

# IEEE 34 node islanding with implemented Artificial Neural Network (ANN)

Demonstration of the use of Artificial Neural Networks (ANNs) in a microgrid. Model developed by Adailton Braga Júnior, Beatriz Cristina Reis Cordeiro, prof. Ciro Egoavil, and prof. Antonio Ricciotti from the *Federal University of Rondonia* for the *10for10 Typhoon HIL Awards* program of 2019.

## Introduction



**Note:** The model described in this application note can be found on Typhoon HIL's Package Manager service, which is accessible as a [tool in the Typhoon HIL Control Center toolchain](#) or on the [Package Manager Website](#).

Artificial Neural Networks (ANNs) are a subset of machine learning, vaguely inspired by biological neural networks, and thus, the functionality of the human brain itself. The key processes in operating of neural networks are the training and testing processes, where the network is trained by processing examples (as input data) and subsequently tested on different sets of data. This model is intended to demonstrate the training and testing processes of an ANN and its implementation for islanding detection in an IEEE34 bus network.

## Model description

The model consists of an IEEE34 bus network with an [Wind Power Plant \(Average\)](#) used as distributed generation (DG). The wind power plant is connected to the network on node 854 through a [Three-phase two-winding transformer](#). The rated power of the wind power plant is 1.5 MVA and the rated voltage is 480 V. It is configured only to supply the grid with active power (meaning that reactive power is always zero). [Figure 1: IEEE34 bus network with a connected wind power plant \(part 1\)](#) on page 1 shows the IEEE34 bus network with the connected wind power plant.

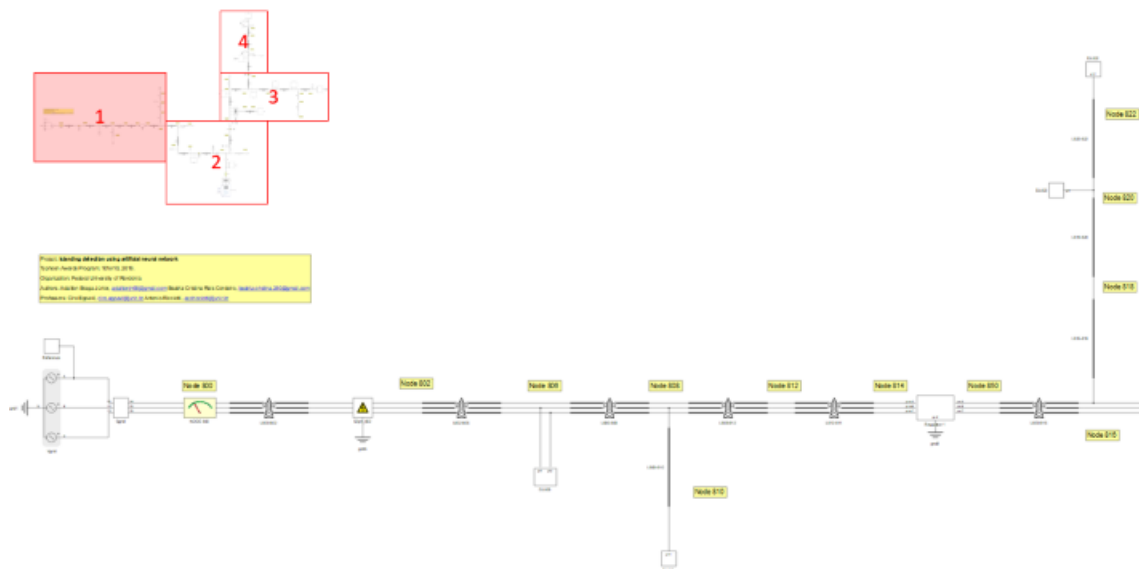
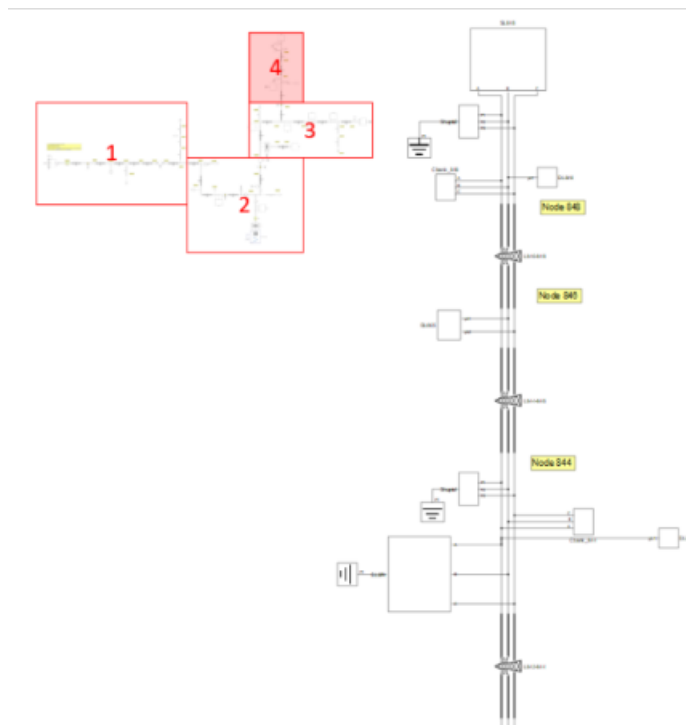


Figure 1: IEEE34 bus network with a connected wind power plant (part 1)





**Figure 4: IEEE34 bus network with a connected wind power plant (part 4)**

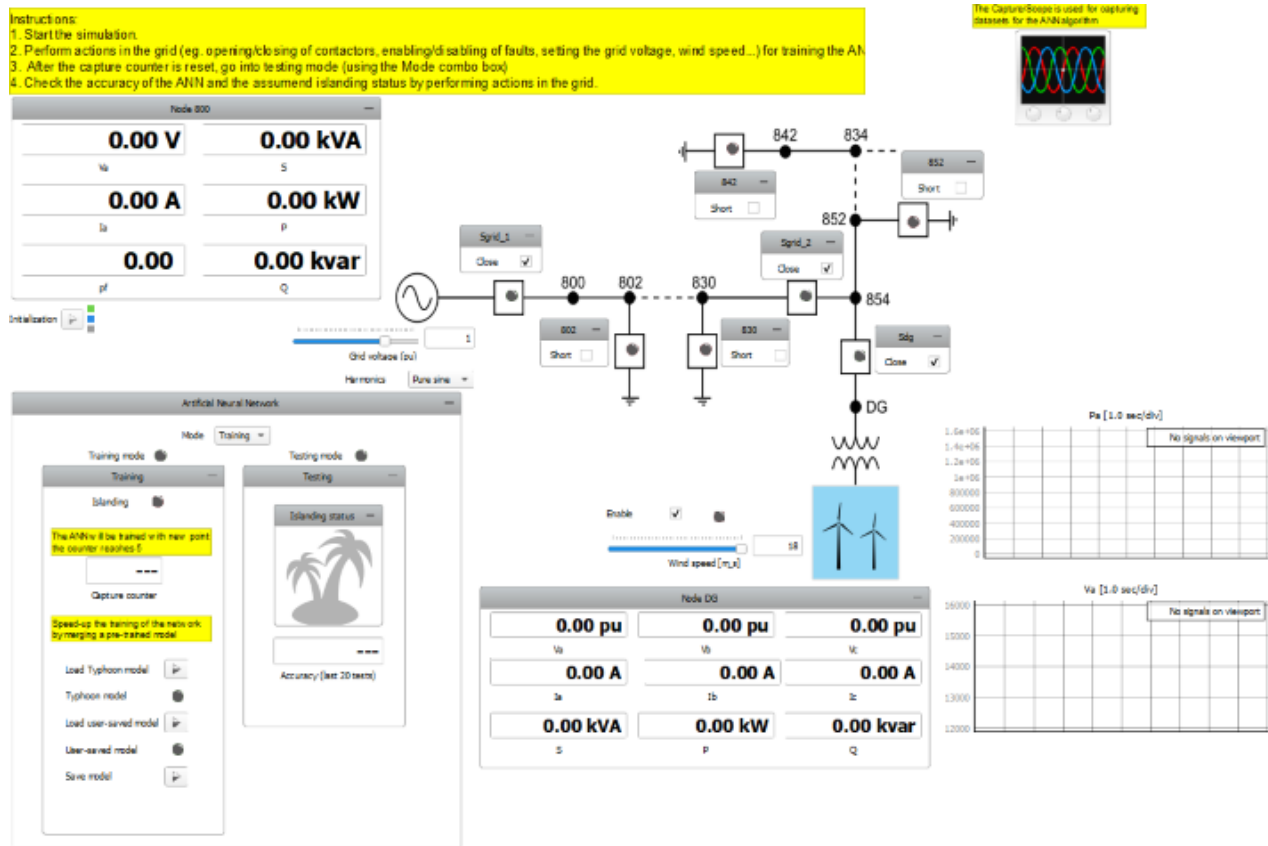
Other components used that are of note are three **TPST (Three-pole single-throw) contactors** used as breakers for generating the state of islanding located at nodes 800 and 854, as well as four **Three-phase ground faults** located in the grid that are used for introducing disturbances during the training and testing processes.

**Table 1: HIL device resource utilization**

No. of processing cores	3
Max. matrix memory utilization	69.26%
Max. time slot utilization	100%
Simulation step, electrical	6μs
Execution rate, signal processing	120μs

## Simulation

This application comes with a pre-built SCADA panel ([Figure 5: Islanding detection in an IEEE34 bus network using an ANN SCADA panel](#) on page 4). The panel offers most essential user interface elements (widgets) to monitor and interact with the simulation in runtime. You can customize it freely to fit your needs.



**Figure 5: Islanding detection in an IEEE34 bus network using an ANN SCADA panel**

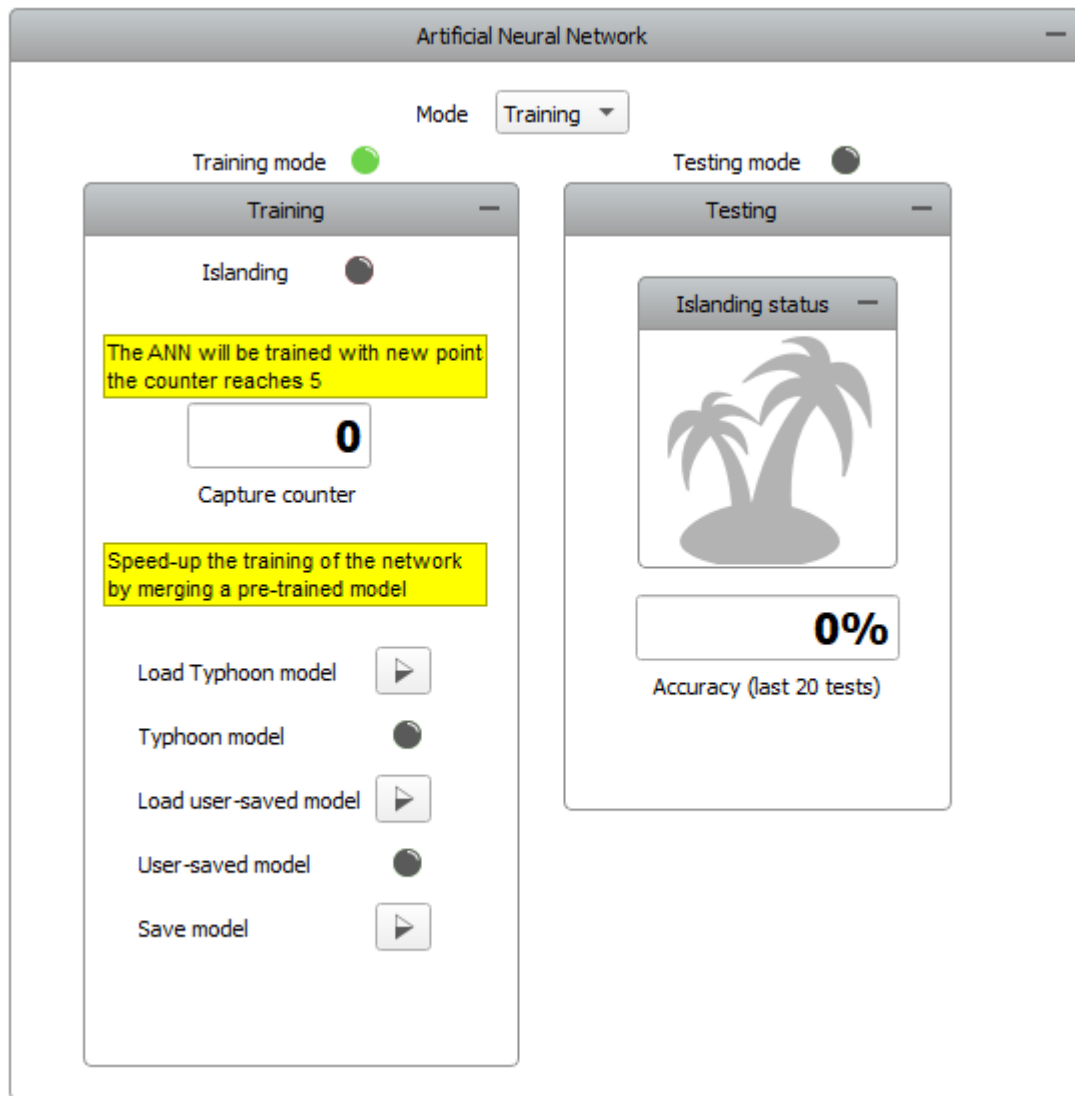
The SCADA panel consists of the following parts:

- Capture/Scope
- One line diagram of the grid
- Artificial Neural Network Group
- Training subpanel
- Testing subpanel
- Node 800 Group
- Node DG Group

The one line diagram displays the state of contactors and faults in the grid (indicated by their corresponding LEDs). Here, you can perform several actions like opening/closing contactors and enabling/disabling faults by using the dedicated checkboxes. Using the dedicated sliders, you are able to set the grid voltage and the wind speed. Using the Harmonics combo box, you are able to enable/disable the presence of harmonics in the grid. Using the Enable checkbox, you can enable/disable the wind power plant. The Node 800 group allows you to observe phase voltage, phase current, apparent power, active power, reactive power, and the power factor on node 800. The Node DG group allows you to observe the values of phase voltages, phase currents, apparent power, active power, and reactive power on node DG.

The Artificial Neural Network group (shown in [Figure 6: The Artificial Neural Network widget group](#) on page 5) consists of 2 main modes:

- Training
- Testing



**Figure 6: The Artificial Neural Network widget group**

The dedicated LEDs for the training and the testing mode indicate the mode that you are currently in, and you are able to switch between these two modes by using the Mode combo box.

The Training subpanel consists of several parts:

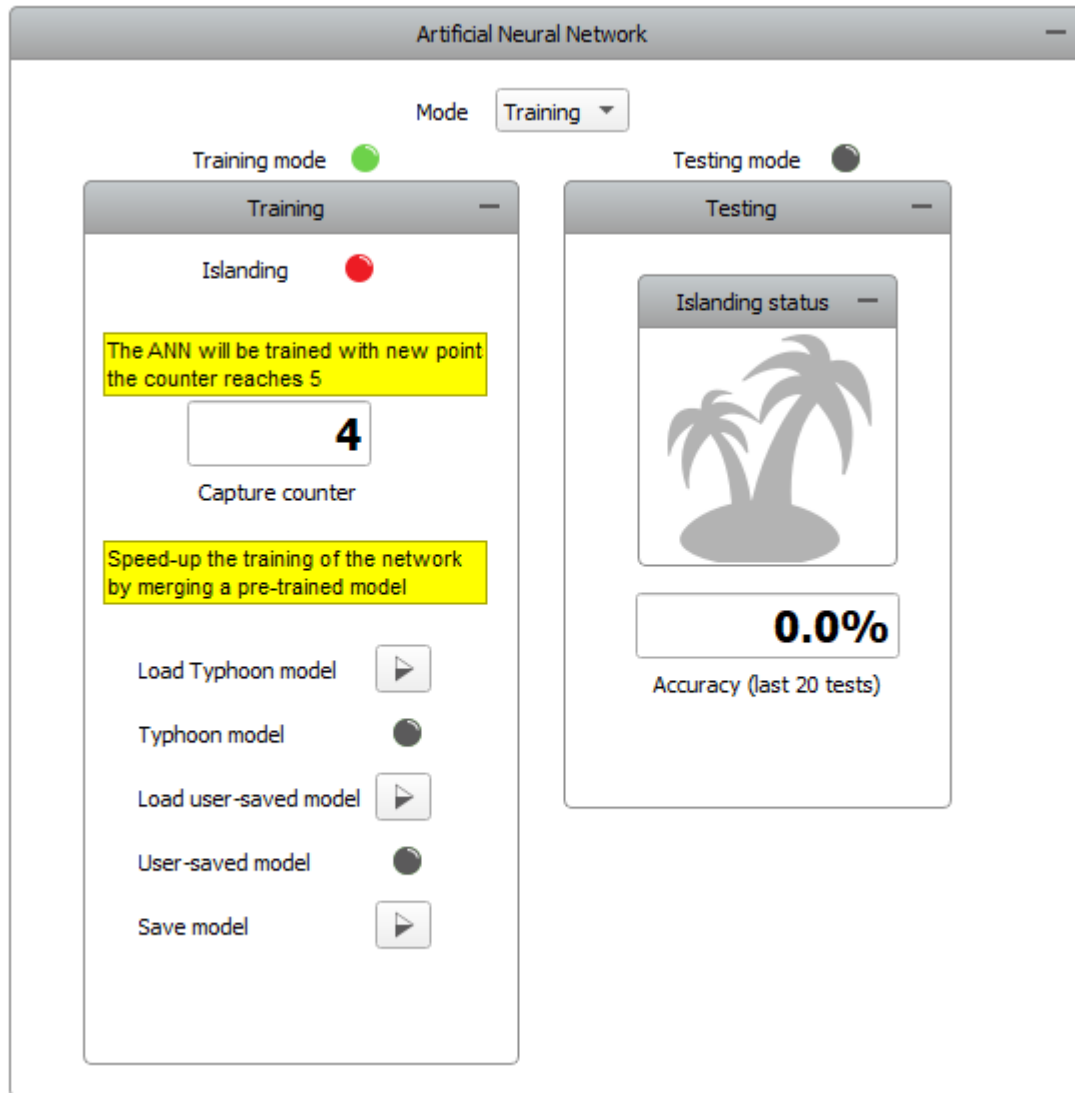
- Islanding LED
- Capture counter
- Load Typhoon model button
- Typhoon model LED
- Load user-saved model button
- User-saved model LED
- Save model button

The islanding LED is used to indicate the state of Islanding. The capture counter is used for displaying the number of data points acquired during one cycle of training.

Upon the start of the simulation, all of the contactors are closed, and all faults are disabled (indicated by their dedicated LEDs).

By default, the model is set to training mode, and it is not possible to go into testing mode until the training process is finished, or an existing model is loaded. With each action performed in the grid (eg. opening/closing of contactors, enabling/disabling of faults, setting the grid voltage, wind speed...), a capture function is performed indicated by the “Capture in progress...” console message. A new set of data points is acquired and the capture counter increases by 1.

After the sixth action in the grid, the counter is reset. The message “New dataset trained.” in the **Message console** indicates that the training process is finished. After that, you are allowed to save the trained model if you choose to, or go into testing mode. [Figure 7: Training mode](#) on page 6 shows the training process.



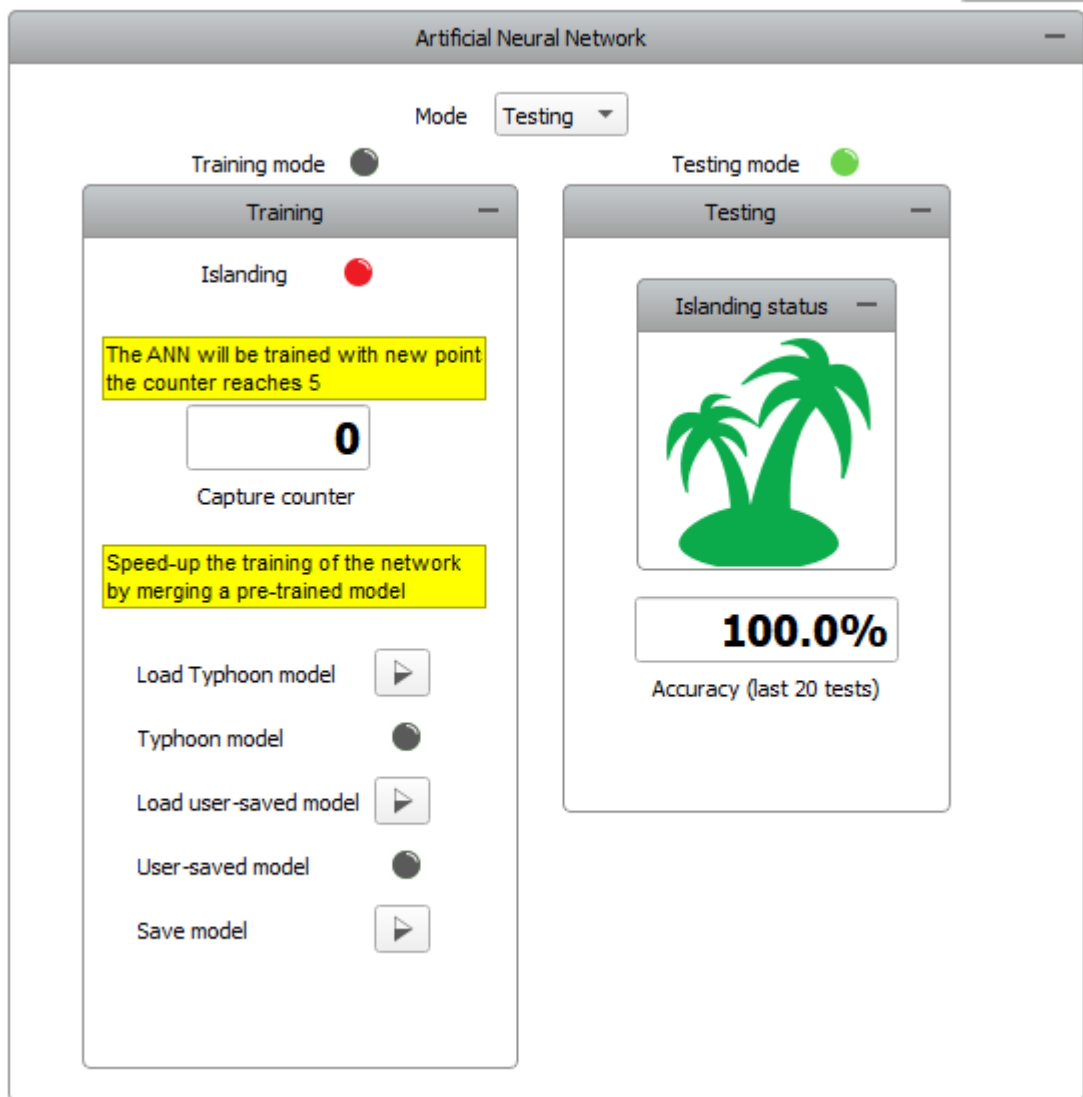
**Figure 7: Training mode**

By using the load button, you are allowed to load a pre-saved model. This model can either be our Typhoon pre-trained model (that you load by clicking on the **Load Typhoon model** button), or you can load your own previously saved model (by clicking on the **Load user-saved model** button). A successful load of a model is indicated by the dedicated Typhoon model and User-saved model LEDs. The **Save** button allows you to save a trained model as a .sav file.

The Testing subpanel consists of the following parts:

- Islanding status
- Accuracy

By performing various actions in the grid (like in training mode), you are able to see the response of the neural network. The islanding status flag is used to indicate the state of islanding state assumed by the trained neural network, and you can check the accuracy of these assumptions (based on the previous 20 tests) below. [Figure 8: Testing mode](#) on page 7 shows the behavior of the testing subpanel while in testing mode.



**Figure 8: Testing mode**

When using a trained model, you are able to go back and forth between training and testing mode so you can further develop the ANN model.

**Table 2: Minimum requirements**

Files	
Typhoon HIL files	examples\models\power systems\ieee 34 node islanding ieee 34 node islanding.tse, ieee 34 node islanding.cus
Minimum hardware requirements	
No. of HIL devices	1

Files	
HIL device model	HIL402
<a href="#">Device configuration</a>	1

**Test automation**

We don't have a test automation for this example yet. Let us know if you wish to contribute and we will gladly have you signed on the application note!

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