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# **Statistical Structure Learning of Smart Grid** for Detection of False Data Injection

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## **Motivation and Introduction**



- Phasor Measurement Units (PMUs)
  - Synchronous with GPS stamp
  - Various applications
  - Will be placed partially along with State Estimators









### **Problem Formulation**

#### Gaussian Markov Random Field

$$f_X(x) \propto exp[-\frac{1}{2}x^T J x + h^T x]$$

$$J(i, j) = 0 \iff (i, j) \notin E$$

$$E(X_i | X_{N(i)}) = E(X_i | X_{-i})$$

DC power flow equations

#### Structure Learning

Conditional Covariance Test (Anandkumar et.al. 2012)

Estimates the structure of underlying graphical model given i.i.d. samples of the r.v.s

**Algorithm 1** Algorithm  $CCT(\mathbf{x}^n; \xi_{n,p}, \eta)$  for structure learning using samples  $\mathbf{x}^n$ .

Initialize  $\widehat{G}_p^n = (V, \emptyset).$ For each  $i, j \in V$ , if

 $\min_{\substack{S \subset V \setminus \{i,j\} \\ i \in \mathcal{I}}} |\widehat{\Sigma}(i,j|S)| > \xi_{n,p},$ 

then add (i, j) to  $\widehat{G}_p^n$ . Output:  $\widehat{G}_p^n$ .

## **Detection Scheme**

- Decentralized scheme
- > Online calculations
- > Markov graph changes under attack
- Mismatch alarm
- > All attack scenarios > MATPOWER for running DC power flow ➢ IEEE 14-bus system & IEEE 30-bus system  $\geq$  100% detection rate,
- 100 90 80 70 % .⊆ 60 50 40 30 20 No. of measurement

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Detection Rate for IEEE 14 bus syster

| $z^{a} = z + a = H(X + d) = HX^{a}$ $\Sigma(X^{a}, X^{a}) = H^{-1}[\Sigma(P, P) + \Sigma(a, a)]H^{-1T}$ $\Sigma(X^{a}, X^{a}) \neq \Sigma(X, X)$ | for IEEE-14 and 50 for IEEE-<br>30.<br>Reason: sparsity | Detection rate is 90% for just<br>30 corrupted samples<br>Considering current sampling<br>rate these values are pretty<br>good. |
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|--|---|---|

## **Discussion & Future Work**

- The first detection scheme for this sophisticated attack
- $\succ$  Computational complexity  $O(p^{\eta+2})$
- > Sample complexity  $\Omega(J_{min}^{-2} \log p)$
- > Apply to bigger networks Readily detects other types of attack > Causality approach with time series analysis