# BF HW 4

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2022-09-28

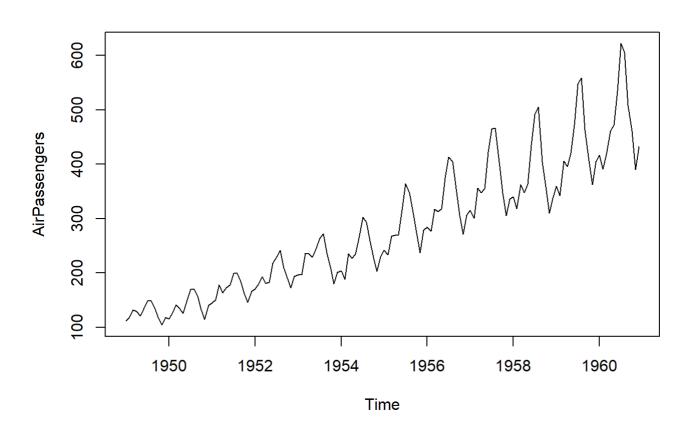
```
library(fpp)
## Loading required package: forecast
## Registered S3 method overwritten by 'quantmod':
     method
##
                       from
##
     as.zoo.data.frame zoo
## Loading required package: fma
## Loading required package: expsmooth
## Loading required package: lmtest
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
## Loading required package: tseries
library(fpp2)
## — Attaching packages —
                                                                     ---- fpp2 2.4 ---
## √ ggplot2 3.3.6
##
## Attaching package: 'fpp2'
```

```
## The following objects are masked from 'package:fpp':
##
## ausair, ausbeer, austa, austourists, debitcards, departures,
## elecequip, euretail, guinearice, oil, sunspotarea, usmelec
```

```
library(TTR)
attributes(AirPassengers)
```

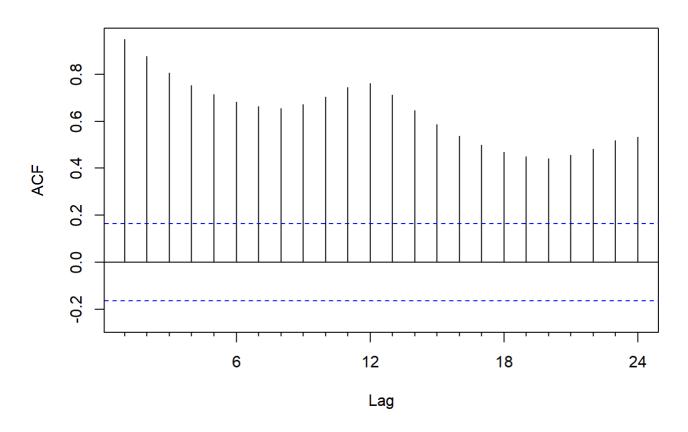
```
## $tsp
## [1] 1949.000 1960.917 12.000
##
## $class
## [1] "ts"
```

```
plot(AirPassengers)
```



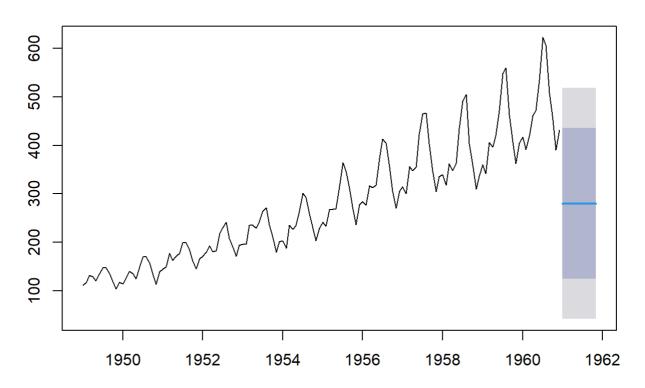
Acf(AirPassengers)

## Series AirPassengers



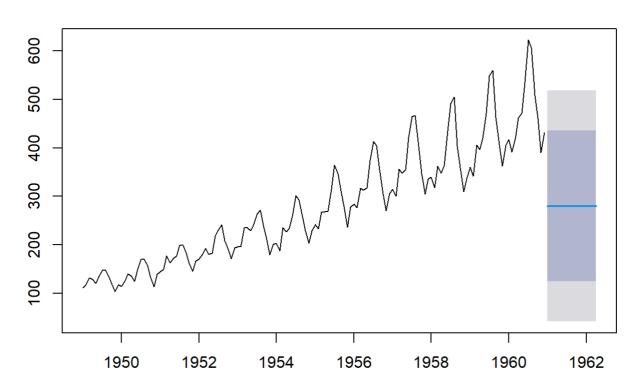
#Mean
mean\_forecast <- meanf(AirPassengers,11)
plot(mean\_forecast)</pre>

## **Forecasts from Mean**



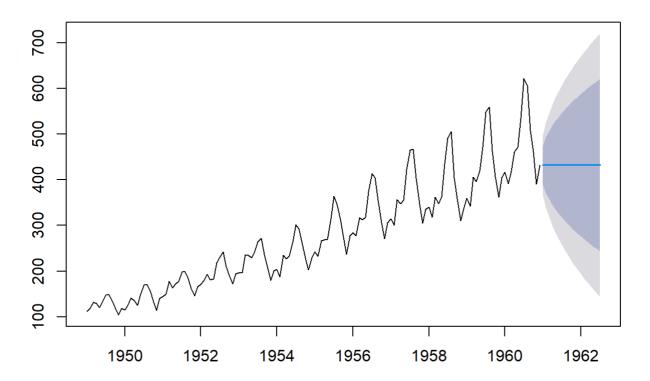
mean\_forecast <- meanf(AirPassengers,16)
plot(mean\_forecast)</pre>

#### **Forecasts from Mean**



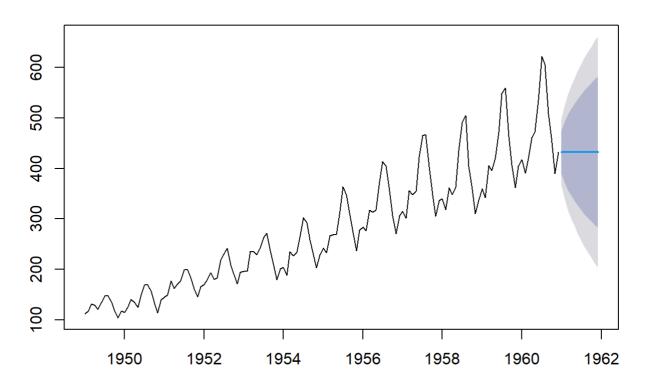
# Naive
naive\_forecast <- naive(AirPassengers,19)
plot(naive\_forecast)</pre>

#### **Forecasts from Naive method**



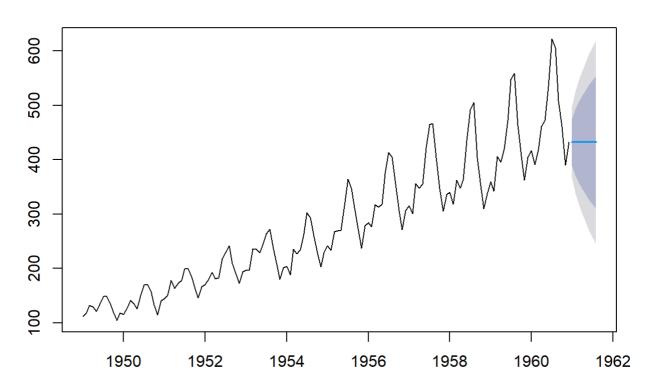
naive\_forecast <- naive(AirPassengers,12)
plot(naive\_forecast)</pre>

## Forecasts from Naive method



naive\_forecast <- naive(AirPassengers,8)
plot(naive\_forecast)</pre>

#### **Forecasts from Naive method**



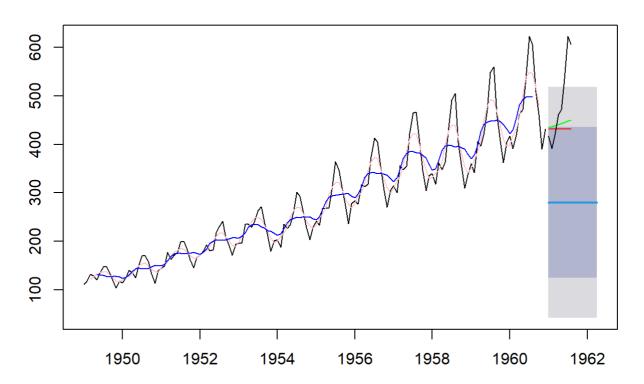
```
# Random Walk
rwf_forecast <- rwf(AirPassengers,8)
rwf_forecast <- rwf(AirPassengers,8, drift=TRUE)
# Seasonal Naive
snaive_forecast <- snaive(AirPassengers,8)

# Moving Averages

MA5_forecast <- ma(AirPassengers,order=5)
MA9_forecast <- ma(AirPassengers,order=9)

#Plots
plot(mean_forecast)
lines(naive_forecast$mean,col="red")
lines(rwf_forecast$mean,col="green")
lines(snaive_forecast$mean,col="black")
lines(MA5_forecast,col="Pink")
lines(MA9_forecast,col="Blue")</pre>
```

#### **Forecasts from Mean**

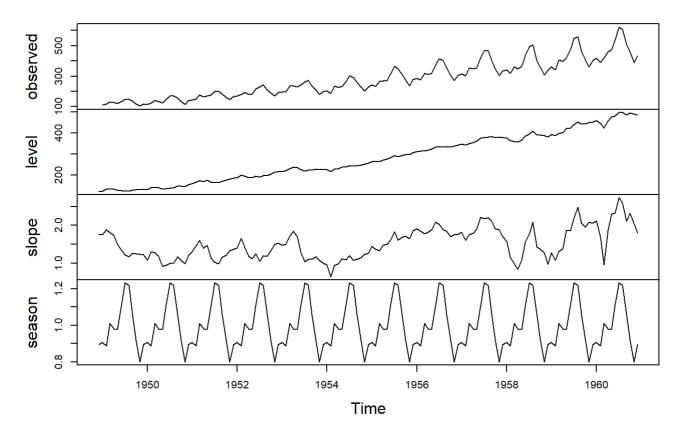


```
# Attributes
attributes(naive_forecast)
```

```
## $names
   [1] "method"
                                               "x"
                                                            "fitted"
                     "model"
                                  "lambda"
                                                                        "residuals"
##
   [7] "series"
                     "mean"
                                  "level"
                                               "lower"
                                                            "upper"
##
## $class
## [1] "forecast"
```

```
# Decomposition
ets_forecast <- ets(AirPassengers)
plot(ets_forecast)</pre>
```

#### Decomposition by ETS(M,Ad,M) method



```
attributes(ets_forecast)
```

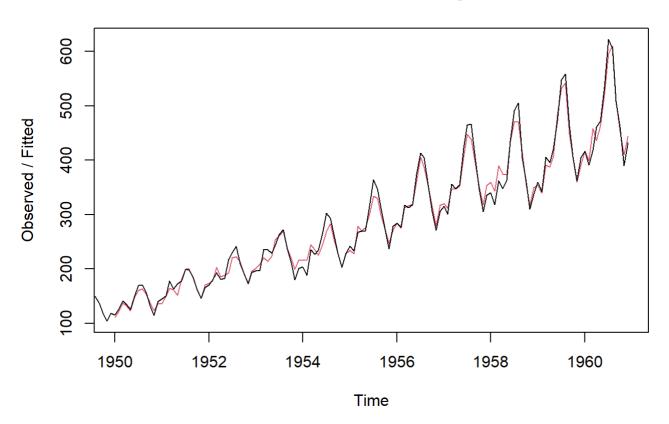
```
## $names
    [1] "loglik"
                      "aic"
                                    "bic"
                                                  "aicc"
                                                                "mse"
   [6] "amse"
                      "fit"
                                    "residuals"
                                                  "fitted"
                                                                "states"
## [11] "par"
                                    "method"
                                                  "series"
                                                                "components"
## [16] "call"
                      "initstate"
                                                  "x"
                                    "sigma2"
##
## $class
## [1] "ets"
```

```
ets_forecast$mse
```

```
## [1] 115.5035
```

```
# HoltWinters
HW_forecast <- HoltWinters(AirPassengers)
plot(HW_forecast)</pre>
```

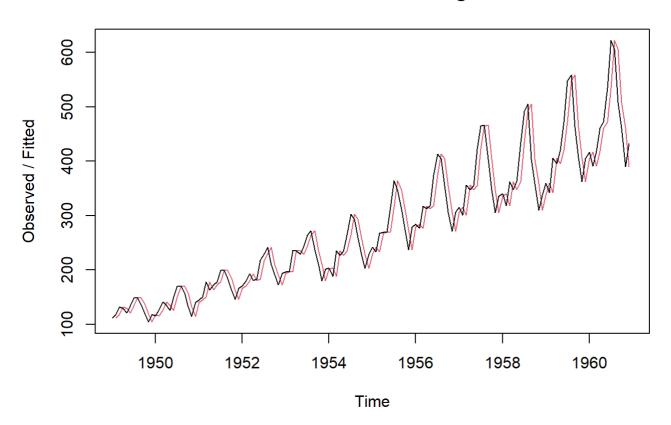
## **Holt-Winters filtering**



```
SSE_Simple <- HoltWinters(AirPassengers, beta=FALSE, gamma=FALSE)
attributes(SSE_Simple)</pre>
```

```
plot(SSE_Simple)
```

#### **Holt-Winters filtering**



```
SSE_Simple$SSE
```

## [1] 162510.6

#### head(SSE\_Simple\$fitted)

```
## Feb 1949 112.0000 112.0000

## Mar 1949 117.9996 117.9996

## Apr 1949 131.9991 131.9991

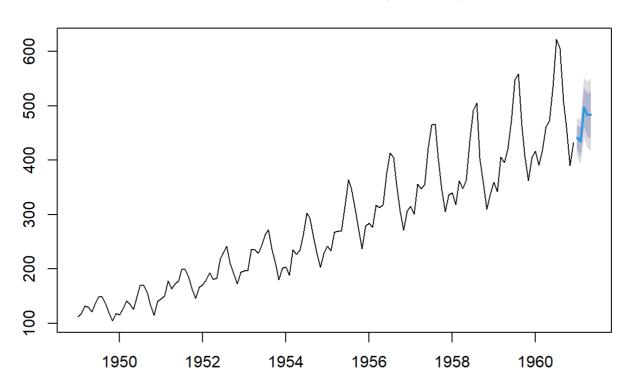
## May 1949 129.0002 129.0002

## Jun 1949 121.0005 121.0005

## Jul 1949 134.9991 134.9991
```

```
#Forecast
forecast_ets_1 <- forecast.ets(ets_forecast, h=5)
plot(forecast_ets_1)
forecast_ets_2 <- forecast(ets_forecast, h=5)
plot(forecast_ets_2)</pre>
```

## Forecasts from ETS(M,Ad,M)



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