# Temporal hierarchies

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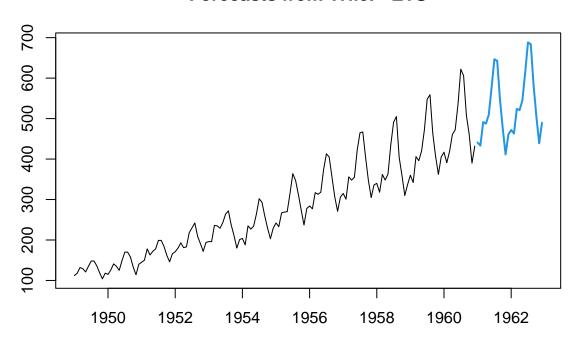
## 1. Data and packages

```
pckg <- c("thief","MAPA","tsutils","abind")</pre>
for (i in 1:length(pckg)){
if(!(pckg[i] %in% rownames(installed.packages()))){
install.packages(pckg[i])
library(pckg[i],character.only = TRUE)
## Warning: package 'thief' was built under R version 4.2.3
## Loading required package: forecast
## Registered S3 method overwritten by 'quantmod':
     method
     as.zoo.data.frame zoo
## Loading required package: parallel
## Loading required package: RColorBrewer
## Loading required package: smooth
## Loading required package: greybox
## Package "greybox", v1.0.5 loaded.
## This is package "smooth", v3.1.6
## Warning: package 'tsutils' was built under R version 4.2.3
y <- AirPassengers
```

#### 2. Temporal hierarchies using the thief package

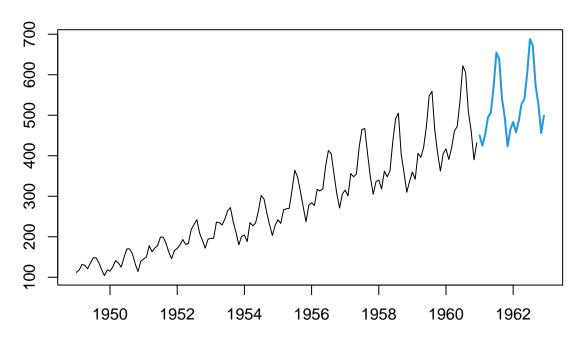
```
frc1 <- thief(y)
plot(frc1)</pre>
```

# Forecasts from THieF-ETS



```
frc2 <- thief(y,usemodel="arima")
plot(frc2)</pre>
```

## Forecasts from THieF-ARIMA



### 3. Manual implementation of THieF

```
S <- tsutils::Sthief(y) # Get the S matrix
ff <- frequency(y) # Get sampling frequency of target series</pre>
AL <- ff/(1:ff) # Calculate frequencies of various aggregation levels
AL <- AL[AL %% 1 == 0] # And exclude those that would not be integer
k <- length(AL) # Find how many are left
Y <- MAPA::tsaggr(y,AL)[[1]]
hrz <- 16 # Target horizon</pre>
hAggr <- (ceiling(hrz/ff)*ff)/AL
hAggr
## [1] 2 4 6 8 12 24
frc <- mse <- list()</pre>
for (i in 1:k){
yTemp <- Y[[i]]
fit <- ets(yTemp)</pre>
mse[[i]] <- fit$mse</pre>
frcTemp <- forecast(fit,h=hAggr[i])$mean</pre>
```

```
# Re-structure forecasts
frc[[i]] <- matrix(frcTemp,ncol=hAggr[1]) # Organised as column per year</pre>
frcAll <- abind(frc,along=1)</pre>
frcAll
##
              [,1]
                        [,2]
## [1,] 6117.6361 6596.1292
## [2,] 3007.6633 3250.5458
## [3,] 3129.1045 3371.9871
## [4,] 1833.0574 1941.5106
## [5,] 2393.3701 2532.2736
## [6,] 1898.4091 2006.5253
## [7,] 1355.3595 1459.7158
## [8,] 1593.3745 1713.7677
## [9,] 1887.3399 2027.3332
## [10,] 1383.1478 1483.8981
## [11,] 867.1471 929.9355
## [12,] 963.5666 1032.5139
## [13,] 1094.7423 1172.1632
## [14,] 1332.7008 1425.8650
## [15,] 1037.7271 1109.4450
## [16,] 881.9057 942.1689
## [17,] 441.8018 459.0139
## [18,] 434.1186 450.6333
## [19,] 496.6300 515.0797
## [20,] 483.2375 500.7700
## [21,] 483.9914 501.1423
## [22,] 551.0244 570.0974
## [23,] 613.1797 633.9130
## [24,] 609.3648 629.4938
## [25,] 530.5408 547.6630
## [26,] 463.0332 477.6340
## [27,] 402.7478 415.1573
## [28,] 451.9694 465.5780
# Structural:
W <- diag(1/rowSums(S))
Gstr <- solve(t(S)%*%W%*%S)%*%t(S)%*%W
# Variance:
mse <- unlist(mse)</pre>
W <- diag(1/mse[rep((1:k),rev(AL))])</pre>
Gvar <- solve(t(S)%*%W%*%S)%*%t(S)%*%W
# Create the bottom level forecasts
frcBRec <- Gstr %*% frcAll</pre>
frcFinal <- as.numeric(frcBRec)[1:hrz]</pre>
# We can also translate this into a time series object
frcFinal <- ts(frcFinal, frequency=frequency(y), start=end(y)[1] + deltat(y)*end(y)[2])</pre>
frcFinal
                                                                    Jul
##
             Jan
                      Feb
```

May

Jun

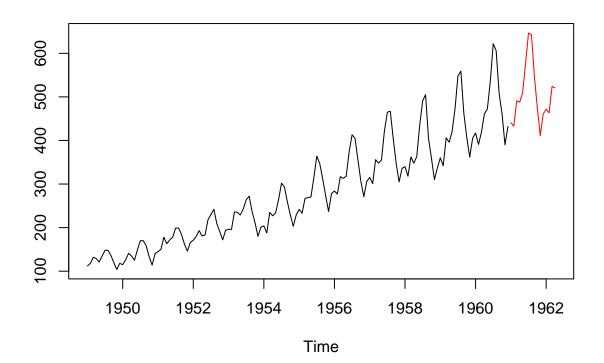
Aug

Apr

Mar

```
## 1961 440.8650 433.1818 491.3771 487.7213 509.7371 576.7701 646.6260 642.8111
## 1962 471.6625 463.2819 524.0035 520.9690
## Sep Oct Nov Dec
## 1961 547.0800 474.7431 411.4239 460.6455
## 1962

ts.plot(y,frcFinal,col=c("black","red"))
```



#### frcARec <- S %\*% frcBRec</pre>

### frcARec

```
##
              [,1]
                        [,2]
    [1,] 6122.9819 6526.9923
##
   [2,] 2939.6523 3138.1222
   [3,] 3183.3296 3388.8700
   [4,] 1853.1451 1979.9169
##
   [5,] 2375.9443 2530.3770
##
  [6,] 1893.8925 2016.6984
## [7,] 1365.4238 1458.9479
## [8,] 1574.2285 1679.1744
## [9,] 1836.5170 1954.9911
## [10,] 1346.8125 1433.8790
## [11,] 874.0467 934.9444
## [12,] 979.0984 1044.9725
```

```
## [13,] 1086.5072 1158.2054
## [14,] 1289.4371 1372.1717
## [15,] 1021.8231 1088.1689
## [16,] 872.0694 928.5295
## [17,] 440.8650 471.6625
## [18,] 433.1818 463.2819
## [19,]
         491.3771 524.0035
## [20,]
         487.7213 520.9690
## [21,] 509.7371
                   544.6251
## [22,]
         576.7701
                   613.5803
## [23,]
         646.6260
                   688.2954
## [24,]
         642.8111
                   683.8763
## [25,]
         547.0800 582.8194
## [26,]
         474.7431 505.3495
## [27,]
         411.4239
                   439.0544
## [28,]
         460.6455 489.4751
```