



Tidy Forecasting in R





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- 1 Why change?
- 2 Example: Australian eating-out expenditure
- 3 Example: Australian prison population
- 4 Example: Half-hourly electricity demand
- 5 Equivalent methods
- 6 More information

forecast package

Pre 2003	Private functions used for consulting project	ts
July/August 2003	ets and thetaf added	
August 2006	v1.0 available on CRAN	
May 2007	auto.arima added	
May 2010	arfima added	
Feb/March 2011	tslm, stlf, naive, snaive added	
August 2011	v3.0. Box Cox transformations added	
December 2011	tbats added	
April 2012	Package moved to github	
November 2012	v4.0. nnetar added	
June 2013	Major speed-up of ets	
February 2016	v7.0. Added ggplot2 graphics	
February 2017	v8.0. Added checkresiduals, tsCV and %>%	
April 2018	v8.3. Added mstl	
June 2018	pprox 100,000 package downloads per month	3

fable package

A replacement for the forecast package.

Why change?

- Interacting with tidyverse packages
- Sub-daily data and multiple seasonal data handled more easily
- Consistency of interface
- Distribution forecasting rather than point+interval
- Flexible transformations
- Extensibility
- Simpler interface for forecast reconciliation
- Boosting, combining and ensemble forecasts
- Designed for forecasting many related time series
- Changes will break too much existing code
- Opportunity to re-think forecasting practice

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fpp2::auscafe

Jan

Feb

##

```
0.342 0.342 0.329 0.339 0.332 0.34
## 1982
  1983 0.369 0.348 0.366 0.351 0.360 0.347 0.364 0.376 0.37
  1984 0.389 0.377 0.398 0.383 0.414 0.382 0.393 0.409 0.39
  1985 0.426 0.392 0.416 0.420 0.446 0.407 0.449 0.466 0.45
  1986 0.504 0.453 0.480 0.497 0.531 0.485 0.526 0.538 0.53
  1987 0.572 0.525 0.544 0.558 0.565 0.542 0.599 0.584 0.59
  1988 0.605 0.586 0.625 0.612 0.630 0.635 0.659 0.656 0.66
  1989 0.733 0.661 0.713 0.694 0.710 0.722 0.741 0.746 0.76
  1990 0.858 0.764 0.840 0.805 0.809 0.799 0.815 0.828 0.81
  1991 0.862 0.771 0.813 0.797 0.821 0.801 0.829 0.854 0.88
  1992 0.938 0.862 0.936 0.932 0.929 0.869 0.891 0.875 0.91
  1993 0.918 0.838 0.870 0.862 0.852 0.828 0.882 0.867 0.90
  1994 0.985 0.902 1.015 0.939 0.941 0.935 1.013 1.018 1.04
  1995 1.076 0.982 1.099 1.068 1.083 1.045 1.094 1.110 1.12
  1996 1.213 1.128 1.180 1.169 1.146 1.109 1.138 1.146 1.10
```

1997 1.180 1.060 1.148 1.141 1.170 1.113 1.165 1.173 1.45 ## 1998 1.186 1.050 1.141 1.107 1.144 1.088 1.162 1.145 1.14

Mar Apr May Jun Jul

Aug Se

```
library(tsibble)
cafe <- as_tsibble(fpp2::auscafe)
cafe</pre>
```

```
## # A tsibble: 426 x 2 [1MONTH]
##
         index value
##
         <mth> <dbl>
##
   1 1982 Apr 0.342
##
   2 1982 May 0.342
   3 1982 Jun 0.329
##
   4 1982 Jul 0.338
##
##
   5 1982 Aug 0.332
##
   6 1982 Sep 0.342
##
   7 1982 Oct 0.358
   8 1982 Nov 0.375
##
##
   9 1982 Dec 0.433
  10 1983 Jan 0.369
## # ... with 416 more rows
```

```
library(fable)
cafe %>% ETS(value)
```

```
## # A tibble: 1 x 2
## data model
## <list> <list>
## 1 <tsibble [426 x 2]> <ETS(M,A,M)>
```

cafe %>% ETS(value) %>% summary()

```
ETS(M,A,M)
##
## Call:
    ETS(data = ., formula = value)
##
##
##
     Smoothing parameters:
       alpha = 0.6263
##
##
       beta = 0.0065
##
       gamma = 0.0755
##
    Initial states:
##
##
    l = 0.3477
##
    b = 0.0038
##
       s = 0.996 \ 0.936 \ 1.01 \ 1.15 \ 1.01 \ 1.01
              0.983 0.991 0.992 0.951 0.997 0.971
##
     sigma: 0.0249
##
##
    AIC AICC BIC
## -319 -318 -250
```

```
cafe %>% ETS(value) %>% forecast()
```

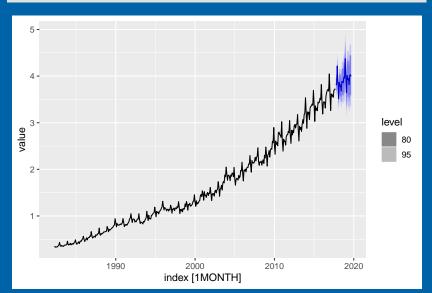
```
cafe %>% ETS(value) %>% forecast() %>%
  summary()
```

```
A tsibble: 24 x 4 [1MONTH]
##
         index
               mean
                              80%
                                           95%
##
         <mth>
               <dbl>
                             <hilo>
                                          <hilo>
##
   1 2017 Oct
               3.83 [3.71, 3.96]80 [3.65,
                                           4.02]95
##
   2 2017 Nov 3.81
                    [3.67, 3.96]80 [3.59,
                                           4.03795
                                           4.49795
##
   3 2017 Dec
               4.22 [4.04, 4.40]80 [3.94,
##
   4 2018 Jan
               3.83
                    [3.64,
                            4.01]80 [3.55,
                                           4.10]95
##
    5 2018 Feb
               3.51 [3.33, 3.70]80
                                   [3.24, 3.79]95
                                           4.20795
##
    6 2018 Mar
               3.87
                    [3.65,
                            4.09]80
                                    [3.54,
##
   7 2018 Apr 3.78
                    [3.56,
                            4.01]80 [3.44, 4.13]95
    8 2018 May
               3.81 [3.57,
                            4.05]80 [3.44,
                                           4.18]95
##
                    [3.43, 3.92]80 [3.30,
##
     2018 Jun
               3.68
                                           4.05195
##
   10 2018 Jul
               3.88
                    [3.60, 4.15]80
                                    [3.46, 4.29]95
##
   # ... with 14 more rows
```

```
cafe %>% ETS(value) %>% forecast() %>%
  summary(level=90)
```

```
# A tsibble: 24 x 3 [1MONTH]
##
        index mean
                             90%
##
        <mth> <dbl> <hilo>
## 1 2017 Oct 3.83 [3.68, 3.99]90
   2 2017 Nov 3.81 [3.63, 4.00]90
##
   3 2017 Dec 4.22 [3.99, 4.45]90
##
##
   4 2018 Jan 3.83 [3.59, 4.06]90
   5 2018 Feb 3.51 [3.28, 3.75]90
##
   6 2018 Mar 3.87 [3.59, 4.15]90
##
##
   7 2018 Apr 3.78 [3.49, 4.07]90
   8 2018 May 3.81 [3.50, 4.12]90
##
##
   9 2018 Jun 3.68 [3.36, 3.99]90
  10 2018 Jul 3.88 [3.53, 4.22]90
  # ... with 14 more rows
```





```
cafe %>% ARIMA(log(value)) %>%
forecast() %>% summary()
```

```
# A tsibble: 24 x 4 [1MONTH]
        index
##
               mean 80%
                                          95%
##
      <mth> <dbl> <hilo>
                                        <hilo>
   1 2017 Oct 1.60 [0.557, 3.09]80 [0.354, 4.86]95
##
##
   2 2017 Nov 1.60 [0.557, 3.09]80 [0.354, 4.86]95
##
   3 2017 Dec
               <u>1.60 [0.5</u>57, 3.09]80 [0.354, 4.86]95
   4 2018 Jan
##
               1.60 [0.557, 3.09]80 [0.354, 4.86]95
##
   5 2018 Feb
               1.60 [0.557, 3.09]80 [0.354, 4.86]95
   6 2018 Mar
               1.60 [0.557, 3.09]80 [0.354, 4.86]95
##
##
   7 2018 Apr
               1.60 [0.557, 3.09]80 [0.354, 4.86]95
##
   8 2018 May
               1.60 [0.557, 3.09]80 [0.354, 4.86]95
##
   9 2018 Jun
               1.60 [0.557, 3.09]80 [0.354, 4.86]95
##
  10 2018 Jul
               1.60 [0.557, 3.09]80 [0.354, 4.86]95
## # ... with 14 more rows
```

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fpp2::prisonLF

```
# A tibble: 1,536 x 5
##
     state gender legal t
                                    count
     <fct> <fct> <fct>
                                   <fdb>>
##
                          <date>
##
   1 ACT Female Remanded 2005-03-01
   2 ACT Female Remanded 2005-06-01
##
   3 ACT Female Remanded 2005-09-01
##
##
   4 ACT Female Remanded 2005-12-01
                                        4
   5 ACT Female Remanded 2006-03-01
##
                                        4
##
   6 ACT
           Female Remanded 2006-06-01
##
   7 ACT
           Female Remanded 2006-09-01
                                        9
           Female Remanded 2006-12-01
##
   8 ACT
                                        6
   9 ACT
           Female Remanded 2007-03-01
                                        4
##
  10 ACT
           Female Remanded 2007-06-01
##
                                        4
## # ... with 1,526 more rows
```

```
prison <- fpp2::prisonLF %>%
  mutate(qtr=yearquarter(t)) %>%
  select(-t) %>%
  as_tsibble(index=qtr, key=id(state,gender,legal))
prison
```

```
## # A tsibble: 1,536 x 5 [1QUARTER]
## # Keys: state, gender, legal [32]
## state gender legal count qtr
## <fct> <fct> <fct> <dbl> <qtr>
##
   1 ACT Female Remanded 2 2005 Q1
   2 ACT Female Remanded 4 2005 Q2
##
##
   3 ACT Female Remanded 1 2005 Q3
   4 ACT Female Remanded 4 2005 Q4
##
   <u>5 ACT</u> Female Remanded
                           4 2006 01
##
   6 ACT Female Remanded
                            6 2006 02
##
##
   7 ACT Female Remanded
                            9 2006 03
```

prison %>% ETS(count)

```
## # A tibble: 32 x 5
     state gender legal
##
                            data
                                              model
     <fct> <fct> <fct>
                            st>
                                              st>
##
##
   1 ACT
           Female Remanded <tsibble [48 x 2]> <ETS(M,A,N)>
   2 ACT
           Female Sentenced <tsibble [48 x 2]> <ETS(A,A,N)>
   3 ACT
           Male
                  Remanded <tsibble [48 x 2]> <ETS(M,N,N)>
##
   4 ACT
           Male Sentenced <tsibble [48 x 2]> <ETS(A,N,N)>
##
   5 NSW
           Female Remanded <tsibble [48 x 2]> <ETS(M,N,M)>
   6 NSW
           Female Sentenced <tsibble [48 x 2]> <ETS(M,N,M)>
##
   7 NSW
                  Remanded <tsibble [48 \times 2] <ETS(M,A,A)>
##
           Male
   8 NSW
           Male Sentenced <tsibble [48 x 2]> <ETS(M,A,A)>
##
   9 NT
           Female Remanded <tsibble [48 x 2]> <ETS(M,N,N)>
## 10 NT
           Female Sentenced <tsibble [48 x 2]> <ETS(M,A,A)>
## # ... with 22 more rows
```

```
prison %>% ETS(count) %>% forecast()
## # A tibble: 32 x 6
     state gender legal
                           data
                                             model
                                                          forecast
     Female Remanded <tsibble [48 x 2]> <ETS(M,A,N)> <tsibble [8 x 3~
   1 ACT
           Female Sentenced <tsibble [48 x 2]> <ETS(A.A.N)> <tsibble [8 x 3~
   2 ACT
   3 ACT
           Male
                 Remanded <tsibble [48 x 2]> <ETS(M,N,N)> <tsibble [8 x 3~
   4 ACT
           Male Sentenced <tsibble [48 x 2]> <ETS(A,N,N)> <tsibble [8 x 3~
   5 NSW
           Female Remanded <tsibble [48 x 2]> <ETS(M.N.M)> <tsibble [8 x 3~
   6 NSW
           Female Sentenced <tsibble [48 x 2]> <ETS(M.N.M)> <tsibble [8 x 3~
                 Remanded <tsibble [48 x 2]> <ETS(M,A,A)> <tsibble [8 x 3~
   7 NSW
           Male
   8 NSW
           Male Sentenced <tsibble [48 x 2]> <ETS(M,A,A)> <tsibble [8 x 3~
           Female Remanded <tsibble [48 x 2]> <ETS(M,N,N)> <tsibble [8 x 3~
   9 NT
           Female Sentenced <tsibble [48 x 2]> <ETS(M,A,A)> <tsibble [8 x 3~
## 10 NT
```

Aggregation and reconciliation not yet implemented.

... with 22 more rows

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Example: Half-hourly electricity demand

elecdemand

```
## # A tsibble: 17,520 x 4 [30MINUTE]
      index
                           Demand Temperature WorkDay
##
      <dttm>
                            <dbl>
                                         <dbl>
                                                 <dbl>
##
##
    1 2014-01-01 00:00:00
                             3.91
                                          18.2
                                                     0
##
    2 2014-01-01 00:30:00
                             3.67
                                          17.9
                                                     0
##
    3 2014-01-01 01:00:00
                             3.50
                                          17.6
                                                     0
    4 2014-01-01 01:30:00
                             3.34
                                          16.8
##
                                                     0
##
    5 2014-01-01 02:00:00
                             3.20
                                          16.3
                                                     0
    6 2014-01-01 02:30:00
                             3.10
                                          16.6
##
                                                     0
##
    7 2014-01-01 03:00:00
                             3.04
                                          16.6
                                                     0
##
    8 2014-01-01 03:30:00
                             3.01
                                          16.7
                                                     0
##
    9 2014-01-01 04:00:00
                             3.02
                                          16.2
                                                     0
##
   10 2014-01-01 04:30:00
                             3.03
                                          16.6
                                                     0
## # ... with 17,510 more rows
```

Example: Half-hourly electricity demand

```
fit2 <- ARIMA(elecdemand,
  Demand ~ Temperature + I(Temperature^2) + WorkDay)
summary(fit2)</pre>
```

```
## Series: Demand
## Regression with ARIMA(0,0,0) errors
##
## Coefficients:
        intercept Temperature I(Temperature^2) WorkDay
##
##
            5.415 -0.195
                                         0.006
                                                 0.733
## s.e. 0.038 0.004
                                         0.000 0.011
##
  sigma^2 estimated as 0.484: log likelihood=-18497
  AIC=37004
             ATCc=37004 BTC=37043
##
  Training set error measures:
##
                    ME RMSE MAE MPE MAPE MASE ACF1
## Training set 1.32e-14 0.695 0.562 -2.36 12.8 2.59 0.971
#forecast(fit2) %>% autoplot(newdata=elecdemandfuture)
```

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Equivalent methods: forecast \longrightarrow **fable**

```
auto, arima \longrightarrow ARIMA
                \longrightarrow ETS
ets
tslm/lm \longrightarrow LM
thats
           \longrightarrow TBATS
nnetar \longrightarrow NNETAR
stlm
                \longrightarrow STL %>%
                      modelcomponents(
                         ETS(seasadj),SNAIVE(season))
```

- All functions have a formula interface with automatic modelling if no formula provided.
- All functions produce mable class objects.
- Some of these functions not yet implemented

Equivalent methods: forecast \longrightarrow **fable**

```
naive → NAIVE %>% forecast
snaive → SNAIVE %>% forecast
thetaf \longrightarrow THETA %>% forecast
stlf \longrightarrow STL %>%
             modelcomponents(
              ETS(seasadj),SNAIVE(season)) %>%
             forecast
    → HW %>% forecast
hw
holt \longrightarrow HOLT %>% forecast
ses \longrightarrow SES %>% forecast
splinef → SPLINE %>% forecast
croston → CROSTON %>% forecast
```

forecast produces fable class objects.

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devtools::install_github("tidyverts/tsibble")
devtools::install_github("tidyverts/fable")







Earo Wang



Mitchell O'Hara-Wild

Follow our progress

- tidyverts.org
- robjhyndman.com/hyndsight