

# Part I: Solving Linear Rational Expectations Models

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# Introduction

- ▶ Why linear solution

- ▶ transparent mechanism
- ▶ computationally efficient
- ▶ facilitate likelihood-based inference

- ▶ Main references

- ▶ Blanchard & Kahn (1980), "*The Solution of Linear Difference Models under Rational Expectations*", *Econometrica*
- ▶ Klein (2000), "*Using the Generalized Schur Form to Solve a Multivariate Linear Rational Expectations Model*", JEDC
- ▶ Sims (2001), "*Solving Linear Rational Expectations Models*", *Computational Economics*

# The Road Ahead...

- ▶ DSGE modeling
  - ▶ small new Keynesian DSGE
  - ▶ FRB-NY medium DSGE
  - ▶ log-linear approximation
- ▶ Sims' (2001) method
  - ▶ MATLAB program
  - ▶ analytical example (see notes)
- ▶ Impulse response functions

# Small New Keynesian DSGE

- Dynamic IS relation

$$\hat{y}_t = \mathbb{E}_t \hat{y}_{t+1} + \hat{g}_t - \mathbb{E}_t \hat{g}_{t+1} - \frac{1}{\tau} (\hat{R}_t - \mathbb{E}_t \hat{\pi}_{t+1} - \mathbb{E}_t \hat{z}_{t+1})$$

- New Keynesian Phillips curve

$$\hat{\pi}_t = \beta \mathbb{E}_t \hat{\pi}_{t+1} + \kappa (\hat{y}_t - \hat{g}_t)$$

- Monetary policy rule

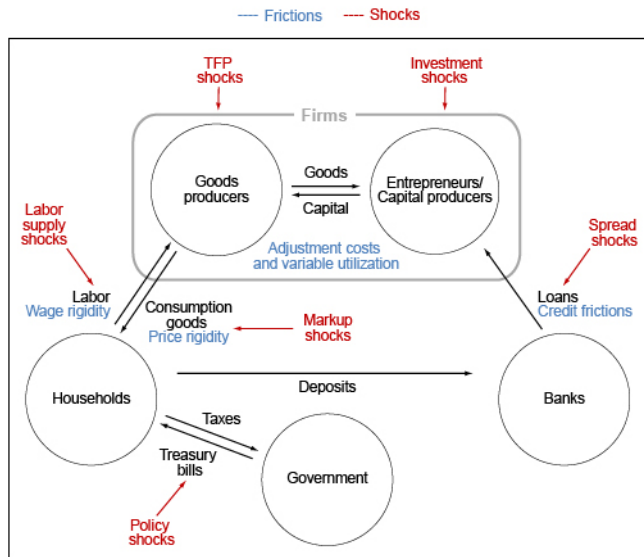
$$\hat{R}_t = \rho_R \hat{R}_{t-1} + (1 - \rho_R) \psi_1 \hat{\pi}_t + (1 - \rho_R) \psi_2 (\hat{y}_t - \hat{g}_t) + \epsilon_{R,t}$$

- Exogenous driving processes

$$\hat{g}_t = \rho_G \hat{g}_{t-1} + \epsilon_{G,t}, \quad \hat{z}_t = \rho_Z \hat{z}_{t-1} + \epsilon_{Z,t}$$

# FRB-NY Medium DSGE

## A Stylized Description of the Model



# Log-Linear Approximation

## 1st-order Taylor expansion

$$f(x_t) \equiv g(\hat{x}_t) = \underbrace{f(\bar{x})}_{g(0)} + \underbrace{f'(\bar{x})\bar{x}}_{g'(0)} \hat{x}_t + o(||\hat{x}_t||)$$

- ▶ Notations

- ▶  $\bar{x}$ : steady state of  $x_t$
- ▶  $\hat{x}_t \equiv \ln x_t - \ln \bar{x}$ : log deviation

- ▶ Why log-linearization

- ▶ level-linearization in economic models is not always well defined or consistent with data
- ▶ interpretation as % change:  $\hat{x}_t \approx (x_t - \bar{x})/\bar{x}$

# Sims' (2001) Method

## Canonical form

$$\Gamma_0(\theta)y_t = \Gamma_1(\theta)y_{t-1} + C(\theta) + \Psi(\theta)z_t + \Pi(\theta)\eta_t$$

### ► Notations

- $\theta$ : structural/deep parameters
- $y_t$ : endogenous variables;  $z_t$ : exogenous shocks
- $\eta_t \equiv y_t - \mathbb{E}_{t-1}y_t$ : expectational errors
- $\Gamma_0, \Gamma_1, C, \Psi, \Pi$ : coefficient matrices

### ► MATLAB program: [simsp.princeton.edu/yftp/gensys](https://simsp.princeton.edu/yftp/gensys)

```
>>[G,C,M,fmat,fwt,ywt,gev,eu] = gensys(G0,G1,  
    CC,Psi,Pi);
```

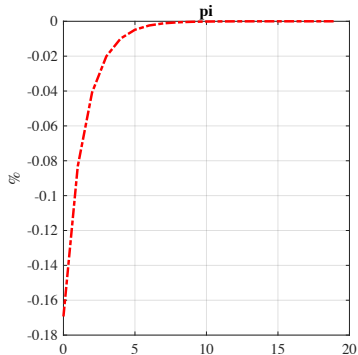
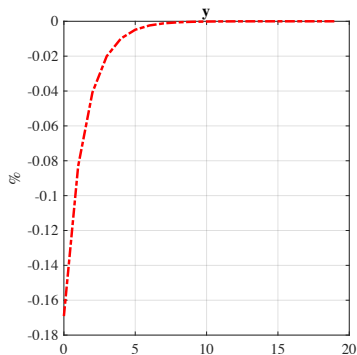
$$\Rightarrow y_t = G y_{t-1} + CC + M z_t + \sum_{k=1}^{\infty} (ywt)(fmat)^{s-1}(fwt) \mathbb{E}_t z_{t+k}$$

# MATLAB Pseudo-Code

```
function [C,G,M,eu] = SolveModel(para,ssp,P,V)
[...]  
% Input equations  
j = j+1;           % new Keynesian Phillips curve  
G0(j,V.model.pi) = 1;  
G0(j,V.model.E_pi) = -ssp(P.beta);  
G0(j,V.model.y) = -para(P.kappa);  
G0(j,V.model.g) = para(P.kappa);  
j = j+1;           % trend shock  
G0(j,V.model.z) = 1;  
G1(j,V.model.z) = para(P.rho_Z);  
Psi(j,V.shock.eps_Z) = 1;  
j = j+1;           % E_pi error  
G0(j,V.model.pi) = 1;  
G1(j,V.model.E_pi) = 1;  
Pi(j,V.fore.pi) = 1;  
% Solve model  
[G,C,M,~,~,~,eu] = gensys(G0,G1,CC,Psi,Pi);
```



# Impulse Response Functions



- Response to one standard deviation monetary shock

$$y_{t+k} - \mathbb{E}_t y_{t+k} = \sum_{j=0}^{k-1} G^j M z_{t+k-j}$$