

# An Efficient Exploring Malicious Meter Inspection in Smart Grid Using an Adaptive Binary Splitting Algorithm

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**Abstract--** Electricity theft is a broad issue that causes gigantic monetary misfortunes for every service organization around the world. The same number of nation's battle to refresh their antique power frameworks to rising smart grids, more and smarter meters are conveyed all through the world. Contrasted and simple meters which can be messed with by just physical assaults, smart meters can be controlled by malicious clients with both physical and digital assaults to steal electricity. Consequently, electricity theft will turn out to be significantly more genuine in a smart grid than in a conventional power framework if a service organizations do not execute compelling arrangements. The objective of this paper is to distinguish every malicious client in an area territory in a smart grid inside the most brief recognition time. To propose an Adaptive Binary Splitting Inspection (ABSI) algorithm which embraces a group testing strategy to find the malicious clients. There are two viewed as inspection procedures in this paper: a filtering technique in which clients will be investigated exclusively, and a binary inquiry strategy by which a particular number of clients will be analyzed all in all. Amid the inspection procedure of our proposed plan, the inspection technique, just as the quantity of clients in the groups to be investigated, are adaptively balanced. Reproduction results demonstrate that the proposed ABSI algorithm beats existing techniques.

**Index Terms—** Electricity Theft, Smart Grid, Security, Group Testing, Scanning Method, Binary Search Method, BCGI (Binary Coded Grouping-based Inspection).

## 1. INTRODUCTION

As a promising force foundation, the smart grid is being acquainted with an ever increasing number of nations, for example, USA, Japan, and China . To make electrical grids "smart", a huge number of current equipment and programming methods are Coordinated into power frameworks. For instance, simple meters in customary power frameworks are

moved up to advanced smart meters, which have capacities of calculation, correspondence, and remote control. In addition, a digital layer is added to the metering framework. Lamentably, while these procedures bring us comfort and proficiency, they likewise empower malicious clients to apply various better approaches to take electricity, where malicious clients are alluded to as the clients taking electricity.

Contrasted with simple meters which can be altered by just physical assaults, for example, straightforwardly taking advantage of electrical cables and bypassing vitality meters, smart meters can likewise be controlled with digital assaults. Practically all service organizations around the world, particularly those in many developing business sector nations experience the ill effects of electricity theft. At present, as indicated by another investigation distributed by Northeast Group, LLC, the world loses \$89.3 billion every year because of electricity theft, among which the best 50 developing business sector nations lose \$58.7 billion every year. Given that service organizations don't execute an effective arrangements, electricity theft will turn out to be considerably more genuine in smart grids than in conventional power frameworks.

The Binary Coded Grouping-based Inspection (BCGI) algorithm group's clients in the NAN dependent on the binary arrangements of distinguishing proof numbers BCGI groups the clients in the NAN. BCGI find malicious clients by just a single inspection step works there is a special malicious client in the NAN. Every inspection box incorporates investigators and sub-controllers. A reviewer box which contains a head auditor and a few sub-controllers. The head investigator is in charge of finding the nearness of grimy clients; the sub-overseers take charges of getting the malicious meters precisely.

For finding messy clients in the briefest identification time, a progression of inspection

methods are proposed in papers. The principal reason for existing is to uncover any noteworthy practices exceedingly identified with electricity theft. Arrangement based and control based techniques are exceedingly costly on the grounds that we need to introduce a focal spectator or controllers for every client premises.

## 2. RELATED WORK

The smart metering framework for neighborhood territory arrange is seen on the Fig1.

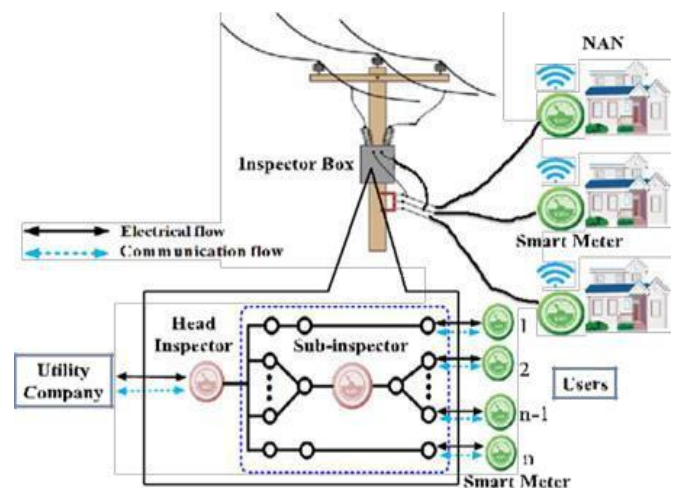


Fig.1 an Outline for the Malicious Meter Inspection

### Neighborhood Area Network (NAN):

In each NAN, there installs an inspector box which contains a head inspector and several sub-inspectors. The head inspector is responsible for detecting the existence of reading anomalies; the sub-inspectors take charges of identifying the malicious meters exactly.

### Inspector Box:

The inspector box consists of two kinds of inspectors: a head inspector responsible for

detecting the existence of reading anomalies, and several sub-inspectors which aim to exactly locating the malicious users in the NAN. We assume that inspectors are either secure or equipped with tamper-resistant components/functions.

- 1) The head inspector monitors all the users statically;
- 2) The set of users monitored by the sub-inspectors can be changed automatically or manually;
- 3) The sub-inspectors can be effortlessly added into or removed from the inspector box

### Utility Company:

Utility companies may install multiple inspector boxes in different neighborhoods and multiple sub-inspectors in each box for shortening the detection time as much as possible. To a large extent, the budget of utility companies determines the number of inspector boxes and the number of sub-inspectors in each box to be installed.

### Smart Meter:

A smart meter is installed at each user's premises for the purpose of recording and then periodically reporting electricity consumptions to utility companies. Let  $n$  and  $U = \{1, 2, \dots, n\}$  denote the total number of users and the set of all users, respectively, in the NAN.

Smart meters are introduced at every client destinations for announcing and recording the electricity utilization of associations. Give  $n$  a chance to be the quantity of clients and  $u$  be the arrangement all things considered,  $u = \{1, 2, \dots, n\}$ . Let  $V_j$  and  $V_j'$  signifies clients'  $v$ 's accounted for perusing advertisement certified electricity uses.  $V_j$

and  $V_j'$  indicates the clients in the NAN as malicious or legit clients. In the event that it is malicious, the electricity take-up is not exactly real uptake. ie,  $V_j < V_j'$  and on the off chance that it is straightforward client truly reports the electricity report insights. The inspection box comprises of two sorts of controllers: that is head assessors and the few overseers. Head controllers identify the peculiarities and sub-investigators finding the malicious clients. In light of the budgetary limitations of associations which decides the quantity of head overseers and sub-assessor.

Give  $k$  a chance to indicate the whole number of sub-monitors in the investigator box. At that point, the arrangement of investigators can be indicated by  $I = \{0, 1, 2, \dots, k\}$ , where controller 0 demonstrates the head investigator and overseers 1, 2, ...  $k$  alludes as sub-controllers. Give  $G_s$  a chance to signify the group of clients checked by overseer  $s$ , sub-assessors, we have  $G_s \subset U$ ,  $\forall s \in I - \{0\}$ . For examiners  $s, t \in I - \{0\}$ , we have  $G_0 \cap G_t = G_t$  and  $G_s \cap G_t = \emptyset$ . For any auditor  $s \in I$ , when it works, it works as pursues:

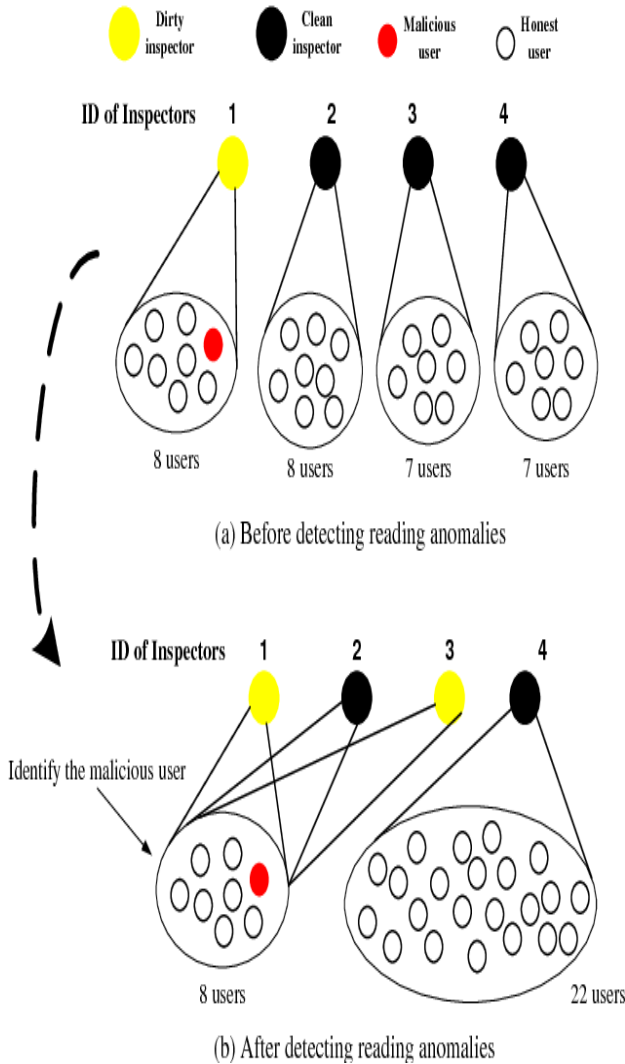
1.  $C_i$  Measures the aggregate sum of electricity appropriated to the clients in  $G_s$ .
2. Receiving these clients' accounted for readings.
3. Calculating is the entire measure of stolen electricity of the considerable number of clients in  $G_s$ , which is recorded by  $b_i$ . At the point when an examiner leads one time of the above tasks, it performs one inspection step. In view of the law of preservation of vitality.

Where  $q_i$  indicates client  $t$ 's accounted for

readings, and  $\delta S$  speaks to the aggregate sum of specialized misfortunes of the clients in  $G_s$ .

$$\text{Energy, } b_i = c_i - q - \delta_i$$

Where  $q_j$  denotes user  $t$ 's reported readings, and  $\delta S$  represents the total amount of technical losses of the users in  $G_s$ .

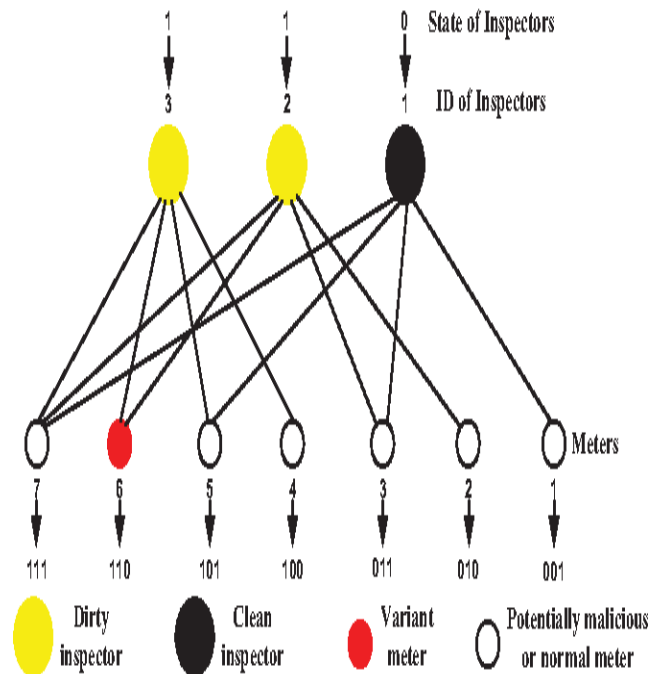


**Fig.2 (a)** Before Detecting Reading Anomalies

**Fig.2 (b)** After Detecting Reading Anomalies

The above Fig.2 (a) & Fig.2 (b) shows that an Adaptive Binary Splitting Inspection (ABSI) algorithm which adopts a group testing method to locate the malicious users. There are two considered

inspection strategies: a scanning method in which users will be inspected individually, and a binary search method by which a specific number of users will be examined as a whole. During the inspection process of our proposed scheme, the inspection strategy as well as the number of users in the groups to be inspected are adaptively adjusted.



**Fig.3** an Illustration of the BCGI Algorithm

The above Fig.3 illustrates about the BCGI (Binary Coded-based Grouping Inspection) algorithm as follows, The BCGI algorithm works only when there is only one malicious user in the NAN, we set  $m=1$ . The total number of users in the NAN ranges from 100 to 400. The sub-inspectors will take more inspection steps using the ABSI algorithm than using the BCGI algorithm. The BCGI algorithm utilizes more inspectors than the ABSI algorithm.

### 3. PROPOSED WORK

Here the proposed framework alludes to the working of ABSC. ABSC utilized for finding the malicious clients from a group. At first set up how the reviewer decide if there is any messy client among the all-out group or in the T. Next characterizes the edge, recorded by  $\omega$ , to help gauge whether there are perusing peculiarities among the clients being observed. In particular, if  $P_s \geq \omega$ ,  $s \in I$ , the reviewer scan reason that there exist malicious clients in Gs.

- If  $P_s > \omega$  and there is just a single client in Gs, this one of a kind client will be distinguished as being malicious.
- If  $P_s \leq \omega$ , all clients in Gs will be proclaimed as being straightforward.
- If  $P_s > \omega$  and Gs contains different clients, we can just reason that somewhere around one malicious client in Gs. The status of any client in Gs can't be resolved quickly, and more inspection steps should be additionally directed.

Give T a chance to signify the arrangement of clients whose status ("fair" or "malicious"). Give M a chance to signify the arrangement of clients in which malicious clients are as of now being recognized. Give H a chance to indicate the arrangement of clients in which officially distinguished the genuine.

- (1)  $U = T \cup M \cup H$ ; (2)  $T \cap M = \emptyset$ ;
- (3)  $W \cap H = \emptyset$ ; (4)  $M \cap H = \emptyset$ ;

An all-out number of at most malicious clients in the NAN and the clients in M are malicious, we can gather that amid the inspection procedure, the

greatest number of malicious clients in T that stay to be distinguished is  $\mu - |M|$ . Among the clients whose status has not been resolved, if all things considered one client out of something like two clients is malicious,

$$\text{i.e., } |T| \geq 2(\mu - |M|) - 1$$

Examining strategy which reviews the clients in one by one just binary pursuit technique which finds the malicious client.

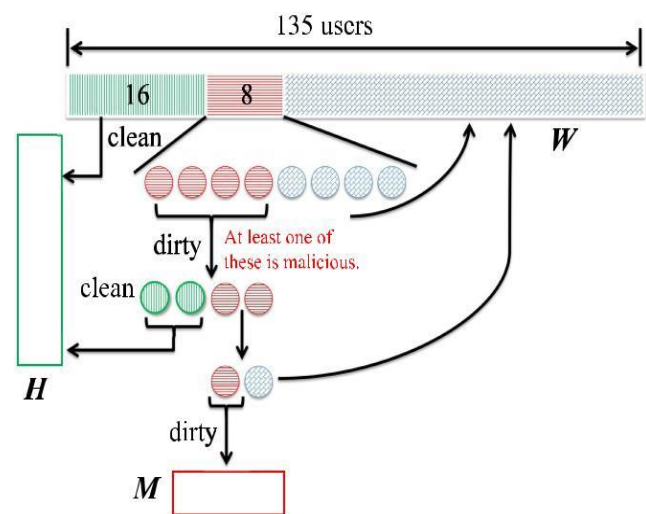


Fig.5 an Illustrate the ABSI Algorithm

Fig.5 shows to illustrate the ABSI algorithm. For example, if all the users being inspected are honest, this inspection is referred to as "clean". Otherwise, it is "dirty." Notably, H denotes the set of users which have already determined as honest; M denotes the set of users which have already determined as being malicious; W denotes the set of users whose status has not yet been identified.

A critical element of ABSC which can adaptively modify their inspection systems by progressively changing the inspection procedures it is possible that it can utilize binary hunt technique or filtering

strategy. Binary hunt technique separates the specific number of clients from T. Examining strategy investigate the whole client one by one, tedious and require more controllers. The significant contrast between this two is binary pursuit technique can review as entire instead of investigating each one in turn.

**3.1 ADJUSTABLE BINARY SPLIT CHECK ALGORITHM**

First, the splitter uses a splitting algorithm category to find the best split for each variable according to the criterion. Next, the variable that has the best split determines the split of the leaf. The splitter uses different algorithms called splitting categories to find the best split for a variable.

**Algorithm1: Adaptive Binary Splitting**

**Inspection (ABSI)**

Require: W

Ensure: M, H

Initialization:  $W \leftarrow U, M \leftarrow \emptyset, H \leftarrow \emptyset$

{M and H are global}

ABSI (W):

**while**  $x_0 > \omega$  **do**

**if**  $|W| \geq 2(\lambda - |M|) - 1$  **then**

    Binary Search (W);

**else**

    Scan (W);

**end if**

**end while**

{end ABSI}

Scan (W):

$G_i, W \leftarrow \text{extract Users (W, 1)}$ ;

{extract 1 user from W to  $G_i$ }

**if**  $x_i > \omega$  **then**

$M \leftarrow M \cup G_i$ ;

**else**

$H \leftarrow H \cup G_i$ ;

**end if**

ABSI (W); {end Scan}

Binary Search (W):

$\alpha \leftarrow \log_2 |W| - \lambda (\lambda - |MM|) + 1$ ;

$G_i, W \leftarrow \text{extract Users (W, } 2\alpha)$ ;

The sub-inspector i conduct one inspection step to obtain  $x_i$ ;

**if**  $x_i > \omega$  **then**

$k \leftarrow 0$ ;

**while**  $k \leq \alpha$  **do**

**if**  $|G_i| == 1$  **then**

$M \leftarrow M \cup G_i$ ;

**break**;

**else**

$G_i, G_i \leftarrow \text{extract Users (} G_i, |G_{2i}|)$ ;

        The sub-inspector i conduct one inspection step to obtain  $X_i$  ;

**if**  $x_i > \omega$  **then**

$G_i \leftarrow G_i; W \leftarrow W \cup G_i$ ;

**else**

$G_i \leftarrow G_i ; H \leftarrow H \cup G_i$ ;

**end if**

$k++$ ;

**end if**

**end while**

**else**

$H \leftarrow H \cup G_i$ ;

**end if**

ABSI (W);

{end Binary Search}

### Algorithm Description:

Its initially exhibit how the investigators judge whether there are malicious clients among the clients being checked by them. Taking into account that specialized misfortunes as a rule can't be acquired precisely in reality, we characterize an edge, recorded by  $\omega$ , to help judge whether there are perusing inconsistencies among the clients being observed. In particular, if  $x_i \geq \omega$ ,  $I \in I$ , the examiner  $I$  can induce that there exist malicious clients in  $G_i$ . Accepting the head examiner for instance, if and just on the off chance that it discovers  $x_0 > \omega$ , it identifies perusing oddities. With respect to the sub-controllers, amid the inspection procedure of discovering every single malicious client, their working techniques are exhibited as follows: (1) For any sub-inspector  $i \in I \setminus \{0\}$ , if and only if  $x_i > \omega$  and there is only one user in  $G_i$ , this unique. Client will be distinguished as being malicious; (2) conversely, for any sub-auditor  $I \in I \setminus \{0\}$ , if  $x_i \leq \omega$ , all clients in  $G_i$  will be proclaimed as being straightforward, paying little respect to the quantity of clients being contained in  $G_i$ ; (3) Especially, for the situations where  $x_i > \omega$  and  $G_i$  contains various clients, we can just presume that there is something like one malicious client in  $G_i$ . For this situation, the status of any client in  $G_i$  can't be resolved promptly, and more inspection steps should be additionally led.

Give  $W$  a chance to signify the arrangement of clients whose status ("genuine" or "malicious") has not yet been resolved. These clients should be

additionally investigated in the accompanying inspection process. Give  $M$  a chance to mean the arrangement of clients which have just been recognized as being malicious. Give  $H$  a chance to signify the arrangement of clients which have just been recognized as genuine. Clearly, we have:

$$(1) U = W \cup M \cup H; (2) W \cap M = \emptyset;$$

$$(3) W \cap H = \emptyset; \text{ and } (4) M \cap H = \emptyset;$$

Note that the sets  $W$ ,  $M$ , and  $H$  powerfully change as the inspection continues. Toward the start of inspection process, we instate  $W = U$ ,  $M = \emptyset$  and  $H = \emptyset$ . At the point when the inspection procedure is done, we have  $W = \emptyset$ , and  $U = M \cup H$ .

### 4. CONCLUSION

In this ABSI algorithm which adaptively modifies the inspection systems. Normal among two clients one being malicious we utilize the binary hunt technique, generally utilize the examining strategy. ABSC gives out the most extreme number of inspection steps. The current frameworks use BCGI algorithm distinguishing the malicious meter submitting electricity theft in neighboring regions. However, BCGI can find the remarkable malicious meter if there is one meter ends up malicious in one revealing period. Its centers around discovering electricity theft. In this, we accept that once malicious clients are found, the service organizations disengage their capacity accounts quickly and don't reestablish electricity until malicious clients wrap up the entire party. ABSC algorithm is an increasingly broad methodology and the smart grid gives client

security, a more prominent number of keen gadgets, the lifetime of intensity frameworks and physical security.

## REFERENCES

- [1] A. H. Nizar, Z. Y. Dong, and Y. Wang, "Power utility nontechnical loss analysis with extreme learning machine method," *IEEE Trans. Power Syst.*, vol. 23, no. 3, pp. 946–955, Aug. 2008.
- [2] B. Krebs, FBI: Smart Meter Hacks likely to spread, 2012, [online] Available: <https://krebsonsecurity.com/2012/04/fbi-smart-meter-hacks-likely-to-spread/>.
- [3] C. H. Lo and N. Ansari, "CONSUMER: A novel hybrid intrusion detection system for distribution networks in smart grid," *IEEE Trans Emerg. Topics Comput.* vol. 1, no. 1, pp. 33–44, Jun. 2013.
- [4] C. C. O. Ramos, A. N. de Sousa, J. PPapa, and A. X. Falcao, "A new approach for nontechnical losses detection based on optimum path forest," *IEEE Trans. Power Syst.*, vol. 26, no. 1, pp. 181–189, Feb. 2011.
- [5] C. Liu, S. Ghosal, Z. Jiang, S. Sarkar, "An unsupervised anomaly detection approach using energy-based spatiotemporal graphical modeling", *Cyber-Phys. Syst.*, vol. 3, no. 1, pp. 66–102, 2017
- [6] D.Z. Du, F. K. Hwang, *Combinatorial Group Testing and Its Applications*, Singapore:World Scientific, vol. 12, 2000.
- [7] E. McCary and Y. Xiao, "Malicious device inspection home area network in smart grids," *Int. J. Sensor Netw.*, vol. 25, no. 1, pp. 45–62, 2017.
- [8] F. K. Hwang, "A method for detecting all defective members in a population by group testing", *J. Amer. Stat. Assoc.*, vol. 67, no. 339, pp. 605–608, 1972.
- [9] G. Xu, W. Yu, D. Griffith, N. Golmie, and P. Moulema, "Toward integrating distributed energy resources and storage devices in smart grid," *IEEE Internet Things J.*, vol. 4, no. 1, pp. 192–204, Feb. 2017.
- [10] Hosni, N. Hamdi, "Distributed cooperative spectrum sensing with wireless sensor network cluster architecture for smart grid communications", *Int. J. Sensor Netw.*, vol. 24, no. 2, pp. 118–124, 2017.
- [11] J.Nagi, K. S. Yap, S. K. Tiong, S. K.Ahmed, and A. M. Mohammad "Detection of abnormalities and electricity theft using genetic support Vector machines," in *Proc. IEEE Region Conf. TENCON*, pp. 1–6. Nov. 2008
- [12] J. Liu, Y. Xiao, S. Li, W. Liang, and C. L. P. Chen, "Cyber security and privacy issues in smart grids," *IEEE Commun. Surveys Tuts.* vol. 14, no. 4, pp. 981–997, 4th Quart. 2012.
- [13] J. Gao, Y. Xiao, J. Liu, W. Liang, C. L. P. Chen, "A survey of communication/networking in smart grids", *Future Generat. Comput. Syst.*, vol. 28, no. 2, pp. 391–404, 2012.
- [14] K. A. Seger, D. J. Icové, "Power theft: The silent crime", *FBI Law Enforcement Bull.*, vol. 57, no. 3, pp. 20–25, Mar. 1988.
- [15] M. Faisal, A. A. Cardenas, "Incomplete clustering of electricity consumption: An empirical analysis with industrial and residential datasets", *Cyber-Phys. Syst.*, vol. 3, no. 1, pp. 42–65, 2017.
- [16] P. Jokar, N. Arianpoo, and V. C. M. Leung, "Electricity theft detection in AMI using customers' consumption patterns," *IEEE Trans. Smart Grid*, vol. 7, no. 1, pp. 216–226, May 2016.
- [17] Q. Yang, J. Yang, W. Yu, D. An, N. Zhang, and W. Zhao, "On false data-injection attacks against power system state estimation: Modeling and countermeasures," *IEEE Trans. Parallel Distrib. Syst.*, vol. 25, no. 3, pp. 717–729, Mar. 2014.



[18] Q. Yang, D. An, R. Min, W. Yu, X. Yang, and W. Zhao, "On optimal PMU placement-based defense against data integrity attacks in smart grid," IEEE Trans. Inf. Forensics Security, vol. 12, no. 7, pp. 1735–1750, Jul. 2017.

[19] R. D. Trevizan et al., "Non-technical losses identification using optimum-path forest and state estimation," in Proc. IEEE Eindhoven PowerTech, Eindhoven, The Netherlands, pp. 1–6, Jun./Jul. 2015.

[20] S. Ma, Y. Yang, Y. Qian, H. Sharif, and M. Alahmad, "Energy harvesting for wireless sensor networks: Applications and challenges in smart grid," Int. J. Sensor Netw., vol. 21, no. 4, pp. 226–241, 2016.

[21] T. Yucelen, W. M. Haddad, and E. M. Feron, "Adaptive control architectures for mitigating sensor attacks in cyber-physical systems," Cyber-Phys. Syst., vol. 2, nos. 1–4, pp. 24–52, 2016.

[22] V. Krishnan, Probability and Random Processes, Hoboken, NJ, USA: Wiley, 2006.

[23] W. Han, Y. Xiao, "Design a fast non-technical loss fraud detector for smart grid", Secur. Commun. Netw., vol. 9, no. 18, pp. 5116-5132, 2016.

[24] X. Xia, W. Xiao, Y. Liang, M. Zheng, "Coded grouping-based inspection algorithms to detect malicious meters in neighborhood area smart grid", Comput. Secur, vol. 77, pp. 547-564, Aug. 2018.

[25] Z. Xiao, Y. Xiao, and D. H. C. Du, "Exploring malicious meter inspection in neighborhood area smart grids," IEEE Trans. Smart Grid, vol. 4, no. 1, pp. 214–226, Mar 2013.

[26] Z. Ling, J. Luo, Y. Xu, C. Gao, K. Wu, X. Fu, "Security vulnerabilities of Internet of Things: A case study of the smart plug system", IEEE Internet Things J., vol. 4, no. 6, pp. 1899-1909, Dec. 2017.

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