

DINAMICA 1.0:
Automated Dynamical Analysis

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1 Introduction

DINAMICA is a tool for automated analysis of the multi-stable dynamics. For this, it uses the differential equation tools supplied with various stochastic validation algorithms. Consequently, DINAMICA can be used as a tool for the comparative analysis between the deterministic and the corresponding stochastic systems.

Simple systems usually do not have very complex dynamics. On the other hand, as the complexity of the system grows the number of possible dynamical regimes increase, leading inevitably to the co-existing of some of the regimes. In rigid terms, for the same parameter set the system demonstrates several possible regimes and, as usual rule for the deterministic systems, different initial conditions lead to the different dynamics.

The usual assumption in physics, chemistry and biology is that the whole system consists of the equal elements of smaller size. Thus, DINAMICA considers the equation supplied by the user to be divided into sub-systems of smaller size. In principle, these sub-systems must be of the same dimension and all equal in other respects (the latter is not required though). Such a system is called a Symmetrically Coupled System.

1.1 Interface

DINAMICA has a primitive interface with no graphics carried out by the program itself. It is easy expandable for using the Gnuplot for visualization of results. The installation process can be done with or without support of the Gnuplot utility. The Gnuplot is freely available from the Internet.

DINAMICA also uses the external library (Gnu Scientific Library, GSL) to perform some basic calculations like integration of the system of differential equations, performing statistics etc. The library can be easily found and downloaded from the Web and subsequently installed.

These two are the main dependencies of DINAMICA which require the user to have them pre-installed. Although the Gnuplot is optional, the GSL is mandatory to have installed on the user's computer.

DINAMICA has been successfully tested on Unix-like OS: Linux, Mac OS X etc. No testing was performed in the Windows environment, though the Windows use of the program is NOT prohibited. Any tests for implementation of the software on the new platforms are much appreciated and all needed help will be provided by the original author.

The most updated information regarding the interface as well as the installation and prerequisites instructions can be found in the README file of DINAMICA package.

1.2 Disclaimer

DINAMICA is distributed as is. The author has no responsibility for the performance of the software nor for any possible harm or damage the software might cause to the platform it is run upon. All the details of the disclaimer are explained and further clarified in the Gnu General Public License.

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The License can be found at <http://www.gnu.org/licenses> or in COPYING file of the package. All the source files of DINAMICA have the Disclaimer in the beginning along with the copyright information and contact details of the original author(s).

1.3 Where To Get

DINAMICA is a part of the free-software community. The description and all source files needed for the end-user utilization and the development are located in the main software forge of the Free Software Foundation — Savannah (<http://savannah.gnu.org>). All the latest updates of the program get uploaded to the Savannah pages dedicated to DINAMICA:

<http://savannah.nongnu.org/projects/din>

No special registration is needed, the software is available right there. Various information about the package and the course of its development can be found on those pages. For example, the download area contains all public releases of the software:

<http://download.savannah.gnu.org/releases/din/>

There is also possibility to become a member of the developers' team if someone is interested in introducing the new features to DINAMICA. All such efforts are appreciated. In this regard, the package file TODO is a good opportunity to see all the new features and capabilities waiting to be realized.

All the changes and updates are recorded in the ChangeLog and NEWS file of the package. ChangeLog files are also organized by year, i.e. all updates introduced in 2012 are in ChangeLog2012. The most recent updates are in ChangeLog.

1.4 Acknowledgements

The author is thankful to all his teachers in the area of dynamical systems and stochastic processes, whom he met during the fulfillment of the Masters and Doctoral thesis. Especially, Prof. Evgenii Volkov (Lebedev Physical Institute, Moscow) and Andre Ribeiro, PhD (Tampere University of Technology).

The special thanks go to N.Devillard, who has developed the interface for using the Gnuplot utility from within a C-program (see <http://ndevilla.free.fr/gnuplot/>). DINAMICA uses this interface.

1.5 Installation

NOTE: the most recent instructions for the installation process and all dependent procedures are located in the package `README` file.

DINAMICA uses AutoConf and AutoMake systems for configuration. The general information on how to configure the package administrated by these two systems is located in the `INSTALL` file of the package.

1.5.1 Prerequisites

1. Unix/Linux OS.
2. gcc compatible C compiler, needed for Dinamica functioning (not only compilation).
3. GNU Scientific Library (GSL) installed.
4. Gnuplot plotting utility installed (optional, but advisable).

1.5.2 Main Steps of Installation Process

1. Download the archive (usually `.zip` or `.tar.gz`) and uncompress it.
2. Configure the system by typing `./configure`. This will check for all the requirements and complain if any of those is not found. You may consider `CPPFLAGS` and `LDFLAGS` variables, as well as `--prefix` option to `./configure`, before configuring the system (see below).
3. Type "make" to compile the `libdin.a` and the `dinamica` itself. The two must appear under the `src/` directory in the root, i.e. where you uncompressed the archive. (It is also important to know why we need these two files for the software to work.)
4. Type `make install` to install `dinamica` executable and `libdin.a` library to the usual destinations (`/usr/local/bin` and `/usr/local/lib`,

respectively). This might require the root password. You may uninstall the program later by typing `make uninstall` to remove those two files from the system. After the installation one might want to remove all the files extracted from the archive.

1.5.3 Important to Know Before Configuring the Package

DINAMICA processes the input from the user (equations, parameters, variables, constants etc.) in the form of the file having `.ode` extension (ode-file, for short). The result of the processing is the output `.c` file with C language definitions and functions for the user system and the binary configuration `.bcf` file. This output `.c` file is then compiled with the DINAMICA library (`libdin.*`) generating the final executable `.din`. This executable is then invoked to read the `.bcf` file and, finally, the program fires up. Schematically this process is shown in Fig. 1.

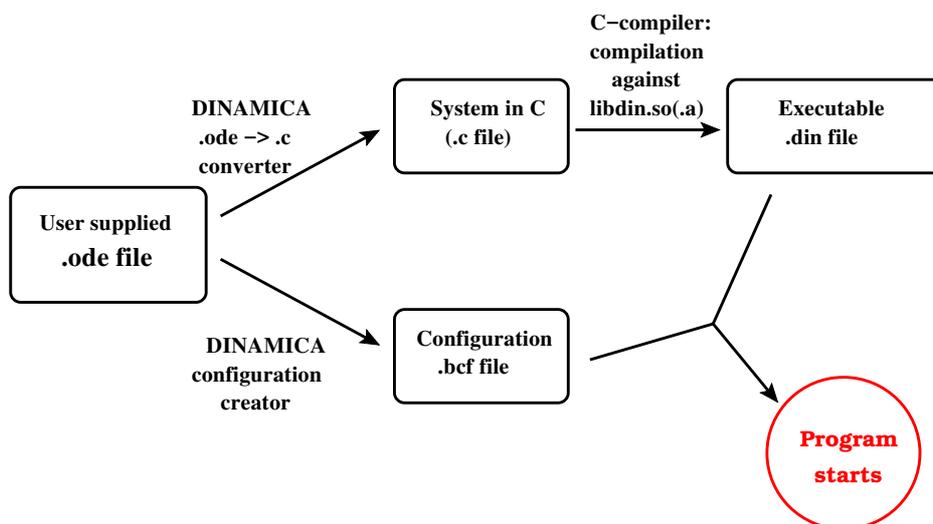


Figure 1: The general scheme representing the DINAMICA preparation procedures before it starts.

It is important to understand that the compilation and linking of the libraries take place during the functioning of DINAMICA. Thus, the compiler and the right path for the required libraries are needed to be properly set when DINAMICA is first compiled. One should take care of this before the configuration starts.

The way DINAMICA can obtain the full set of the paths is to specify `CPPFLAGS`, `LDFLAGS` and `--prefix` option, together or one by one when needed. These three variables are passed to the DINAMICA compilation command, so they are crucial. The current directory where DINAMICA is invoked is always checked for the dinamica library (`libdin.so` or `libdin.a`).

CPPFLAGS. This variable is important for the preprocessor, a program checking the included, so called header, files. During the configuration process the `./configure` script checks for several `.h` files from the GSL and the standard C libraries whether they are available. Sometimes it fails to find them in the standard locations. In this case, the `CPPFLAGS` is needed. The usage is simple: if one knows that the GSL headers are located, for example, in `/usr/local/include/`, e.g. the full path to `gsl_odeiv2.h` file is `/usr/local/include/gsl/gsl_odeiv2.h` (similarly for all other `gsl *.h` files), then one could type at the shell prompt:

```
CPPFLAGS=/usr/local/include ./configure
```

This sets the environment variable for the `./configure` script. Next, in `DINAMICA` this variable will be set after the `-I` flag of the compiler, i.e. will also be used as a path to the header files to find (since `DINAMICA` uses the same set of header files to compile the user-defined `.c` file). Note, that the path to the GSL header files is always `gsl/*.h`, so omit the `gsl/` directory when specifying the `CPPFLAGS` like in the example provided.

LDFLAGS. This variable shows the path to the GSL library (and possibly the standard C libraries). If the `gsl` library files are located in `/usr/local/lib` then combining with `CPPFLAGS` one could type

```
CPPFLAGS=/usr/local/include LDFLAGS=/usr/local/lib ./configure
```

which should do the trick. Additionally, `DINAMICA` will compile the output `.c` file against the `libdin.so/libdin.a` using this variable or the one specified by `--prefix` (see below) to find the `libdin.so/libdin.a` library. This value will go to the `-L` flag of the compiler, which specifies the path to the libraries to be used in the compilation.

--prefix. `./configure` script accepts the `--prefix` option, which specifies the installation directory for `make install` command. Namely,

```
./configure --prefix=/usr/local
```

would install all the files produced by the package to the corresponding subdirectories of `/usr/local`. For example, binary files, i.e. `dinamica`, would go into `/usr/local/bin` and libraries, i.e. `libdin.a`, — into `/usr/local/lib`. This variable is set after the `-L` flag to the compiler in `DINAMICA`.

1.6 Running `DINAMICA`

The most simple invocation of `DINAMICA` is to type at the command line:

```
dinamica <your_ode_file>.ode
```

This will fire up the program, if the configuration and installation processes went well. <your_ode_file>.ode is the .ode file containing the system to be analyzed. This file should be prepared by the user.

The program starts by showing some information about the system it has read from the .ode file, a little report on the compilation of the transformed .c file against the DINAMICA library (libdin.so) and some miscellaneous information. A typical output looks like:

```
<here is your prompt>$ dinamica bruss2.ode
Starting to check system's specification:
Rebuilding the function:
f(x,y)=a-(b+1)*x+y*x^2
Rebuilding the function:
g(x,y)=b*x-y*x^2
u1' = f(u1,v1)
v1' = g(u1,v1)+dv*(v2-v1)
u2' = f(u2,v2)
v2' = g(u2,v2)+dv*(v1-v2)
Starting transfer to 'bruss2.c'...
Done.
Active parameters: 'b', 'a', 'dv',
Writing 'bruss2.bcf'...Done.
Preparing for 'bruss2.c' compiling...
Compiling with: 'gcc -Wall bruss2.c
-I/opt/local/include -L/opt/local/lib -L./
-o bruss2.din -ldin -lgsl -lgslcblas -lm '
Starting bruss2.din...
./bruss2.din bruss2.bcf
```

DINAMICA Ver. 1.0 (<dl.sv.nongnu.org/releases/din/>)
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This software uses the gnuplot_i library written by N.Devillard
(see <<http://ndevilla.free.fr/gnuplot/>>).

This program comes with ABSOLUTELY NO WARRANTY;
for details type 'warranty' or simply 'w'
This is free software, and you are welcome to redistribute it
under certain conditions; see GNU General Public License for details.

Report bugs to <elias.potapov@gmail.com>

Reading 'bruss2.bcf'...Done.

>

Then the DINAMICA's own command line (>) is shown up.

DINAMICA is organized in menus, which have a certain hierarchy that does not go deeper than 2-3 levels starting from the main menu and ending at submenus. The ubiquitous commands are `ls` and `sh/show` (not everywhere). The former shows the list of possible menu items and the command shortcuts to access them, while the latter shows the information corresponding to a certain type of menu.

The menu items are shown such that the shortest command abbreviation for invocation of this particular menu is surrounded with parentheses. For example, menu item `(N)umerics` can be accessed by typing `n` at the DINAMICA prompt. The main menu look like:

```
>ls
MAIN menu:
(R)un*
(R)un (t)ransient*
(R)un (i)nitiaL*
(C)alculate*/
(F)ile/
(N)umerics/
(P)arameters*
(V)ariables*
P(e)riodics/
(G)raphics/
(I)nitiaLs*
(T)rajectory/
C(o)ntinue/
R(a)ndom/
(Er)rors/
(S)ingularity/
(L)yapunov/
(R)un (l)inear*
```

Some more examples to clarify the concept, to access `File` menu one should type in `f` at the prompt, to run system for the transient amount of time type in `rt` etc.

The main menu also has the sign showing the type of the submenu. A slash '/' at the end of menu item tells a user that this is a regular menu, while an asterisk '*' tells about a command to be invoked.

Every command released at the prompt needs an 'Enter' key hit at the end to be accepted by the DINAMICA interpreter. One can use several commands in a row to access more items in the submenus. For example, to show the numerics information of the systems one could type in `n`, 'Enter', `sh` which will first bring the user to the `Numerics` submenu and then in that particular submenu the `show` command is invoked. Alternatively, the same result in a rather faster way can be achieved by typing `n sh` at the prompt. This is quite self-explanatory.

```
>n sh
* Dimension: 4
* Number of systems: 2
* Number of parameters: 3
* Number of user functions: 2
* Number of auxillary entities: 0
*****
Total time: 50
Transient time: 50
Step: 0.02
Writing step: 1
Sampling frequency: 1.00
Method: rkf45
Langevin flag: false
```

2 Input Files

2.1 ODE Files

This type of files is the main source for DINAMICA. Certainly, men must tell software what to do, that is why `.ode` files preparation mostly lies upon the user's shoulders. First, we will present some general examples of using `.ode` files, which will provide one with a good basis to start his/her own research almost immediately once the examples are understood. Next, we will try to draw the detailed explanation on how the `.ode` files are constructed and what is the main syntax and usual pitfalls one might encounter during the declaration of his/her own system.

2.1.1 Two Start-off Examples

Let's start from the simple example we used before for invocation of DINAMICA:

```

# bruss2.ode
# two Brusselator coupled by the diffusion equations

%system 2

du1/dt=f(u1,v1)
dv1/dt=g(u1,v1)+dv*(v2-v1)
du2/dt=f(u2,v2)
dv2/dt=g(u2,v2)+dv*(v1-v2)

f(x,y)=a-(b+1)*x+y*x^2
g(x,y)=b*x-y*x^2

jac u1=-(b+1)+2*v1*u1-du,u1^2,du,0
jac v1=b-2*v1*u1,-u1^2-dv,0,dv
jac u2=-(b+1)+2*v2*u2+du,u1^2,-du,0
jac v2=b-2*v2*u2,-u1^2-dv,0,dv

par b=2.5,a=1,dv=.57
init u1=10,u2=1,v1=1.5,v2=15
@ total = 50,yax=v1,yax2=v2
done

```

This file specifies the systems of two diffusively coupled Brusselators. The system of Ordinary Differential Equations (ODE's) is written down in a self-explanatory way. The variables of the system are u_1 , v_1 , u_2 and v_2 . Everything that is not a variable in the ODE's is either parameter or function/auxillary entity. Thus, dv is a parameter, while f and g are the functions whose definition follows the ODE specification. The function declaration introduce new parameters: a and b . The function arguments list is within the parentheses, hence (x,y) denotes two arguments to the functions.

There is a possibility to include Jacobian into the system declaration. This is used by several solvers. However, it is not necessary since DINAMICA has a capability of calculating the Jacobian numerically. That is done automatically, when the program cannot find the user-supplied Jacobian. The Jacobian is included through `jac` statement followed by a variable name whose differential equation is going to be subject for differentiation. All different derivatives are separated by commas ', ' . Thus, the Jacobian matrix is formed.

`par` statement specifies the initial values of the parameters of the system. If some of the parameters are omitted here, their values equal to zero by default.

`init` statement does the same job as `par`, but for the initial values of the variables. Again, omitted variable values default to zero.

@ sign denotes the line with internal parameters for DINAMICA. In the above example, `total` means the total time of integration, `yax` denotes the variable to be plotted on the Y-axis, `yax2` denotes the second variable to be plotted along with the first one on the Y-axis.

`done` statement is not necessary, but shows the hereditary connection of DINAMICA to the Bard Ermentrout's XPPAUT software. Actually, the ODE syntax is mainly like in XPPAUT (see <http://www.math.pitt.edu/~bard/xpp/xpp.html>).

As you might have noted the # sign starts the comments. The very important DINAMICA directive is `%system` which defines number of physical sub-systems that the whole system has. In the example above, this number is 2, meaning that there are 2 Brusselators coupled with each other.

This example must provide a start-off principles of defining the ODE systems through the `.ode` file.

The next example include the definition of the discrete stochastic system whose dynamics is going to be compared against the deterministic system of ODE's.

```
# Toggle Switch example
# two mutually inhibiting proteins x and y

x'=alpha/(1+y^n)-d*x
y'=alpha/(1+x^n)-d*y

init x=10,y=0
par alpha=1,n=2,d=0.1

g:alpha/(1+x^n);+y
g:alpha/(1+y^n);+x
g:d*x;-x
g:d*y;-y

@method=complex,method2=rkf45,sf=1,total=10000
done
```

This system has two ODE's describing the dynamics of two protein species in a Genetic Toggle Switch. This system is characterized with the two stable states and possibility to switch between them. Here you can find another type of variables/ODE definition — through `x'` notation. This totally equals the dx/dt notation.

The main difference as compared to the first example in this section is the `'g':`'s statements closer to the end of the file. These statements define the Gillespie procedure for solving systems possessing the discrete and stochastic dynamics. The syntax goes as follows: `g` is a keyword, everything

between ':' and the following ';' is known as the propensity of the chemical reaction and, finally, everything after ';' is the update vector, i.e. the vector (whose elements are separated by commas ',') containing the information on how the numbers of the species involved in the reaction change after the reaction takes place. In our example, first reaction produces (+y) one molecule y, while the third one removes (-x) one x molecule from the reaction space.

Under the section of internal parameters (after @) you can find `method=complex` directive which tells DINAMICA to use both stochastic discrete and normal ODE integration methods altogether and compare the results. Additionally, `sf=1` statement tells DINAMICA to use `sampling frequency` for the stochastic simulation equal to 1 (depending on the units used in the system, it could be seconds, years or ages).

Finally, `done` finishes the input.

2.1.2 The Syntax in Details

The `ode` files are the main source for DINAMICA running. The program cannot start without an appropriate `ode` file supplied to it. In this section we discuss the syntax of the `ode` files in some greater detail. The general rule for `ode` file syntax is the every new statement line is separated from others by a `newline` symbol.

Comments. Comments in the `ode` files are marked by # sign and everything that follows the sign until the end of line is ignored by the DINAMICA parser.

Differential equations. The Ordinary Differential Equations (ODE) determine the list of all variables of the system as well as the law governing its dynamics. The main syntax for defining ODE's is:

$$\begin{aligned} dx/dt &= \dots \\ y' &= \dots \end{aligned}$$

Thus, x and y are declared as variables and the corresponding Right Hand Sides (RHS) of the ODE's are determined.

The RHS's are transferred as they are to the corresponding `.c` file for further compilation (see Fig. 1). So the operator precedence is determined exactly the same way it is determined for C-language mathematical expressions. The C-language principles can be found elsewhere, e.g. [1].

For example, the following system

$$\begin{aligned} dx/dt &= 1 + x^2 - y*x \\ y' &= (x+1)*y - y^2*x \end{aligned}$$

will be converted to the corresponding C-function:

```
f[0] = 1 + pow(x[0],2) - x[1]*x[0];
f[1] = (x[0]+1)*x[1] - pow(x[1],2)*x[0];
```

where \mathbf{f} is an array of the RHS functions, while \mathbf{x} is an array of variables in standard C language notation and indexing, starting from 0 for the 1st equation, 1 for the 2nd and so on. Note also that the `pow` function from the `math` C library is used for the power sign $^$. The `math` library is included into DINAMICA automatically.

Importantly, DINAMICA first checks the derivatives and extracts the variable names. Then the actual transferring of the RHS equations starts. All unknown symbols at this moment are set to be the parameters that can be freely varied from within the program.

Physical systems. DINAMICA deals only with symmetrical dynamical systems or, at least, those having the same number of variables. Here, we call a sub-system, defined by a subset of variables of the whole system, a physical system. For example, there can be two oscillators coupled to each other. Importantly, these sub-systems are mathematically inside a single system, however, physically they connote different systems.

Number of physical systems is defined by the `%system` directive that can appear anywhere in the text. For example,

```
%system 2
```

defines 2 physical sub-systems.

There is no way to define physical sub-systems in the iteration style. This should be fixed in the future releases. In contrast, DINAMICA given N sub-systems divides the whole systems into N equal parts starting from the first variable and finishing by the last. If one has defined 6 variables x_1, x_2, x_3, x_4, x_5 , and x_6 and $N = 2$, then variables x_1, x_2 , and x_3 denote sub-system 1, whereas x_4, x_5 , and x_6 belong to the sub-system 2. Similarly, the same system can be divided into $N = 3$ sub-systems. However, $N = 4$ will not work and DINAMICA will report the error in this case, since the sub-systems must contain the same number of variables.

Parameters. Parameter values are set through `par` statements:

```
par a=1,b=5
```

The above sets value for the parameter a to 1 and that of b to 5.

All undeclared alpha-numeric entities in the ODE declarations are set to be parameters. The only way to assign a parameter value is via the `par` statement. Thus, any parameter that did not appear under the statement will be assigned a value of zero.

Initial conditions. The initial conditions for the deterministic simulations can be set with `init` statement. The line has to be of the form:

```
init x=0,y=0
```

specifying the initial conditions for x and y , respectively. Several `init` statements can be on different lines.

The initial condition can be a parameter. In this case the variable's initial condition and parameter receive a link, which is preserved throughout the running session of DINAMICA. If the user simulate a trajectory from the initial condition the link tells which parameter value to take for which variable. This has the preference over other methods of setting the initial conditions, for example, from a file.

Functions. Functions can be defined in the ode file facilitating a simpler and more structured way of representing the system.

```
f(x) = ...  
g(x,y) = ...  
w(x,y,z) = ...
```

The above examples set the functions for the subsequent use in ODE equations. The syntax includes the function name followed by the list of arguments (in parenthesis), separated with comma. Consider the following self-explanatory example:

```
f(x) = x^2  
u' = v*f(u)  
v' = u
```

The expression `f(u)` is expanded to `u^2` at the moment of the transfer to the C-file. Note, that the argument list can contain similar names as those used in the ODE specification, i.e. the name space of variables of a function is separate from that of the system. For instance, the following is the same as previous with `u = y` and `v = x`:

```
f(x) = x^2  
y' = x*f(y)  
x' = y
```

Auxillary entities. DINAMICA allows for using auxillary entities that are some constant expressions. The auxillary functions can be used in the equations and other mathematical expressions. The main purpose of using the auxillary entities is to provide a capability to track certain mathematical relations during the simulations. However, this capability is not fully supported yet and general advice is to avoid using auxiliaries.

The main syntax for the auxillary entity is:

```
aux R=x*y/(1+a)
```

The keyword `aux` signifies the beginning of the auxiliary statement.

Jacobian. Some of the numerical methods require calculation of Jacobian matrix. This can be done numerically (by default). However, if the user doubts the correct numerical estimates of the Jacobian and if the RHS derivatives are virtually easy to calculate analytically, it is possible to supply the Jacobian matrix to DINAMICA.

The general form of the Jacobian matrix is:

$$\begin{vmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \cdots & \cdots & \cdots & \cdots \\ \cdots & \cdots & \cdots & \cdots \\ \cdots & \cdots & \cdots & \cdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \cdots & \frac{\partial f_n}{\partial x_n} \end{vmatrix}$$

where f_i is a RHS function of i -th variable x_i . There are total n number of ODE's and, obviously, the same number of variables.

The keyword for entering the Jacobian is `jac`. After the keyword the variable name, whose differential equation derivatives to be considered on the line. Then, after the variable name, the equal sign '=' goes, followed by the expressions of the derivatives separated by comma ','. For example,

```
jac u = 2*u-1,v*u-1
```

determines the Jacobian entry for the `u` variable.

IMPORTANT: `jac` statement **MUST** be put in the ode file after ALL ODE declarations. Since for the very first entry the system parser must know how many variables are in the system, in other terms, what is the dimension of the system.

In the considered above example, we do not know what is the number of the variable `u` and hence cannot determine derivatives over which variables are presented. All we know is that there is a variable `u` in the system and the dimension of the system is 2 (two entries separated by the comma).

Here is the full functional example of Jacobian statement use:

```
u'=a-(b+1)*u+v*u^2
dv/dt=b*u-v*u^2
```

```
jac u = -(b+1)+2*v*u,u^2
jac v = b-2*v*u, -u^2
```

The entries `jac u` and `jac v` can be put in any order, since the variable name after `jac` determines the equation.

Noise terms. The so called Ito differential equations can be studied in DINAMICA. For this one has to specify the noise terms in the RHS functions. The Ito differential equations (sometimes referred to as Langevin equations) has a specific form of the noise term, that is, some **amplitude** times the white noise = standard normal distribution (with 0 mean and variance of 1) multiplied by square root of the time step, namely:

$$Noise = A \times \sqrt{dt} \times N(0,1),$$

where A is the amplitude, dt is a time step and $N(0,1)$ is the standard normal distribution. The amplitude can be set in DINAMICA using the keyword **lang**:

```
lang u=0.1,v=0.01
```

The variable name signifies the ODE equation to add the noise term to. The expressions right to the equal sign can contain normal mathematical operators and parameters.

In order to simulate the stochastic trajectory defined by Langevin equations one has to enable the corresponding simulation methods: simple Euler method or Milstein method (both are fixed step size algorithms). Additionally, the so called **Langevin flag** must be turned on. All these options can be found under the **Numerics** menu.

Stochastic discrete algorithm. DINAMICA allows for using the discrete stochastic algorithm when simulating the trajectories [2]. The algorithm was formulated by D. Gillespie and is of great use for the systems with small number of molecules. Generally, the algorithm requires the so called **propensity** functions to be defined and the **update vector** of the chemical reactions. The propensity is analog of the chemical reaction rate. The update vector defines which chemicals are consumed and which are produced as a result of the reaction. DINAMICA has the straightforward algorithm (referred to in [2] as “Direct” method) for simulating the discrete stochastic trajectory.

For example, a chemical reaction $A + B \xrightarrow{k} C$ will be written in DINAMICA ode file as:

```
gill:k*A*B;-A,-B,+C
```

The keyword for the Gillespie algorithm instruction is **gill**, or simply **g**, followed by the colon **’:**. Then the propensity is defined as in the example above **k*A*B**, followed by the semicolon **’;**. Finally, the update vector is specified, showing how many molecules were removed (**-**) from the reaction space and how many were added (**+**, the plus sign can be omitted) for each chemical species involved in the reaction.

NOTE: the variable name space used in this approach is global, that is the same as for the ODE equations. Thus, to define a variable to use for the discrete method one needs to define the corresponding ODE. This way DINAMICA becomes an excellent tool for qualitative and quantitative analysis and comparison of both discrete and continuous deterministic trajectories. If the ODE system being compared with the corresponding discrete stochastic system does not contain a variable to be used one can always write an “empty” ODE to declare a variable to the program:

```
dx/dt = 0
```

NOTE also, that DINAMICA does not prepare the propensity functions for the user. Thus, for example, for the chemical equation $A + A \xrightarrow{c} B$ the propensity must be equal to $c \times A \times (A - 1)/2$ and be set explicitly by the user:

```
g: c*A*(A-1)/2; -2A, B
```

Additionally, the update vector should not contain mathematical expressions, i.e. in the example above two molecules of A are consumed, which is set by $-2A$ and NOT $-2*A$.

Technical parameters. Technical parameters are those you specify when dealing with the numerical simulations (time of simulation, step etc.), graphics (what to plot, type of the plots) and some other. This list can be extended significantly, since DINAMICA contains a lot of different technical parameters related to many analysis procedures. Every new release of the software should extend the list significantly. Though not all the parameters can be defined from within an ode file, almost all of them can be set from within the running program.

The technical parameters to DINAMICA are usually presented in a form `<par>=<val>` after @ sign beginning a line, where `<par>` is a name of the parameter and `<val>` is its value. The pairs can be separated by comma inside a single line. For example,

```
@total=100,dt=.2
```

total defines the total time of integration/simulation for the trajectories.

dt defines the time step for numerical integrators. The adaptive step size algorithms take this as an initial guess for the step.

trans determines so called transient time, that is a time interval that can be run without producing any output. This is usually used for getting the final attractor of the system.

method/m defines the method of integration. The possible options can be: **eu** (simple Euler fixed step size procedure), **run-kut4** (Runge-Kutta (4,5) fixed step size method), **rkf45** (embedded Runge-Kutta-Fehlberg (4,5)

adaptive size method), **rk8pd** (embedded Runge-Kutta Prince-Dormand (8,9) adaptive step method), **rkck** (embedded Runge-Kutta Cash-Karp (4,5) adaptive step method), **bsimp** (Implicit Bulirsch-Stoer method, requiring Jacobian calculation), **discrete** (Stochastic Discrete Algorithm from [2]), **milst** (Milstein method for stochastic ODE's, langevin flag (see below) must be turned on), **complex** (complex method calculating deterministic trajectory and one or several stochastic ones combining them into a single output).

method2 defines the method for the deterministic run if **method=complex**.

ws defines the writing step to the output trajectory file. For example, **ws=100** will force the integrator to put every 100-th point into the output file.

sf is sampling frequency for the stochastic discrete algorithm [2]. Determines how often to write the output. For example, **sf=5** will force the simulator to produce output every 5 sec/min/hours or any other time units implicitly assumed in the model under study.

lf/lang is a Langevin flag. If set to non-zero value, tells the program to augment the ODE equations with the noise terms, whose amplitudes must be defined in the ode file or inside the program.

graph/gf is a graphics flag. Non-zero value indicates that the user wants to use the Gnuplot output. If the program was compiled with Gnuplot support this option is by default true. Usually used to suppress the output.

xaxis/xax is a variable name or index for plotting at the X-axis. **O** indicates time (default), 1,2,3 etc indicate 1st, 2nd, 3rd etc variable.

yaxis/yax/yaxis1/yax1 is a variable name or index for plotting at the Y-axis. First variable is taken by default.

yaxis2/yax2 is an additional variable to plot at the Y-axis.

yaxis3/yax3 is yet another additional variable to plot at the Y-axis. Three is the maximum number of plotted variables at Y-axis.

permeth/pm is either 1 or 0 determining the way the periods are calculated from the simulated trajectory. 0 indicates the default method of Poincare sections, 1 indicates the autocorrelation method. The former produces set of values for period, whereas the latter produces a single value for a trajectory.

pervar/pv is a variable name or index telling which variable to use to assess the system's period.

cross/c is a Poincare section level. This value determines the fraction between the max and min of the trajectory. Namely, the section level is determined by the formula: Poincare section = $(\max(\mathbf{X}) - \min(\mathbf{X}))/C + \min(\mathbf{X})$, where \mathbf{X} is an array of points defining the trajectory and C is the cross value defined by **cross**.

Finishing input. The keyword **done** is reserved for the final statement in the ode file. This is **optional** and kept for the backwards compatibility as

well as compatibility with XPPAUT ode files [3].

3 Simulation Methods

The simulation methods in DINAMICA can be divided roughly into two big classes: deterministic and stochastic.

3.1 Deterministic Methods

These methods are primarily obtained from the GNU Scientific Library (GSL) [4]. The methods are intended for the numerical simulation of the system of ODE's. Although there are a plenty of methods for the purpose in the GSL library, we have opted to use several methods. For the very same reason, the DINAMICA set of algorithms can be easily extended using the GSL.

Euler 2-nd order. The simplest fixed step size approach to solve the system of ODE's. The routine for the procedure is not from the GSL library.

Runge-Kutta 4-th order. Explicit (classical) 4-th order Runge-Kutta algorithm. Own routine.

Runge-Kutta-Fehlberg 4-th order. Explicit embedded Runge-Kutta-Fehlberg (4,5) method. This method is a good general-purpose integrator. This is from GSL: `gsl_odeiv2_step_rkf45` step function (see GSL reference available at [4]).

Runge-Kutta Prince-Dormand 8-th order. Explicit embedded Runge-Kutta Prince-Dormand (8, 9) method. From GSL: `gsl_odeiv2_step_rk8pd`.

Runge-Kutta Cash-Karp 4-th order. Explicit embedded Runge-Kutta Cash-Karp (4,5) method. From GSL: `gsl_odeiv2_step_rkck` step function.

Implicit Bulirsch-Stoer method. Implicit Bulirsch-Stoer method of Bader and Deuffhard. The method is generally suitable for stiff problems. This stepper requires the Jacobian. From GSL: `gsl_odeiv2_step_bsimp`.

3.2 Stochastic Methods

These methods can be both continuous and discrete in nature.

Discrete. The discrete stochastic simulation follows the algorithm described by D. Gillespie in [2] ("Direct" method for the Monte-Carlo step).

Continuous. This option refers to the Ito stochastic differential equations, also known as Langevin equations. The general form of the equation of this type is:

$$\frac{dx}{dt} = f(x) + g(x) \times dW, \quad (1)$$

where $f(x)$ is a normal RHS of an ODE, $g(x)$ is the function determining the amplitude of the noise and dW is a Wiener process.

For solving this type of a problem one can choose either the Euler procedure or the special Milstein method for solving Ito differential equations. For both toggling of the Langevin flag (see manual on `Numerics` menu) is required.

3.3 Complex Method

Complex method is a special feature of DINAMICA, which allows for simultaneous simulation of both stochastic and deterministic counterparts of a system, further facilitating the comparative analysis of them. This method is designed so that it first simulates the deterministic part of the system and then several times the stochastic part, be it discrete or continuous.

The complex methods requires specifying the deterministic algorithm. The stochastic algorithm is chosen automatically based on whether the langevin flag or the discrete algorithm are set.

4 Dynamical Analysis Methods: Trajectory System

For the analysis of the dynamics DINAMICA uses the simulated time series. At the moment, all methodology described in this section refers only to the deterministic time series that is obtained from the simulation of a system described in terms of Ordinary Differential Equations (ODE's). However, the actual core algorithms for the dynamical analysis do not need a special means for trajectory generation, only the deterministic property is required.

Note, that the analysis can be applied to any system described in terms of ODE's, hence, it is irrelevant what is the independent variable of the system is. Throughout this manual we use term "time" as the independent variable when describing the algorithms, however, it can be a spatial variable, for example, for some specific systems. Moreover, we assume that the independent variable is increasing from the beginning to the end of the simulated trajectory as, usually, happens with time variable. Thus, for example, the point i of the trajectory takes place for the larger value of time (or other independent variable) than the point $i - 1$.

The set of ODE's represents the system under study mathematically. The system, however, can be further divided into physical sub-systems representing the actual coupled physical elements. Although indiscernible mathematically such systems as a whole may produce a significant differences in the dynamics of the physical sub-systems (see e.g. [10]).

First, DINAMICA analyzes the dynamics of 1D system: either the whole system, if the system's physical dimension is equal to 1, or every sub-system (number of physical sub-systems is determined in the `%system` directive in the `.ode` file). Next, the whole system analysis is fulfilled producing the overall dynamics report.

The analysis is based upon the concept of *slope*. The next section explains in detail the concept. Then, we present the slope algorithm for 1D systems (meaning systems with the only physical dimension) and, finally, the procedure for ND systems.

4.1 The Slope Concept

Once the deterministic time series is simulated it can be analyzed through the DINAMICA TRAJECTORY-system (or T-system). The elementary unit the whole analysis relies upon is *slope*. The slope is a part of the calculated trajectory (or the time series) from the beginning of section, where the trajectory's points start to decline over time, up to the beginning of section, where the points start to ascend over time, or vice versa. The concept is illustrated in Fig. 2.

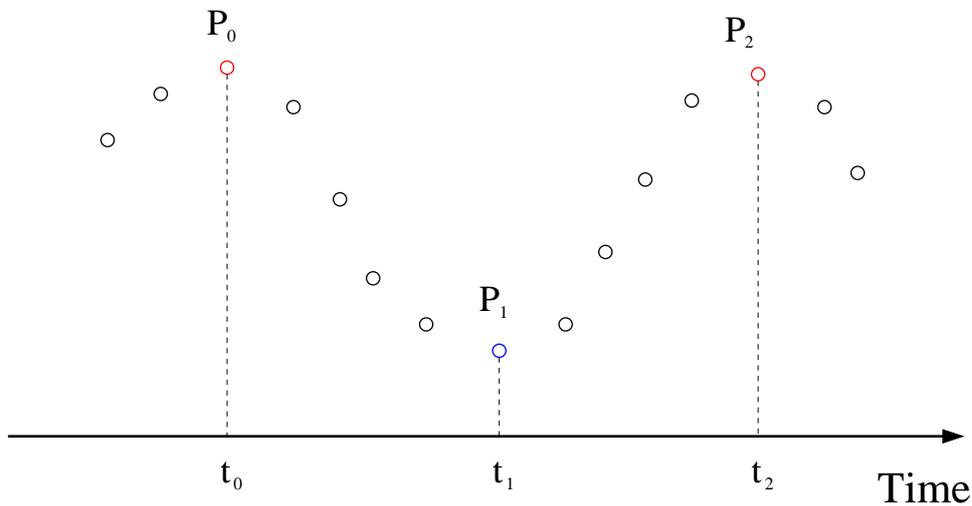


Figure 2: The graphical representation of the slope concept

The set of calculated points belonging to the trajectory has *turning points*, at which the trajectory changes its direction from pointing upwards

to pointing downwards, or vice versa. Strictly speaking in terms of continuous trajectory and infinitesimal time step, at those points the corresponding variable’s derivative becomes equal zero.

In Fig. 2 these points are depicted in red and blue (P_0 , P_1 and P_2) demarcating two slopes: the first contains all the points from point P_0 up to P_1 inclusively and the second — from P_1 to P_2 .

Practically speaking, all kinds of trajectories contain the slopes, since in the most cases the calculated points have at least small discrepancies in their values caused by the computer representation of the real numbers. Nevertheless, in those really rare cases, when two adjacent points have values that are the same (up to the all representation bits of the numbers), the T-system determines the *plateau*, and the beginning and/or end of the slope is calculated as the point in the center of plateau (if the plateau contains even number of points the ceiling operation is taken).

Thus, one can distinguish between *ascending* and *descending* slopes. In Fig. 2 the first slope is descending (from P_0 to P_1), while the second one (from P_1 to P_2) is ascending. For the obvious reasons, the ascending and descending slopes alternate over time. Additionally, the end of each slope is the beginning of the next one. For this reason, for example, the T-system of DINAMICA stores only the time that the slope starts at. Using the example depicted in Fig. 2, only t_0 and t_1 are stored for the two slopes shown.

There are three entities that describe a slope in DINAMICA. The slope *amplitude* S^A is the difference between the variable values at the beginning and at the end of the slope. The slope *base* S^B is the lowest point of the slope. The beginning of the slope in time S^T is the third entity.

Continuing with the example of Fig. 2, the slope appearing earlier in time has amplitude equal to $(P_0 - P_1)$, base — P_1 and the start time — t_0 , while the later appearing slope has amplitude equal to $(P_2 - P_1)$, base — P_1 and the start time — t_1 . As a result, two adjacent slopes (descending and then ascending) have the common base.

4.2 1D Dynamical Analysis

DINAMICA processes the simulated time series of a single variable within the first physical dimension of the system. All dynamics detection is based on the slope analysis of the simulated trajectory. The process could be virtually divided into the *preparation* stage and actual slope algorithm.

The preparation process starts with the calculation of the peaks and troughs of the simulated trajectory (in Fig. 2 points P_0 and P_2 are peaks and point P_1 is a trough). Then, based on the peak and trough information the slopes of the trajectory are calculated. Next, the slopes are checked to represent real slopes of the trajectory and not small “fluctuations” of it. These small fluctuations appear in systems with very different timescales, e.g. when there are very slow and very fast equations in the system. The

“fluctuations” appear due to this difference of the time scales and *not* due to any stochastic forces applied to the system. Perhaps, a suitable integration algorithm accounting for the stiffness of the system might resolve this problem. In any case, this “sanity” checking of the slopes does not hamper the whole analysis and, hence, is included in the preliminary preparation stage of the process.

Before moving to the section explaining the slope algorithm we need to take a look at the dynamical regimes and the definitions used in DINAMICA for the dynamical processing.

4.2.1 Dynamical Regimes of 1D Systems

1D (here we refer to the physical dimension) systems can in principle have two types of dynamics: *stationary* and *non-stationary* behavior. The former refers to so called steady state of a dynamical system, at which the system eventually ends up at a stationary point. The latter, oppositely, connotes a behavior that does not rest at any particular point. The non-stationary behavior contains many sub-classes of behaviors, ranging from pure oscillatory dynamics with distinguished period and amplitude to chaos with no certainty in the period and amplitude.

NOTE #1: the DINAMICA slope algorithm determines the *genuine* dynamics of a system in a sense of the realized (really appearing) behavior. For example, there is no way in the slope algorithm to distinguish between oscillations induced by the harmonic oscillator and the limit cycle oscillations (the first one is the linear system possessing the “center” fixed point with pure imaginary Lyapunov numbers, while the second is the non-linear system containing the Hopf bifurcation).

NOTE #2: DINAMICA is capable of determining the chaotic behavior — a dynamical regime in which the deterministic laws of motion produce unpredictable behavior. But this capability of the program goes beyond the slope algorithm, which most likely would produce the “unknown” or *period- n* oscillatory regime (see below), where n is very large, in the case of chaos.

Thus, the non-stationary dynamical behavior is either oscillations with some predictable periodicity or chaos. The oscillatory regime can be characterized with its amplitude and period. While the amplitude value is not qualitatively useful, the period might be composed of several sub-periods, denoting, if they are present, the additional frequency of oscillations. This takes place, for instance, in case of the torus attractor.

Here, we define *period- n* oscillations referring to the oscillatory trajectory with every oscillation being of the same amplitude and the same level as those of the n -th oscillation *prior* or *next* to the given one. For example, Fig. 3 shows two sinusoidal oscillation trends. The first one is described by the expression $\sin(x)$ and the second one — by $\sin(x) + \sin(x/2)$. The former has one frequency of oscillations and the latter one has a period composed

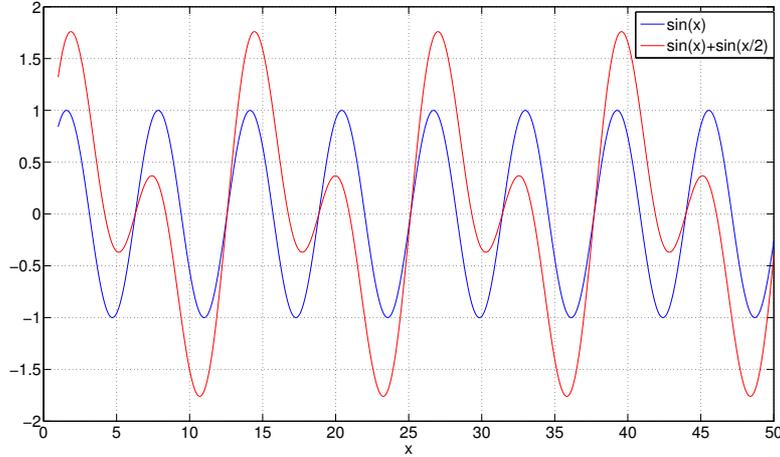


Figure 3: Sinusoidal oscillations as an example of period-1 (blue) and period-2 (red) oscillatory dynamics.

of two sub-periods. The $\sin(x)$ oscillatory regime is period-1 oscillations, the $\sin(x) + \sin(x/2)$ oscillatory regime is period-2 oscillations.

The summary of all dynamical regimes detectable within the frame of the slope algorithm goes in Tab. 1. Here we also present the numerical codes for each possible regimes: 0 means stationary dynamics, $1 \dots n$ means period-1...period- n oscillatory regime and -1 stands for the unknown or undetermined regime.

Regime	Name(numerical code)
Stationary state	SS(0)
Period- n oscillations	OS(n)
Unknown/undetermined	$-(-1)$

Table 1: 1D dynamical regimes detected by the slope algorithm.

4.2.2 The slope algorithm

Note that the dynamics of a system cannot be stationary and non-stationary at the same time. Thus, these two pathways of analysis are separated from each other in the slope algorithm (SA).

Stationary dynamics. The stationary dynamics can be easily detected when the system is already at the rest state, meaning it does not deviate from the state significantly. This can be checked by comparing the very last slope's amplitude to the system-wide absolute error level (which is defined

by the user). If the amplitude is so small (less than the error level), then the decision is made right away — $SS(0)$ (see Tab. 1). This can be expressed by the formula:

$$S_{last}^A < \epsilon_{abs}, \quad (2)$$

where S_{last}^A is the amplitude of the last slope in the trajectory, i.e. containing the largest value for the time (S^T) of the slope. The value of ϵ_{abs} is highly system-dependent and changes, for example, with different units of the system. Thus, it is highly recommended to check this value before any serious studies of the dynamics of the system of interest.

The more complex situation of the stationary dynamics detection takes place, when the algorithm receives the time series where the system does not appear to be in its final resting state. But, since the system has it, it must converge to it. The convergence is manifested in the consistent decrease in the slope amplitudes when moving from the first slopes to the last ones. So the slopes successively demonstrate a decrease in their amplitudes as time goes on (or any other independent variable). In this case, the SA checks the slope amplitudes moving from the last to the first slope. If, starting from the last slope, at least 50% (default value) of all the amplitudes increase sequentially (in the reversed direction), then SA detects the $SS(0)$ regime (Tab. 1). This increase must not be interrupted anywhere, in other words, approximately the last half of the trajectory shows the monotonic decay in the slope amplitudes in direct time (or other independent variable).

The tolerance ϵ_{rel} determining the criterion for the descending of the slope amplitudes is relative and also specified by the user. Thus, the $(i + 1)$ amplitude is considered to be decreasing compared to the previous i -th amplitude, if the following condition is fulfilled:

$$(S_i^A - S_{i+1}^A) > \epsilon_{rel} \cdot S_i^A \quad (3)$$

This means, that from slope i to slope $i + 1$ the amplitude decreases for more than $100 \times \epsilon_{rel}\%$. The usual value for ϵ_{rel} is 0.01 (1%). Note that for the increasing amplitudes the condition cannot be ever fulfilled, since the left hand difference of the amplitudes becomes negative in this case, while the right hand product is always positive.

Let M be the following:

$$M = \max(i \mid (S_i^A - S_{i+1}^A) \leq \epsilon_{rel} \cdot S_i^A \mid ((S_i^T < S_{i+1}^T) \mid (i < i + 1))), \quad (4)$$

i.e. the maximum index of the slope where the condition (3) for the slope decrease is violated, given that the time is always increasing in the time series (this must be applied to any other independent variable too) and the smaller slope index corresponds to its smaller time value.

Finally, given the total number of slope amplitudes in the simulated time series is equal N , the criterion for the stationary dynamical regime is:

$$\frac{N - M}{N} > \epsilon_{thr}, \quad (5)$$

where the $\epsilon_{thr} = 0.5$ (50%) by default and $N - M$ gives the number of slopes sequentially decreasing at the end of the simulated time series.

Summarizing, the stationary (resting) regimes get detected in two cases:

1. The last amplitude is smaller than the absolute tolerance (error) given by the user.
2. At least, $100 \times \epsilon_{thr}\%$ (50%) of the slope amplitudes sequentially *increase* as calculated *from the end to the beginning* of the trajectory.

For the examples of the SS detection look, for instance, into Sections 6.1 and 6.2.

Non-Stationary dynamics. There are two ways in DINAMICA to determine the oscillatory regime of any periodicity given the period (return time) is constructed in a constant number of subperiods (see Sec. 4.2.1).

The **first way** to detect OS dynamics is when the system of interest is on the attractor, that is one can find the similar slopes in the trajectory. The algorithm here tries to find a slope that is similar (in amplitude and base) to the very last slope of the trajectory. The comparison is made with the relative tolerance level ϵ_{rel} that is used in the SS detection (by decreasing slope amplitudes). Therefore, the error level ϵ_{rel} is somewhat a demarcation threshold between similar and different slopes. That is, the slopes are similar if the following two criteria are fulfilled:

$$\begin{cases} |S_i^A - S_k^A| < \epsilon_{rel} \cdot S_k^A \\ |S_i^B - S_k^B| < \epsilon_{rel} \cdot S_k^A \end{cases} \quad (6)$$

Note, that here we use the *absolute* difference between the amplitude and base values, since the approaching of the oscillatory attractor can be both from higher slope amplitudes and lower slope amplitudes to the last one. Additionally, we use slope bases for the comparison, since the amplitudes are not enough for the purpose. For example, the amplitudes can be similar, but the slopes are located in different parts of the phase space.

For the first trial of finding similar slope k (see eq. (6)) takes the value of N (the number of slopes in the trajectory), while i moves from $N - 1$ to 1 with step 2 (since a period is composed with 2 consecutive slopes) until the conditions (6) are met. Once the conditions are satisfied the algorithm proposes the lag (i.e. number of slopes in the period, see more on the system

lag in Sec. 4.3.1) and decreases k by one. For every new value of k algorithm tries to find the lag again to support the proposal found for the previous value of k . The algorithm continues decreasing k until the newly proposed lag is not equal to the one determined for the previous k . The algorithm reports how many k iterations supported the first proposal of the lag, which is the relevant information for the user.

When the lag is zero for the first value of k , the **second** procedure is to be carried out. It checks for monotonic increase in the slope amplitudes as the time goes on, in a similar way (but with inverted conditions) the algorithm checks the monotonic decrease in the case of SS detection.

The following condition determines the increase in the slope amplitudes:

$$S_{i+1}^A - S_i^A > \epsilon_{rel} \cdot S_i^A \quad (7)$$

The algorithm calculates the fraction of slopes satisfying the condition (7) at the end of the trajectory. The fraction must be larger than the predefined error level ϵ_{thr} that is equal 0.5 by default. Note, that this threshold error can be different from that used in the SS detection. This is fulfilled with the set of expressions (8), where M denotes the max index of a slope where the condition (7) is violated, then the number of the increasing slope amplitudes at the end of the trajectory equals $N - M$, where N is total number of the slopes in the trajectory. Finally, the comparison with the threshold value is done.

$$\left\{ \begin{array}{l} M = \max (i \mid (S_{i+1}^A - S_i^A) \leq \epsilon_{rel} \cdot S_i^A \mid ((S_i^T < S_{i+1}^T) \mid (i < i + 1)) , \\ \frac{N - M}{N} > \epsilon_{thr} \end{array} \right. \quad (8)$$

Summarizing, the non-stationary (oscillatory) regimes get detected in two cases:

1. One can find a slope similar (in amplitude and base) to the very last slope anywhere in the given trajectory.
2. At least, $100 \times \epsilon_{thr}\%$ (50%) of the slope amplitudes sequentially *decrease* as calculated *from the end to the beginning* of the trajectory.

The examples of the OS detection can be found, for instance, in Sections 6.2 and 6.3.

4.2.3 General Comments on Slope Algorithm

Here we list all the characteristics of the slope algorithm, common pitfalls in the analysis, known problems, limitations and assumptions behind the algorithm. Users of DINAMICA are assumed to be familiar with these notes.

1. The piece of the trajectory supplied to the T-system must be sufficient for the analysis. It also must represent the dynamics of the system under study at the most relevant scale.
2. The slope algorithm is a heuristic procedure, that is it detects the actual realized dynamics as it is provided by the user. It does not calculate any analytical characteristics from the rigid and solid theory to make a decision on the dynamics. Thus, the scope and scale of the supplied trajectories and the system-wide constants must be controlled by the user of the program. For example, the error level for the stationary dynamics detection is fully dependent on the system under analysis: it can be 100 or 0.0001 and only units of the model and common sense tell what the appropriate level is. Moreover, the part of the trajectory presented to the algorithm could be just a tiny piece of a larger picture: what if the trajectory will go extremely high with increasing slope amplitudes right after the time window with consistent decrease in the amplitudes? Such regimes do exist. This is a philosophical issue, which can be partially overcome with the analytical knowledge like Lyapunov exponents and linear stability analysis in the case of steady states. DINAMICA does NOT do this kind of analysis. However, the core slope algorithm can be easily extended to calculate analytical characteristics (like Lyapunov exponents or Floque multipliers) to further support the made decision. Thus, the user can do the analytical assessments on the steady state solutions in a separate algorithm, implemented in DINAMICA (see the (S)ingularity menu).
3. 1D detection of the dynamics is represented with 4 distinct procedures: two for SS and two for OS. Given the error levels supplied by the user are set in correspondence with the scale of the model, the first procedures for both SS and OS detection methods are the most precise or, in other words, trustworthy. The second procedures for both SS and OS detection predict the most probable dynamical outcome of the model. For this reason, they produce many numerical assessment entities to be analyzed by the user.
4. The slope algorithm proceeds as follows. First, it tries to identify SS regime by the absolute tolerance way (1st procedure). If that fails, it goes for the rest three procedures (2nd SS and 1st and 2nd OS). The next goes the 1st OS procedure. If that fails, the SA proceeds to the 2nd SS procedure (identification of the amplitude decrease). If this does not produce any result, the algorithm proceeds to the 2nd OS detection procedure (monotonic increase in amplitudes).
5. OS detection relies only upon the monotonic increase (2nd procedure)

in the slope amplitudes and not upon the decrease. However, approaching the oscillatory attractor can be both from the space lying outside the closed curve of the attractor in the phase space and from within. Thus, the decrease in the slope amplitudes can signify the approaching to the oscillatory attractor, but can be detected with 2nd SS procedure checking for the monotonic decrease in the amplitudes. The workaround here is the appropriate absolute error level for the SS detection and longer time series, since the longer time series can reveal either further decrease of the amplitudes or arriving to the oscillatory attractor. The ration between the last slope and the absolute error level is shown when SS is detected via the 2nd procedure. This indicates how close the final trajectory amplitude to the actual stationary state.

6. It is possible, given the time of the slopes, to calculate approximate rate of changing of the slope amplitude over time. Hence, the additional integration might be invoked from within the T-system to further clarify the regime, if the user has supplied a non-sufficient piece of the trajectory.

4.3 ND Dynamical Analysis

Once the slope information about each subsystem is gathered, it is supplied to the N-dimensional algorithm. ND analysis is based on the comparison of some entities of each system one with another so that all systems are compared. For example, if one has three subsystems and an entity A is compared, then A_1 is compared with A_2 , A_2 compared with A_3 and A_3 compared with A_1 (where A_i stands for A entity in the subsystem i). In other words, the entity undergoes all possible pairwise comparisons. In the system with M subsystems the number of comparisons is $M \cdot (M - 1)/2$.

The comparison is made by taking the ratio between two entities being compared. The maximal ratio out of all $M \cdot (M - 1)/2$ comparisons is called **gain**. Gain is always ≥ 1 since the larger value is divided by the smaller one in all comparisons unless the two comparable entities are equal.

Overall, there are two test in the ND algorithm: **homogeneity** and **phase** tests. The former checks the levels of the variables in different subsystems, whereas the latter checks the time moments of the slopes in order to understand the phase shift of the subsystems in respect to each other.

The overall dynamics has its numerical code. The code is a simple average of numerical codes of the subsystems (see the codes in Sec. 4.2.1). Thus, if all subsystems possess the same dynamics and, hence, the same dynamical numerical code, the overall dynamical code will be the same. If, for some reason, the numerical codes in the subsystems do not match each other the final overall numerical code will be non-integer number indicating

the **mixed** regime. This is, in general, considered to be the error in the dynamics checking. However, there are certain meaningful exceptions.

The preliminary test checks whether all sub-systems possess the similar dynamics. The **mixed** regimes, i.e. when some sub-systems have one behavior, whereas others have other behavior, are possible. In general, the mixed regimes are indicators of a strange behavior, however, some particular examples show this type of dynamical regime, for example, when there are two limit cycles in a systems separated in the phase space and one of them having extremely small amplitude, thus, getting detected as a steady state.

4.3.1 System lag

If all sub-systems are detected to be in one dynamical regime and this regime is not SS, the system lag needs to be determined. The system lag is a shift between the sub-systems as measured by the number of slopes. Practically, there are $M \cdot (M - 1)/2$ comparisons in a system with M sub-systems. Here, we refer to the system lag as a maximal lag between sub-systems.

The SA utilizes the iteration procedure to go through all sub-systems. In a single iteration, every sub-system's last slope is compared with the slopes of another sub-system starting with the last and proceeding backwards to the first slope until the **equal amplitude slopes** are found. Additionally, the equal amplitude slopes must have the same direction, pointing either downwards (descending slope) or upwards (ascending slope). Once these two conditions are satisfied the system lag is reported in a number of slopes from the current one to the one having similar slope characteristics.

The concept is further clarified by Fig. 4. The red system j runs in synchrony with the black system i . So the first comparison of the amplitudes $|A_0^j - A_0^i| < \epsilon_{rel} \cdot A_0^i$ and slopes' direction give the positive outcome, thus, the amplitudes and directions are equal. Hence, the lag is equal 0 and there is no need to compare other amplitudes of these two sub-systems. The iteration switch to the next pair of sub-systems.

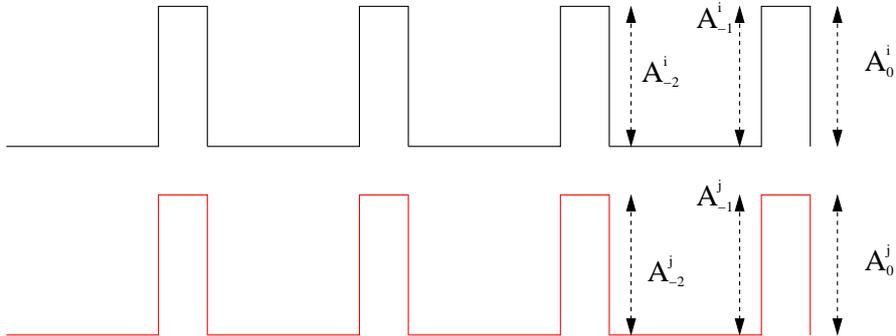


Figure 4: Zero system lag.

If the comparison fails, the further comparisons are carried out, i.e. $|A_{-1}^j - A_0^i| < \epsilon_{rel} \cdot A_0^i$, $|A_{-2}^j - A_0^i| < \epsilon_{rel} \cdot A_0^i$ etc., accompanied with similar slope direction comparisons, until the conditions are satisfied. Note the lag is reported as a number of slopes.

Consider the next figure (Fig. 5), which depicts a hypothetical system with the system lag of 2 slopes. The inequality $|A_{-2}^j - A_0^i| < \epsilon_{rel} \cdot A_0^i$ is fulfilled and the directions of the last slope of the system i and the slope -2 of the system j coincide (both descending).

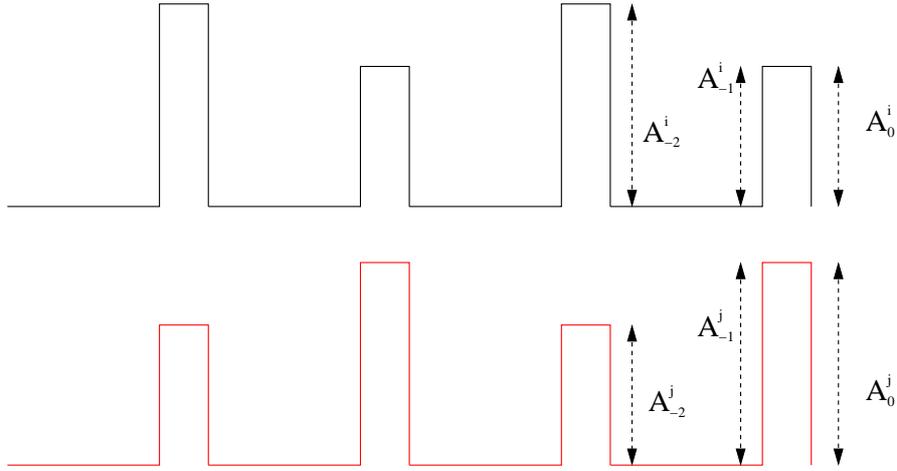


Figure 5: The system lag equals 2.

There is an important difference between the system lag and the phase of non-stationary dynamical regime. The phase refers to the actual time shift between the sub-systems and determined further in the T-system using the system lag. In order to stress the difference consider the following Fig. 6. The figure clearly shows the non-zero time shift between the systems, however, the system lag is still zero.

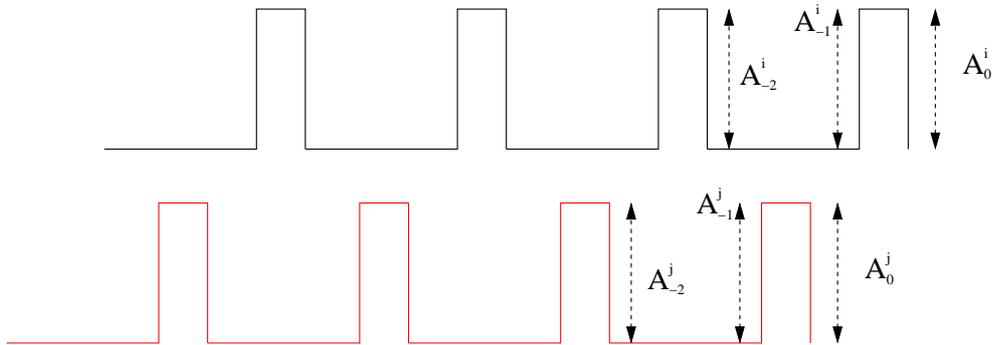


Figure 6: The zero lag and non-zero phase (shifted sub-systems).

Once the system lag is determined, the phase and the amplitude/base gains can be correctly computed, since now the T-system is aware of what slopes to compare, when calculating these entities.

4.3.2 Phase test

Phase is an actual time shift between the sub-systems. In DINAMICA it is reported as the fraction of the period of any OS-regime. Note, that the SS regimes do not have phase.

Given the system lag, the phase can be simply calculated as the difference of the corresponding slopes time moments S^T . The maximal phase is reported.

If the phase shift is not zero, solution is considered to be **out-of-phase** oscillations. Otherwise, solution is **in-phase**, meaning the synchronous oscillations in all subsystems.

The examples shown in Figs. 4, 5, 6 have 0T, 0.5T, and 0.5T phase shifts (T is a period), respectively.

4.3.3 Homogeneity test

For the homogeneity test the maximal gains of the base and amplitude are calculated. If the base gain OR the amplitude gain is larger than $1 + \epsilon_{rel}$, then the **in-homogeneous** solution is detected. Otherwise, solution is **homogeneous**. Note that the same ϵ_{rel} was used in the comparisons of 1D dynamical checking algorithms.

For OS type regime the base and amplitude gains are calculated accounting for the system lag information. For other types of dynamical behavior these gains are calculated on the last slope amplitudes of all sub-systems.

4.3.4 Report on the Overall Dynamics

The report of the overall dynamics of the multi-component system contains much useful information. First, the code name for the regime comes: **SS**, **OS**, **mixed** or **-** (undetermined) and some additional information about the detected regimes. At the end of the report the information about homogeneity comes. Moreover, OS regime has additional information about phase and period of oscillations.

Here is the example of global SS detected:

```
SS(0)/IH(bg=1.12,ag=3.54)
```

This means that the solution was in-homogeneous with base gain equal 1.12 and amplitude gain equal to 3.54.

Homogeneous SS solution gets detected with the following:

```
SS(0)/H(bg=1,ag=1)
```

Oscillatory regime (OS) gets detected with the following information:

OS-1(T=35.4)/IP(0T)/H(bg=1,ag=1)

This indicates that the OS-1 regimes was detected with period 35.4, as determined by simply subtracting the corresponding slopes' time moments accounting for the system lag information. The regime happened to be the in-phase regime (IP) with maximal phase shift of 0 periods T (0T). Finally, the homogeneity information is given.

Another type of OS detection might be:

OS-2(T=5.32)/OP(0.5T)/H(bg=1,ag=1)

This means that period-2 oscillations were detected with period 5.32. The solution is out-of-phase with the phase shift of half of the period. It is a homogeneous solution.

The OS solution, of course, can be non-homogeneous as well. In this case, the report would show IH mark with larger than 1 values of **bg** (base gain) and/or **ag** (amplitude gain).

5 Other Dynamical Analysis Methods

6 Examples

6.1 Single Gene Expression

(NOTE: the ode-file of the system can be found in the package `ode/` folder under the name `single.ode`)

Here we present the simplest model representing a gene producing mRNA and, finally, protein, which then represses its own synthesis. Formally, mRNA and protein numbers are the only variables of the model. The ODE's representing the system are:

$$\begin{aligned} dA/dt &= \frac{\alpha}{1 + \left(\frac{B}{K_h}\right)^n} - d_m A \\ dB/dt &= gA - d_p B \end{aligned} \tag{9}$$

This system is merely representing self-repressing gene. The repression is carried out through the Hill-function with a Hill-coefficient n . K_h is affinity of the repressor protein to the corresponding operator site of its own gene, which is expressed as ratio between the rate constants of binding and un-binding reactions of the repressor to/from the operator site. The mRNA and protein are degraded linearly with rate constants d_m and d_p , respectively. The translation process, synthesis of the protein with the mRNA, is also linearly modelled with rate constant g .

For some set of the parameters one can find a stable steady state in the system. The kinetics of the two variables of the system is depicted in Fig. 7. Being run on the trajectory, the T-system of DINAMICA reports:

```
***Checking variable a
SS(0): abs.tol = 0.001.
```

This reads as follows. The T-system checked variable `a` and made a decision on the dynamics regime — the steady state (SS), as determined by the final slope amplitude that happened to be less than the user-defined error level `abs.tol=0.001`.

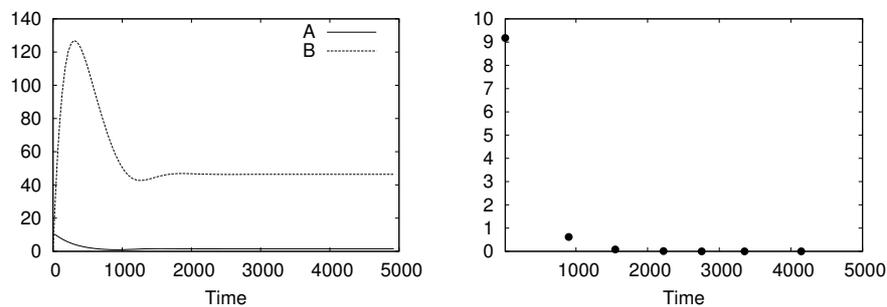


Figure 7: The time series of mRNA (A) and protein (B) of the system of a single self-repressing gene (left) and the corresponding slope amplitudes as calculated from A 's trajectory (right). Parameters used for the simulations: $\alpha = 0.05$, $K_h = 20$, $n = 2.6$, $d_m = 0.0033$, $g = 0.1$, $d_p = 0.0033$

The result shown in Fig. 7 and the dynamics detected can be further supported with the stability analysis of the fixed points of the system. This can be done under the `Singularity` menu.

```
1) iter = 5 F(x) = -2.640e-08 0.000e+00
Solution N 1
U(1)=1.530145
U(2)=46.368037
Analytical J=
-0.003300 -0.000255
0.100000 -0.003300
L_1=-0.0033 + 0.00504524i
L_2=-0.0033 + -0.00504524i
Stable
```

The first line of the output shows the iteration of the algorithm at which the solution was found and the RHS function values at the found point. It is clear that the RHS functions are practically zeros. Thus, the algorithm indicates one solution to be $U(1) = A = 1.53$ and $U(2) = B = 46.37$. Real

parts of the linearized system's eigenvalues L_1 and L_2 are negative indicating the stable steady state. The Jacobian is also shown: **Analytical** means that the user has supplied the Jacobian in the ode-file, while **Numerical** means that it was computed numerically.

6.2 Brusselator

(NOTE: the system analyzed here is located in the package `ode/` directory under the name `bruss.ode`)

The Brusselator system is a model corresponding to a non-existing chemical system of interaction of two molecules. Its major dynamical regime is limit-cycle oscillations. The system of equations of the Brusselator is the following:

$$\begin{aligned}\frac{dU}{dt} &= A - (B + 1)U + VU^2 \\ \frac{dV}{dt} &= BU - VU^2\end{aligned}\tag{10}$$

Analytically, the Hopf bifurcation at which the oscillations emerge takes place for $B > A^2 + 1$, that is in this parameter region the fixed point becomes unstable. We have shown the results of the dynamics test for parameter set: $A = 1$ and $B = 1.9$. This allows having the system slowly approaching the steady state Fig. 8.

Fig. 8 clearly demonstrates the consistent decrease in the slope amplitudes leading to the steady state point of the Brusselator, which has started from a non-resting state. The two criteria for the detection of the stationary dynamics can be fulfilled together or one by one. Namely, the last amplitude can be lower than the user-defined error level ϵ_{abs} , indicating the steady state or, alternatively, since all the amplitudes decrease over time, the steady state can be detected through the second method by calculating the decrease of the slope amplitudes.

If we make the dynamics test on the trajectory obtained from the initial conditions $U(0) = 1$ and $V(0) = 1$ and total time of integration 100 (see Fig. 8), the DINAMICA output would be:

```
***Checking variable u
Last ampl. (0.005568) / Abs. tol. (0.001) = 5.568
SS(0): 31 of 31(100%) slopes demonstrate monotonical decrease
SS(0): total decrease=99.4198%,rel.tol=0.01
```

This reads as follows. The program detected the monotonic decrease in the slope amplitudes of the trajectory of the variable u . All 31 slopes demonstrate the decrease as determined with the relative error level `rel.tol=0.01` (1%). Total decrease from the first slope to the last is 99.42%. The last

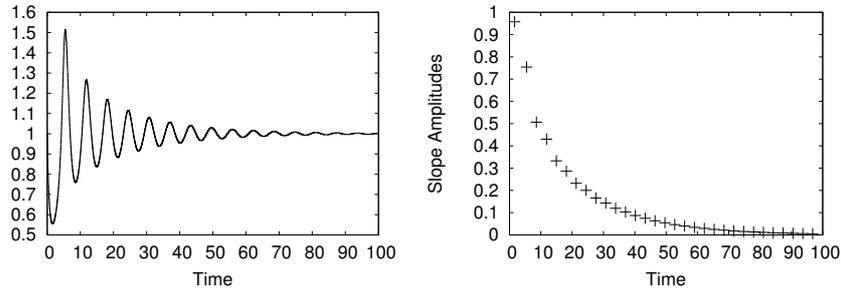


Figure 8: The stationary dynamics of U variable (see eq. (10)) of the Brusselator (left) and the corresponding slope amplitudes (right) over time. The parameter set for the Brusselator equations: $A = 1$ and $B = 1.9$

slope shows only 5.6 times higher amplitude than the absolute error level `abs.tol=0.001`.

This means that the trajectory is almost at the final resting state. The `SS(0)` regime is determined based on the monotonic decrease of the slope amplitudes (the second way). However, the ratio of the last amplitude and the absolute tolerance tells us that the first pathway of determining `SS` could be reached if one had a slightly longer integration time or a larger absolute error.

For example, if we make the absolute error at least 5.568 times larger than the current value, e.g. `abs.tol=0.006`, then the `SS(0)` is determined by this criterion only, like in the previous example of the single gene expression model:

```
***Checking variable u
SS(0): abs.tol = 0.006.
```

Alternatively, the longer integration time would determine the `SS(0)` without changing the absolute tolerance level. Since the bifurcation point leading to the oscillatory regime is quite close to the point $A = 1$ and $B = 1.9$ in the parameter space, the trajectory converge to the steady state for a quite long time for the given system.

Finally, we could set the parameters of the Brusselator system to obtain an oscillatory dynamics. For example, $A = 1$ and $B = 3$ make the fixed point unstable and oscillations emerge. The trajectory is shown on Fig. 9.

The trajectory test being run on the time series shown in Fig. 9 gives the following result:

```
***Checking variable u
Lag is 1: 24 of 25 (96%) comparisons support the conclusion.
OS(1): period-1,rel.tol=0.01.
```

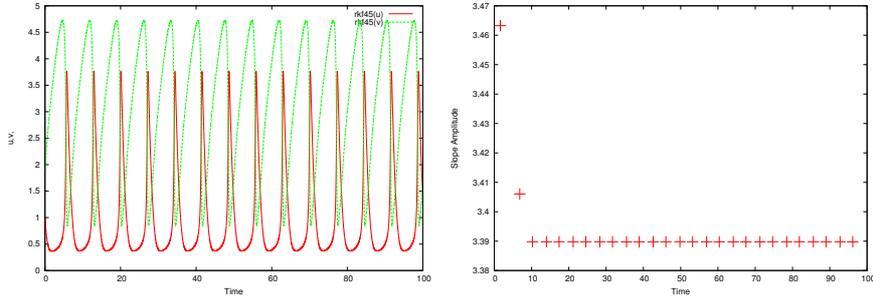


Figure 9: The time series of the variables of the Brusselator (left) for $A = 1$ and $B = 3$ and the corresponding slope amplitude values (right).

This reads as follows. The period-1 oscillations are detected and this is supported by the mutual comparison of 24 out of 25 slope amplitudes and slope bases. The decision is made with the relative tolerance equal to 0.01 (1%).

6.3 Repressilator

(NOTE: the file with the Repressilator system described in this section is shipped with the package archive and located in the `ode/` directory under the name `rep.ode`)

The Repressilator is a synthetic genetic circuit consisting of three genes and exhibiting an oscillatory behavior. The latter is possible due to the circular arrangement of the three negative feedback loops introduced by the proteins of the Repressilator genes. The first model of the Repressilator was proposed by the authors who experimentally constructed the genetic network [5, 6, 7]. The system of dimensionless ODE's is the following:

$$\begin{aligned}
 da/dt &= \frac{\alpha}{1 + C^n} - a & dA/dt &= \beta(a - A) \\
 db/dt &= \frac{\alpha}{1 + A^n} - b & dB/dt &= \beta(b - B) \\
 dc/dt &= \frac{\alpha}{1 + B^n} - c & dC/dt &= \beta(c - C)
 \end{aligned} \tag{11}$$

The system (11) represents the dynamics of mRNA (lowercase letters) and protein (uppercase letters) species of three genes A, B, and C. The repression of each gene's mRNA is represented by the Hill function that is dependent on the corresponding protein numbers. Additionally, each mRNA is degraded that is represented by the negative terms of the equations. The proteins are synthesized and degraded linearly.

The logic and the design behind the scheme imply the oscillatory behavior of the circuit. However, there must be some steady state, which then loses stability for the oscillations to emerge. This state appears for a small

value of the transcription rate α . The increase in α makes the system lose its steady state stability via the Hopf bifurcation and the limit cycle oscillations emerge Fig. 10.

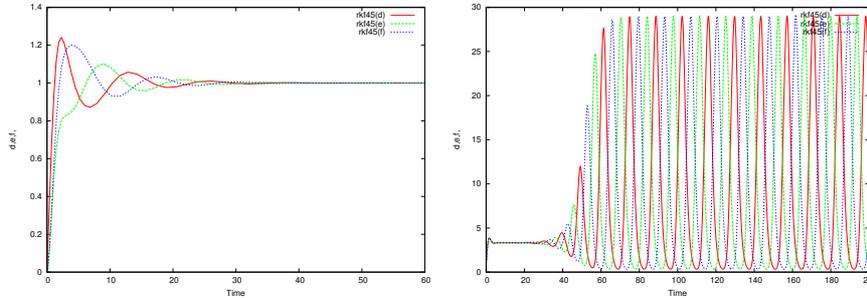


Figure 10: The stationary (left, $\alpha = 2$) and oscillatory (right, $\alpha = 40$) regime of the Repressilator model. The variables A, B, and C, corresponding to the proteins, are shown. Other parameters used in the simulation: $n = 2$, $\beta = 1$.

The T-system test being run on the trajectory shown in the left panel of Fig. 10 shows SS with the absolute tolerance equal 0.001:

```
***Checking variable e
SS(0): abs.tol = 0.001.
```

T-system detects the OS being run on the trajectory shown in the right panel of Fig. 10:

```
***Checking variable e
Lag is 1: 17 of 30 (56.6667%) comparisons support the conclusion.
OS(1): period-1,rel.tol=0.01.
```

The slope amplitudes course over time clearly demonstrate how the trajectory settles down on the limit cycle attractor. First, the amplitudes are small and then start increasing until they end up on the plateau (Fig. 11). Thus, the T-system reports 17/30 amplitude comparisons to support period-1 oscillation decision, which means that approximately half (56.7%) of the trajectory demonstrates pure limit cycle oscillations. This is seen from Fig. 11 where the slope amplitudes are equal from about time 60 onward.

6.4 Two coupled Brusselators

(NOTE: the `ode` file for the systems is called `bruss2.ode` and can be found in the `ode/` folder of the package.)

The system of two identical diffusively coupled Brusselators gives rise to the whole set of various dynamical behaviors. We mostly rely in our analysis on [8], which gives the extensive bifurcation analysis of the system.

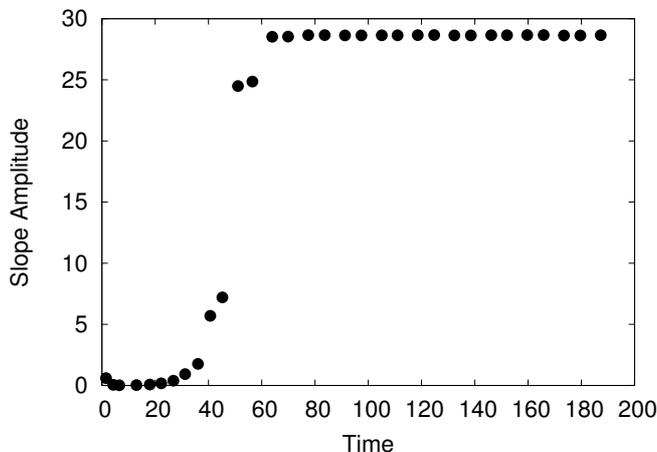


Figure 11: The slope amplitudes of the Repressilator system showing the transition to the oscillatory dynamics. Parameter set: $\alpha = 40$, $n = 2$, $\beta = 1$ (eq. (11)).

The bifurcation diagrams and results of this work can serve as a guide for finding dynamical behaviors of the system.

The system of equations describing the behavior of coupled Brusselators is the following:

$$\begin{aligned}
 \frac{dx_1}{dt} &= A - (B + 1)x_1 + x_1^2 y_1 \\
 \frac{dy_1}{dt} &= Bx_1 - x_1^2 y_1 + D(y_2 - y_1) \\
 \frac{dx_2}{dt} &= A - (B + 1)x_2 + x_2^2 y_2 \\
 \frac{dy_2}{dt} &= Bx_2 - x_2^2 y_2 + D(y_1 - y_2)
 \end{aligned} \tag{12}$$

As one can see from the system (12) the coupling is realized through one of the variables of the system y_i . Here we can see the 2-dimensional physical system, each consisting with 2 variables x_i and y_i , where i refers to the number of the system.

Given that the single Brusselator demonstrates the steady state dynamics and oscillations appearing through the Hopf bifurcation, we can assume that the homogeneous solutions of the systems (12) that correspond to these two regimes must exist.

Homogeneous Steady State. For some initial conditions and parameters $A = 1$, $B = 1.5$, and $D = 0.57$ the system (12) converges to a steady state Fig. 12.

The T-system of DINAMICA reports:

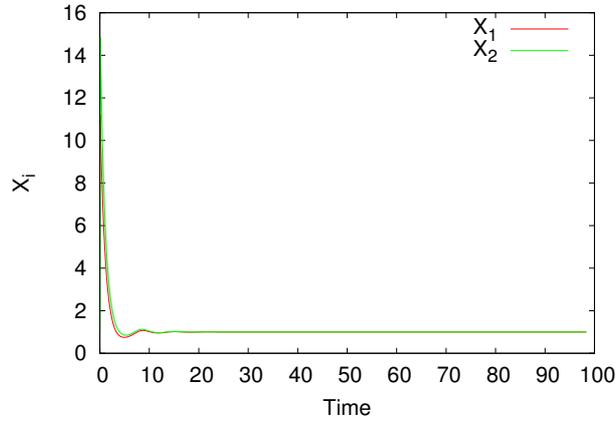


Figure 12: The steady state of the system (12). Parameters: $A = 1$, $B = 1.5$, $D = 0.57$. The initial condition for the simulation was: $x_1 = 10$, $y_1 = 1.5$, $x_2 = 1$, $y_2 = 15$.

```
***Checking sub-system 1(u1):
SS(0): abs.tol = 0.001.
```

```
***Checking sub-system 2(u2):
SS(0): abs.tol = 0.001.
```

```
=====
DYNAMICS:
SS(0)/H(bg=1,ag=1)
```

So, the both subsystems were determined to have SS(0) and the total regime is homogeneous SS(0). The overall report on the system can be seen under the title DYNAMICS:. Homogeneity is seen by amplitude (ag) and base (bg) gains both equal to 1 (about the base and amplitude gains see the Section 4.3).

Homogeneous In-Phase Oscillations. Increasing parameter B to 2.5 and starting from the steady state obtained in the previous section as the initial condition, one can see the synchronous oscillatory dynamics of the system (12). The solution is shown in Fig. 13.

The T-system of DINAMICA reports:

```
***Checking sub-system 1(u1):
Lag is 1: 25 of 27 (92.5926%) comparisons support the conclusion.
OS(1): period-1,rel.tol=0.01.
```

```
***Checking sub-system 2(u2):
Lag is 1: 25 of 27 (92.5926%) comparisons support the conclusion.
```

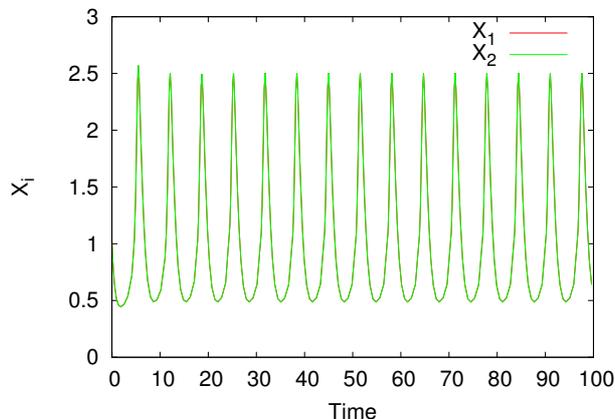


Figure 13: The synchronous oscillatory solution of system of eqs. (12). Parameter used in simulation: $A = 1$, $B = 2.5$, $D = 0.57$. The initial condition was the steady state obtained in the previous section.

```
OS(1): period-1,rel.tol=0.01.
=====
DYNAMICS:
OS-1(T=6.591)/IP(0T)/H(bg=1,ag=1)
```

The report reads as follows. Each subsystem has been determined to have period-1 oscillations, i.e. OS(1), with the relative error equal to 1%. More than 92% of the slope comparisons support the period-1 conclusion for both sub-systems. Thus, it is believed to be true single period oscillations. The overall dynamics checking does not reject the subsystem's conclusions about the regime and furthermore shows that the solution is synchronous (in-phase, IP) with period equal to 6.591 and 0 phase shift (0T) as measured in the period units. Moreover, the regime is homogeneous: amplitude and base gains are both equal to 1.

In-homogeneous Steady State. The work [8] reports that for the IP solution parameter set we used in the previous section, i.e. $A = 1$, $B = 2.5$, $D = 0.57$, there is another stable solution that is in-homogeneous steady state.

Here, in order to find the unstable solution, we will use the random initial condition approach, which is an additional technique utilized by the T-system of DINAMICA. The random initial condition technique assigns random initial conditions and tries out the trajectory. This method is available under the DINAMICA's R(a)ndom menu.

Here, we opted to throw 10 random initial conditions for the system (12) with the span equal to 10 (for more details about the random initial conditions see the appropriate section of the DINAMICA's manual) that was

found to be sufficient. For each initial condition T-system performs the dynamical test and reports the result for every such test as well as the overall statistics at the end of the calculation. Additionally the analyzed trajectories are plotted. This example's plot can be seen in Fig. 14.

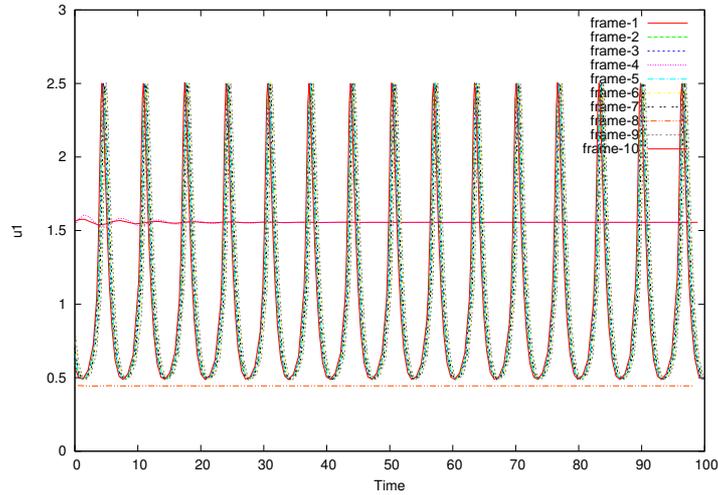


Figure 14: The output of the random initial condition trials. 10 random trials are plotted for the parameter set: $A = 1$, $B = 2.5$, and $D = 0.57$.

The report looks as following:

Dynamics report:

```
#1) OS-1(T=6.5894)/IP(OT)/H(bg=1,ag=1)
#2) SS(0)//IH(bg=1.53092,ag=2.38667)
#3) OS-1(T=6.5628)/IP(OT)/H(bg=1,ag=1)
#4) SS(0)//IH(bg=1.53091,ag=1.59574)
#5) OS-1(T=6.5888)/IP(OT)/H(bg=1,ag=1)
#6) OS-2(T=13.0331)/IP(OT)/H(bg=1,ag=1)
#7) OS-1(T=6.5762)/IP(OT)/H(bg=1,ag=1)
#8) OS-1(T=6.5635)/IP(OT)/H(bg=1,ag=1)
#9) SS(0)//IH(bg=1.53093,ag=2.63333)
#10) OS-1(T=6.6523)/IP(OT)/H(bg=1,ag=1)
Abs.tol = 0.001, rel.tol = 0.01
```

Regimes statistics:

```
Number of regimes: 10
Steady States: 3 (30%)
Oscillatory: 7 (70%)
Homogeneous: 7 (70%)
In-homogeneous: 3 (30%)
```

Homogeneous oscillatory: 7 (70%, 100% of homogeneous)
 In-homogeneous oscillatory: 0 (0%, 0% of in-homogeneous)
 In-phase oscillatory: 7 (70%, 100% of oscillatory)
 Out-of-phase oscillatory: 0 (0%, 0% of oscillatory)
 Mixed: 0 (0%)
 Undetermined: 0 (0%)

One can see both from the report and from the output figure (Fig. 14) that the system has determined two stable dynamical regimes: in-phase oscillatory and in-homogeneous steady state. The latter is characterized with different steady state levels of the subsystems in the phase space. For example, the second solution in the report shows the base gain equal to around 1.5 meaning that the levels of the last slopes in the two subsystems differ: one level is 1.5 times higher than the other. Note, that the amplitude gain for the SS comparisons is not important, since the steady state level amplitudes eventually have to approach, by the definition, an infinitesimal value. Finally, the statistics is shown, where different kinds of comparisons are brought together to exemplify the diversity of behaviors of the system under study.

Out-of-phase oscillations. According to [8] for $A = 2$, $B = 7$ and $D = 0.24$ the system of two coupled Brusselators contain the out-of-phase oscillation with half a period phase shift. In order to find the solution we resort to the same random initial condition technique as in the previous section, since the solution co-exist with the in-phase oscillations. The result is shown in Fig. 15, where only three trial results are given: one in-phase and two out-of-phase oscillatory regimes.

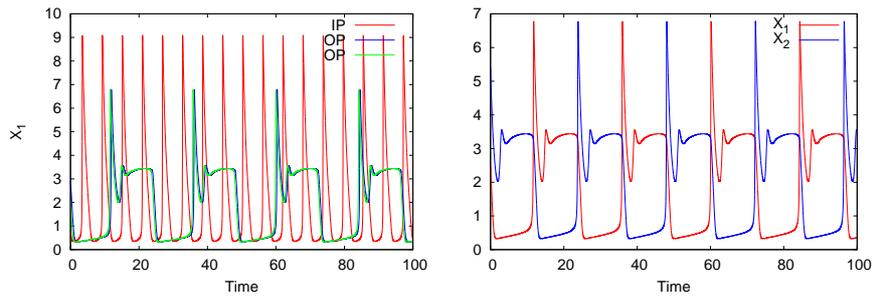


Figure 15: The random initial conditions to find the out-of-phase oscillations. Left: only three trials are shown giving in-phase (IP) and two out-of-phase (OP) trajectories. Right: out-of-phase solution with half of the period phase shift. Parameters for the system (12): $A = 2$, $B = 7$, and $D = 0.24$.

The dynamical test for the 10 random trials results in:

Dynamics report:

```

#1) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
#2) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
#3) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
#4) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
#5) OS-4(T=24.198)/OP(0.499959T)/H(bg=1,ag=1.00012)
#6) OS-4(T=24.198)/OP(0.499959T)/H(bg=1,ag=1.00008)
#7) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
#8) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
#9) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
#10) OS-1(T=5.86)/IP(OT)/H(bg=1,ag=1)
Abs.tol = 0.001, rel.tol = 0.01
-----
Regimes statistics:
-----
Number of regimes: 10
Steady States: 0 (0%)
Oscillatory: 10 (100%)
Homogeneous: 10 (100%)
In-homogeneous: 0 (0%)
Homogeneous oscillatory: 10 (100%, 100% of homogeneous)
In-homogeneous oscillatory: 0 (0%, NAN% of in-homogeneous)
In-phase oscillatory: 8 (80%, 80% of oscillatory)
Out-of-phase oscillatory: 2 (20%, 20% of oscillatory)
Mixed: 0 (0%)
Undetermined: 0 (0%)

```

From the reports it is easy to see that there were found 2 OS-4 dynamical regimes (#5 and #6) with period of 24.2. These oscillatory behaviors are classified as the homogeneous out-of-phase oscillations with phase shift of approximately half of the period. All other 8 trials gave the normal homogeneous in-phase oscillations.

Fig. 15 clearly shows the curly shape of the out-of-phase oscillations. There are 4 subperiods for the oscillatory trend. Although it cannot be clearly discerned from the figure, the 4-th subperiod contains practically small slopes in it.

In-homogeneous oscillations. The system (12) of two coupled Brusselators comprise one more interesting dynamical regime, that is in-homogeneous oscillations. The general characteristic of the behavior is the different levels of oscillations in the sub-systems. It usually takes place when the Hopf bifurcation appears on the in-homogeneous steady state solutions. The waveform of the in-homogeneous OS for the system (12) is shown in Fig. 16.

The trajectory shown in Fig. 16 is analyzed by T-system of DINAMICA giving the following output:

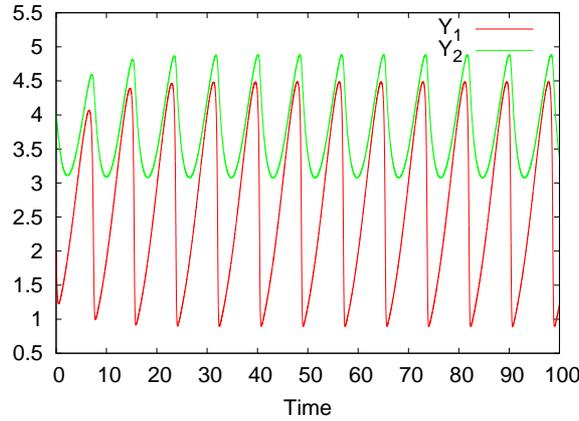


Figure 16: The in-homogeneous oscillatory solution of two coupled Brusselators. Parameters: $A = 1$, $B = 3.2$, $D = 0.57$.

```
***Checking sub-system 1(y1):
# subperiods is 1: 18 of 22 (81.8182%) comparisons support the conclusion.
OS(1): period=1,rel.tol=0.01.
```

```
***Checking sub-system 2(y2):
# subperiods is 1: 17 of 21 (80.9524%) comparisons support the conclusion.
OS(1): period=1,rel.tol=0.01.
```

```
Could not determine the lag[0] => IH?
#0) Systems lag = -1
=====
DYNAMICS:
OS-1(T=8.322)/OP(-1T)/IH(bg=3.43325,ag=1.99052)
```

T-system has checked the first sub-system (on the variable y_1) and reported about OS-1 regime found. Similarly, for the second sub-systems (y_2). Finally, the lag between the two sub-systems could not be found due to the inhomogeneity of the two. Thus, the system lag is set to be -1 as a convention for the not-found lag.

Additionally, in-homogeneous oscillatory solutions cannot be determined to have a phase shift one from another. Hence, the T-system uses conventionally the negative number (usually close to -1) for the phase shift of such dynamical behaviors. However, the base and amplitude gains can be well determined and they are reported.

6.5 Three coupled Brusselators

(NOTE: the three coupled Brusselators system is defined in `bruss3.ode` file of the package and can be found in `ode/` folder.)

Here we consider another example of coupled Brusselators, where the coupling is carried out through the common media via the slow variable y_i and three oscillators are coupled. In general, this type of the system does not differ from the previously considered 2 Brusselators, except for the number of them, since the system (12) can be brought to the form of the following system:

$$\begin{aligned}\frac{dx_i}{dt} &= A - (B + 1)x_i + x_i^2 y_i \\ \frac{dy_i}{dt} &= Bx_i - x_i^2 y_i + D\left(\frac{1}{N} \sum_{j=1}^N y_j - y_i\right)\end{aligned}\tag{13}$$

Eq. (13) represents the dynamical system of a single i -th oscillator, where the total number of oscillators $N = 3$. This system is known to have so called “wave” solution where the trajectory is a period-2 oscillation for each of the oscillators and the phase shift between the oscillators is $T/3$ (Fig. 17).

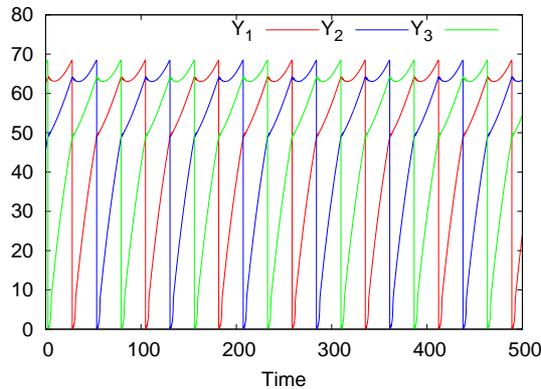


Figure 17: The so called “wave” solution to the eq. (13). Parameters used: $A = 1$, $B = 15.4$, $D = 0.046$.

T-system reports on the “wave” solution (as shown in Fig. 17, but with time 1000):

```
***Checking sub-system 1(y1):
# subperiods is 2: 42 of 49 (85.7143%) comparisons support the conclusion.
OS(2): period-2,rel.tol=0.01.
```

```
***Checking sub-system 2(y2):
# subperiods is 2: 42 of 49 (85.7143%) comparisons support the conclusion.
```

OS(2): period-2,rel.tol=0.01.

***Checking sub-system 3(y3):

subperiods is 2: 46 of 49 (93.8776%) comparisons support the conclusion.

OS(2): period-2,rel.tol=0.01.

period 1: 76.766 = 25.38 + 51.386

period 2: 76.889 = 25.43 + 51.459

period 3: 76.848 = 51.445 + 25.403

phase shift 1: 0.334341

phase shift 2: 0.332856

phase shift 3: 0.667018

=====

DYNAMICS:

OS-2(T=76.766)/OP(0.666662T)/H(bg=1.00006,ag=1.00007)

The report above reads as follows. Each sub-system was determined to have OS-2 periodic trajectory. Periods of each of the sub-system were about 76.8, divided into two subperiods: 25.4 and 51.4. Phase shifts were 0.33, 0.33 and 0.67. The first two are expected phase shifts as was noted above for the “wave” trajectory, however, 0.67 phase shift is also reported, for the mutual comparison is carried out between **all** possible pairs of oscillators, and, of course, there is a pair of oscillators that has a phase shift of double of the minimal one 0.33.

In general, this is the case for N oscillators’ system with equal phase shift between the “closest” oscillators, i.e. system is symmetric with regards to the phase shift. The minimal phase shift is going to represent this “closest” components and there will be also phase shifts of $i \times$ minimal-phase-shift, where i runs from 1 to $N - 1$.

Another example of the phase symmetric oscillations but with different periodicity inside each oscillator is shown in Fig. 18.

The shown in Fig. 18 regime is OS-3 oscillations as T-system reports:

***Checking sub-system 1(y1):

subperiods is 3: 24 of 29 (82.7586%) comparisons support the conclusion.

OS(3): period-3,rel.tol=0.01.

***Checking sub-system 2(y2):

subperiods is 3: 24 of 29 (82.7586%) comparisons support the conclusion.

OS(3): period-3,rel.tol=0.01.

***Checking sub-system 3(y3):

subperiods is 3: 24 of 29 (82.7586%) comparisons support the conclusion.

OS(3): period-3,rel.tol=0.01.

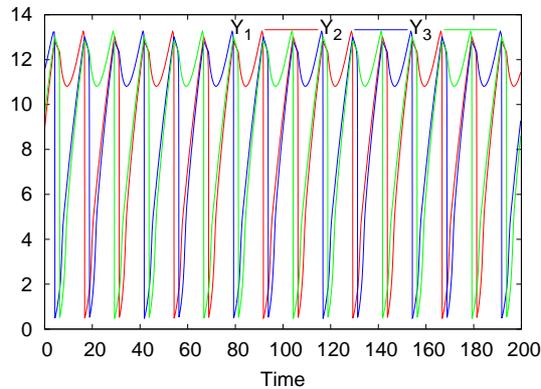


Figure 18: The OS-3 solution with $T/3$ phase shift. Parameters: $A = 1$, $B = 6$, $D = 0.234$

```

period 1: 37.493 = 12.498 + 13.05 + 11.945
period 2: 37.498 = 11.946 + 12.497 + 13.055
period 3: 37.493 = 13.055 + 11.94 + 12.498
phase shift 1: 0.333342
phase shift 2: 0.666569
phase shift 3: 0.333476

```

=====

DYNAMICS:

OS-3(T=37.493)/OP(0.666569T)/H(bg=1.00001,ag=1.00008)

Again, one can see the 0.33 to be the minimal phase shift between the oscillators, but the period is now composed of 3 subperiods, which is reliably determined for each of the sub-systems.

All regimes demonstrated in this section are homogeneous.

6.6 Two coupled genetic Repressilators

(NOTE: the two coupled Repressilators system can be found in `repB2.ode` under the `ode/` directory of the package.)

Recent example of the genetic dynamical system possessing a multitude of dynamical behaviors has been reported in [9]. The system is a further development of the original Repressilator system (see eq. (11) and [5]) and

contains the following equations [9, 10, 11]:

$$\begin{aligned}
\frac{da_i}{dt} &= \frac{\alpha}{1 + C_i^n} - a_i & \frac{dA_i}{dt} &= \beta_a(a_i - A_i) \\
\frac{db_i}{dt} &= \frac{\alpha}{1 + A_i^n} - b_i & \frac{dB_i}{dt} &= \beta_b(b_i - B_i) \\
\frac{dc_i}{dt} &= \frac{\alpha}{1 + B_i^n} - c_i + \kappa \frac{S_i}{1 + S_i} & \frac{dC_i}{dt} &= \beta_c(c_i - C_i) \\
\frac{dS_i}{dt} &= -k_{s0}S_i + k_{s1}B_i - \eta(S_i - Q\bar{S}),
\end{aligned} \tag{14}$$

where $\bar{S} = \frac{1}{N} \sum_{j=1}^N S_j$ (N is the number of oscillators).

The system (14) has the original Repressilator's three genes augmented with a small diffusive molecule carrying out the coupling function. The molecule is represented by the term S_i and affects the expression rate of the mRNA c_i . The coupling itself is realized through the diffusion of S_i . Furthermore, the contribution of all Repressilator cells is represented by the average concentration of S_i molecules \bar{S} in all cells, thus, having an assumption of fast mixing of S molecules in the media. Since the mechanism is borrowed from a clear natural example of so called quorum sensing found in bacteria, the average concentration of S molecules is further multiplied with the quorum sensing coefficient Q that takes the values from the interval $[0, 1]$. The Q parameter determines how close the cells are in the media. For example, if $Q = 1$ all cells are tightly connected and once S molecule is produced in cell i it is immediately available to all other cells. On the contrary, $Q = 0$ indicates the full dilution of S molecules in the media, i.e. cells are far away from each other. For further details, see [10].

There is a whole set of the dynamical behaviors the model demonstrates, e.g. as the quorum sensing parameter Q varies. We will follow the results reported in [10]. Until a further notice made, the default parameter set we use is: $N = 2$, $\alpha = 216$, $n = 2.6$, $\beta_a = 0.85$, $\beta_b = 0.1$, $\beta_c = 0.1$, $\kappa = 25$, $k_{s0} = 1$, $k_{s1} = 0.01$, $\eta = 2$ and we vary Q in the range $[0, 1]$.

Anti-phase oscillations. The only pure oscillatory regime the system (14) has is the anti-phase oscillations, i.e. oscillations with phase shift of half a period. So, given the default parameter set and $Q = 0.1$ we observe the anti-phase kinetics (Fig. 19).

The T-system being run on the results shown in Fig. 19 reports:

```

***Checking sub-system 1(a1):
# subperiods is 1: 16 of 17 (94.1176%) comparisons support the conclusion.
OS(1): period=1,rel.tol=0.01.

```

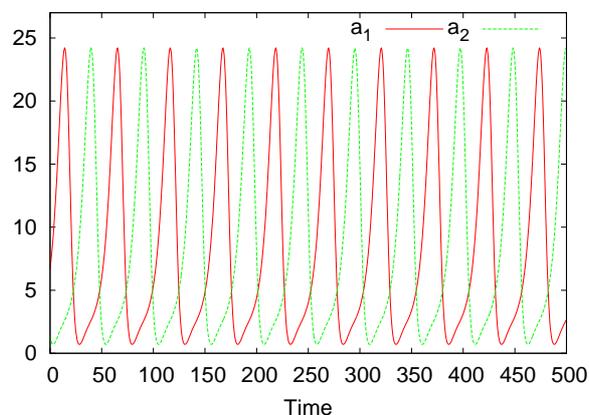


Figure 19: The anti-phase solution of the system (14). The default parameter set and $Q = 0.1$.

```
***Checking sub-system 2(a2):
# subperiods is 1: 15 of 16 (93.75%) comparisons support the conclusion.
OS(1): period=1,rel.tol=0.01.
```

```
DYNAMICS:
OS-1(T=51.08)/OP(0.5T)/H(bg=1,ag=1)
```

The above report shows the homogeneous oscillations with phase shift of 0.5 of the period, that is 51.08. Both sub-systems are reported to have OS-1 regime.

Oscillations co-exist with homogeneous steady state. If we set $Q = 0.2$ the dynamical picture of the system changes and new dynamics emerge. Namely, the stable homogeneous steady state solution in addition to the oscillatory anti-phase solution.

Let us try to identify the solution using the random initial conditions. In the following we opted to have 300 randomly thrown initial points for simulations:

```
Abs.tol = 0.001, rel.tol = 0.01
-----
Regimes statistics:
-----
Number of regimes: 300
Steady States: 298 (99.3333%)
Oscillatory: 2 (0.666667%)
Periodicity (unique): 1
Homogeneous steady state: 298 (99.3333%)
```

```

In-homogeneous steady state: 0 (0%)
Homogeneous oscillatory: 2 (0.666667%, 0.666667% of homogeneous)
In-homogeneous oscillatory: 0 (0%, NAN% of in-homogeneous)
In-phase oscillatory: 0 (0%, 0% of oscillatory)
Out-of-phase oscillatory: 2 (0.666667%, 100% of oscillatory)
Mixed: 0 (0%)
Undetermined: 0 (0%)

```

So the homogeneous steady state is a dominant regime (with the throwing radius of 200) with 298 appearances, whereas the anti-phase solution appeared only twice. The report also specified the solutions found:

```

...
#228) SS(0)/H(bg=1,ag=1)
#229) OS-1(T=49.381)/OP(0.500142T)/H(bg=1.00007,ag=1)
...

```

Thus, the same OS-1 solution found as in the previous section, but with slightly smaller period, and the steady state solution.

Three co-existing solutions. If we increase Q even further up to 0.3, we can reach the state of three different dynamical solutions co-existing with each other. Namely, these are anti-phase oscillations, homogeneous steady state and in-homogeneous limit cycle as reported in [10].

Again, using the random initial conditions we can find the aforementioned regimes. As shown below, the homogeneous steady state is dominating over other two dynamical behaviors and the in-homogeneous limit cycle appears in the test 28 times vs. 1000 random runs. Finally, there is only one initial condition led to the homogeneous anti-phase solution.

```
Abs.tol = 0.001, rel.tol = 0.01
```

```
-----
Regimes statistics:
-----
```

```

Number of regimes: 1000
Steady States: 971 (97.1%)
Oscillatory: 29 (2.9%)
Periodicity (unique): 0 1
Homogeneous steady state: 971 (97.1%)
In-homogeneous steady state: 0 (0%)
Homogeneous oscillatory: 1 (0.1%, 0.102881% of homogeneous)
In-homogeneous oscillatory: 28 (2.8%, 100% of in-homogeneous)
In-phase oscillatory: 0 (0%, 0% of oscillatory)
Out-of-phase oscillatory: 1 (0.1%, 3.44828% of oscillatory)
Mixed: 0 (0%)
Undetermined: 0 (0%)

```

The solutions found were determined with the following characteristics:

```

...
#4) OS-1(T=47.778)/OP(0.5T)/H(bg=1,ag=1)
#5) SS(0)/H(bg=1,ag=NAN)
...
#25) OS-1(T=33.743)/IH(bg=4.0355,ag=2805.98)
...

```

As one can see, there is again the dominating homogeneous steady state solution. The anti-phase solution is characterized with further decreased period. The new in-homogeneous oscillatory solution is characterized with period around 33.7 and large amplitude and base gains. This solution of the system is manifested with one system fully oscillating, while the other is oscillating with very small amplitude, resembling a steady state (this is why the amplitude gain is so big). This solution is shown in Fig. 20.

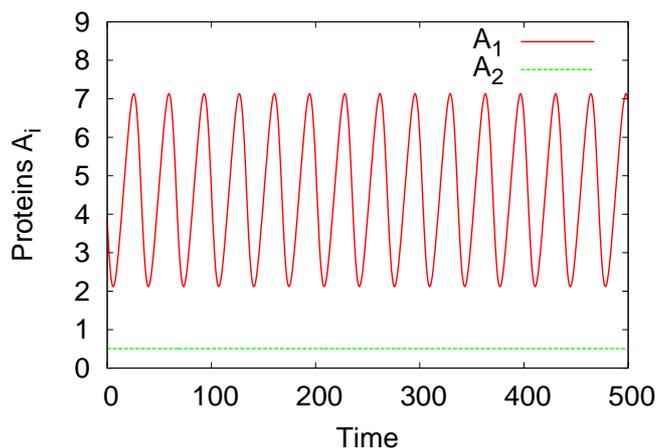


Figure 20: The in-homogeneous limit cycle oscillations.

In-homogeneous steady state appears. For $Q = 0.5$ the system has three solutions, but instead of in-homogeneous limit cycle there appears in-homogeneous steady state.

Thousand randomly thrown initial conditions give the following output:

```
Abs.tol = 0.001, rel.tol = 0.01
```

```
-----
```

```
Regimes statistics:
```

```
-----
```

```
Number of regimes: 1000
```

```
Steady States: 999 (99.9%)
```

```

Oscillatory: 1 (0.1%)
Periodicity (unique): 1
Homogeneous steady state: 995 (99.5%)
In-homogeneous steady state: 4 (0.4%)
Homogeneous oscillatory: 1 (0.1%, 0.100402% of homogeneous)
In-homogeneous oscillatory: 0 (0%, 0% of in-homogeneous)
In-phase oscillatory: 0 (0%, 0% of oscillatory)
Out-of-phase oscillatory: 1 (0.1%, 100% of oscillatory)
Mixed: 0 (0%)
Undetermined: 0 (0%)

```

There are 999 steady states, among which 995 are homogeneous and 4 are in-homogeneous. There is 1 anti-phase homogeneous oscillatory solution found. The found solutions are further explained in detail:

```

...
#399) OS-1(T=44.611)/OP(0.499989T)/H(bg=1,ag=1)
#400) SS(0)/H(bg=1,ag=NAN)
...
#792) SS(0)/IH(bg=8.66739,ag=INF)
...

```

The period of the anti-phase is further decreased to 44.6. In-homogeneous steady state is around 8.7 times higher in one cell as compared to the other.

Chaos regime. According to the study performed in [10] there is an interval of parameter Q values, where the anti-phase becomes a chaotic dynamical regime. This regime cannot be found using the slope algorithm of the T-system, since there is no stable pattern in the time series, when the system is on the chaotic attractor. This regime, however, will be determined as either **Mixed** or **Undetermined** under the T-system checking.

We put $Q = 0.7$, simulate and check the dynamics for 10000 randomly set initial conditions. The output is the following:

```

Abs.tol = 0.001, rel.tol = 0.01
-----
Regimes statistics:
-----
Number of regimes: 10000
Steady States: 9989 (99.89%)
Oscillatory: 0 (0%)
Periodicity (unique):
Homogeneous steady state: 9989 (99.89%)
In-homogeneous steady state: 0 (0%)
Homogeneous oscillatory: 0 (0%, 0% of homogeneous)

```

In-homogeneous oscillatory: 0 (0%, 0% of in-homogeneous)
Mixed: 11 (0.11%)
Undetermined: 0 (0%)

The system detected 11 **mixed regime** trajectories (this indicates presence of both stationary and non-stationary dynamics according to the T-system). One of the mixed regimes found is depicted in Fig. 21, which is referred to as **chaos** in dynamical studies. Another possibility for the chaotic regime to be detected as **Undetermined**. In general, there is no way to detect chaos with the T-system.

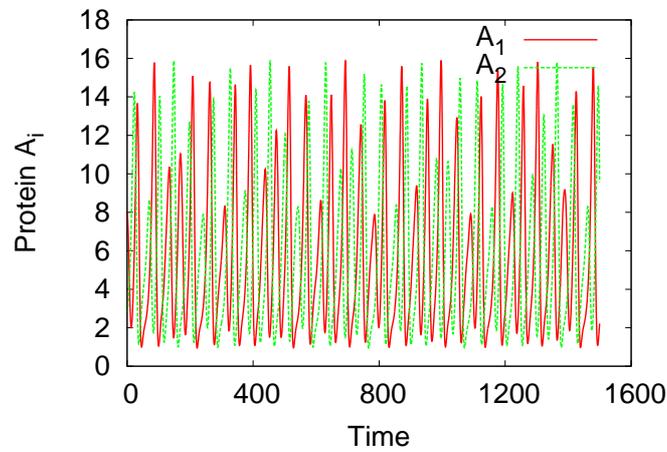


Figure 21: The chaotic regime.

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