

Analysis of Cyber Attacks in Power Grids with Increasing Renewable Energy Penetration

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I. INTRODUCTION

With the application of advanced computer and communication technologies, traditional power systems are converting to cyber-physical power systems. [1] Several secondary systems, like SCADA (Supervisory Control and Data Acquisition), WAMS (Wide Area Monitoring System), AMI (Advanced Metering Infrastructure), and smart substations, are the main cyber-physical subsystems that could be vulnerable to cyber-attacks. [1], [3] Mathematically modeled cyber-attack types such as FDIA (False Data Injection Attack) [1], [2], [3], [4], DoS (Denial of Service) [1], [3], replay attacks, etc. can be modeled using the PSCAD power system model with the addition of renewable power source models to the system. The goal of this project is to develop a PSCAD power system model with increasing renewable power sources, simulate the different kinds of cyber-attacks mentioned above in the PSCAD power system model, collect faulty data, and develop a deep learning model to detect and act against the cyber-attacks.

II. LITERATURE REVIEW

To overcome the lack of labeled data and address the class imbalance, Generative Adversarial Networks (GANs) have emerged as a contemporary approach. GANs enable the reproduction of synthetic samples that maintain the distribution of original samples.[5] The scenarios include short-circuit faults, line maintenance, and various cyberattacks like RTCI, RSC, and DI attacks. Each scenario represents a specific challenge in power grid security.[6],[7] Detailed analysis of cyber-attacks involving photovoltaic (PV) systems, such as extra reactive power compensation, PV farm inverter attacks, and linear load cutoff. These attacks simulate scenarios, where false data injection and data integrity attacks compromise power system stability.[6] Research gaps can be identified as follows.

- Only focusing on one type of attack
- Real scenarios are not simulated.
- Renewable energy penetration less considered.

III. MATERIALS AND METHODS

The study utilized software such as PSCAD and MATLAB Simulink are used to design a power system model with renewable energy sources, generate healthy data for seven transmission lines, simulate fault data injection and load alternating attack types, and collect data through simulations.

Fault data injection attack is simulated using the below model. [Fig.1] This model can generate random values for voltage, frequency, and phase angle and through the solar plant phasor measurement units these data were injected to the system communication network. Due to these types of fault data, some of diesel generators change their characteristics. Due to those changes, system instabilities occurred and by measuring wave patterns in transmission lines, fault types can be identified.

Load Alternating attack [Fig.2] model can cut off the loads in system by accessing through the solar system communication channel. The attack is directly done to breaker relay system and breakers can be controlled using the model.



Fig. 1. Fault data injection model



Fig. 2. Load alternating attack model

IV. RESULTS AND DISCUSSION

Data was collected for healthy power system wave patterns, Fault data injection attacked system wave patterns and Load alternating attacked system wave patterns for the Instantaneous current, Instantaneous voltage, RMS voltage, Active Power, Reactive Power, and Phase angle.

A. Healthy data wave patterns



Fig. 3. Healthy data wave pattern for instantaneous current



Fig. 4. Healthy data wave pattern for active and reactive power

B. Fault data injection attack wave patterns



Fig. 5. Fault data injection attack wave pattern for instantaneous current



Fig. 6. Fault data injection attack wave pattern for active and reactive power

C. Load alternating attack wave patterns



Fig. 7. Figure 1: Load alternating attack wave pattern for instantaneous current



Fig. 8. Load alternating attack wave pattern for active and reactive power

V. CONCLUSION

Due to rapid development of cyber – physical power systems, attack detection and defense is becoming more complex day by day. Using the data collected from the power system model with and without attacks, our aim is to build a deep learning based neural network model to identify the attack type.

Future works to achieve the objectives:

- Analyze the attack patterns with common faults and improve the data set.
- Analyze the data set and feature extraction for neural network model.
- Develop the neural network model and optimize the model.

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