

Introduction to DSGE Modeling

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- Estimated dynamic stochastic general equilibrium (DSGE) models are now widely used for
 - empirical research in macroeconomics;
 - quantitative policy analysis and prediction at central banks.
- We will consider a prototypical New Keynesian DSGE model...

- ① What is the optimal target inflation rate?
- ② Was high inflation and output volatility in the 1970s due to loose monetary policy?
- ③ Effects of the zero lower bound on nominal interest rates on monetary policy.
- ④ How large are government spending multipliers?
- ⑤ Fiscal policy rules and the effect of a change in the labor tax rate.

A Small-Scale New Keynesian DSGE Model

- The model consists of
 - households;
 - final goods producing firms;
 - intermediate goods producing firms;
 - central bank and fiscal authority;
 - exogenous shock processes
- Let's take a look at the decision problems faced by economic agents...

- Households maximize

$$\mathbb{E}_\tau \left[\sum_{t=\tau}^{\infty} \beta^{(t-\tau)} \left\{ \ln C_t - \frac{\phi_t}{1+\nu} L_t^{1+\nu} \right\} \right]$$

- subject to the constraints:

$$P_t C_t + B_{t+1} \leq P_t W_t L_t + \Pi_t + R_{t-1} B_t - T_t + \Omega_t.$$

- In a nutshell:
 - household cares about the future: intertemporal optimization
 - household likes consumption
 - household does not like to work...
 - there is a budget constraint: can't spend more than you earn and borrow; have to pay taxes;

- Households maximize

$$\mathbb{E}_\tau \left[\sum_{t=\tau}^{\infty} \beta^{(t-\tau)} \left\{ \ln C_t - \frac{\phi_t}{1+\nu} L_t^{1+\nu} \right\} \right]$$

- subject to the constraints:

$$P_t C_t + B_{t+1} \leq P_t W_t L_t + \Pi_t + R_{t-1} B_t - T_t + \Omega_t.$$

- Possible modifications/generalizations:
 - let households own shares to the capital stock;
 - introduce money explicitly: cash-in-advance versus money in the utility function;
 - make taxes distortionary;
 - introduce differentiated labor.

Households: First-Order Conditions

- Households maximize

$$\mathbb{E}_\tau \left[\sum_{t=\tau}^{\infty} \beta^{(t-\tau)} \left\{ \ln C_t - \frac{\phi_t}{1+\nu} L_t^{1+\nu} \right\} \right]$$

- subject to the constraints:

$$P_t C_t + B_{t+1} \leq P_t W_t L_t + \Pi_t + R_{t-1} B_t - T_t + \Omega_t.$$

- Introduce Lagrange multiplier μ_t for budget constraint.
- Lagrangian

$$\mathcal{L} = \mathbb{E}_\tau \left[\sum_{t=\tau}^{\infty} \beta^{(t-\tau)} \left\{ \ln C_t - \frac{\phi_t}{1+\nu} L_t^{1+\nu} - \mu_t \left(P_t C_t + B_{t+1} - [P_t W_t L_t + \Pi_t + R_{t-1} B_t - T_t + \Omega_t] \right) \right\} \right]$$

Households: First-Order Conditions

- Lagrangian

$$\mathcal{L} = \mathbb{E}_\tau \left[\sum_{t=\tau}^{\infty} \beta^{(t-\tau)} \left\{ \ln C_t - \frac{\phi_t}{1+\nu} L_t^{1+\nu} - \mu_t \left(P_t C_t + B_{t+1} - [P_t W_t L_t + \Pi_t + R_{t-1} B_t - T_t + \Omega_t] \right) \right\} \right]$$

- First-order condition for C_t :

$$\frac{1}{C_t} = \mu_t P_t$$

- First-order condition for B_{t+1} :

$$\mu_t = \beta \mathbb{E}_t [\mu_{t+1} R_t]$$

- Combine to **consumption Euler equation** (define $\pi_{t+1} = P_{t+1}/P_t$):

$$\frac{1}{C_t} = \beta \mathbb{E}_t \left[\frac{1}{C_{t+1}} \frac{R_t}{\pi_{t+1}} \right]$$

- Lagrangian

$$\mathcal{L} = \mathbb{E}_\tau \left[\sum_{t=\tau}^{\infty} \beta^{(t-\tau)} \left\{ \ln C_t - \frac{\phi_t}{1+\nu} L_t^{1+\nu} - \mu_t \left(P_t C_t + B_{t+1} - [P_t W_t L_t + \Pi_t + R_{t-1} B_t - T_t + \Omega_t] \right) \right\} \right]$$

- Labor supply – first-order condition for L_t :

$$\phi_t L_t^\nu = \mu_t P_t W_t = \frac{W_t}{C_t}.$$

A Small-Scale New Keynesian DSGE Model

- households;
- **final goods producing firms;**
- intermediate goods producing firms;
- central bank and fiscal authority;
- exogenous shock processes

- Production: (these guys just buy and combine intermediate goods)

$$Y_t = \left[\int_0^1 Y_t(i)^{\frac{1}{1+\lambda}} di \right]^{1+\lambda_t}$$

- Profits

$$Y_t P_t - \int Y_t(i) P_t(i) di = \left[\int_0^1 Y_t(i)^{\frac{1}{1+\lambda_t}} di \right]^{1+\lambda_t} P_t - \int Y_t(i) P_t(i) di.$$

- Take prices as given and maximize profits by choosing optimal inputs $Y_t(i)$:

$$P_t(i) = P_t Y_t^{\lambda_t/(1+\lambda_t)} Y_t(i)^{-\lambda_t/(1+\lambda_t)} \implies Y_t(i) = \left(\frac{P_t(i)}{P_t} \right)^{-\frac{1+\lambda_t}{\lambda_t}} Y_t$$

- Free entry leads to zero profits:

$$Y_t P_t = \int Y_t(i) P_t(i) di \implies P_t = \left[\int_0^1 P_t(i)^{-\frac{1}{\lambda_t}} di \right]^{-\lambda_t}.$$

- Aggregate inflation is defined as $\pi_t = P_t/P_{t-1}$.

A Small-Scale New Keynesian DSGE Model

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Intermediate Goods Production

- Production (these guys hire to produce something):

$$Y_t(i) = \max \left\{ A_t L_t(i) - \mathcal{F}, 0 \right\}.$$

- Firms are monopolistically competitive; face downward sloping demand curve:

$$Y_t(i) = \left(\frac{P_t(i)}{P_t} \right)^{-\frac{1+\lambda_t}{\lambda_t}} Y_t.$$

- Firms set prices to maximize profits, but there is a friction:
 - firms can only re-optimize their prices with probability $1 - \zeta_p$;
 - remaining $1 - \zeta_p$ firms adjust their prices by $\bar{\pi}$
- Once prices are set, firms have to produce whatever quantity is demanded.

- Define the real marginal costs of producing a unit Y_{it} as

$$MC_t = \frac{W_t}{A_t}$$

- Decision problem ($\beta^s \Xi_{t+s|t}$ is today's value of a future dollar)

$$\begin{aligned} \max_{\tilde{P}_t(i)} \quad & \mathbb{E}_t \left\{ \sum_{s=0}^{\infty} \zeta_p^s \beta^s \Xi_{t+s|t} Y_{t+s}(i) \left[\tilde{P}_t(i) \bar{\pi}^s - P_{t+s} MC_{t+s} \right] \right\} \\ \text{s.t.} \quad & Y_{t+s}(i) = \left(\frac{\tilde{P}_t(i) \bar{\pi}^s}{P_{t+s}} \right)^{-\frac{1+\lambda_t}{\lambda_t}} Y_{t+s} \end{aligned}$$

- Differentiate with respect to $\tilde{P}_t(i)$ to obtain first-order condition for optimal price.

- First-order condition to determine $\tilde{P}_t(i)$:

$$\mathbb{E}_t \left\{ \sum_{s=0}^{\infty} \zeta_p^s \beta^s \Xi_{t+s|t} \left(\frac{\partial Y_{t+s}(i)}{\partial \tilde{P}_t(i)} (\tilde{P}_t(i) \bar{\pi}^s - P_{t+s} MC_{t+s}) + Y_{t+s}(i) \bar{\pi}^s \right) \right\} = 0,$$

- where

$$\frac{\partial Y_{t+s}(i)}{\partial \tilde{P}_t(i)} = -\frac{1 + \lambda_t}{\lambda_t} \frac{\bar{\pi}^s}{P_{t+s}} \left(\frac{\tilde{P}_t(i) \bar{\pi}^s}{P_{t+s}} \right)^{-\frac{1+\lambda_t}{\lambda_t} - 1} \quad Y_{t+s} = -\frac{1 + \lambda_t}{\lambda_t} \frac{1}{\tilde{P}_t(i)} Y_{t+s}(i)$$

- Assume all optimizing firms choose the same price: $\tilde{P}_t(i) = \tilde{P}_t$.

- Divide FOC by P_t and impose symmetry. Let $\tilde{p}_t = \tilde{P}_t/P_t$.
- First-order condition to determine \tilde{p}_t :

$$\mathbb{E}_t \left\{ \sum_{s=0}^{\infty} \zeta_p^s \beta^s \frac{\bar{\Xi}_{t+s|t}}{\lambda_t \tilde{p}_t} \left(\frac{\tilde{p}_t \bar{\pi}^s}{\prod_{j=1}^s \pi_{t+j}} \right)^{-\frac{1+\lambda_t}{\lambda_t}} Y_{t+s} \left[\tilde{p}_t \bar{\pi}^s - (1 + \lambda_t) \left(\prod_{j=1}^s \pi_{t+j} \right) MC_{t+s} \right] \right\} = 0,$$

- New Keynesian Phillips curve: relationship between \tilde{p}_t , inflation π_t , and real marginal costs MC_t .

Intermediate Goods Production

- Recall from final goods producers:

$$P_t = \left[\int_0^1 P_t(i)^{-\frac{1}{\lambda_t}} di \right]^{-\lambda_t}.$$

- Fraction ζ_p will index previous price $P_{t-1}(i)$ by inflation, whereas fraction $(1 - \zeta_p)$ will charge \tilde{P}_t :

$$\begin{aligned} P_t &= \left[(1 - \zeta_p) \tilde{P}_t^{-\frac{1}{\lambda_t}} + \zeta_p \bar{\pi}^{-\frac{1}{\lambda_t}} \int_0^1 P_{t-1}(i)^{-\frac{1}{\lambda_t}} di \right]^{-\lambda_t} \\ &= \left[(1 - \zeta_p) \tilde{P}_t^{-\frac{1}{\lambda_t}} + \zeta_p \bar{\pi}^{-\frac{1}{\lambda_t}} P_{t-1}^{-\frac{1}{\lambda_t}} \right]^{-\lambda_t} \end{aligned}$$

- Inflation satisfies (let $\tilde{p}_t = \tilde{P}_t/P_t$):

$$\pi_t = \left[(1 - \zeta_p) (\pi_t \tilde{p}_t)^{-\frac{1}{\lambda_t}} + \zeta_p \bar{\pi}^{-\frac{1}{\lambda_t}} \right]^{-\lambda_t}$$

- Most complicated part of the model...
- generates a relationship between real marginal costs and inflation.
- So, it connects nominal and real side of the economy.
- **Exercise:** if $\zeta_p = 0$ prices are flexible. Simplify the formulas!

A Small-Scale New Keynesian DSGE Model

- households;
- final goods producing firms;
- intermediate goods producing firms;
- **central bank and fiscal authority;**
- exogenous shock processes

- We did not specify a money demand equation, but we could. It would depend on the nominal interest rate. The higher R_t , the lower the demand for money.
- Central bank prints enough money so that demand is satisfied at **interest rate implied by monetary policy rule**:

$$R_t = R_{*,t}^{1-\rho_R} R_{t-1}^{\rho_R} \exp\{\sigma_{R\epsilon_{R,t}}\}, \quad R_{*,t} = (r\pi_*) \left(\frac{\pi_t}{\pi_*}\right)^{\psi_1} \left(\frac{Y_t}{\gamma Y_{t-1}}\right)^{\psi_2}$$

- r is equilibrium real rate.
- π_* is target inflation rate.
- $\epsilon_{R,t}$ is exogenous monetary policy shock. Interpretation?

- For now, it's passive and not very interesting.
- Budget constraint:

$$P_t G_t + R_{t-1} B_t + M_t = T_t + B_t + M_{t+1}$$

- Lump-sum taxes/transfer balance the budget in every period. Seigniorage does not matter.
- Government spending is exogenous. Re-scale:

$$G_t = \left(1 - \frac{1}{g_t}\right) Y_t.$$

A Small-Scale New Keynesian DSGE Model

- households;
- final goods producing firms;
- intermediate goods producing firms;
- central bank and fiscal authority;
- **exogenous shock processes.**

Exogenous shock processes

- Total factor productivity A_t .
- Preference / labor demand shifter ϕ_t .
- Mark-up shock λ_t .
- Monetary policy shock $\epsilon_{R,t}$.
- Government spending shock g_t .
- We will specify exogenous laws of motions for these processes, e.g.,

$$\ln g_t = (1 - \rho_g) \ln g^* + \rho_g \ln g_{t-1} + \sigma_g \epsilon_{g,t}, \quad \epsilon_{g,t} \sim N(0, 1).$$

Aggregate Resource Constraint

- Combine household and government budget constraints:

$$P_t C_t + P_t G_t = P_t W_t \int L_t(i) di + \int \Pi_t(i) di$$

- Final goods producers make zero profits, which implies:

$$P_t Y_t = \int P_t(i) Y_t(i) di.$$

- Profits of intermediate goods producers:

$$\begin{aligned} \int \Pi_t(i) di &= \int Y_t(i) P_t(i) di - P_t W_t \int L_t(i) di - \mathcal{F} \\ &= P_t Y_t - P_t W_t L_t - \mathcal{F}. \end{aligned}$$

- Thus, assuming $\mathcal{F} = 0$:

$$C_t + G_t = Y_t.$$

Aggregate Resource Constraint

- Production:

$$Y_t(i) = A_t L_t(i)$$

- Using the demand function for $Y_t(i)$ we can write

$$Y_t \left(\frac{P_t(i)}{P_t} \right)^{-\frac{1+\lambda_t}{\lambda_t}} = A_t L_t(i).$$

- Integrating over the firms i yields:

$$Y_t = \frac{1}{D_t} A_t L_t, \quad D_t = \int \left(\frac{P_t(i)}{P_t} \right)^{-\frac{1+\lambda_t}{\lambda_t}} di \geq 1$$

- Price dispersion creates a loss of output!

Evolution of Price Dispersion

- Recall

$$D_t = \int \left(\frac{P_t(i)}{P_t} \right)^{-\frac{1+\lambda_t}{\lambda_t}} di$$

- A fraction of ζ_p firms changes its price in each period. Thus,

$$\begin{aligned} D_t &= (1 - \zeta_p) \sum_{j=0}^{\infty} \zeta^j \left(\frac{\bar{\pi}^j \tilde{P}_{t-j}}{\pi_t \pi_{t-1} \cdots \pi_{t-j+1} P_{t-j}} \right)^{-\frac{1+\lambda_t}{\lambda_t}} \\ &= (1 - \zeta_p) \sum_{j=0}^{\infty} \zeta^j \left(\frac{\bar{\pi}^j}{\pi_t \pi_{t-1} \cdots \pi_{t-j+1}} \tilde{p}_t \right)^{-\frac{1+\lambda_t}{\lambda_t}} \end{aligned}$$

- Firms discount future profits using the households stochastic discount factor:

$$\Xi_{t+s|t} = \frac{C_t}{C_{t+1}}$$

- We now have a small-scale New Keynesian DSGE model! What are the policy trade-offs? What policies can we study?
- **Monetary policy:**
 - systematic part (react to inflation and output growth): what happens if we change inflation target π^* ? What happens if CB reacts more aggressively to inflation deviations?
 - discretionary component: what happens if CB raises interest rates in an unanticipated fashion, i.e., $\epsilon_{R,t} > 0$?
- **Fiscal policy:**
 - systematic part: what happens if g^* increases?
 - unanticipated: reaction to $\epsilon_{g,t}$.
- To answer other questions, we need to enrich the model:
 - ZLB constraint;
 - role for unconventional monetary policy;
 - distortionary taxes;
 - more interesting debt dynamics.

- After deriving the equilibrium conditions of the model, we now need to solve for the dynamics of the endogenous variables.
- System of nonlinear expectational difference equations;
- Find solution(s) of system of expectational difference equations:
 - global (nonlinear) approximation methods;
 - local approximation near steady state.
- We will focus on log-linear approximations around the steady state.
- Many more details in FVRRS.

Our Goal: State-space Representation of DSGE Model

- $n_y \times 1$ vector of observables:

$$y_t = M'_y [\log(X_t/X_{t-1}), \log lsh_t, \log \pi_t, \log R_t]'$$

- $n_s \times 1$ vector of econometric state variables s_t

$$s_t = [\phi_t, \lambda_t, z_t, \epsilon_{R,t}, \hat{x}_{t-1}]'$$

- DSGE model parameters:

$$\theta = [\beta, \gamma, \lambda, \pi^*, \zeta_p, \nu, \rho_\phi, \rho_\lambda, \rho_z, \sigma_\phi, \sigma_\lambda, \sigma_z, \sigma_R]'$$

- Measurement equation:

$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t.$$

- State-transition equation:

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t, \quad \epsilon_t = [\epsilon_{\phi,t}, \epsilon_{\lambda,t}, \epsilon_{z,t}, \epsilon_{R,t}]'$$

Our Goal: State-Space Representation of DSGE Model

State-space representation:

$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t$$

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t$$

System matrices:

$$\Psi_0(\theta) = M'_y \begin{bmatrix} \log \gamma \\ \log(lsh) \\ \log \pi^* \\ \log(\pi^* \gamma / \beta) \end{bmatrix}, \quad x_\phi = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\phi}, \quad x_\lambda = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\lambda}, \quad x_z = \frac{\rho_z \psi_p}{1 - \psi_p \rho_z}, \quad x_{\epsilon_R} = -\psi_p \sigma_R$$

$$\Psi_1(\theta) = M'_y \begin{bmatrix} x_\phi & x_\lambda & x_z + 1 & x_{\epsilon_R} & -1 \\ 1 + (1 + \nu)x_\phi & (1 + \nu)x_\lambda & (1 + \nu)x_z & (1 + \nu)x_{\epsilon_R} & 0 \\ \frac{\kappa_p}{1 - \beta \rho_\phi} (1 + (1 + \nu)x_\phi) & \frac{\kappa_p}{1 - \beta \rho_\lambda} (1 + (1 + \nu)x_\lambda) & \frac{\kappa_p}{1 - \beta \rho_z} (1 + \nu)x_z & +\kappa_p (1 + \nu)x_{\epsilon_R} & 0 \\ \frac{\kappa_p / \beta}{1 - \beta \rho_\phi} (1 + (1 + \nu)x_\phi) & \frac{\kappa_p / \beta}{1 - \beta \rho_\lambda} (1 + (1 + \nu)x_\lambda) & \frac{\kappa_p / \beta}{1 - \beta \rho_z} (1 + \nu)x_z & (\kappa_p (1 + \nu)x_{\epsilon_R} / \beta + \sigma_R) & 0 \end{bmatrix}$$

$$\Phi_1(\theta) = \begin{bmatrix} \rho_\phi & 0 & 0 & 0 & 0 \\ 0 & \rho_\lambda & 0 & 0 & 0 \\ 0 & 0 & \rho_z & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ x_\phi & x_\lambda & x_z & x_{\epsilon_R} & 0 \end{bmatrix}, \quad \Phi_\epsilon(\theta) = \begin{bmatrix} \sigma_\phi & 0 & 0 & 0 \\ 0 & \sigma_\lambda & 0 & 0 \\ 0 & 0 & \sigma_z & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

M'_y is an $n_y \times 4$ selection matrix that selects rows of Ψ_0 and Ψ_1 .

Steady State

- **Shut down aggregate uncertainty:** set all shock standard deviations $\sigma. = 0$.
- **Technology:**

$$\ln A_t = \ln \gamma + \ln A_{t-1} + z_t, \quad z_t = \rho_z z_{t-1} + \sigma_z \epsilon_{z,t}.$$

Set $\sigma_z = 0$: $\ln A_t^* = \gamma t$.

- **Preferences:**

$$\ln \phi_t = (1 - \rho_\phi) \ln \phi + \rho_\phi \ln \phi_{t-1} + \sigma_\phi \epsilon_{\phi,t}.$$

- **Mark-up:**

$$\ln \lambda_t = (1 - \rho_\lambda) \ln \lambda + \rho_\lambda \ln \lambda_{t-1} + \sigma_\lambda \epsilon_{\lambda,t}.$$

- **Government Spending:**

$$\ln g_t = (1 - \rho_g) \ln g^* + \rho_g \ln g_{t-1} + \sigma_g \epsilon_{g,t}$$

- Problem: this economy grows... which does not lead to a **steady** state.
- Solution: detrend model variables by A_t .
- Model has steady state in terms of detrended variables.

- Recall:

$$\frac{1}{C_t} = \beta \mathbb{E}_t \left[\frac{1}{C_{t+1}} \frac{R_t}{\pi_{t+1}} \right]$$

- Rewrite:

$$\frac{A_t}{C_t} = \beta \mathbb{E}_t \left[\frac{A_{t+1}}{C_{t+1}} \frac{A_t}{A_{t+1}} \frac{R_t}{\pi_{t+1}} \right] \quad \implies \quad \frac{1}{C_t} = \beta \mathbb{E}_t \left[\frac{1}{C_{t+1}} \frac{1}{\gamma e^{z_{t+1}}} \frac{R_t}{\pi_{t+1}} \right]$$

- Steady state:

$$R = \pi \frac{\gamma}{\beta} = \pi r.$$

- Recall:

$$\phi_t L_t^\nu = \frac{W_t}{C_t}$$

- Rewrite:

$$\phi_t L_t^\nu = \frac{W_t/A_t}{C_t/A_t} \implies \phi_t L_t^\nu = \frac{w_t}{c_t}$$

- Steady state:

$$\phi L^\nu = \frac{w}{c}.$$

Intermediate Goods Production

- Recall:

$$MC_t = \frac{W_t}{A_t}.$$

- Steady state:

$$mc = w.$$

- Recall:

$$\pi_t = \left[(1 - \zeta_p)(\pi_t \tilde{p}_t)^{-\frac{1}{\lambda_t}} + \zeta_p \bar{\pi}^{-\frac{1}{\lambda_t}} \right]^{-\lambda_t}.$$

- Steady state:

$$\pi = \left[(1 - \zeta_p)(\pi \tilde{p})^{-\frac{1}{\lambda}} + \zeta_p \bar{\pi}^{-\frac{1}{\lambda}} \right]^{-\lambda}.$$

- Recall:

$$C_t \mathbb{E}_t \left\{ \sum_{s=0}^{\infty} \zeta_p^s \beta^s \frac{Y_{t+s}/C_{t+s}}{\lambda_t \tilde{p}_t} \left(\frac{\tilde{p}_t \bar{\pi}^s}{\prod_{j=1}^s \pi_{t+j}} \right)^{-\frac{1+\lambda_t}{\lambda_t}} \left[\tilde{p}_t \bar{\pi}^s - (1 + \lambda_t) \left(\prod_{j=1}^s \pi_{t+j} \right) MC_{t+s} \right] \right\} = 0,$$

- Steady state:

$$\frac{c/y}{\lambda \tilde{p}} \tilde{p}^{-\frac{1+\lambda}{\lambda}} \left\{ \sum_{s=0}^{\infty} \zeta_p^s \beta^s \left(\frac{\bar{\pi}^s}{\pi^s} \right)^{-\frac{1+\lambda}{\lambda}} [\tilde{p}_t \bar{\pi}^s - (1 + \lambda) \pi^s mc] \right\} = 0,$$

- **Monetary policy rule:**

$$R = r\pi_* \left(\frac{\pi}{\pi_*} \right)^{\psi_1}$$

- **Government spending:**

$$g = \left(1 - \frac{1}{g_*} \right) y$$

Aggregate Resource Constraint and Price Dispersion

- **Market clearing:**

$$c + \left(1 - \frac{1}{g_*}\right) y = y \quad \implies \quad c = \frac{1}{g_*} y.$$

- **Aggregate production:**

$$y = \frac{1}{D} L.$$

- **Price dispersion:**

$$D = (1 - \zeta_p) \sum_{j=0}^{\infty} \zeta_p^j \left(\frac{\bar{\pi}^j}{\pi^j} \tilde{p} \right)^{-\frac{1+\lambda}{\lambda}} = \tilde{p}^{-\frac{1+\lambda}{\lambda}} \frac{1 - \zeta_p}{1 - \zeta_p (\bar{\pi}/\tilde{\pi})^{-\frac{1+\lambda}{\lambda}}}.$$

Combining Bits and Pieces

- Steady state equations are quite complicated.
- **Special case:** $\bar{\pi} = \pi_*$, i.e., price setters index prices by target inflation rate.
- Verify that $\pi = \pi_* = \bar{\pi}$ is an equilibrium:
 - Policy rule and Euler equation imply $R = \pi r$, where $r = \gamma/\beta$.
 - For $\bar{\pi} = \pi$ the condition

$$\pi = \left[(1 - \zeta_p)(\pi \tilde{p})^{-\frac{1}{\lambda}} + \zeta_p \bar{\pi}^{-\frac{1}{\lambda}} \right]^{-\lambda}.$$

implies $\tilde{p} = 1$.

- Thus, there is no steady state price dispersion: $D = 1$.
- The firms' FOC imply that

$$mc = w = \frac{1}{1 + \lambda} \implies \tilde{p} = (1 + \lambda)mc.$$

- Using $c = y/g_*$ and $y = l$, the households' labor supply condition implies

$$\phi y^\nu = \frac{w}{c} = \frac{1}{1 + \lambda} \frac{g_*}{y} \implies y = \left(\frac{g_*}{\phi(1 + \lambda)} \right)^{1/(1+\nu)}.$$

- Change the target inflation rate π_* , assuming that indexation to $\bar{\pi}$ does not change.
Crucial parameter: ζ_p .
- Change the amount of government spending through g_* and compute long-run multipliers.
Crucial parameter ν .
- Estimate model to obtain policy-effect relevant parameters.
- Parameter uncertainty translates into policy uncertainty.

(Log) Linearization Around Steady State

- We will now approximate the equilibrium dynamics of the model.
- Taylor series expansion around around the steady state.
- Linear rational exectations system:

$$\widehat{c}_t = \mathbb{E}_{t+1}[\widehat{c}_{t+1}] - \left(\widehat{R}_t - \mathbb{E}[\widehat{\pi}_{t+1}] \right) + \mathbb{E}_t[z_{t+1}]$$

$$\widehat{\pi}_t = \beta \mathbb{E}_t[\widehat{\pi}_{t+1}] + \kappa_p (\widehat{ish}_t + \lambda_t)$$

$$\widehat{R}_t = \psi_1 \widehat{\pi}_t + \psi_2 (\widehat{y}_t - \widehat{y}_{t-1} + z_t) + \sigma_{R \in R, t}$$

$$\widehat{ish}_t = (1 + \nu) \widehat{c}_t + \nu \widehat{g}_t + \phi_t$$

$$\widehat{y}_t = \widehat{c}_t + \widehat{g}_t$$

State-space Representation of DSGE Model

- $n_y \times 1$ vector of observables:

$$y_t = M'_y [\log(X_t/X_{t-1}), \log lsh_t, \log \pi_t, \log R_t]'$$

- $n_s \times 1$ vector of econometric state variables s_t

$$s_t = [\phi_t, \lambda_t, z_t, \epsilon_{R,t}, \hat{x}_{t-1}]'$$

- DSGE model parameters:

$$\theta = [\beta, \gamma, \lambda, \pi^*, \zeta_p, \nu, \rho_\phi, \rho_\lambda, \rho_z, \sigma_\phi, \sigma_\lambda, \sigma_z, \sigma_R]'$$

- Measurement equation:

$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t.$$

- State-transition equation:

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t, \quad \epsilon_t = [\epsilon_{\phi,t}, \epsilon_{\lambda,t}, \epsilon_{z,t}, \epsilon_{R,t}]'$$

State-Space Representation of DSGE Model

State-space representation:

$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t$$

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t$$

System matrices:

$$\Psi_0(\theta) = M'_y \begin{bmatrix} \log \gamma \\ \log(lsh) \\ \log \pi^* \\ \log(\pi^* \gamma / \beta) \end{bmatrix}, \quad x_\phi = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\phi}, \quad x_\lambda = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\lambda}, \quad x_z = \frac{\rho_z \psi_p}{1 - \psi_p \rho_z}, \quad x_{\epsilon R} = -\psi_p \sigma_R$$

$$\Psi_1(\theta) = M'_y \begin{bmatrix} x_\phi & x_\lambda & x_z + 1 & x_{\epsilon R} & -1 \\ 1 + (1 + \nu)x_\phi & (1 + \nu)x_\lambda & (1 + \nu)x_z & (1 + \nu)x_{\epsilon R} & 0 \\ \frac{\kappa_p}{1 - \beta \rho_\phi} (1 + (1 + \nu)x_\phi) & \frac{\kappa_p}{1 - \beta \rho_\lambda} (1 + (1 + \nu)x_\lambda) & \frac{\kappa_p}{1 - \beta \rho_z} (1 + \nu)x_z & +\kappa_p (1 + \nu)x_{\epsilon R} & 0 \\ \frac{\kappa_p / \beta}{1 - \beta \rho_\phi} (1 + (1 + \nu)x_\phi) & \frac{\kappa_p / \beta}{1 - \beta \rho_\lambda} (1 + (1 + \nu)x_\lambda) & \frac{\kappa_p / \beta}{1 - \beta \rho_z} (1 + \nu)x_z & (\kappa_p (1 + \nu)x_{\epsilon R} / \beta + \sigma_R) & 0 \end{bmatrix}$$

$$\Phi_1(\theta) = \begin{bmatrix} \rho_\phi & 0 & 0 & 0 & 0 \\ 0 & \rho_\lambda & 0 & 0 & 0 \\ 0 & 0 & \rho_z & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ x_\phi & x_\lambda & x_z & x_{\epsilon R} & 0 \end{bmatrix}, \quad \Phi_\epsilon(\theta) = \begin{bmatrix} \sigma_\phi & 0 & 0 & 0 \\ 0 & \sigma_\lambda & 0 & 0 \\ 0 & 0 & \sigma_z & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

M'_y is an $n_y \times 4$ selection matrix that selects rows of Ψ_0 and Ψ_1 .

What is a Local Approximation?

- In a nutshell... consider the backward-looking model

$$y_t = f(y_{t-1}, \sigma\epsilon_t). \quad (1)$$

- Suppose there is a steady state y^* satisfies $y^* = f(y^*, 0)$.
- Guess that the solution to (1) is of the form

$$y_t = y^* + \sigma y_t^{(1)} + o(\sigma). \quad (2)$$

- Taylor series expansion of $f(\cdot)$ around steady state:

$$f(y_{t-1}, \sigma\epsilon_t) = y^* + f_y y_{t-1} + f_\epsilon \sigma\epsilon_t + o(|y_{t-1}|) + o(\sigma)$$

- Now plug-in conjectured solution (2) into (1) using approx of $f(\cdot)$:

$$y^* + \sigma y_t^{(1)} + o(\sigma) = y^* + f_y \sigma y_{t-1}^{(1)} + f_\epsilon \sigma\epsilon_t + o(\sigma)$$

- Deduce that $y_t^{(1)} = f_y y_{t-1}^{(1)} + f_\epsilon \epsilon_t$.

What is a Log-Linear Approximation?

- Consider Cobb-Douglas production function: $Y_t = Z_t K_t^\alpha H_t^{1-\alpha}$.

- **Linearization** around Y_* , Z_* , K_* , H_* :

$$Y_t - Y_* = K_*^\alpha H_*^{1-\alpha} (Z_t - Z_*) + \alpha Z_* K_*^{\alpha-1} H_*^{1-\alpha} (K_t - K_*) \\ + (1 - \alpha) Z_* K_*^\alpha H_*^{-\alpha} (H_t - H_*)$$

- **Log-linearization**: Let $f(x) = f(e^v)$ and linearize with respect to v :

$$f(e^v) \approx f(e^{v_*}) + e^{v_*} f'(e^{v_*})(v - v_*).$$

Thus:

$$f(x) \approx f(x_*) + x_* f'(x_*) (\ln x / x_*) = f(x_*) + f'(x_*) \hat{x}$$

- Cobb-Douglas production function:

$$\tilde{Y}_t = \hat{Z}_t + \alpha \hat{K}_t + (1 - \alpha) \hat{H}_t$$

Let's Try the Log-linearizations

- **Euler Equation:**

$$\frac{1}{c_t} = \beta \mathbb{E}_t \left[\frac{1}{c_{t+1}} \frac{1}{\gamma e^{z_{t+1}}} \frac{R_t}{\pi_{t+1}} \right].$$

- **Log-linearized:**

$$-\widehat{c}_t = \mathbb{E}_t \left[-\widehat{c}_{t+1} - z_{t+1} + \widehat{R}_t - \widehat{\pi}_{t+1} \right] \implies \widehat{c}_t = \mathbb{E}_t[\widehat{c}_{t+1}] - (\widehat{R}_t - \mathbb{E}[\widehat{\pi}_{t+1}]) + \mathbb{E}_t[z_{t+1}].$$

- **Labor Supply:**

$$\phi_t L_t^\nu = \frac{w_t}{c_t}.$$

- **Log-linearized:**

$$\widehat{\phi}_t + \nu \widehat{L}_t = \widehat{w}_t - \widehat{c}_t$$

Let's Try the Log-linearizations

- **Aggregate Resource Constraint:**

$$y_t = \frac{L_t}{D_t}, \quad c_t + \left(1 - \frac{1}{g_t}\right) y_t = y_t \quad \implies c_t g_t = y_t.$$

- **Log-linearized:**

$$\hat{y}_t = \hat{L}_t - \hat{D}_t, \quad \hat{c}_t + \hat{g}_t = \hat{y}_t.$$

- **Monetary Policy Rule:**

$$R_t = R_{*,t}^{1-\rho_R} R_{t-1}^{\rho_R} \exp\{\sigma_R \epsilon_{R,t}\}, \quad R_{*,t} = (r\pi_*) \left(\frac{\pi_t}{\pi_*}\right)^{\psi_1} \left(\frac{Y_t}{\gamma Y_{t-1}}\right)^{\psi_2}.$$

- **Log-linearized**

$$\hat{R}_t = (1 - \rho_R) \hat{R}_{*,t} + \rho_R \hat{R}_{t-1} + \sigma_R \epsilon_{R,t}, \quad \hat{R}_{*,t} = \psi_1 \hat{\pi}_t + \psi_2 [\hat{y}_t - \hat{y}_{t-1} + z_t].$$

New Keynesian Phillips Curve

- This is fairly complicated... let's focus on the result.

- **Assume:** $\pi = \bar{\pi} = \pi_*$

- Note that

$$\widehat{mc}_t = \widehat{w}_t = \widehat{ls}h_t.$$

- **Log-linearized:**

$$\widehat{\pi}_t = \beta \mathbb{E}_t[\widehat{\pi}_{t+1}] + \kappa_p(\widehat{ls}h_t + \lambda_t), \quad \kappa_p = \frac{(1 - \zeta_p \beta)(1 - \zeta_p)}{\zeta_p}.$$

- We also get $\widehat{D}_t = 0$.

Combining Bits and Pieces

- **Notation:** write x_t instead of y_t for output.
- **Assume:** $\pi = \bar{\pi} = \pi_*$, $\psi_1 = 1/\beta$, $\psi_2 = 0$, $\rho_R = 0$.
- **Linear rational expectations (LRE) system:**

$$\widehat{c}_t = \mathbb{E}_{t+1}[\widehat{c}_{t+1}] - \left(\widehat{R}_t - \mathbb{E}[\widehat{\pi}_{t+1}] \right) + \mathbb{E}_t[z_{t+1}]$$

$$\widehat{\pi}_t = \beta \mathbb{E}_t[\widehat{\pi}_{t+1}] + \kappa_p (\widehat{ish}_t + \lambda_t)$$

$$\widehat{R}_t = \frac{1}{\beta} \widehat{\pi}_t + \sigma_R \epsilon_{R,t}$$

$$\widehat{ish}_t = (1 + \nu) \widehat{c}_t + \nu \widehat{g}_t + \phi_t$$

$$\widehat{x}_t = \widehat{c}_t + \widehat{g}_t$$

$$\widehat{g}_t = \rho_g \widehat{g}_{t-1} + \sigma_g \epsilon_{g,t}$$

$$\phi_t = \rho_\phi \phi_{t-1} + \sigma_\phi \epsilon_{\phi,t}$$

$$\lambda_t = \rho_\lambda \lambda_{t-1} + \sigma_\lambda \epsilon_{\lambda,t}$$

$$z_t = \rho_z z_{t-1} + \sigma_z \epsilon_{z,t}$$

How Can One Solve LRE Systems? A Simple Example

Simple model:

$$y_t = \frac{1}{\theta} \mathbb{E}_t[y_{t+1}] + \epsilon_t, \quad \epsilon_t \sim iid(0, 1), \quad \theta \in \Theta = [0, 2].$$

- **Method 1:** Introduce conditional expectation $\xi_t = \mathbb{E}_t[y_{t+1}]$ and forecast error $\eta_t = y_t - \xi_{t-1}$:

$$\xi_t = \theta \xi_{t-1} - \theta \epsilon_t + \theta \eta_t.$$

Nonexplosive solutions:

- **Determinacy:** $\theta > 1$. The only stable solution:

$$\xi_t = 0, \quad \eta_t = \epsilon_t \quad \implies \quad y_t = \epsilon_t$$

- **Indeterminacy:** $\theta \leq 1$ the stability requirement imposes no restrictions on forecast error:

$$\eta_t = \tilde{M}\epsilon_t + \zeta_t \quad \implies \quad y_t = \theta y_{t-1} + \tilde{M}\epsilon_t + \zeta_t - \theta \epsilon_{t-1}$$

How Can One Solve LRE Systems? A Simple Example

Simple model:

$$y_t = \frac{1}{\theta} \mathbb{E}_t[y_{t+1}] + \epsilon_t, \quad \epsilon_t \sim iid(0, 1), \quad \theta \in \Theta = [0, 2].$$

- **Method 2:** Construct nonexplosive solutions as follows:

- **Determinacy:** $\theta > 1$. Solve equation forward:

$$y_t = \epsilon_t + \frac{1}{\theta} \mathbb{E}_t \left[\frac{1}{\theta} \mathbb{E}_{t+1}[y_{t+2}] + \epsilon_{t+1} \right] = \sum_{s=0}^{\infty} \mathbb{E}_t \left[\left(\frac{1}{\theta} \right)^s \epsilon_{t+s} \right] = \epsilon_t.$$

- **Indeterminacy:** $\theta \leq 1$. Express model in terms of $\xi_t = \mathbb{E}_t[y_{t+1}]$ and solve backward (as in previous slide).

How Can One Solve LRE Systems? A Simple Example

Simple model:

$$y_t = \frac{1}{\theta} \mathbb{E}_t[y_{t+1}] + \epsilon_t, \quad \epsilon_t \sim iid(0, 1), \quad \theta \in \Theta = [0, 2].$$

- **Method 3:** Undetermined coefficients. Guess that $y_t = \gamma_1 y_{t-1} + \gamma_2 \epsilon_t + \gamma_3 \epsilon_{t-1}$. Thus,

$$y_t = \frac{1}{\theta} \mathbb{E}_t[\gamma_1 y_t + \gamma_2 \epsilon_{t+1} + \gamma_3 \epsilon_t] + \epsilon_t$$

Nonexplosive solutions:

- **Indeterminacy:** $\theta \leq 1$

$$y_t : \gamma_1 = \gamma_1^2 / \theta \implies \gamma_1 = 0 \text{ or } \gamma_1 = \theta$$

$$\epsilon_t : \gamma_2 \text{ is unrestricted}$$

$$\epsilon_{t-1} : 0 = \gamma_3 / \theta + 1 \implies \gamma_3 = 0 \text{ or } \gamma_3 = -\theta$$

- **Determinacy:** $\theta > 1$. We cannot set $\gamma_1 = \theta$. Thus,

$$\gamma_1 = 0, \quad \gamma_2 = 1, \quad \gamma_3 = 0.$$

More generally...

- Linearized DSGE leads to linear rational expectations (LRE) system.
- Sims (2002) provides solution algorithm for canonical form

$$\Gamma_0(\theta)s_t = \Gamma_1(\theta)s_{t-1} + \Psi\epsilon_t + \Pi\eta_t$$

where

- s_t is a vector of model variables, ϵ_t is a vector of exogenous shocks,
 - η_t is a vector of RE errors with elements $\eta_t^x = \hat{x}_t - \mathbb{E}_{t-1}[\hat{x}_t]$, and
 - s_t contains (among others) the conditional expectation terms $\mathbb{E}_t[\tilde{x}_{t+1}]$.
- Overall the solution in terms of s_t is of the form

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t.$$

- Other solution methods for LREs: Blanchard and Kahn (1980), King and Watson (1998), Uhlig (1999), Anderson (2000), Klein (2000), Christiano (2002).

- Assumption: $\psi_2 = 1/\beta$, $\hat{g}_t = 0$.
- Eliminate nominal interest rate from the consumption Euler equation using policy rule

$$\hat{x}_t = \mathbb{E}_{t+1}[\hat{x}_{t+1}] - \left(\frac{1}{\beta} \hat{\pi}_t + \sigma_R \epsilon_{R,t} - \mathbb{E}[\hat{\pi}_{t+1}] \right) + \mathbb{E}_t[z_{t+1}].$$

- Rewrite NKPC:

$$\frac{1}{\beta} \hat{\pi}_t - \mathbb{E}_t[\hat{\pi}_{t+1}] = \frac{\kappa_p}{\beta} ((1 + \nu)\hat{x}_t + \phi_t + \lambda_t).$$

Substitute NKPC into consumption Euler equation:

$$\hat{x}_t = \psi_p \mathbb{E}_t[\hat{x}_{t+1}] - \frac{\kappa_p \psi_p}{\beta} (\phi_t + \lambda_t) + \psi_p \mathbb{E}_t[z_{t+1}] - \psi_p \sigma_{R \in R, t},$$

where $0 \leq \psi_p \leq 1$ is given by

$$\psi_p = \left(1 + \frac{\kappa_p}{\beta} (1 + \nu) \right)^{-1}.$$

Solving our LRE Model – Output

- **Recall:**

$$\hat{x}_t = \psi_p \mathbb{E}_t[\hat{x}_{t+1}] - \frac{\kappa_p \psi_p}{\beta} (\phi_t + \lambda_t) + \psi_p \mathbb{E}_t[Z_{t+1}] - \psi_p \sigma_R \epsilon_{R,t},$$

- We now need to find a law of motion for output (and, equivalently, consumption) of the form

$$\hat{x}_t = \hat{x}(\phi_t, \lambda_t, Z_t, \epsilon_{R,t}) = x_\phi \phi_t + x_\lambda \lambda_t + x_Z Z_t + x_{\epsilon_R} \epsilon_{R,t}$$

- that solves the functional equation:

$$\begin{aligned} 0 = & \mathbb{E}_t \left[\hat{x}(\phi_t, \lambda_t, Z_t, \epsilon_{R,t}) \right. \\ & - \psi_p \hat{x}(\rho_\phi \phi_t + \sigma_\phi \epsilon_{\phi,t+1}, \rho_\lambda \lambda_t + \sigma_\lambda \epsilon_{\lambda,t+1}, \rho_Z Z_t + \sigma_Z \epsilon_{Z,t+1}, \epsilon_{R,t+1}) \\ & \left. + \frac{\kappa_p \psi_p}{\beta} (\phi_t + \lambda_t) - \psi_p Z_{t+1} + \psi_p \sigma_R \epsilon_{R,t} \right]. \end{aligned}$$

- Decision rule for output:

$$\hat{x}_t = \hat{x}(\phi_t, \lambda_t, z_t, \epsilon_{R,t}) = x_\phi \phi_t + x_\lambda \lambda_t + x_z z_t + x_{\epsilon_R} \epsilon_{R,t}$$

- where

$$x_\phi = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\phi}, \quad x_\lambda = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\lambda}, \quad x_z = \frac{\rho_z \psi_p}{1 - \psi_p \rho_z} z_t, \quad x_{\epsilon_R} = -\psi_p \sigma_R.$$

- **Recall:** $\widehat{lsh}_t = (1 + \nu)\widehat{x}_t + \phi_t$.

- Deduce

$$\widehat{lsh}_t = [1 + (1 + \nu)x_\phi]\phi_t + (1 + \nu)x_\lambda\lambda_t + (1 + \nu)x_z z_t + (1 + \nu)x_{\epsilon_R}\epsilon_{R,t}.$$

The NKPC yields the following functional equation:

$$0 = \mathbb{E}_t \left[\widehat{\pi}(\phi_t, \lambda_t, z_t, \epsilon_{R,t}) - \beta \widehat{\pi}(\rho_\phi \phi_t + \sigma_\phi \epsilon_{\phi,t+1}, \rho_\lambda \lambda_t + \sigma_\lambda \epsilon_{\lambda,t+1}, \rho_z z_t + \sigma_z \epsilon_{z,t+1}, \epsilon_{R,t+1}) - \kappa_p \widehat{ish}(\phi_t, \lambda_t, z_t, \epsilon_{R,t}) - \kappa_p \lambda_t \right],$$

where $\widehat{ish}(\cdot)$ was given on previous slide.

The solution takes the form

$$\begin{aligned} \widehat{\pi}_t = & \frac{\kappa_p}{1 - \beta \rho_\phi} [1 + (1 + \nu) x_\phi] \phi_t + \frac{\kappa_p}{1 - \beta \rho_\lambda} [1 + (1 + \nu) x_\lambda] \lambda_t \\ & + \frac{\kappa_p}{(1 - \beta \rho_z)} (1 + \nu) x_z z_t + \kappa_p (1 + \nu) x_{\epsilon_R} \epsilon_{R,t}. \end{aligned}$$

Combining the decision rule for inflation with the monetary policy rule yields

$$\begin{aligned}\widehat{R}_t &= \frac{\kappa_p/\beta}{1 - \beta\rho_\phi} [1 + (1 + \nu)x_\phi] \phi_t + \frac{\kappa_p/\beta}{1 - \beta\rho_\lambda} [1 + (1 + \nu)x_\lambda] \lambda_t \\ &\quad + \frac{\kappa_p/\beta}{1 - \beta\rho_z} (1 + \nu)x_z z_t + [\kappa_p(1 + \nu)x_{\epsilon_R}/\beta + \sigma_R] \epsilon_{R,t}.\end{aligned}$$

- To confront the model with data, one has to account for the presence of the model-implied stochastic trend in aggregate output and to add the steady states to all model variables.
- **Measurement equations:**

$$\log(X_t/X_{t-1}) = \hat{x}_t - \hat{x}_{t-1} + z_t + \log \gamma$$

$$\log(lsh_t) = \widehat{lsh}_t + \log(lsh)$$

$$\log \pi_t = \hat{\pi}_t + \log \pi^*$$

$$\log R_t = \widehat{R}_t + \log(\pi^* \gamma / \beta).$$

State-space Representation of DSGE Model

- $n_y \times 1$ vector of observables:

$$y_t = M'_y [\log(X_t/X_{t-1}), \log lsh_t, \log \pi_t, \log R_t]'$$

- $n_s \times 1$ vector of econometric state variables s_t

$$s_t = [\phi_t, \lambda_t, z_t, \epsilon_{R,t}, \hat{x}_{t-1}]'$$

- DSGE model parameters:

$$\theta = [\beta, \gamma, \lambda, \pi^*, \zeta_p, \nu, \rho_\phi, \rho_\lambda, \rho_z, \sigma_\phi, \sigma_\lambda, \sigma_z, \sigma_R]'$$

- Measurement equation:

$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t.$$

- State-transition equation:

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t, \quad \epsilon_t = [\epsilon_{\phi,t}, \epsilon_{\lambda,t}, \epsilon_{z,t}, \epsilon_{R,t}]'$$

State-Space Representation of DSGE Model

State-space representation:

$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t$$

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t$$

System matrices:

$$\Psi_0(\theta) = M'_y \begin{bmatrix} \log \gamma \\ \log(lsh) \\ \log \pi^* \\ \log(\pi^* \gamma / \beta) \end{bmatrix}, \quad x_\phi = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\phi}, \quad x_\lambda = -\frac{\kappa_p \psi_p / \beta}{1 - \psi_p \rho_\lambda}, \quad x_z = \frac{\rho_z \psi_p}{1 - \psi_p \rho_z}, \quad x_{\epsilon_R} = -\psi_p \sigma_R$$

$$\Psi_1(\theta) = M'_y \begin{bmatrix} x_\phi & x_\lambda & x_z + 1 & x_{\epsilon_R} & -1 \\ 1 + (1 + \nu)x_\phi & (1 + \nu)x_\lambda & (1 + \nu)x_z & (1 + \nu)x_{\epsilon_R} & 0 \\ \frac{\kappa_p}{1 - \beta \rho_\phi} (1 + (1 + \nu)x_\phi) & \frac{\kappa_p}{1 - \beta \rho_\lambda} (1 + (1 + \nu)x_\lambda) & \frac{\kappa_p}{1 - \beta \rho_z} (1 + \nu)x_z & +\kappa_p (1 + \nu)x_{\epsilon_R} & 0 \\ \frac{\kappa_p / \beta}{1 - \beta \rho_\phi} (1 + (1 + \nu)x_\phi) & \frac{\kappa_p / \beta}{1 - \beta \rho_\lambda} (1 + (1 + \nu)x_\lambda) & \frac{\kappa_p / \beta}{1 - \beta \rho_z} (1 + \nu)x_z & (\kappa_p (1 + \nu)x_{\epsilon_R} / \beta + \sigma_R) & 0 \end{bmatrix}$$

$$\Phi_1(\theta) = \begin{bmatrix} \rho_\phi & 0 & 0 & 0 & 0 \\ 0 & \rho_\lambda & 0 & 0 & 0 \\ 0 & 0 & \rho_z & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ x_\phi & x_\lambda & x_z & x_{\epsilon_R} & 0 \end{bmatrix}, \quad \Phi_\epsilon(\theta) = \begin{bmatrix} \sigma_\phi & 0 & 0 & 0 \\ 0 & \sigma_\lambda & 0 & 0 \\ 0 & 0 & \sigma_z & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

M'_y is an $n_y \times 4$ selection matrix that selects rows of Ψ_0 and Ψ_1 .

- We want to understand the implications of the DSGE model.
- We could simulate data from the state-space representation

$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t$$

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t$$

- using:

$$\epsilon_t \sim iidN(0, I).$$

- But some calculations are better done analytically.

- What is the correlation between consumption growth this quarter and four quarters ago?
- What is the correlation between inflation and interest rates?
- Does the labor share predict consumption growth one-year ahead?

DSGE Model Implications: Fix Parameters

Parameter	Value	Parameter	Value
β	1/1.01	γ	$\exp(0.005)$
λ	0.15	π^*	$\exp(0.005)$
ζ_p	0.65	ν	0
ρ_ϕ	0.94	ρ_λ	0.88
ρ_z	0.13		
σ_ϕ	0.01	σ_λ	0.01
σ_z	0.01	σ_R	0.01

DSGE Model Implications: Autocovariances

- State-space representation:

$$\begin{aligned}y_t &= \Psi_0(\theta) + \Psi_1(\theta)s_t \\s_t &= \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t\end{aligned}$$

- Notation:

$$\Gamma_{yy}(h) = \mathbb{E}[y_t y_{t-h}'], \quad \Gamma_{ss}(h) = \mathbb{E}[s_t s_{t-h}'], \quad \text{and} \quad \Gamma_{ys}(h) = \mathbb{E}[y_t s_{t-h}']$$

- Covariance matrix of s_t is solution to Lyapunov equation:

$$\Gamma_{ss}(0) = \Phi_1 \Gamma_{ss}(0) \Phi_1' + \Phi_\epsilon \Phi_\epsilon'$$

- Autocovariance matrices for $h \neq 0$:

$$\Gamma_{ss}(h) = \Phi_1^h \Gamma_{ss}(0)$$

- Using the measurement equation, we deduce that

$$\Gamma_{yy}(h) = \Psi_1 \Gamma_{ss}(h) \Psi_1', \quad \Gamma_{ys}(h) = \Psi_1 \Gamma_{ss}(h)$$

DSGE Model Implications: Autocovariances

- ① Fix a set of DSGE model parameters.
- ② Solve model and compute matrices $\Psi_0(\theta)$, $\Psi_1(\theta)$, $\Phi_1(\theta)$, $\Phi_\epsilon(\theta)$ in state-space representation:

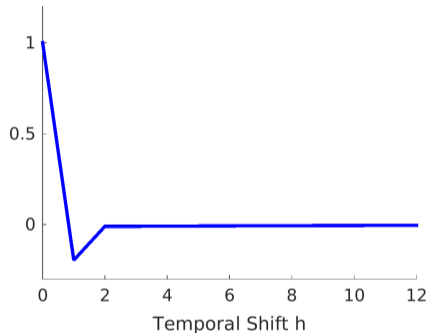
$$y_t = \Psi_0(\theta) + \Psi_1(\theta)s_t$$

$$s_t = \Phi_1(\theta)s_{t-1} + \Phi_\epsilon(\theta)\epsilon_t$$

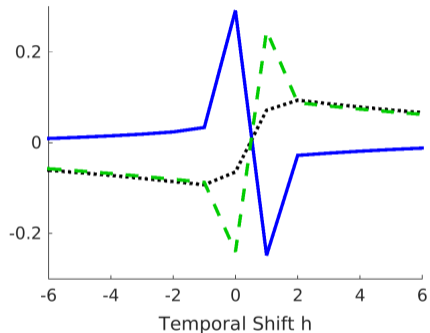
- ③ Compute autocovariances based on $\Psi_0(\theta)$, $\Psi_1(\theta)$, $\Phi_1(\theta)$, $\Phi_\epsilon(\theta)$.

Compute Autocovariances and Plot

$$\text{Corr}(\log(X_t/X_{t-1}), \log(X_{t-h}/X_{t-h-1}))$$



$$\text{Corr}(\log(X_t/X_{t-1}), \log Z_{t-h})$$



Notes: Right panel: correlations of output growth with labor share (solid), inflation (dotted), and interest rates (dashed).

- Fluctuations in the model are driven by shocks:
 - technology growth z_t ;
 - mark-up λ_t ;
 - preference ϕ_t ;
 - monetary policy $\epsilon_{R,t}$;
 - government spending \widehat{g}_t .
- Shocks generate uncertainty about future macroeconomic outcomes
- Question: what is the contribution of monetary policy shocks to forecast errors in inflation rates?

DSGE Model Implications: Forecast Error Variance Decomposition

- The law of motion for s_t can be expressed as the infinite-order vector moving average (MA)

$$y_t = \Psi_0 + \Psi_1 \sum_{s=0}^{\infty} \Phi_1^s \Phi_{\epsilon} \epsilon_{t-s}.$$

- h -step-ahead forecast error is

$$e_{t|t-h} = y_t - \mathbb{E}_{t-h}[y_t] = \Psi_1 \sum_{s=0}^{h-1} \Phi_1^s \Phi_{\epsilon} \epsilon_{t-s}.$$

- h -step-ahead forecast error covariance matrix is

$$\mathbb{E}[e_{t|t-h} e'_{t|t-h}] = \Psi_1 \left(\sum_{s=0}^{h-1} \Phi_1^s \Phi_{\epsilon} \Phi'_{\epsilon} \Phi_1^{s'} \right) \Psi_1' \quad \text{with} \quad \lim_{h \rightarrow \infty} \mathbb{E}[e_{t|t-h} e'_{t|t-h}] = \Gamma_{ss}(0).$$

DSGE Model Implications: Forecast Error Variance Decomposition

- Recall $\mathbb{E}[\epsilon_t \epsilon_t'] = I$. Let $I^{(j)}$ be defined by setting all but the j -th diagonal element of the identity matrix I to zero:

$$I = \sum_{j=1}^{n_\epsilon} I^{(j)}.$$

- Express the contribution of shock j to the forecast error for y_t as

$$e_{t|t-h}^{(j)} = \Psi_1 \sum_{s=0}^{h-1} \Phi_1^s \Phi_\epsilon I^{(j)} \epsilon_{t-s}.$$

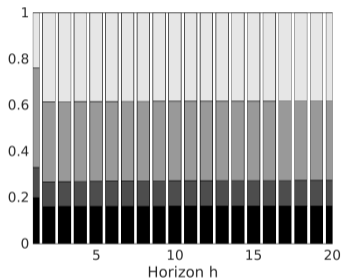
- The contribution of shock j to the forecast error variance of observation $y_{i,t}$ is

$$\text{FEVD}(i, j, h) = \frac{\left[\Psi_1 \left(\sum_{s=0}^{h-1} \Phi_1^s \Phi_\epsilon I^{(j)} \Phi_\epsilon' \Phi_1^{s'} \right) \Psi_1' \right]_{ii}}{\left[\Psi_1 \left(\sum_{s=0}^{h-1} \Phi_1^s \Phi_\epsilon \Phi_\epsilon' \Phi_1^{s'} \right) \Psi_1' \right]_{ii}},$$

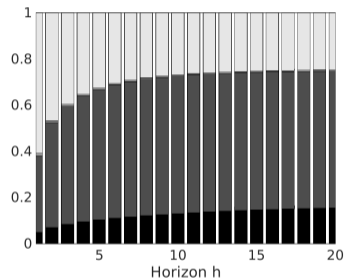
where $[A]_{ij}$ denotes element (i, j) of a matrix A .

Compute Forecast Error Variance Decomposition and Plot

Output Growth $\log(X_t/X_{t-1})$



Labor Share $\log lsh_t$



Notes: The stacked bar plots represent the cumulative forecast error variance decomposition. The bars, from darkest to lightest, represent the contributions of ϕ_t , λ_t , z_t , and $\epsilon_{R,t}$.

DSGE Model Implications: Impulse Response Functions

- What is the dynamic effect of a 25 basis point unanticipated reduction in the interest rate?
- What is the effect of an unanticipated increase in technology growth?

DSGE Model Implications: Impulse Response Functions

- **Definition:**

$$\text{IRF}(i, j, h | s_{t-1}) = \mathbb{E}[y_{i,t+h} | s_{t-1}, \epsilon_{j,t} = 1] - \mathbb{E}[y_{i,t+h} | s_{t-1}].$$

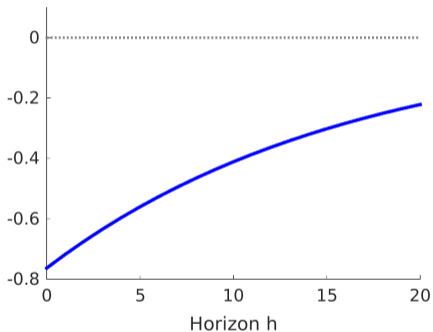
- Both expectations are conditional on the initial state s_{t-1} and integrate over current and future realizations of the shocks ϵ_t .
- First term also conditions on $\epsilon_{j,t} = 1$, whereas the second term averages of $\epsilon_{j,t}$.
- In a linearized DSGE model we have $\mathbb{E}[\epsilon_{t+h} | s_{t-1}] = 0$ for $h = 0, 1, \dots$ and deduce

$$\text{IRF}(., j, h) = \Psi_1 \frac{\partial}{\partial \epsilon_{j,t}} s_{t+h} = \Psi_1 \Phi_1^h [\Phi_\epsilon]_{.j},$$

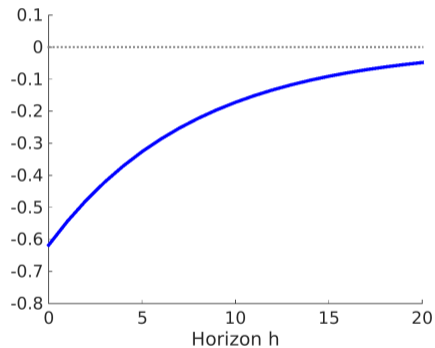
where $[A]_{.j}$ is the j -th column of a matrix A . We dropped s_{t-1} from the conditioning set to simplify the notation.

Compute and Plot Impulse Responses of Log Output $100 \log(X_{t+h}/X_t)$

Preference Innov. $\epsilon_{\phi,t}$

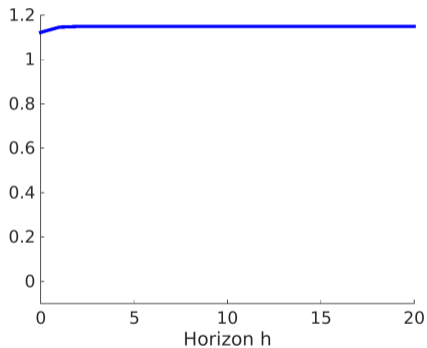


Mark-Up Innov $\epsilon_{\lambda,t}$

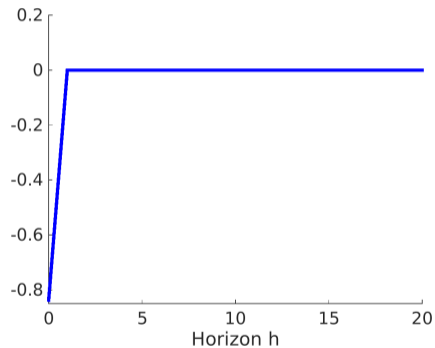


Compute and Plot Impulse Responses of Log Output $100 \log(X_{t+h}/X_t)$

Techn. Growth Innov. $\epsilon_{z,t}$



Monetary Policy Innov. $\epsilon_{R,t}$



DSGE Model Implications

- Formulas for autocovariance functions, spectra, and impulse response functions for a linearized DSGE model can be derived analytically.
- Analytical expressions can then be numerically evaluated for different vectors of parameter values θ .
- For general nonlinear general nonlinear DSGE model, the implied moments have to be computed using Monte Carlo simulation.
- Let $Y_{1:T}^*$ denote a sequence of observations simulated from the state-space representation of the DSGE model by drawing an initial state vector s_0 and innovations ϵ_t from their model-implied distributions, then

$$\frac{1}{T} \sum_{t=1}^T y_t^* \xrightarrow{a.s.} \mathbb{E}[y_t].$$

- The downside of Monte Carlo approximations is that they are associated with a simulation error.