ETS

Annual oil production in Saudi Arabia

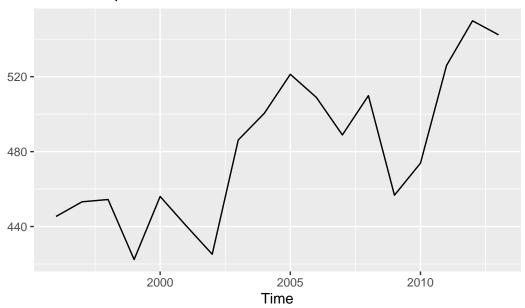


Figure 1: 1996 Saudi Arabia

```
ses() 2014 \sim 2016
```

```
ses(oil_1996, h = 3) \%>\%
  summary()
##
## Forecast method: Simple exponential smoothing
##
## Model Information:
## Simple exponential smoothing
##
## Call:
    ses(y = oil_1996, h = 3)
##
##
##
     Smoothing parameters:
##
       alpha = 0.8339
##
##
     Initial states:
       l = 446.5868
##
##
##
     sigma: 29.8282
```

```
##
  ##
           AIC
                  AICc
                             BIC
  ## 178.1430 179.8573 180.8141
  ##
  ## Error measures:
                                                   MPE
  ##
                         ME
                                RMSE
                                          MAE
                                                           MAPE
                                                                      MASE
                                                                                  ACF1
  ## Training set 6.401975 28.12234 22.2587 1.097574 4.610635 0.9256774 -0.03377748
  ##
  ## Forecasts:
                                                Lo 95
  ##
          Point Forecast
                             Lo 80
                                      Hi 80
                542.6806 504.4541 580.9070 484.2183 601.1429
  ## 2015
                 542.6806 492.9073 592.4539 466.5589 618.8023
  ## 2016
                542.6806 483.5747 601.7864 452.2860 633.0752
Level
           \alpha = 0.8339
                            level
                                                  Figure 2
  ses(oil_1996, h = 3) \%
    autoplot() +
```

Forecasts from Simple exponential smoothing

labs(y = NULL)

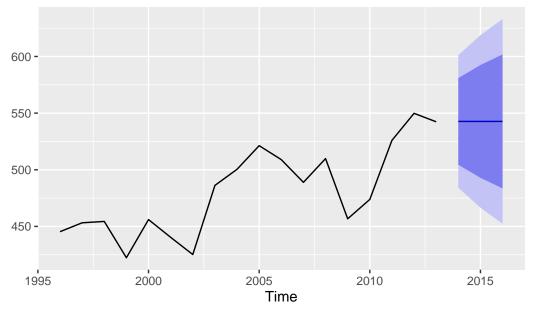


Figure 2: 1996 Saudi Arabia 2014

level , h . 80% , 95%

.

• Trend method : fpp2::ausair

 $\verb"fpp2::ausair" 1970 2016" . Holt's linear trend damped Holt's trend$

.

Figure 3 . . .

Air Transport Passengers Australia

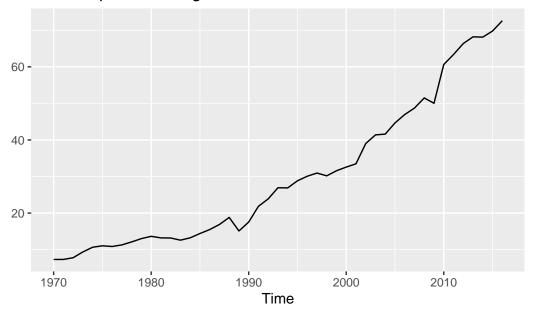


Figure 3: 1970 2016

Holt's linear trend holt() , 15 Figure 4 .

```
holt(ausair, h = 15) %>%
  autoplot()
```

Forecasts from Holt's method

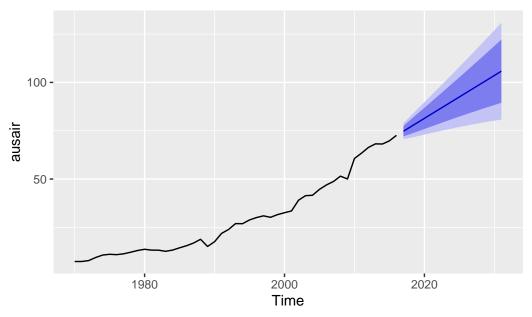


Figure 4: ausair Holt's linear trend method

```
Damped Holt's trend
```

Figure 5 .

```
holt(ausair, h = 15, damped = TRUE) %>%
  autoplot()
```

Forecasts from Damped Holt's method

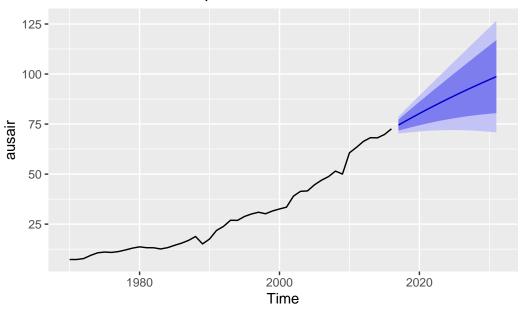


Figure 5: ausair damped Holt's trend method

Figure 4 Holt's linear trend method , damped Holt's trend method Figure 5 .

• Holt-Winters' seasonal method : fpp2::austourists

austourists 199 2015 . Figure 6 .

```
autoplot(austourists) +
  labs(y = NULL, title = "International Tourists to Australia")
```

International Tourists to Australia

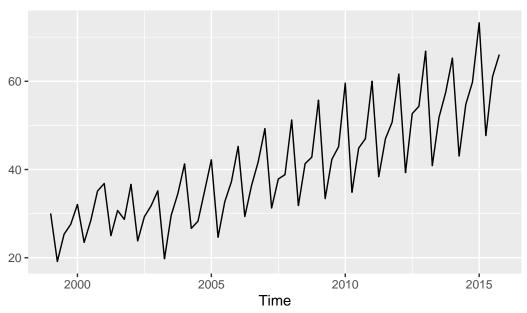


Figure 6: 199 2015

Holt-Winters' seasonal

Figure 7

Forecasts from Holt-Winters' additive method

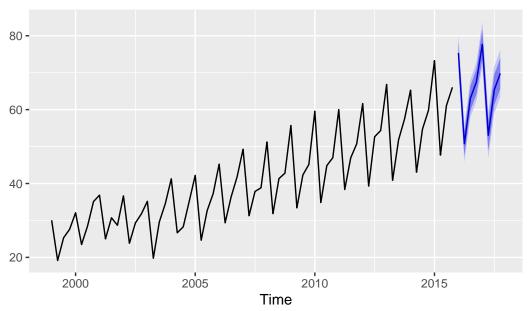


Figure 7: austourists Holt-Winters' additive seasonal method

Figure 8

```
hw(austourists, seasonal = "multiplicative") %>%
autoplot() + labs(y = NULL)
```

Forecasts from Holt-Winters' multiplicative method

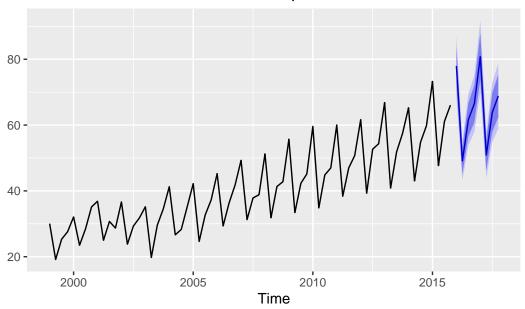


Figure 8: austourists Holt-Winters' multiplicative seasonal method

ETS

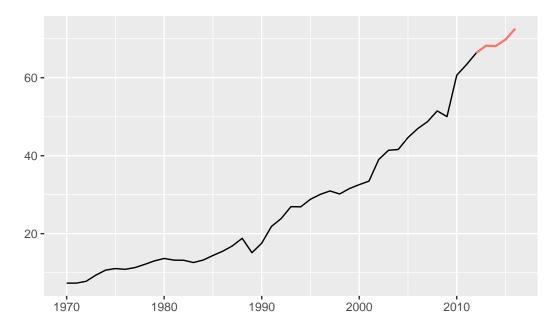


Figure 9: ausair

ets() ETS ,

```
fit_air <- ets(train_air)</pre>
fit_air
## ETS(M, A, N)
##
## Call:
##
    ets(y = train\_air)
##
     Smoothing parameters:
##
       alpha = 0.9999
##
      beta = 0.024
##
##
     Initial states:
##
       l = 6.5399
##
       b = 0.7358
##
##
     sigma: 0.08
##
##
      AIC AICC BIC
##
```

206.1828 207.8495 214.8712

figure 10 . ,

autoplot(fit_air)

Components of ETS(M,A,N) method

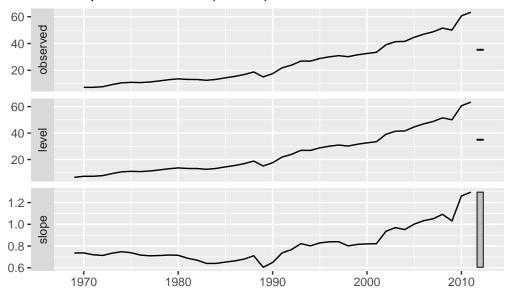
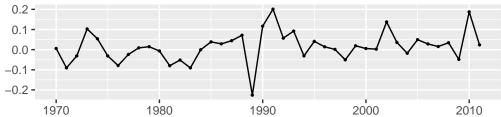


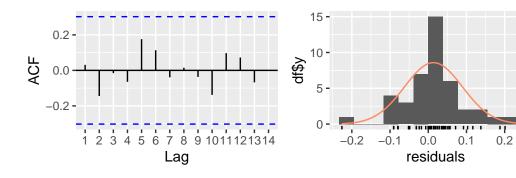
Figure 10: ETS

checkresiduals() . . .

checkresiduals(fit_air)

Residuals from ETS(M,A,N)





```
##
## Ljung-Box test
##
## data: Residuals from ETS(M,A,N)
## Q* = 3.5236, df = 8, p-value = 0.8973
##
## Model df: 0. Total lags used: 8
```

. autoplot() forecast() fc_air training data . include = 0 . Figure 11

```
library(patchwork)
p1 <- autoplot(fc_air) +</pre>
  autolayer(test_air, color = "red", size = .8) +
  labs(y = NULL, x = NULL)
p2 <- autoplot(fc_air, include = 0) +</pre>
  autolayer(test_air, color = "red", size=.8) +
  labs(y = NULL, x = NULL)
p1 + p2
```

Forecasts from ETS(M,A,N) Forecasts from ETS(M,A,N) 90 -75 **-**80 -70 -50 -60 -25 -50 -1980 1990 2010 2012 2013 2014 2015 2016 1970 2000

ETS Figure 11: fpp2::ausair

```
2: 1999 2015
                          (fpp2::austourists)
```

austourists 1999 2015. 2013 4 training data 2014 1test data

```
train_tour <- window(austourists, end = c(2013, 4))</pre>
test_tour <- window(austourists, start = c(2014, 1))
```

Figure 12 . Test data

```
autoplot(window(austourists, end = c(2014,1))) +
 autolayer(window(austourists, start = c(2014,1)), size = .8) +
```

```
labs(y = NULL) +
theme(legend.position = "none")
```

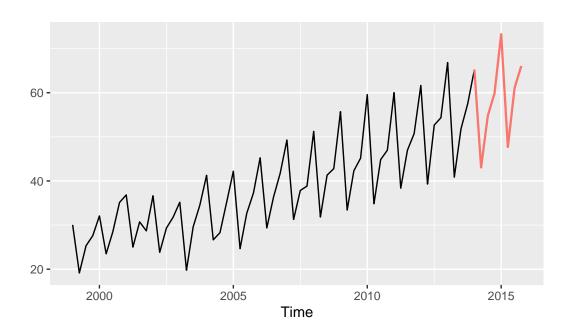


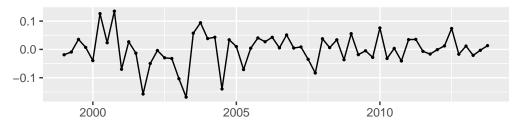
Figure 12: austourists

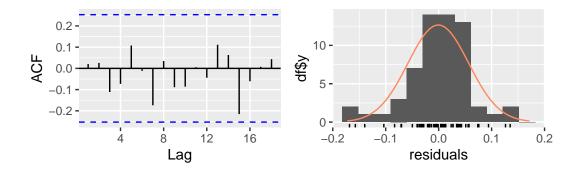
```
ets()
```

```
fit_tour <- ets(train_tour)</pre>
fit_tour
## ETS(M, A, M)
##
## Call:
##
    ets(y = train_tour)
##
##
     Smoothing\ parameters:
       alpha = 0.4189
##
       beta = 1e-04
##
##
       gamma = 1e-04
##
##
     Initial states:
        l = 24.2672
##
```

```
b = 0.5179
  ##
         s = 1.0367 \ 0.9578 \ 0.7697 \ 1.2358
  ##
  ##
  ##
      sigma: 0.0612
  ##
          AIC
                AICc
  ##
                         BIC
  ## 353.3882 356.9882 372.2373
                                                           ETS
ETS(M,A,M)
 . ets() Box\_Cox lambda 0
  fit_lntour <- ets(train_tour, lambda = 0)</pre>
  fit_lntour
  ## ETS(A,A,A)
  ##
  ## Call:
  ## ets(y = train_tour, lambda = 0)
  ##
  ##
      Box-Cox transformation: lambda= 0
  ##
  ##
      Smoothing parameters:
  ##
        alpha = 0.337
        beta = 1e-04
  ##
         gamma = 0.0137
  ##
  ##
  ##
      Initial states:
  ##
         l = 3.2161
        b = 0.0122
  ##
  ##
        s = 0.055 - 0.0254 - 0.2477 0.2181
  ##
  ##
      sigma: 0.0639
  ##
  ##
          AIC
                  AICc
                             BIC
  ## -75.06808 -71.46808 -56.21898
 fit_tour
  checkresiduals(fit_tour)
```

Residuals from ETS(M,A,M)



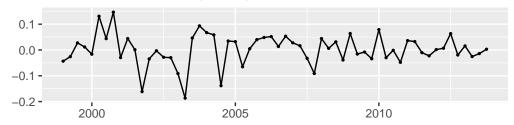


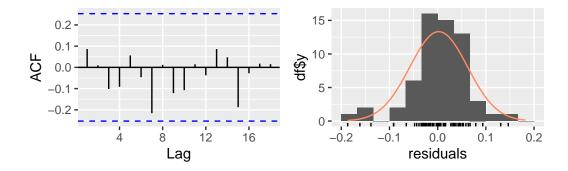
```
##
## Ljung-Box test
##
## data: Residuals from ETS(M,A,M)
## Q* = 4.2317, df = 8, p-value = 0.8356
##
## Model df: 0. Total lags used: 8

fit_lntour .
```

checkresiduals(fit_lntour)

Residuals from ETS(A,A,A)





```
##
## Ljung-Box test
##
## data: Residuals from ETS(A,A,A)
## Q* = 5.355, df = 8, p-value = 0.719
##
## Model df: 0. Total lags used: 8
```

. , . . .

```
fc_tour <- forecast(fit_tour, h = length(test_tour))
fc_lntour <- forecast(fit_lntour, h = length(test_tour))</pre>
```

```
accuracy(fc_tour, test_tour)
##
                                          MAE
                                                     MPE
                                                              MAPE
                         ME
                                RMSE
                                                                        MASE
## Training set 0.004642325 1.966538 1.474615 -0.4054666 4.196005 0.5206241
## Test set
                1.541295202 2.989673 2.414226 2.6022154 3.958845 0.8523613
##
                      ACF1 Theil's U
## Training set -0.0243205
## Test set
                 0.5001355 0.2077632
accuracy(fc_lntour, test_tour)
##
                       ME
                              RMSE
                                        MAE
                                                   MPE
                                                            MAPE
                                                                      MASE
```

```
## Training set 0.1640202 2.014686 1.562901 0.04389638 4.399828 0.5517943
  ## Test set 0.6115065 2.128486 1.722829 1.03590923 2.853023 0.6082580
                        ACF1 Theil's U
  ##
  ## Training set 0.01619145
  ## Test set 0.47705974 0.1348746
 ETS(A,A,A) fit_lntour test data
                                                         Figure 13
autolayer() PI = FALSE
  library(patchwork)
  p1 <- autoplot(train_tour) +</pre>
    autolayer(test_tour, series = "Test data") +
    autolayer(fc_tour, PI = FALSE, series = "ETS(M,A,M)") +
    autolayer(fc_lntour, PI = FALSE, series = "ETS(A,A,A)") +
    labs(y = NULL, x = NULL, color = NULL) +
    theme(legend.position = "top")
  p2 <- autoplot(test_tour, series = "Test data", size = .8) +</pre>
    autolayer(fc_tour, PI = FALSE, series = "ETS(M,A,M)",
              size = .8) +
    autolayer(fc_lntour, PI = FALSE, series = "ETS(A,A,A)",
              size = .8) +
    labs(y = NULL, x = NULL, color = NULL) +
    theme(legend.position = "top")
  p1 + p2
```

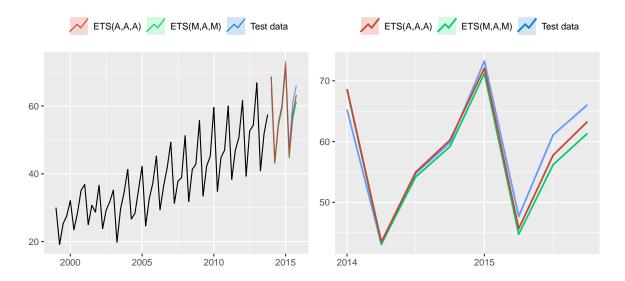


Figure 13: austourists

```
ETS(A,A,A) fit_lntour test data
```

Figure 14

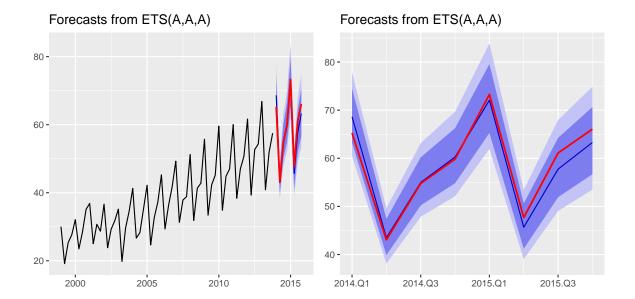


Figure 14: austourists

```
• 3: 1965 1 1992 7 (fma::dole)

fma::dole 1965 1 1992 7 . 2 test data .

train_d <- window(dole, end = c(1990, 7))
test_d <- window(dole, start = c(1990, 8))

Figure 15 . Test data . 1990 , test data
.

autoplot(train_d) +
autolayer(test_d, show.legend=FALSE, size = .8) +
```

labs(y = NULL, x = NULL)

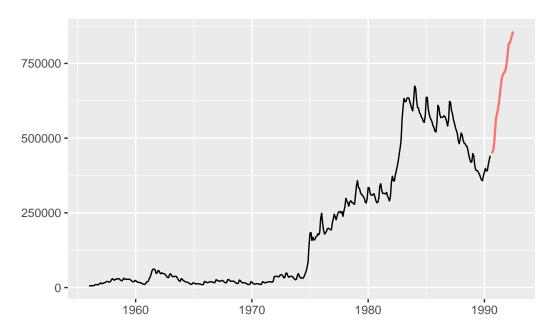


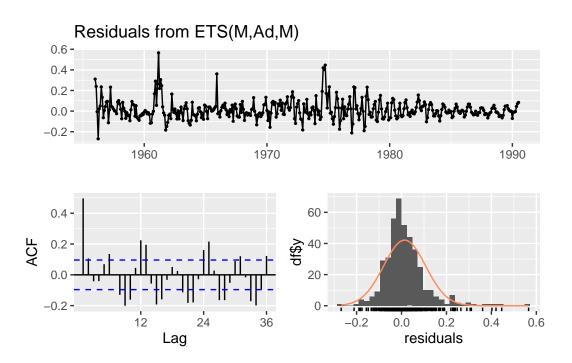
Figure 15: dole

ETS

```
fit_d <- ets(train_d)</pre>
fit_d
## ETS(M, Ad, M)
##
## Call:
    ets(y = train_d)
##
##
##
     Smoothing parameters:
       alpha = 0.7057
##
       beta = 0.1262
##
       gamma = 0.2942
##
##
       phi = 0.8701
##
     Initial states:
##
##
       l = 2693.6084
       b = 838.4198
##
       s = 1.0776\ 0.9108\ 0.9286\ 0.9993\ 1.0254\ 1.0275
##
               1.0028 0.9466 1.0225 0.9982 1.0038 1.0568
##
```

```
##
## sigma: 0.0965
##
## AIC AICc BIC
## 9930.864 9932.591 10003.373
, , (point forecast) , . .
```

checkresiduals(fit_d)



```
##
## Ljung-Box test
##
## data: Residuals from ETS(M,Ad,M)
## Q* = 270.13, df = 24, p-value < 2.2e-16
##
## Model df: 0. Total lags used: 24</pre>
```

22

```
fc_d <- forecast(fit_d, h = length(test_d))</pre>
accuracy(fc_d, test_d)
                                                       MPE
##
                                RMSE
                                                                MAPE
                                                                          MASE
                        ME
                                            MAE
                  307.438 16094.96
## Training set
                                       9474.828 0.5940649 6.112239 0.2965093
## Test set
              208048.806 234353.38 208048.806 28.7917875 28.791788 6.5107678
                     ACF1 Theil's U
##
## Training set 0.5103798
## Test set
               0.8895083 8.715368
```

MASE . Figure 16

```
autoplot(fc_d) +
  autolayer(test_d, color = "red", size = .8) +
  labs(y = NULL, x = NULL)
```

Forecasts from ETS(M,Ad,M)

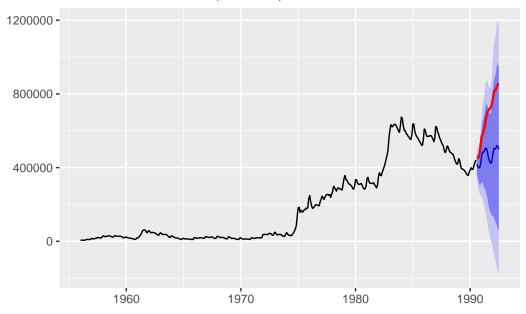


Figure 16: dole

Test data test data . , training data . test data . 1 test data . subset() . end start .

```
train_d_1 <- subset(dole, end = length(dole) - 12)</pre>
test_d_1 <- subset(dole, start = length(dole) - 11)</pre>
training data
                      1
fc_d_1 <- train_d_1 %>%
  ets() %>%
  forecast(h = length(test_d_1))
accuracy(fc_d_1, test_d_1)
                                                    MPE
                                                                                 ACF1
                               RMSE
                                          MAE
                                                            MAPE
                                                                       MASE
## Training set 419.4624 18184.20 10856.63 0.5111136 6.268750 0.3014767 0.5191139
## Test set
                5143.1303 46330.29 40948.53 1.0361996 5.319935 1.1370959 0.8678915
##
                 Theil's \it U
## Training set
                        NA
## Test set
                 2.792915
     Figure 17)
library(patchwork)
p1 <- autoplot(fc_d_1) +
  autolayer(test_d_1, color = "red", size = .8) +
  labs(y = NULL, x = NULL)
p2 <- autoplot(fc_d_1, include = 0) +</pre>
  autolayer(test_d_1, color = "red", size = .8) +
  labs(y = NULL, x = NULL)
p1 + p2
```

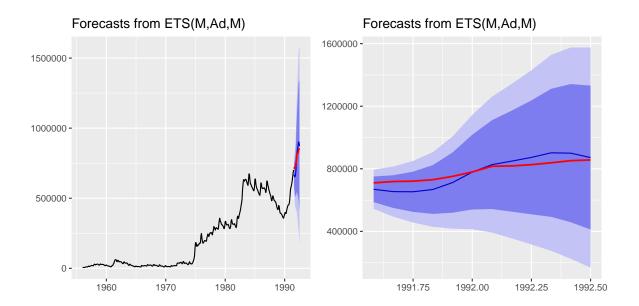


Figure 17: dole

• 4: 2014 4 30 1 Hyndsight (fpp2::hyndsight)

hyndsight 2014 4 30 1 Hyndman . Figure 18

autoplot(hyndsight) + labs(x = NULL, y = NULL)

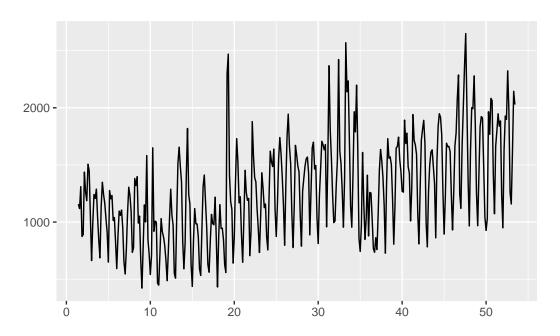


Figure 18: hyndsight

. 2014 4 30

m = 7

```
lubridate::wday(as.Date("2014-4-30"), label = TRUE)
  ## [1]
  ## Levels:
  start(hyndsight); end(hyndsight)
  ## [1] 1 4
  ## [1] 53 4
            , 14() 53 4() . Training data 14 48 4 , test data
2014 4 30
48 5 53 4 5
     window()
                                       subset()
                                                           35
                                                              test data
  train_hyn <- subset(hyndsight, end = length(hyndsight)-35)</pre>
  test_hyn <- subset(hyndsight, start = length(hyndsight)-34)</pre>
ETS \qquad \quad . \qquad ,
```

```
fit_hyn <- ets(train_hyn)</pre>
fit_hyn
## ETS(A, N, A)
##
## Call:
## ets(y = train_hyn)
##
##
    Smoothing parameters:
##
      alpha = 0.4426
      gamma = 1e-04
##
##
    Initial states:
##
      l = 1173.6676
##
      s = 296.9907 - 34.3415 - 457.6839 - 271.7606 63.9589 172.9509
##
              229.8854
##
   sigma: 232.0085
##
##
       AIC AICc
##
                       BIC
## 5519.446 5520.136 5557.437
fc_hyn <- forecast(fit_hyn, h = length(test_hyn))</pre>
accuracy(fc_hyn, test_hyn)
##
                       ME
                              RMSE
                                        MAE
                                                  MPE
                                                          MAPE
                                                                    MASE
## Training set 3.727848 228.8229 164.1102 -2.026679 13.86475 0.7404020
## Test set -73.167844 231.3494 180.9392 -8.044075 13.09378 0.8163279
                    ACF1 Theil's U
## Training set 0.1874059
## Test set
              0.3712518 0.5878683
                                    . 80% 95%
                                                                  , 80%
 Figure 19
               . test data ,
   , 95%
autoplot(fc_hyn, include = 0) +
 autolayer(test_hyn, color = "red", size = .8) +
 labs(x = NULL, y = NULL)
```

Forecasts from ETS(A,N,A)

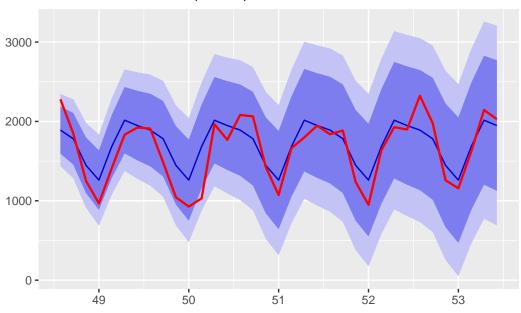


Figure 19: hynsight

- 1. ETS , . .
- fma::chicken
- fma::ibmclose
- 2. Seoul_temp.xlsx Suwon_temp.xlsx , https://raw.githubusercontent.com/yjyjp . ETS . Test data 2 .
- 3. Baek_pm10.xlsx Seoul_pm10.xlsx, Suwon_pm10.xlsx , https://raw.githubusercontent.com/yjyjpark/TS-with-R/main/Data/ . Test data 2 . .