# Treinamento LSD

Aulas 2-10 – 2ª Parte

Baseado nos slides de Marco Valente

# Agenda

- Aula 2: Linear model and LSD basics
- Aula 3: Random walk model and LSD model structure
- Aula 4: Logistic model and chaotic behavior
- Aula 5: Replicator dynamics model
- Aula 6: Extended replicator dynamics model
- Aula 7: Network externalities model Part 1
- Aula 8: Network externalities model Part 2
- Aula 9: Consumers' model Part 1
- Aula 10: Consumers' model Part 2

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The code refers to the elements by their labels, relying on the LSD system to associate, at run time, the actual values required in the equation's computation.

In the process the lesson will also provide tips on several aspects of L<sup>SD</sup> interfaces and internal working.

# LSD modular features

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The separation of data from the code greatly simplifies the writing and re-use of code:

- Avoid mistakes due to wrong referencing, e.g. of writing myvector[j,i] instead of myvector[i,j].
- Avoid adapting the same computational content to different model structures, such as replacing myvector[i,j] with myvector[i,j,h]

#### **Random Walk**

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In this lesson we consider different models sharing an identical computational content (i.e. equations) and will see how different model structures produces intuitively different configurations.

We consider a model representing a *random walk* expressed by two equations:

$$X_t = X_{t-1} + RE$$
  
 $RE = U(min, Max)$ 

where U(m, M) generates a (pseudo-)random values as if they were drawn from a uniform function in the range set by the parameters min and Max. The Uniform function, as many other functions, is provided by the system in the LMM (Use: Insert Lsd Script).

### Random Walk

#### Code for variable *X*.

```
EQUATION("X")
/*
A variable moving as a random walk
*/
v[0]=V("RE");
v[1]=VL("X",1);
v[2]=v[0]+v[1];
RESULT(v[2])
```

#### **Random Event**

#### Code for variable RE.

```
EQUATION("RE")
/*
A random event
*/
v[0]=V("min");
v[1]=V("Max");
v[3]=UNIFORM(v[0],v[1]);
RESULT(v[3])
```

As initial model structure let's consider the model that may be expressed formally as:

$$X_t = X_{t-1} + RE$$
  
 $RE_t = U(min, Max)$ 

That is, we define a model composed by a single object containing the two variables and the two parameters.

To express such a model structure use the L<sup>SD</sup> model browser generated with the equations. Then follow these steps:

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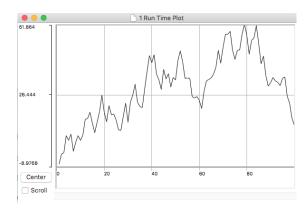
- o create an object, call it Obj1.
- 2 Move the Browser to show the content of the newly created object.
- Insert in the object all the elements: variables *X* (1 lag) and *RE* (0 lag); parameters *min* and *Max*.
- Using **Data/Initial Values** set  $X_0 = 0$  and min = -10 and Max = 10.

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- Insert in the object all the elements: variables *X* (1 lag) and *RE* (0 lag); parameters *min* and *Max*.
- Using **Data/Initial Values** set  $X_0 = 0$  and min = -10 and Max = 10.
- 5 Double-click on X and check on the option Run Time Plot.

Beware of using the same spelling in the model configuration as in the equations' code In case of errors correct the equations or the element in the structure. In the latter case double-click on the misspelled element, click on **Properties** and edit the label..

Running the simulation you will obtain *exactly* the following Run Time Plot window



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To replicate the simulation you cannot simply use  $\mathbf{Run}$  if you already executed a simulation. The reason is that the values stored in the model (e.g. for X) refer to the last time step of the simulation, which generally differ from those defined to start a simulation. Therefore to repeat a simulation run you need, firstly, the reload a fresh configuration (menu **File/Reload**), and then execute a new simulation (menu **Run/Run**).

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Before starting a simulation run, use menu **Run/Sim.Settings** and change the value in the entry labeled **Initial seed**. This will produce completely different random values.

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Using the standard formalism we may express the model as follows:

$$X_t^i = X_{t-1}^i + RE_t^i$$
  
 $RE_t^i = U(min^i, Max^i)$ 

where each element is indexed with i.

To generate the new model follow these steps:

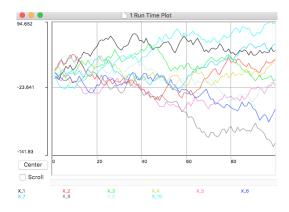
Load a fresh configuration of Sim1 and save it as Sim2, to avoid overwriting the first configuration.

To generate the new model follow these steps:

- Load a fresh configuration of Sim1 and save it as Sim2, to avoid overwriting the first configuration.
- Use menu Data/Set Number of Objects and set to 10 the copies of Obj1.

The new model will generate the following result.

#### Simulation with 10 copies.



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This is possible because the  $L^{SD}$  simulation manager automatically associate each copy of a variable (e.g. X) to a specific copy of the elements required for its computation (in our case RE).

The modeler needs not to specify the association of the necessary elements in the equation, since LSD automatically deduce that the "correct" elements to use are those in the same object of the variable computed.

Note that even using the same seed (random sequence) none of the 10 series will be identical to the result from the configuration using a single object.

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The reason is that the same series of random values will be spread over 10 series. For example, the second random value of the sequence is used, in Sim1, as the random event at time t=2, while in Sim2 it is used for the variable in the second object at time t=1.

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We can therefore consider the implementation of a model where the parameters are the same for all the different series, a model that in formal terms would be expressed as:

$$X_t^i = X_{t-1}^i + RE_t^i$$
  
 $RE_t^i = U(min, Max)$ 

where parameters *min* and *Max* are **not** indexed, meaning they are common to all the copies of *RE*.

To implement this configuration load a fresh configuration of Sim2 and save it as Sim3 and then makes the following operations.

Move the Browser to show the content of Obj1.

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- Oouble-click on parameter min. In the resulting window click on Properties. Use option Move to another Object and select Obj0 as object of destination.

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- Repeat the same operation for Max.

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Differently from the execution of Sim2, using Sim3 the system cannot find the parameters min and Max in the same object of the copy of RE under computation.

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Differently from the execution of Sim2, using Sim3 the system cannot find the parameters min and Max in the same object of the copy of RE under computation.

When this is the case, the LSD simulation manager automatically starts a search for the required element. The search starts from the object containing the variable under computation, and then moves, firstly, in its descendants (if any) and then its parent, exploring potentially the whole model structure until the element is found.

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As an example, consider the replication of Sim3 in two distinct settings. Formally, you may express the model as follows:

$$egin{array}{lll} egin{array}{lll} egin{array}{lll} X_t^{j,i} & = & X_{t-1}^{j,i} + RE_t^{j,i} \ RE_t^{j,i} & = & U(min^j, Max^j) \end{array}$$

where j refers to one of two different setting. In practice, the configuration needs to express two independent "branches" formed each by a single **Obj0** containing a set of objects **Obj1**.

To express this configuration you need not change the computational content (the equations) but only adapt the model structure:

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- Move the browser to show Obj0.
- Select Data/Initial Values. Assign to the parameters min and Max in the second copy the values -1 and 1, respectively.
- Double-click on variable X in the Browser and check on the option Save: save these series for later analysis.

We now have effectively two models as in Sim3 running in parallel. The equation for each *RE* will make use of the appropriate copy of parameters depending on the branch it is part of, thanks to the automatic LSD system of retrieving elements.

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To verify the results open the module **Data/Analysis of Results**. The series available are marked with two digits, referring respectively to the two layers of objects in the model.

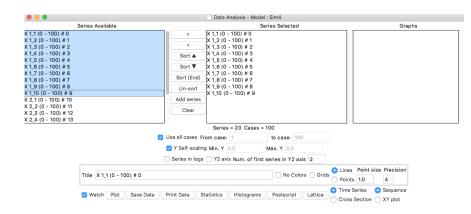
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To verify the results open the module **Data/Analysis of Results**. The series available are marked with two digits, referring respectively to the two layers of objects in the model.

Select the series separately from the two branches, and you will see that the same overall dynamics will extend over different ranges, as implied by the different values of parameters.

#### **Pseudo random numbers**

Analysis of result for parallel and independent settings.



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The model would then formally become:

$$X_t^{j,i} = X_{t-1}^{j,i} + RE^j$$
  
 $RE^j = U(min^j, Max^j)$ 

where RE lacks a time subscript indicating that it does not generate a value at each time step t, but needs to supply fresh values every time it is applied.

As in all the previous experiments we need not to modify the equations, but only manipulate the model structure.

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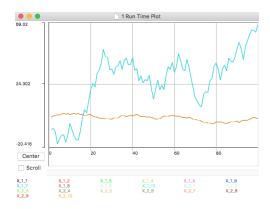
As in all the previous experiments we need not to modify the equations, but only manipulate the model structure.

- Load a fresh configuration Sim4 and save it as Sim5.
- 2 Move the Browser to show Obj1.
- Ouble-click on RE.
- Click on Properties. Move the element to Obj0.

The new configuration shares the same copy of *RE* for the whole branch. But this generates the **wrong results**. Try to execute a simulation run and it will produce different results from Sim4.

### Pseudo random numbers

All 10 series for each branch produce *identical* results, and hence overlapping graphical series.



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The different copies of X will all be able to reach the value of RE stored in their **Obj0**, but this value will be identical for all the copies of X, hence generating identical random walks.

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If we want that the same computation is replicated every time it is requested, and not only once at each time step, we need to define it as a **function**. A function is like a variable, but its values cannot be stored for analysis since it can produce multiple (or none) values at each time step.

To fix the Sim5 double click on variable **RE** and then click on properties. In the resulting window select the option **Function** and confirm.

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Replicating now the simulation will generate exactly the same results as using Sim4.

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We will run the model including 100,000 distinct instances of random walks over 1,000 time steps. LSD can execute simulations under a stripped down mode, renouncing any graphical representation, and hence being as fast as a pure C++ compiled code.

• Load a fresh configuration Sim5 and save it as Sim6.

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- Open Data/Set Number of Objects. Set to 1 the number of Obj0. Set to 100,000 the number of Obj1.

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- Ouble-click on X and deselect the option Run Time Plot. Ensure that the option for saving is on.

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- Move the Browser to show Obj1.
- Oouble-click on X and deselect the option Run Time Plot. Ensure that the option for saving is on.
- In menu Run/Sim.Settings set the number of Simulation Steps to 1,000.

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- Move the Browser to show Obj1.
- Oouble-click on X and deselect the option Run Time Plot. Ensure that the option for saving is on.
- In menu Run/Sim.Settings set the number of Simulation Steps to 1,000.
- Run the simulation. It will take a minute or two to compute the 10<sup>8</sup> computations. Press on **Fast** in the **Log** window to skip the messages on time steps completed, further reducing the execution time.

At the end of the simulation open the module **Data/Analysis of Results**. Select all the series available and place them in the **Series Selected** box (**Tip**: right-click on one of the **X** series and see the **Help** for details).

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Don't try to plot the time series graph, this will freeze the graphical engine of the LSD windowing system for several minutes. Rather, select the option **Cross Section** (bottom right), and click on the button **Histograms**.

The resulting box will ask for a few options. Accept all the default values, or click on **Help** for details.

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The graph produced on confirmation represents a frequency histogram of evenly spaces classes of values between the minimum and the maximum values from all the series at the last time step of the simulation run.

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The graph produced on confirmation represents a frequency histogram of evenly spaces classes of values between the minimum and the maximum values from all the series at the last time step of the simulation run.

As expected, the graph produces a neat Gaussian function, centered on the expected value of 0.

#### **Pseudo random numbers**

Histogram of end-of-simulation frequency classes.

