

# **GUI Editor User Guide**

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#### Revision

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0.1.0	October 12, 2001	Harry Glasgow	Pre-release draft
0.1.5	October 15, 2001	Paolo Marrone	Added a parameter file on the command line to reflect the last change of the editor package
0.1.6	October 21, 2001	Paolo Marrone	<ul> <li>Added the description of the XML parameter file</li> <li>Added the description of the Teacher component and of the Control Panel</li> </ul>
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0.5.8	January 13, 2002	Paolo Marrone	<ul> <li>Added the drawing tool buttons</li> <li>Added the official URL of Joone (<u>www.joone.org</u>)</li> </ul>
0.5.9	January 22, 2002	Harry Glasgow	Added plugin sections
0.6	February 12, 2002	Harry Glasgow	<ul> <li>Added the use of the Monitor plugins</li> <li>Added the 'Add Noise' and 'reset Input Streams' menu items</li> </ul>
0.6.5	April 9, 2002	Paolo Marrone	<ul> <li>Added the new layers (XL I/O, nested ANN, Switch I/O)</li> <li>Updated the list of libraries needed to run the editor</li> <li>Enhanced the explanation of the Delay Layer</li> <li>Added the Help Contents menu item</li> </ul>
0.6.6	May 15, 2002	Paolo Marrone	<ul> <li>Added the Learning Switch component and the validation parameter in the Control Panel</li> <li>Added the use of the Scripting Plugin</li> </ul>
0.7.0	September 02, 2002	Harry Glasgow	<ul> <li>Added details of new sections for the two-part menu, the Macro Editor pane and the charting module.</li> <li>Added the description of the Context and Logarithmic layers</li> </ul>

# 1 Introduction

The Java Object Oriented Neural Engine (Joone) is a system of Java modules that provide a framework for developing, teaching and testing neural networks.

Joone comprises two elements, a core engine and a GUI editor. The core engine is used by the editor to process neural networks. This guide focuses chiefly upon the editor component.

### 1.1 Intended Audience

This document is intended for non-programmers, persons interested in developing neural networks with the Joone editor.

# 2 Download and Installation

### 2.1 Java

To run Joone, Java 2 (release 1.2 or above) needs to be installed on your computer. This is available from Sun Microsystems at <u>http://java.sun.com/j2se/</u>. Installation of Java falls outside the scope of this document but is relatively straightforward.

### 2.2 Joone Files

The compiled Joone project files are available from Source Forge at <u>http://sourceforge.net/projects/joone/</u>. The Joone engine and editor zip files are available from the project's Summary page. Both these files are required to run the Joone editor and should be included in your computer's Java Classpath.

### 2.3 Other Files

The Joone editor makes use of the following libraries:

- *jhotdraw.jar* from JhotDraw 5.2, a Java based drawing package
- *xalan.jar* and *crimson.jar* from JAXP, the Java XML processing package
- *jh.jar* from JavaHelp 1.1 or above
- *poi-hssf.jar, poi-poifs.jar, poi-util.jar* and *hssf-serializer.jar* from Jakarta HSSF project, the library to read/write Excel files
- *bsh-core.jar* from BeanShell (<u>www.beanshell.org</u>), a Java scripting engine

The **joone-ext.zip** package contains a version of the above libraries.

### 2.3.1 Note

- JoonEdit does not work with the new JhotDraw 5.3, only the 5.2 version. A new version of JoonEdit will soon be released that works with JhotDraw 5.3.
- Sun Microsystems has released Java Standard Edition 1.4, which includes JAXP, so it may not be necessary to include this separately for Java editions 1.4 and above.

## 2.4 Running the Joone Editor

To run the Joone editor, put all the above packages on the classpath, then JoonEdit class should be invoked with the following command:

java org.joone.edit.JoonEdit <parameters.xml>

where <parameters.xml> is the optional XML parameters file with the complete path (see /org/joone/data/layers.xml for an example). If the parameter is not specified, the editor will use a default set of parameters in the joone-edit.jar file

On a Windows platform, Click Start then Run and type in the above command, then click OK. Alternatively you can use the RunEditor.bat contained in the executable distribution package.

# 3 Menu

The menu of the Joone editor accesses standard operations.

### 3.1 File

Projects built with the Joone editor can be saved and reopened at a later date. Projects are saved as serialized files with the file extension of 'ser'. Only one project can be edited at a time. Joone will prompt to save an edited project prior to closing, or opening a new project.



### 3.1.1 New

Creates a new Joone project. Any existing project changes are lost.

### 3.1.2 Open

Opens an existing Joone project. This will replace any existing project.

### 3.1.3 Save

Allows the current Joone project to be saved as a serializable file.

### 3.1.4 Save As

Allows a Joone project to be saved as a serializable file with a different name or path.

### 3.1.5 Export Neural Net

Allows the exporting of a neural net in a serialized form (.snet). This is provided for use in the distributed environment (or for other future uses).

## 3.1.6 Print

Prints a graphical representation of the current project.

### 3.1.7 Exit

Exits the Joone editor.

## 3.2 Edit

Joone allows various editing and positional actions to be performed on a component. Edit options may not be available if the required number of components is not selected. Copy

and paste operations are only available with drawing tools and not with Joone components.

### 3.2.1 Cut

Deletes a selected component from the screen but keeps it in memory for Paste operations.

### 3.2.2 Copy

Copies a selected component from the screen to memory.

### 3.2.3 Paste

Copies a selected component from memory to the screen.

### 3.2.4 Duplicate

Duplicates a selected component.

### 3.2.5 Delete

Deletes a selected component from the screen.

### 3.2.6 Group

This menu item groups a number of components together so that they can be manipulated as a single component.

### 3.2.7 Ungroup

This ungroups a previously grouped set of components.

### 3.2.8 Sent to Back

Positions a selected component so that other components that overlap it are drawn over this one.

### 3.2.9 Bring to Front

Positions a selected component so that this component is drawn over other components that it overlaps.

### 3.3 Align

Several components can be selected concurrently by either clicking on each required component while holding down the Shift key, or by dragging a rectangle around a group of components. Once a number of components are selected, alignment menu options can be applied to align the controls relative to each other.

### 3.3.1 Toggle Snap to Grid

Turns on/off the alignment of the components on a fixed grid facilitating the arrangement of the objects on the drawing area.

### 3.3.2 Left

Aligns selected components vertically along their left hand edge.

### 3.3.3 Center

Aligns selected components vertically along their center line.

### 3.3.4 Right

Aligns selected components vertically along their right hand edge.

## 3.3.5 Тор

Aligns selected components horizontally along their top edge.

### 3.3.6 Middle

Aligns selected components horizontally along their middle.

### 3.3.7 Bottom

Aligns selected components horizontally along their bottom edge.

## 3.4 Attributes

The attributes of the drawing tools (see below) can be modified with the following commands. The following attributes can be changed:

### 3.4.1 Figure Attributes:

- Fill Color
- Pen Color
- Line's Arrows

### 3.4.2 Text Attributes

- Font
- Font Size
- Font Style
- Text Color

These commands do not affect the attributes of the neural network components (layers and connections).

## 3.5 Control

### 3.5.1 Control Panel

Properties window					
Controls					
Run	Conti	nue	Pause		
Epochs: 0					
1	RMSE: 0	).0			
le	arning	False 💌			
learnin	g rate	0.0			
m.o.	mentum	0.0			
input pa	tterns	0			
pre-le	arning	0			
	epochs	0			
vali	dation	False	-		

The Control Panel is the tool that controls the behaviour of the neural net. It contains three buttons:

Run:	Starts the neural net beginning from the first pattern of the input data set.			
Continue:	e: Restarts the neural net from the last pattern processed.			
Stop:	Stops the neural net.			
The Control	Panel parameters are:			
Epochs:	The total number of the cycles for which the net is to be trained.			
Input Patter	<b>ns</b> : The total number of input rows for which the net is to be trained. This can be different from the number of rows read from the FileInput component (lastRow $-$ firstRow $+$ 1).			
Momentum:	The value of the momentum (see the literature about the back-propagation algorithm).			
Learning Ra	te: The value of the learning rate (see the literature about the back-propagation algorithm).			
Learning:	True if the net is to be trained, otherwise set false.			
Validation:	True if the net is to be tested on a validation data set. Used ONLY in conjunction with a Learning Switch component inserted in the net.			
Pre-Learnin	<b>g</b> : The number of initial cycles skipped from the learning algorithm. Normally this parameter is zero, and is used when there is a DelayLayer component in the net. In this case pre-learning must be set equal to the number of taps of that component allowing its buffer to become filled before the learning cycle starts.			

To better explain the use of the Pre-Learning parameter, it serves to avoid making changes to the values of biases and synapses in the presence of a delayed layer. This is because if these values are altered from the first cycle of the learning process, this would adjust them using a wrong input pattern, obtained before the full input 'temporal window' is presented to the network. For instance, if an input pattern is composed of one column as follows:

0.2

0.5

0.1

and an input delay layer with taps = 3 is present, then when the network might only have read the first two input values (0.2 and 0.5), and the output of this first layer would be:



In this case the network would learn the wrong  $\{0, 0, 0.2, 0.5\}$  pattern. Thus the Pre-Learning parameter must be set equal to the taps parameter so that the network starts to learn only when all the 'taps' values have been read.

### 3.5.2 Add Noise

This adds a random noise component to the net and is useful for allowing the net to exit from a local minimum. At the end of a round of training, adding noise may 'jolt' the netork out of a local minimum so that further training produces a better network.

### 3.5.3 Randomize

This reset the weights of a neural network, initializing it to a random state.

### 3.5.4 Reset Input Streams

This command resets all the buffered input streams in input and teacher layers, permitting the reloading of their buffers with the input data. This is useful after the contents of some files have changed and it is necessary to reload the data.

### 3.5.5 Macro Editor

A powerful scripting management engine is provided in the Joone Editor. Before describing this feature, it is important to understand the concepts underlying the scripting engine.

There are two types of script in Joone:

*Event-driven scripts* These are executed when a neural network's event is raised. *User-driven scripts* These are executed manually by the user.

Both of the above type of scripts are contained within the neural network, and are serialized, stored and transported along with the neural network that contains them (in the same way that macros defined in a MS Word<sup>®</sup> of Excel<sup>®</sup> document are). In this manner the defined scripts can be executed even when the net is running on remote machines outside the GUI editor.

It is possible to load, edit, save and run any script (in this document also referred to as Macro) using a user friendly graphical interface. The Macro Editor can be opened with the Control $\rightarrow$ Macro Editor menu item. A window will be shown, as depicted in the following figure:



In the Macro Editor, the user can define both of the two script types. On the left side of the editor there is the list of the available scripts for the neural network. For a new network, the list already contains the definitions of all the available *event-driven* scripts. To edit a script, the user simply clicks on the desired script on the list and inserts the code in the text area on the right.

The text area has some useful features to help with the writing of the BeanShell code:

- Coloured code based on the Java syntax (comments, keywords, literals, etc.)
- Highlighting of matching opening/closing brackets and parenthesise.
- Highlighting of the current edited row of code.
- Different colours used to emphasize unclosed strings.

```
/* this is a comment
  on more lines */
void someFunction (int max) {
    newNN = new NeuralNetwork();
    // This is a comment on a single line
    for (int i=0; i < max; ++i) {
        layer = new SigmoidLayer();
        layer.setLayerName("This is a not-closed string !);
        newNN.addLayer(layer);
    }
    newNN.setLayerName("This is a correctly closed string");
}
```

Note in the above figure the bold Java keywords "**new**" and "**for**", the small box indicating the corresponding open bracket matching the one near to the cursor, the greyed row indicating the current edited row, and the red colour used to indicate an unterminated string.

The *event-driven* scripts can not be renamed nor deleted from the list. If the user does not want to define an *event-driven* script, s/he can just remove or comment the corresponding code in the text area.

The *user-driven* scripts can be added, renamed or deleted by choosing the corresponding menu item in the 'Macro' menu.

#### 3.5.5.1 The Macro Editor Menu

Here are details of all the menu items of the Macro Editor frame.

#### File

Import Macro	The content of a text file can be imported into the text area
	of the selected script. The old text will be replaced.
Save Macro as	The content of the selected script can be exported into a
	text file.

	Close	Closes the Macro Editor window.
Edit		
	Cut	Copies the selected text into the clipboard and delete it from the text area (you can also use Ctrl-X).
	Сору	Copies the selected text into the clipboard (you can also use Ctrl-C).
	Paste	Pastes the content of the clipboard into the text area starting at the cursor position (you can also use Ctrl-V).
	Select All	Selects all the content of the text area.
Macr	0	
	Enable Scripting	If checked, this enables the execution of the <i>event-driven</i> scripts for the edited neural network. If not checked, all the events of the net will be ignored.
	Add	Adds a new <i>user-driven</i> script (the user cannot insert new <i>event-driven</i> scripts).
	Remove	Removes the selected <i>user-driven</i> script. Disabled for <i>event-driven</i> scripts.
	Rename	Permits renaming of the selected <i>user-driven</i> script. Disabled for <i>event-driven</i> scripts.
	Run	Runs the selected script.
	Set Rate	Sets the execution rate (the number of free training cycles between two execution calls) for <i>cyclic event-driven</i> scripts. The <i>cyclic event-driven</i> scripts are the 'cycleTerminated' and the 'errorChanged' scripts.
		<b>Warning:</b> The default value of rate for a new network is 1 (one), but it is recommended tat the value be set to between 10 and 100 (or even more) to ensure that there is sufficient processing time available for the running of neural network.

#### 3.5.5.2 Macro Scripting Features

The following section describes some characteristics of the scripting feature added to Joone's engine.

#### How to use internal Joone objects

To obtain a reference to the internal neural network's objects use:

- **jNet** to access to the edited org.joone.net.NeuralNet object
- **jMon** to access to the contained org.joone.engine.Monitor object

For example:

- **jNet**.getLayers().size() returns the number of the layers contained in the neural network.
- **jMon**.getGlobalError() returns the last RMSE value from the net.

A list of the callable public methods for the above two objects is available in the project's javadoc html files. To use the objects from the Joone libraries, it is not necessary to import the corresponding package. The following packages are imported automatically for you by the script's engine:

```
org.joone.engine.*
org.joone.engine.learning.*;
org.joone.edit.*
org.joone.util.*
org.joone.net.*
org.joone.io.*
org.joone.script.*;
```

For instance, to create a new sigmoid layer, simply write:

```
newLayer = new SigmoidLayer();
```

#### How to call another script from within a macro

Within a macro it is possible to call another script contained in the neural network (both *user* and *event-driven* scripts).

To do this, use the following code:

```
name = "NameOfTheMacroToCall";
macro = jNet.getMacroPlugin().getMacroManager().getMacro(name);
eval(macro);
```

#### The scope of the script's variables

All the scripts defined in a neural network (both *user* and *event-driven* scripts) share the same namespace and actual-values storage. Thus a global variable declared and initialised in *script\_A* can be accessed in *script\_B*.

For example:

1. Add a macro named 'macro1' and insert into it the code: myVariable = 5;

2. Add a macro named 'macro2' and insert into it the code: print("The value is:
 "+ myVariable);

Run first the macro1 and then the macro2; you will see into the standard output console the result: "The value is: 5"

For further details about scope and the use of variables, see the BeanShell manual.

### 3.6 Look'n'Feel

The default 'look and feel' of the Joone editor is Metal. Currently there is no way to make another style the default.

### 3.6.1 Metal

Selecting this menu option displays the editor in the Metal GUI style.

### 3.6.2 CDE/Motif

Selecting this menu option displays the editor in the CDE/Motif GUI style.

### 3.6.3 Windows

Selecting this menu option displays the editor in the Windows GUI style.

## 3.7 Help

### 3.7.1 About Joone

This option displays the Joone Editor About screen. The current version of the Editor and of the Engine being used is also displayed. Version numbers are of the form 'major-release.minor-release.build', i.e. 1.2.5. If an older, incompatible engine version is detected, a warning will be displayed in the About screen, as backward compatibility is not guaranteed between editor and engine versions.

### 3.7.2 Help Contents

This option displays the on line help of the editor.

# 4 Toolbar

The tool bar buttons are divided into two palettes. One contains all the drawing buttons, while the other contains all the construction components, as shown in the following figure:



The content of the drawing palette is determined by the Joone application while the component panel is configurable by modifying the layers.xml file. Please note that not all of the following components may appear by default in the Joone Editor application due to limited space on the tool bars. See the chapter on the XML parameter file for details on how to alter the items that appear in the toolbar.

## 4.1 Selection Tool

### $\Diamond$

The Selection Tool allows Layers to be selected and dragged around the screen. It is also use to create links between Layers.

# 4.2 Label and Connected Label

# A A

The Label tool allows text comments to be placed on the screen.

The Connected Label tool allows the addition of text to each drawing tool. The attached text will follow the figure's movements.

## 4.3 Drawing Tools



These tools permit the addition of several figures to the drawing panel of the GUI editor. They will be saved along with the neural network (Save menu item) and then restored on the drawing panel (Open menu item).

## 4.4 Input Layers



The New File Input Layer, New URL Layer, New Excel Input Layer, the Switch Input Layer and the Learning Switch tools allow new input layers to be added to the screen.

## 4.5 Layers

### ┉<sup>┿</sup>╷┙<sup>┿</sup>╷<sup>┿</sup>╷┙<sup>┿</sup>

The New Linear Layer, New Sigmoid Layer, New Tanh Layer, New Logarithmic Layer, New Context Layer, New Nested ANN Layer and New Delay Layer tools allow new processing layers to be added to the screen.

### 4.6 Output Layers

🤨 🔁 🗶 🗅

The New Switch Output Layer, New File Output Layer, New Excel Output Layer and New Teacher Layer tools allow new output layers to be added to the screen.

## 4.7 Charting Component

 $\sim$ 

This component belongs to the output components family, so it can be used anywhere that makes sense to insert an output component. For instance, it can be attached to the output of a Layer, or to the output of a Teacher component.

The charting component has the following properties:

maxXaxis:	The maximum value of the X axis. Set this value to the maximum number of the samples required to visualise in the chart.
maxYaxis:	The maximum value of the Y axis. Set this value to maximum value you expect to display in the chart.
name:	The name of the chart component.
resizable:	If true, the window containing the chart will be resizable. The resizing of
	the window will be rescale the entire chart, adapting itself to the new size
	of the frame.
show:	Used to show or hide the window containing the chart.
title:	The text shown in the title bar of the window containing the chart.
serie:	Indicates what series (column) in a multicolumn output data is to be displayed

All the above properties are updateable at any time, including during the running of the network making it possible to show the chart at several sizes or resolutions. This is an example of the use of the charting component:



In the above example, the charting component is attached to the output of a teacher component to capture and display the RMSE values during the training phase. The maxXaxis property is set to the number of the training epochs, while the maxYaxis is set to the max value we want to show in the chart.

The user can change either of these values at any time to show a zoomed view of the chart.

## 4.8 Synapses

## **₩₩**<sup>1</sup>

These components allow the user to choose the type of synapse that connects two layers. In the toolbar there are three buttons as above, representing respectively a Full Synapse, a Direct Synapse and a Delayed Synapse. More type will be added to future versions of Joone.

- A Full Synapse will connect every output connection of one layer to this input of every neuron in another layer.
- A Direct Synapse will connect each output connection of one layer to exactly one neuron in the other layer. The number of outputs of the first layer must match the number of neurons in the second layer, or an exception will be generated when the net is started.
- A Delayed Synapse behaves as a full Synapse where each connection is implemented with a FIRFilter object. In this connection is implemented the temporal backpropagation algorithm by Eric A. Wan, as in *'Time Series Prediction by Using a Connectionist Network with Internal Delay Lines*' in Time Series Prediction. Forecasting the Future and Understanding the Past, by A.Weigend and N.Gershenfeld. Addison-Wesley, 1994.

To use these components, the firstly selects the tool button corresponding to the synapse required, then drags a line from one layer to another in the drawing area. The new inserted synapse will be shown with an arrow containing a small box at its centre, as in the following figure:



The box contains a label indicating the type and state of the synapse. The available types are shown the following table:



If the little box is greyed, then the synapse is disabled, indicating that this branch of the net is interrupted. To disable a synapse, right-click it and set the 'Enabled' property in the property panel to **false**. This feature is very useful in the designing of a neural network to try several architectures on the fly.

The neural network contained in the file org/joone/samples/synapses/synapses.ser contains an example using synapses currently available in JoonEdit.

#### Warning:

- Use the above tool buttons ONLY to connect two layers and **NOT** to connect any other component such as input or output components, plugins, etc. Doing otherwise will cause an exception.
- The basic method to connect two layers is to simply drag an arrow from the right handle of a layer to the other, which default to a Full Synapse.

• Neural networks saved with a previous version of JoonEdit will continue to work but the synapses will be shown unlabelled. This should not create confusion because in previous releases of the Joone editor only full synapses was created. Popup menu will work on older versions.

### 4.8.1 The Pop-Up Menu for all Connection Types

Right-clicking on any connection displays a pop-up menu as per other components.



The menu contains two items:

**Properties**To show the property panel for the synapse selected.**Delete**To delete the connection.

NB

- The Properties item is not shown if the line selected does not contain a synapse, for instance the arrow connecting an input component to a layer.
- The Delete item is included with all pop-up menus.

## 4.9 Plugins

## ∽√ 🔽 ₩ 😗 🧐

The New Center Zero Layer, New Normalize Layer and the New Turning Point Layer tools, as well as Linear, Dynamic Annealing Monitor allow new plugin layers to be added to the screen.

## 5 Layers

There are numerous layer types available in the Joone editor.

### 5.1 Processing Layers

These layers contain processing neurons. They consist of a number of neurons as set by the layer's row parameter. The differences between layer types are described in this section.

### 5.1.1 Linear

The output of a linear layer neuron is the sum of the input values, scaled by the beta parameter. No transfer function is applied to limit the output value, the layer weights are always unity and are not altered during the learning phase.

### 5.1.2 Sigmoid

The output of a sigmoid layer neuron is the sum of the weighted input values, applied to a sigmoid function. This function is expressed mathematically as:

 $y = 1 / (1 + e^{-x})$ 

This has the effect of smoothly limiting the output within the range 0 and 1.

#### 5.1.3 Tanh

The tanh layer is similar to the sigmoid layer except that the applied function is a hyperbolic tangent function. This function is expressed as:

$$y = (e^{x} - e^{-x}) / (e^{x} + e^{-x})$$

This has the effect of smoothly limiting the output within the range -1 and 1.

During training, the weights of the Sigmoid and Tanh layers are adjusted to match teacher layer data to the network output.

### 5.1.4 Logarithmic

The logarithmic layer is similar to the sigmoid layer except that the applied function is a logarithmic function. This function is expressed as:

У	=	log(1	+	x)	if	х	>=	: 0
У	=	log(1	—	x)	if	х	<	0

This layer is useful to avoid to saturate the input of the layer in presence of input values near the extreme points 0 and 1.

### 5.1.5 Context

The context layer is similar to the linear layer except that it has an auto-recurrent connection between its output and input, like depicted in the following figure:



Its activation function is expressed as:  $y = b * (x + y^{(t-1)} * w)$ 

where :

b = the beta parameter (inherited from the linear layer)

w = the fixed weight of the recurrent connection (not learned)

The w parameter is named 'timeConstant' in the property panel because it back-

propagates the past output signals and, as its value is less than one, the contribute of the past signals decays slowly toward zero at each cycle.

In this manner the context layer has a own 'memory' embedded mechanism. This layer is used in recurrent neural networks like the Jordan-Elman NNs.

#### 5.1.6 Nested ANN

The nested neural network layer permits an entire neural network to be added to the network being edited. Using this component, it is possible to build modular neural networks, constituted of several pre-built neural networks, allowing complex, compound 'organisms' to be created. The parameter 'Nested ANN' must be filled with the name of a serialized neural network (saved with the File->Export neural net menu item).

**Note:** A neural network to be used in a nested ANN component, must be composed solely of processing elements and not file I/O and/or Teacher layers.

### 5.1.7 Delay

The delay layer applies the sum of the input values to a delay line, so that the output of each neuron is delayed a number of iterations specified by the taps parameter.

To understand the meaning of the taps parameter, the following picture that contains two different delay layers, one with 1 rows and 3 taps, and another with 2 rows and 3 taps:



the delay layer has:

- the number of inputs equal to the rows parameter
- the number of outputs equal to the rows \* (taps + 1)

The taps parameter indicates the number of output delayed cycles for each row of neurons, plus one because the delayed layer also presents the actual input sum signal Xn(t) to the output. During a training phase, error values are fed backwards through the delay layer as required.

This layer is very useful to train a neural network to predict a time-series, giving it a 'temporal window' of the input raw data.

### 5.2 I/O Layers

The I/O (Input / Output) layers represent interfaces between the processing layers of a neural network and the external environment, providing the net with the data needed for processing and/or training.

### 5.2.1 File Input

The file input layer allows data in a file to be applied to a network for processing. Data for processing is expected as a number of rows of semicolon-separated columns of values. For example, the following is a set of three rows of four columns:

0.2;0.5;0.6;0.4 0.3;-0.35;0.23;0.29 0.7;0.99;0.56;0.4

Each value in a row will be made available as an output of the file layer, and the rows will be processed sequentially by successive processing steps of the network.

As some files may contain information additional to the required data, the parameters firstRow, lastRow, firstCol and lastCol may be used to define the range of useable data. The filename parameter specifies the file that is to be read from.

### 5.2.2 URL Input

The URL input layer allows data from a remote location to be applied to a network for processing. The allowed protocols are http and ftp. The data format is the same as for the FileInput layer.

### 5.2.3 Excel Input

The Excel Input layer permits data from an Excel file to be applied to a neural network for processing. Its 'sheet' parameter allows the name of the sheet to be chosen from which the input data is read.

### 5.2.4 Switch Input

The switch input allows the choice of which input component is to be connected to the neural network, choosing between all the input components attached to it. The user, after having attached several input components to its input, can set the 'active input' parameter with the name of the chosen component that is to be connected to the net. The 'default input' parameter must be filled with the name of the default component (the one activated when the user selects the 'Control->Reset Input Streams' menu item).

The switch input component, along with the output switch layer, permits dynamic changing of the architecture of the neural network, changing the input and/or output data layers attached to the neural network at any time. This is useful to switch the input source, for instance, between the file containing the training data set and the file containing the validation data set to test the training of the neural network, as depicted in the following screen shot:



### 5.2.5 Learning Switch

The learning switch is a special implementation of the Switch Input component and can be used to attach both a training data set and a validation data set to the neural net. In this way the user can test the generalization property of a neural network using a different data set to the one used during the training phase.

The training input data set can be attached by dragging an arrow from the input component to the learning switch, while the validation input data set can be attached simply by dragging an arrow from the red square on top of the learning switch to the input component containing the validation data set. To switch between them, simply change the value of the 'validation' parameter shown in the Control Panel.

This component has two properties:

- trainingPatterns: Must be set to the number of the input patterns constituting the training data set.
- validationPatterns: Must be set to the number of the input patterns constituting the validation data set.

Both these above parameters are obtained from the following formula

*lastRow - firstRow + 1* 

where the last/firstRow variables contain the values of the corresponding parameters of the attached input components.

The following figure depicts the use of this component:



**Warning**: Because a validation data set will also be required for the desired data, this component must be inserted both before the input layer of the neural network and between the Teacher layer and the *desired* input data sets.

### 5.2.6 File Output

The file output layer is used to convert the results of a processing layer to a text file. The filename parameter specifies the file that the results are to be written to. Results are written in the same semicolon-separated form as file input layers.

### 5.2.7 Excel Output

The Excel output layer is used to write the results of a processing layer to an Excel formatted file. The filename parameter specifies the file that the results are to be written to. The 'sheet' parameter allows the name of the sheet to be chosen, to which the input data is to be written.

### 5.2.8 Switch Output

The switch output permits the choice of which output component is to be connected to the neural network, choosing between all the output attached components. The user, after having attached several components to its output, can set the 'active output' parameter with the name of the chosen component that is to be connected to the net. The 'default output' parameter must be filled with the name of the default component (which one activated when the user selects the 'Control->Reset Input Streams' menu item).

### 5.2.9 Teacher

The Teacher layer allows the training of a neural net, using the back-propagation learning algorithm. It calculates the difference between the actual output of the net and the expected value from an input source. It provides this difference to the output layer for the training.

To train a net, add a Teacher component, connecting it to the output layer of the net, and then connect an input layer component to it, linked to a source containing the expected data (see figure).



# 6 Plugins

There are three types of pre-processing plugins for the input data, and two monitor plugins. A connection to a plugin can be added by dragging an arrow from the magenta square handle on the bottom side of an input layer, as depicted in the following figure:

	ile Input
0	
C	enter on Zero
In	nutPlugin 8

### 6.1 Pre-Processing Plugins

There are three pre-processing plugins implemented, but others can be implemented by extending the org.joone.util.ConverterPlugIn class:

Centre on Zero	This plugin centres the entire data set around the zero axis
	by subtracting the average value.
Normalizer	This plugin can normalize an input data stream within a
	range determined by its <i>min</i> and <i>max</i> parameters.
<b>Turning Points Extractor</b>	This plugin extracts the turning points of a time series,
	generating a useful input signal for a neural net,
	emphasising the relative max and min of a time series
	(very useful to extract buy and sell instants for stock
	forecasting). Its minChangePercentage parameter indicates
	what the minimum change around a turning point should
	be to consider it a real change of direction of the time
	series. Setting this parameter to a relative high value helps
	to reject the noise of the input data.

Every plugin has a common parameter named *serie*. This indicates what series (column) in a multicolumn input data is to be affected (0 = all series).

A plugin can be attached to an input layer, or to another plugin so that pre-processing modules can be cascaded.

If both centre on zero and normalize processing is required for an input stream, the centre on zero plugin can be connected to a file input layer, and then a normalizer plugin attached to this, as shown in the following figure:



#### 6.2 Monitor Plugins

There are also two Monitor Plugins. These are useful for dynamically controlling the behaviour of control panel parameters (parameters contained in the org.joone.engine.Monitor object).

The **Linear Annealing** plugin changes the values of the learning rate (LR) and the momentum parameters linearly during training. The values vary from an initial value to a final value linearly, and the step is determined by the following formulas:

step = (FinalValue - InitValue) / numberOfEpochs
LR = LR - step

The **Dynamic Annealing** plugin controls the change of the learning rate based on the difference between the last two global error (E) values as follows:

If E(t) > E(t-1) then LR = LR \* (1 - step/100%).

If  $E(t) \le E(t-1)$  then LR remains unchanged.

The 'rate' parameter indicates how many epochs occur between an annealing change. These plugins are useful to implement the annealing (hardening) of a neural network, changing the learning rate during the training process.

With the Linear Annealing plugin, the LR starts with a large value, allowing the network to quickly find a good minimum, and then the LR reduces permitting the found minimum to be fine tuned toward the best value, with little the risk of escaping from a good minimum by a large LR.

The Dynamic Annealing plugin is an enhancement to the Linear concept, reducing the LR only as required, when the global error of the neural net augments are larger (worse) than the previous step's error. This may at first appear counter-intuitive, but it allows a good minimum to be found quickly and then helps to prevent its loss.

### 6.2.1 The Annealing Concept



To explain why the learning rate has to diminish as the error increases, look at the above figure:

All the weights of a network represent an error surface of n-dimensions (for simplicity, in the figure there are only two dimensions). To train a network means to modify the connection weights so as to find the best group of values that give the minimum error for certain input patterns.

In the above figure, the red ball represents the actual error. It 'runs' on the error surface during the training process, approaching the minimum error. Its velocity is proportionate to the value of the learning rate, so if this velocity is too high, the ball can overstep the absolute minimum and become trapped in a relative minimum.

To avoid this side effect, the velocity (learning rate) of the ball needs to be reduced as the error becomes worse (the grey ball).

## 7 Basic Tutorial

This tutorial creates a simple network connecting a file input layer containing four examples of two input values to a file output layer via a linear layer containing two neurons.

- 1. Using a text editor, create a new file and add four lines as follows:
  - 0.2;0.3 0.4;0.5 0.6;0.8 0.9;1.0
- 2. Save the file to disk (e.g. c:\temp\sample.txt).

- 3. Start JoonEdit and insert a new linear layer. Click on this layer to display the properties page. Set the rows value to 2.
- 4. Insert a File Input layer to the left of the linear layer, then click on it to display the properties window:
  - Set the firstCol parameter to 1.
  - Set the lastCol parameter to 2.
  - Enter c:\temp\sample.txt in the fileName parameter.
  - Leave the firstRow as 1 and the lastRow as 0 so that the input layer will read all the rows in the file.
- 5. Connect the input layer to the linear layer dragging a line from the little circle on the right hand side of the input layer, releasing the mouse button when the arrow is on the linear layer.
- 6. Now insert a File Output layer to the right of the linear layer, click on it and insert into the properties window:
  - "c:\temp\sampleout.txt" on the fileName parameter.
- 7. Connect the linear layer to the file output layer by dragging a line from the little circle on the right hand side of the linear layer, releasing the mouse button when the arrow is on the file output layer.
- 8. At this stage the screen should look similar to this:



- 9. Click on the 'Net->Control Panel' menu item to display the control panel. Insert the following:
  - Set the totCicles parameter to 1. This will process the file once.
  - Set the patterns parameter to 4. This sets the number of example rows to read.
  - Leave the learningRate and the momentum fields unchanged. These parameters are used for training a net.
  - Also set the learning parameter to FALSE, as the net is not being trained.
- 10. Click the START button
- 11. A file named c:\temp\sampleout.txt will be written with the results.
- 12. If you want, you can save the net to disk with the 'File->Save As' menu item, and reload it later with 'File->Open'.

## 8 An Advanced Example: The XOR Problem

This example illustrates a more complete construction of a neural net to teach the classical XOR problem.

In this example, the net is required to learn the following XOR truth table:

Input 1	Input 2	Output
0	0	0
0	1	1
1	0	1
1	1	0

So, we must create a file containing these values:

0.0;0.0;0.0 0.0;1.0;1.0 1.0;0.0;1.0 1.0;1.0;0.0

A semicolon needs to separate each column. The decimal point is not mandatory if the numbers are integer.

Create this file with a text editor and save it on the file system (for instance  $c:\joone\xor.txt$  in a Windows environment).

Now we'll build a neural net like this:



Run the editor, and execute these following steps:

- 1. Add a new sigmoid layer is and set its layerName to 'Input' and the rows parameter to 2
- 2. Add a new sigmoid layer, and set its layerName to 'Hidden' and the rows parameter to 3
- 3. Add a new sigmoid layer, and set its layerName to 'Output', leaving the rows parameter as 1

- 4. Connect the input layer to the hidden layer by dragging a line from the little circle on the right hand side of the input layer, releasing the mouse button when the arrow is on the hidden layer
- 5. Repeat the above step connecting the hidden layer to the output layer

At this stage the screen should look similar to this:



- 6. Insert a File Input layer by to the left of the input layer, then click on it to display the properties window:
  - Set the firstCol parameter to 1
  - Set the lastCol parameter to 2
  - Enter c:\joone\xor.txt in the fileName parameter
  - Leave the firstRow as 1 and the lastRow as 0 so that the input layer will read all the rows in the file
- 7. Connect the File Input to the input layer
- 8. Insert a Teacher layer to the right of the output layer
- 9. Connect the output layer to the Teacher layer

Now we must provide the desired data to the teacher (the last column of the file xor.txt) to train the net:

- 10. Insert a File Input layer is above of the Teacher layer, then click on it to display the properties window:
  - Set the firstCol parameter to 3
  - Set the lastCol parameter to 3
  - Enter c:\joone\xor.txt in the fileName parameter
  - Leave the firstRow as 1 and the lastRow as 0 so that the input layer will read all the rows in the file
  - Set the name parameter to 'Desired data'
- 11. Connect the Teacher layer to that last File Input layer by dragging a line from the little red square on the top side of the Teacher layer, releasing the mouse button when the yellow arrow is on the last inserted File Input layer

At this stage the screen should look similar to this:



- 12. Click on the 'Net->Control Panel' menu item to display the control panel. Insert the following:
  - $\circ~$  Set the totCicles parameter to 10,000. This will process the file 10,000 times
  - $\circ$  Set the patterns parameter to 4. This sets the number of example rows to read.
  - $\circ$  Set the learningRate parameter to 0.8 and the momentum parameter to 0.3
  - Set the learning parameter to TRUE, as the net must be trained.
  - Set the validation parameter to FALSE, as the net is not being tested

13. Click the START button, and you'll see the training process starting.

The Control Panel shows the cycles remaining and the current error of the net. At the end of the last cycle, the error should be very small (less than 0.1), otherwise click on the Net->Randomize menu item (to add a little 'noise' to the weights of the net) and click the START button again. If required, you can save the net with the 'File->Save As...' menu item, so you can reuse the net later by loading it from the file system.

### 8.1 Testing the Trained Net

To test the trained net:

- 14. Add an Output File layer on the right of the output layer, click on it and insert into the properties window:
  - "c:\temp\xorout.txt" on the fileName parameter
- 15. Connect the output layer of the net to the File Output layer.
- 16. Select the line that connects the output layer to the Teacher layer and click on the Edit->Delete to disconnect the Teacher from the neural net.
- 17. On the Control Panel change the following:
  - Set the totCicles parameter to 1. This will process the input file once.
  - Set the learning parameter to FALSE, as the net is not being trained.
- 18. Click on the START button.
- 19. Open the xorout.txt file with an editor, and you'll see a result like this (the values can change from one run to another, depending on the initial random values of the weights):
  - 0.02592995527027603 0.9664584492704527 0.9648193164739925 0.03994103766843536

This result shows that the neural net has learned the XOR problem, providing the correct results:

- a value near zero when the input columns are equal to 0;0 and 1;1
- a value near one when the input columns are equal to 0;1 and 1;0

### 9 The XML Parameter File

This section explains how to modify the tool palette to personalize and extend the editor with new components.

The following sections are provided in the layers.xml file, provided with the editor package. By default he Joone Editor will look for the file org/joone/data/layers.xml as the parameter file, but this behaviour can be overridden by specifying the parameter file as a parameter to the run command:

java org.joone.edit.JoonEdit /some\_path/your\_parameter\_file.xml

The layers.xml file provided with Joone includes all the available component types in the Joone project although some of the less-used ones are commented out.

#### 9.1 The <buttons> ... </buttons> Section

This section contains all the buttons that are present in the toolbar.

<layer type="..." class="..." image="..." />

This section describes a layer. Its parameters are:

type:	The name of the layer that will be shown when the mouse passes over it.
class:	The name of the class (complete with the package 'xxx.yyy.zzz.classname'). The
	class must extend the org.joone.engine.Layer class and must be in the JVM
	classpath.
image:	The name of the image of the toolbar button. The name of the file should be:
	"org.joone.images." + image_name + ".gif", so the searched file name will be:
	/org/joone/images/image_name.gif (only gif images are allowed in the actual
	version of the editor).

<input\_layer type="..." class="..." image="..." />

This section describes an input layer. All parameters are the same as above. The input layer class must extend the org.joone.io.StreamInputSynapse and must be in the classpath.

<input\_switch type="..." class="..." image="..." />

This section describes an input switch layer. All parameters are the same as above. The input switch layer class must extend the org.joone.engine.InputSwitchSynapse and must be in the classpath.

<learning\_switch type="..." class="..." image="..." />

This section describes a learning switch layer.

All parameters are the same as above. The learning switch layer class must extend the org.joone.engine.learning.LearningSwitch and must be in the classpath.

<output\_layer type="..." class="..." image="..." />

This section describes an output layer. All parameters are the same as above. The output layer class must extend the org.joone.io.StreamOutputSynapse and must be in the classpath.

<output\_switch type="..." class="..." image="..." />

This section describes an output switch layer. All parameters are the same as above. The output switch layer class must extend the org.joone.engine.OutputSwitchSynapse and must be in the classpath.

< teacher\_layer type="..." class="..." image="..." />

This section describes a teacher layer. All parameters are the same as above. The teacher layer class must extend the org.joone.engine.learning.TeachingSynapse and must be in the classpath.

<input\_plugin type="..." class="..." image="..." />

This section describes an input plugin for the data pre-processing. All parameters are the same as above. The input plugin class must extend the org.joone.util.ConverterPlugin class and must be in the classpath.

<monitor\_plugin type="..." class="..." image="..." />

This section describes a monitor plugin to control the behaviour of the training process. All parameters are the same as above. The monitor plugin class must extend the org.joone.util.MonitorPlugin class and must be in the classpath.

<synapse type="..." class="..." label="..." image="..." />

This section describes a synapse to connect two layers together. This tag has the same properties as the other components, plus a 'label' property to set the label shown in the little box. The label is not mandatory but its use is recommended to visually distinguish various types of synapse displayed in the drawing area. If no label is specified, the synapses will be drawn as an arrow. The synapse component must extend the org.joone.engine.Synapse class.

#### 9.2 The <options> ... </options> Section

This section contains all the parameters that control the behaviour of the GUI editor. Currently it contains the following tags:

<refreshing\_rate value="nnn" />

This parameter controls the refreshing rate of the fields shown in the control panel during the training. To speed up the training process, set this value to a high value (100 or more).

#### 9.3 Separators

Small spaces can be inserted between toolbar buttons by inserting separator tags at the required position in the layers file thus: <separator/>

#### Temporarily Removing Items 9.4

Any item can be temporarily removed from the toolbar by enclosing the item in comment tags <!-- and -->. For example, to temporarily remove a separator from the toolbar, edit the separator tag...

<separator/>

...to...

<!--<separator/>--> ... or alternatively... <!--<separator/-->

# **10 Online Resources**

### 10.1 Joone

Home page: <u>http://www.joone.org/</u> Summary page: <u>http://sourceforge.net/projects/joone/</u>

10.2 Artificial Neural Networks Technology

 $\underline{http://www.dacs.dtic.mil/techs/neural/neural\_ToC.html}$ 

10.3 Java

http://java.sun.com/

10.4 JAXP

http://java.sun.com/xml

### 10.5 JHotDraw

http://sourceforge.net/projects/jhotdraw/

10.6 Source Forge

http://www.sourceforge.net

## 10.7 Sun Microsystems

http://www.sun.com/

# 11 Glossary

### 11.1 Ann / NN

Artificial Neural Network. Synonymous throughout this document with the term neural network, an interconnected network of simulated neurons.

### 11.2 Classpath

The environment variable listing the files and directories that Java will search for runtime classes. On a Windows system, this variable is available by right-clicking the My Computer icon on the desktop, selecting the Advanced tab of properties, selecting Environment Variables. A semicolon should separate different entries in the Classpath.

### 11.3 GUI

Graphical User Interface.

### 11.4 Jar File

Similar to a Zip file.

### 11.5 JAXP

Java API for XML Parsing. Java's set of packages for processing XML files.

### 11.6 Layer

A collection of neurons that make up a Neural Network. The output of each neuron in a layer is linked to the input of every neuron in connected layers.

### 11.7 Neuron

An independent processing unit, similar to a neuron in the brain. A neuron accepts inputs from the outputs of other neurons and produces a result from these.

### 11.8 Neural Network

Computer models based on the neural structure of the brain that are able to learn from experience.

### 11.9 PE

Processing Element, i.e. a node (the neuron) constituting a layer.

### 11.10 Swing

Java's GUI component set. The Joone editor uses these components as the graphical framework of the GUI.

### 11.11 XML

Extensible Markup Language. The Joone editor uses this language to process some parameter files.

## 11.12 Zip File

A compressed set of files. Java is able to read zipped files at run time, allowing Java programs to be deployed as a small number of files.

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