

**tssmooth** — Smooth and forecast univariate time-series data

[Description](#)[Syntax](#)[Remarks and examples](#)[References](#)[Also see](#)

## Description

`tssmooth` creates new variable *newvar* and fills it in by passing the specified expression (usually a variable name) through the requested smoother.

## Syntax

```
tssmooth smoother [type] newvar = exp [if] [in] [, ...]
```

Smoother category	<i>smoother</i>
Moving average	<a href="#">ma</a>
Recursive	
exponential	<a href="#">exponential</a>
double exponential	<a href="#">dexponential</a>
nonseasonal Holt–Winters	<a href="#">hwinters</a>
seasonal Holt–Winters	<a href="#">shwinters</a>
Nonlinear filter	<a href="#">nl</a>

## Remarks and examples

[stata.com](#)

The recursive smoothers may also be used for forecasting univariate time series; indeed, the Holt–Winters methods are used almost exclusively for this. All can perform dynamic out-of-sample forecasts, and the smoothing parameters may be chosen to minimize the in-sample sum-of-squared prediction errors.

The moving-average and nonlinear smoothers are generally used to extract the trend—or signal—from a time series while omitting the high-frequency or noise components.

All smoothers work both with time-series data and panel data. When used with panel data, the calculation is performed separately within panel.

Several texts provide good introductions to the methods available in `tssmooth`. [Chatfield \(2004\)](#) discusses how these methods fit into time-series analysis in general. [Abraham and Ledolter \(1983\)](#); [Montgomery, Johnson, and Gardiner \(1990\)](#); [Bowerman, O’Connell, and Koehler \(2005\)](#); and [Chatfield \(2001\)](#) discuss using these methods for modern time-series forecasting. [Beckett \(2020\)](#) includes a Stata-centric discussion of these techniques. As he emphasizes, these methods often work as well as more complicated methods and are easier to explain to lay audiences. Do not dismiss these techniques as being too simplistic or inferior.

### References

- Abraham, B., and J. Ledolter. 1983. *Statistical Methods for Forecasting*. New York: Wiley.
- Beckett, S. 2020. *Introduction to Time Series Using Stata*. Rev. ed. College Station, TX: Stata Press.
- Bowerman, B. L., R. T. O'Connell, and A. B. Koehler. 2005. *Forecasting, Time Series, and Regression: An Applied Approach*. 4th ed. Pacific Grove, CA: Brooks/Cole.
- Chatfield, C. 2001. *Time-Series Forecasting*. London: Chapman & Hall/CRC.
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- Montgomery, D. C., L. A. Johnson, and J. S. Gardiner. 1990. *Forecasting and Time Series Analysis*. 2nd ed. New York: McGraw–Hill.
- Winters, P. R. 1960. Forecasting sales by exponentially weighted moving averages. *Management Science* 6: 324–342. <https://doi.org/10.1287/mnsc.6.3.324>.

### Also see

- [TS] **arima** — ARIMA, ARMAX, and other dynamic regression models
- [TS] **sspace** — State-space models
- [TS] **tsfilter** — Filter a time series for cyclical components
- [TS] **tsset** — Declare data to be time-series data
- [R] **smooth** — Robust nonlinear smoother