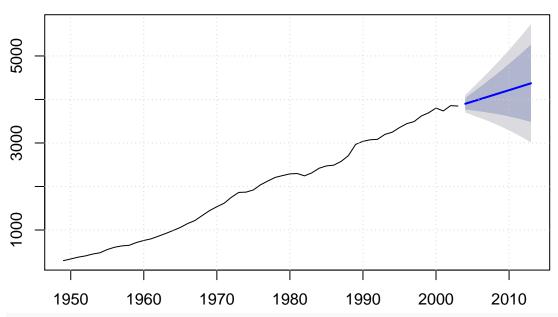
```
# ets1.r
# datasets from expsmooth
# install.packages("expsmooth")
library(expsmooth)
## Loading required package: forecast
## Registered S3 method overwritten by 'xts':
    method
               from
##
     as.zoo.xts zoo
## Registered S3 method overwritten by 'quantmod':
                       from
##
     as.zoo.data.frame zoo
## Registered S3 methods overwritten by 'forecast':
                        from
##
    fitted.fracdiff
                       fracdiff
##
    residuals.fracdiff fracdiff
data(package='expsmooth')
# Data sets in package 'expsmooth':
# ausgdp
                             Quarterly Australian GDP
# bonds
                             Monthly US government bond yields
# canqas
                            Monthly Canadian gas production
# carparts
                            Monthly sales car parts
# dji
                            Monthly Dow Jones Index
# djiclose
                            Monthly Dow Jones Index: closing
# enplanements
                            Monthly US domestic enplanements
# fmsales
                            Weekly FM sales
# freight
                            Annual US new freight cars
# frexport
                            Quarterly French exports
# qasprice
                            US gasoline prices
# hospital
                            Monthly patient count
# jewelry
                             Weekly jewelry sales
# mcopper
                             Monthly copper prices
# msales
                             Monthly product sales
# partx
                             Monthly sales of an automobile part
# ukcars
                             Quarterly UK passenger car production
# unemp.cci
                            Unemployment and the CCI
                            Quarterly US GDP
# usqdp
# usnetelec
                             Annual US net electricity generation
# utility
                            Hourly utility demand
# vehicles
                            Hourly vehicle counts
# visitors
                            Monthly Australian overseas vistors
# xrates
                             Monthly exchange rates
library(forecast)
data(usnetelec)
data(bonds)
data(ukcars)
data(visitors)
```

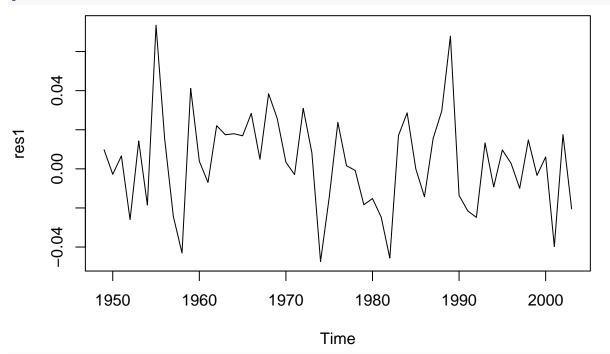
```
# models
etsfit1 = ets(usnetelec)
etsfit1
## ETS(M,A,N)
##
## Call:
## ets(y = usnetelec)
##
##
    Smoothing parameters:
##
      alpha = 0.9999
##
      beta = 0.2191
##
##
   Initial states:
##
     1 = 254.9338
##
      b = 38.3125
##
##
   sigma: 0.0259
##
             AICc
##
       AIC
                       BIC
## 634.0437 635.2682 644.0803
etsfit2 = ets(bonds)
etsfit2
## ETS(A,Ad,N)
##
## Call:
## ets(y = bonds)
##
##
   Smoothing parameters:
##
     alpha = 0.9999
##
      beta = 0.0954
##
      phi = 0.8026
##
##
   Initial states:
##
     1 = 5.3252
##
      b = 0.5934
##
## sigma: 0.2428
##
##
       AIC
              AICc
## 256.5383 257.2502 273.5082
etsfit3 = ets(ukcars)
etsfit3
## ETS(A,N,A)
##
## Call:
## ets(y = ukcars)
##
## Smoothing parameters:
##
      alpha = 0.6199
```

```
##
       gamma = 1e-04
##
##
     Initial states:
       1 = 314.2568
##
##
       s = -1.7579 - 44.9601 21.1956 25.5223
##
     sigma: 25.9302
##
##
        AIC
                AICc
                          BIC
## 1277.752 1278.819 1296.844
etsfit4 = ets(visitors)
etsfit4
## ETS(M,A,M)
##
## Call:
    ets(y = visitors)
##
##
     Smoothing parameters:
##
       alpha = 0.6146
##
       beta = 2e-04
##
       gamma = 0.192
##
##
     Initial states:
##
      1 = 92.9631
##
       b = 2.2221
       s = 0.9378 \ 1.0666 \ 1.0669 \ 0.9625 \ 1.3768 \ 1.113
##
              1.0012 0.8219 0.9317 1.0046 0.8755 0.8413
##
##
##
     sigma: 0.0536
##
##
        AIC
                AICc
                          BIC
## 2603.654 2606.411 2662.825
#
forecast(etsfit1)
        Point Forecast
                          Lo 80
                                   Hi 80
                                             Lo 95
              3900.329 3770.801 4029.857 3702.233 4098.425
## 2004
## 2005
              3952.650 3747.279 4158.022 3638.562 4266.738
## 2006
              4004.972 3725.589 4284.355 3577.692 4432.251
## 2007
              4057.293 3701.885 4412.701 3513.743 4600.842
## 2008
              4109.614 3674.968 4544.259 3444.881 4774.347
## 2009
              4161.935 3644.367 4679.503 3370.383 4953.487
## 2010
              4214.256 3609.881 4818.632 3289.944 5138.569
## 2011
              4266.577 3571.428 4961.726 3203.439 5329.716
## 2012
              4318.898 3528.985 5108.812 3110.830 5526.967
## 2013
              4371.220 3482.552 5259.888 3012.119 5730.320
plot(forecast(etsfit1),main='US 10-year bond yield')
grid()
```

US 10-year bond yield

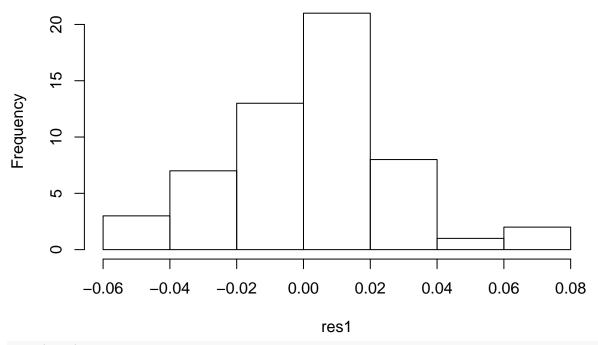


residuals
res1 = residuals(etsfit1)
plot(res1,main='')



hist(res1)

Histogram of res1

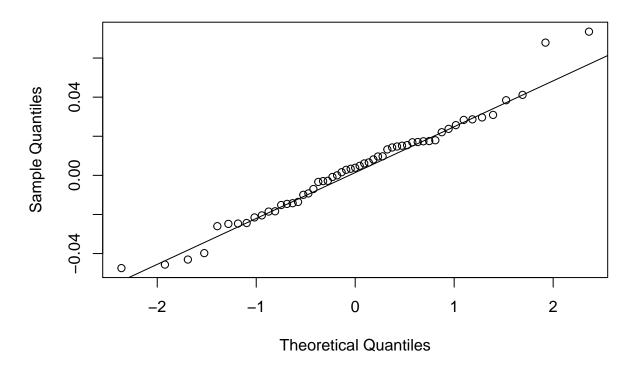


mean(res1)

[1] 0.003243211

qqnorm(res1)
qqline(res1)

Normal Q-Q Plot



class(res1)

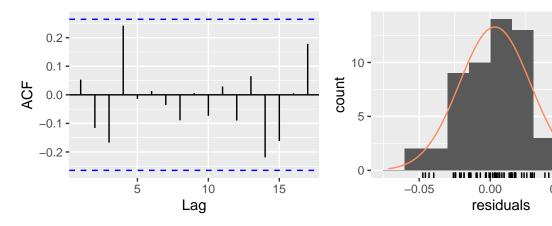
[1] "ts"

checkresiduals(etsfit1)

Residuals from ETS(M,A,N)



0.05



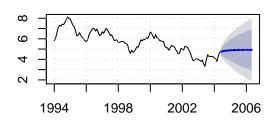
```
##
## Ljung-Box test
##
## data: Residuals from ETS(M,A,N)
## Q* = 7.2548, df = 6, p-value = 0.2979
##
## Model df: 4. Total lags used: 10
```

```
par(mfrow=c(2,2))
plot(forecast(etsfit1),main=''); grid(); title('US 10-year bond yield')
plot(forecast(etsfit2),main=''); grid(); title('US net electricity generation')
plot(forecast(etsfit3),main=''); grid(); title('UK passenger motor vehicle production')
plot(forecast(etsfit4),main=''); grid(); title('Visitors to Australia')
```

US 10-year bond yield

1950 1970 1990 2010

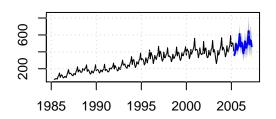
US net electricity generation



UK passenger motor vehicle productio

1980 1990 2000

Visitors to Australia



```
# validation
str(usnetelec)
```

```
## Time-Series [1:55] from 1949 to 2003: 296 334 375 404 447 ...

length(usnetelec)
```

```
## [1] 55
```

```
# train performance
model1 = ets(usnetelec[1:45])
accuracy(model1)
```

```
## ME RMSE MAE MPE MAPE MASE ACF1
## Training set 3.952422 48.8502 33.89778 0.3530148 2.051756 0.4957958 0.225129
# test performance
test = ets(usnetelec[46:55], model = model1)
```

Model is being refit with current smoothing parameters but initial states are being re-estimated. ## Set 'use.initial.values=TRUE' if you want to re-use existing initial values.

```
accuracy(test)
```

```
## Training set -13.0306 64.02833 45.85866 -0.3446059 1.234965 0.5484756 ## Training set -0.5879419
```