



# Tidy Forecasting in R



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# forecast package

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



A replacement for the forecast package.

## Why change?

- Integrating with tidyverse packages
- Designed for forecasting many related time series
- Consistency of interface using formulas
- Distribution forecasting rather than point + interval
- Flexible transformations
- Sub-daily data and multiple seasonal data handled more easily
- Simpler interface for forecast reconciliation

# Example: Australian eating-out expenditure

fpp2: auscafe

##	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep
## 1982				0.342	0.342	0.329	0.339	0.332	0.342
## 1983	0.369	0.348	0.366	0.351	0.360	0.347	0.364	0.376	0.378
## 1984	0.389	0.377	0.398	0.383	0.414	0.382	0.393	0.409	0.395
## 1985	0.426	0.392	0.416	0.420	0.446	0.407	0.449	0.466	0.455
## 1986	0.504	0.453	0.480	0.497	0.531	0.485	0.526	0.538	0.537
## 1987	0.572	0.525	0.544	0.558	0.565	0.542	0.599	0.584	0.593
## 1988	0.605	0.586	0.625	0.612	0.630	0.635	0.659	0.656	0.660
## 1989	0.733	0.661	0.713	0.694	0.710	0.722	0.741	0.746	0.767
## 1990	0.858	0.764	0.840	0.805	0.809	0.799	0.815	0.828	0.812
## 1991	0.862	0.771	0.813	0.797	0.821	0.801	0.829	0.854	0.882
## 1992	0.938	0.862	0.936	0.932	0.929	0.869	0.891	0.875	0.914
## 1993	0.918	0.838	0.870	0.862	0.852	0.828	0.882	0.867	0.905
## 1994	0.985	0.902	1.015	0.939	0.941	0.935	1.013	1.018	1.041
## 1995	1.076	0.982	1.099	1.068	1.083	1.045	1.094	1.110	1.126
## 1996	1.213	1.128	1.180	1.169	1.146	1.109	1.138	1.146	1.105
## 1997	1.180	1.060	1.148	1.141	1.170	1.113	1.165	1.173	1.154
## 1998	1.186	1.050	1.141	1.107	1.144	1.088	1.162	1.145	1.149
## 1999	1.244	1.124	1.245	1.236	1.271	1.208	1.219	1.234	1.261
## 2000	1.297	1.207	1.325	1.252	1.282	1.275	1.318	1.329	1.432

# Example: Australian eating-out expenditure

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

```
## # 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
```

# Example: Australian eating-out expenditure

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

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

# Example: Australian eating-out expenditure

```
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
```

# Example: Australian eating-out expenditure

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

```
## # A tibble: 1 forecast [1MONTH]
##   data          model      forecast
##   <list>        <model>    <fc>
## 1 <tsibble [426 x 2]> ETS(M,A,M) ~N [h=24]
```



# Example: Australian eating-out expenditure

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

```
## # A tsibble: 24 x 4 [1MONTH]  
##       index mean      80%      95%  
##       <mtm> <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, 4.03]95  
## 3 2017 Dec   4.22 [4.04, 4.40]80 [3.94, 4.49]95  
## 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  
## 6 2018 Mar   3.87 [3.65, 4.09]80 [3.54, 4.20]95  
## 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  
## 9 2018 Jun   3.68 [3.43, 3.92]80 [3.30, 4.05]95  
## 10 2018 Jul  3.88 [3.60, 4.15]80 [3.46, 4.29]95  
## # ... with 14 more rows
```

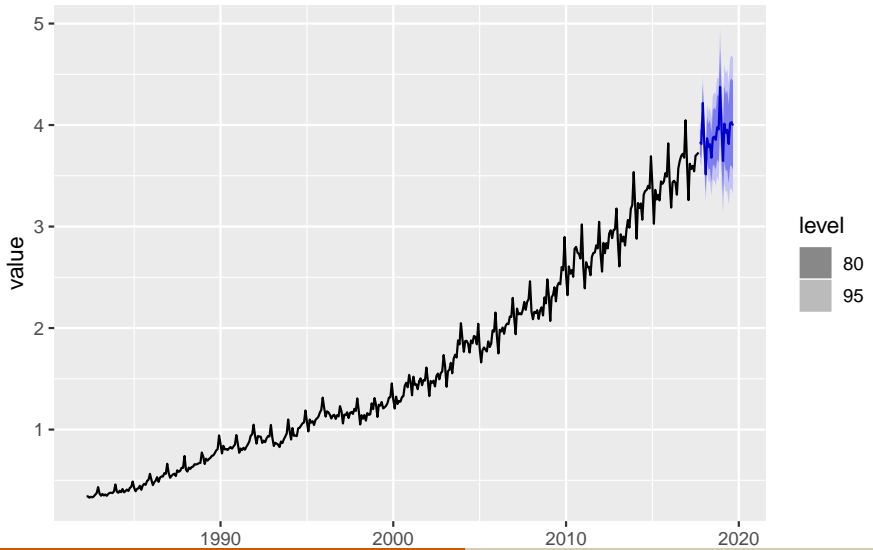
# Example: Australian eating-out expenditure

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

```
## # A tsibble: 24 x 3 [1MONTH]  
##       index mean      90%  
##       <mtm> <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
```

# Example: Australian eating-out expenditure

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



# Example: Australian eating-out expenditure

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

```
## # A tsibble: 24 x 4 [1MONTH]  
##       index mean      80%      95%  
##       <mth> <dbl>    <hilo>    <hilo>  
## 1 2017 Oct  3.81 [3.70, 3.93]80 [3.64, 3.99]95  
## 2 2017 Nov  3.79 [3.65, 3.93]80 [3.58, 4.00]95  
## 3 2017 Dec  4.17 [3.99, 4.34]80 [3.91, 4.43]95  
## 4 2018 Jan  3.73 [3.55, 3.90]80 [3.46, 4.00]95  
## 5 2018 Feb  3.40 [3.22, 3.57]80 [3.14, 3.67]95  
## 6 2018 Mar  3.77 [3.56, 3.99]80 [3.46, 4.10]95  
## 7 2018 Apr  3.70 [3.48, 3.93]80 [3.37, 4.05]95  
## 8 2018 May  3.76 [3.52, 4.00]80 [3.40, 4.13]95  
## 9 2018 Jun  3.66 [3.41, 3.90]80 [3.29, 4.04]95  
## 10 2018 Jul  3.88 [3.61, 4.15]80 [3.48, 4.31]95  
## # ... with 14 more rows
```

# Example: Australian prison population

```
fpp2::prisonLF
```

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

# Example: Australian prison population

```
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]  
## # Key:      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  
## 5 ACT   Female Remanded     4 2006 Q1  
## 6 ACT   Female Remanded     6 2006 Q2  
## 7 ACT   Female Remanded     9 2006 Q3
```

# Example: Australian prison population

```
prison %>% ETS(count)
```

```
## # A mable: 32 models [1QUARTER]
## # Key:      state, gender, legal [32]
##   state gender legal      data      model
##   <fct> <fct>  <fct>    <list>    <model>
## 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   Male   Remanded <tsibble [48 x 2]> ETS(M,A,A)
## 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
```

# Example: Australian prison population

```
prison %>% ETS(count) %>% forecast()
```

```
## # A tibble: 32 forecasts [1QUARTER]
## # Key:      state, gender, legal [32]
##   state gender legal      data      model      forecast
##   <fct> <fct>  <fct>    <list>    <model>    <fc>
## 1 ACT   Female Remanded <tsibble [48 x 2]> ETS(M,A,N) ~N [h=8]
## 2 ACT   Female Sentenced <tsibble [48 x 2]> ETS(A,A,N) ~N [h=8]
## 3 ACT   Male   Remanded <tsibble [48 x 2]> ETS(M,N,N) ~N [h=8]
## 4 ACT   Male   Sentenced <tsibble [48 x 2]> ETS(A,N,N) ~N [h=8]
## 5 NSW   Female Remanded <tsibble [48 x 2]> ETS(M,N,M) ~N [h=8]
## 6 NSW   Female Sentenced <tsibble [48 x 2]> ETS(M,N,M) ~N [h=8]
## 7 NSW   Male   Remanded <tsibble [48 x 2]> ETS(M,A,A) ~N [h=8]
## 8 NSW   Male   Sentenced <tsibble [48 x 2]> ETS(M,A,A) ~N [h=8]
## 9 NT    Female Remanded <tsibble [48 x 2]> ETS(M,N,N) ~N [h=8]
## 10 NT   Female Sentenced <tsibble [48 x 2]> ETS(M,A,A) ~N [h=8]
## # ... with 22 more rows
```

Aggregation and reconciliation not yet implemented.



# Example: Half-hourly electricity demand

elecdemand

```
## # A tsibble: 17,520 x 4 [30MINUTE]
##   index          Demand Temperature WorkDay
##   <dtm>          <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)
```

```
## Series: Demand  
## Regression with ARIMA(1,1,0)(2,0,2)[2] errors  
##  
## Coefficients:  
##          ar1      sar1      sar2      sma1      sma2  Temperature  
##          0.853   -0.181    0.523   -0.066   -0.792         -0.009  
## s.e.      0.005    0.015    0.012    0.012    0.011         0.002  
##          I(Temperature^2)  WorkDay  
##                          0      0.016  
## s.e.                     0      0.006  
##  
## sigma^2 estimated as 0.00846:  log likelihood=16949  
## AIC=-33881  AICc=-33881  BIC=-33811  
##  
## Training set error measures:  
##                          ME  RMSE      MAE      MPE  MAPE  MASE  ACF1  
## Training set 6.51e-06 0.092 0.0634 0.00633 1.39 0.292 0.103  
  
forecast(fit2, newdata=elecdemandfuture) %>% autoplot()
```

## Equivalent methods: forecast → fable

`auto.arima` → ARIMA

`ets` → ETS

`tslm/lm` → LM

`tbats` → TBATS

`nnetar` → NNAR

`stlm` → 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 → fable

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

■ forecast produces fable class objects.

# More information



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

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



Di Cook



Earo Wang



Mitchell O'Hara-Wild

## Follow our progress

- [tidyverts.org](https://tidyverts.org)
- [robjhyndman.com/hyndsight](https://robjhyndman.com/hyndsight)