

Paper Presentation 2 - Privacy in the smart grid

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Smart Grid Privacy via Anonymization of Smart Metering Data

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Abstract—The security and privacy of future smart grid and smart metering networks is important to their rollout and eventual acceptance by the public. Research in this area is ongoing and smart meter users will need to be reassured that their data is secure. This paper describes a method for securely anonymizing frequent (for example, every few minutes) electrical metering data by a smart meter. Although such frequent metering data may be required by a utility or electrical distribution network for operational reasons, this data may not necessarily need to be attributable to a specific smart meter or customer. It does, however, need to be securely attributable to a specific location (e.g. a group of houses or apartments) within the electricity distribution network. The method described in this paper provides a first step towards mechanisms for authorized anonymous meter readings which are difficult to associate with a particular smart meter or customer. This method does not preclude the provision of attributable metering data that is required for other purposes such as billing, account management or marketing research purposes.

1. INTRODUCTION

Historically, the electrical grid of each country has been a “broadcast” grid, where a few central power generation (i.e. power stations) produce electricity to cover demand in a country or region, and distribute this electricity to the end users via a large network of cables and transformers. This model has served well for the last century or so, there is a growing need to reform the world’s electrical grids, both from an aging infrastructure point of view and to address new environmental and societal challenges. In response to this need, national governments and relevant stakeholders are making significant efforts in the development of future electrical grids or “Smart Grids”, see examples in [1] and [2].

Development of this new grid will require significant efforts in technology development, standards, policy and regulatory activities because of its inherent complexity. Smart Metering [3] is a key component of the future smart grid.

Security and privacy are considered to be of prime importance to smart grids, given how easily large networks, such as the Internet, can be hacked. This paper focuses on the privacy aspect of smart metering data, discussing its importance and vulnerabilities and proposes a solution for anonymizing high-frequency metering data through the use of a pseudonym ID without compromising the operation of the utility and/or distribution network.

Analysis of the Impact of Data Granularity on Privacy for the Smart Grid

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ABSTRACT

The upgrade of the electricity network to the “smart grid” has been intensified in the last years. The new automated devices being deployed gather large quantities of data for each offer premises of a new metered grid but also raise privacy concerns among customers and energy distributors. In this paper, we focus on the energy consumption traces that smart meters generate and especially on the risk of being able to identify individual customers given a large dataset of those traces. This is a question raised in the recent literature on smart or important privacy research topic. We present an overview of the current research regarding privacy in the Advanced Metering Infrastructure. We make a formalization of the problem of de-anonymization by matching low-frequency and high-frequency smart metering datasets and we also build a threat model related to this problem. Finally, we investigate the characteristics of these datasets in order to make them more resilient to the de-anonymization process. Our methodology can be used by electricity companies to better understand the properties of their smart metering datasets and the conditions under which such datasets can be released to third parties.

Keywords

Smart grid data privacy; Advanced Metering Infrastructure (AMI) data characteristics; Smart meter privacy; Smart metering data

Categories and Subject Descriptors

H.4 [Information Systems Applications]: Miscellaneous; K.4 [Computers and society]: Public Policy Issues; Smart Metering Privacy

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I. INTRODUCTION

In any new domain where significantly more data starts being produced, the privacy of the customer who produces these data may be at risk. This is also the case in the new smart grid which is the name used for the modern electrical grid. One of the main differences between the traditional electrical grid and the new smart grid is the large number of computing and communication devices being installed in different parts of the grid and that are connected through an overlay communication network, their main purpose is to make the grid monitoring and operational processes more accurate and more efficient.

This computing and communication device are deployed in all of the three main sections of the electrical network: the generation section, the transmission section and the distribution section. Specifically, in the distribution section, the traditional electro-mechanical meters that used to monitor the electrical energy consumed by the end customers are replaced by the new so-called smart meters. The smart meters, together with other devices that monitor, gather and send their data to the energy distributor’s central location form the Advanced Metering Infrastructure (AMI). The AMI offers two-way communication between the central control system and the smart meters, resulting in better remote functionality of the smart meters, such as remote shut-off commands and control of demand-side electricity load and generation. Figure 1 presents an overview of the AMI, together with a categorization of the different types of communication media (radio, wired, fiber-optic) and protocols used (HART, Power Line Communication, ZigBee, GPRS) in suggested applications.

As a consequence of the upgrade to the smart grid, significantly more data is collected and analyzed, for example in the AMI where more parameters than just the the electrical energy consumed by customers are recorded, at a higher frequency than before. It is estimated that the size of the smart grid will be larger than the total size of the Internet, the quantity of data produced will be considerable. These data are expected to play a key role in the development of the smart grid and will improve the balance between energy production and energy consumption by making a significant contribution in improving electrical grid stability and energy efficiency.

*http://www.east.com/830/-115128,-3-1024102-64.html

Smart Metering De-Pseudonymization

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ABSTRACT

Consumption traces collected by Smart Meters are highly privacy sensitive data. For this reason, current best practice is to store and process such data in pseudonymized form, equipping identity information from the consumption traces. However, even the consumption traces alone provide more valuable information if combined with limited external information. Based on this observation, we identify two attack vectors: one aimed at identifying and behavior pattern matching that allow effective de-pseudonymization. Using a practical evaluation with real-life consumption traces of 32 households, we verify the feasibility of our techniques and show that the attacks are robust against common countermeasures, such as reduction in resolution or frequent re-pseudonymization.

1. INTRODUCTION

The deployment of Smart Metering – the digital recording and processing of electricity consumption – is over increasing. A Smart Meter is an electrical meter that records a fine-grained consumption trace of a household and sends it to the respective electricity supplier. These consumption traces, in contrast to traditional single annual consumption values, allow the realization of time-of-use tariffs and demand response schemes.

This flexibility, however, comes at a price. Every activity that takes place in the household and makes use of electrical appliances is reflected in the consumption trace. In consequence, Smart Metering has repeatedly been called a privacy intrusion into households [7, 8] and a large body of previous work [5, 6, 11, 12, 14, 15, 20] has been concerned with limiting privacy information from energy consumption traces.

Based on the identified privacy implications, there is consensus that consumption data of Smart Metering needs to be adequately protected. Such protection entails the protection during storage by the supplier and during the use of the data by the supplier and 3rd party contractors. Pseudonymization is one of the most common techniques used to protect data from unauthorized access.

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anonymization of consumption traces is considered an effective defense against privacy attacks and allows for retaining the identity of the household and its consumption trace. The consumer’s identity can be stored independently from consumption traces, only linked by the pseudonym. In such a scenario, the privacy-trading method developed in previous work can be applied by the owner of both the identity database and the consumption traces. An attacker faces two problems, if he has only access to pseudonymized traces: First, detection from pseudonymized consumption traces is more prone as an identity information can be used as contextual data. Second and more important, all information inferred from consumption traces cannot be attributed to a specific household due to the unavailability introduced by pseudonymization. This makes consumption traces and its contained information unattractive for targeted abuse and appropiately the consumer’s privacy is protected.

In this paper, we develop two attack vectors targeting the privacy of pseudonymized consumption traces. The first attack allows to create a link between a household’s identity and its consumption trace, and therefore enables an attacker to undo pseudonymization. If successful, the attack allows all existing detection attacks to be applied again. The second method enables an attacker to track the origin of a consumption trace across re-pseudonymization or across different databases. For conducting these attacks in practice, we provide a data analysis framework that allows an attacker to apply either method to consumption databases. The paper’s main contributions are as follows:

1. An abstract definition of attack vectors on the unavailability of pseudonymized Smart Metering consumption traces.
2. A machine learning framework for the analysis of consumption traces and subsequent execution of above-mentioned attack vector.
3. Experimental findings about the anomaly detection in consumption traces and the tracking of consumption traces across pseudonyms.

An evaluation of different mitigation techniques with respect to their effectiveness against these attacks is provided in Section 5. The structure of this paper is as follows: In Section 2 we provide an overview of the terminology used in this paper. Section 3 describes the two attack vectors that we identified as de-pseudonymization attacks. In Section 4, we present

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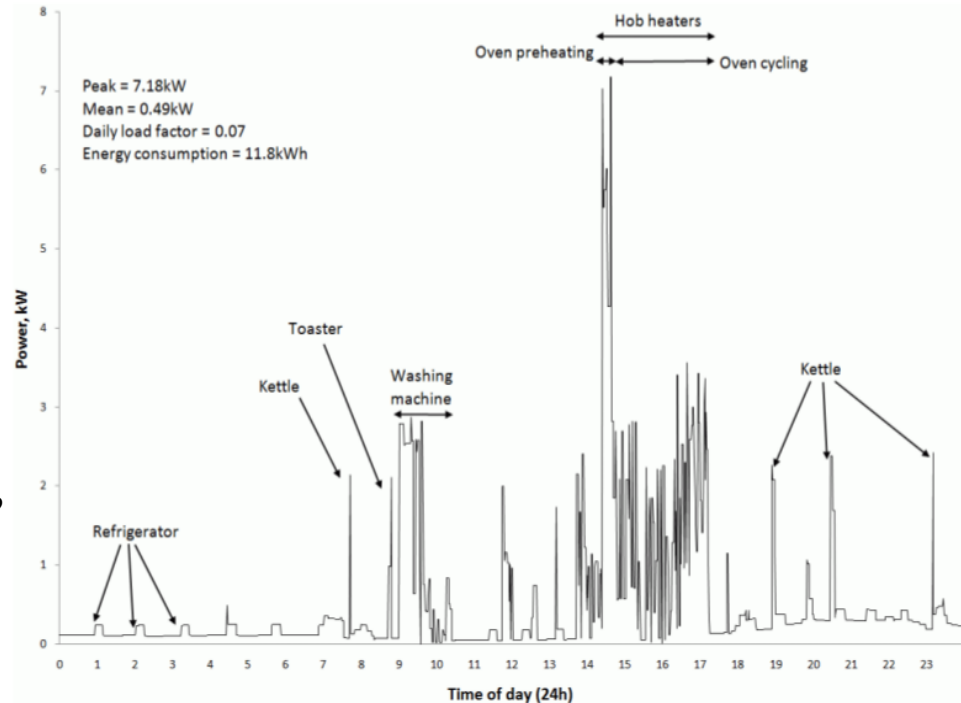
Tudor et al. - Analysis of the
impact of data granularity on
privacy for the smart grid

Jawurek et al. - Smart
metering de-
pseudonymization

Problem Description

- “High Frequency” metering data.
 - About every 5 minute
 - Electric data from home
- “Low Frequency” metering data.
 - Weekly/Monthly
 - Meter reading for billing

How can we anonymize high frequency data?



Picture: E. L. Quinn, “Privacy and the New Energy Infrastructure”,
Social

Science Research Network (SSRN), February 2009

Method(1)

HFID = High Frequency ID

LFID = Low Frequency ID

- HFID should never be known to the power company or the smart meter installer
- HFID hardcoded by the manufacturer
 - 3rd party escrow
 - Manufacturer is not expected to manage any data
 - Manufacturer requires a strong data privacy policy to ensure the secret of the relation between LFID and HFID
- Secure protocol setup mechanism
- The protocol is not perfect w.r.t privacy protection but described as a step in the right direction

Method(2)

- Client Data Profile(CDP)
 - Initial process done to identify the client
 - Client <-> Power Company
 - LFID included
- Anonymous Data Profile(ADP)
 - Initiated after the CDP process.
 - Power Company <-> Escrow
 - Escrow <-> Client
 - HFID included

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Development of this new grid will require significant efforts in technology development, standards, policy and regulatory activities because of its inherent complexity. Smart Metering [3] is a key component of the future smart grid. Security and privacy are considered to be of prime importance to smart grids, given how easily large networks, such as the Internet, can be hacked. This paper focuses on the privacy aspect of smart metering data, discussing its importance and vulnerabilities and proposes a solution for anonymizing high-frequency metering data through the use of a pseudonym ID without compromising the operation of the utility and/or distribution network.

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1. INTRODUCTION

In any new domain where significantly more data starts being produced, the privacy of the customer who produces these data may be at risk. This is also the case in the new smart grid which is the name used for the modern electrical grid. One of the main differences between the traditional electrical grid and the new smart grid is the large number of computing and communication devices being installed in different parts of the grid and that are connected through an overlay communication network; their main purpose is to transfer the grid monitoring and operational processes more accurate and more efficient.

This computing and communication device are deployed in all of the three main sections of the electrical network: the generation section, the transmission section and the distribution section. Specifically, in the distribution section, the traditional electro-mechanical meters that used to monitor the electrical energy consumed by the end customers are replaced by the new so-called smart meters. The smart meters, together with other devices that monitor, gather and send their data to the energy distributor’s central location form the Advanced Metering Infrastructure (AMI). The AMI offers two-way communication between the central control system and the smart meters, resulting in better remote functionality of the smart meters, such as remote shut-off commands and control of demand-side electricity load and generation. Figure 1 presents an overview of the AMI, together with a categorization of the different types of communication media (radio, wired, fiber-optic) and protocols used (Ethernet, Power Line Communication, ZigBee, GPRS) in suggested applications.

As a consequence of the upgrade to the smart grid, significantly more data is collected and analyzed, for example in the AMI where more parameters than just the the electrical energy consumed by customers are recorded, at a higher frequency than before. It is estimated that the size of the smart grid will be larger than the total size of the Internet, the quantity of data produced will be considerable. These data are expected to play a key role in the development of the smart grid and will improve the balance between energy production and energy consumption by making a significant contribution in improving electrical grid stability and energy efficiency.

*http://www.east.com/830/-151128,-3-10241052-64.html

Smart Metering De-Pseudonymization

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ABSTRACT

Consumption traces collected by Smart Meters are highly privacy sensitive data. For this reason, current best practice is to store and process such data in pseudonymized form, equipping identity information from the consumption traces. However, even the consumption traces alone provide more valuable information if combined with limited external information. Based on this observation, we identify two attack vectors: one aimed at the identity and behavior pattern matching that allow effective de-pseudonymization. Using a practical evaluation with real-life consumption traces of 32 households, we verify the feasibility of our techniques and show that the attack is robust against common countermeasures, such as resolution reduction or frequent re-pseudonymization.

1. INTRODUCTION

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This flexibility, however, comes at a price. Every activity that takes place in the household and makes use of electrical appliances is reflected in the consumption trace. In consequence, Smart Metering has repeatedly been called a privacy intrusion into households [7, 8] and a large body of previous work [5, 6, 11, 12, 14, 15, 20] has been concerned with limiting privacy intrusion by energy consumption traces.

Based on the identified privacy implications, there is consensus that consumption data of Smart Metering needs to be adequately protected. Such protection entails the protection during storage by the supplier and during the use of the data by the supplier and 3rd party contractors. Pseudonymization is the most common technique for this purpose.

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anonymization of consumption traces is considered an effective defense against privacy attacks, and allows for retaining the identity of the household and its consumption trace. The consumer’s identity can be linked independently from consumption traces, only linked by the pseudonym. In such a scenario, the privacy-trading method developed in previous work can only be applied by the owner of both the identity database and the consumption traces.

An attacker faces two problems, if he has only access to pseudonymized traces. First, distinction from pseudonymized consumption traces to reveal precise an identity information can be used as contextual data. Second and more important, all information inferred from consumption traces must be attributed to a specific household due to the unidirectionality introduced by pseudonymization. This makes consumption traces and its contained information much harder to target alone and aggregate the consumers’ privacy is protected.

In this paper, we develop two attack vectors targeting the privacy of pseudonymized consumption traces. The first attack allows to create a link between a household’s identity and its consumption trace, and therefore enables an attacker to undo pseudonymization. If successful, the attack allows all existing detection attacks to be applied again. The second method enables an attacker to track the origin of a consumption trace across re-pseudonymization or across different databases. For conducting these attacks in practice, we provide a data analysis framework that allows an attacker to apply either method to consumption databases. The paper’s main contributions are as follows:

1. An abstract definition of attack vectors on the unidirectionality of pseudonymized Smart Metering consumption traces.
2. A machine learning framework for the analysis of consumption traces and subsequent execution of above-mentioned attack vector.
3. Experimental findings about the anomaly detection in the consumption traces and the tracking of consumption traces across pseudonyms.

4. An evaluation of different mitigation techniques with respect to their effectiveness against these attacks.

The structure of this paper is as follows. In Section 2, we provide an overview of the terminology used in this paper. Section 3 describes the two attack vectors that we identified as de-pseudonymization attacks. In Section 4, we present

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Problem Description

- Matching high-frequent data with low-frequent data => Customer Identity
- $\text{Sum}(\text{High Frequent Data for Time Period}) = \text{Low Frequent data}$

Method

- What if the granularity is rounded to every 10 kWh instead of 1 kWh

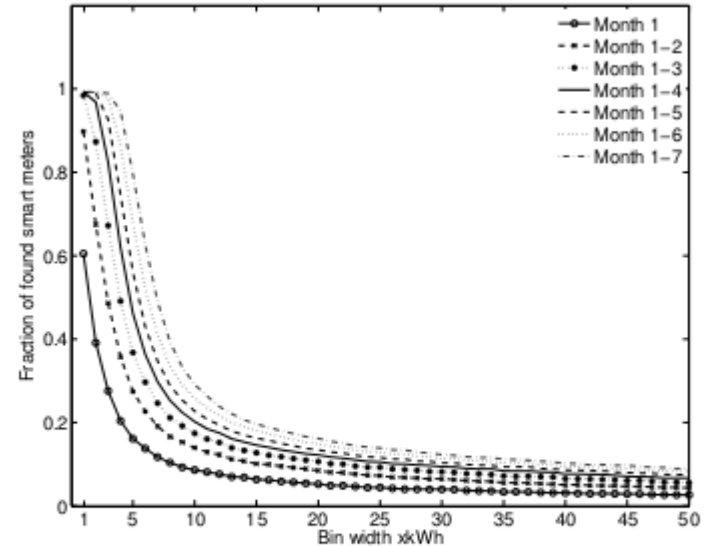


Figure 5: Fraction of unique smart meters - seven months of data - dataset case

Time period	Newly found smart meters		Total found smart meters %	
	Simulation	Evaluation	Simulation	Evaluation
m_1	18461	11698	95.4%	60.5%
m_2	871	5655	99.9%	89.7%
m_3	2	1669	100 %	98.3%
m_4	0	155	100 %	99.1%
m_5	0	11	100 %	99.2%
m_6	0	11	100 %	99.3%
m_7	0	10	100 %	99.3%
Total	19334	19209	100 %	99.3%

Table 3: Expected number of identified smart meters for a reporting granularity of 1 kWh

Time period	Newly found smart meters		Total found smart meters %	
	Simulation	Evaluation	Simulation	Evaluation
m_1	12182	1670	63.0%	8.6%
m_2	6029	1027	94.1%	13.9%
m_3	1093	671	99.8%	17.4%
m_4	30	543	100 %	20.2%
m_5	0	487	100 %	22.7%
m_6	0	579	100 %	25.7%
m_7	0	651	100 %	29.1%
Total	19334	5628	100 %	29.1%

Table 4: Expected number of identified smart meters for a reporting granularity of 10 kWh

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1. INTRODUCTION

Historically, the electrical grid of each country has been a “broadcast” grid, where a few central power generation (i.e. power stations) produce electricity to cover demand in a country or region, and distribute this electricity to the end users via a large network of cables and transformers. This model has served well for the last century or so, there is a growing need to reform the world’s electrical grids, both from an aging infrastructure point of view and to address new environmental and societal challenges.

An example of this is given in Fig. 1, reproduced from [4], which discusses these privacy concerns at length with regards to expected and/or projected availability of high-frequency metering data. Another good argument for privacy is given in a recent paper on “Digital homes” and smartification [5]. There is a rich literature in load signature algorithms which use energy measurements to extract detailed information regarding domestic appliance usage. This research is typically termed NALM (Non-Intrusive Appliance Load Monitoring), which is discussed in [6]. There is an active line of research in the construction and supply of appliance libraries and detection algorithms, see, for example, [7].

In this paper, we address the privacy problems by anonymizing smart metering data so that information gleaned from it cannot easily be associated with an identified person, the utility and/or distribution network.

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I. INTRODUCTION

In any new domain where significantly more data starts being produced, the privacy of the customer who produces these data may be at risk. This is also the case in the new smart grid which is the name used for the modern electrical grid. One of the main differences between the traditional electrical grid and the new smart grid is the large number of computing and communication devices being installed in different parts of the grid and that are connected through an overlay communication network; their main purpose is to make the grid monitoring and operational processes more accurate and more efficient.

This computing and communication device are deployed in all of the three main sections of the electrical network: the generation section, the transmission section and the distribution section. Specifically, in the distribution section, the traditional electro-mechanical meters that used to monitor the electrical energy consumed by the end customers are replaced by the new so-called smart meters. The smart meters, together with other devices that monitor, gather and send their data to the energy distributor’s central location form the Advanced Metering Infrastructure (AMI). The AMI offers two-way communication between the central control system and the smart meters, resulting in better remote functionality of the smart meters, such as remote shut-off commands and control of demand-side electricity load and generation. Figure 1 presents an overview of the AMI, together with a classification of the different types of communication media (radio, wired, fiber-optic) and protocols used (HART, Power Line Communication, ZigBee, GPRS) in suggested applications.

As a consequence of the upgrade to the smart grid, significantly more data is collected and analyzed, for example in the AMI where more parameters than just the the electrical energy consumed by customers are recorded, at a higher frequency than before. It is estimated that the size of the smart grid will be larger than the total population and the quantity of data produced will be considerable. These data are expected to play a key role in the development of the smart grid and will improve the balance between energy production and energy consumption by making a significant contribution in improving electrical grid stability and energy efficiency.

*http://www.east.com/830/-/11128/-/11041102-64.
HAL

Smart Metering De-Pseudonymization

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ABSTRACT

Consumption traces collected by Smart Meters are highly privacy sensitive data. For this reason, current best practice is to store and process such data in pseudonymized form, equipping identity information from the consumption traces. However, even the consumption traces alone provide more valuable information if combined with limited external information. Based on this observation, we identify two attack vectors: an attacker can combine and behave pattern matching that allow effective de-pseudonymization. Using a practical evaluation with real-world consumption traces of households, we verify the feasibility of our techniques and show that the attacks are robust against common countermeasures, such as resolution reduction or frequent re-pseudonymization.

1. INTRODUCTION

The deployment of Smart Metering – the digital recording and processing of electricity consumption – is ever increasing. A Smart Meter is an electrical meter that records a fine-grained consumption trace of a household and sends it to the respective electricity supplier. These consumption traces, in contrast to traditional single-point consumption values, allow the realisation of time-of-use tariffs and demand response schemes.

This flexibility, however, comes at a price. Every activity that takes place in the household and makes use of electrical appliances is reflected in the consumption trace. In consequence, Smart Metering has repeatedly been called a privacy intrusion into households [7, 8] and a large body of previous work [5, 6, 11, 12, 14, 15, 20] has been concerned with limiting privacy intrusion by energy consumption traces.

Based on the identified privacy implications, there is consensus that consumption data of Smart Metering needs to be adequately protected. Such protection entails the protection during storage by the supplier and during the use of the data by the supplier and 3rd party contractors. Pseudonymization is a well-known technique for this purpose.

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anonymization of consumption traces is considered an effective defense against privacy attacks, and allows for retaining the identity of the household and its consumption trace. The consumer’s identity can be stored independently from consumption traces, only linked by the pseudonym. In such a scenario, the privacy-trading method developed in previous work can be applied by the owner of the household identity database and the consumption traces. An attacker faces two problems, if he has only access to pseudonymized traces: First, detection from pseudonymized consumption traces is more prone as an identity information can be used as contextual data. Second and more important, all information inferred from consumption traces cannot be attributed to a specific household due to the unavailability introduced by pseudonymization. This makes consumption traces and its contained information unattractive for targeted abuse and appropiately the consumers’ privacy is protected.

In this paper, we develop two attack vectors targeting the privacy of pseudonymized consumption traces. The first attack allows to create a link between a household’s identity and its consumption trace, and therefore enables an attacker to undo pseudonymization. If successful, the attack allows all existing detection attacks to be applied again. The second method enables an attacker to track the origin of a consumption trace across re-pseudonymization or across different databases. For conducting these attacks in practice, we provide a data analysis framework that allows an attacker to apply either method to consumption databases. The paper’s main contributions are as follows:

1. An abstract definition of attack vectors on the unavailability of pseudonymized Smart Metering consumption traces.
2. A machine learning framework for the analysis of consumption traces and subsequent execution of above-mentioned attack vectors.
3. Experimental findings about the anomaly detection in consumption traces and the tracking of consumption traces across pseudonyms.

An evaluation of different mitigation techniques with respect to their effectiveness against these attacks. This is the main part of this paper and is best read as in Section 2. In Section 2 we provide an overview of the terminology used in this paper. In Section 3 we describe the two attack vectors that we identified as de-pseudonymization attacks. In Section 4, we present

Costas Efthymiou and
Georgios Kalogeris - Smart
Grid Privacy via
Anonymization of Smart
Metering Data

Tudor et al. - Analysis of the
impact of data granularity on
privacy for the smart grid

Jawurek et al. - Smart
metering de-
pseudonymization

Two types of attack

Linking by behaviour anomaly

Unique event creates a peak or valley in the consumption trace

Linking by Behavior Pattern

Tracks the origin of a consumption trace

- Multiple pseudonyms
- Multiple databases

Possible ways to protect against the attacks

- Create new pseudonyms more often to confuse the attacker and harder to track
 - Overhead for storage
 - Maybe the attacker can follow the trace anyway?
- Lower Resolution of Smart metering
 - Proved in the paper that the linking accuracy drops significantly

Not discussed in the papers

- Proper protection during storage of the data
- Cryptographic methods
- **Politics:** Under what circumstances should the identity be revealed?
 - Court order, police suspect something illegal
 - Employer spy on workers who called in sick
 - Power theft

Questions?