

UPEC 2018

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Stochastic Technical Losses Analysis of Smart Grids under Uncertain Demand

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Universidad Carlos III de Madrid
 Naturgy (Gas Natural Fenosa)









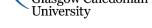


Contents



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- 1. Introduction
- 2. Methodology
- 3. Case Study
- 4. Conclusions





Introduction

1. Introduction

- Technical Losses in Low Voltage (LV) distribution networks
 power flow (measurements)
- LV distribution networks are still in transition (not 100% Smart Grids). Power losses can not be substracted from the measurements.
- Problem: Presence of non-telemetered customers = Uncertain Demand
- Losses accuracy → load demand resolution
- high-resolution load demand profiles would allow a more accurate technical losses analysis



Introduction



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1. Introduction

Upper Level

Middle Level

Lower Level

High resolution

Monthly Energy Estimation

Daily Energy Estimation

Hourly Energy Estimation

Intra-Hour load demand



Introduction



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1. Introduction

 Solution proposed: a stochastic top-down approach for uncertain load demand estimation





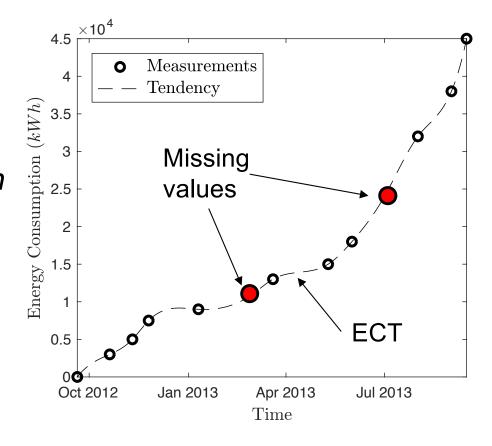
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2. Methodology

Upper Level

- Non- Telemetered customers
- Energy Consumption
 Tendency (ETC) Curve

Monthly Energy Estimation







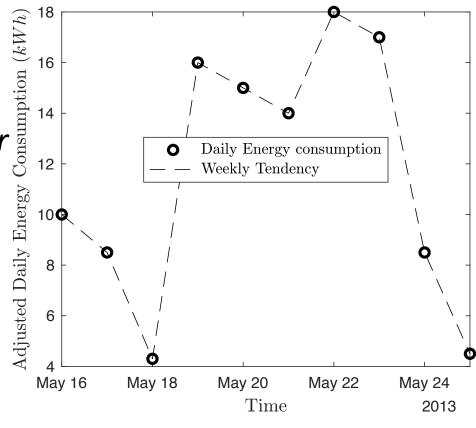
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2. Methodology

Middle Level

Daily Energy Estimation

- Weekly Energy
 Consuption (WEC) profile
- Supervisor Smart Meter In Secondary Substation





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2. Methodology

Hourly Energy Estimation

Lower Level

NLP Optimization : Minimize cuadratic error Minimize: Objective function

$$+\sum_{d\in\Omega_d}\sum_{h=1}^{24}\left(\sum_{i=\Omega_c}P_{(i,h,d)}-\widehat{P}_{(h,d)}^s\right)^2$$
(1a)

Subject to:

Constraints

$$\sum_{i \in \Omega_c} \sum_{h=1}^{24} P_{(i,h,d)} = \sum_{i \in \Omega_c} E_{(i,d)} \,\,\forall \, d \in \Omega_d \tag{1b}$$

$$\sum_{i \in \Omega_c} \sum_{h=1}^{24} P_{(i,h,d)} \le \sum_{h=1}^{24} P_{(s,h,d)}; \forall d \in \Omega_d$$

$$\sum_{i \in \Omega_c} E_{(i,d)} \le \hat{E}_{(d)}^s; \forall d \in \Omega_d$$
(1c)

$$\sum_{i \in \Omega_c} E_{(i,d)} \le \widehat{E}_{(d)}^s; \forall d \in \Omega_d$$
 (1d)



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2. Methodology

High resolution

Intra-Hour load demand

Markov Chains

$$\Pr \{X_{t+1} = j | X_0 = i_0, ..., X_{t-1} = i_{t-1}, X_t = i\}$$

$$= \Pr \{X_{t+1} = j | X_t = i\} = p^{(t)}_{i,j}$$

$$[p_{1,1} \cdots p_{1,k}]$$
(2)

$$P = \begin{bmatrix} p_{1,1} & \cdots & p_{1,k} \\ \vdots & \ddots & \vdots \\ p_{k,1} & \cdots & p_{k,k} \end{bmatrix}$$
(3)

$$\hat{f}(\tau) = \frac{1}{n\hat{h}} \int_{1}^{24} \varphi(\tau) \left(\frac{\tau - S_d(t)dt}{\hat{h}} \right) \tag{6}$$

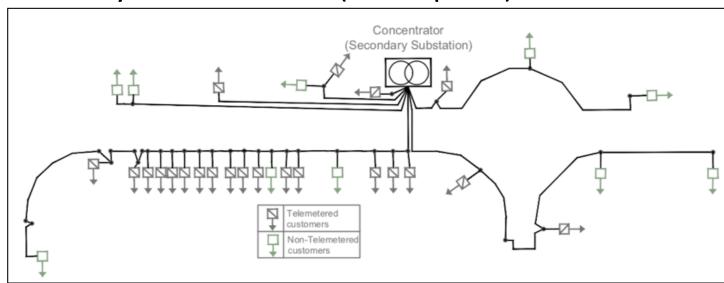




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3. Case Study

- Demonstration Smart Grid project OSIRIS (Spain)
 - Utility Gas Natural Fenosa (Naturgy)
 - 1 substation 630 kVA (8 Feeders)
 - 32 Customers (10 non-telemetered)
 - Total power contracted: 442 kW
 - Yearly data: 2013-14 (incomplete)



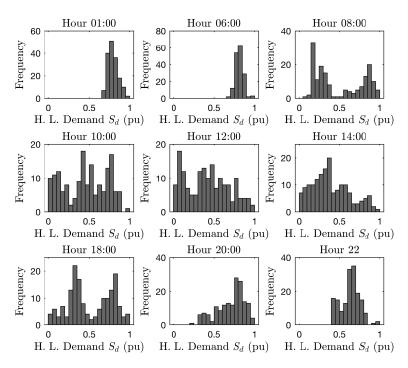




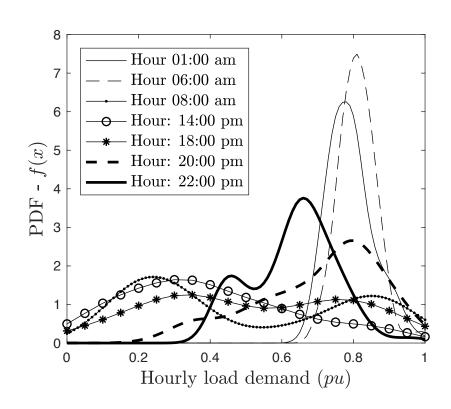
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3. Case Study

Statiscial study of load demand



PDFs estimated

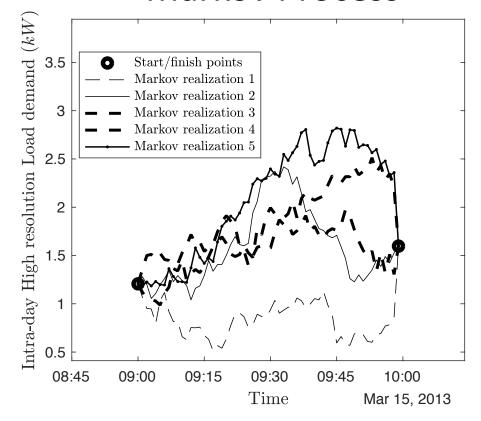




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3. Case Study

Realizations of the Markov Process

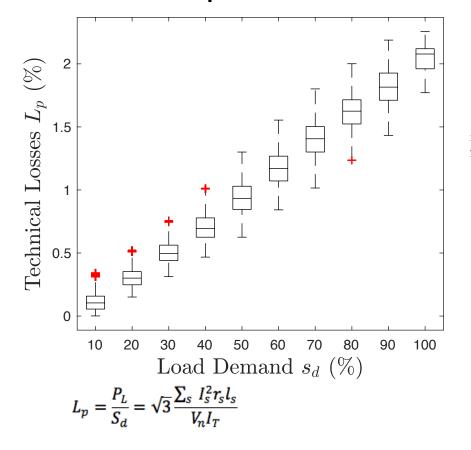




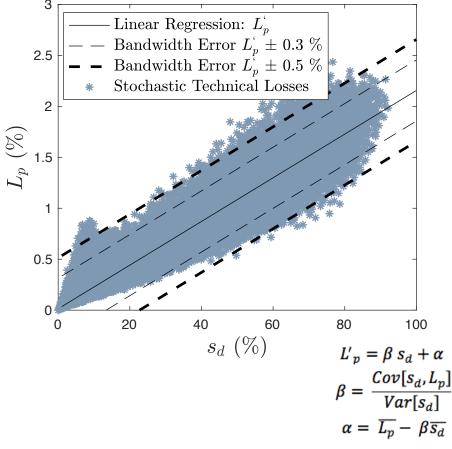
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3. Case Study

Technical Losses Calculated With power flow



Stochastic Losses Obtained

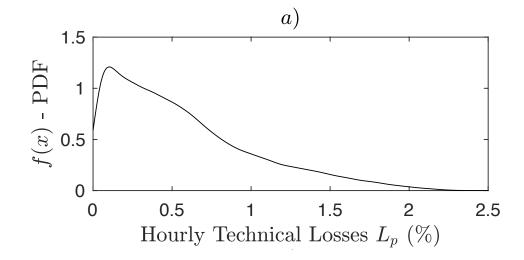




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3. Case Study

Probability Distribution Losses





Conclusions



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4. Conclusions

- A stochastic top-down method to analyze Technical Losses in (still in transition) LV distribution Networks have been proposed
- Presence of non-telemeterd customers in LV networks as well as missing load demands have been taked into account
- Missing hourly load demand values have been estimated through an NLP optimization process



Conclusions



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4. Conclusions

- Intra-hour high-resolution load demand profiles have been synthetically generated by means of Markov Chains
- For every stochastic realization of the Markov process, a balanced three-phase load flow analysis has been carried out to obtain the network losses
- Losses obtained have been statistically analyzed using a linear regression fitting to find the expected value



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Thank You

Questions?

Jose Angel Velasco











