

# Introduction to Dynare and Examples

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# What is Dynare?

- Dynare is a Matlab front end to solve and simulate dynamic models
- Either deterministic or stochastic
- Developed by Michel Juillard at CEPREMAP
- website: <http://www.dynare.org/>

## What is Dynare?

- Dynare runs inside the computation platform MATLAB, or the open-source alternative Octave.
- Both Dynare and Octave are available for free download, the former running under MATLAB or Octave on all platforms and the latter in separate versions for Windows, Mac, or Linux.
- You will need to run your simulations on a computer on which Dynare and either MATLAB or Octave are installed (Read instruction in readme file to add Dynare to Matlab.)

## How does it work?

- 1 Input your model, parameter values, and instructors to Dynare through a text file with a .mod suffix.
  - The Quick Start guide, Tutorial, User Guide, and Manual available at the Dynare Web site
  - website: <http://www.dynare.org/> provide details.
- 2 Template for several .mod files so you will be able to run the code
- 3 Note that the .mod file must be in plain text format. You can use a simple text editor such as Notepad to create such a file.
  - If you instead use Word or another advanced word processor, you must set the file format to plain text when you save or Dynare won't be able to read your file.

## Structure of .mod file

The .mod file consists of five parts:

- The **preamble** section
  - defines your variables and parameters.
- The **model** section
  - contains the equations of your model in a simple algebraic notation similar to standard programming languages.
- The **steady state** section
  - gives the steady-state (initial) values of the model solution.
- The **shocks** section
  - in which you tell Dynare about the deterministic or random shocks that you want to simulate.
- The **computation** section
  - tells Dynare to simulate the model and report the results.

## How does it work?

Before we get into details, note several features of the Dynare program:

- Variables are case-sensitive.
- All commands end with a semi-colon.
- Blocks of commands such as model equations, steady-state (initial) values, and shocks end with the `end;` statement.
- Options for any command are entered in parentheses immediate after the name of the command.

# General Structure

- \* Preamble
  - Declaration of endogenous, exogenous variables and parameters
  - Assignment of parameter values
- \* Declaration of model
  - Start: model;
  - End: end;
  - In between: all equilibrium conditions
- \* Initial conditions
  - Start: initval
  - Provide initial conditions for steady state of model
  - End: end

# General Structure

- \* Specify shocks
  - Start: Shocks
  - Define (all non-zero entries of) Variance-Covariance matrix of shocks
  - End: end
  
- \* Solution of Model
  - Use command `stoch simul`
  - More on options later



# Timing Convention

- Pre-determined variables (e.g., capital stock) dated  $t-1$  in time  $t$  equation
- Way to tell Dynare which variables are state variables
- Need to rewrite set of equations
- Lag capital stock in all equations

## More on conventions

### \* Timing conventions

- If variable  $x$  is decided in period  $t$ , write  $x$
- If variable  $x$  is decided in period  $t - 1$ , write  $x(-1)$
- If variable  $x$  is decided in period  $t + 1$ , write  $x(+1)$
- If variable  $x$  is decided in period  $t + 2$ , introduce auxiliary variable

### \* Solutions

- Dynare default: linear approximation of levels of variables
- Linear approx. in logs convenient as IRFs are in percentage terms
- Define variables as  $\exp(x)$
- Now  $x$  will be interpreted as log of variable

## Structure of the mod file: Preamble

Assume your model takes the form

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

with  $e_t \sim N(0, \sigma^2)$

- Variable:  $x_t$
- Exogenous Variable:  $e_t$
- Parameters:  $\rho$  and  $\sigma^2$ 
  - If  $x_t$  captures technology,  $\rho$  is a parameter capturing the persistence of technological progress

## Structure of the mod file: Preamble

Define variables and parameters

3 major instructions:

- 1 var: Define variables
- 2 varexo: Define (truly) exogenous variables
- 3 parameters: Declare parameters
  - ... assign values to parameters

## Structure of the mod file: Preamble

### An example

```
// Simple AR(1) model
// This version: 01/16/17
var x;
varexo e;
parameters rho,sigma;
rho      = 0.90;
sigma    = 0.01;
```

## Structure of the mod file: Model

Define model equations

1 major instruction:

```
model;  
...  
end;
```

Write equations as they appear in *natural* language

# Structure of the mod file: Model

## An example

AR(1) example:  $x_t = \rho x_{t-1} + e_t$

Model writes:

```
model;  
x=rho*x(-1)+e;  
end;
```

## Structure of the mod file: Steady State

Compute the long-run of the model

- That is: Where its deterministic dynamics will converge
- Why? Because it will take a (non-)linear approximation around this long run

Structure:

```
initval;  
...  
end;  
steady;  
check;
```



## Structure of the mod file: Steady State

Steady computes the long run of the model using a non-linear solver

- Close to the Newton algorithm (more sophisticated though!)

It therefore needs initial conditions

- That's the role of the `initval; ... end; statement.`

Better give (very) good initial conditions for all variables

## Structure of the mod file: Steady State

What if you forget steady

- It will not compute the steady state.

What happens then?

- Simulations start from values specified in the `initval; ... end;` statement.

`check` is optional. It checks the dynamic stability of the system

## Structure of the mod file: Steady State

### An example

AR(1) example:  $x_t = \rho x_{t-1} + e_t$

In deterministic steady state:  $e_t = e^* = 0$

$$x^* = \rho x^* \implies x^* = 0$$

```
initval;  
e    = 0;  
x    = 0;  
end;  
steady;  
check;
```

## Structure of the mod file: Shocks

Define the properties of the exogenous shocks

- Exogenous shocks are Gaussian innovations with  $N(0, \Sigma)$

Structure:

```
shocks;  
var ...;  
stderr ...;  
end;
```

## Structure of the mod file: Shocks

AR(1) example:  $x_t = \rho x_{t-1} + e_t$

```
shocks;  
var e;  
stderr se;  
end;
```

## Structure of the mod file: Solution

Final step: Compute the solution and produce some output

Solution method

- Deterministic model: Relaxation method
- Stochastic model: First or Second order perturbation method

Then compute some moments and impulse responses.

```
stoch_simul(...) ...;
```

## Structure of the mod file: Solution

AR(1) example:  $x_t = \rho x_{t-1} + e_t$

Therefore (because the model is linear)

```
stoch_simul(linear);
```

# Structure of the mod file: Solution

## Options of the `stoch_simul` option

### Solver

- `linear`: In case of a linear model.
- `order = 1 or 2` : order of Taylor approximation (default = 2).

### Output (prints everything by default)

- `noprnt`: cancel any printing.
- `nocorr`: doesn't print the correlation matrix.
- `nofunctions`: doesn't print the approximated solution.
- `nomoments`: doesn't print moments of the endogenous variables.
- `ar = INTEGER`:  
Order of autocorrelation coefficients to compute (5)



## Structure of the mod file: Solution

### Impulse Response Functions

- `irf` = INTEGER: number of periods on which to compute the IRFs (Setting `IRF=0`, suppresses the plotting of IRFs).
- `relative irf` requests the computation of normalized IRFs in percentage of the standard error of each shock.

### Simulations

- `periods` = INTEGER: specifies the number of periods to use in simulations (default = 0).
- `replic` = INTEGER: number of simulated series used to compute the IRFs (default = 1 if order = 1, and 50 otherwise).
- `drop` = INTEGER: number of points dropped in simulations (default = 100).

```
// AR(1) model
// Name of the variable
var x;
// Name of the exogenous variable
varexo e;
// Parameters of the model
parameters rho se;
rho = 0.95;
se = 0.02;
model;
x = rho*x(-1)+e;
end;

initval;
e=0;
x=0;
end;
steady;
check;
shocks;
var e; stderr se;
end;
stoch_simul(linear);
```

## STEADY-STATE RESULTS:

$x = 0$

## EIGENVALUES:

Modulus	Real	Imaginary
0.95	0.95	0

There are 0 eigenvalue(s) larger than 1 in modulus  
for 0 forward-looking variable(s)

The rank condition is verified.

## MODEL SUMMARY

Number of variables: 1

Number of stochastic shocks: 1

Number of state variables: 1

Number of jumpers: 0

Number of static variables: 0

## MATRIX OF COVARIANCE OF EXOGENOUS SHOCKS

Variables	e
e	0.000400

## POLICY AND TRANSITION FUNCTIONS

	x
x(-1)	0.950000
e	1.000000

## THEORETICAL MOMENTS

VARIABLE	MEAN	STD. DEV.	VARIANCE
x	0.0000	0.0641	0.0041

## MATRIX OF CORRELATIONS

Variables	x
x	1.0000

## COEFFICIENTS OF AUTOCORRELATION

Order	1	2	3	4	5
x	0.9500	0.9025	0.8574	0.8145	0.7738

## Samuelson's accelerator model

Consider the Samuelson's accelerator model (backward looking model)

$$\begin{aligned}C_t &= \beta Y_{t-1} & G_t &= \rho G_{t-1} + (1 - \rho) \bar{G} + \eta_t \\ I_t &= \alpha (C_t - C_{t-1}) & Y_t &= C_t + I_t + G_t\end{aligned}$$

- Consumption is 80% of income
- Government spending have a strong persistence (0.9)
- $\bar{G} = 1$
- Investment in period  $t$  equals 5% more than the variation in consumption between  $t$  and  $t - 1$
- $\eta_t \sim iid N(0, 0.02^2)$ .

## Samuelson's accelerator model

- Variables:  $C_t, I_t, Y_t, G_t$
  - Exogenous Variable:  $\eta_t$
  - Parameters:  $\alpha, \beta, \rho, \sigma$
- A) Write the Dynare code to perform a stochastic simulation of this model for 2000 periods and generate impulse response functions for the endogenous variables for 60 periods.
- B) Present the matrix of correlation coefficients of the endogenous variables.
- C) (make sure the model is stable!) Try run the code with  $\beta = 2$

## Forward looking models

Consider a (linear) rational expectations model

- postpone theory to tomorrow!

The model writes

$$y_t = aE_t y_{t+1} + b x_t$$

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

with  $\varepsilon_t \sim N(0, \sigma^2)$

- Variable:  $y_t, x_t$
- Exogenous Variable:  $\varepsilon_t$
- Parameters:  $a, b, \rho$  and  $\sigma$

## Forward looking models

- $\varepsilon_t$  is exogenous
- $x_t$  is a predetermined variable
- $y_t$  is a jump variable (it's a control variable)

$$y_t = \sum_{j=0}^{\infty} a^j E_t x_{t+j}$$

- \* Blanchard and Kahn (Econometrica 1980)
- \* Fundamentally forward looking model!
- \* Dynare knows how to solve it!



## Backward-Forward looking models

Mix of jump and predetermined (endogenous) variables

The model writes

$$E_t y_{t+1} - (\lambda + \mu)y_t + \lambda \mu y_{t-1} = b x_t$$

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

with  $\varepsilon_t \sim N(0, \sigma^2)$

- Variable:  $y_t, x_t$
- Exogenous Variable:  $\varepsilon_t$
- Parameters:  $\lambda, \mu, b, \rho$  and  $\sigma$

## Backward-Forward looking models

- $\varepsilon_t$  is exogenous
  - $x_t$  is a predetermined variable
  - $y_t$  is a jump variable but it has also a predetermined component.
- \* Blanchard and Kahn (Econometrica 1980) again!

## Deterministic vs. Stochastic models

The important question: **is your model stochastic or deterministic?**

- The distinction hinges on whether future shocks are known.

**Deterministic models:** the occurrence of all future shocks is known exactly at the time of computing the model's solution.

**Stochastic models:** only the distribution of future shocks is known.

## Deterministic vs. Stochastic models

Consider a shock to a model's innovation only in period 1.

In a deterministic context

- Agents will take their decisions knowing that future values of the innovations *will be zero in all periods to come.*

In a stochastic context

- Agents will take their decisions knowing that the future value of innovations *are random but will have zero mean.*

The solution method for each of these model types differs significantly.

## Deterministic vs. Stochastic models

In deterministic models

- A highly accurate solution can be found by numerical methods.
- Solution: a series of numbers that match a given set of equations.

If an agent has perfect foresight

- Agent can specify today - at the time of her decision - what each of her precise actions will be in the future.

# Deterministic vs. Stochastic models

## Stochastic environment

- The best the agent can do is specify a decision, policy or feedback rule for the future:
  - What will her optimal actions be contingent on each possible realization of shocks.
- We search for a function satisfying the model's first order conditions.
  - This function may be non-linear and thus needs to be approximated.

## Deterministic models have the following characteristics:

1. DSGE literature has gained attention in economics, **deterministic models have become somewhat rare**. Examples include OLG models without aggregate uncertainty.
2. **Solution does not require linearization.**
  - It doesn't even really need a steady state.
  - Numerical simulation to find the exact paths of endogenous variables that meet the model's first order conditions and shock structure.
  - This solution method can therefore be useful when the economy is far away from steady state (when linearization offers a poor approximation).

## Deterministic models have the following characteristics:

3. These models are usually introduced to **study the impact of a change in regime** (i.e., introduction of a new tax).
4. **Full information, perfect foresight and no uncertainty around shocks.**
5. **Shocks** can hit the economy today or at any time in the future, in which case they **would be expected with perfect foresight.**
  - Shocks can also last one or several periods.
  - Most often models introduce a positive shock today and zero shocks thereafter (with certainty).



## Stochastic models have the following characteristics:

- 1 These types of models tend to be **more popular in the literature**. Examples include most RBC, or new Keynesian monetary models.
- 2 In these models, **shocks hit today (with a surprise)**, but thereafter their expected value is zero.
  - Expected future shocks, or permanent changes in the exogenous variables cannot be handled due to the use of Taylor approximations around a steady state.
- 3 When these models are linearized to the first order, **agents behave as if future shocks were equal to zero** (since their expectation is null), which is the certainty equivalence property.
  - This is an often overlooked point in the literature which misleads readers in supposing their models may be deterministic.