



CPS: Breakthrough: Securing Smart Grid by Understanding Communications Infrastructure Dependencies

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Overview

Objectives:

- Characterize inter-dependence between electrical grid and communication systems
- Secure protocols for state estimation more robust in transmission and distribution network.
- Assure Data Integrity from Advanced metering infrastructure and customer network.
- Build various models for attack mitigation.
- Validate with micro-grid test-bed and real datasets

Research Methodologies

- Intentional Islanding
- Stochastic decision processes
- Failure prediction models
- Software attestation
- Power flow inconsistency detection
- Cloud-based state estimation
- Steganography-based tamper detection
- Vulnerability assessment
- Analysis of protocol anomalies

Scientific Impact:

- Anomaly detection and trust models for attack mitigation
- Situation-aware models for threat monitoring, analytics, decision control

Challenges: Inter-dependence, IoT Robustness, Cyber-Physical, Big Data

- Integrity mechanism for protection and state estimation

Broader Impacts:

- Influencing the standards
- Multi-disciplinary training in CPS security
- Experiential learning in real micro-grid facility.
- Outreach and research demo
- Generalization to other CPS

Expected Results

- Mitigation of Cascading Failures
- Attack Detection and Isolation
- Improved System Robustness & Control

PMU data → State Estimation → Bad Data Detector → Fail → Bad Data Identification and Removal → Pass

S. Tan, D. De, W. Song and S. K. Das, "Security Advances in Smart Grid: A Data Driven Approach," *IEEE Communications Surveys and Tutorials*, 2017.

Silent Perturbation of State Estimation

Goal:

- Damage power equipment
- Increase system operation costs
- Disproportionate power generation/dispatch or energy routing
- Cause economic loss

How?

- Perturbing the state estimation
- Fooling the system operator to make unnecessary and costly actions, such as generator rescheduling and load shedding.

Assumptions:

- IEC TR 61850-90-2 allows sending protection messages in plaintext.
- Active adversary with MitM attack capability.
- Adversarial knowledge:
 - Known: bad data detection threshold, i.e., # of states and measurements, topology of the power grid
 - Unknown: accurate knowledge of Jacobian measurement matrix.

Smart Meter Data Falsification

Organized, Persistent Adversaries

- Circumvent cryptographic defense
- Compromise a large # of meters
- Attacks persist and evolve
- Mask easy consistency check
- Knowledge of business and revenue models

Challenges

- Consumption exhibits inherent fluctuations
- Distinguishing between legitimate and malicious changes
- Large no. of compromised nodes with smaller margin of false data
- Various falsification types

Attack Models

- Additive
- Deductive
- Camouflage
- Conflict
- Incremental Evolving, On-Off attacks, Omission, Order Aware Falsification Strategies

Proposed Trust Model

Robust consensus formed through anomaly detection

Use a Kullback-Leibler Divergence between historical and current proximity distributions of smart meter to the robust consensus R_t

$$X_i(t) = \begin{cases} 1 & \text{if } p_i^t(rep) \in \{\mu_i \pm \sigma_i\}; \\ 0 & \text{otherwise} \end{cases} \quad \text{Historical Proximity}$$

$$Y_i(t) = \begin{cases} 1 & \text{if } p_i^t(rep) \in \{\mu_{R_t} \pm \sigma_{R_t}\}; \\ 0 & \text{otherwise} \end{cases} \quad \text{Current Proximity}$$

$X_i(t) \rightarrow$ probability parameter r

$Y_i(t) \rightarrow$ probability parameter q

$$D_i(X_i||Y_i) = (1-r) \times \ln\left(\frac{1-r}{1-q}\right) + p \times \ln\left(\frac{r}{q}\right) \rightarrow \text{Kernel Methods}$$

Integrity of Protection Messages

Challenges

- Most recent mp in substations use ARM Cortex-M cores
 - Cannot meet 4ms requirement for hash based integrity checking or encryption
- Need a very light weight but secure mechanism.

Our Approach

- Permutation only encryption

Algorithm

- Generate 16-bit Fletcher checksum
- Generate a set of random numbers based on a seed
- Sort the numbers & use them as offsets for checksum bits
- Hide checksum bits in the message

Key management

- Initially communicated to all receivers securely.
- Salted with status and renegotiated when counter rolls over.

Security Analysis

- 96 bit security
- Key salting ensures security against known/chosen plaintext attacks
- Success probability before the key changes is negligible.
- Secure from off-path attacks

Performance Analysis

- Real implementation on a 48 MHz ARM cortex mp

Algorithm	Speed (KB/s)
Proposed Method	424
MDS	147
ChaCha20-Poly1305	94
AES-128-CCM	70
AES-128-EAX	70
AES-128-GCM	41

Publication: Kant, K. and Jolfaei, A. 2017. A Lightweight Integrity Protection Scheme for Fast Communications in Smart Grid, 14th International Conference on Security and Cryptography (SECRYPT), Jul. 24–26, Madrid, Spain, pp. 31-42.

Attack Procedure

Bypass bad data detection

- Malicious measurements pass the bad measurement detection if the L_2 norm of the attack vector \leq the bad data detection threshold.

Adversary reconstructs the entries of measurement Jacobian matrix within the maximum error margin of a small percentage.

- Small perturbations in the measurements can lead to a large drift in the state value if the smallest singular value of the Jacobian measurement matrix is small.

It is theoretically/practically impossible to spoof a large number of measurements at once.

- States are perturbed partially/gradually in different rounds of state estimation.

Drift state values within a desired range

- Linear unidirectional changes in voltage magnitudes and phase angles.
- Impulsive and/or oscillatory modifications.
- The acceptable range of voltage amplitude variation is within $\pm 5\%$.

Proposed Framework: Overview

Light weight, Real Time Anomaly Detection; Not privacy intrusive; Works for various attack types; Distinguish between legitimate and malicious changes; Suitable for both isolated and organized rivals

Performance Evaluation

Detection = 100%
False Alarm = 8.3%

Detection = 99%
False Alarm = 9.2%

We use real data set from PECAN Street Project (SmartGridGov) and Irish Data Sets.

We emulate attacks on real data fed to a virtual simulated AMI

We observe clear difference between compromised and non-compromised nodes.

Results[1] are better due to the robustness of statistical measures used in various steps

Bhattacharjee, Thakur, Silvestri, Das, et al. "Statistical Security Incident Forensics against Data Falsification in Smart Grid Advanced Metering Infrastructure," *ACM CODASPY*, 2017.

Smart Grid Management

Kant, K. and Jolfaei, A. 2017. On the Silent Perturbation of State Estimation in Smart Grid, *IEEE Journal of Selected Topics in Signal Processing*, Under Review.

Evaluation

Proposed Anomaly Detection

HM to AM absolute difference (AD) is a stable invariant

AD metric sharply increases for all types of Data Falsification

A rise in the absolute difference of HM and AM is a reliable indication of occurrence of organized falsification

Anomaly detection metric is stable without any smoothing average technique

The property holds for all different datasets studied. e.g. Texas, Ireland datasets

AD	AM	HM	GM	Conclusion
Increased	Increased	Increased	Increased	Additive
Increased	Decreased	Decreased	Decreased	Deductive
Increased	Same	Any	Any	Camouflage
Increased	Don't Care	Don't Care	Don't Care	Conflict
Same	Any	Any	Any	No Attack

Ongoing Research

Stealthy and Persistent Attacks

- Advanced Information Theoretic Approaches beyond divergence measures
- State-less (short term) and State-full (long term) Detectors.
- Decrease the false alarm rates without sacrificing detection rate.
- Margins of false data below 400, Unsupervised and scalable.

Robust state estimation

- Silent state perturbation mechanisms with partial knowledge of network parameters
- Mitigation mechanisms

PMU data falsification

- Identify compromised meters
- Formalize supervised and unsupervised learning techniques