```
#STL
```

Seasonal Decomposition of Time Series

"Seasonal and Trend decomposition using Loess"

```
#load packages
library(readr)
library(plyr)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:plyr':
##
       arrange, count, desc, failwith, id, mutate, rename, summarise,
##
##
       summarize
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(ggplot2)
library(tsibble)
##
## Attaching package: 'tsibble'
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, union
library(fable)
## Loading required package: fabletools
library(zoo)
##
## Attaching package: 'zoo'
## The following object is masked from 'package:tsibble':
##
##
       index
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
```

```
library(forecast)
## Registered S3 method overwritten by 'quantmod':
    method
                      from
##
     as.zoo.data.frame zoo
library(fpp3)
## -- Attaching packages ------ fpp3 0.4.0 --
## v tibble
                3.1.2
                         v tsibbledata 0.3.0
                          v feasts
## v tidyr
               1.1.3
                                    0.2.2
## v lubridate 1.7.10
## -- Conflicts -----
                                          ----- fpp3_conflicts --
## x dplyr::arrange() masks plyr::arrange()
## x dplyr::count() masks plyr::count()
## x lubridate::date() masks base::date()
## x dplyr::failwith() masks plyr::failwith()
## x dplyr::filter()
                          masks stats::filter()
## x dplyr::id()
                          masks plyr::id()
## x zoo::index()
                         masks tsibble::index()
## x tsibble::intersect() masks base::intersect()
## x lubridate::interval() masks tsibble::interval()
## x dplyr::lag()
                        masks stats::lag()
## x dplyr::mutate()
                         masks plyr::mutate()
## x dplyr::rename()
                         masks plyr::rename()
## x tsibble::setdiff()
                          masks base::setdiff()
## x dplyr::summarise()
                          masks plyr::summarise()
## x dplyr::summarize()
                          masks plyr::summarize()
## x tsibble::union()
                          masks base::union()
df <- read_csv(file = "data/data_interpolated_with_diesel.csv")</pre>
## Warning: Missing column names filled in: 'X1' [1]
##
## -- Column specification -------
##
    X1 = col_double(),
##
    Mode = col_character(),
##
    ORegionDAT = col_character(),
##
    DRegionDAT = col_character(),
##
    yw = col_character(),
##
     sanitized_cost = col_double(),
##
    prcp = col_double(),
##
    tavg = col_double(),
    tmax = col_double(),
##
##
    tmin = col_double(),
##
    approx_cost = col_double(),
    diesel_price = col_double()
## )
```

```
ts <- df %>%
  mutate(yw = yearweek(yw)) %>%
  as_tsibble(key = c(Mode,ORegionDAT, DRegionDAT), index = yw)
#select diesel prices in just one lane (so it's not repeated)
dieselTS <- ts %>% select(diesel_price) %>% filter(Mode == "R", ORegionDAT == "CA_FRS", DRegionDAT == "
dieselTS
## # A tsibble: 235 x 5 [1W]
               Mode, ORegionDAT, DRegionDAT [1]
## # Key:
##
      diesel_price
                         yw Mode ORegionDAT DRegionDAT
##
            <dbl> <week> <chr> <chr>
                                             <chr>>
## 1
              2.59 2017 W01 R
                                  CA_FRS
                                             IL_CHI
                                             IL_CHI
## 2
              2.60 2017 W02 R
                                  CA FRS
## 3
              2.58 2017 W03 R
                                 CA_FRS
                                             IL_CHI
                                 CA_FRS
## 4
              2.57 2017 W04 R
                                             IL\_CHI
## 5
              2.56 2017 W05 R
                                  CA_FRS
                                             IL_CHI
## 6
              2.56 2017 W06 R
                                  CA_FRS
                                             IL\_CHI
                                  CA_FRS
                                             IL_CHI
## 7
              2.56 2017 W07 R
## 8
              2.57 2017 W08 R
                                  CA_FRS
                                             IL_CHI
                                             {\tt IL\_CHI}
## 9
              2.58 2017 W09 R
                                  CA_FRS
              2.58 2017 W10 R
                                  CA_FRS
                                             IL_CHI
## 10
## # ... with 225 more rows
#create training set - up through 2020 of the time series
train <- dieselTS %>%
 filter_index(~ "2019 W52")
#create and plot STL decompositions of training set
dcmp <- train %>% model(STL(diesel_price))
components(dcmp) %>% autoplot()
```

STL decomposition diesel_price = trend + season_year + remainder 3.4 -3.2 -3.0 -2.8 -2.6 -3.2 -3.0 -2.8 -2.6 -0.10 -0.05 -0.00 --0.05 **-**0.1 -0.0 -

```
#fit model
dcmp <- train %>%
  model(stlf = decomposition_model(
    STL(diesel_price),
    SNAIVE(season_year),
    ARIMA(season_adjust)
))
```

2019 W01

2020 W01

Warning in sqrt(diag(best\$var.coef)): NaNs produced

2018 W01

-0.1 **-**

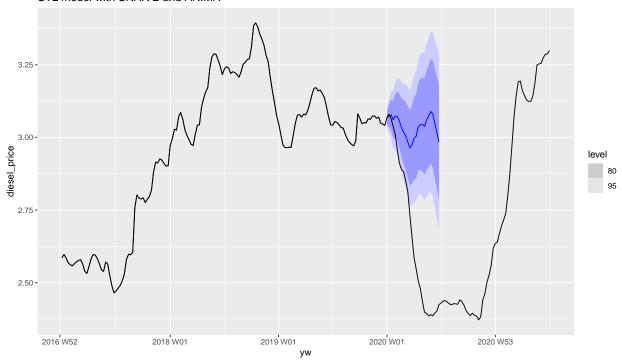
2016 W52

```
fc <- dcmp %>%
forecast(h=26)
```

```
#plot
fc %>%
  autoplot(train) +
  autolayer(dieselTS, colour = "black") +
  labs(title="STL model with SNAIVE and ARIMA")
```

Plot variable not specified, automatically selected '.vars = diesel_price'





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train <- dieselTS %>%
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#create and plot STL decompositions of training set
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STL decomposition

diesel_price = trend + season_year + remainder 3.25 -3.00 -2.75 -2.50 -3.00 -2.75 -2.50 -2.25 -0.10 -0.05 -0.00 --0.05 -0.2 -0.1 -0.0 --0.1 --0.2 **-**2016 W52 2018 W01 2019 W01 2020 W01 2020 W53 yw

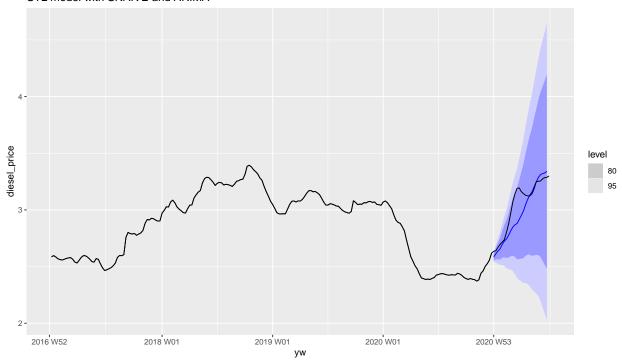
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STL model with SNAIVE and ARIMA



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STL decomposition

diesel_price = trend + season_year + remainder 3.25 -3.00 -2.75 -2.50 -3.2 -3.0 -2.8 -2.6 -2.4 -0.04 -0.00 --0.04 -0.0 --0.2 **-**2016 W52 2018 W01 2019 W01 2020 W01 2020 W53

```
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