# Team AquaFu (Aperio)

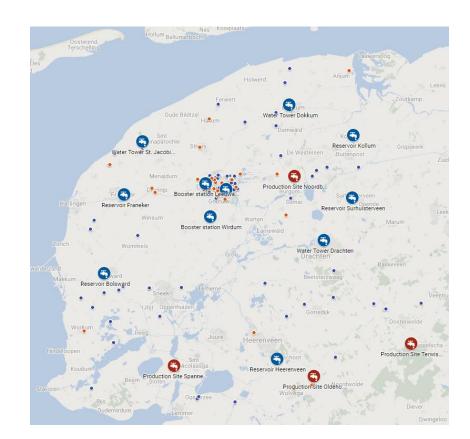
Data Science on Water Data

# Analysis of water quality problems - a challenge

Water Quality sensors,
 Flow sensors

 Sensors of different types on different pipes.

Unlabeled data.



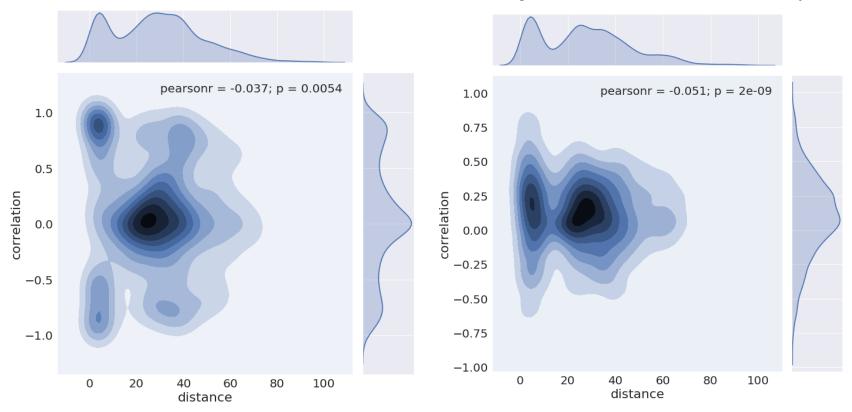
#### Tools

Python - Pandas, scikit-learn, jupyter notebook

Python - OSISOFT PI Web API library (written by Marcos Loeff)

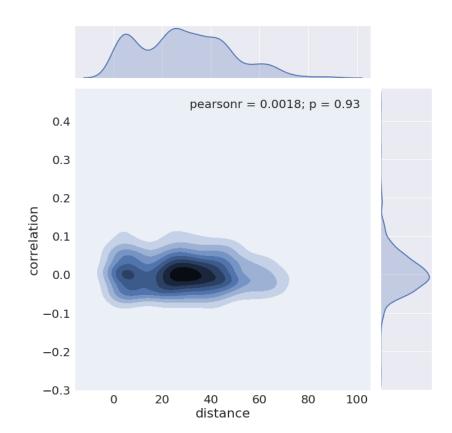
Elasticsearch + Kibana. Python code written to Export PI to Elasticsearch

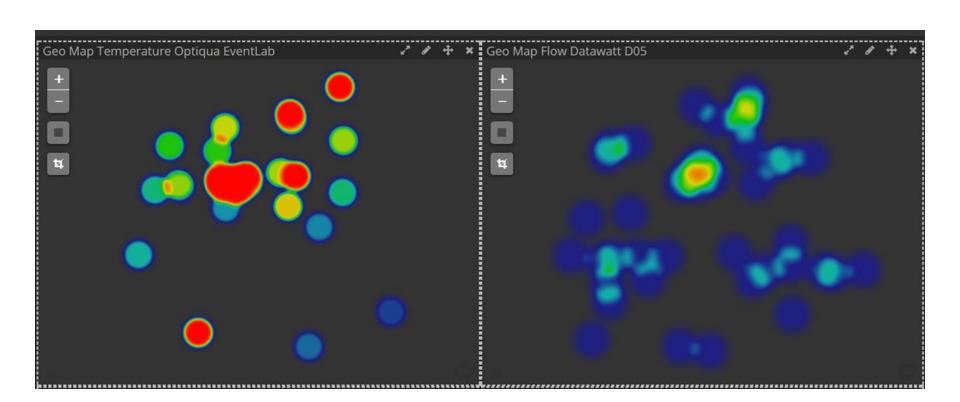
### Correlate the Flow Rate & Temperature vs Dist (km)



# Correlate the Water Quality (Optical Refraction)

Distance - km



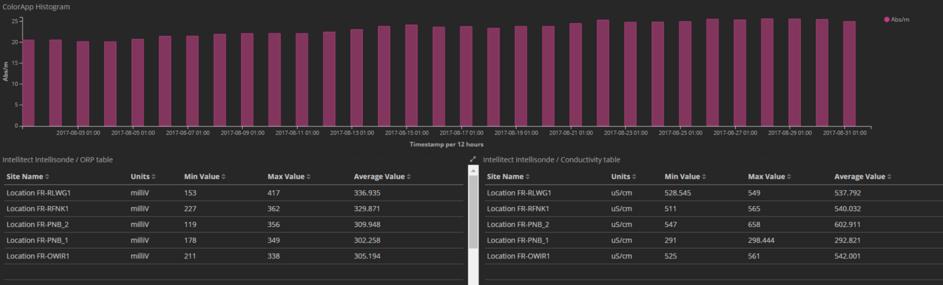


#### Timestamp per 12 hours





Timestamp per 12 hours



¥

Average Value \$

0.003

0.002

0.006

Intellitect I
Site Nam
Location
Optiqua Ex

Location FR-MOZSI

Location FR-MOZSI

Location FR-MOZSI

Location FR-MOZBG

Location FR-MOZBG

Location FR-MOZBG

Location FR-MOZB2

Location FR-MOZB2

Units 0

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NOR2

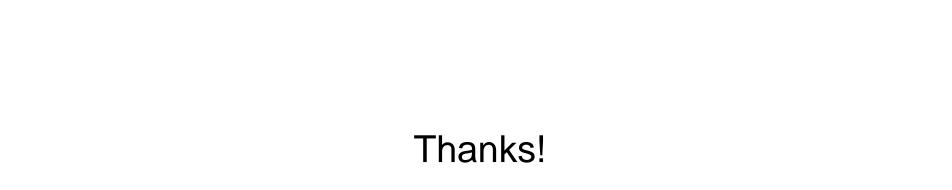
NOR3

NOR3

Max Value \$

0.086

0.066





# Life On







# A box







Our Thinking...









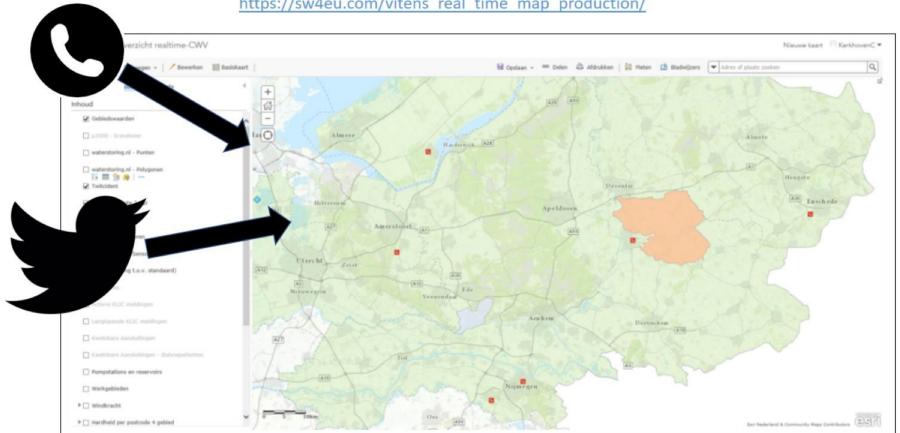


Water Environment for Smart Technology



#### Our Inspiration - Existing "Real-Time" Map

https://sw4eu.com/vitens real time map production/





#### Our Inspiration – Amazon "Dash" (Dash Replenishment Service)



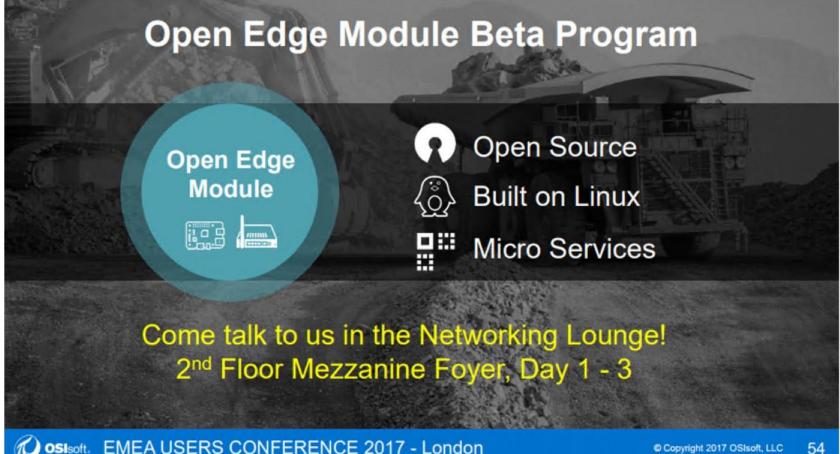


#### Our Inspiration – Amazon "Dash" (Dash Replenishment Service)



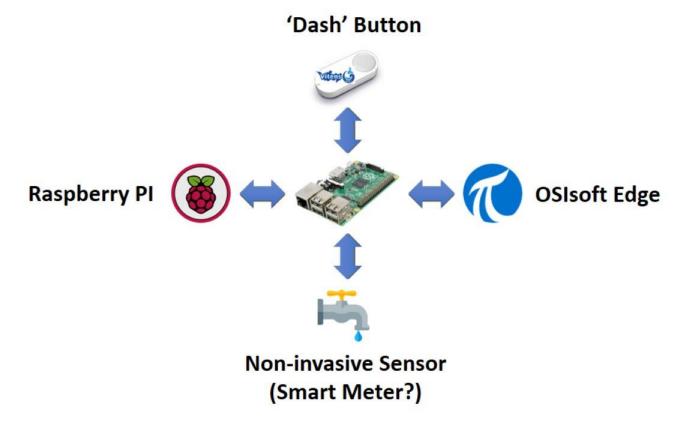
"Never forget a consumer's telemetry again"

















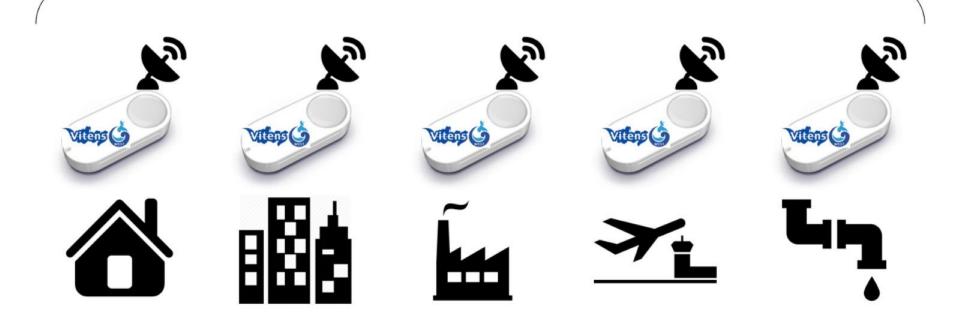






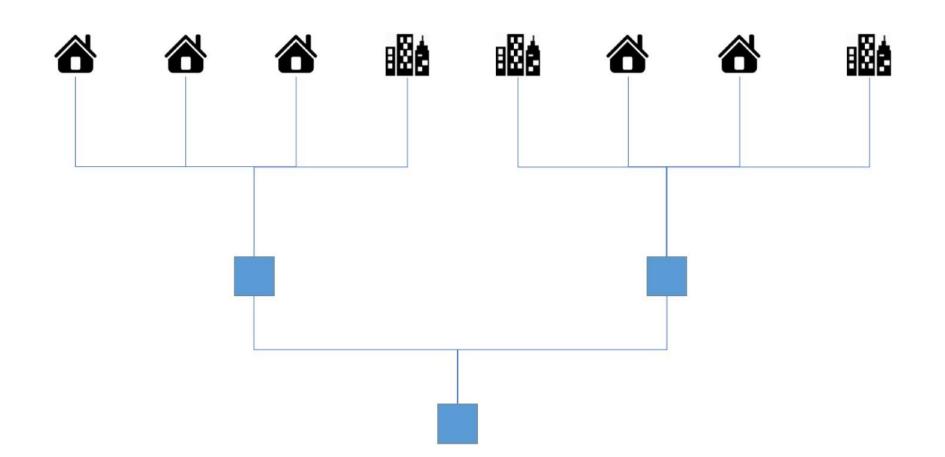






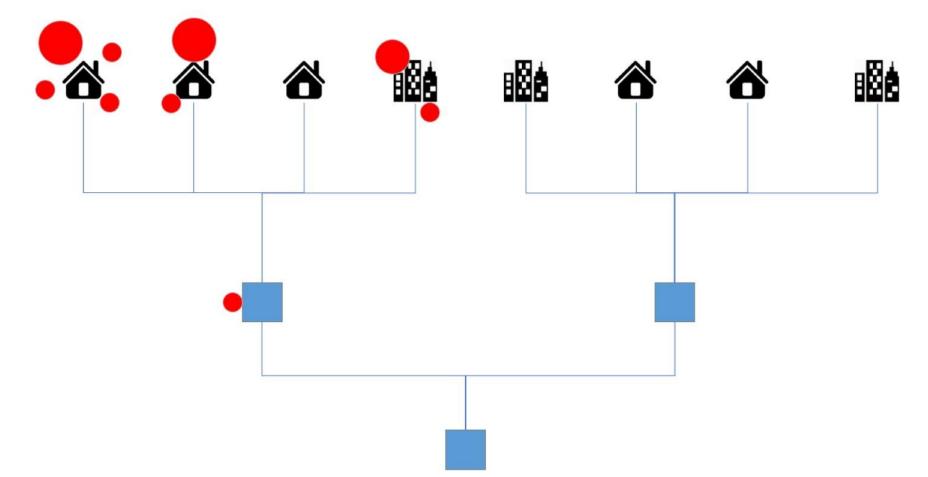






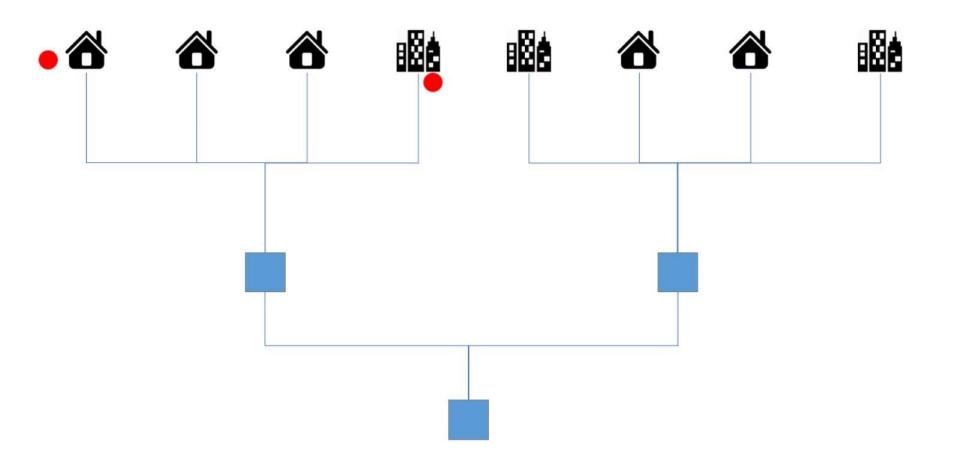


#### **Conceptual Overview - Expanded**



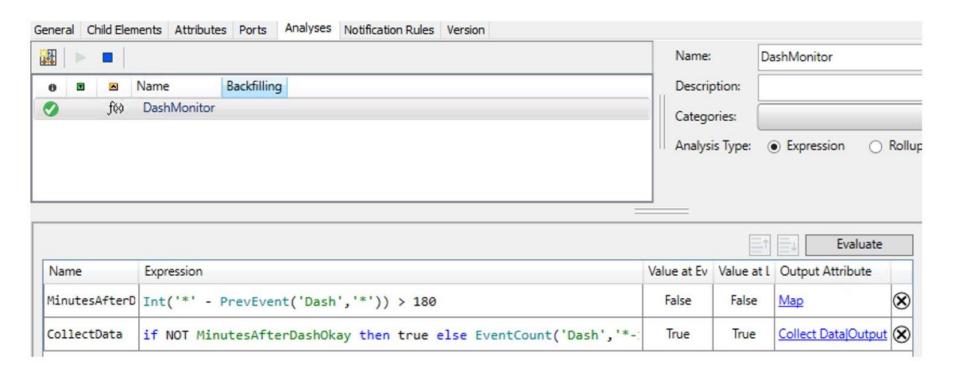






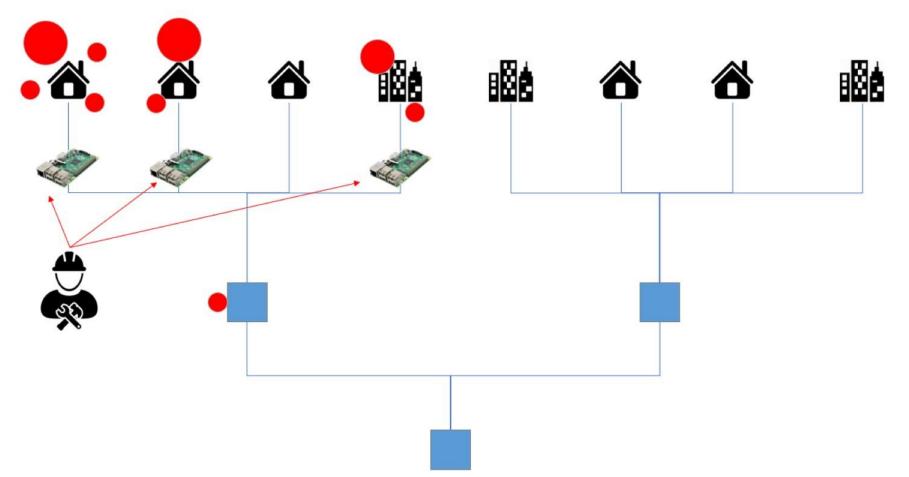


#### AF Analyses – Calculation output determines when to collect



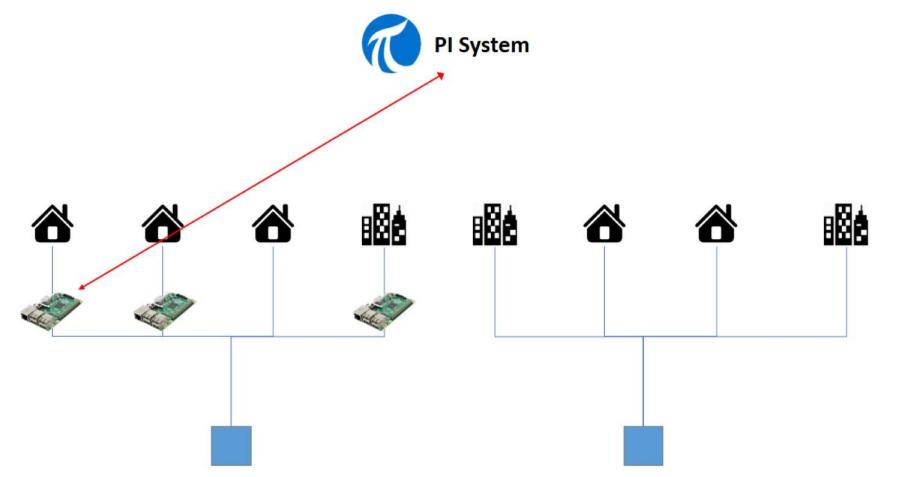








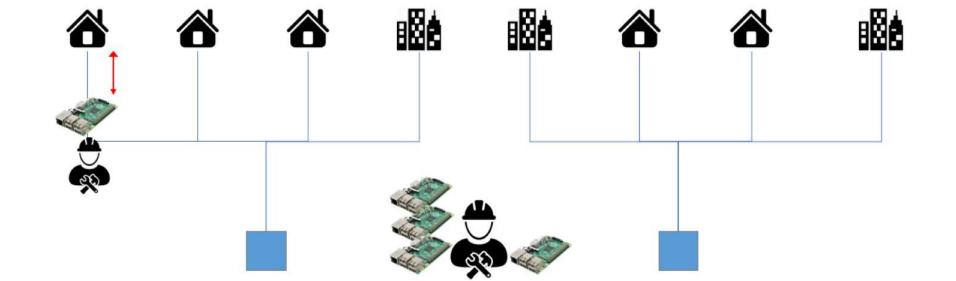






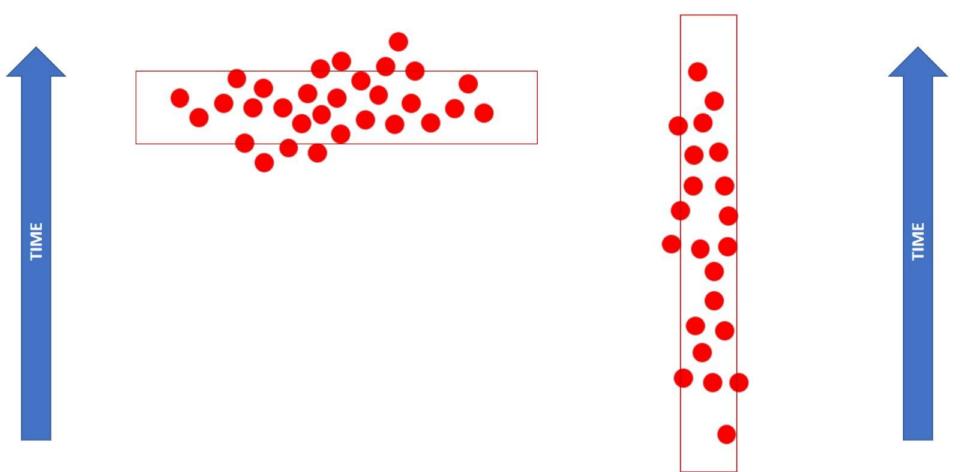






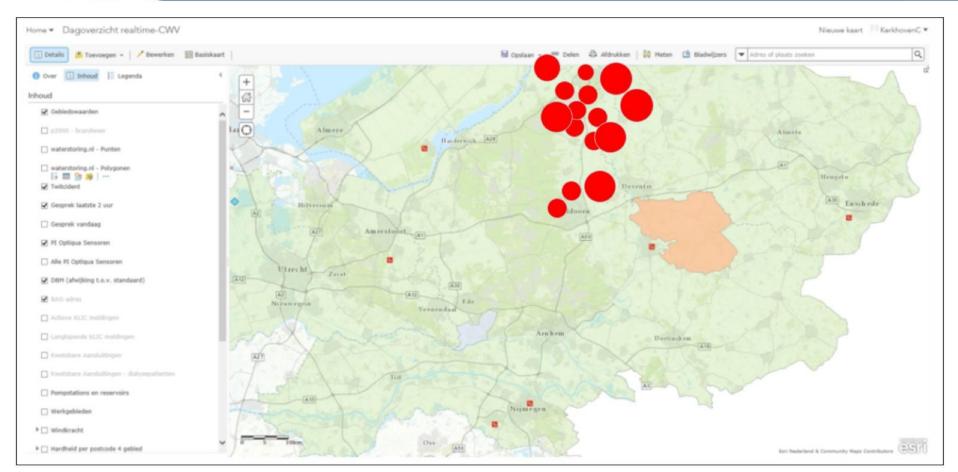


#### **Acute vs Chronic Dashes**



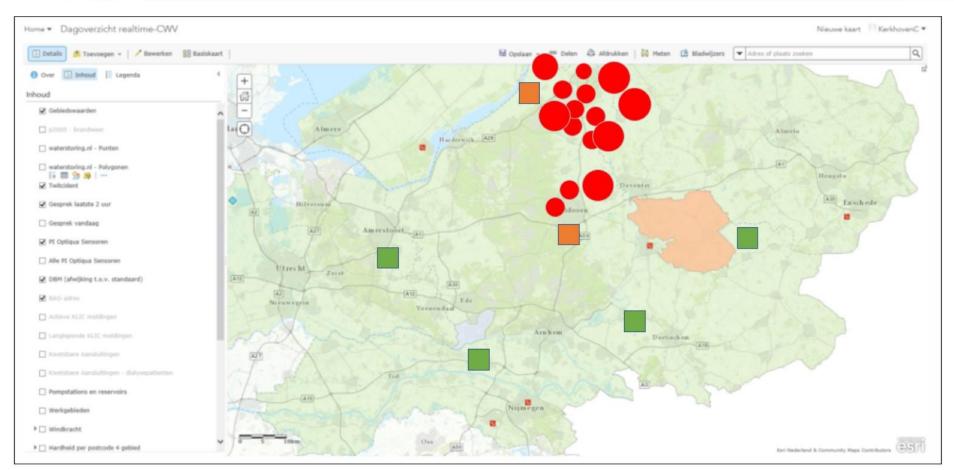


#### Overlay of "Dashes" on Water Network



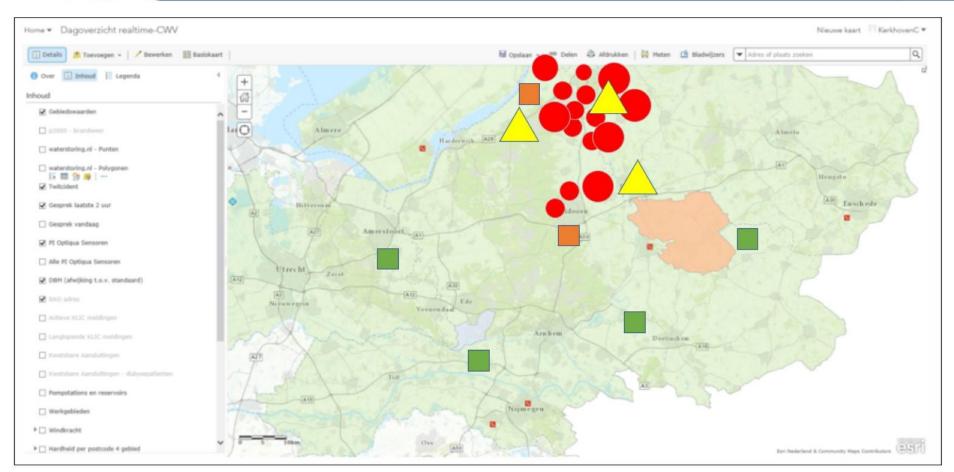


#### Overlay of "Dashes" on Water Network



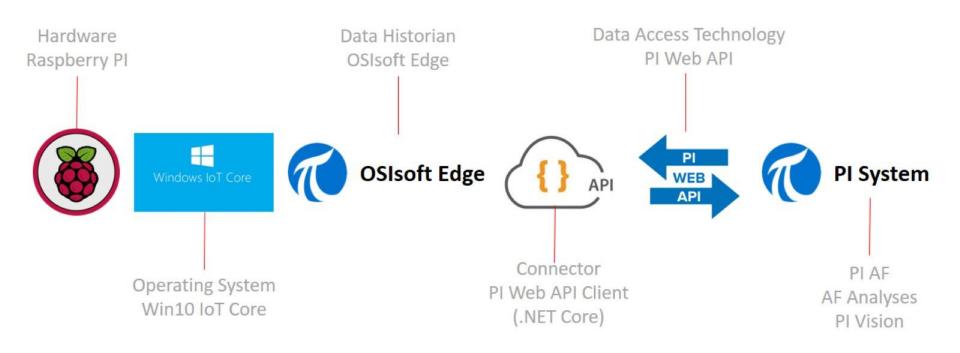


#### Overlay of "Dashes" on Water Network













**Live Demonstration** 



Near Real-Time feedback for customer as a "reward".



Behaviour based messaging.











Enrich existing data with a new source of consumer feedback and telemetry data.

Low cost and higher benefit to Vitens

- More rapid
- More consistent
- Less variable dataset

Modular and configurable based on where deployed in the water network.

Mobile solution – can be deployed multiple times, moved around the water network.



## Why Vitens should do this!

	Phone Call	Social Media	Dash Button
Customer Experience	Not easy. Time Consuming. Reluctance/inconvenient.	More attractive, easier than phone call.  Some generations may struggle with this approach.	Simple, quick, easy.  Convenient.  Feels more connected.  Physical presence.
Vitens	Staff for call centre. Flood or drought – in terms of calls. Variable/inconsistent feedback on issue.	Potential negative impact on Vitens – social trending of bad issues.  Semi-automated feedback processing.  Variable/inconsistent feedback on issue.	Instant feedback.  Automated processing.  Consistent telemetry data.

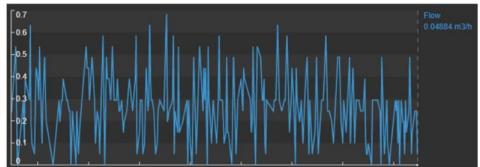






"You are currently in a queue, please hold...
You are currently in a queue, please hold...
You are currently in a queue, please hold...
You are currently in a queue, please hold..."











Convenient.

Instant.

Listened to.





Connected.



Engaged.

# **PyVision**

**Advanced Anomaly Detection for Water Quality Monitoring** 

## What

- Replace the monitor lizard
- Advanced Anomaly Detection at all locations
- Automatically create alarms & notifications



MONITOR LIZARD

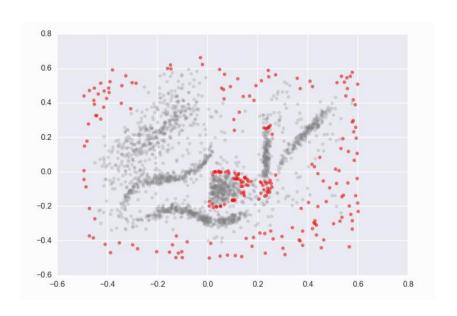
## Why

- Ensuring reliable service
- Enabling fastermaintenance response
- Reduce likelihood of false alarms by combining multiple sensors

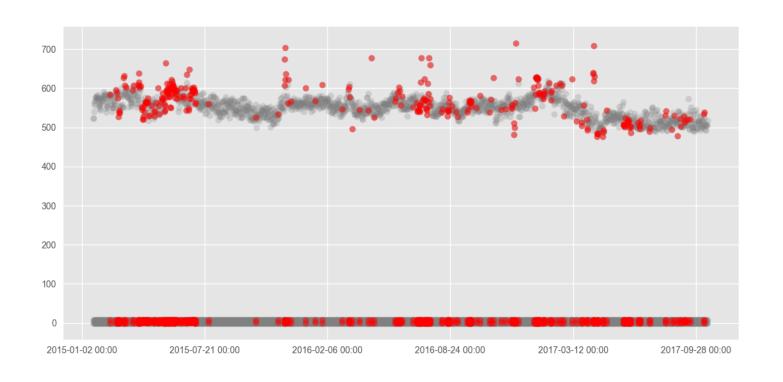


## How

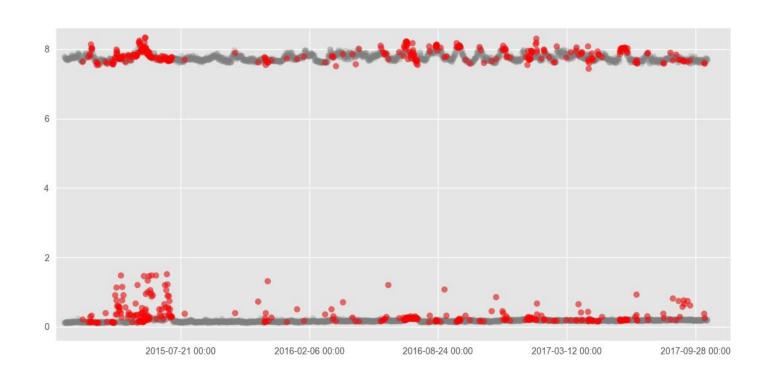
- HDBSCAN (hierarchical density-based spatial clustering w/ noise)
- Looking for spatial density-adaptive outliers in multiple dimensions



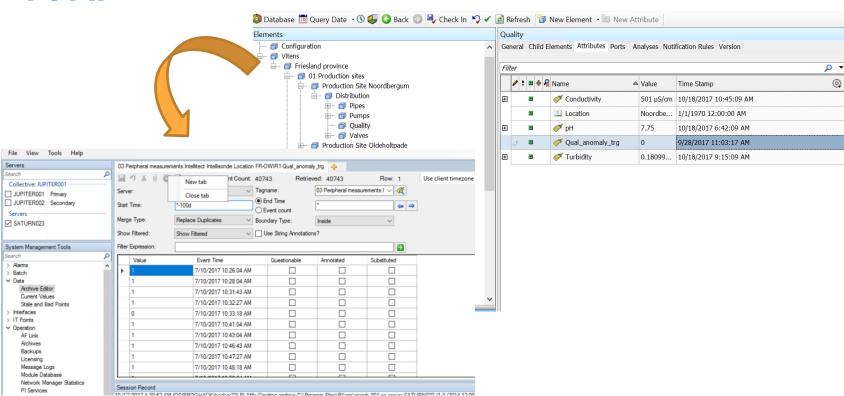
## **Production Site Noordbergum - conductivity**



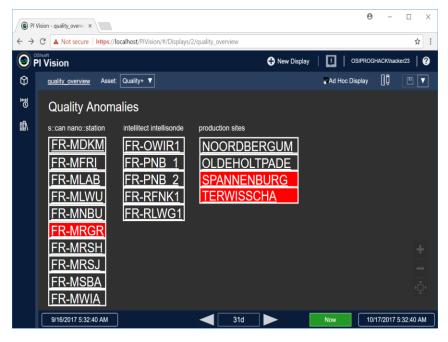
## **Production Site Noordbergum – pH & turbidity**

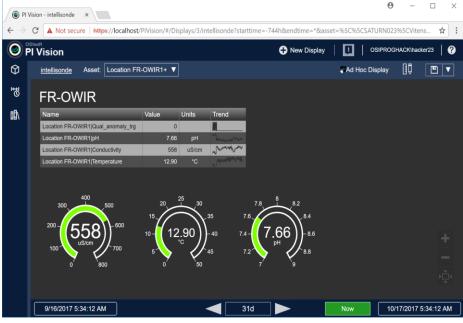


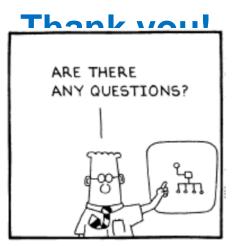
## PI AF



## **PI Vision Dashboard**

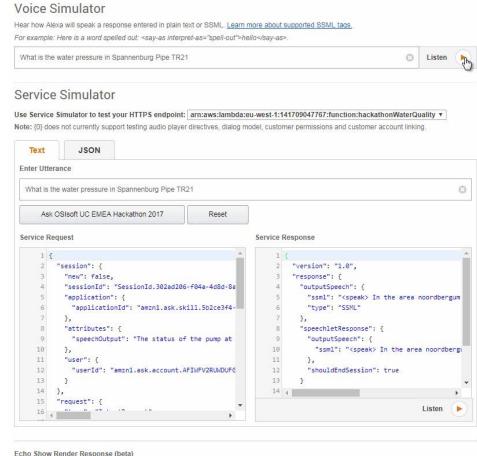






## **OSIsoft EMEA 2017 Hackathon event**





## **OSIsoft EMEA 2017 Hackathon event**



"Objects on the presentation might look older than it appear"









## Who are we?



**Marek**Data Plumber



**Jakub** Coder



Alexa "her"



**Przemyslaw** Coder



**Gregor** The other



## The challenge

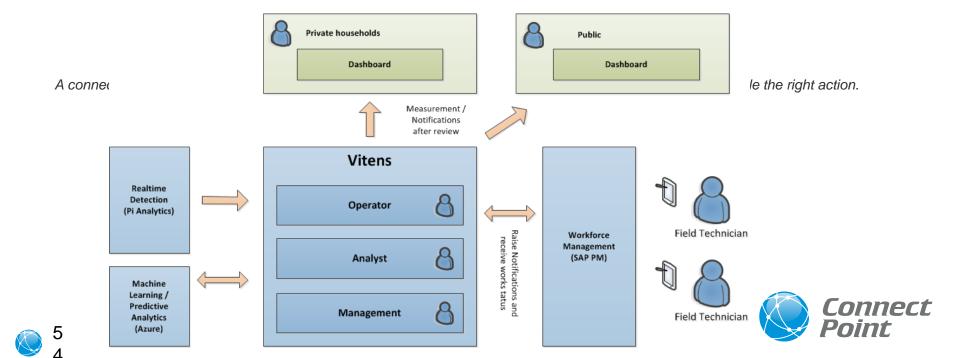






of thei.

*future* 











When the data indicates an issue that can cause a problem, it is assigned to the right work force and affected parties are notified if required.



Interventions are performed to resolve the issue.



Analytics in real time identify anomalies. Machine learning analytics predicts by correlating signals and past experience Connected system **enabling learning cycles** where gained knowledge can be apply to future issue to **detect earlier** or **resolve faster** 





Data is acquired and transmitted to a central location



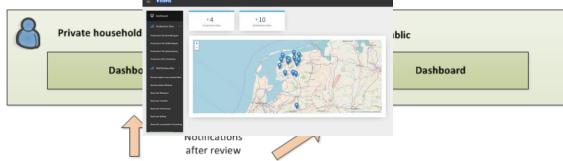
Solutions are recorded.
Outage and incidents are
prevented and private
households dissatisfaction
eliminated

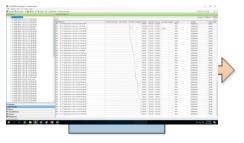






## What we did













## Used technologies

#### PI AssetFramework

-Analysis

-EventFrames

#### Webtechnologies

-PI Web API

-Realtime data (channels/websockets)

-Batch queries

-EventFrames (fetch and acknowledge)



#### **Data Analytics**

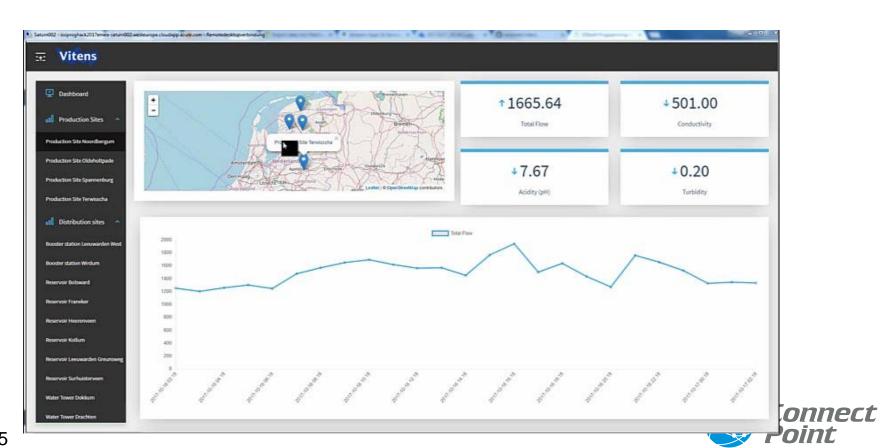
- Tableau Desktop
- Microsoft Azure Machine Learning Studio
- SQL
- PI OLEDB

#### **Extended Userexperience**

- Alexa
  - Alexa Skills Kit
  - AWS Lambda Function (Node.js)

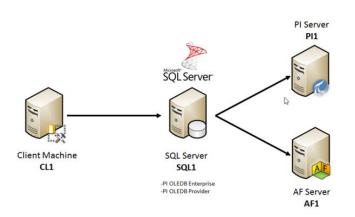


## Dashboard

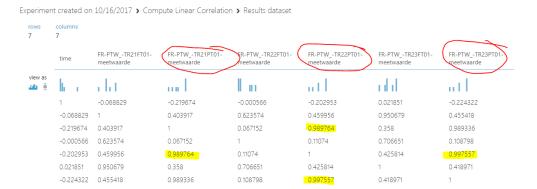




## Data Analytics (Cloud)



Using data extracts via PI OLEDB. Visualizing the data with Tableau or experimenting with Microsoft Azure Machine Learning Studio by apply existing algorithms.

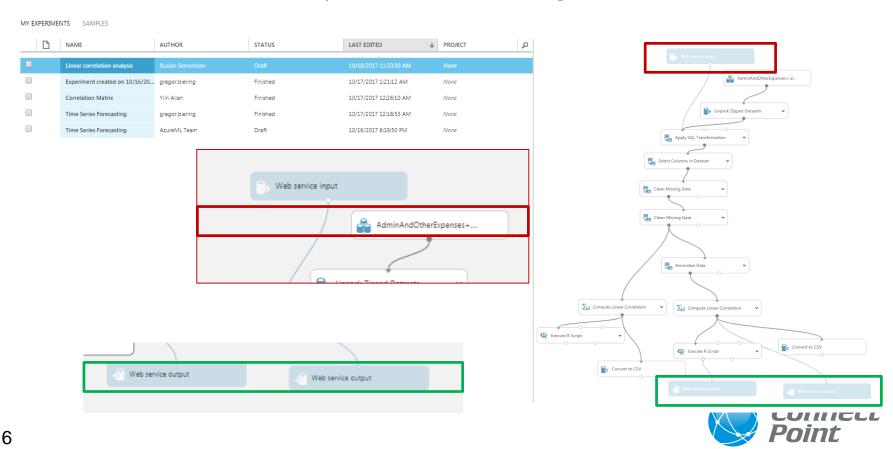




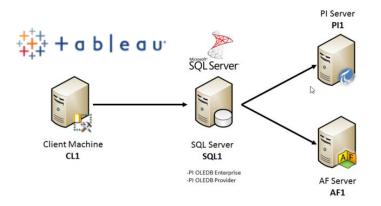




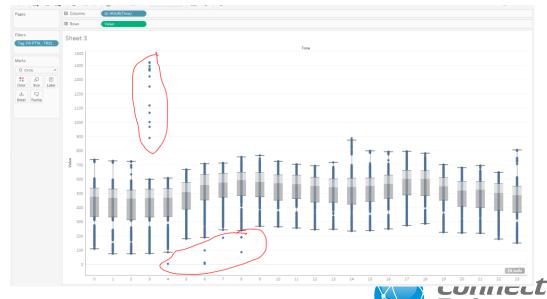
## Data Analytics (Cloud-Integration)



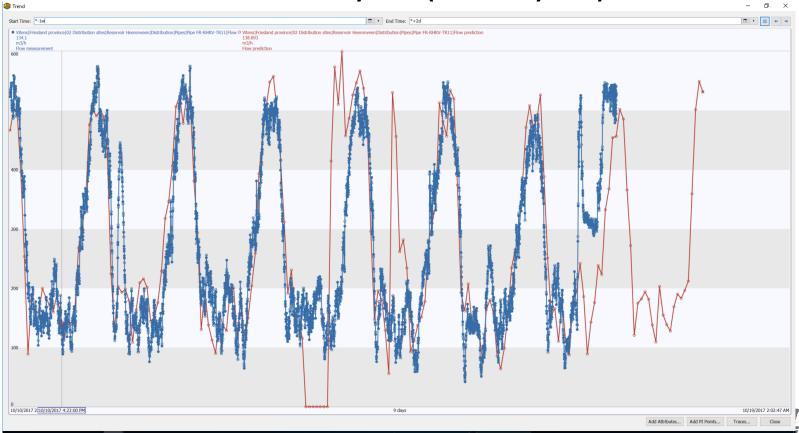
## Data Analytics (Tableau)



Using data extracts via PI OLEDB. Visualizing the data with Tableau or experimenting with Microsoft Azure Machine Learning Studio by apply existing algorithms.



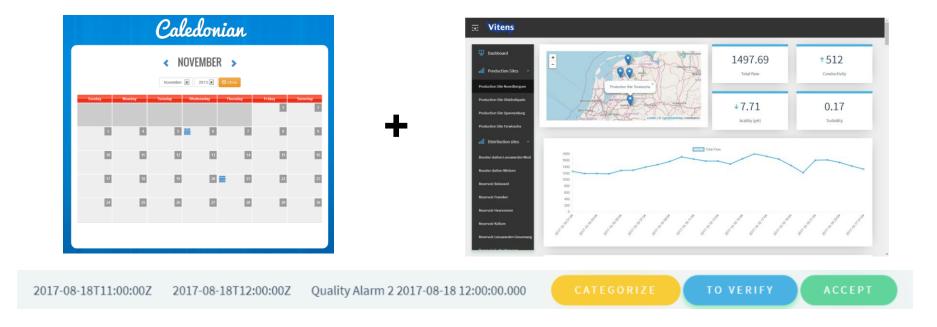
Data Analytics (PI Analytics)







### Consider all kind of data



By categorizing issues / anomalies not just to operative events but public events too (e.g. football, vacation) is it possible to predict data anomalies by just checking the calendar?

# How this will help Vitens

- •A connected system provides a better overview and enables a seamless workflow by having systems integrated
- Using existing powerfull tools to prevent implementation of things that already exists (e.g.
   Cortana Intelligence Gallery)
- Enabling data exploration by using Tableau for vizualisation
- •Interaction with the system in new way (Alexa)
- Saving costs predicting events faster and assigning field forces at the right time

# **Thanks**

## E.ON Climate & Renewables -

E.ON is an international, privately-owned energy supplier focused on 3 areas

Energy networks

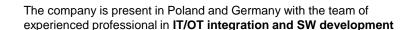
Customer solutions

It operates in over 30 countries and serves over 33 million customers

The Company employs above 50 000 people world-wide and appually achieves revenues above 100 billions Furn

F.ON Climate & Renewables (FC&R), headquartered in Essen, Germany, is responsible for F.ON's industrial-scale renewable energy activities





ConnectPoint started co-operation with E.ON Climate & Renewables in 2013 and since that time the company conducted above 20 projects in areas like

Custom Software development

Connect Point is a Software vendor and System Integrator specialized in Connect Point is a Software vendor and System Integrator specialized in dust vendor specialized in dust

- System Integration in IT/OT area
- Business Intelligence applications development (e.g. OSIsoft PI)



# Find out more.

If you would like to find out more about how ConnectPoint can help transform your utility service with intelligent network capability, get in touch with

