DSGE-Models

Linearization and Solution Methods

Dr. Andrea Beccarini Willi Mutschler, M.Sc.

Institute of Econometrics and Economic Statistics
University of Münster
willi.mutschler@uni-muenster.de

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Previously...

- Theory and intuition behind the Smets/Wouters' model as a prototype of current DSGE-models.
- Derivation of the structural form and log-linearization.

Insight

A DSGE-model consists of a set of expected, nonlinear optimality conditions and transition equations for stochastic processes, which one has to solve.

• A DSGE-model can in general be written as:

General form

$$\Gamma(E_t x_{t+1}, x_t, v_{t+1} | \mu) = \Gamma(x_{t+1}, x_t, v_{t+1}, \eta_{t+1} | \mu) = 0,$$
 (1)

with $\mathbf{x_t}$: $(n \times 1)$ -vector of stationary variables, v_t : $(m \times 1)$ -vector of structural shocks, μ : $(k \times 1)$ -vector of parameters.

General Form

- Rational Expectations: $\eta_{t+1} = E_t x_{t+1} x_{t+1}$.
- Non-predictable expectation error η_{t+1} occurs due to the realization of structural shocks: $\eta_t = f(v_t)$.
- $\mathbf{x_t}$ consist of n_c control variables and n_s state variables.
- Control variables are denoted by c_t: optimal behavior of the agents as
 a function of the current state of the economy.
- ullet State variables are denoted by $ullet_t$. They consist of exogenous independent of the decisions of the agents and endogenous state variables, that can be influenced by the agents.
- **s**_t is a function of previous states and current shocks.

Solution of a DSGE-model

• To solve such a rational expectations model means to find so-called policy-functions c and s, that solve (at least approximately) the system of equations Γ :

Policy-functions

$$\mathbf{c_t} = c(\mathbf{s_t}), \qquad \mathbf{s_t} = s(\mathbf{s_{t-1}}, v_t).$$

- DSGE-models can be interpreted as state-space-models.
- One distinguishes between linear and non-linear methods:
 - Linear methods: Blanchard/Khan (1980), Binder and Pesaran (1997),
 Christiano (2002), King and Watson (1998), Klein (2000), Sims (2001) and Uhlig (1999).
 - Frequently used nonlinear method: Schmitt-Grohé/Uribe (2004). Excellent overview: Heer and Maussner (2009).

Repetition: Exercise 1

Assume the following stochastic growth-model:

$$\begin{split} \max_{c_t,k_{t+1}} E_0 \sum_{t=0}^\infty \beta^t \textit{U}(c_t) & \text{ with } \textit{U}(c_t) = \frac{c_t^{1-\sigma}-1}{1-\sigma}, \\ c_t + k_{t+1} &= z_t f(k_t) + (1-\delta)k_t & \text{ with } f(k_t) = k_t^\alpha, \ k_o \ \text{given}, \\ \textit{log}(z_t) &= \rho \textit{log}(z_{t-1}) + \varepsilon_t & \text{ with } \varepsilon_t \sim \textit{WN}(0,\sigma_\varepsilon), \ 0 < \rho < 1. \end{split}$$

- (a) What is the unconditional expectation of $log(z_t)$, what of z_t ? Find an expression for the unconditional variance of $log(z_t)$?
- (b) Derive the first-order-conditions (FOC) by setting up the Bellman-equation and using methods of dynamic programming. What are state and control variables?
- (c) Derive the first-order-conditions (FOC) by maximizing the lagrangian.
- (d) Derive the steady-state and linearize the FOC.

Linear Solution Methods

Pros:

- Simple linear state-space representation of the model, which in many cases is sufficiently exact.
- One can use the Kalman-filter to empirically evaluate the system.

Cons:

- One looses important information during the linearization.
- Higher moments play an important role for analyzing markets, risk, welfare, etc.
- An approximation to, say, the second order can yield different results, because the variance of future shocks matters (risk premium).

Certainty-equivalence-property

In stochastic rational expectations models agents take the effect of future shocks into account. For a linearization to the first-order these expectations are zero, thus, they don't matter for the decision rules.

Linear Solution Methods

- First linearize or log-linearize the general form (1) around the deterministic *steady-state*.
- Together with the transition equation of the stochastic processes one gets the reduced-form model:

$$\mathbf{A}\mathbf{x}_{t+1} = \mathbf{B}\mathbf{x}_t + \mathbf{C}\boldsymbol{\upsilon}_{t+1} + \mathbf{D}\boldsymbol{\eta}_{t+1} + \mathbf{E}.$$

- The matrices A, B, C and D are functions of the structural parameters μ ; E is a vector of constants (mostly zero).
- The solution of this linear representation has a VAR(1)-form:

$$\mathbf{x}_{t+1} = \mathbf{F}(\boldsymbol{\mu})\mathbf{x}_t + \mathbf{G}(\boldsymbol{\mu})\boldsymbol{v}_{t+1}, \tag{2}$$

with **F** and **G** being functions of the parameters μ .

- ullet The motor of the model is the vector of exogenous shocks $v_{
 m t}$.
- (2) describes thus, the fluctuations around the steady-state as well as the decision rules given the stochastic innovations.

Concepts

Notation

Predetermined variables: $E_t X_{1,t+1} = X_{1,t+1}$, Non-predetermined variables: $E_t X_{2,t+1} = E_t X_{2,t+1}$.

Unitary matrices

 $\mathbf{M}'\mathbf{M}=\mathbf{M}\mathbf{M}'=\mathbf{I}$ are the complex analogous to orthogonal matrices. They are diagonalizable.

 The method of Sims (2001) begins with a QZ-factorization (General Schur decomposition), in which the matrices A and B are transformed into unitary upper triangular matrices:

$$A = Q'\Lambda Z', \qquad B = Q'\Omega Z'.$$

• Λ and Ω are upper triangular matrices with the generalized Eigenvalues of A and B, and they are sorted in increasing order from left to right.

 The Eigenvalues determine if a system of equations converges or explodes.

Blanchard/Khan-conditions

The number of Eigenvalues, that are in absolute terms greater than 1, must be equal to the number of non-predetermined variables, in order to get a stable solution (saddle-path).

 Let z_{t+1} = Z'x_{t+1}, then the system can be divided into a non-explosive (upper) and an explosive part (lower):

$$\begin{bmatrix} \mathbf{\Lambda}_{11} & \mathbf{\Lambda}_{12} \\ \mathbf{0} & \mathbf{\Lambda}_{22} \end{bmatrix} \begin{pmatrix} \mathbf{z}_{1,t+1} \\ \mathbf{z}_{2,t+1} \end{pmatrix} = \begin{bmatrix} \mathbf{\Omega}_{11} & \mathbf{\Omega}_{12} \\ \mathbf{0} & \mathbf{\Omega}_{22} \end{bmatrix} \begin{pmatrix} \mathbf{z}_{1,t} \\ \mathbf{z}_{2,t} \end{pmatrix} + \begin{pmatrix} \mathbf{Q}_{1} \\ \mathbf{Q}_{2} \end{pmatrix} \begin{bmatrix} \mathbf{E}_{1} + \mathbf{C}_{1} v_{1,t+1} + \mathbf{D}_{1} \eta_{1,t+1} \\ \mathbf{E}_{2} + \mathbf{C}_{2} v_{2,t+1} + \mathbf{D}_{2} \eta_{2,t+1} \end{bmatrix}.$$
(3)

- The difference equations belonging to the Eigenvalues greater than 1 are solved forwards.
- Remark: $\lim_{t\to\infty} \left(\Omega_{22}^{-1} \mathbf{\Lambda}_{22}\right)^t \mathbf{z}_{2,\mathbf{t}} = \mathbf{0}$ and for all s>0: $E_t v_{2,\mathbf{t}+\mathbf{s}} = E_t \eta_{2,\mathbf{t}+\mathbf{s}} = 0$ (Expectations don't matter)

$$\begin{split} \mathbf{z}_{2,t} &= \Omega_{22}^{-1} \mathbf{\Lambda}_{22} \mathbf{z}_{2,t+1} - \Omega_{22}^{-1} \mathbf{Q}_2 \left[\mathbf{E}_2 + \mathbf{C}_2 \upsilon_{2,t+1} + \mathbf{D}_2 \eta_{2,t+1} \right] \\ &= -\sum_{i=0}^{\infty} \left(\Omega_{22}^{-1} \mathbf{\Lambda}_{22} \right)^i \Omega_{22}^{-1} \mathbf{Q}_2 \left[\mathbf{E}_2 + \mathbf{C}_2 \upsilon_{2,t+1+i} + \mathbf{D}_2 \eta_{2,t+1+i} \right] \\ &= -\sum_{i=0}^{\infty} \left(\Omega_{22}^{-1} \mathbf{\Lambda}_{22} \right)^i \Omega_{22}^{-1} \mathbf{Q}_2 \mathbf{E}_2 \\ &= -\left[\mathbf{I} - \Omega_{22}^{-1} \mathbf{\Lambda}_{22} \right]^{-1} \Omega_{22}^{-1} \mathbf{Q}_2 \mathbf{E}_2 = \left[\mathbf{\Lambda}_{22} - \Omega_{22} \right]^{-1} \mathbf{Q}_2 \mathbf{E}_2. \end{split}$$

- The difference equations belonging to the Eigenvalues less or equal than 1 are solved backwards.
- Remark: Systematical relationship betweend the expectation errors $\eta_{1,t} \& \eta_{2,t}$.

Sufficient condition for a stable saddle-path

$$\mathbf{Q_1D} = \mathbf{\Phi}\mathbf{Q_2D}.$$

- Φ is of dimension $n_s \times n_c$ (with $\mathbf{z_{1,t}} : n_s \times 1$ and $\mathbf{z_{2,t}} : n_c \times 1$).
- (3) can be rewritten as:

$$\underbrace{\begin{bmatrix} \mathbf{I} & -\mathbf{\Phi} \\ n_s \times n_s & n_s \times n_c \end{bmatrix}}_{n_s \times (n_s + n_c)} \begin{bmatrix} \mathbf{\Lambda}_{11} & \mathbf{\Lambda}_{12} \\ \mathbf{0} & \mathbf{\Lambda}_{22} \end{bmatrix} \begin{pmatrix} \mathbf{z}_{1,t} \\ \mathbf{z}_{2,t} \end{pmatrix} =$$

$$\begin{bmatrix} \textbf{I} & -\boldsymbol{\Phi} \end{bmatrix} \begin{bmatrix} \Omega_{11} & \Omega_{12} \\ 0 & \Omega_{22} \end{bmatrix} \begin{pmatrix} \textbf{z}_{1,t-1} \\ \textbf{z}_{2:t-1} \end{pmatrix} + \begin{bmatrix} \textbf{I} & -\boldsymbol{\Phi} \end{bmatrix} \begin{pmatrix} \textbf{Q}_1 \\ \textbf{Q}_2 \end{pmatrix} \begin{bmatrix} \textbf{E} + \textbf{C} \upsilon_t + \textbf{D} \eta_t \end{bmatrix}.$$

$$\begin{split} \Leftrightarrow \begin{bmatrix} \Lambda_{11} & \Lambda_{12} - \Phi \Lambda_{22} \end{bmatrix} \begin{pmatrix} z_{1,t} \\ z_{2,t} \end{pmatrix} &= \begin{bmatrix} \Omega_{11} & \Omega_{12} - \Phi \Omega_{22} \end{bmatrix} \begin{pmatrix} z_{1,t-1} \\ z_{2,t-1} \end{pmatrix} \\ &+ \begin{bmatrix} Q_1 - \Phi Q_2 \end{bmatrix} \begin{bmatrix} \mathsf{E} + \mathsf{C} \upsilon_t \end{bmatrix} + \underbrace{\begin{bmatrix} (Q_1 - \Phi Q_2) \, \mathsf{D} \eta_t \end{bmatrix}}_{=0}. \end{split}$$

The algorithm

QZ-factorization:
$$\mathbf{A} = \mathbf{Q}' \mathbf{\Lambda} \mathbf{Z}', \, \mathbf{B} = \mathbf{Q}' \mathbf{\Omega} \mathbf{Z}', \, \mathbf{x}_t = \mathbf{Z} \begin{pmatrix} \mathbf{z}_{1,t} \\ \mathbf{z}_{2,t} \end{pmatrix}$$

$$\begin{split} \textbf{z}_{1,t} &= -\left. \boldsymbol{\Lambda}_{11}^{-1} \left(\boldsymbol{\Lambda}_{12} - \boldsymbol{\Phi} \boldsymbol{\Lambda}_{22} \right) \textbf{z}_{2,t} + \boldsymbol{\Lambda}_{11}^{-1} \boldsymbol{\Omega}_{11} \textbf{z}_{1,t-1} \right. \\ &+ \left. \boldsymbol{\Lambda}_{11}^{-1} \left(\boldsymbol{\Omega}_{12} - \boldsymbol{\Phi} \boldsymbol{\Omega}_{22} \right) \textbf{z}_{2,t-1} + \boldsymbol{\Lambda}_{11}^{-1} \left(\boldsymbol{Q}_1 - \boldsymbol{\Phi} \boldsymbol{Q}_2 \right) (\textbf{E} + \textbf{C} \boldsymbol{\upsilon}_t) \,, \\ \textbf{z}_{2,t} &= \left(\boldsymbol{\Lambda}_{22} - \boldsymbol{\Omega}_{22} \right)^{-1} \boldsymbol{Q}_2 \textbf{E}. \end{split}$$

Exercise 1: continued

The Sims (2001)-algorithm is available for different program packages, and it is implemented in Dynare.

(e) Rewrite the linearized model into the form

$$\mathbf{A}\mathbf{x}_{\mathsf{t}+1} = \mathbf{B}\mathbf{x}_{\mathsf{t}} + \mathbf{C}\upsilon_{\mathsf{t}+1} + \mathbf{D}\eta_{\mathsf{t}+1} + \mathbf{E}$$

(f) Solve for the *policy-function* by using the Sims (2001)- Algorithm. Assume the following values for the parameters:

$$\beta = 0.99$$
, $\alpha = 0.36$, $\sigma = 2$, $\delta = 0.025$, $\rho = 0.9$.

Homework

- Install the newest version of Matlab.
- https://zivdav.uni-muenster.de/ddfs/Soft.ZIV/TheMathWorks/.
- Because of license issues you have to be connected to the university either via WLAN or VPN.
- Download all files from http://sims.princeton.edu/yftp/gensys/mfiles/ into a folder called gensys. Make this folder available to Matlab (File-SetPath-Add Folder).
- Install the newest version of Dynare (http://www.dynare.org/). Make the Dynare folder c: \dynare\aktuelle Version\matlab) available to Matlab (File-SetPath-Add Folder).