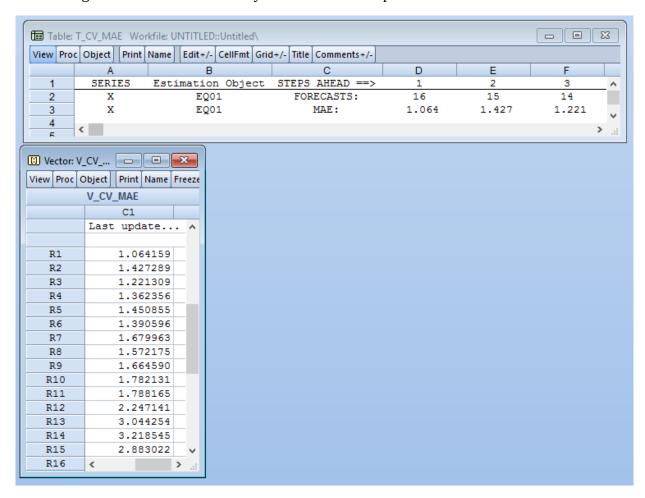
#### Perform time-series cross-validation.

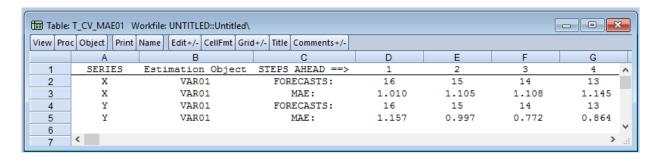
tscval performs rolling estimation and out-of-sample forecast evaluation for equation and VAR objects. If called from an equation object, tscval will return cross-validation results for forecasts of the base (e.g. non-transformed) forms of the dependent variable. If called from a VAR object, tscval will return cross-validation results for forecasts of the base forms of all endogenous variables.

The add-in will return table objects and vector objects (one per variable and error measure) containing cross-validation results by horizon. For example:



The object t\_cv\_mae shows MAE results from the equation object "EQ\_01". The table shows that the average 1-period-ahead MAE was 1.064, and that this average was calculated over 16 forecasts. The vector object v\_cv\_mae mirrors the table data, and is provided for the convenience of users who might want to use cross-validation results for model weighting or selection.

For VAR output, tscval will return one table per error type, with the endogenous variables stacked on top of each other. For example:



The object t\_cv\_mae01 shows MAE results from the VAR object "VAR01". The table shows that the 1-period-ahead MAE for series X was 1.010 and the 1-period-ahead MAE for series Y was 1.157. For VAR objects, tscval will return one vector object per endogenous variable and error type.

For both equation and VAR cross-validation, the returned objects contain two useful metadata attributes to aid in manipulation/combination of the results. The table and vector objects all contain the attribute "Estimation\_Object", which gives the name of the workfile object (equation or VAR) used to produce forecasts. In addition, vector objects contain the attribute "Series", which gives the name of the workfile series object to which the error estimates pertain.

# **Syntax**

{%equation}.tscval(options)
{%var}.tscval(options)

## **Options**

SAMPLE = arg	The range within which cross-validation is performed. Arguments should be strings in valid EViews sample form (e.g., "1999m01 2015m12"). If omitted, SAMPLE will default to the workfile sample
H = arg	Maximum proportion of the training range (specified in SAMPLE) to holdout. Arguments should be real numbers which satisfy 0 < H <= 1. If omitted, H will default to 0.1 (10% of the training range).
ERR = arg(s)	One or more error measures to return. <i>arg</i> takes values of "MSE" (mean squared error), "MAE" (mean absolute error), "RMSE" (root mean squared error), "MSFE" (mean squared forecast error), "medAE" (median absolute error), "MAPE" (mean absolute percent error),

	"SMAPE" (symmetric MAPE), "MPE" (mean percent error), "MSPE" (mean squared percent error), "RMSPE" (root mean squared percent error), "medPE" (median percent error), "medSPE" (median squared percent error), "SIGN" (count of times that the model forecast the correct direction of change over a given horizon), and "SIGNP" (SIGN, expressed as a percentage of forecasts produced). To specify multiple errors, pass in any combination of the arguments above as a space-delimited list (e.g. "MAE MAPE MSE"). Defaults to "MAE".
KEEP_MATS = arg	Return raw matrices of all the errors and forecasts ("t"). If ignored or anything other than "t" is passed, raw matrices will not be returned in the workfile.

## **Examples**

The commands

```
equation eq01.1s d(ip) c d(gdp) eq01.tscval
```

estimate an equation object called EQ01 and generate cross-validation results using the add-in defaults. The workfile will contain a table object and vector object containing out-of-sample forecast error results for ip (the difference operator will be unwound in forecasting).

You may wish to customize the results more. For example, the commands:

estimate an equation object called EQ01 and perform time-series cross-validation. Given the settings from SAMPLE and H, the first equation in the cross-validation exercise will be trained on the sample "2006m01 2011m12" (60% of the argument passed to SAMPLE) and tested over the sample "2012m01 2015m12" (the holdout sample, the remaining 40% of the argument passed to SAMPLE). Three table objects and three vector objects will be created, one for each error measure.

The commands

```
var var01.LS 1 2 LOG(X) LOG(Y) @ C @LAG(LOG(Z),1) @TREND
var01.tscval
```

will produce a similar output with a table for each error type selected in the program command.

Each table will contain error estimates for every endogenous variable in the VAR object.

### **Contact Information**

Please send any questions, comments, criticisms, or complaints to <u>jaylamb20@gmail.com</u>. If you'd like to contribute to the project, please feel free to send a pull request to <a href="https://github.com/jameslamb/ML4EVIEWS/tree/master/tscval">https://github.com/jameslamb/ML4EVIEWS/tree/master/tscval</a>.

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