

# Experimental Setup and Learning- Based AI Model for Developing Accurate PV Inverter Models

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*Photo by Dennis Schroeder, NREL 55200*



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#### Energy Systems Integration Facility

**The Energy Systems Integration Facility (ESIF) is a national user facility located in Golden, Colorado, on the campus of the National Renewable Energy Laboratory (NREL).**



<http://www.nrel.gov/esif> *Photo by NREL*

#### Controller- and Power-Hardware-in-the-Loop

NREL's megawatt-scale controller- and power-hardware-in-the-loop (CHIL/PHIL) capabilities allow researchers and manufacturers to test energy technologies at full power in real-time grid simulations to safely evaluate performance and reliability.





Microgrids **Cosimulation** 



*Photos by NREL*

Power system studies

## Experimental Setup

- Controllable AC supply.
- Controllable DC supply.
- High bandwidth measurements.
- High bandwidth, highly reliable, long duration data storage.
- Automated experimental setup.



*Inverter under test experimental setup*

## Experimental Setup

- Three-phase PV inverter.
- Split-phase PV inverter.
- Battery inverter (not presented in the paper).
- Python script to synchronize the experimental setup.
- Experiments are usually completed in less than three hours.
- Data collected less than a few gigabytes.
- Voltage run and frequency run.



*Experiment flowchart*

(a) Three-phase PV inverter



- Voltage magnitude run
	- RMS change
	- Irradiance change





#### (a) Three-phase PV inverter



- Voltage magnitude run
	- RMS change
	- Irradiance change



- Voltage frequency run
	- Frequency change
	- Irradiance change
	- Frequency measurement of voltage
	- Fourier transform time prioritized and amplitude prioritized



- Voltage frequency run
	- Frequency change
	- Irradiance change
	- Frequency measurement of current
	- Fourier transform time prioritized and amplitude prioritized



### Machine Learning Model

- Inputs: three-phase (a,b,c) voltage, DC current, DC Voltage
- Outputs: three-phase (a,b,c) current
- ANN topology: fully-connected with input/output layers and two 100-node hidden layers using ReLU activations
- Training: Adam optimizer, mse loss function, terminated training after 100 epochs



#### Voltage step experiments outputs



#### Frequency step experiments outputs



#### Future work: machine learning and data collection

- Experimental design: ensure sufficient frequency and voltage coverage in data to train ANN model representation of inverters
- Invest more time into the model, e.g., hyperparameter tuning, adding regularization, etc.
- Represent inverters using more advance ML models: given timeseries data, explore RNNs, Convolutional NNs, and SciML techniques such as:
	- Operator learning (DeepONets or Fouier Neural Operators)
	- Physics Informed Neural Networks (PINNs)

## Takeaways and future directions

- Reduced inverter modeling time frame from 12-18 months to 3- 6 months.
- Targeting to reduce this to days.
- Currently, experimental data is moved to computational setup after the experiments.
- Targeting to train models while the experiment is running.
- How to handle uncertainties?
- Use of design of experiments to determine the experimental sequence.

#### References

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# Thank You

#### **www.nrel.gov**

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