Examen 1

Luis Eduardo Jiménez del Muro

Librerias

```
library(tidyverse)
library(fpp3)
```

Graficar datos

```
data <- tidyquant::tq_get(
    x = "RIVN",
    from = today() - 365,
    to = today() - 1

) |>
    select(symbol, adjusted) |>
    mutate(
        t = seq_len(n())
    ) |>
    as_tsibble(index = t, key = symbol) |>
    relocate(symbol, t)
```

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

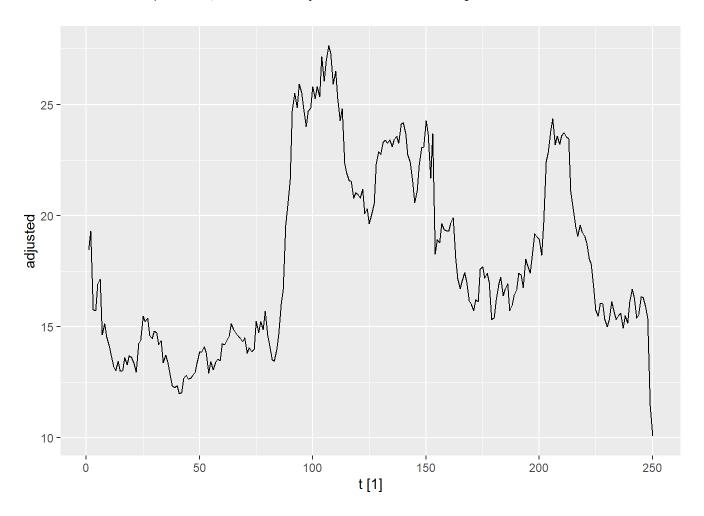
```
data
```

```
# A tsibble: 250 x 3 [1]
# Key:
             symbol [1]
              t adjusted
   symbol
   <chr> <int>
                   <dbl>
1 RIVN
                    18.5
              1
 2 RIVN
              2
                    19.3
 3 RIVN
              3
                    15.8
                    15.7
4 RIVN
              4
 5 RIVN
              5
                    16.9
6 RIVN
              6
                    17.1
7 RIVN
              7
                    14.6
8 RIVN
              8
                    15.1
9 RIVN
              9
                    14.5
10 RIVN
             10
                    14.2
# i 240 more rows
```

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autoplot(data)

Plot variable not specified, automatically selected `.vars = adjusted`



Obtener Lambda

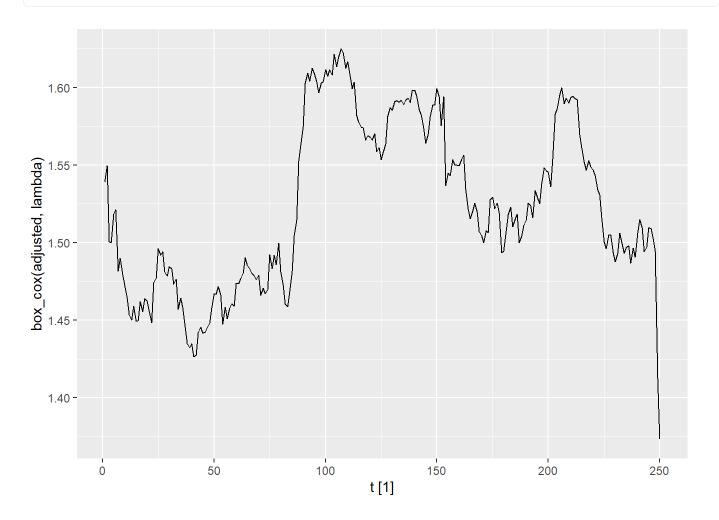
```
data |>
   features(adjusted, features=guerrero)
```

Grafica con box-cox

```
lambda <- -0.4972675
```

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```
data |>
  autoplot(box_cox(adjusted, lambda))
```



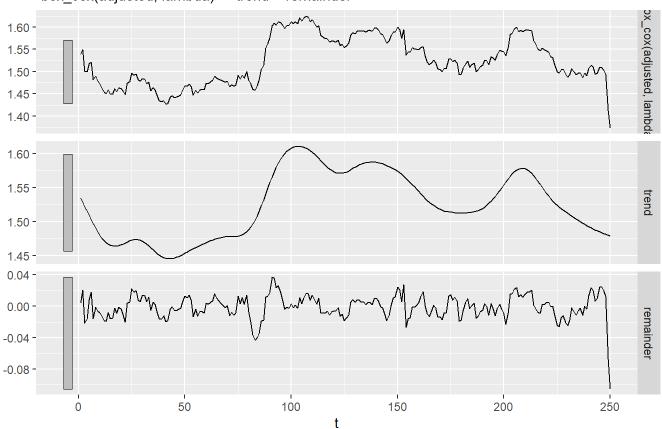
Descomponer para ver si vale la pena

```
data |>
  model(
    stl = STL(box_cox(adjusted, lambda), robust = TRUE)
  ) |>
  components() |>
  autoplot()
```

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STL decomposition

`box_cox(adjusted, lambda)` = trend + remainder



Seleccionar datos de entrenamiento

```
data_train <- data |>
  filter_index(. ~ "200")

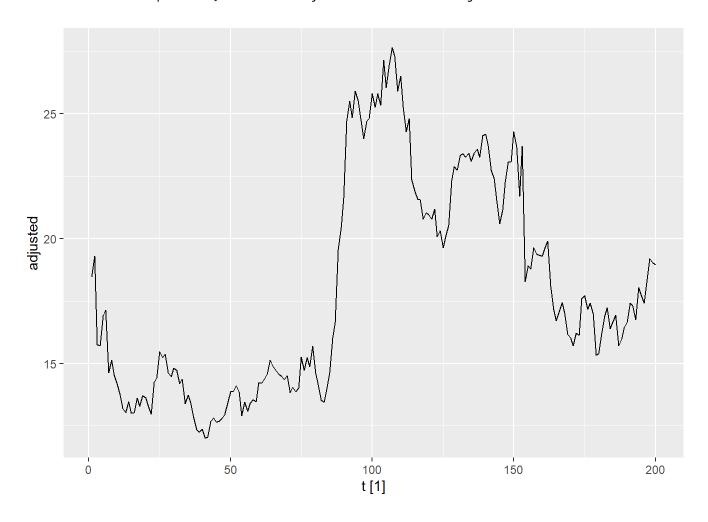
data_train
```

```
# A tsibble: 200 x 3 [1]
              symbol [1]
# Key:
   symbol
               t adjusted
   <chr>>
          <int>
                    <dbl>
1 RIVN
               1
                     18.5
               2
 2 RIVN
                     19.3
 3 RIVN
               3
                     15.8
4 RIVN
               4
                     15.7
 5 RIVN
               5
                     16.9
                     17.1
 6 RIVN
               6
               7
7 RIVN
                     14.6
8 RIVN
               8
                     15.1
               9
                     14.5
9 RIVN
10 RIVN
              10
                     14.2
# i 190 more rows
```

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```
data_train |>
  autoplot()
```

Plot variable not specified, automatically selected `.vars = adjusted`



Datos de entrenamiento con box-cox

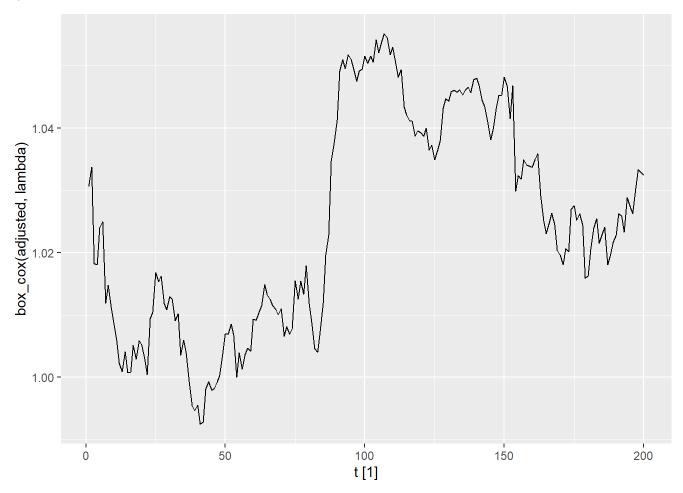
```
lambda <- data_train |>
  features(adjusted, features = guerrero) |>
  pull(lambda_guerrero)

lambda
```

[1] -0.8999268

```
data_train |>
  autoplot(box_cox(adjusted, lambda))
```

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Hacer los modelos para comparar cual vale mas la pena

```
data_fit <- data_train |>
  model(
    media = MEAN(box_cox(adjusted, lambda)),
    naive = NAIVE(box_cox(adjusted, lambda)),
    drift = RW(box_cox(adjusted, lambda) ~ drift()),
    # No se usa snaive porque no tiene estacionalidad
  )

data_fit
```

```
# A mable: 1 x 4
# Key: symbol [1]
symbol media naive drift
<chr> <model> <model> <model>
1 RIVN <MEAN> <NAIVE> <RW w/ drift>

accuracy(data_fit)
```

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```
# A tibble: 3 \times 11
  symbol .model .type
                               ME RMSE
                                          MAE
                                                 MPE MAPE MASE RMSSE
                                                                         ACF1
  <chr> <chr> <chr>
                            <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
                                                                        <dbl>
                                  4.37 3.61 -0.254 19.5
                                                           5.98 4.92 0.978
1 RIVN media Training 0.892
2 RIVN naive Training 0.00251
                                  0.889 0.604 -0.100 3.34 1
                                                                 1
                                                                      -0.0811
3 RIVN drift Training 0.0000728 0.889 0.604 -0.113 3.34 1.00 1.00 -0.0811
El que tiene menos error es naive con mae
```

Preparar datos para el pronóstico

```
data_fit <- data_train |>
    model(
        naive = NAIVE(box_cox(adjusted, lambda)),
    )

data_fit
```

```
# A mable: 1 x 2
# Key: symbol [1]
  symbol naive
  <chr> <model>
1 RIVN <NAIVE>
```

P-value

```
data_fit |>
  augment() |>
  features(.innov, ljung_box, lag = 10, dof = 0)
```

Como el p-values es mayor a α ; $p_{value}=0.05$, **NO RECHAZAMOS** $H_0:$ "Los residuos son ruido blanco"

Gráficas de residuos

```
data_fit |>
  select(naive) |>
  gg_tsresiduals()
```

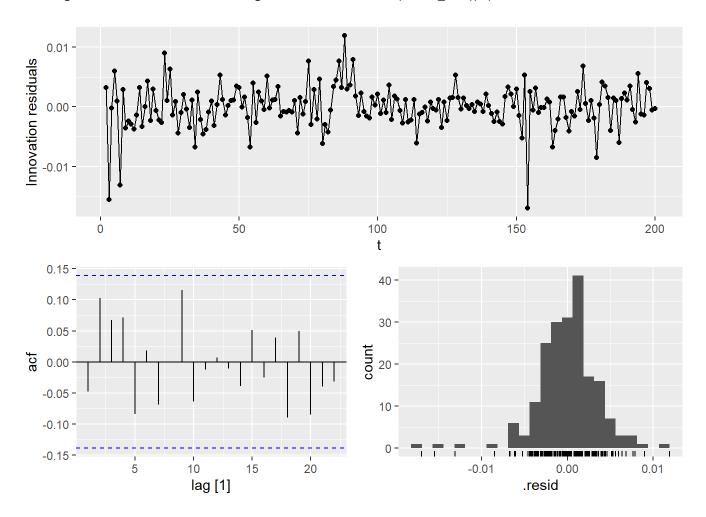
Warning: Removed 1 row containing missing values (`geom_line()`).

Warning: Removed 1 rows containing missing values (`geom_point()`).

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Warning: Removed 1 rows containing non-finite values (`stat_bin()`).

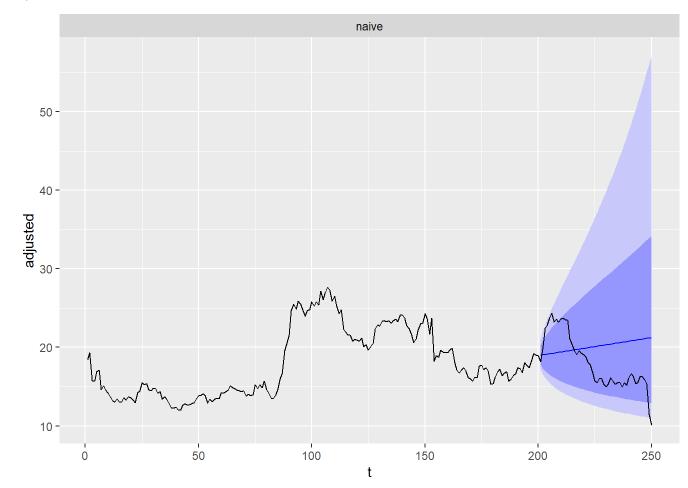


Pronóstico

```
data_fc <- data_fit |>
  forecast(h = 50)

data_fc |>
  autoplot(data) +
  facet_wrap(~ .model) +
  theme(legend.position = "none")
```

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