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## Study Committee C2

**Power System Operation and Control** 

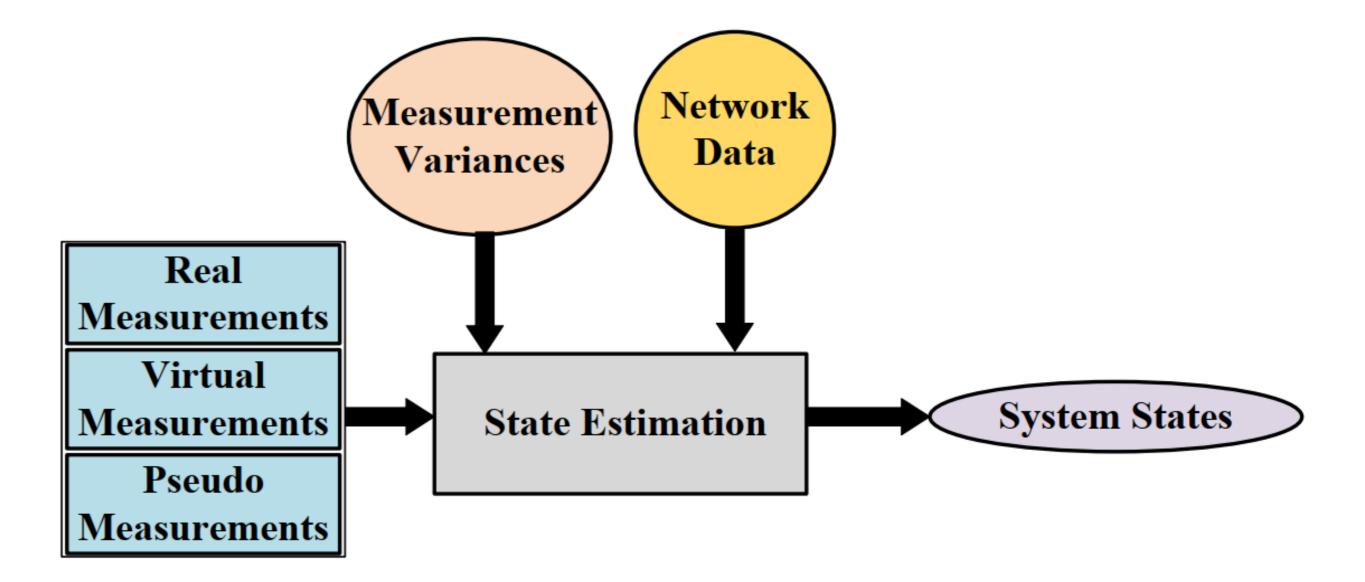
Paper ID-10548

# State Estimation in Medium Voltage Distribution Networks using **Pseudo-Measurements**

Sai Suprabhath NIBHANUPUDI<sup>1,2\*</sup>, Anton ISHCHENKO<sup>2</sup>, Simon TINDEMANS<sup>1</sup>, Peter PALENSKY<sup>1</sup>

<sup>1</sup>Delft University of Technology, <sup>2</sup>Phase to Phase BV





- Changing grid topology.
- Rise in Distributed Energy Resources (DERs) penetration.
- Important to continuously monitor the grid state.
- Distribution System State Estimator (DSSE) is of prominent interest.

#### The challenge

- Grid modelling has lot of assumptions.
- Better load modelling techniques needed for better idea.
- Planning for future scenarios.

### **Distribution System State Estimation**

Figure 2. DSSE Functional Block

# The Approach

- Currently, LV network is not modelled accurately and considered as a black box.
- Stochastic LV network power flow is proposed here.
- Detailed cable models are considered.
- Gaussian Mixture (GM) load model is used to represent

# (DSSE)

#### **Problem Formulation**

- Power system state need to be always known.
- Voltage and Current based approaches are common to formulate the states.
- Voltage based approach is taken here owing to faster convergence.
- Various inputs can be considered.

- Conventional Weighted Least Squares method is employed here.
- Variance of the input are the weights.

- individual household (probability density functions of powers).
- Each appliance (PV,EV, heat pump) is modelled with their own time-varying PDF's (in Gaia software).
- Latin Hypercube based Monte Carlo random sampling used to perform power flow calculations.
- Normal approximation is used for aggregate load.

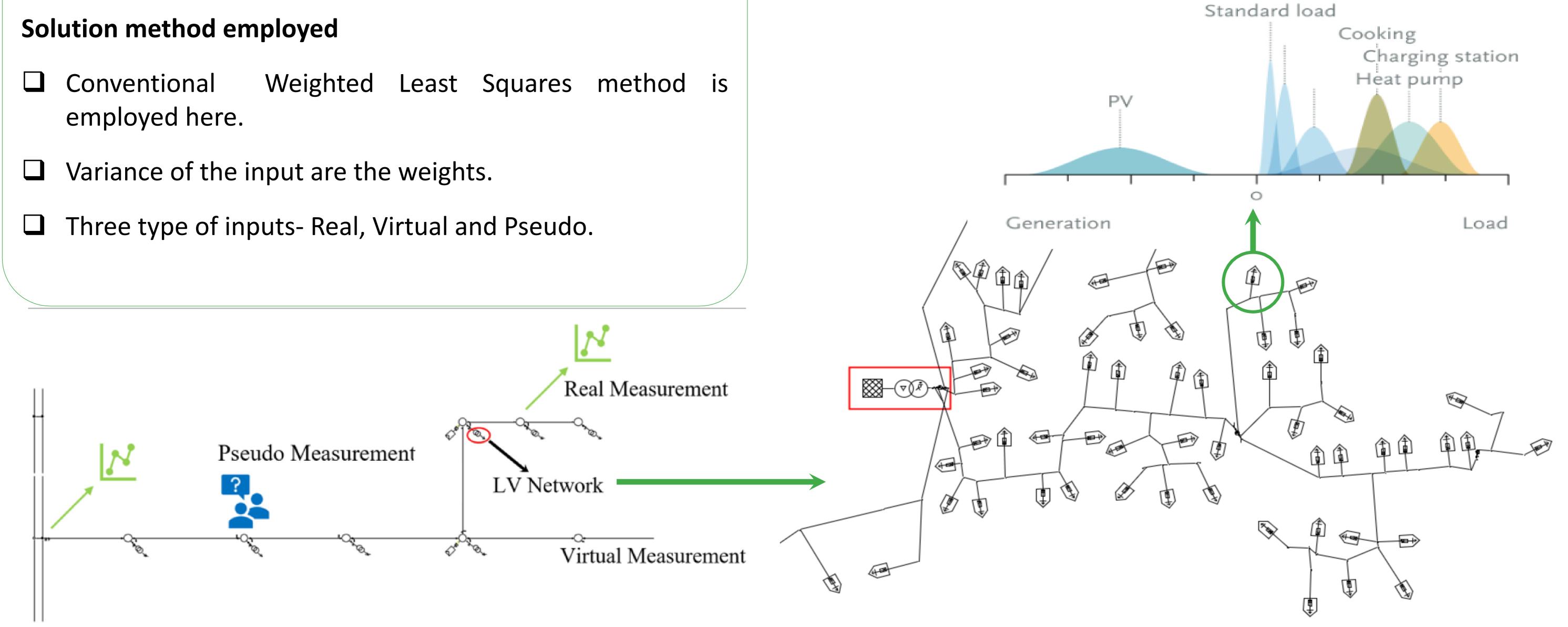


Figure 1. Grid Scenario

Figure 3. A sample LV network and Household level PDF's







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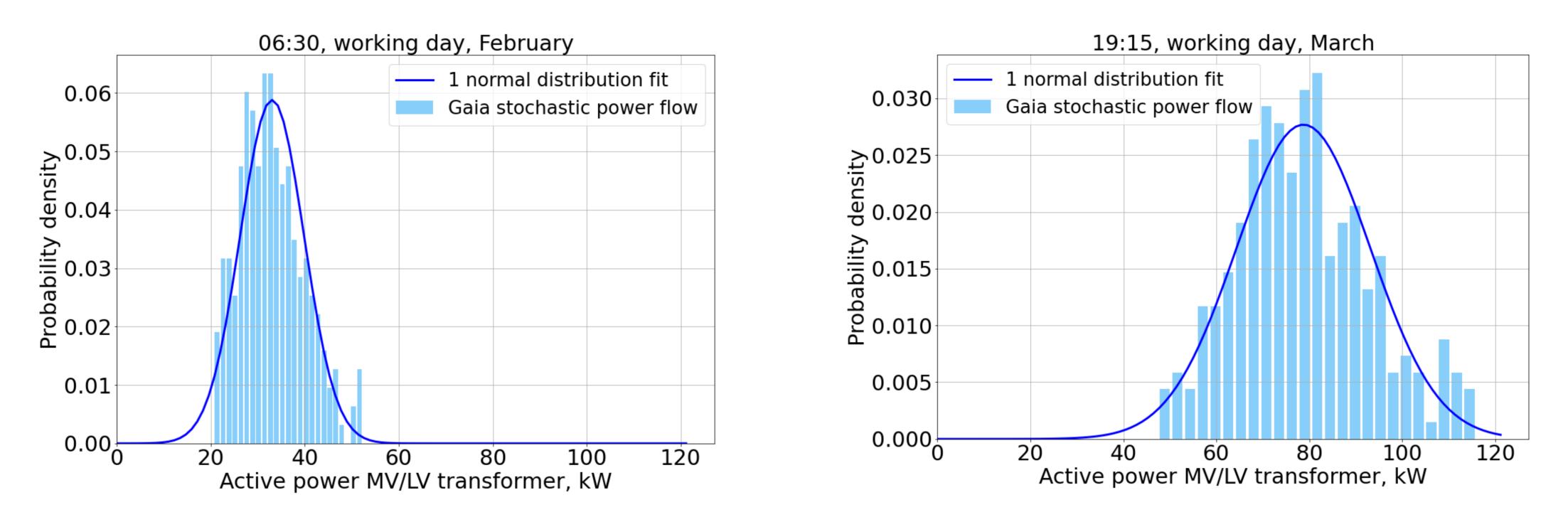
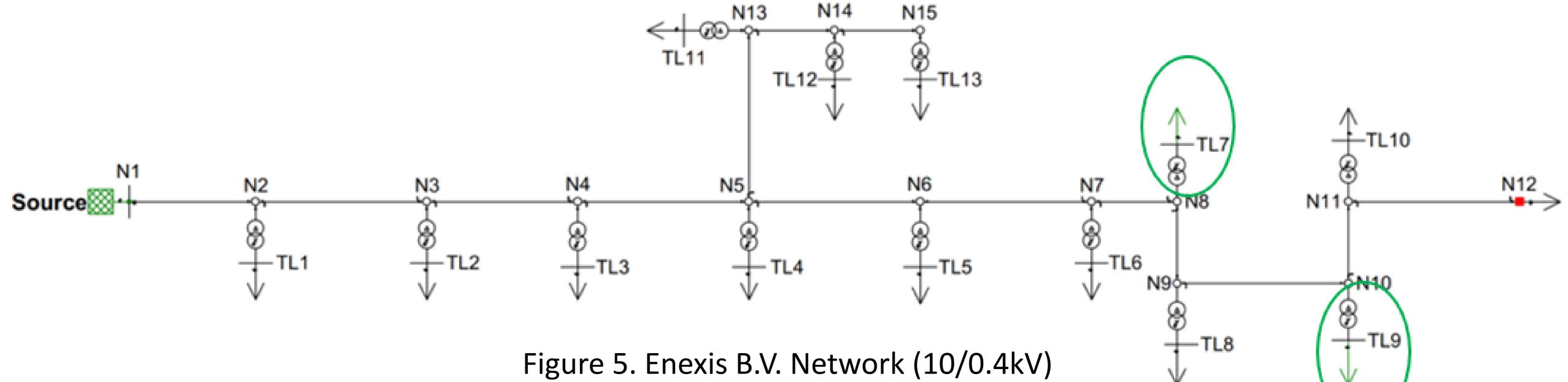


Figure 4. Normal distribution fitting of aggregated households



#### $\mathbf{v}$

### **Objects of investigation**

Two typical Dutch Distribution Networks are considered and tested with this algorithm.

#### **Enexis BV network description**

- 28 nodes : 13 connected to LV network equivalent loads and one to a MV level consumer (N12).
- General overview of the loading profile generated.
- This implies, historical data previously available is considered to generate a approximated loading profile behavior.
- Network and stochastic household behavior modelled using Gaia (Phase2Phase).

#### **Studies**

#### **Enexis Network**

• Only measurement at node TL5 is considered.

- Proposed LV stochastic load flow model is used for other nodes.
- □ The nodes in green have measurement data and are used for validation purpose.
- The validation is done for the peak load moment in April 2021 at 18:00 as shown in Figure 6.

#### **Stedin Network**

Two scenarios are considered for thirty days period with measurements being 5 min apart.

#### **Stedin BV network description**

- 28 nodes : 15 connected to household loads and one to a MV level consumer (see paper for more).
- General overview of the loading profile over the entire year generated by the company experts.

Assumption: Grid topology is constant as well as free from faults and failure in components.

• Scenario 1 takes power injections as input and scenario 2 also considers voltage measurements at the nodes.









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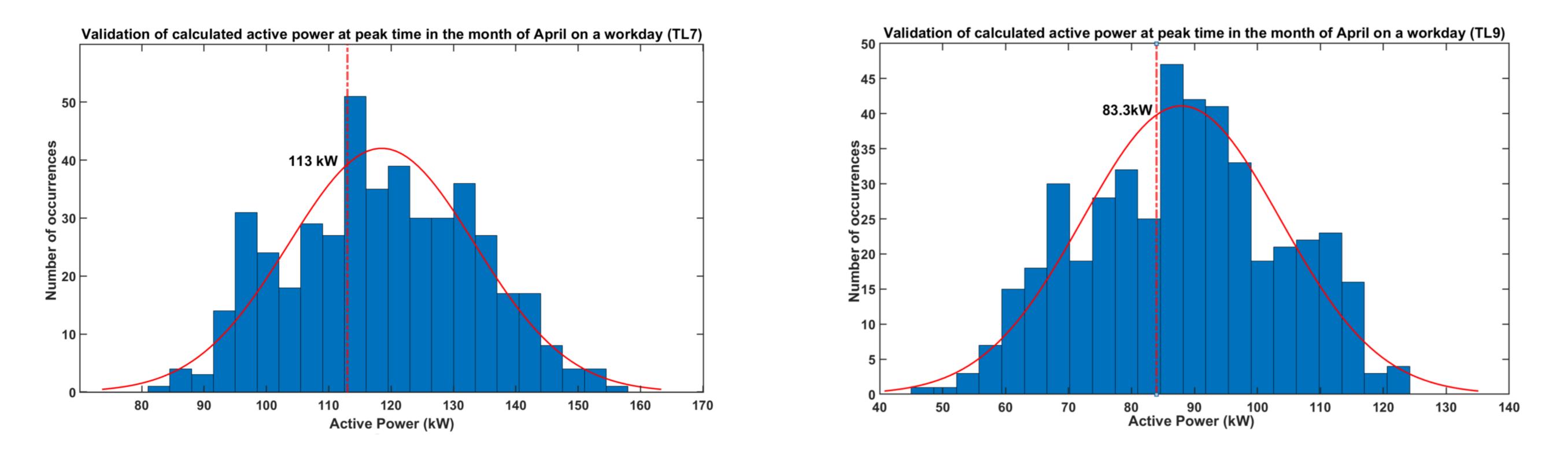
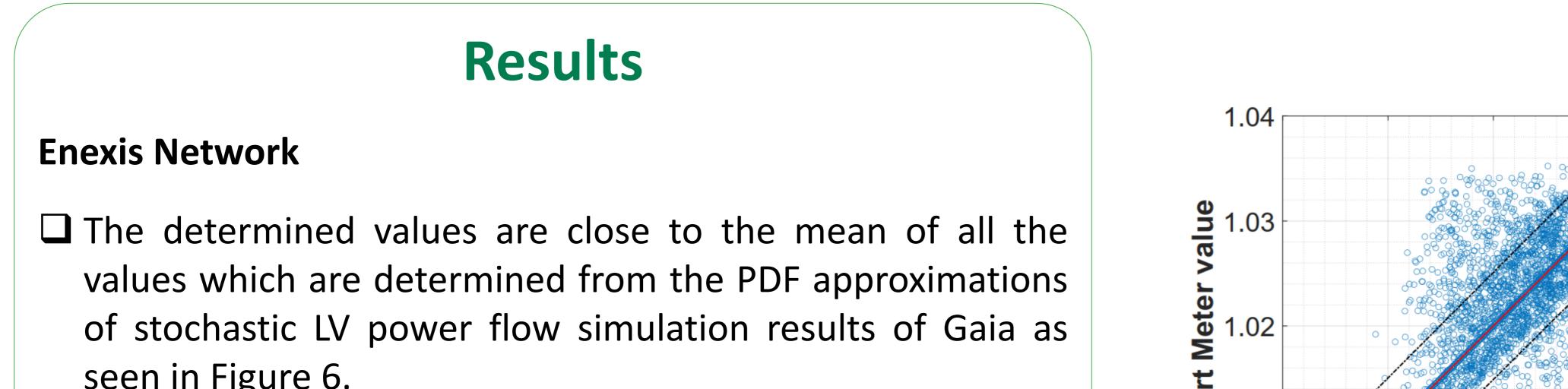
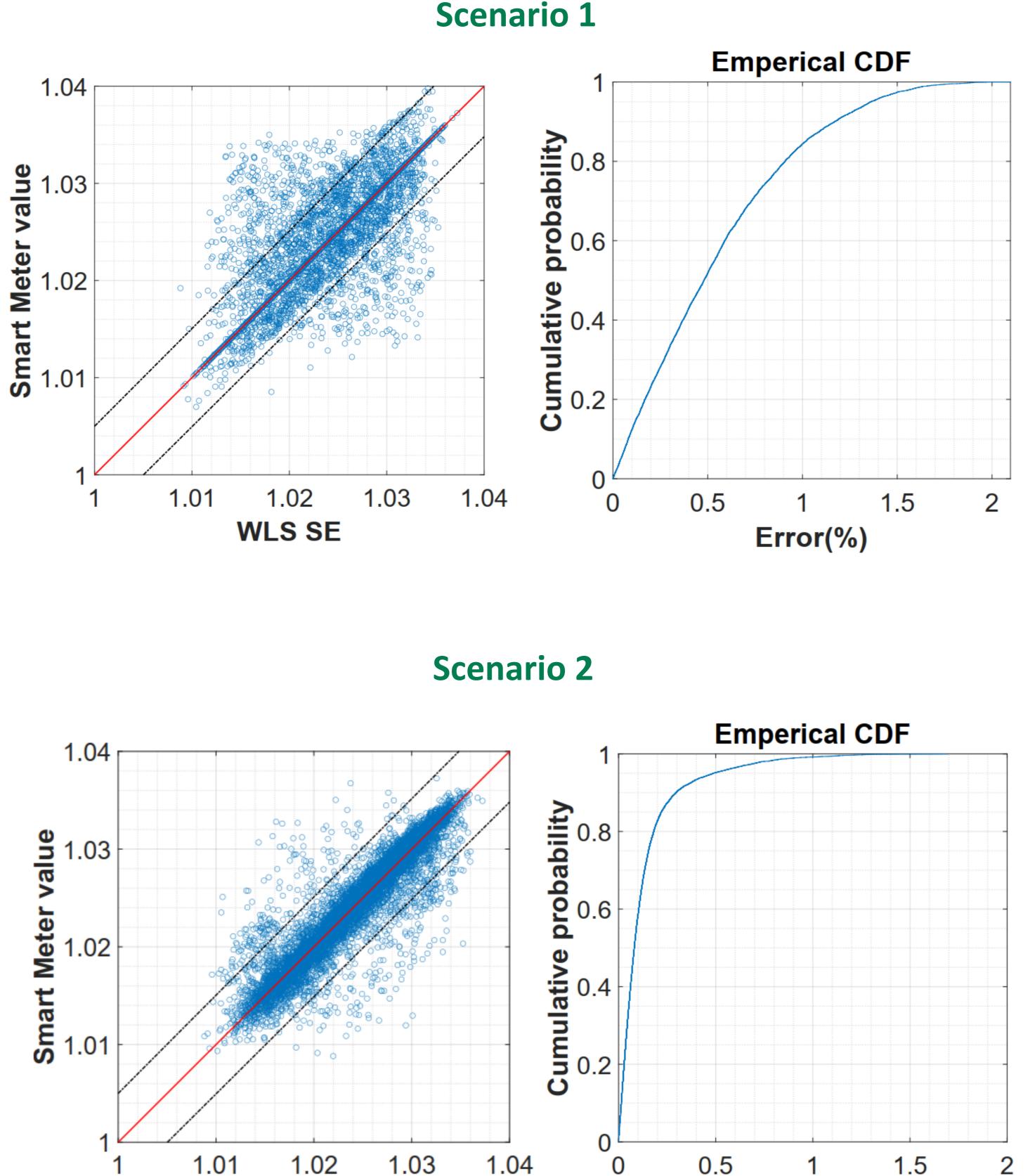


Figure 6. Validation of active power at peak time for Enexis BV network





#### **Stedin Network**

seen in Figure 6.

- Grigure 7 shows the difference in results for the two scenarios.
- It shows that the type of input taken affects the accuracy of determined states.

### **Conclusions/Discussions**

- The working of WLS algorithm on the networks using synthetic data and measuring devices data is verified.
- The effect of type of input measurement on state estimation precision has been evidently demonstrated.
- A detailed way of generating pseudo-measurements for the nodes with missing data is proposed, and a validation

#### with available data is done.

WLS SE

1.03

1.02

1.01

Error(%)

.5

Figure 7. SE result comparison scenarios for Stedin BV network

1.04

0

## References

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