

```
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'  
  
## 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(TTR)  
library(imputeTS)
```

```
## Registered S3 method overwritten by 'quantmod':  
##   method           from  
##   as.zoo.data.frame zoo
```

```
library(forecast)
```

```
#read in the interpolated data  
data <- readr::read_csv(file = 'data/data_interpolated.csv')
```

```
## Warning: Missing column names filled in: 'X1' [1]
```

```
##  
## -- Column specification -----  
## cols(  
##   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()  
## )
```

```
# filter to just refriddered trucks in CHI  
df_R_IL <- data %>%  
  filter(data$Mode == "R", data$DRegionDAT == "IL_CHI")
```

```

# convert
train <- ts(df_R_IL$approx_cost,frequency = 52,start = c(2017,01),end = c(2019,01))

#convert df_R_IL to time series to use autolayer when plotting later
ts <- ts(df_R_IL$approx_cost,frequency = 52,start = c(2017,01),end = c(2019,52))

#Here we want to use imputeTS. This imputation algorithm fills in all missing values in a time series.
train_1= na_interpolation(train)

#create tbats model and forecast
tbats_mod <- tbats(train_1)
tbats_for = forecast(tbats_mod, h=52)

#plot forecast with original data autolayer
autoplot(tbats_for) +
  autolayer(ts, color = "BLACK") +
  labs(
    y = "Approximate Cost",
    title = "Forecasts for weekly cost (CA_FRS to IL_CHI)"
  )

```

