

# Conditional Forecasts

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# Introduction

- A message of these notes is that the following, apparently simple, question is in fact ambiguous:

‘what will happen if the interest rate is increased by 25 basis points (bps) over the following year?’

- Three interpretations:
  - Will set interest rate higher by 25 bps next year, come what may (‘Odyssean policy’).
  - Expect a configuration of shocks next year which (given the usual monetary policy) will drive up the interest rate by 25 bps.
  - Expect a sequence of shocks to monetary policy (which the public thinks are drawn randomly) that will cause the interest rate to be higher by 25 bps, regardless of the pattern of non-monetary shocks.
- Each interpretation has a different implication for forecasts.

## More Examples

- Suppose you have two sets of variables:  $y_t$  and  $x_t$ .
- Would like to forecast  $y_{T+j}$ ,  $j = 1, 2, \dots, f$ , conditional on specified future values of at least a subset of  $x_{T+j}$ ,  $j = 1, 2, \dots, f$ .
- Example 1:
  - $x_t$  denotes the foreign variables in a small open economy model used in the monetary policy division's model and  $y_t$  denotes the domestic variables.
  - The analyst has been given forecasts for at least a subset of elements of  $x_{T+j}$ ,  $j = 1, \dots, f$ , by another division and has been asked to 'determine the distribution of future  $y_t$  under the assumption that what the other division forecasts actually happens'.
- Example 2: There is concern that oil prices,  $x_t$ , will rise in the future and you want to know what that implies for the other variables,  $y_t$ .
- Example 3: what happens if there is a shock to foreigners' appetite for domestic versus foreign assets?

# Simple Example

- Suppose  $x_t$  and  $y_t$  are scalars, with the following representation:

$$y_t = a\mu_t + bv_t$$

$$x_t = \mu_t + v_t,$$

where  $\mu_t$  and  $v_t$  are iid over time and:

$$\mu_t \sim N(0, \sigma_\mu^2), \quad v_t \sim N(0, \sigma_v^2),$$

$$Ev_t\mu_{t-j} = 0, \quad j \in (-\infty, \infty).$$

- Suppose that somehow, we've figured out the model parameters:

$$\sigma_\mu, \sigma_v, a, b.$$

## Simple Example

- At date  $T$  we are asked to construct the distribution of  $Y$ ,

$$Y = \begin{bmatrix} y_{T+1} \\ \vdots \\ y_{T+f} \end{bmatrix},$$

subject to a specific sequence of values of  $x_t$  :

$$X = \begin{bmatrix} x_{T+1} \\ \vdots \\ x_{T+f} \end{bmatrix}.$$

- We expect that  $Y$  has a multivariate Normal distribution with mean a function of  $X$  and variance a function of the model parameters:

$$\sigma_{\mu}, \sigma_{\nu}, a, b.$$

# Simple Example: Signal Extraction

- The example is sufficiently simple that the distribution of  $y_t$ ,  $t > T$  is simply a function of  $x_t$  (no dynamics). For example,

$$E [Y|X] = \begin{bmatrix} E [y_{T+1}|x_{T+1}] \\ \vdots \\ E [y_{T+f}|x_{T+f}] \end{bmatrix}$$

- In general the problem is more complicated and the distribution of  $y_t$ ,  $t > T$ , is a function of the whole of  $X$ , as well as past data.
- Still, the intuition from the example is sufficient to interpret results from Dynare (see below).
- Conditional on  $x_t$ , restriction across shocks:

$$x_t = \mu_t + v_t.$$

- Drawing an inference from  $x_t$  about the density of  $\mu_t$  or  $v_t$  is called a *signal extraction problem*.

# Conditional Expectation of $Y$

- We have:

$$\begin{aligned} E[y_t|x_t] &= aE[\mu_t|x_t] + bE[v_t|x_t] \\ &= [a\beta + b(1 - \beta)]x_t \end{aligned}$$

$$\beta = \frac{\sigma_\mu^2}{\sigma_\mu^2 + \sigma_v^2}.$$

- Basic idea: knowledge of  $x_t$  shrinks set of shocks,  $\mu_t, v_t$ , that can occur.
  - Given  $x_t$ , the most likely value of the shocks is (using the Normality assumption): [▶ details](#)

$$\hat{\mu}_t = \operatorname{argmax}_{\mu_t} f(\mu_t|x_t) = \frac{\sigma_\mu^2}{\sigma_\mu^2 + \sigma_v^2}x_t, \quad \hat{v}_t = \frac{\sigma_v^2}{\sigma_\mu^2 + \sigma_v^2}x_t$$

- Standard signal extraction.

# Getting a Conditional Expectation in Dynare

- Setup:
  - put NaN (MATLAB for 'not a number') in each entry of  $Y$ .
  - put numbers in the elements of  $X$  corresponding to the conditioning information that you have. If there are elements of  $x_t$  that you know nothing about, put in NaN.
- Provide the historical data on  $x_t$  and  $y_t$  and the 'data',  $Y$  and  $X$ .
  - You now have one big data set with missing observations.
  - Implement Dynare's 'estimate' command.
  - Provide all your data to Dynare in the usual way.
  - Set `mode_compute=0` and `mh_replic=0`, but include the 'smoother' command.
  - The structure, `oo_.SmoothedVariables`, contains the results Dynare's Kalman smoother.
    - All entries with NaN will be replaced by estimates.



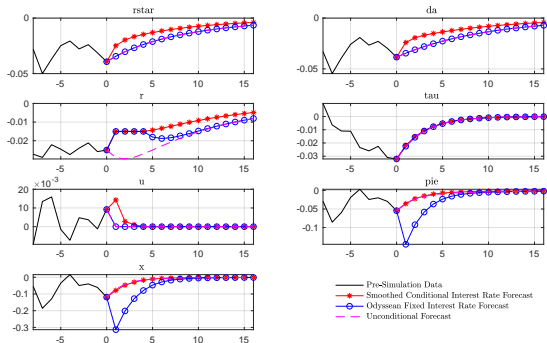
# Experiments to Illustrate the Results

- Take our simple NK model (DS version), with an iid monetary policy shock,  $u_t$ .
- Assign parameter values:

$$\beta = 0.99, \phi_x = 0, \phi_\pi = 1.5, \alpha = 0.85, \rho = 0.9, \lambda = 0.7, \varphi = 1, \theta = 1, \sigma_a = \sigma_\tau = \sigma_u = 0.01.$$

# Odyssean Policy versus Smoothing

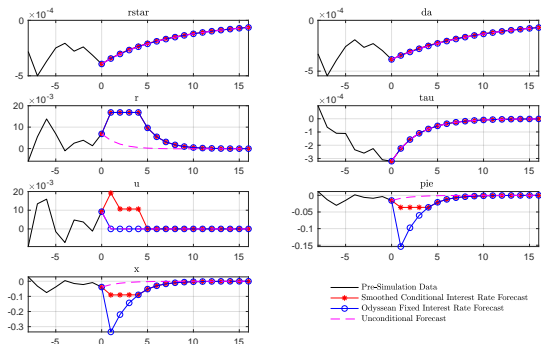
Figure: Baseline Parameterization



Note: with the Odyssean policy and unconditional forecast, all future shocks are set to zero. With the conditional expectation, the procedure picks the most likely configuration of shocks, that explain the fixed interest rate. Note the Kalman smoother does not employ the  $\tau$  shock at all. The most important shock for  $r$  is the  $\varepsilon_t^d$  shock, and that's the one that the smoother puts the most weight on.

# Odyssean Policy versus Smoothing

Figure: Only the Monetary Policy Shock Matters



Note: Here,  $\sigma_u = \sigma_\tau = 0.0001$ , and  $\sigma_{\tilde{u}} = 0.01$ . This has no impact on unconditional forecast and Odyssean policy. There is a substantial impact on the conditional expectation. Given the specification of shock variances, Dynare has been 'tricked' into thinking the monetary policy shock is the only important shock. So, it uses the monetary policy shock to get the interest rate to follow its path. The impact of the Odyssean policy is bigger on output because in that case the monetary authority commits to sticking to the interest rate path, while in the conditional expectations, people are fooled in each period to think that the interest rate is high only temporarily (the monetary policy shock is - according to their belief - only temporary).

# Conclusion

- We've described three ways to interpret the question, 'what will happen if the interest rate remains fixed over the next year or two?'
  - Also, have addressed a large class of conditional forecasting questions.
- The Dynare code, `smallNK_model.mod` implements all three.
  - It is contained in the file, `forecasting.zip`.

## Simple Example: Bayes' Rule

- Let  $p(x)$ ,  $p(\mu)$  denote the marginal density of  $x_t$  and  $\mu_t$  respectively (from here on, we drop the  $t$  subscript).
- Let  $p(x|\mu)$  denote the distribution of  $x$  conditional on observing  $\mu$ .
- Our assumptions imply:

$$p(x|\mu) = \frac{1}{\sqrt{2\pi\sigma_v^2}} \exp \left\{ -\frac{1}{2} \frac{(x - \mu)^2}{\sigma_v^2} \right\}$$

$$p(\mu) = \frac{1}{\sqrt{2\pi\sigma_\mu^2}} \exp \left\{ -\frac{1}{2} \frac{\mu^2}{\sigma_\mu^2} \right\}$$

$$p(x) = \frac{1}{\sqrt{2\pi(\sigma_\mu^2 + \sigma_v^2)}} \exp \left\{ -\frac{1}{2} \frac{x^2}{\sigma_\mu^2 + \sigma_v^2} \right\}.$$

## Simple Example: Bayes' Rule

- With a little algebra, the marginal density of  $\mu$  given  $x$  :

$$p(\mu|x) = \frac{p(x|\mu)p(\mu)}{p(x)} = \frac{1}{\sqrt{2\pi}s^2} \exp \left\{ -\frac{1}{2} \frac{\left( \mu - \frac{\sigma_\mu^2}{\sigma_\mu^2 + \sigma_v^2} x \right)^2}{s^2} \right\},$$

where  $\beta x$  is the value of  $\mu$  that maximizes  $p(\mu|x)$  and

$$s^2 = \frac{\sigma_\mu^2 \sigma_v^2}{\sigma_\mu^2 + \sigma_v^2}, \quad E[\mu|x] = \beta x, \quad \beta = \frac{\sigma_\mu^2}{\sigma_\mu^2 + \sigma_v^2}.$$

- Similarly,

$$E[v|x] = (1 - \beta) x$$