

Off-line Data Validation for Water Network Modeling Studies

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1.- Motivation & Objective

Motivation

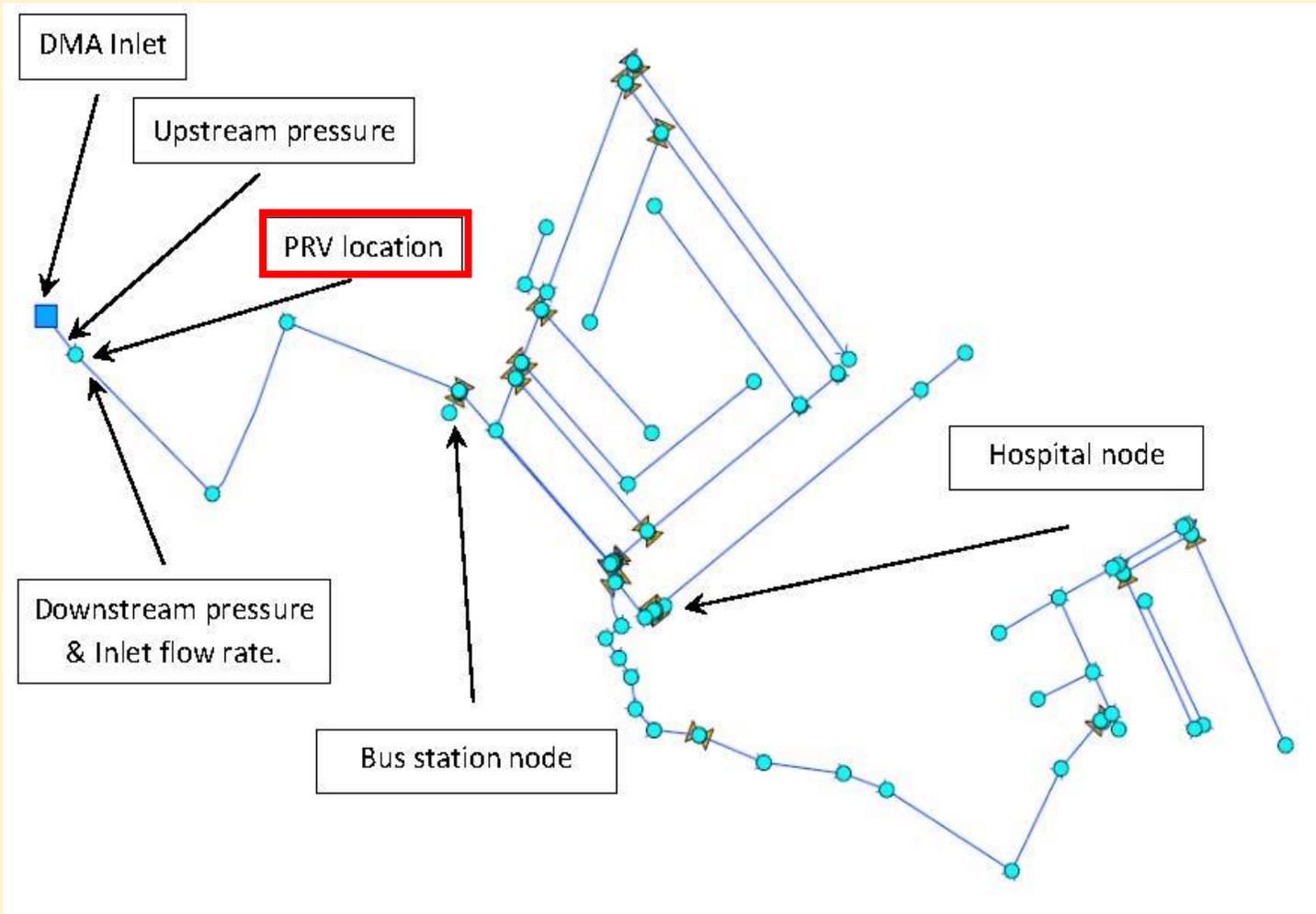
- Water network (WN) operating studies are significantly affected by the real data quality.
- If raw data are not validated before they are used, the resulting studies and models could not be representative of the real behavior of the WN.

Objective

- The application of an off-line semi- automatic classifier that separates data of nominal & abnormal events into WNs.
- A simplified procedure to validate raw data of WNs by using machine learning techniques.



2.- Case study: DMA El Charro



Characteristics	Quantity
Pipelines:	75
Nodes	90
Supply reservoir capacity:	1000 (m ³)
Customers:	2000
Average consumption per year:	3 (lps)
Installed Sensors:	1 upstream pressure transducer (kg/cm ²) 1 downstream pressure transducer (kg/cm ²) 1 inlet flowmeter (lps)
<u>Valves:</u>	1 pressure reducing valve [PRV]



2.- Case study: DMA El Charro

- Web platform & monitoring station



Detalles

Dispositivo

Estacion del Charro ▾

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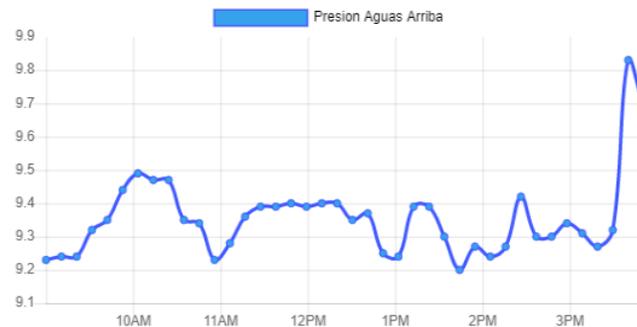
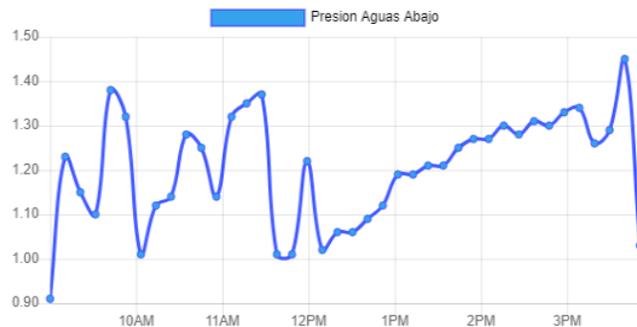
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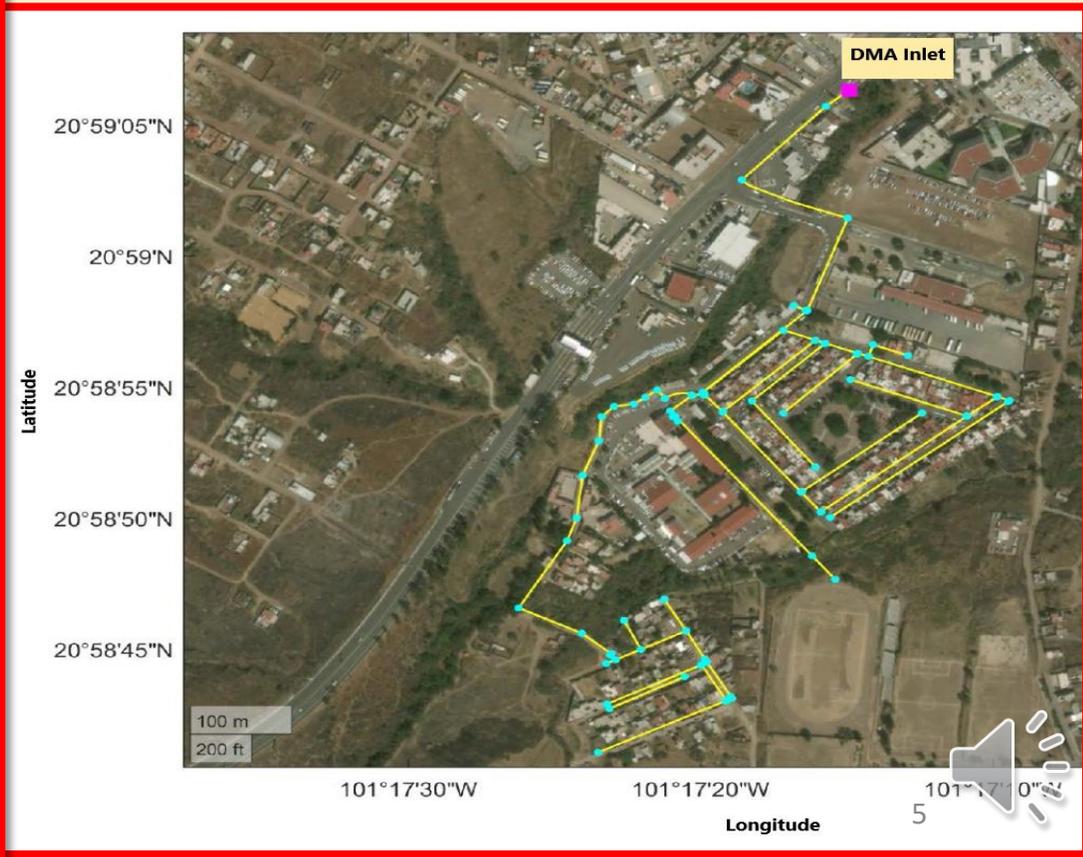
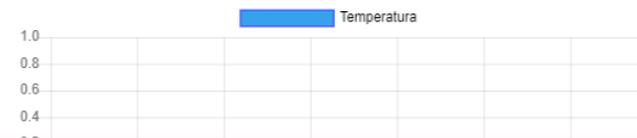


Aplicar

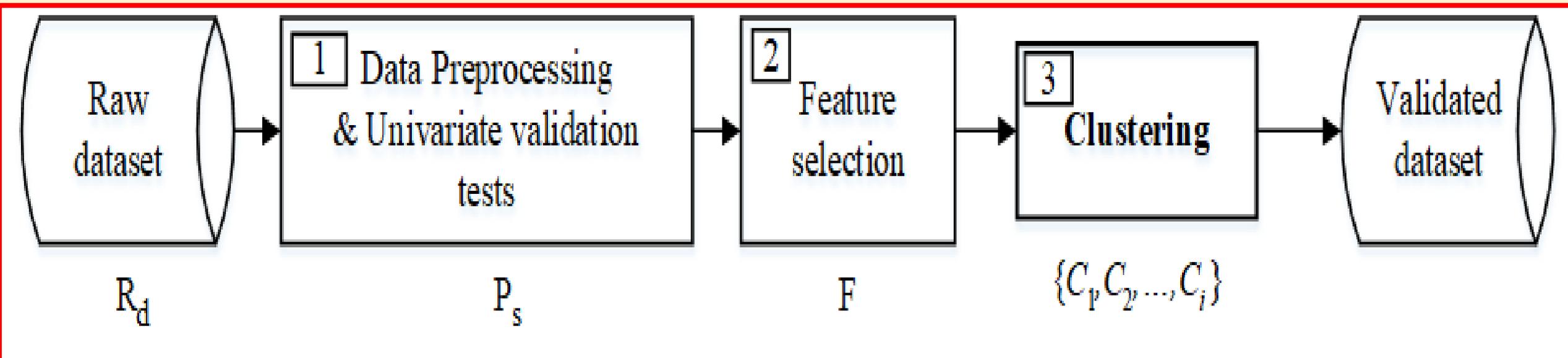
Parámetros Cuantitativos



Parámetros Cualitativos

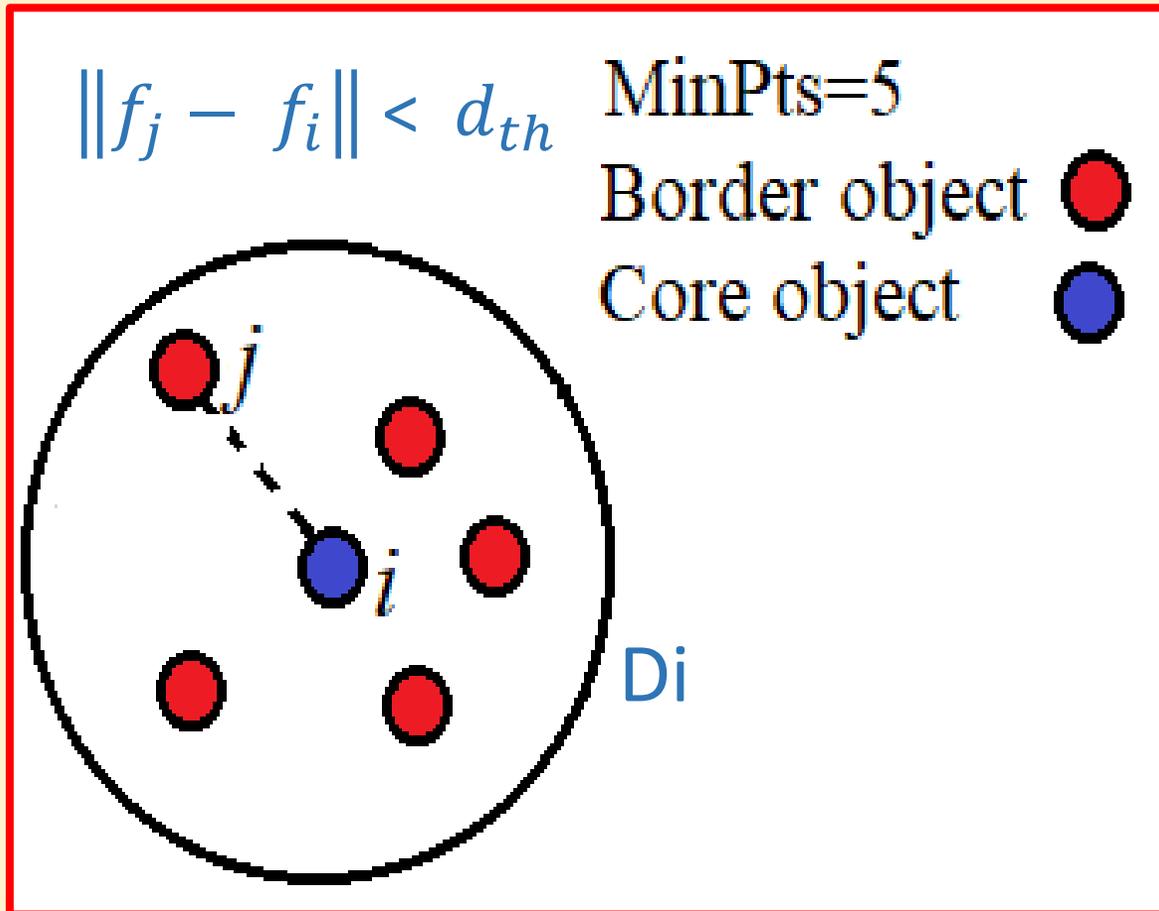


3.- Off-line Semi Automatic Validation Scheme

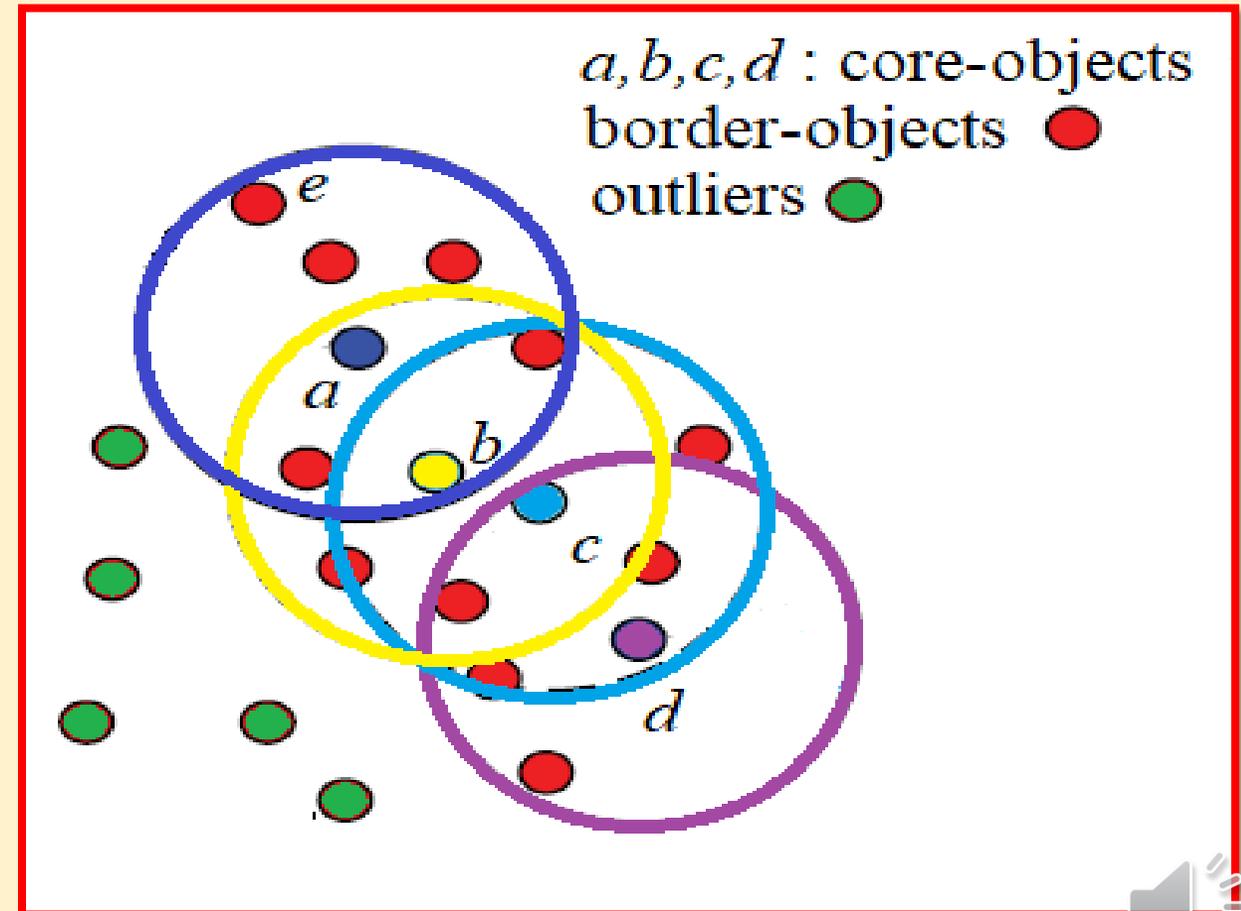


4.-Density-Based Spatial Clustering (DBSCAN)

- Object with its Neighborhood



- Density-based Cluster and outliers



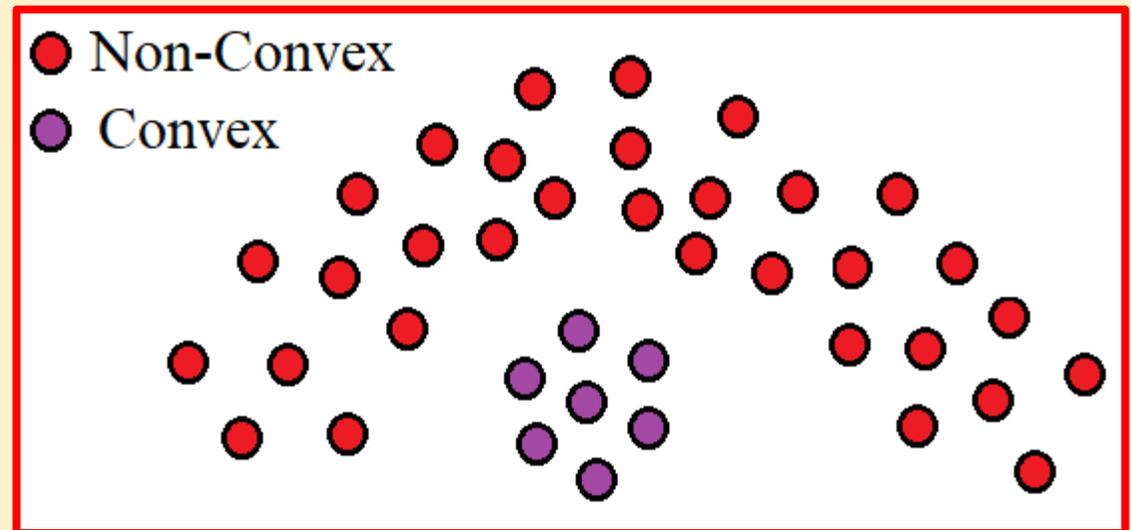
4. DBSCAN: Algorithm and properties

Algorithm

```
Data:  $F$ ,  $MinPts$ ,  $d_{th}$ ,  $C$ : set of clusters,  $No$ : set of noise objects,  $i$ : number of clusters
Label all objects as not classified,  $C = \emptyset$ ,  $No = \emptyset$ ,  $i = 0$ ;
for  $f_j \in F$  do
  if  $f_j$  is not classified then
     $DR_j = DensReach(f_j)$ 
    if  $|DR_j| > 1$  then
      Form a new cluster with all density-reachable objects
      Label cluster' objects as classified
       $C_i = DR_j$ ,  $C = \{C, C_i\}$ ,  $i = i + 1$ 
    if  $f_j$  is not a border-object then
       $No = No \cup f_j$ 
  Label  $f_j$  as classified
end
```

Properties

- Clustering of objects with non-convex shapes

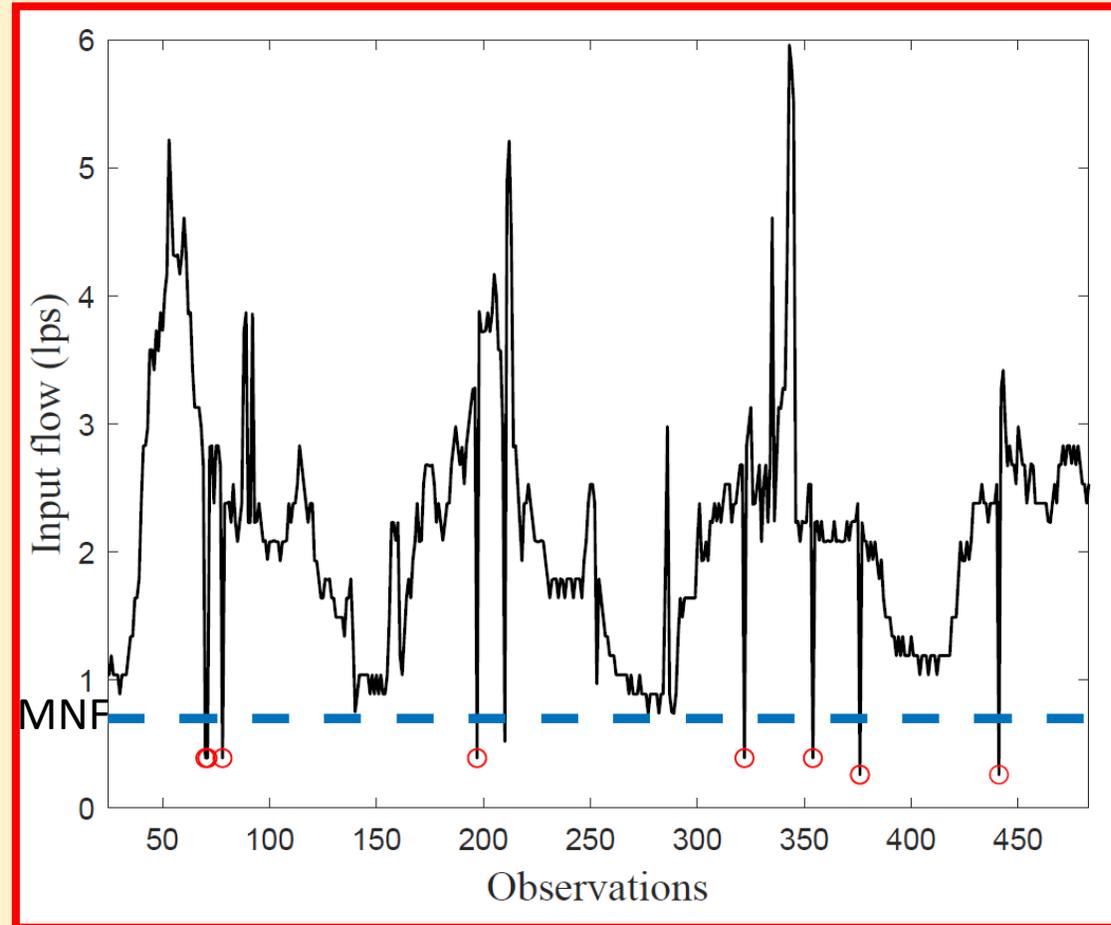


- Isolation of outliers from clusters

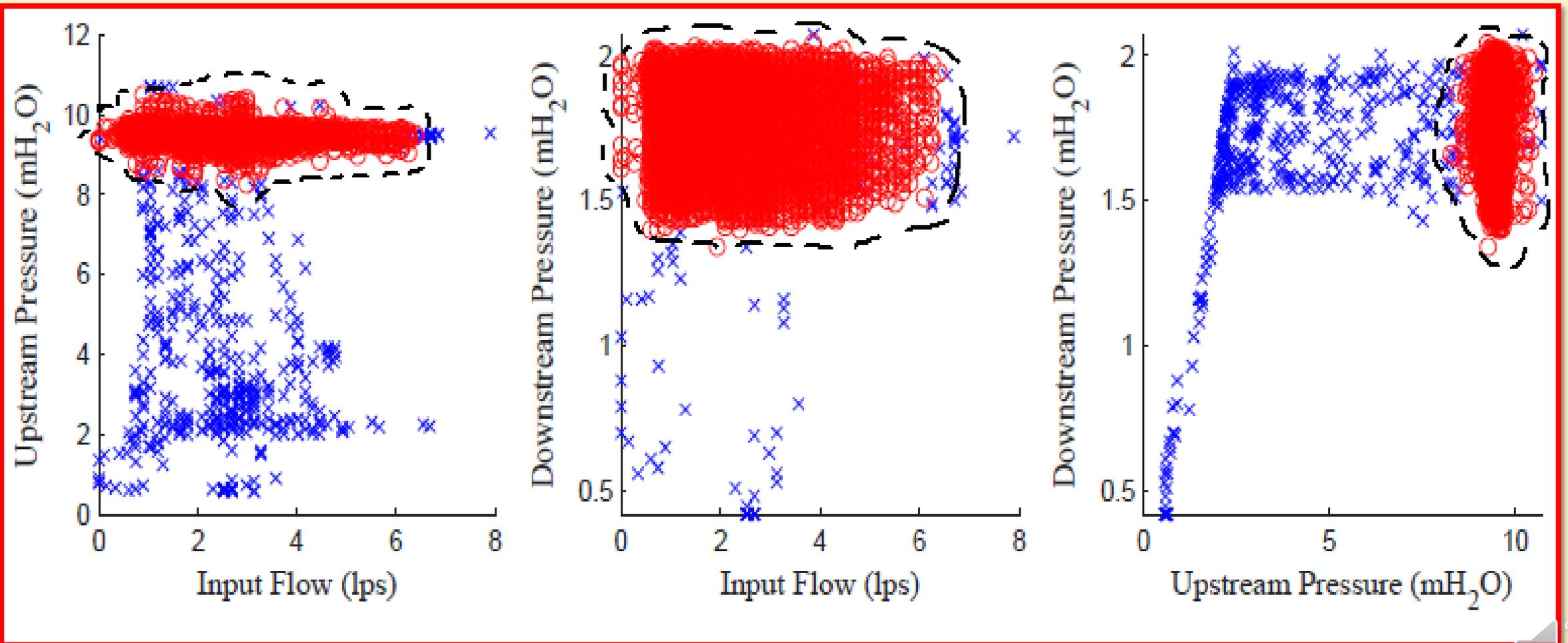


5.- Results & Discussion

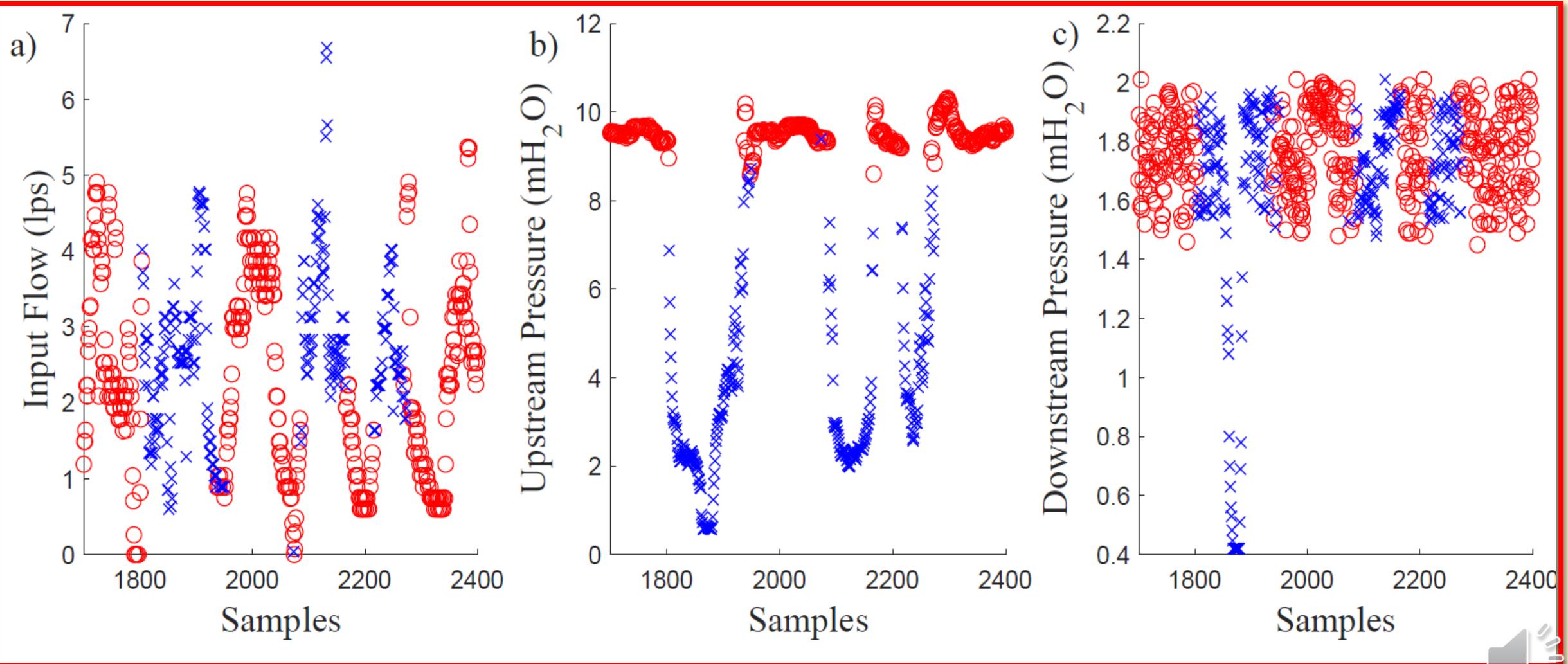
- Preprocessing Tasks



5. Clusters of Normal & Abnormal Data



5.- Draining of the reservoir



5.- Conclusions

- An off-line approach to data validation in WN is introduced.
- The core of the proposal is the application of an unsupervised clustering method without feature definition for the diverse data sets to be identified.
- The application of the cluster algorithm to the DMA El Charro allowed the identification of a systematic anomaly: the reservoir draining.
- Given the results, the network operators concluded the convenience of the pressure reducing valve for the DMA.

Thanks to you for the attention!
&
we are open to questions by email

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