

# Session : DSGE models and rational expectations

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# **Dynamic Stochastic General Equilibrium**

# Plan of part III of the course “Rational Expectation and DSGE models” - *Lectures 4,5,6,7*

## Lecture 4 (This lecture)

- Introduction to DSGE models (examples, linearizing)
- Solving a DSGE model (i.e. solving for unobserved expectations)

## Lecture 5-6 (HLB) Estimation of DSGE models

## Lecture 7 (AP) Simulation and use of DSGE models with Dynare.

# Plan

Rational expectations

DSGE models

Solving DSGE models

Wrapping-up

# Rational expectation models

Rational expectation models (e.g. DSGE) differ from

- Traditional structural models, "Cowles Commission"-type (taught by P.O. Beffy in Lecture 1 and 2)
- Structural VAR models (Lecture 3)

**Common point:** dynamic time-series models

**Key differences:**

- RE models make expectations explicit
- RE models are more structural than "structural VARs" (which impose only few restrictions)
- RE models are in general derived from micro foundations.

**Proeminent examples: DSGE models**

NB: expectations are (most often) non-observed variables.

→ *rational* expectations; help circumvent this

# Expectations in macro models

- Expectations are everywhere in macro!
- Consumption-saving decision:  
 $U'(c_t) = \beta(1 + r_t)E_t U'(c_{t+1})$  ( $c_t$  consumption,  $r_t$  real interest rate)
- asset yields; in the risk-neutral case:  $r_t = \frac{E_t p_{t+1} - p_t}{p_t} + \frac{d_t}{p_t}$   
( $r_t$  riskless rate,  $d_t$  dividend,  $p_t$  stock price)
- money demand (Cagan)  $\ln(M_t/P_t) = -\alpha E_t(\Delta P_{t+1}/P_t) + \varepsilon_t$

# The rational expectation assumption

**Notation:**  $E_t Y_{t+1}$  expectation of  $Y_{t+1}$  formulated at date  $t$

**Rational Expectations :**

$$E_t Y_{t+1} = E(Y_{t+1}|I_t),$$

$E(.|I_t)$  mathematical expectation conditional on  $I_t$ , information set available at date  $t$ ,

# Rational Expectation Hypothesis: a discussion

- Justifications:

- RE expectations are "model-consistent" : no reason to assume economic agents have less knowledge than the econometrician about the structure of the model
- Lucas critique (Lucas, 1976) : traditional reduced form models are non invariant to an economic policy intervention

- Remarks:

- RE is a strong assumption
- rational expectations  $\neq$  perfect expectations
- RE can be tested (direct tests, using expectations data like surveys – or indirect tests)



## Some alternatives to rational expectations

- *Adaptatives* expectations

$$E_t Y_{t+1} = \lambda Y_t + (1 - \lambda) E_{t-1} Y_t,$$

- *Naive expectations*, an extreme particular case:  $E_t Y_{t+1} = Y_t$  (case  $\lambda = 1$ ).
- Many others: rational learning, limited rationality...

## Some alternatives to rational expectations (continued)

Expectations data that can be used for direct tests of the REH  $E_t Y_{t+1}$  Such data are either directly *observed* or built.

### Examples:

- quantitative or qualitative answers to business survey (in France surveys of consumers or firms by Insee)
- expectations data inferred from financial market data (e.g. inflation swap)

### Some limits:

- assumptions underlying quantification
- directly observed expectations are most often forecast of some specific agents (e.g. professional forecasters like OECD experts or Consensus Forecast)
- in practice, expectations are most often unserved

**These alternatives are not developed in this lecture**

$$E_t Y_{t+1} = E(Y_{t+1}|I_t),$$

We will assume for  $E(\bullet|I_t)$  that agents know the model structure and past data when forming expectations.

Expectations will be *internally consistent*

Rational expectations

DSGE models

- A simple RBC model

- The simplest neo-Keynesian model

- (large) DSGE models in public administrations and international institutions

- Linearizing

Solving DSGE models

Wrapping-up

# This course is not about deriving DSGE models

For detailed material on the derivation of such models, see

- 2A *Macroéconomie 2: fluctuations* with Franck Malherbet
- 3A Monetary Economics with Olivier Loisel
- 3A Structural macroeconomics with Edouard Challe
- appendix slides for this class from last year by Benoît Campagne (Smets and Wouters, 2003)

# This course focuses on linear models

- linear models ... or linearized around the stationary equilibrium
- Alternative to linearization:
  - value function iteration,
  - second-order (or higher order) approximations

These approaches will not be detailed here.

See (Canova, 2011), (DeJong and Dave, 2011), (Juillard and Ocktan, 2008).

NB: in some cases you may lose relevant information by linearizing (Lindé and Trabandt, 2018)

# What is this course about?

This course will show in the next session how linearized DSGE models can be put under the form of a restricted SVAR.

Before that we will describe DSGE models as macroeconomic tools.

## DSGE models

### A simple RBC model

The simplest neo-Keynesian model

(large) DSGE models in public administrations and international institutions

Linearizing



## A simple RBC model

Keynes-Ramsey's rules:  $\frac{1}{c_t} = \tilde{\beta} E_t \left[ \frac{1}{c_{t+1}(1 + r_{t+1})} \right]$

Hours worked:  $H_t = 1 - \phi \frac{c_t}{w_t}$

Production function:  $y_t = A_t k_t^\alpha H_t^{1-\alpha}$

Real interest rate:  $r_t = \alpha \frac{y_t}{k_t} - \delta$

Real wage:  $w_t = (1 - \alpha) \frac{y_t}{H_t}$

Investment:  $i_t = \gamma k_{t+1} - (1\delta)k_t$

Market clearing:  $y_t = c_t + i_t$

Productivity shock:  $\log(A_t) = \eta \log(A_{t-1} + (1 - \eta) \log(\bar{A})) + \varepsilon_t$

cf. Franck Malherbet's class on *Fluctuations* (or King and Rebelo, 1999)

## DSGE models

- A simple RBC model

- The simplest neo-Keynesian model**

- (large) DSGE models in public administrations and international institutions

- Linearizing

# The simplest neo-Keynesian model

$$\tilde{y}_t = E_t(\tilde{y}_{t+1}) - \frac{1}{\sigma}(i_t - E_t(\pi_{t+1}) - r_t^n) \quad (1)$$

$$\pi_t = \beta E_t(\pi_{t+1}) + \kappa \tilde{y}_t \quad (2)$$

$$i_t = \phi_p \pi_t + \phi_y \tilde{y}_t + \varepsilon_t^i \quad (3)$$

cf. Olivier Loisel's class on *Monetary economics* (or [Jordi Galí \(2015\)](#). *Monetary policy, inflation, and the business cycle: an introduction to the new Keynesian framework and its applications*. Chap. 3)

There is a bit of a *dogmatic* controversy between both types of models...

### Don't get involved

Keep a **scientific** approach: the model does or does not fit the data, channel X or Y is or is not important in explaining macro fluctuations...

You should not care whether Prof. W or Z had the right intuition decades ago!

Anyway, the *truth* will (dis)agree partially with both so be **nuanced**.

## DSGE models

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- Linearizing

# The Smets-Wouters/Christiano Eichenbaum and Evans core

The most popular mid-size model is (Smets and Wouters, 2003; Christiano, Eichenbaum, and Evans, 2005) (see Campagne's annex or Challe's class).

- Closed economy
- Capital and labour
- Nominal rigidities (monetary policy has a role) and real rigidities
- Exogenous fiscal policy
- 12 equations - 10 shocks - 7 observables

## DSGE in policy institutions

DSGE models are very popular in central banks and other economic policy institutions. The most popular mid-size model is Smets-Wouters/CEE (see Campagne's annex or Challe's class).

- IMF-GIMF (Kumhof et al., 2010)
- IMF-GEM (Bayoumi et al., 2004)
- European Commission-[Quest](#) (Ratto, Roeger, and Veld, 2009)
- ECB-NAWM (Warne, Coenen, and Christoffel, 2008)
- ECB-EAGLE (Gomes, Jacquinet, and Pisani, 2012)
- Fed-Sigma (Erceg, Guerrieri, and Gust, 2006)
- DG Trésor-Omega3 (Carton and Guyon, 2007)
- Insee-Mélèze (Campagne and Poissonnier, 2016)
- ...

There are libraries of models [macromodelbase.com](http://macromodelbase.com) and private collections ([J. Pfeifer](#))

# Main blocks in a DSGE model

## ■ Firms

- Production function
- Balancing factors of production
- Set prices / marginal cost
- Banking sector (optional)

## ■ Households

- Consume
- Save-Invest
- Supply labour
- Negotiate wages / marginal productivity
- Heterogenous agents (optional)

## ■ Policy (fiscal and monetary, optional)

## ■ *Laws of nature*

- Market clearing
- Capital dynamics
- Other countries and trade (optional)



What would you want to include in a DSGE model?

## A general form

A generalized DSGE model (non linear)

$$E_t f(y_{t+1}, y_t, y_{t-1}, v_t) = 0 \quad (4)$$

with ... endogenous variables, ... exogenous variables, and  $f$  containing ... equations for *ldots* endogenous variables.

In general only a subset  $y_{t+1}^+$  ( $y_{t-1}^-$ ) of the ..... variables are forward (reps. backward) looking.

The endogenous variables need to be solved for, as a function of the *predetermined variables* and the shocks:  $y_t = g(\dots, \dots)$  with  $g$  the ".....".

## DSGE models

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- Linearizing

# Why linear?

The simplest RBC and neo-Keynesian model are "naturally" linear when taken in logs.

For large DSGE models this cannot be true.

We perform a first order Taylor development around a **Steady State**.

You need to think twice about trends and the definition of equilibrium in your model.

## (Standard) Notations

$X_t$  a model variable,  $\bar{X}$  its steady state,  $\hat{X}_t = \frac{X_t - \bar{X}}{\bar{X}}$  the deviation from the steady state.

In a linearized model you assume that  $\hat{X} \ll 1$  (e.g. 5% at most) so that  $\hat{X}^2 \simeq 0$  (is negligible).

$\hat{X}_t = \frac{X_t - \bar{X}}{\bar{X}} \simeq \log(X_t) - \log(\bar{X})$  (but I recommend that you do not use the log definition of  $\hat{X}$ , cf. *pen& paper*). You will see why in a minute.

$$Y = f(X) \implies \hat{Y} = \frac{f'(\bar{X})\bar{X}}{\bar{Y}} \hat{X}$$

Let's take 2 simple equations:

$$Y_t = K_t^\alpha (A_t L_t)^{1-\alpha} \quad (5)$$

and the growth rate

$$dY_t = \frac{Y_t - Y_{t-1}}{Y_{t-1}} \approx \log \left( \frac{Y_t}{Y_{t-1}} \right) \quad (6)$$

Let's assume that  $A_t = A_0 e^{at}$  is a deterministic trend. We denote  $y_t = Y_t/A_t$  and  $k_t = K_t/A_t$  the stationary component of output and capital.  $L_t$  is stationary.

Compute a log-linearisation of the 2 equations above

- with  $\hat{y}_t = \log(y_t/\bar{y})$
- with  $\hat{y}_t = \frac{y_t - \bar{y}}{\bar{y}}$

The first equation gives the same result in both cases:

$$y_t = k_t^\alpha L_t^{1-\alpha} \quad ; \quad \bar{y} = \bar{k}^\alpha \bar{L}^{1-\alpha} \quad ; \quad \hat{y}_t = \alpha \hat{k}_t + (1 - \alpha) \hat{L}_t \quad (7)$$

The second does not:

$$dY_t \approx \log \left( \frac{Y_t}{Y_{t-1}} \right) = \log(y_t) - \log(y_{t-1}) + a = \hat{y}_t - \hat{y}_{t-1} + a \quad (8)$$

$$\begin{aligned} dY_t &= \frac{Y_t - Y_{t-1}}{Y_{t-1}} = \frac{y_t e^a - y_{t-1}}{y_{t-1}} = \frac{e^a(1 + \hat{y}_t)}{(1 + \hat{y}_{t-1})} - 1 \\ &\approx e^a(1 + \hat{y}_t - \hat{y}_{t-1}) - 1 = e^a(\hat{y}_t - \hat{y}_{t-1}) + (e^a - 1) \end{aligned} \quad (9)$$

The generalized DSGE model (non linear)

$$E_t f(y_{t+1}, y_t, y_{t-1}, v_t) = 0 \quad (10)$$

can then be turned into a matrix expression (linear)

$$AY_t = BE_t Y_{t+1} + CY_{t-1} + DV_t \quad (11)$$

see (Villemot, 2011) in the Dynare documentation to understand how it's done automatically.

Note for the future, linearized model makes an implicit certainty equivalence assumption.



# Plan

Rational expectations

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# Solving rational expectation models

- **A benchmark (simple) example:**

$$y_t = \beta E_t y_{t+1} + \gamma x_t + \varepsilon_t$$

- "Forward" solution :

$$y_t = \beta E_t (\beta E_{t+1} y_{t+2} + \gamma x_{t+1} + \varepsilon_{t+1}) + \gamma x_t + \varepsilon_t$$

$$y_t = \gamma \sum_{i=0}^T \beta^i E_t x_{t+i} + \beta^{T+1} E_t y_{t+T+1} + \varepsilon_t$$

- The "fundamental" solution (in the case  $\beta < 1$ ) :

$$y_t = \gamma \sum_{i=0}^{\infty} \beta^i E_t x_{t+i} + \varepsilon_t$$

- An interpretation: Discounted Present Value

## Computing a closed form

Let's assume the model gives us a backward dynamic for  $x_t$   
e.g. AR(1)  $x_t = \rho x_{t-1} + v_t$

**Property:**

$$E_t x_{t+i} = \rho^i x_t$$

**Implied reduced form:**

$$y_t = \gamma \frac{1}{(1 - \beta\rho)} x_t + \varepsilon_t$$

## Model specification

$$y_t = bE_t y_{t+1} + a y_{t-1} + c x_t + \varepsilon_t$$

$$x_t = \rho x_{t-1} + u_t$$

## Examples :

- new “hybrid” Phillips curve;
- consumption Euler equation with habit formation in utility,
- models with adjustment costs,...

# Example of a resolution method:

the undetermined coefficient approach

**"Guess"** the form of the solution:

$$y_t = \varphi y_{t-1} + \alpha x_t + \tilde{\varepsilon}_t$$

This implies

$$\begin{aligned} E_t y_{t+1} &= \varphi y_t + \alpha E_t x_{t+1} = \varphi(\varphi y_{t-1} + \alpha x_t + \tilde{\varepsilon}_t) + \alpha \rho x_t \\ &= \varphi^2 y_{t-1} + \alpha(\varphi + \rho)x_t + \varphi \tilde{\varepsilon}_t \end{aligned}$$

Substitute this in  $y_t = bE_t y_{t+1} + ay_{t-1} + cx_t + \varepsilon_t$

This yields a second degree equation for  $\varphi$ :

$$\varphi = b\varphi^2 + a$$

And  $\alpha = c + b\alpha(\varphi + \rho)$

We choose the root such that  $\varphi < 1$

It results that  $\varphi = \frac{1 - \sqrt{1 - 4ba}}{2b}$ , and  $\alpha = \frac{c}{1 - b(\varphi + \rho)}$

$$y_t = \left( \frac{1 - \sqrt{1 - 4ba}}{2b} \right) y_{t-1} + \left( \frac{c}{1 - b(\varphi + \rho)} \right) x_t + \left( \frac{1}{1 - b\varphi} \right) \varepsilon_t$$

**Shortcoming:** this approach is not systematic

## The multivariate case

- A general formulation for a multivariate model:

$$\begin{aligned}AY_t &= BE_t Y_{t+1} + CY_{t-1} + DX_t \\ X_t &= \Phi X_{t-1} + V_t\end{aligned}$$

where  $A, B, C, D, \Phi$  are matrices,  $Y_t$ , a vector of endogenous variables,  $X_t$ , a vector of exogenous variables (e.g. shocks),  $V_t$  i.i.d. innovation

- Other general formulations are possible:

$$\begin{aligned}AY_t &= BE_t Y_{t+1} + DX_t \\ X_t &= \Phi X_{t-1} + V_t\end{aligned}$$

The two formulations are equivalent up to a redefinition of vector  $Y_t$ .

- In general, no analytical solution is available

**Various procedures that produce numerical solutions:**

- (Uhlig, 1995): undetermined coefficient approach
- Using matrices decompositions (Jordan, Schur, QZ...) to provide forward solutions: (Blanchard and Kahn, 1980), (Klein, 2000), (Sims, 2002)



## Sketch of the undetermined coefficient approach (Uhlig, 1995).

- Canonical formulation of a multivariate model

$$FE_t Y_{t+1} + GY_t + HY_{t-1} + MX_t = 0$$

$$X_t = NX_{t-1} + V_t$$

where  $F, G, H, M, N$  are matrices,  $X_t$  is a vector of endogenous variables

**"Guess" the solution:**

$$Y_t = PY_{t-1} + QX_t$$

$P$  and  $Q$  fulfill:

$$FP^2 + GP + H = 0 \quad (12)$$

$$(FQ + L)N + (FP + G)Q + M = 0 \quad (13)$$

→ Numerical solution of quadratic equation (12)

# The Blanchard and Kahn (1980) method for the Multivariate case

- Formulation of the model

$$E_t Y_{t+1} = AY_t + CX_t$$

$$X_t = \Phi X_{t-1} + V_t$$

→ Distinguish in  $Y_t$  between variables

$y_{1t}$  predetermined ( $n_1$ ) and

$y_{2t}$  non-predetermined ( $\equiv$  no initial condition) .

→  $y_{1t}$  with size  $n_1$ ,  $y_{2t}$  with size  $n_2$

## Note: unicity/stability of the solution

### Blanchard and Kahn (1980) Conditions :

In the above model:

Let  $n$  the number of eigenvalues of matrix  $A$  that are larger than 1

If  $n =$  number of non-predetermined variables (ie  $n_2$ ) : then the stable solution is unique (saddle-path solution)

If  $n > n_2$  ( number of non-predetermined variables) : instability (unit-root, explosive dynamics for the variables)

If  $n < n_2$  number of non-predetermined variables : multiplicity of solutions

## Example: the simple model

$$y_t = \beta E_t y_{t+1} + \gamma x_t + \varepsilon_t$$

$$x_t = \rho x_{t-1} + v_t$$

with  $x_t$  predetermined variable,  $y_t$  non predetermined variable,  
 $Y_t = (x_t, y_t)'$

$$E_t Y_{t+1} = A Y_t + V_t$$

with  $A = \begin{bmatrix} \rho & 0 \\ -\gamma/\beta & 1/\beta \end{bmatrix}$   
Eigenvalues of  $A$  are  $\rho, (1/\beta)$ .

Unicity and stability if  $-1 < \rho < 1, -1 < \beta < 1$

If  $\beta > 1$ , "rational stationary bubbles" are possible (multiplicity of solutions)

Example of a bubble:  $b_t = (1/\beta)b_{t-1} + e_t$

# Blanchard and Kahn (1980) method

- Formulation of the model

$$E_t Y_{t+1} = AY_t + CX_t$$

$$X_t = \Phi X_{t-1} + V_t$$

- We seek a solution that has the form :

$$y_{2t} = Hy_{1t} + NX_t$$

$$y_{1t} = My_{1t-1} + LX_{t-1}$$

## Blanchard and Kahn (1980) method

The system writes:

$$\begin{bmatrix} y_{1t+1} \\ E_t y_{2t+1} \end{bmatrix} = A \begin{bmatrix} y_{1t} \\ y_{2t} \end{bmatrix} + C X_t$$

Perform a Jordan decomposition of  $A$  :

$$A = \Lambda^{-1} J \Lambda$$

$J$  is a bloc-diagonal matrix containing the eigenvalues of  $A$

$$J = \begin{bmatrix} J_1 & 0 \\ 0 & J_2 \end{bmatrix}$$

where  $J_1$  collects eigenvalues smaller than 1

and  $J_2$  collects eigenvalues larger than 1

Note  $J$  is diagonal if eigenvalues are distinct,

otherwise  $J$  contains both "0"'s and "1"'s on the line above the diagonal.

## Implementing Blanchard and Kahn (1980)

The system can be re written as:

$$\Lambda \begin{bmatrix} y_{1t+1} \\ E_t y_{2t+1} \end{bmatrix} = J \Lambda \begin{bmatrix} y_{1t} \\ y_{2t} \end{bmatrix} + \Lambda \begin{bmatrix} C_1 \\ C_2 \end{bmatrix} X_t$$

We define auxiliary variables

$$\begin{bmatrix} \tilde{y}_{1t} \\ \tilde{y}_{2t} \end{bmatrix} = \begin{bmatrix} \Lambda_{11} & \Lambda_{12} \\ \Lambda_{21} & \Lambda_{22} \end{bmatrix} \begin{bmatrix} y_{1t} \\ y_{2t} \end{bmatrix}$$

Then we can write a “decoupled” system.

$$\begin{bmatrix} \tilde{y}_{1t+1} \\ E_t \tilde{y}_{2t+1} \end{bmatrix} = \begin{bmatrix} J_1 & 0 \\ 0 & J_2 \end{bmatrix} \begin{bmatrix} \tilde{y}_{1t} \\ \tilde{y}_{2t} \end{bmatrix} + \begin{bmatrix} D_1 \\ D_2 \end{bmatrix} X_t$$



The sub-system associated with eigenvalues larger than one can be solved forward :

$$E_t \tilde{y}_{2t+1} = J_2 \tilde{y}_{2t} + D_2 X_t$$

hence

$$\tilde{y}_{2t} = J_2^{-1} E_t \tilde{y}_{2t+1} - J_2^{-1} D_2 X_t$$

Solving:

$$\tilde{y}_{2t} = - \sum_{k=0}^{\infty} J_2^{-1-k} D_2 E_t X_{t+k}$$

Using the properties of the forcing process  $X_t$  :  
 $\text{VAR}(1) \ E_t X_{t+k} = \Phi^k X_t$

$$\tilde{y}_{2t} = - \sum_{k=0}^{\infty} J_2^{-1-k} D_2 \Phi^k X_t$$

An explicit form for this sum (using properties of the Kronecker product):

$$\tilde{y}_{2t} = -J_2^{-1}(I - \Phi' \otimes J_2^{-1})D_2 X_t$$

Expression as a function of the predetermined variables

$$y_{2t} = \Lambda_{22}^{-1} \tilde{y}_{2t} - \Lambda_{22}^{-1} \Lambda_{21} y_{1t}$$

So

$$y_{2t} = -\Lambda_{22}^{-1} J_2^{-1} (I - \Phi' \otimes J_2^{-1}) D_2 X_t - \Lambda_{22}^{-1} \Lambda_{21} y_{1t}$$

This expression is indeed of the form

$$y_{2t} = H y_{1t} + N X_t$$

## **Last step (!)**

Solving for predetermined variables  $y_{1t}$

$$y_{1t+1} = A_{11}y_{1t} + A_{12}y_{2t} + D_1X_t$$

where  $A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$

Replacing  $y_{2t}$  with its solved form, the process for  $y_{1t}$  is indeed of the form:

$$y_{1t} = My_{1t-1} + LX_{t-1}$$

# Extensions and alternatives to Blanchard-Kahn method

- (Sims, 2002), (Klein, 2000): use of Schur and QZ decomposition

$$BE_t Y_{t+1} = AY_t + CX_t$$

$$X_t = \Phi X_{t-1} + V_t$$

→ In a general model  $B$  may be non invertible  
(if  $B$  invertible, back to B-K case with  $\tilde{A} = B^{-1}A$ )

→ Sims method does not require specification of  
predetermined/non-predetermined variables

→ The various procedures are available in the form of Matlab,  
Gauss,... routines

→ The Sims/Klein method is implemented in the **Dynare** toolbox

- (Anderson and Moore, 1985) : multiple leads and lags of endogenous variable

$$Y_t = \sum_{j=1}^J A_j E_t Y_{t+j} + \sum_{i=1}^I B_i Y_{t-i} + V_t$$

AIM Algorithm (Federal Reserve) with Matlab, GAUSS

- DYNARE allows the user to write multiple leads and lags

# Plan

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## Bottom line of this lecture

- Start with a non linear model

$$E_t f(y_{t+1}, y_t, y_{t-1}, v_t) = 0$$

- Linearize it into

$$AY_t = BE_t Y_{t+1} + CY_{t-1} + DV_t$$

- Obtain the reduced form

$$Y_t = MY_{t-1} + D\eta_t$$

You obtain a **constrained VAR model** • In practice done by computer routines (eg Dynare)

- From there you can do the same as with a SVAR
- + you can derive normative results



# What are A,B,C,D numerically?

## Calibration

- Associated to RBC models.
- Main principle of calibration: set values of parameters (including shocks standard deviations) Then compare model predictions with second moments of the data
- Cf. DeJong and Dave (2011) chap.6
- This approach is not strictly speaking an econometric one.

It can be viewed as a particular case, and less formalized version of these approaches: Minimum Distance Estimation, Bayesian Approach






**Next Lectures with HLB, you will discuss the estimation of such models**

# Prepare for the next Lectures

- Lecture 5-6: have a look at (Christiano, Eichenbaum, and Evans, 2005)
- Lecture 7: Install [Dynare](#) + Matlab/Octave (if not already done)

try the examples provided by the dynare team

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## An alternative solution method : Factorisation

Rely on "Forward " operator:

$$Fy_t = E_ty_{t+1}$$

Benchmark equation writes:

$$y_t - bFy_t - aLy_t = cx_t + \varepsilon_t$$

hence

$$P(F)Ly_t = -(1/b)(cx_t + \varepsilon_t)$$

where  $P(F) = F^2 - (1/b)F + (a/b)$

$P(F)$  can be factored  $P(F) = (F - \varphi_1)(F - \varphi_2)$  with  $\varphi_1 < 1$

Dividing by polynomial  $(F - \varphi_2)$

→ Solving "forward" for the root  $\varphi_2 > 1$

$$(1 - \varphi_1 L)y_t = \frac{1}{b\varphi_2} \frac{1}{1 - (1/\varphi_2)F} (\beta x_t + \varepsilon_t)$$

where  $\varphi_1 = \frac{1 - \sqrt{1 - 4ba}}{2b}$  (note  $\varphi_1 = \varphi$ , same as the undetermined coefficient approach)

and

$$y_t = \varphi_1 y_{t-1} - \frac{1}{b\varphi_2} \sum_{k=0}^{\infty} (1/\varphi_2)^k E_t(cx_{t+k} + \varepsilon_{t+k})$$

Using  $E_t(x_{t+k}) = \rho^h x_t$ , we find (as in the UC approach):

$$\alpha = \frac{c}{1 - b(\varphi_1 + \rho)}$$