





ABM Macro Lab: Agent-based Modelling Tools

Session 01

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Outline

Introduction

Crash course on LSD

Logistic chaotic model

The industry model

Introduction

Objectives

Throughout the sessions, we will

- 1. Understand how LSD works.
- 2. Implement a chaotic model¹
- 3. Understand and implement the Dosi, Pereira, and Virgillito (2017) model on LSD
- 4. Analyze the model results on LSD

¹Based on the slides of LSD manual folder

Structure of the sessions

- Session 1 Presents the general structure of LSD and perform some simulations
- Session 2 Write the industry model equations in LSD
- Session 3 Complete the scripts (if necessary), run the simulation and the analysis of results
 - Bonus A primer on sensitivity analysis

Repository structure

This presentation can be found on this repository with another examples:

https://github.com/gpetrini/CodingSession

Sim*.lsd Are the configuration files

Important: Clone/Download this folder under LSD/Work/

Missed something?

*_complete.cpp contains the complete equation file for reference.

Crash course on LSD

C++ for LSD I

LSD adds macros to C++. As a consequence, some basic knowledge on C++ is usefull. What do we need to know here?

In python, R

```
x = 5 # A comment
y = 2.5
## We do not need to
## work with pointers
```

In C++

```
int x = 5; // A comment
double y = 2.5;
object *agent; // Latter
```

Compilation

Different from python and R, we need to compile our code before using it.

C++ for LSD II

To avoid initializing everything, LSD has some already initialized variables:

```
v[0] = 10; // We can assing number to
v[1] = 50; // a vector (always available)
T; // Current simulation time
i = 1; // i,h,j,k can be used for integers
cur; cur1; // Pre-allocated pointers
```

Those are local variables that we can use on equations.

Debbuging

LSD has an internal debbuger that allow us to look to local variables on the fly.

Macros I



Figure 1: General structure of most LSD macros

- VLS(POS, "NAME", 1) Returns the value of NAME at lag 1, position POS
- \bullet SUMS(POS, "NAME") Sums the value of variable NAME at position POS
- SEARCHS(PARENT, "AGT") Searches for agent AGT starting from PARENT
- WRITELS(POS, "NAME", 1) Overwrites NAME at lag 1 at position POS

Macros II

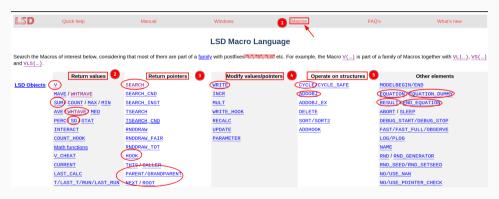


Figure 2: Other LSD macros

Macro example I

How can we write the following equation using LSD syntax?

$$X_t = X_{t-1} + m$$

```
EQUATION("X") // This is a single-line comment
/*
This is a multi-line comment
*/
v[0] = VL("X",1); //past value of X, lagged of 1 period
v[1] = V("m"); //current value of m (variable or parameter)
v[2] = v[0] + v[1]; // v[n] are local variables
RESULT(v[2]) // Specify the output of the function
```

Variable or parameter?

As a rule of thumb, variables have an EQUATION associated, parameters do not.

Macro example II

How can we write the following equation using LSD syntax?

$$Y_t = \sum_{i=1}^{n} X_{n,t} + W_{n,t-1}$$

```
EQUATION("Y")
v[0] = 0; // Initialize the Accumulation
CYCLE(cur, "Firm") { // Similar to a for-loop in other languages
    v[1] = VS(cur, "X"); // cur points to a "Firm" object
    v[2] = VLS(cur, "W", 1); // cur is a locally defined pointer
    v[3] = v[1] + v[2]:
   v[0] = v[0] + v[3]:
RESULT(v[0])
```

Macro example III

We can also write on an alternative way

$$Y_t = \sum_{i=1}^{n} X_{n,t} + W_{n,t-1}$$

```
EQUATION("Y")
// If X and W are bellow Y on the tree (later)
v[0] = SUM("X");
v[1] = SUML("W", 1);
v[2] = v[0] + v[1];
RESULT(v[2])
```

Equation file

Any equation file (.cpp) must contain the following text:

```
// File created using org-mode tangle feature.
#include "fun_head.h" // This is mandatory
MODEL BEGIN
// Your code goes here
MODELEND
void close sim(void) {}
```

In our session, we will copy-and-paste the initialization equation and continue from there.

Model structure and data organization I

Besides the equation files (.cpp or .h), LSD defines the model structures on a different file (extension .lsd). This special file has:

- Variables and parameters names
- Model structure (where elements are)
- Initial and parameter values
- Simulation settings
- Number of objects
- Variables to plot, analyze, debug, etc

LSD and OOP

This structure ensure the modeler to think in terms of data structure.

Logistic chaotic model

Start

Let's start with the file on Chaotic_Model/fun_chaos.cpp

```
#include "fun_head.h" // This is mandatory

MODELBEGIN

// Your code goes here

MODELEND
void close_sim(void) {}
```

Equation

Consider the model made of the single equation

$$X_t = m \cdot X_{t-1} \cdot (1 - X_{t-1})$$

To implement this model follow this steps:

- 1. Insert the equation's code for the model.
- 2. Compile and run the model (menu Model/Run).
- 3. Using the Lsd model program generate one object and place in it variable X with 1 lag and parameter m.

Equation on LSD

The model equation can be written as follows:

$$X_t = m \cdot X_{t-1} \cdot (1 - X_{t-1})$$

```
EQUATION("X")

v[0] = VL("X",1);

v[1] = V("m");

v[2] = v[1]*v[0]*(1-v[0]);

RESULT(v[2])
```

Configuring the model structure

- Compile (Model > Compile and run ...).
- Wait to open a new window.
- Create a descending object called Obj1 (Model > Add Object)
- Double click on Obj1
- Add the Variable X (max. lags = 1, $X_0 = 0.5$) and Parameter (m = 2)
- Save this configuration as Sim1.lsd
- Mark X to be saved (F5) and run
- Analyze the results

Logistic chaotic model IV

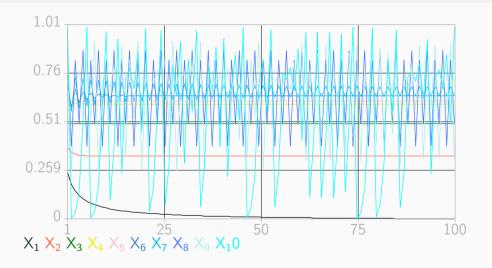
So far, there is nothing new if you use other programming language. However, we can leverage from the fact that LSD is OOP.

- Use menu Data/Set Number of Objects and set to 10 the copies of Obj1
- Select *m* and click on Data>Initial Values ... and set different values for each instance from (1...3.99)
- Run and plot X

Key takeaways

The source code remains untouched, while we produced a completelly different structure. This is the benefit of isolating the code and the structure.

Results I



Sensitivity to m

The function produces extremely different outcomes depending on the value of m.

- We will create a large number of series computed independently.
- We will set *m* to a slightly different value for each series.
- We will set the initial points for the X to random values.

Exploiting the chaotic behaviour

- Load Sim1.1sd and save it as Sim2.1sd
- Generate 10000 copies of Obj1
- Open the initial values interface with Data/Initial Values...
- Use the buttons Set All on the side of *m* and *X* to assign values to all the elements.
 - m: Range between 2.8 and 3.99.
 - X: Random (Uniform) between 0.01 and 0.99.
- Run>Simulation Settings... to set 1000 time steps.
- Mark *m* to be saved
- Save, run, analyze

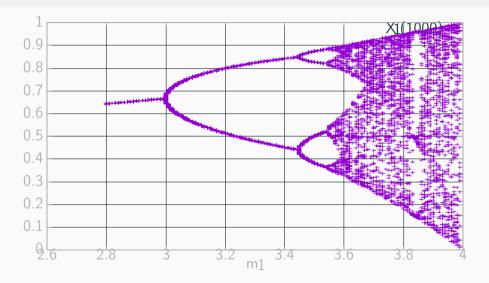
Plotting

- Open Analysis of results
- Select all *m* and *X* series
- On the bottom right, select Cross section and XY plot
- Select Points
- Plot for the last timestep (1000, default)
- Hit continue and wait

Key takeaways

Once again, the code remains untouched. Later, we will check other features of LSD that benefit from this design principle.

Results II



The industry model

Where are we and where are we going?

So far, we implemented a small non-economic model. On the following lectures, we will focus on the industry model.

Dosi, Pereira, and Virgillito (2017) design objective: simplest industrial dynamics model capturing most stylized facts:

- Persistent heterogeneity in productivity and all other performance variables
- Persistent market share and entry-exit turbulence
- Skewed size distributions
- Fat-tailed distribution of growth rates
- Scaling of the growth-variance relationship

Equations

Idiosyncratic learning process:	$a_{i,t} =$	$a_{i,t-1} \cdot (1 + \eta \cdot heta_{i,t})$
Learning shocks	$ heta_{i,t} \sim$	$Beta(eta_1,eta_2)$
Market selection	$s_{i,t} =$	$s_{i,t-1} \cdot \left(1 + A \cdot rac{a_{i,t} - ar{a}_t}{ar{a}_t} ight)$
Average productivity	$ar{a}_t =$	$\sum_{i=1}^{NF} s_{i,t-1} \cdot a_{i,t}$
Exit condition	$s_{i,t} <$	S _{min}
Entrant productivity	$a_{j,t} =$	$ar{a}_t \cdot (1 + \eta \cdot heta_{i,t})$
Entrant market-share	$s_{j,t} =$	1/NF
Market concentration index	$HHI_t =$	$\sum_{i=1}^{NF} (s_i)^2$
Market-share adjustment	$s_i \mapsto$	$s_i \cdot \frac{1}{\sum_{i=1}^{NF} s_i} \Rightarrow \sum_{i=1}^{NF} s_i = 1$
Fixed number of firms	$\#\{1,\ldots,n\} =$	NF

Graph of industry model

