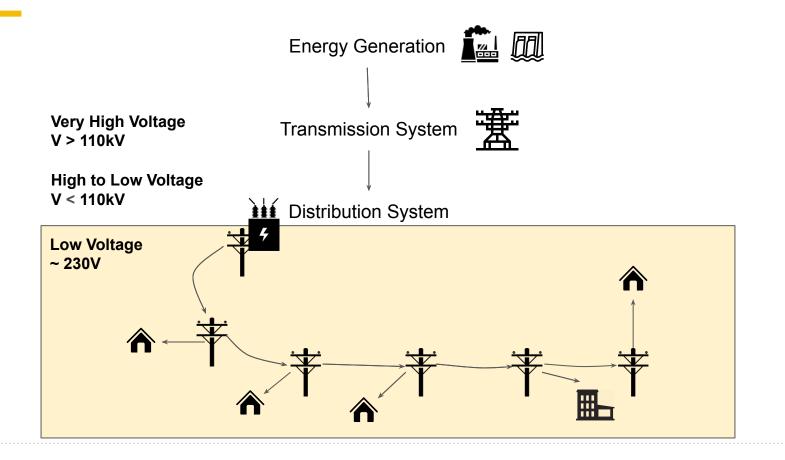
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Road to **ZEN**

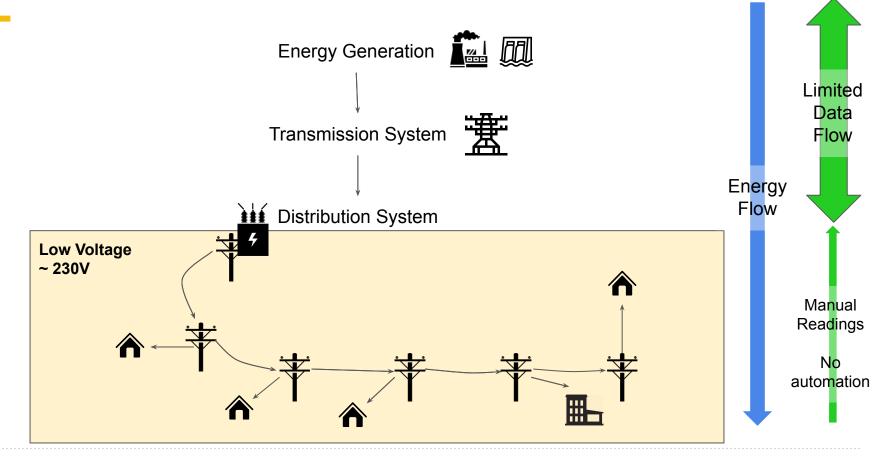
Zero Emission Neighbourhoods

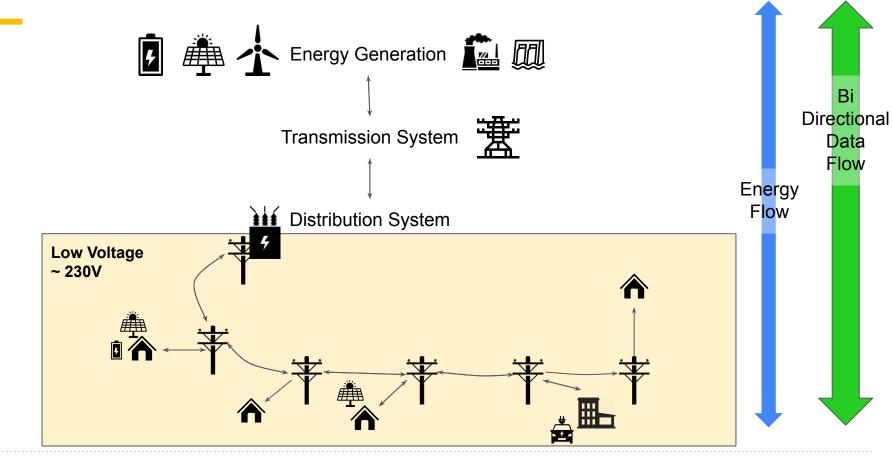
Bruno Galhardo (bgalhardo@eneida.io)

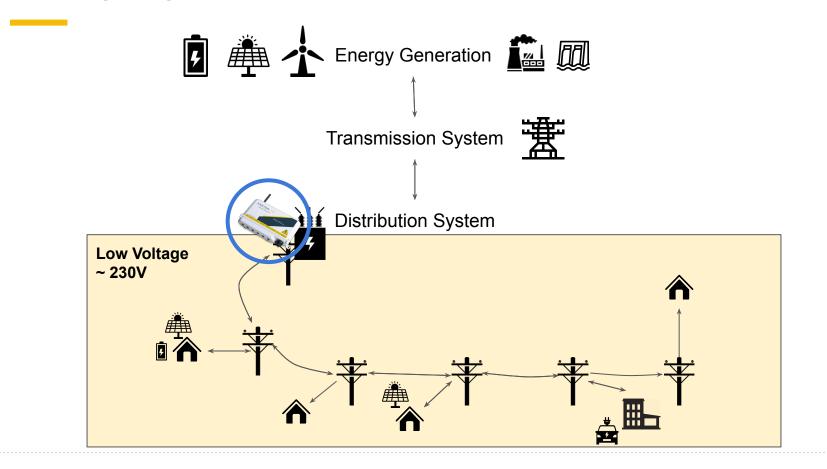
TRADITIONAL GRID



TRADITIONAL GRID





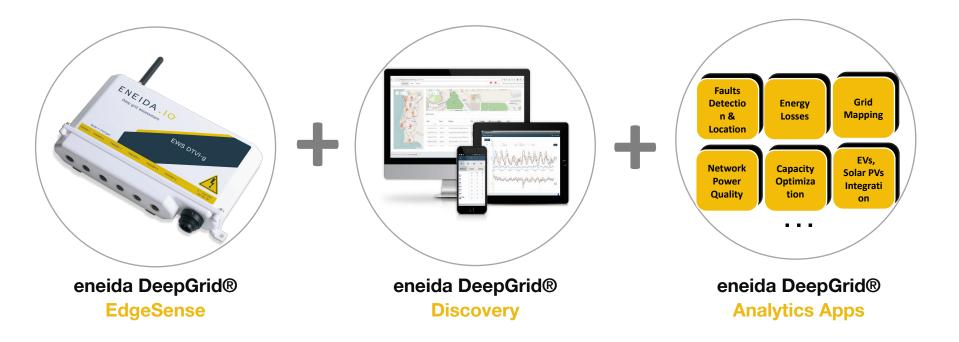


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GOAL: Monitor and optimize Low Voltage Grids



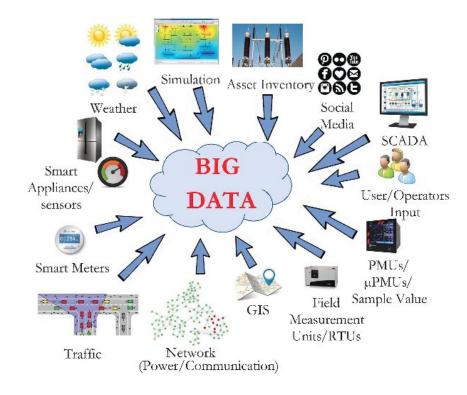
SOLUTION VISIBILITY, CONTROL E INTELLIGENCE OF THE LOW VOLTAGE GRID

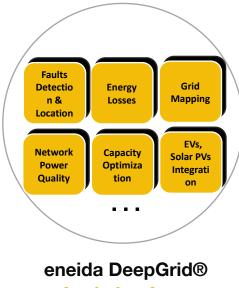


SOLUTION

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VISIBILITY, CONTROL E INTELLIGENCE OF THE LOW VOLTAGE GRID





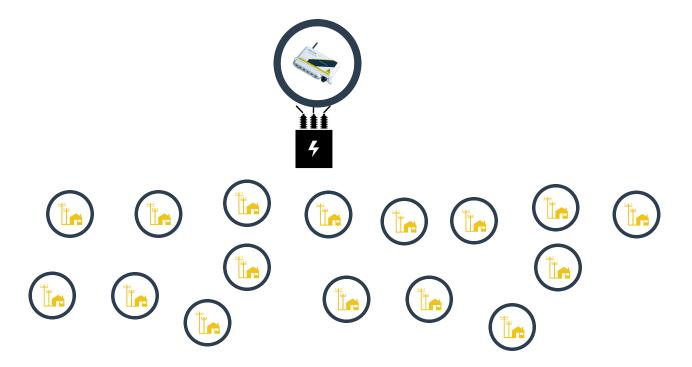
Analytics Apps

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Deep grid assessment

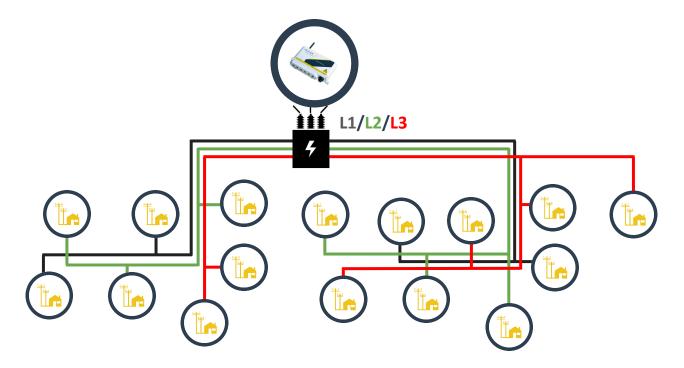
The Problem

Phase/Circuit at each location is unknown



The Problem

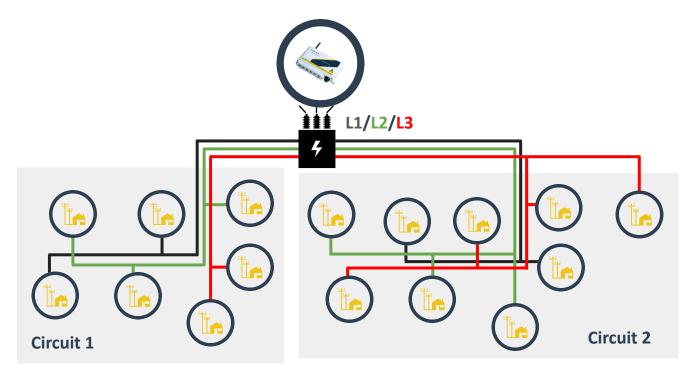
Phase/Circuit at each location is unknown



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The Problem

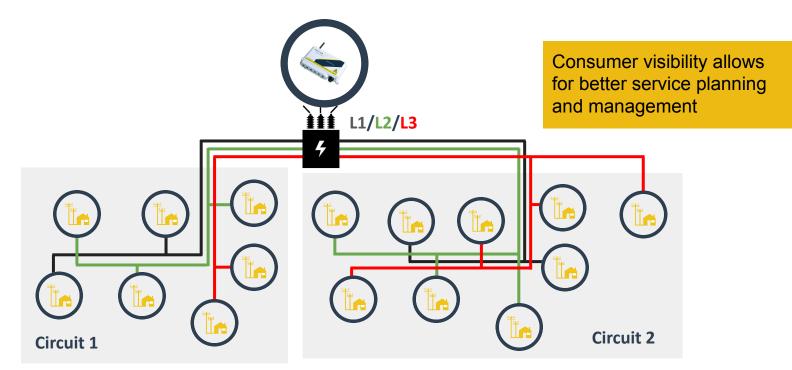
Phase/Circuit at each location is unknown



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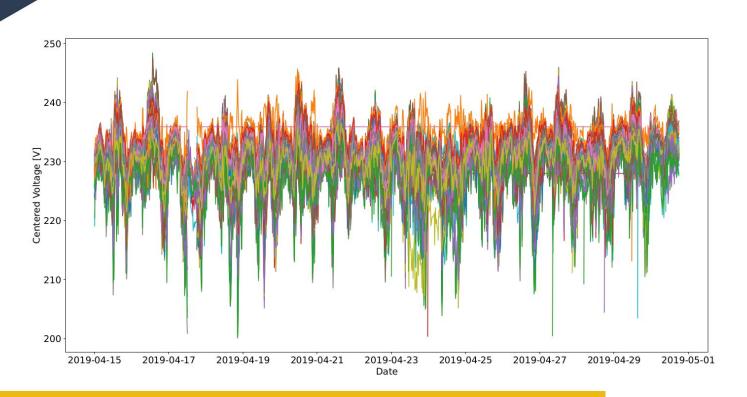
The Problem

Phase/Circuit at each location is unknown



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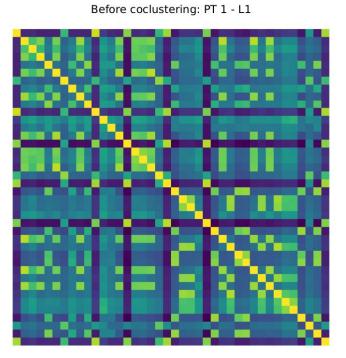
The Problem

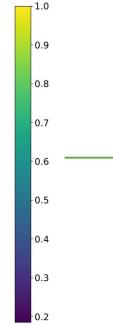


Household meters in the same neighborhood measure very similar Voltages

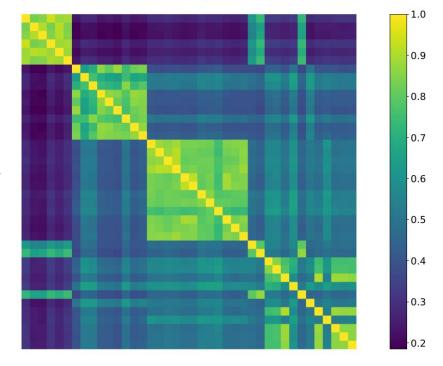
ENEIDA.IO

Co-Clustering: Circuit Attribution



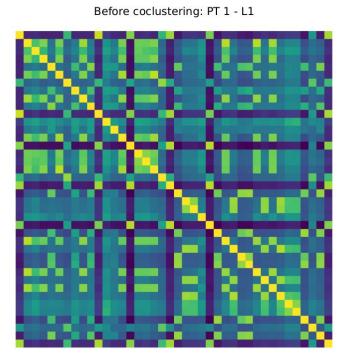


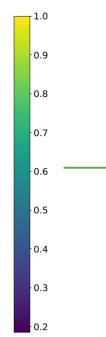


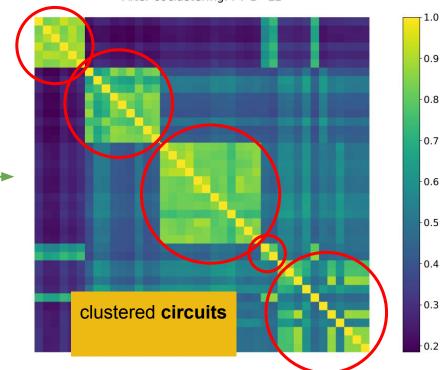


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Co-Clustering: Circuit Attribution



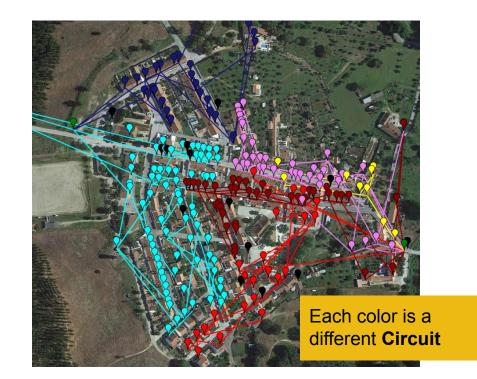




After coclustering: PT 1 - L1

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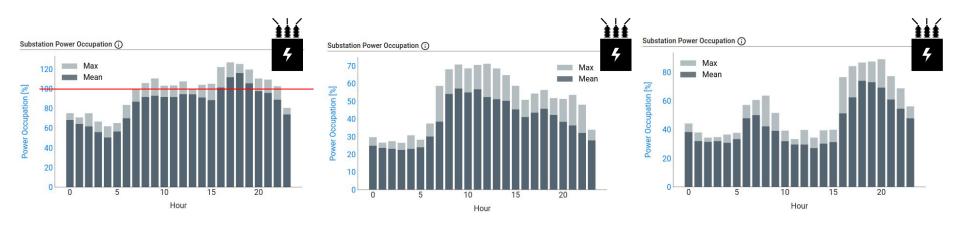
Circuit assignment



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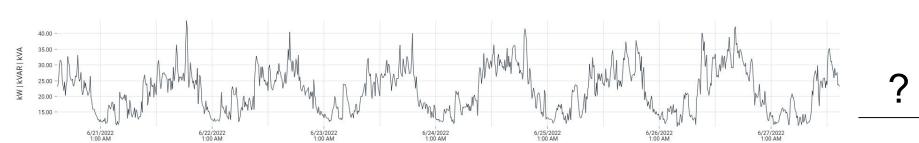
Deep grid assessment

The Problem



The operator wants to know when a given transformer will not meet the demand

The Problem



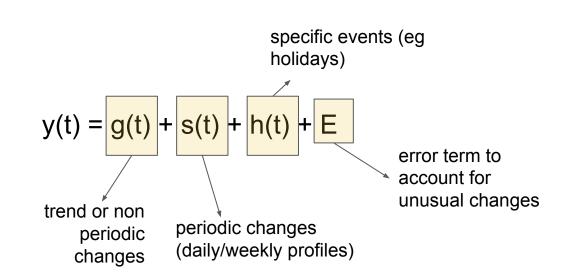
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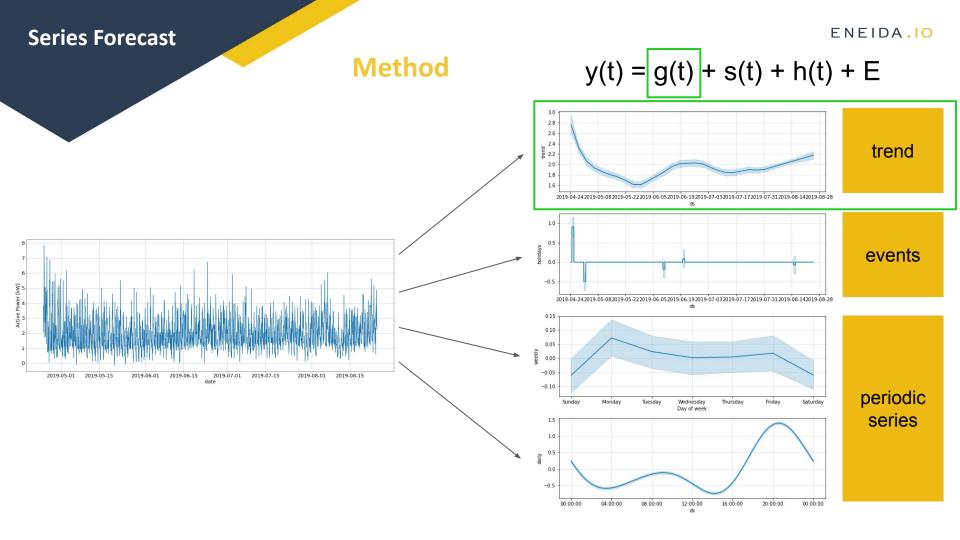
The Problem

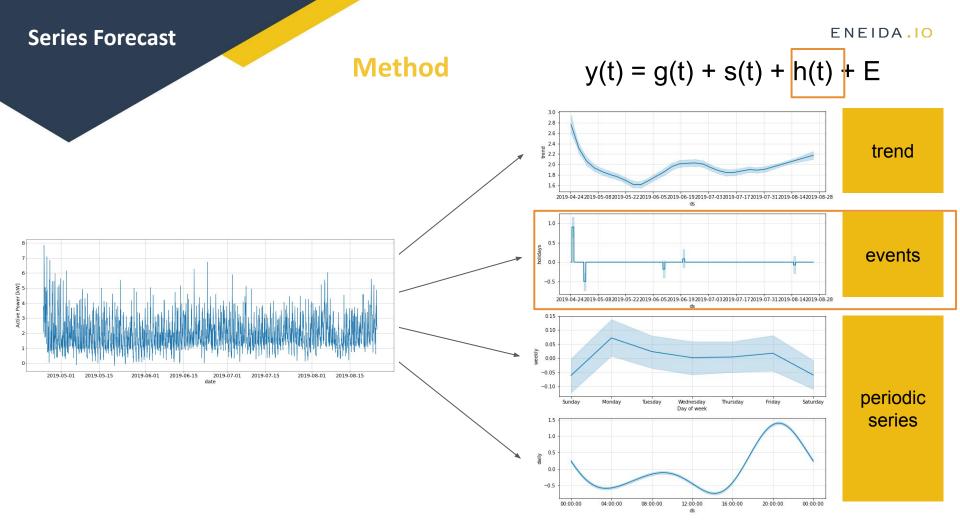


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Method











Power [kW]

2019-05-01

2019-05-15

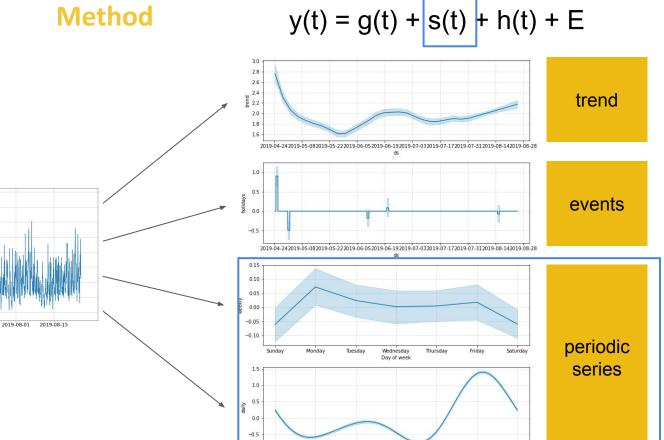
2019-06-01

2019-06-15

2019-07-01

date

2019-07-15



00:00:00

04:00:00

08:00:00

12:00:00

ds

16:00:00

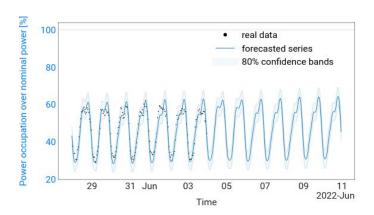
20:00:00

00:00:00

ENEIDA.IO

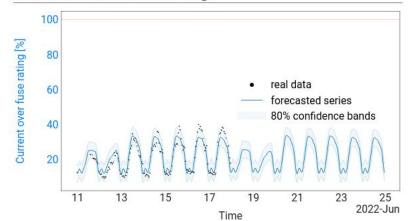
Result

Substation Power Occupation Forecast (i)



real data 100 . forecasted series Current over fuse rating [%] 80% confidence bands 80 60 40 20 0 25 2022-Jun 11 13 15 17 19 21 23 Time

L3 - Feeder/phase Overcurrent Forecasting ()

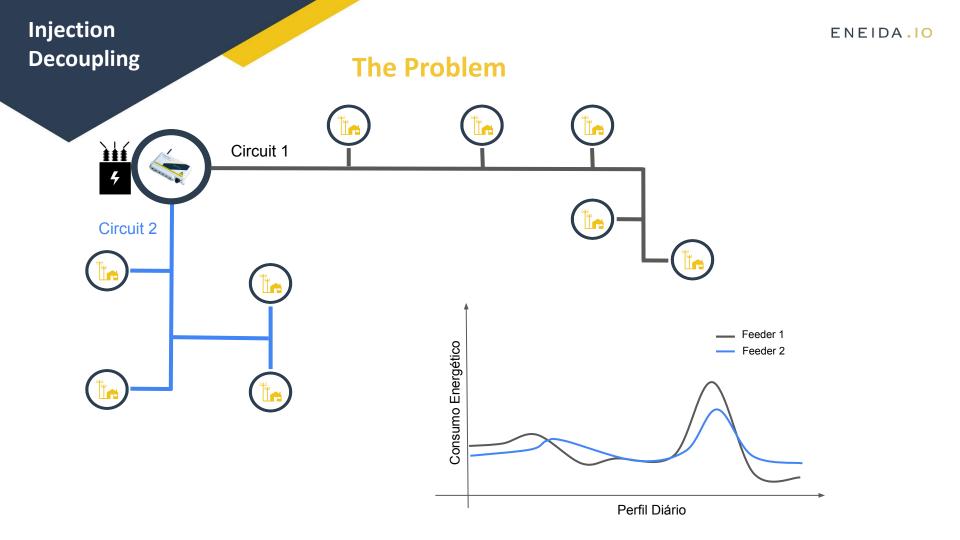


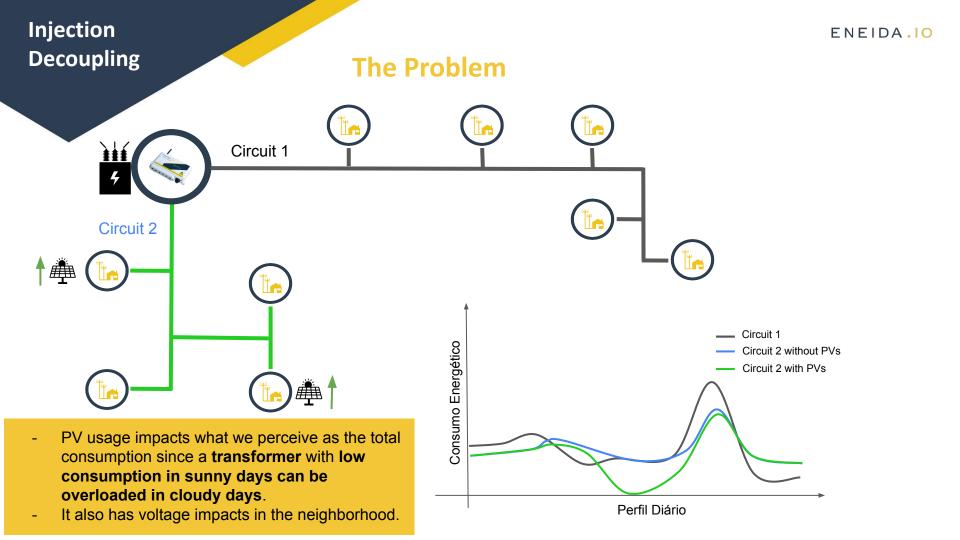
L3 - Feeder/phase Overcurrent Forecasting (i)

Injection Decoupling

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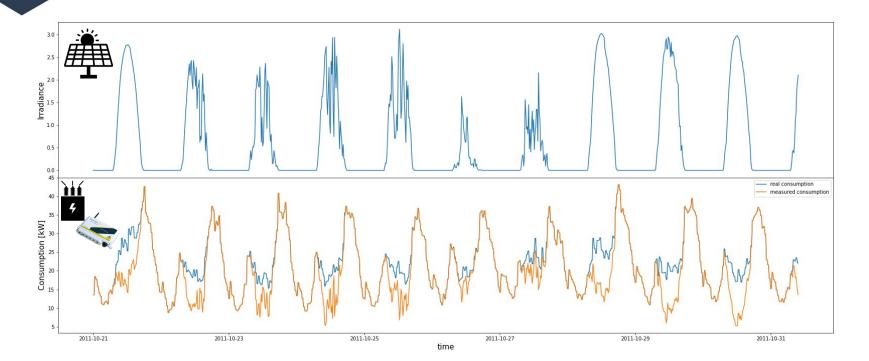
Deep grid assessment





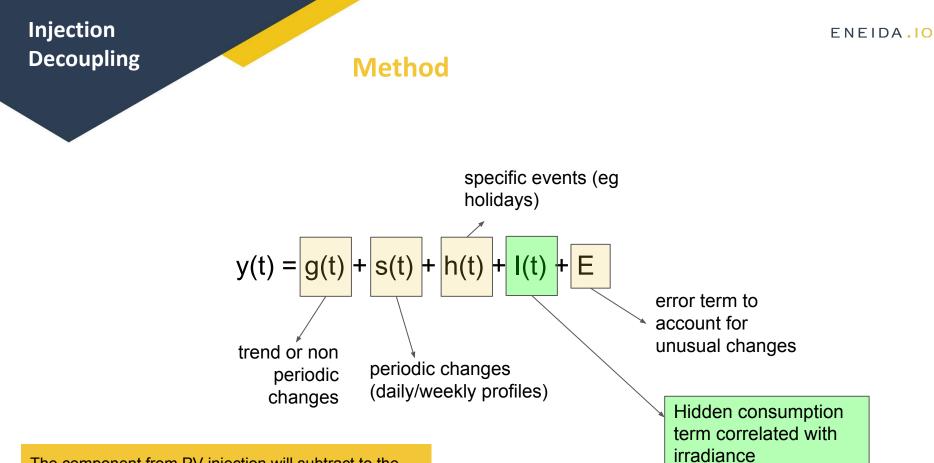
Injection Decoupling

The Problem



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We are interested in the real (bottom blue) consumption so that we can know the transformer peak demand

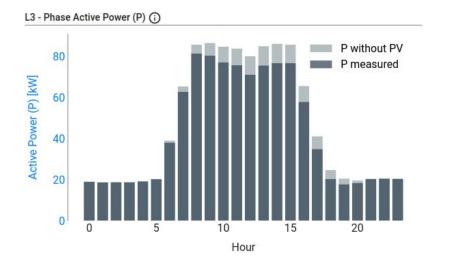


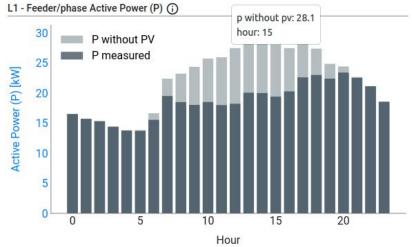
The component from PV injection will subtract to the consumption series (consumers will use own produced energy) and will be invisible for outside measurement

Injection Decoupling

Result

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Deep grid assessment

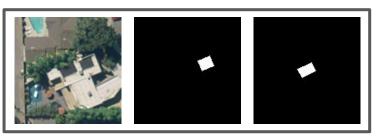
The Problem

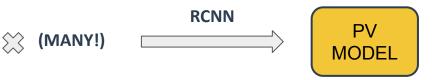
- instead of looking to the consumption series to find hidden consumption from injection we look at satellite imagery to find the PVs directly
- The amount of injection (hidden consumption) in the grid will be proportional to the PV area



Method

TRAINING:





PREDICTION:



User can **validate** identified PVs, **remove** wrong ids and **delineate** missing PVs.

Valid PV locations will be added to endpoint and images added to training database.

Manual Classification

Manual classification

delete current shape delete all shapes



Results

PV Finder

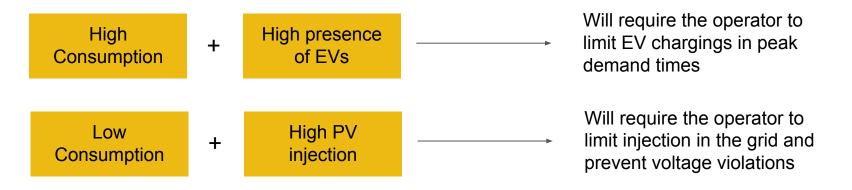
draw new shape delete current shape





CONCLUSION MORE LV GRID CHALLENGES

In the near future the outputs of these apps (and others) will be used to act directly on the grid. For example:



Other applications being developed inhouse:

- Fault detection and location
- Prediction of faults before they happen
- Energy theft detection and location
- EV charging detection and location

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Road to ZEN

Zero Emission Neighbourhoods

Bruno Galhardo bgalhardo@eneida.io

datascience@eneida.io

CERANAL
