 2e-16	***
## season158	4.195e+01	5.317e+00	7.889	2.43e-14	***
## season159	2.595e+01	5.317e+00	4.880	1.48e-06	***
## season160	2.945e+01	5.317e+00	5.539	5.26e-08	***
## season161	2.545e+01	5.317e+00	4.786	2.32e-06	***
## season162	2.845e+01	5.318e+00	5.350	1.42e-07	***
## season163	2.545e+01	5.318e+00	4.786	2.33e-06	***
## season164	2.845e+01	5.318e+00	5.350	1.42e-07	***
## season165	2.195e+01	5.318e+00	4.128	4.38e-05	***
## season166	2.145e+01	5.318e+00	4.034	6.48e-05	***
## season167	2.795e+01	5.318e+00	5.256	2.30e-07	***
## season168	2.395e+01	5.318e+00	4.504	8.57e-06	***
## season169	2.945e+01	5.318e+00	5.538	5.28e-08	***
## season170	3.195e+01	5.318e+00	6.008	3.94e-09	***
## season171	3.395e+01	5.318e+00	6.384	4.39e-10	***
## season172	4.495e+01	5.318e+00	8.452	4.23e-16	***

## season173	4.795e+01	5.318e+00	9.016	< 2e-16	***
## season174	7.245e+01	5.318e+00	13.624	< 2e-16	***
## season175	7.895e+01	5.318e+00	14.846	< 2e-16	***
## season176	8.495e+01	5.318e+00	15.974	< 2e-16	***
## season177	6.195e+01	5.318e+00	11.649	< 2e-16	***
## season178	7.295e+01	5.318e+00	13.717	< 2e-16	***
## season179	5.945e+01	5.318e+00	11.178	< 2e-16	***
## season180	6.044e+01	5.318e+00	11.366	< 2e-16	***
## season181	4.694e+01	5.318e+00	8.828	< 2e-16	***
## season182	3.544e+01	5.318e+00	6.665	7.94e-11	***
## season183	3.444e+01	5.318e+00	6.477	2.51e-10	***
## season184	3.094e+01	5.318e+00	5.819	1.14e-08	***
## season185	2.694e+01	5.318e+00	5.066	5.98e-07	***
## season186	2.844e+01	5.318e+00	5.348	1.43e-07	***
## season187	1.344e+01	5.318e+00	2.528	0.011829	*
## season188	7.442e+00	5.318e+00	1.399	0.162387	
## season189	4.420e-01	5.318e+00	0.083	0.933800	
## season190	-1.058e+00	5.318e+00	-0.199	0.842354	
## season191	-2.059e+00	5.318e+00	-0.387	0.698875	
## season192	-5.893e-02	5.318e+00	-0.011	0.991165	
## season193	9.408e-01	5.318e+00	0.177	0.859672	
## season194	7.440e+00	5.318e+00	1.399	0.162502	
## season195	6.440e+00	5.318e+00	1.211	0.226563	
## season196	1.144e+01	5.318e+00	2.151	0.032016	*
## season197	7.940e+00	5.318e+00	1.493	0.136187	
## season198	5.439e+00	5.318e+00	1.023	0.306996	
## season199	9.439e+00	5.318e+00	1.775	0.076625	.
## season200	1.044e+01	5.318e+00	1.963	0.050305	.
## season201	1.944e+01	5.318e+00	3.655	0.000288	***
## season202	1.794e+01	5.318e+00	3.373	0.000810	***
## season203	1.944e+01	5.318e+00	3.655	0.000288	***
## season204	1.544e+01	5.318e+00	2.903	0.003887	**
## season205	2.144e+01	5.318e+00	4.031	6.55e-05	***
## season206	1.744e+01	5.319e+00	3.279	0.001126	**
## season207	1.994e+01	5.319e+00	3.748	0.000202	***
## season208	2.644e+01	5.319e+00	4.971	9.58e-07	***
## season209	2.794e+01	5.319e+00	5.252	2.34e-07	***
## season210	2.844e+01	5.319e+00	5.346	1.44e-07	***
## season211	2.844e+01	5.319e+00	5.346	1.44e-07	***
## season212	3.293e+01	5.319e+00	6.192	1.36e-09	***
## season213	4.243e+01	5.319e+00	7.978	1.30e-14	***
## season214	4.193e+01	5.319e+00	7.884	2.52e-14	***
## season215	4.243e+01	5.319e+00	7.978	1.30e-14	***
## season216	3.443e+01	5.319e+00	6.474	2.55e-10	***
## season217	3.943e+01	5.319e+00	7.414	6.35e-13	***
## season218	4.293e+01	5.319e+00	8.072	6.67e-15	***
## season219	4.993e+01	5.319e+00	9.388	< 2e-16	***
## season220	3.943e+01	5.319e+00	7.414	6.36e-13	***
## season221	4.293e+01	5.319e+00	8.072	6.69e-15	***
## season222	3.793e+01	5.319e+00	7.131	4.11e-12	***
## season223	3.643e+01	5.319e+00	6.849	2.51e-11	***
## season224	2.943e+01	5.319e+00	5.533	5.41e-08	***
## season225	3.843e+01	5.319e+00	7.225	2.22e-12	***
## season226	2.743e+01	5.319e+00	5.157	3.80e-07	***

```

## season227 2.993e+01 5.319e+00 5.627 3.27e-08 ***
## season228 2.693e+01 5.319e+00 5.063 6.08e-07 ***
## season229 3.543e+01 5.319e+00 6.661 8.15e-11 ***
## season230 3.193e+01 5.319e+00 6.003 4.06e-09 ***
## season231 4.093e+01 5.319e+00 7.695 9.41e-14 ***
## season232 4.593e+01 5.319e+00 8.634 < 2e-16 ***
## season233 5.893e+01 5.319e+00 11.078 < 2e-16 ***
## season234 5.293e+01 5.319e+00 9.950 < 2e-16 ***
## season235 4.493e+01 5.319e+00 8.446 4.42e-16 ***
## season236 5.593e+01 5.319e+00 10.514 < 2e-16 ***
## season237 5.543e+01 5.319e+00 10.420 < 2e-16 ***
## season238 4.593e+01 5.319e+00 8.634 < 2e-16 ***
## season239 5.293e+01 5.319e+00 9.950 < 2e-16 ***
## season240 4.993e+01 5.319e+00 9.386 < 2e-16 ***
## season241 5.143e+01 5.319e+00 9.667 < 2e-16 ***
## season242 3.993e+01 5.320e+00 7.506 3.42e-13 ***
## season243 3.943e+01 5.320e+00 7.411 6.46e-13 ***
## season244 3.843e+01 5.320e+00 7.223 2.25e-12 ***
## season245 2.542e+01 5.320e+00 4.779 2.40e-06 ***
## season246 3.142e+01 5.320e+00 5.907 6.97e-09 ***
## season247 2.542e+01 5.320e+00 4.779 2.40e-06 ***
## season248 2.292e+01 5.320e+00 4.309 2.02e-05 ***
## season249 1.792e+01 5.320e+00 3.369 0.000820 ***
## season250 9.923e+00 5.320e+00 1.865 0.062798 .
## season251 6.923e+00 5.320e+00 1.301 0.193821
## season252 1.423e+00 5.320e+00 0.267 0.789280
## season253 -2.078e+00 5.320e+00 -0.391 0.696308
## season254 -2.078e+00 5.320e+00 -0.391 0.696267
## season255 -1.078e+00 5.320e+00 -0.203 0.839461
## season256 -2.579e+00 5.320e+00 -0.485 0.628118
## season257 -1.579e+00 5.320e+00 -0.297 0.766758
## season258 -1.079e+00 5.320e+00 -0.203 0.839328
## season259 4.204e-01 5.320e+00 0.079 0.937050
## season260 -2.580e+00 5.320e+00 -0.485 0.627961
## season261 1.420e+00 5.320e+00 0.267 0.789693
## season262 -2.081e+00 5.320e+00 -0.391 0.695938
## season263 -8.083e-02 5.320e+00 -0.015 0.987885
## season264 4.189e-01 5.320e+00 0.079 0.937283
## season265 -8.145e-02 5.320e+00 -0.015 0.987793
## season266 4.182e-01 5.320e+00 0.079 0.937376
## season267 -8.206e-02 5.320e+00 -0.015 0.987700
## season268 4.176e-01 5.320e+00 0.078 0.937469
## season269 -8.268e-02 5.320e+00 -0.016 0.987608
## season270 2.917e+00 5.320e+00 0.548 0.583785
## season271 -8.330e-02 5.320e+00 -0.016 0.987516
## season272 -5.836e-01 5.320e+00 -0.110 0.912704
## season273 2.916e+00 5.320e+00 0.548 0.583911
## season274 4.916e+00 5.321e+00 0.924 0.356031
## season275 1.915e+00 5.321e+00 0.360 0.719011
## season276 3.415e+00 5.321e+00 0.642 0.521288
## season277 3.415e+00 5.321e+00 0.642 0.521329
## season278 8.415e+00 5.321e+00 1.581 0.114486
## season279 6.414e+00 5.321e+00 1.206 0.228649
## season280 6.914e+00 5.321e+00 1.299 0.194477

```

## season281	9.914e+00	5.321e+00	1.863	0.063101	.
## season282	6.413e+00	5.321e+00	1.205	0.228725	
## season283	7.413e+00	5.321e+00	1.393	0.164262	
## season284	3.913e+00	5.321e+00	0.735	0.462520	
## season285	3.912e+00	5.321e+00	0.735	0.462558	
## season286	3.412e+00	5.321e+00	0.641	0.521692	
## season287	5.912e+00	5.321e+00	1.111	0.267162	
## season288	4.411e+00	5.321e+00	0.829	0.407518	
## season289	2.911e+00	5.321e+00	0.547	0.584587	
## season290	9.108e-01	5.321e+00	0.171	0.864164	
## season291	2.411e+00	5.321e+00	0.453	0.650763	
## season292	1.410e+00	5.321e+00	0.265	0.791117	
## season293	-1.590e+00	5.321e+00	-0.299	0.765217	
## season294	-5.904e-01	5.321e+00	-0.111	0.911706	
## season295	4.093e-01	5.321e+00	0.077	0.938724	
## season296	-5.910e-01	5.321e+00	-0.111	0.911615	
## season297	9.087e-01	5.321e+00	0.171	0.864490	
## season298	1.908e+00	5.321e+00	0.359	0.720049	
## season299	1.408e+00	5.321e+00	0.265	0.791439	
## season300	1.908e+00	5.321e+00	0.359	0.720139	
## season301	-5.926e-01	5.321e+00	-0.111	0.911389	
## season302	-5.929e-01	5.322e+00	-0.111	0.911343	
## season303	-1.093e+00	5.322e+00	-0.205	0.837337	
## season304	-5.935e-01	5.322e+00	-0.112	0.911253	
## season305	-1.094e+00	5.322e+00	-0.206	0.837248	
## season306	-1.094e+00	5.322e+00	-0.206	0.837204	
## season307	-1.594e+00	5.322e+00	-0.300	0.764621	
## season308	-1.595e+00	5.322e+00	-0.300	0.764578	
## season309	-1.595e+00	5.322e+00	-0.300	0.764536	
## season310	-1.095e+00	5.322e+00	-0.206	0.837028	
## season311	-5.956e-01	5.322e+00	-0.112	0.910935	
## season312	-2.096e+00	5.322e+00	-0.394	0.693894	
## season313	-5.963e-01	5.322e+00	-0.112	0.910845	
## season314	-3.097e+00	5.322e+00	-0.582	0.560969	
## season315	-3.097e+00	5.322e+00	-0.582	0.560933	
## season316	-3.097e+00	5.322e+00	-0.582	0.560897	
## season317	-3.097e+00	5.322e+00	-0.582	0.560861	
## season318	-3.098e+00	5.322e+00	-0.582	0.560825	
## season319	-3.098e+00	5.322e+00	-0.582	0.560789	
## season320	-3.098e+00	5.322e+00	-0.582	0.560753	
## season321	-3.099e+00	5.322e+00	-0.582	0.560716	
## season322	-3.099e+00	5.322e+00	-0.582	0.560680	
## season323	-2.599e+00	5.322e+00	-0.488	0.625521	
## season324	-3.100e+00	5.322e+00	-0.582	0.560608	
## season325	-3.100e+00	5.322e+00	-0.582	0.560572	
## season326	-3.100e+00	5.322e+00	-0.582	0.560536	
## season327	-2.601e+00	5.322e+00	-0.489	0.625368	
## season328	-3.101e+00	5.323e+00	-0.583	0.560464	
## season329	-2.101e+00	5.323e+00	-0.395	0.693204	
## season330	-2.102e+00	5.323e+00	-0.395	0.693163	
## season331	-2.602e+00	5.323e+00	-0.489	0.625214	
## season332	-2.602e+00	5.323e+00	-0.489	0.625176	
## season333	-2.602e+00	5.323e+00	-0.489	0.625138	
## season334	-2.103e+00	5.323e+00	-0.395	0.693001	

```

## season335 -3.103e+00 5.323e+00 -0.583 0.560212
## season336 -2.603e+00 5.323e+00 -0.489 0.625023
## season337 -1.037e-01 5.323e+00 -0.019 0.984471
## season338 -1.604e+00 5.323e+00 -0.301 0.763304
## season339 -2.104e+00 5.323e+00 -0.395 0.692799
## season340 -1.605e+00 5.323e+00 -0.301 0.763220
## season341 -1.105e+00 5.323e+00 -0.208 0.835663
## season342 -1.105e+00 5.323e+00 -0.208 0.835619
## season343 -1.106e+00 5.323e+00 -0.208 0.835575
## season344 -3.106e+00 5.323e+00 -0.583 0.559889
## season345 -6.061e-01 5.323e+00 -0.114 0.909397
## season346 -1.606e+00 5.323e+00 -0.302 0.762966
## season347 -1.607e+00 5.323e+00 -0.302 0.762923
## season348 -2.607e+00 5.323e+00 -0.490 0.624564
## season349 -1.607e+00 5.323e+00 -0.302 0.762839
## season350 8.923e-01 5.323e+00 0.168 0.866958
## season351 -2.608e+00 5.323e+00 -0.490 0.624449
## season352 -2.108e+00 5.324e+00 -0.396 0.692274
## season353 -2.109e+00 5.324e+00 -0.396 0.692234
## season354 -6.089e-01 5.324e+00 -0.114 0.908990
## season355 -1.609e+00 5.324e+00 -0.302 0.762585
## season356 -6.095e-01 5.324e+00 -0.114 0.908900
## season357 -2.610e+00 5.324e+00 -0.490 0.624220
## season358 -1.110e+00 5.324e+00 -0.209 0.834917
## season359 -1.110e+00 5.324e+00 -0.209 0.834873
## season360 -2.111e+00 5.324e+00 -0.396 0.691952
## season361 -2.111e+00 5.324e+00 -0.397 0.691911
## season362 -2.111e+00 5.324e+00 -0.397 0.691871
## season363 -2.112e+00 5.324e+00 -0.397 0.691831
## season364 -6.120e-01 5.324e+00 -0.115 0.908539
## season365 -2.112e+00 5.324e+00 -0.397 0.691750
## season366 -1.113e+00 5.324e+00 -0.209 0.834566
## season367 -6.129e-01 5.324e+00 -0.115 0.908404
## season368 8.868e-01 5.324e+00 0.167 0.867797
## season369 -2.614e+00 5.324e+00 -0.491 0.623764
## season370 -1.114e+00 5.324e+00 -0.209 0.834391
## season371 3.859e-01 5.324e+00 0.072 0.942262
## season372 -2.614e+00 5.324e+00 -0.491 0.623650
## season373 -2.615e+00 5.324e+00 -0.491 0.623612
## season374 -2.615e+00 5.325e+00 -0.491 0.623574
## season375 -6.154e-01 5.325e+00 -0.116 0.908043
## season376 -2.616e+00 5.325e+00 -0.491 0.623498
## season377 -3.116e+00 5.325e+00 -0.585 0.558712
## season378 -3.116e+00 5.325e+00 -0.585 0.558677
## season379 -1.166e-01 5.325e+00 -0.022 0.982537
## season380 -6.169e-01 5.325e+00 -0.116 0.907817
## season381 2.383e+00 5.325e+00 0.447 0.654749
## season382 3.882e+00 5.325e+00 0.729 0.466320
## season383 8.382e+00 5.325e+00 1.574 0.116177
## season384 1.438e+01 5.325e+00 2.701 0.007184 **
## season385 2.238e+01 5.325e+00 4.203 3.19e-05 ***
## season386 1.488e+01 5.325e+00 2.795 0.005424 **
## season387 1.488e+01 5.325e+00 2.794 0.005426 **
## season388 1.438e+01 5.325e+00 2.700 0.007191 **

```

```

## season389 1.788e+01 5.325e+00 3.358 0.000854 ***
## season390 1.738e+01 5.325e+00 3.264 0.001186 **
## season391 1.738e+01 5.325e+00 3.264 0.001186 **
## season392 1.788e+01 5.325e+00 3.357 0.000855 ***
## season393 1.838e+01 5.325e+00 3.451 0.000612 ***
## season394 2.088e+01 5.325e+00 3.921 0.000102 ***
## season395 2.488e+01 5.326e+00 4.672 3.98e-06 ***
## season396 2.988e+01 5.326e+00 5.610 3.58e-08 ***
## season397 2.638e+01 5.326e+00 4.953 1.04e-06 ***
## season398 3.338e+01 5.326e+00 6.267 8.77e-10 ***
## season399 2.738e+01 5.326e+00 5.141 4.13e-07 ***
## season400 3.288e+01 5.326e+00 6.173 1.52e-09 ***
## season401 3.688e+01 5.326e+00 6.924 1.56e-11 ***
## season402 3.638e+01 5.326e+00 6.830 2.83e-11 ***
## season403 4.488e+01 5.326e+00 8.426 5.13e-16 ***
## season404 4.388e+01 5.326e+00 8.238 2.02e-15 ***
## season405 5.238e+01 5.326e+00 9.834 < 2e-16 ***
## season406 4.588e+01 5.326e+00 8.613 < 2e-16 ***
## season407 6.387e+01 5.326e+00 11.993 < 2e-16 ***
## season408 4.937e+01 5.326e+00 9.270 < 2e-16 ***
## season409 4.187e+01 5.326e+00 7.862 2.94e-14 ***
## season410 5.787e+01 5.326e+00 10.866 < 2e-16 ***
## season411 4.487e+01 5.326e+00 8.425 5.17e-16 ***
## season412 3.537e+01 5.326e+00 6.641 9.20e-11 ***
## season413 3.137e+01 5.326e+00 5.890 7.68e-09 ***
## season414 3.187e+01 5.326e+00 5.984 4.52e-09 ***
## season415 3.637e+01 5.327e+00 6.829 2.86e-11 ***
## season416 3.637e+01 5.327e+00 6.828 2.86e-11 ***
## season417 2.837e+01 5.327e+00 5.326 1.60e-07 ***
## season418 3.637e+01 5.327e+00 6.828 2.87e-11 ***
## season419 2.737e+01 5.327e+00 5.138 4.17e-07 ***
## season420 3.687e+01 5.327e+00 6.922 1.58e-11 ***
## season421 3.087e+01 5.327e+00 5.795 1.30e-08 ***
## season422 4.637e+01 5.327e+00 8.705 < 2e-16 ***
## season423 4.837e+01 5.327e+00 9.080 < 2e-16 ***
## season424 6.137e+01 5.327e+00 11.520 < 2e-16 ***
## season425 6.087e+01 5.327e+00 11.426 < 2e-16 ***
## season426 1.034e+02 5.327e+00 19.404 < 2e-16 ***
## season427 1.084e+02 5.327e+00 20.343 < 2e-16 ***
## season428 1.229e+02 5.327e+00 23.064 < 2e-16 ***
## season429 9.587e+01 5.327e+00 17.996 < 2e-16 ***
## season430 8.737e+01 5.327e+00 16.400 < 2e-16 ***
## season431 7.087e+01 5.327e+00 13.303 < 2e-16 ***
## season432 6.287e+01 5.327e+00 11.801 < 2e-16 ***
## season433 5.937e+01 5.327e+00 11.144 < 2e-16 ***
## season434 5.137e+01 5.328e+00 9.642 < 2e-16 ***
## season435 4.287e+01 5.328e+00 8.046 8.02e-15 ***
## season436 2.937e+01 5.328e+00 5.512 6.05e-08 ***
## season437 1.837e+01 5.328e+00 3.447 0.000621 ***
## season438 2.187e+01 5.328e+00 4.104 4.84e-05 ***
## season439 1.236e+01 5.328e+00 2.321 0.020752 *
## season440 7.365e+00 5.328e+00 1.382 0.167587
## season441 3.864e+00 5.328e+00 0.725 0.468662
## ---

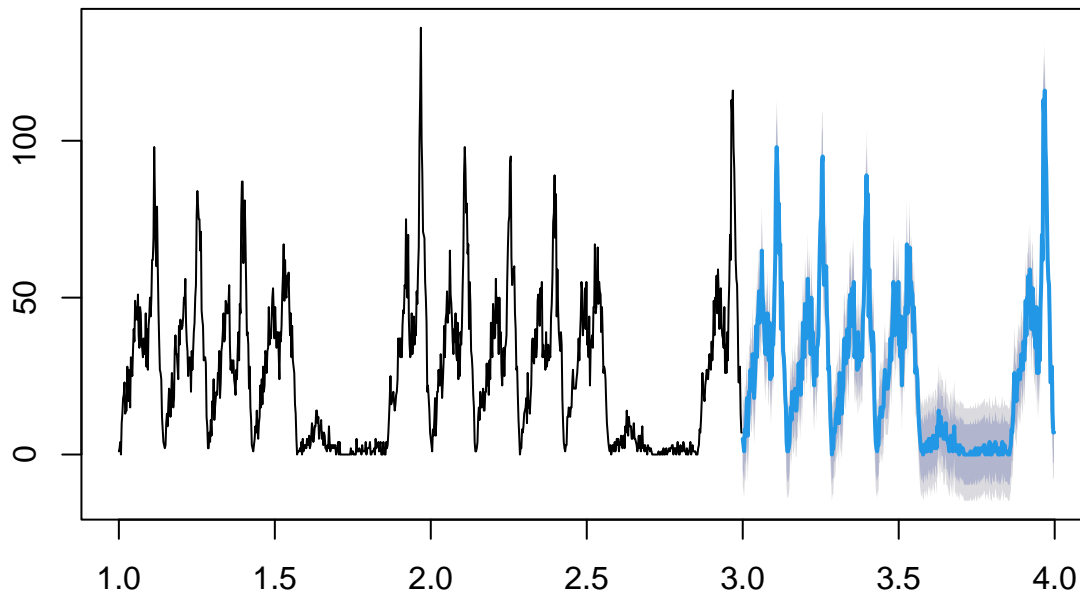
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.316 on 440 degrees of freedom
## Multiple R-squared:  0.9739, Adjusted R-squared:  0.9478
## F-statistic: 37.29 on 441 and 440 DF,  p-value: < 2.2e-16

#accuracy(snaive(train.msts, h = 441), td.msts)
sn.train <- snaive(train.msts, h = 441)
sn.pred <- forecast(sn.train, h=441)

plot(sn.pred)
```

## Forecasts from Seasonal naive method



```
line(td.ts)

##
## Call:
## line(td.ts)
##
## Coefficients:
## [1] 23.59070 -0.07143

accuracy(sn.pred, td.msts)
```

```
##              ME      RMSE      MAE  MPE  MAPE      MASE      ACF1
## Training set 0.13605442  7.510537  5.106576 -Inf  Inf  1.000000  0.1342442
## Test set    0.06575964 10.804278  5.970522 -Inf  Inf  1.169183  0.3648308
##              Theil's U
## Training set      NA
## Test set          0
```

## Final Model



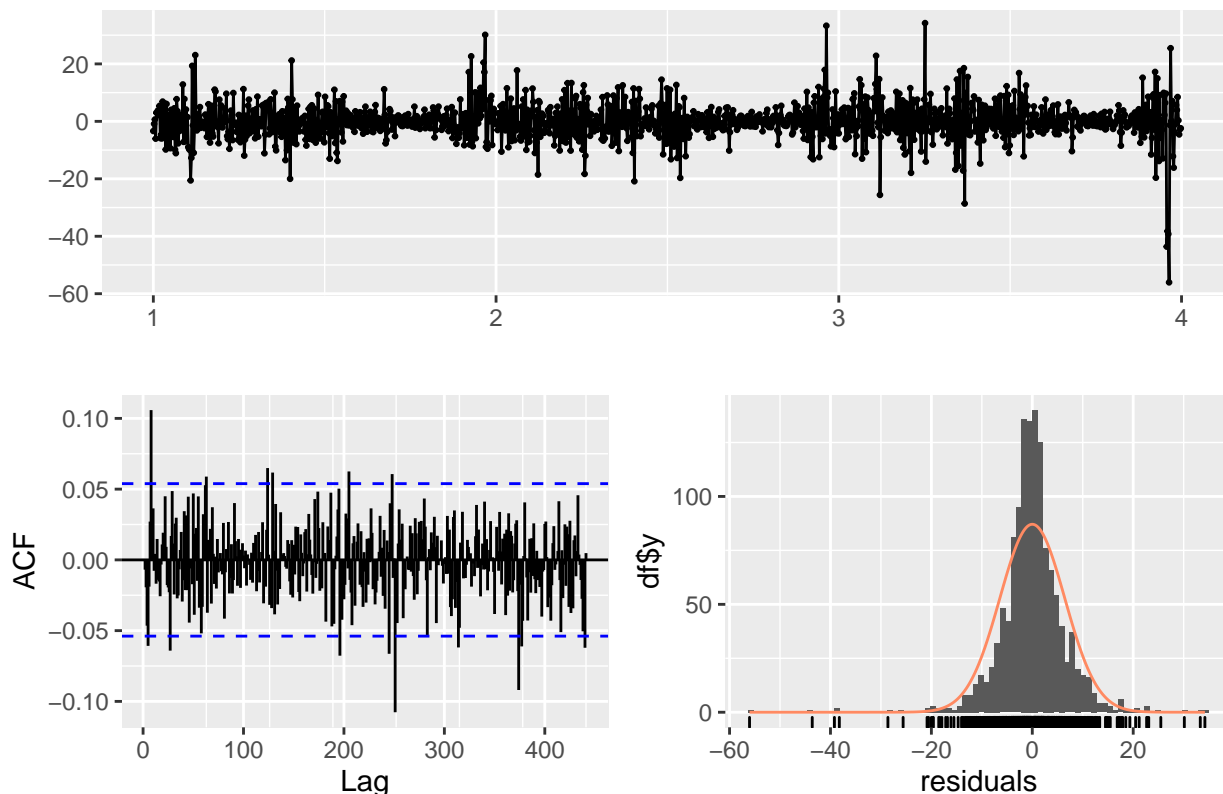
## ARIMA(0,0,3)

```
train2.msts <- msts(td$DEMAND[1:1323], seasonal.periods = c(63, 441)) # full period to train on

arima2.mod <- Arima(train2.msts, order = c(0, 0, 3),
                    xreg = fourier(train2.msts, K = c(20, 40)))
# new model trained on full dataset

arima2.pred <- forecast(arima2.mod, xreg = fourier(train2.msts, K = c(20, 40), h=189))
checkresiduals(arima2.pred)
```

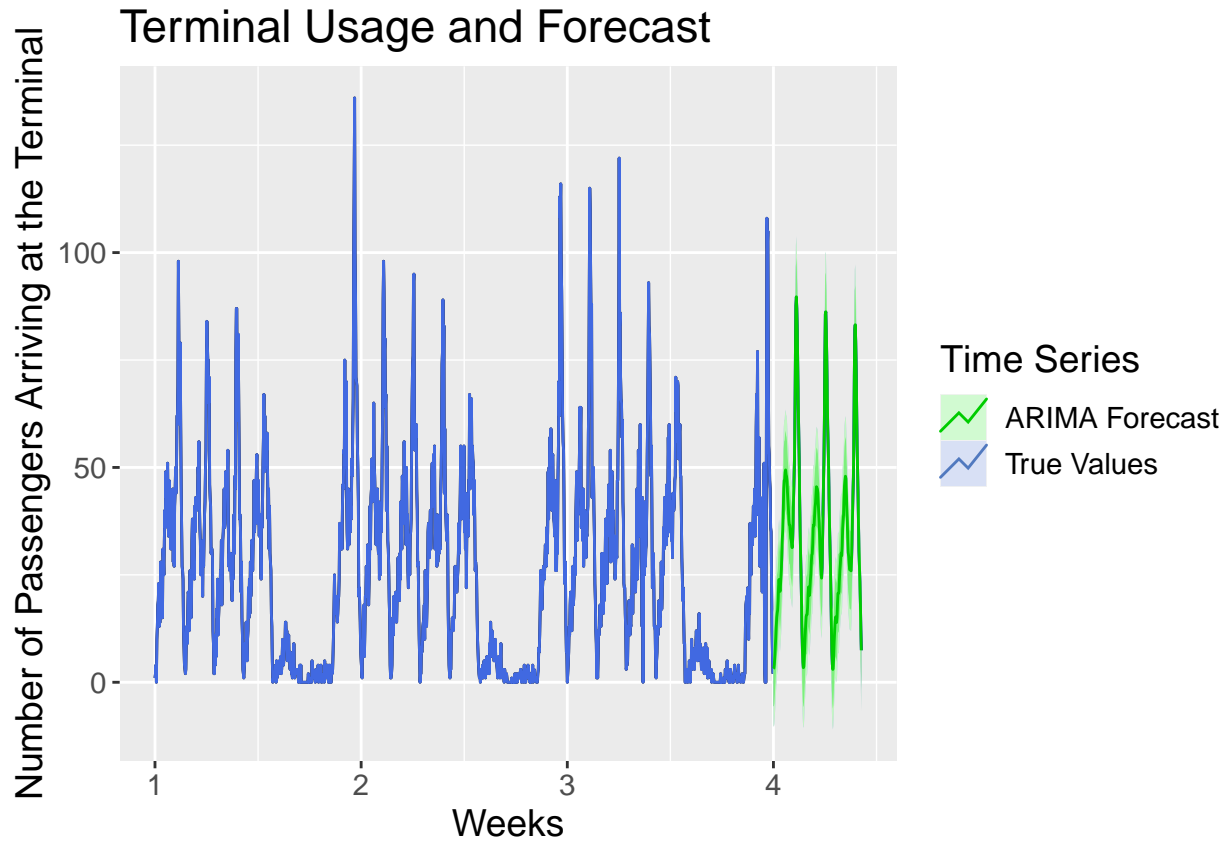
### Residuals from Regression with ARIMA(0,0,3) errors



```
##
##  Ljung-Box test
##
## data:  Residuals from Regression with ARIMA(0,0,3) errors
## Q* = 278.88, df = 151, p-value = 1.168e-09
##
## Model df: 114.    Total lags used: 265

# Plot results
clrs2 <- c("green", "royalblue")
autoplot(arima2.pred, series = "ARIMA") +
  ggtitle("Terminal Usage and Forecast") +
  xlab("Weeks") +
  ylab("Number of Passengers Arriving at the Terminal") +
  autolayer(td.msts, series = "True Values") +
  autolayer(arima2.pred, series = "ARIMA Forecast") +
  guides(colour=guide_legend(title="Time Series")) +
```

```
scale_color_manual(values=clrs2) +
theme(text = element_text(size = 14))
```



```
arima2.pred
```

##	Point	Forecast	Lo 80	Hi 80	Lo 95	Hi 95
##	4.000000	3.147802	-5.4740511	11.76965	-10.03818417	16.33379
##	4.002268	3.299259	-5.6207894	12.21931	-10.34277802	16.94130
##	4.004535	4.217660	-4.9031792	13.33850	-9.73146004	18.16678
##	4.006803	6.159160	-2.9669461	15.28527	-7.79801462	20.11633
##	4.009070	8.457134	-0.6689722	17.58324	-5.50004072	22.41431
##	4.011338	10.554903	1.4287968	19.68101	-3.40227177	24.51208
##	4.013605	12.942924	3.8168187	22.06903	-1.01424990	26.90010
##	4.015873	15.082156	5.9560503	24.20826	1.12498175	29.03933
##	4.018141	16.080024	6.9539179	25.20613	2.12284933	30.03720
##	4.020408	16.795741	7.6696353	25.92185	2.83856677	30.75292
##	4.022676	18.932786	9.8066803	28.05889	4.97561169	32.88996
##	4.024943	22.283095	13.1569895	31.40920	8.32592090	36.24027
##	4.027211	23.935027	14.8089211	33.06113	9.97785251	37.89220
##	4.029478	22.466115	13.3400092	31.59222	8.50894065	36.42329
##	4.031746	21.146586	12.0204799	30.27269	7.18941131	35.10376
##	4.034014	22.421420	13.2953147	31.54753	8.46424616	36.37859
##	4.036281	24.302288	15.1761827	33.42839	10.34511410	38.25946
##	4.038549	26.350204	17.2240980	35.47631	12.39302948	40.30738
##	4.040816	30.060591	20.9344855	39.18670	16.10341695	44.01777
##	4.043084	33.404508	24.2784026	42.53061	19.44733399	47.36168
##	4.045351	35.517268	26.3911623	44.64337	21.56009373	49.47444

## 4.047619	39.793324	30.6672181	48.91943	25.83614956	53.75050
## 4.049887	45.111114	35.9850082	54.23722	31.15393961	59.06829
## 4.052154	46.951811	37.8257048	56.07792	32.99463624	60.90898
## 4.054422	47.191329	38.0652232	56.31743	33.23415464	61.14850
## 4.056689	48.852275	39.7261693	57.97838	34.89510077	62.80945
## 4.058957	49.425975	40.2998695	58.55208	35.46880096	63.38315
## 4.061224	48.335977	39.2098713	57.46208	34.37880277	62.29315
## 4.063492	47.936913	38.8108077	57.06302	33.97973912	61.89409
## 4.065760	46.892637	37.7665315	56.01874	32.93546297	60.84981
## 4.068027	44.223135	35.0970293	53.34924	30.26596072	58.18031
## 4.070295	42.184721	33.0586152	51.31083	28.22754666	56.14190
## 4.072562	40.407118	31.2810121	49.53322	26.44994357	54.36429
## 4.074830	37.852573	28.7264676	46.97868	23.89539898	51.80975
## 4.077098	36.649189	27.5230833	45.77529	22.69201468	50.60636
## 4.079365	36.554993	27.4288873	45.68110	22.59781871	50.51217
## 4.081633	35.109647	25.9835417	44.23575	21.15247308	49.06682
## 4.083900	33.497617	24.3715110	42.62372	19.54044247	47.45479
## 4.086168	33.003755	23.8776491	42.12986	19.04658055	46.96093
## 4.088435	32.081077	22.9549713	41.20718	18.12390269	46.03825
## 4.090703	31.316218	22.1901122	40.44232	17.35904364	45.27339
## 4.092971	33.017867	23.8917615	42.14397	19.06069295	46.97504
## 4.095238	37.255391	28.1292856	46.38150	23.29821699	51.21257
## 4.097506	42.676494	33.5503880	51.80260	28.71931944	56.63367
## 4.099773	47.468401	38.3422950	56.59451	33.51122639	61.42558
## 4.102041	52.085987	42.9598816	61.21209	38.12881302	66.04316
## 4.104308	61.561096	52.4349903	70.68720	47.60392175	75.51827
## 4.106576	76.456438	67.3303318	85.58254	62.49926324	90.41361
## 4.108844	87.652760	78.5266539	96.77887	73.69558537	101.60993
## 4.111111	89.710227	80.5841216	98.83633	75.75305305	103.66740
## 4.113379	86.935428	77.8093220	96.06153	72.97825340	100.89260
## 4.115646	81.825183	72.6990776	90.95129	67.86800907	95.78236
## 4.117914	73.511607	64.3855009	82.63771	59.55443232	87.46878
## 4.120181	65.157958	56.0318518	74.28406	51.20078320	79.11513
## 4.122449	59.091637	49.9655315	68.21774	45.13446290	73.04881
## 4.124717	52.245297	43.1191914	61.37140	38.28812287	66.20247
## 4.126984	43.464544	34.3384380	52.59065	29.50736945	57.42172
## 4.129252	35.727138	26.6010325	44.85324	21.76996395	49.68431
## 4.131519	30.167806	21.0417002	39.29391	16.21063159	44.12498
## 4.133787	25.095960	15.9698539	34.22207	11.13878538	39.05313
## 4.136054	19.113417	9.9873117	28.23952	5.15624315	33.07059
## 4.138322	12.636125	3.5100195	21.76223	-1.32104904	26.59330
## 4.140590	7.533243	-1.5928630	16.65935	-6.42393152	21.49042
## 4.142857	4.694489	-4.4316172	13.82059	-9.26268576	18.65166
## 4.145125	3.383415	-5.7426903	12.50952	-10.57375887	17.34059
## 4.147392	3.664663	-5.4614428	12.79077	-10.29251139	17.62184
## 4.149660	5.839989	-3.2861166	14.96609	-8.11718516	19.79716
## 4.151927	8.603051	-0.5230551	17.72916	-5.35412372	22.56022
## 4.154195	11.043636	1.9175299	20.16974	-2.91353870	25.00081
## 4.156463	13.470600	4.3444942	22.59671	-0.48657437	27.42777
## 4.158730	15.296741	6.1706356	24.42285	1.33956706	29.25392
## 4.160998	15.716152	6.5900458	24.84226	1.75897721	29.67333
## 4.163265	15.768256	6.6421498	24.89436	1.81108120	29.72543
## 4.165533	17.362482	8.2363758	26.48859	3.40530723	31.31966
## 4.167800	20.450607	11.3245011	29.57671	6.49343255	34.40778

## 4.170068	22.180586	13.0544798	31.30669	8.22341121	36.13776
## 4.172336	21.067733	11.9416275	30.19384	7.11055893	35.02491
## 4.174603	20.213598	11.0874921	29.33970	6.25642354	34.17077
## 4.176871	21.833962	12.7078561	30.96007	7.87678755	35.79114
## 4.179138	23.713103	14.5869976	32.83921	9.75592907	37.67028
## 4.181406	25.247849	16.1217430	34.37395	11.29067440	39.20502
## 4.183673	27.877637	18.7515314	37.00374	13.92046279	41.83481
## 4.185941	29.644554	20.5184487	38.77066	15.68738012	43.60173
## 4.188209	29.872141	20.7460352	38.99825	15.91496663	43.82932
## 4.190476	32.226199	23.1000934	41.35230	18.26902487	46.18337
## 4.192744	35.890239	26.7641332	45.01634	21.93306459	49.84741
## 4.195011	36.630103	27.5039977	45.75621	22.67292915	50.58728
## 4.197279	36.536807	27.4107016	45.66291	22.57963301	50.49398
## 4.199546	38.741442	29.6153363	47.86755	24.78426771	52.69862
## 4.201814	40.716575	31.5904689	49.84268	26.75940034	54.67375
## 4.204082	41.737372	32.6112666	50.86348	27.78019804	55.69455
## 4.206349	43.896464	34.7703579	53.02257	29.93928929	57.85364
## 4.208617	45.515240	36.3891341	54.64135	31.55806555	59.47241
## 4.210884	45.238399	36.1122931	54.36450	31.28122453	59.19557
## 4.213152	44.971415	35.8453089	54.09752	31.01424030	58.92859
## 4.215420	44.077082	34.9509762	53.20319	30.11990767	58.03426
## 4.217687	41.386363	32.2602572	50.51247	27.42918868	55.34354
## 4.219955	39.064644	29.9385385	48.19075	25.10746992	53.02182
## 4.222222	37.080171	27.9540648	46.20628	23.12299620	51.03734
## 4.224490	33.326126	24.2000207	42.45223	19.36895216	47.28330
## 4.226757	29.421772	20.2956664	38.54788	15.46459778	43.37895
## 4.229025	27.083381	17.9572754	36.20949	13.12620684	41.04056
## 4.231293	25.101733	15.9756275	34.22784	11.14455892	39.05891
## 4.233560	24.232433	15.1063276	33.35854	10.27525903	38.18961
## 4.235828	26.743795	17.6176888	35.86990	12.78662021	40.70097
## 4.238095	32.462083	23.3359770	41.58819	18.50490845	46.41926
## 4.240363	39.644455	30.5183497	48.77056	25.68728117	53.60163
## 4.242630	46.032916	36.9068104	55.15902	32.07574186	59.99009
## 4.244898	51.690866	42.5647605	60.81697	37.73369193	65.64804
## 4.247166	61.409223	52.2831175	70.53533	47.45204897	75.36640
## 4.249433	75.719983	66.5938776	84.84609	61.76280906	89.67716
## 4.251701	85.689368	76.5632618	94.81547	71.73219320	99.64654
## 4.253968	86.228971	77.1028648	95.35508	72.27179619	100.18614
## 4.256236	82.069097	72.9429911	91.19520	68.11192253	96.02627
## 4.258503	76.090445	66.9643389	85.21655	62.13327031	90.04762
## 4.260771	67.669021	58.5429147	76.79513	53.71184617	81.62619
## 4.263039	60.011228	50.8851222	69.13733	46.05405366	73.96840
## 4.265306	55.281890	46.1557845	64.40800	41.32471598	69.23906
## 4.267574	50.096814	40.9707082	59.22292	36.13963958	64.05399
## 4.269841	42.922894	33.7967879	52.04900	28.96571928	56.88007
## 4.272109	36.403518	27.2774119	45.52962	22.44634336	50.36069
## 4.274376	31.472533	22.3464269	40.59864	17.51535832	45.42971
## 4.276644	26.418630	17.2925244	35.54474	12.46145579	40.37580
## 4.278912	19.994759	10.8686531	29.12086	6.03758453	33.95193
## 4.281179	12.880751	3.7546449	22.00686	-1.07642371	26.83792
## 4.283447	7.231986	-1.8941198	16.35809	-6.72518840	21.18916
## 4.285714	4.153735	-4.9723703	13.27984	-9.80343891	18.11091
## 4.287982	2.987965	-6.1381409	12.11407	-10.96920952	16.94514
## 4.290249	3.721118	-5.4049876	12.84722	-10.23605615	17.67829

## 4.292517	6.453129	-2.6729765	15.57924	-7.50404507	20.41030
## 4.294785	9.626346	0.5002404	18.75245	-4.33082820	23.58352
## 4.297052	12.120600	2.9944942	21.24671	-1.83657434	26.07777
## 4.299320	14.153723	5.0276168	23.27983	0.19654828	28.11090
## 4.301587	15.204871	6.0787651	24.33098	1.24769653	29.16205
## 4.303855	14.677737	5.5516315	23.80384	0.72056295	28.63491
## 4.306122	13.901334	4.7752280	23.02744	-0.05584058	27.85851
## 4.308390	15.061886	5.9357807	24.18799	1.10471216	29.01906
## 4.310658	18.282618	9.1565122	27.40872	4.32544364	32.23979
## 4.312925	20.715728	11.5896227	29.84183	6.75855416	34.67290
## 4.315193	20.703916	11.5778101	29.83002	6.74674150	34.66109
## 4.317460	21.047000	11.9208941	30.17311	7.08982556	35.00417
## 4.319728	23.621919	14.4958133	32.74802	9.66474475	37.57909
## 4.321995	25.945245	16.8191390	35.07135	11.98807046	39.90242
## 4.324263	27.307403	18.1812973	36.43351	13.35022872	41.26458
## 4.326531	29.244842	20.1187365	38.37095	15.28766789	43.20202
## 4.328798	30.078180	20.9520739	39.20429	16.12100535	44.03535
## 4.331066	29.508856	20.3827506	38.63496	15.55168204	43.46603
## 4.333333	31.560502	22.4343959	40.68661	17.60332736	45.51768
## 4.335601	35.633442	26.5073360	44.75955	21.67626747	49.59062
## 4.337868	37.484863	28.3587576	46.61097	23.52768907	51.44204
## 4.340136	38.952113	29.8260068	48.07822	24.99493819	52.90929
## 4.342404	42.722775	33.5966691	51.84888	28.76560056	56.67995
## 4.344671	45.753373	36.6272675	54.87948	31.79619897	59.71055
## 4.346939	46.877951	37.7518456	56.00406	32.92077701	60.83513
## 4.349206	47.958191	38.8320857	57.08430	34.00101713	61.91537
## 4.351474	47.377383	38.2512775	56.50349	33.42020889	61.33456
## 4.353741	44.138732	35.0126265	53.26484	30.18155791	58.09591
## 4.356009	40.723636	31.5975306	49.84974	26.76646207	54.68081
## 4.358277	37.145990	28.0198845	46.27210	23.18881590	51.10316
## 4.360544	32.796271	23.6701648	41.92238	18.83909625	46.75344
## 4.362812	30.162711	21.0366047	39.28882	16.20553618	44.11988
## 4.365079	29.216870	20.0907638	38.34298	15.25969526	43.17404
## 4.367347	27.538179	18.4120730	36.66428	13.58100442	41.49535
## 4.369615	26.201806	17.0756999	35.32791	12.24463138	40.15898
## 4.371882	26.295695	17.1695892	35.42180	12.33852066	40.25287
## 4.374150	26.062105	16.9359995	35.18821	12.10493089	40.01928
## 4.376417	25.924062	16.7979567	35.05017	11.96688814	39.88124
## 4.378685	28.104488	18.9783826	37.23059	14.14731404	42.06166
## 4.380952	32.660268	23.5341625	41.78637	18.70309391	46.61744
## 4.383220	38.268209	29.1421035	47.39432	24.31103492	52.22538
## 4.385488	43.144510	34.0184038	52.27062	29.18733523	57.10168
## 4.387755	47.740458	38.6143522	56.86656	33.78328358	61.69763
## 4.390023	57.042841	47.9167349	66.16895	43.08566631	71.00001
## 4.392290	71.546486	62.4203806	80.67259	57.58931203	85.50366
## 4.394558	82.080210	72.9541039	91.20632	68.12303533	96.03738
## 4.396825	83.211113	74.0850076	92.33722	69.25393904	97.16829
## 4.399093	79.340112	70.2140058	88.46622	65.38293720	93.29729
## 4.401361	73.145673	64.0195673	82.27178	59.18849870	87.10285
## 4.403628	63.995854	54.8697480	73.12196	50.03867947	77.95303
## 4.405896	55.285679	46.1595730	64.41178	41.32850438	69.24285
## 4.408163	49.503463	40.3773571	58.62957	35.54628858	63.46064
## 4.410431	43.618496	34.4923901	52.74460	29.66132149	57.57567
## 4.412698	36.367008	27.2409023	45.49311	22.40983374	50.32418

## 4.414966	30.490383	21.3642772	39.61649	16.53320863	44.44756
## 4.417234	26.816023	17.6899168	35.94213	12.85884823	40.77320
## 4.419501	23.352498	14.2263926	32.47860	9.39532401	37.30967
## 4.421769	18.483199	9.3570936	27.60931	4.52602501	32.44037
## 4.424036	12.539683	3.4135774	21.66579	-1.41749117	26.49686
## 4.426304	7.452686	-1.6734200	16.57879	-6.50448855	21.40986

## Final comments

The final forecast used an ARIMA(0, 0, 3) model and looks to have done a good job on the data taking into consideration morning and evening peaks as well as the weekly seasonality.