## Forecaster Toolbox

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## Libraries

```
library(knitr)
library(kableExtra)
library(tidyverse)
library(gridExtra)
library(fpp2)
library(readxl)
```

### Exercise 3.1

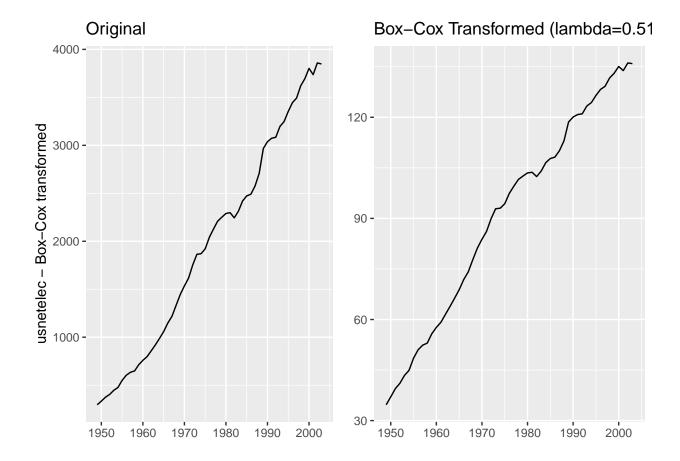
For the following series, find an appropriate Box-Cox transformation in order to stabilise the variance.

Since there are four items, I created a loop to contrast the original with the Box-Cox transformation of the data.

```
boxcox_trans_loop <- function(item, title){
  lambda <- BoxCox.lambda(item)
  Visulization1 <- autoplot(item) + ggtitle("Original") + ylab(title) + theme(axis.title.x = element_bl
  Visulization2 <- autoplot(BoxCox(item, lambda)) + ggtitle(paste0("Box-Cox Transformed (lambda=", roungrid.arrange(Visulization1, Visulization2, ncol = 2)
}</pre>
```

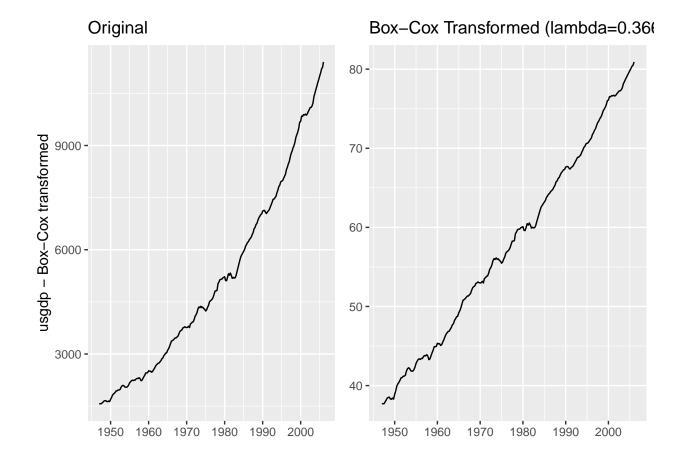
### usnetelec

```
boxcox_trans_loop(usnetelec, "usnetelec - Box-Cox transformed")
```



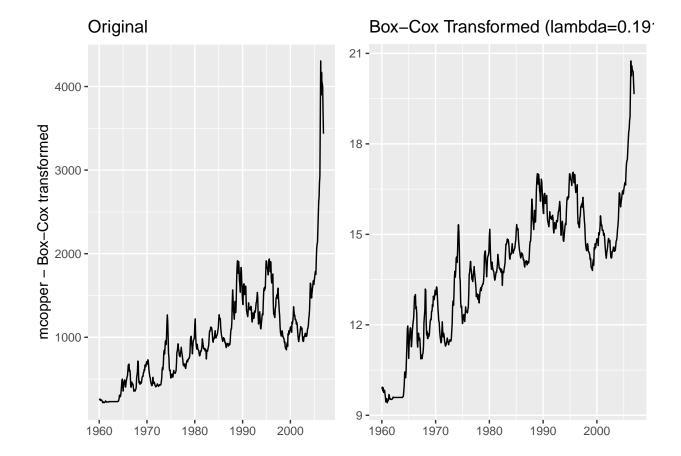
# usgdp

boxcox\_trans\_loop(usgdp, "usgdp - Box-Cox transformed")



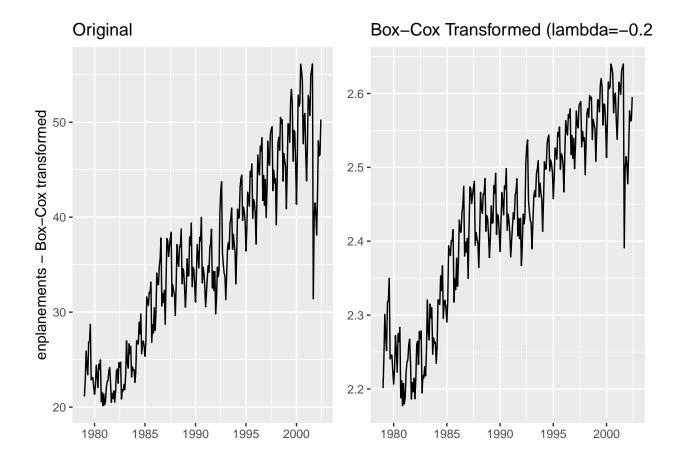
## mcopper

boxcox\_trans\_loop(mcopper, "mcopper - Box-Cox transformed")



## enplanements

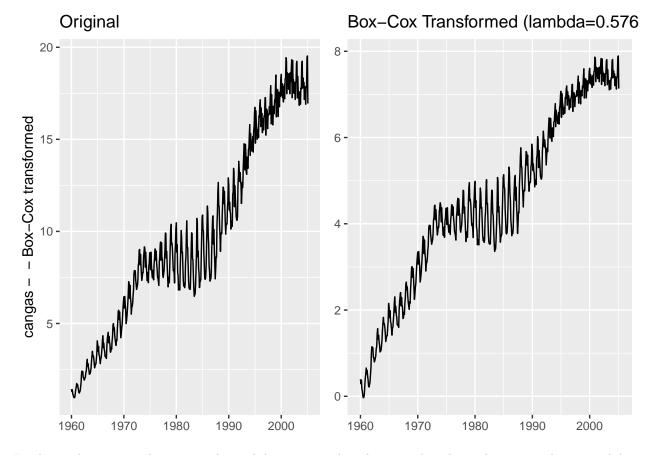
boxcox\_trans\_loop(enplanements, "enplanements - Box-Cox transformed")



# Exercise 3.2

Why is a Box-Cox transformation unhelpful for the cangas data?

boxcox\_trans\_loop(cangas, "cangas - - Box-Cox transformed")



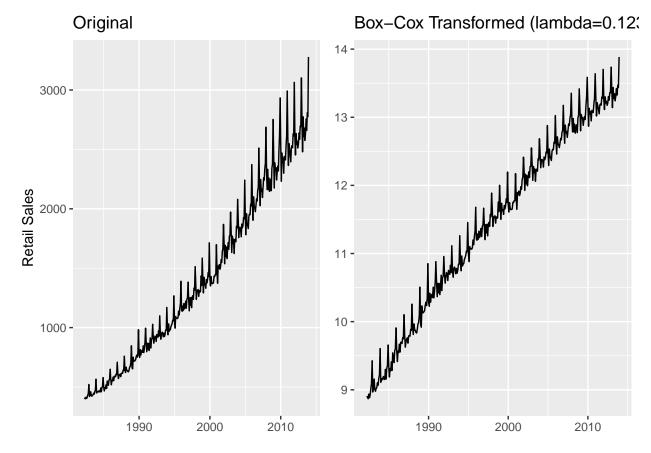
In above plot, we see that seasonal variability remains largely same thoughout the years, where variability increased after 1975 and reduced after 1990.

So to stabilise the variance, the transformation will need to "stretch" the seasonal variability before 1975; "shrink" the variability between 1975 and 1985; and then again "stretch" the variability after 1990.

## Exercise 3.3

What Box-Cox transformation would you select for your retail data?

```
retaildata <- read_excel("retail.xlsx", skip = 1)
myts <- ts(retaildata[, "A3349398A"], frequency = 12, start = c(1982, 4))
# myts <- ts(retaildata[, "A3349709X"], frequency = 12, start = c(1982, 4))
# myts <- ts(retaildata[, "A3349413L"], frequency = 12, start = c(1982, 4))
# myts <- ts(retaildata[, "A3349335T"], frequency = 12, start = c(1982, 4))
# myts <- ts(retaildata[, "A3349627V"], frequency = 12, start = c(1982, 4))
# myts <- ts(retaildata[, "A3349338X"], frequency = 12, start = c(1982, 4))
# myts <- ts(retaildata[, "A3349398A"], frequency = 12, start = c(1982, 4))
boxcox_trans_loop(myts, "Retail Sales")</pre>
```



In the original data, variance increased from 1990 to 2010+. But, in Box-cox transformed data, variance seems to have become uniform. The value for Lambda is .1232. So, this transformation appears to be effective.

## Exercise 3.8

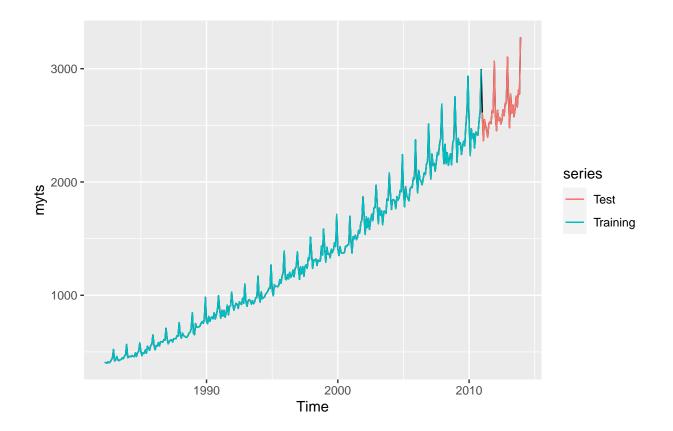
For your retail time series:

a) Split the data into two parts using

```
myts.train <- window(myts, end = c(2010, 12))
myts.test <- window(myts, start = 2011)</pre>
```

b) Check that your data have been split appropriately by producing the following plot.

```
autoplot(myts) +
autolayer(myts.train, series = "Training") +
autolayer(myts.test, series = "Test")
```



c) Calculate forecasts using snaive applied to myts.train.

### fc <- snaive(myts.train)</pre>

d) Compare the accuracy of your forecasts against the actual values stored in myts.test.

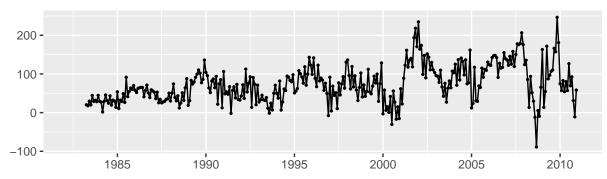
### accuracy(fc, myts.test)

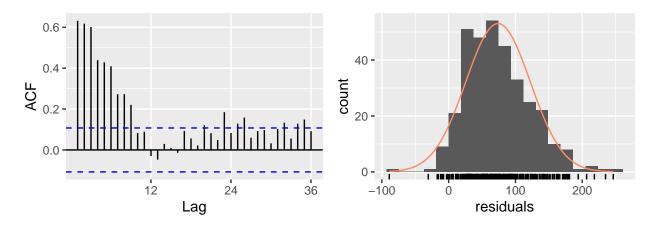
```
##
                       ME
                               RMSE
                                          MAE
                                                   MPE
                                                            MAPE
                                                                     MASE
                                                                               ACF1
## Training set 73.94114
                           88.31208 75.13514 6.068915 6.134838 1.000000 0.6312891
                115.00000 127.92727 115.00000 4.459712 4.459712 1.530576 0.2653013
## Test set
                Theil's U
##
## Training set
                       NA
## Test set
                0.7267171
```

e) Check the residuals.

#### checkresiduals(fc)

## Residuals from Seasonal naive method





```
##
## Ljung-Box test
##
## data: Residuals from Seasonal naive method
## Q* = 671.41, df = 24, p-value < 2.2e-16
##
## Model df: 0. Total lags used: 24</pre>
```

Do the residuals appear to be uncorrelated and normally distributed?

Yes, the residuals seem to be normally distributed but with a bit of a positive skew. The residuals don't appear to be uncorrelated. The Ljung-Box test shows a p value of that less than 0.05, which suggests incomplete information, and therefore the seasonal naive model is not reliable model.

f) How sensitive are the accuracy measures to the training/test split?

The accuracy measures are quite sensitive to the training/test split. The values are significantly different between the two. Probably the model doesn't generalize well.

Marker: 624-01