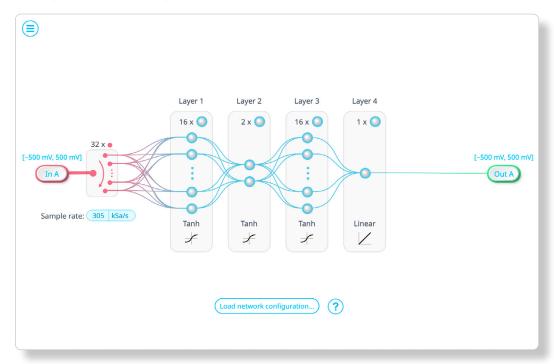


The Moku:Delta Neural Network enables real-time deployment of user-defined models within Multi-instrument Mode. Networks are designed and trained offline in Python, then compiled into Liquid Instruments Neural Network (.linn) files containing weights, biases, and activation functions for accurate hardware reconstruction. Running on the FPGA, the Neural Network delivers low-latency inference, deterministic feedback, and supports up to five hidden layers with 100 nodes per layer and five activation functions. Integrated operation with other instruments allows users to capture, analyse, and respond to signals in real time, making it ideal for applications such as adaptive control and quantum experiments.



Max number of layers

Max neurons per layer 100 Maximum throughput
Up to 305 kHz sample
rate in all modes and
network sizes

Number inputs
1 to 4 in parallel,
1 to 100 in serial

Number of outputs 1 to 4 in parallel, 1 to 100 in serial

18-bit fixed-point representation

### **Features**

- Training
  - Collect training data on any Moku device or simulate a desired complex process
  - Train the Neural Network offline
  - Upload the network configuration to a Moku:Delta device for real-time performance
- Datalog network outputs with the Moku Data Logger, Oscilloscope, or other instruments
- Examples available, including a user guide

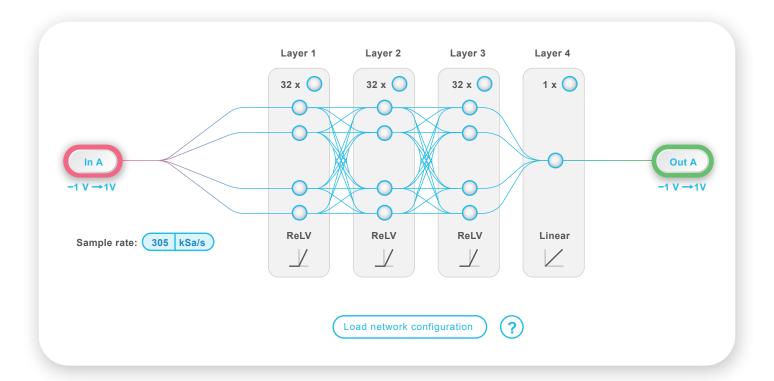
## **Specifications**

- · Activation functions:
  - Nonlinear: ReLU, Softsign, Tanh, Sigmoid
  - · Linear and (future) arbitrary
- Microsecond latency

- Control systems
- Sensor conditioning
- · Signal classification
- Signal denoising



The Moku:Pro Neural Network enables fast, agile implementation and evaluation of user-defined neural networks utilizing Multi-instrument Mode for Moku:Pro. Design and train your network configuration offline. Then, upload it to the Moku app for real-time Neural Network operation.



Max numb of layers 5

Max neurons per layer 100 Maximum throughput Up to 305 kHz sample rate in all modes and network sizes Number inputs
1 to 4 in parallel,
1 to 100 in serial

Number of outputs 1 to 4 in parallel, 1 to 100 in serial

18-bit fixed-point representation

### **Features**

- Training
  - Collect training data on any Moku device or simulate a desired complex process
  - Train the Neural Network offline
  - Upload the network configuration to a Moku:Pro device for real-time performance
- Datalog network outputs with the Moku Data Logger, Oscilloscope, or other instruments
- Examples available, including a user guide

# **Specifications**

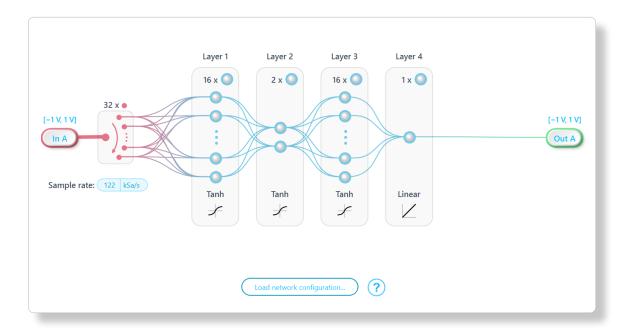
- · Activation functions:
  - Nonlinear: ReLU, Softsign, Tanh, Sigmoid
  - Linear and (future) arbitrary
- Microsecond latency

- Control systems
- Sensor conditioning
- Signal classification
- · Signal denoising





The Moku:Lab Neural Network delivers deterministic performance for real-time applications by running inference directly on the FPGA with sub-100 microsecond latency and 18-bit fixed-point precision. You can implement fast control loops, perform on-device sensor conditioning, or run classification and denoising models. The FPGA executes each layer in parallel using your selected activation functions, allowing you to evaluate compact networks of up to five layers and eighty neurons per layer with consistent timing. This capability enables practical deployment of embedded inference workloads, from adaptive filtering to predictive or state-estimation tasks, as part of an integrated Moku test setup.



Max number of layers

Max neurons per layer **80** 

Up to 122 kHz sample rate in all modes and

network sizes

1 or 2 in parallel, 1 to 80 in serial Number of outputs 1 or 2 in parallel, 1 to 80 in serial

18-bit fixed-point representation

#### **Features**

- Collect training data on any Moku device or simulate a desired complex process
- Train the Neural Network offline
- Upload the network configuration to a Moku:Lab device for real-time performance
- Datalog network outputs with the Moku Data Logger, Oscilloscope, or other instruments

## **Specifications**

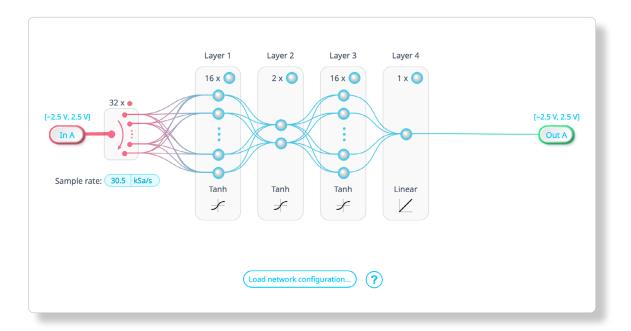
- · Activation functions:
  - Nonlinear: ReLU, Softsign, Tanh, Sigmoid
  - · Linear and (future) arbitrary
- Max neurons per layer
  - 2-slot mode: 80 neurons
  - 3-slot mode: 50 neurons
- Input sampling rate: up to 122 kSa/s
- Sub-100 microsecond latency

- Adaptive filtering
- Anomaly detection
- Real-time control systems
- · Sensor conditioning
- Signal classification
- Signal denoising





The Moku:Go Neural Network delivers deterministic performance for real-time applications by running inference directly on the FPGA with sub-millisecond latency and 18-bit fixed-point precision. You can implement control loops, perform on-device sensor conditioning, or run classification and denoising models. The FPGA executes each layer in parallel using your selected activation functions, allowing you to evaluate compact networks of up to five layers and eighty neurons per layer with consistent timing. This capability enables practical deployment of embedded inference workloads, from adaptive filtering to predictive or state-estimation tasks, as part of an integrated Moku test setup.



Max number of layers
5

Max neuron per layer **80** 

Up to 30.5 kHz sample rate in all modes and network sizes

Number of inputs 1 or 2 in parallel, 1 to 80 in serial Number of outputs 1 or 2 in parallel, 1 to 80 in serial

18-bit fixed-point representation

#### **Features**

- Collect training data on any Moku device or simulate a desired complex process
- Train the Neural Network offline
- Upload the network configuration to a Moku:Go device for real-time performance
- Datalog network outputs with the Moku Data Logger, Oscilloscope, or other instruments

## **Specifications**

- · Activation functions:
  - Nonlinear: ReLU, Softsign, Tanh, Sigmoid
  - · Linear and (future) arbitrary
- Max neurons per layer
  - 2-slot mode: 80 neurons
  - 3-slot mode: 50 neurons
- Input sampling rate: up to 30.5 kSa/s
- Sub-millisecond latency

- Adaptive filtering
- Anomaly detection
- Real-time control systems
- · Sensor conditioning
- Signal classification
- Signal denoising