

intro 4c — Including a lag of a state variable

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Description

Lags of state variables are not allowed to be included in the structural model specified in the `dsge` command. These lags do not fit into the form of a structural model that can be solved for the state-space form. However, this restriction does not prevent us from fitting models that involve the lag of a state variable. This entry shows how to define a new state variable and rewrite the equations in the required structural form.

Remarks and examples

[stata.com](#)

Remarks are presented under the following headings:

A model with a lagged state variable
Parameter estimation

A model with a lagged state variable

The model in (1)–(5) is a standard monetary model in which the state driving output growth y_t is a second-order autoregressive AR(2) process instead of a first-order autoregressive process.

$$p_t = \beta E_t(p_{t+1}) + \kappa y_t \quad (1)$$

$$y_t = E_t(y_{t+1}) - \{r_t - E_t(p_{t+1}) - z_t\} \quad (2)$$

$$r_t = \frac{1}{\beta} p_t + u_t \quad (3)$$

$$z_{t+1} = \rho_{z1} z_t + \rho_{z2} z_{t-1} + \epsilon_{t+1} \quad (4)$$

$$u_{t+1} = \rho_u u_t + \xi_{t+1} \quad (5)$$

Equation (1) specifies the structural equation for inflation p_t . Inflation is a linear combination of expected future inflation $E_t(p_{t+1})$ and output growth y_t . Equation (2) specifies the structural equation for output growth. Output growth is a linear combination of expected future output growth $E_t(y_{t+1})$, the interest rate r_t , expected future inflation, and the state z_t . The state z_t is an AR(2) process that drives output growth. Equation (3) specifies the structural equation for the interest rate. The interest rate depends on the inflation rate and the state u_t . The state u_t is an AR(1) process that drives the interest rate. The control variables in this model are p_t , y_t , and r_t . The state variables are u_t and z_t .

The AR(2) term in (4) is a problematic term because a lag of a state variable does not fit into the structure required to solve the model for the state-space form. We accommodate this term by defining a new state variable Lz_t that equals z_{t-1} and replacing z_{t-1} in (4) with this new state variable. We define a new state instead of new control because the lagged state is exogenous. These changes yield the model in (6)–(11).

$$p_t = \beta E_t(p_{t+1}) + \kappa y_t \quad (6)$$

$$y_t = E_t(y_{t+1}) - \{r_t - E_t(p_{t+1}) - z_t\} \quad (7)$$

$$r_t = \frac{1}{\beta} p_t + u_t \quad (8)$$

$$z_{t+1} = \rho_{z1} z_t + \rho_{z2} L z_t + \epsilon_{t+1} \quad (9)$$

$$u_{t+1} = \rho_u u_t + \xi_{t+1} \quad (10)$$

$$L z_{t+1} = z_t \quad (11)$$

We will estimate the parameters in the model in (6)–(11).

Parameter estimation

We estimate the parameters in the model in (6)–(11) using data on U.S. interest rate r and inflation p .

```
. use http://www.stata-press.com/data/r15/usmacro2
(Federal Reserve Economic Data - St. Louis Fed, 2017-01-15)

. dsge (p = {beta}*E(F.p) + {kappa}*y)
> (y = E(F.y) -(r - E(f.p) - z), unobserved)
> (r = (1/{beta})*p + u)
> (F.u = {rho_u}*u, state)
> (F.z = {rho_z1}*z + {rho_z2}*Lz, state)
> (F.Lz = z, state noshock)
(setting technique to bfgs)
Iteration 0: log likelihood = -9116.0985
Iteration 1: log likelihood = -997.95396 (backed up)
Iteration 2: log likelihood = -881.53897 (backed up)
Iteration 3: log likelihood = -836.47478 (backed up)
Iteration 4: log likelihood = -769.37111 (backed up)
(switching technique to nr)
Iteration 5: log likelihood = -768.80775 (not concave)
Iteration 6: log likelihood = -760.17544 (not concave)
Iteration 7: log likelihood = -755.99765 (not concave)
Iteration 8: log likelihood = -754.93526 (not concave)
Iteration 9: log likelihood = -754.42155
Iteration 10: log likelihood = -753.8269
Iteration 11: log likelihood = -753.20702
Iteration 12: log likelihood = -753.08071
Iteration 13: log likelihood = -753.07788
Iteration 14: log likelihood = -753.07788
```

DSGE model

Sample: 1955q1 - 2015q4

Number of obs = 244

Log likelihood = -753.07788

| | OIM | | z | P> z | [95% Conf. Interval] | |
|-------------|----------|-----------|-------|-------|----------------------|----------|
| | Coef. | Std. Err. | | | | |
| <hr/> | | | | | | |
| /structural | | | | | | |
| beta | .5154874 | .0770496 | 6.69 | 0.000 | .364473 | .6665019 |
| kappa | .1662425 | .0468473 | 3.55 | 0.000 | .0744234 | .2580616 |
| rho | .6979877 | .0452071 | 15.44 | 0.000 | .6093834 | .7865921 |
| rhoz1 | .6735462 | .2666028 | 2.53 | 0.012 | .1510143 | 1.196078 |
| rhoz2 | .2710021 | .2564596 | 1.06 | 0.291 | -.2316495 | .7736537 |
| <hr/> | | | | | | |
| sd(e.u) | 2.315262 | .2992183 | | | 1.728805 | 2.901719 |
| sd(e.z) | .7720678 | .1716214 | | | .4356961 | 1.10844 |

Looking at the confidence interval on rhoz2, we find no evidence that the AR(2) term is needed in this model.

Also see

[DSGE] [intro 2](#) — Learning the syntax

[DSGE] [intro 4](#) — Writing a DSGE in a solvable form