

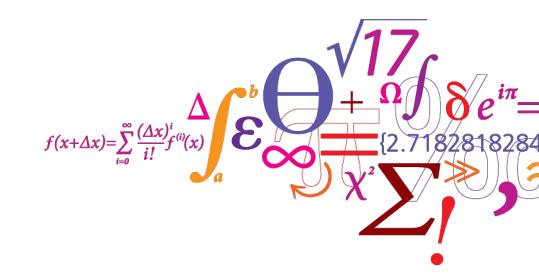
# Classification of electricity consumption using smart meter data

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In collaboration with:







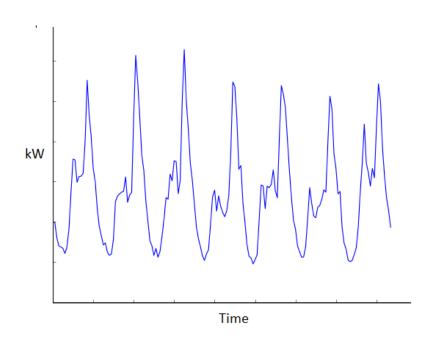
#### DTU Management Engineering

Department of Management Engineering



# **Smart Meters and Analytics**







### **Presentation Outline**

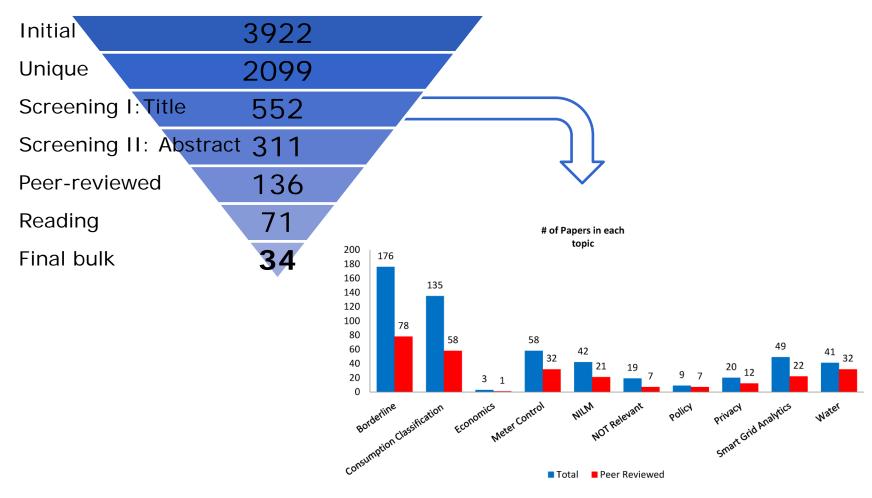
- Systematic Literature Review
  - Structured Literature Review of Electricity Consumption Classification Using Smart Meter Data
  - http://www.mdpi.com/1996-1073/10/5/584
- Analysis
  - SydEnergi data



# classification profiling segmentation learning feature load big Statistics monitoring machine time series mining statistical Residential clustering extraction electricity construction consumption Smart Customer Analysis Smart analytics



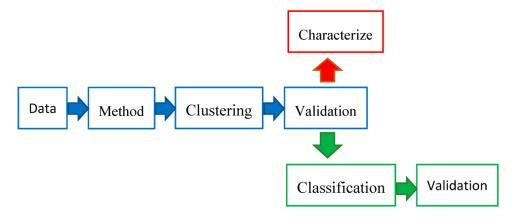
### **Smart Meter Papers**





## Common Clustering methods and workflow

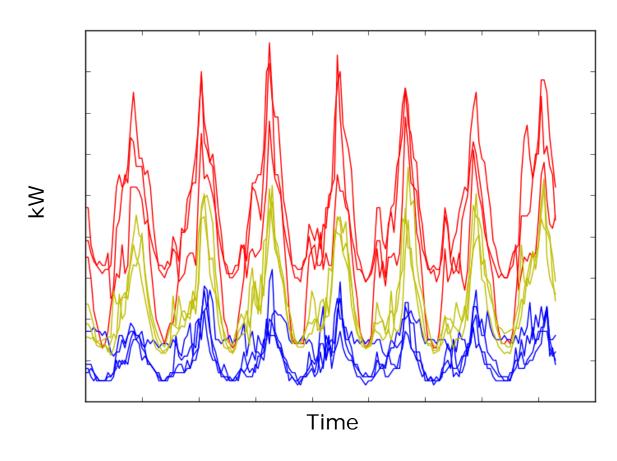
Metode	Total
K-Means	65%
Hierarchical	45%
Fuzzy K-Means	12%
Follow-The-leader	9%
Mixture Model	9%
K-Medoid	6%
Neural Network	6%
Fast Fourier Transform	3%





## Is it possible to classify smart meter data ...?

Classification 10 meters 3 classes k-means



(ENSYMORA data)



# SydEnergi Smart Meter Data

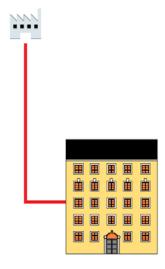


# **SydEnergi**



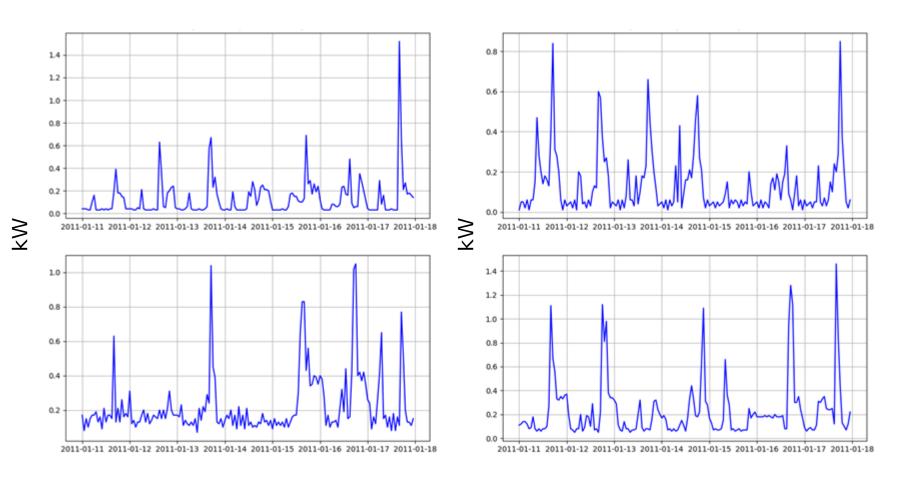
#### Subset:

100 Apartments with district heating in Esbjerg.



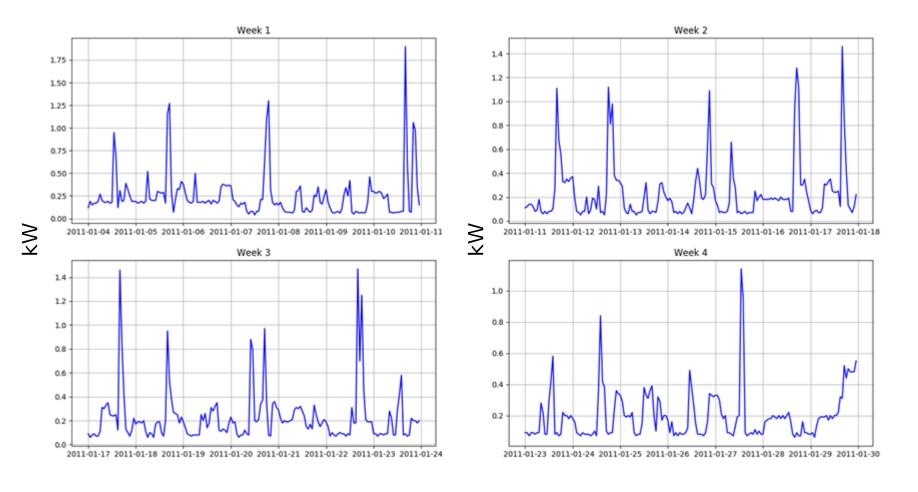


# January 2011 (11th-17th) 4 meters (mon-sun)





## January 2011 1 meter

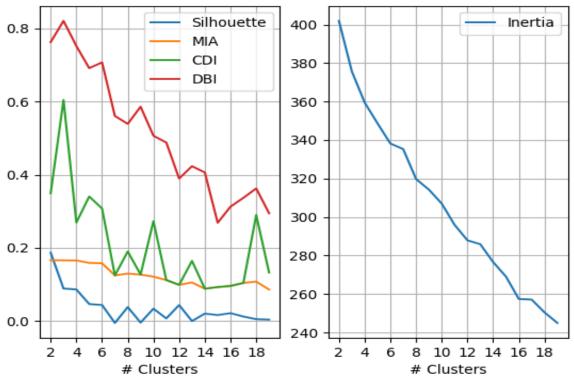




## **Optimum number of clusters**

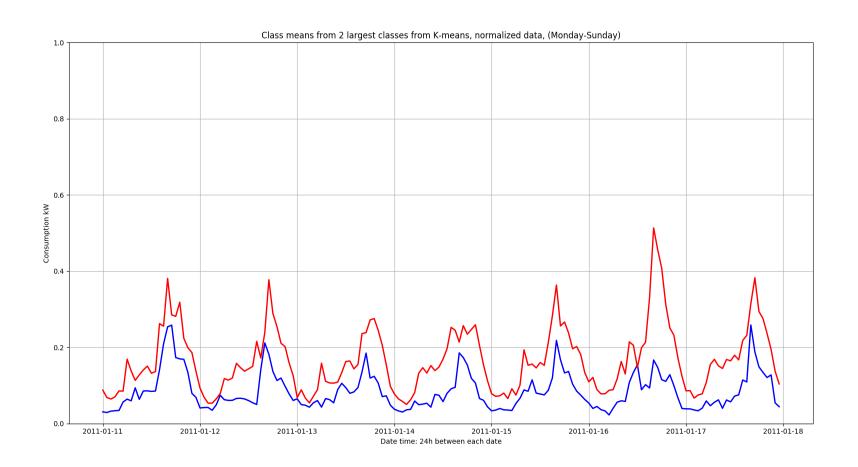
K-means classification: 2-20 classes, 100 meters





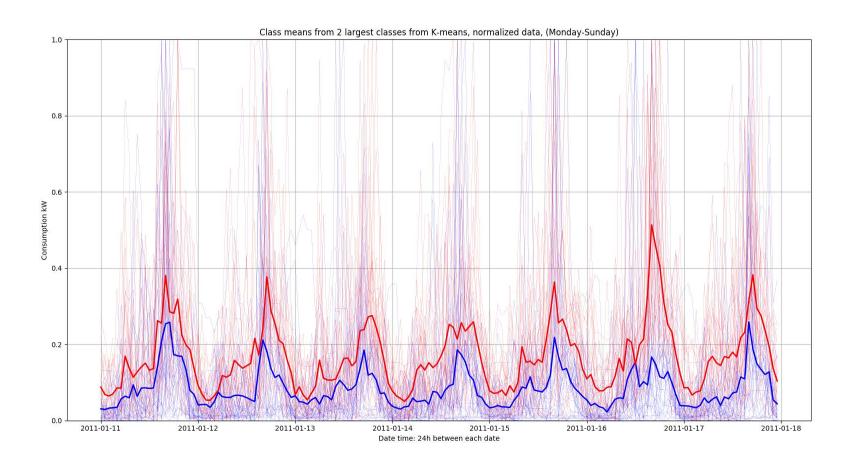


# Mean of 2 largest classes





# 2 largest classes all meters





#### Conclusion

- Classification is possible
  - Cluster dispersion is large
  - Not much help from reducing time window, month...week...day...
- Potential:
  - Adding Socio-economicdemographic and housing data
  - Feature extraction of meter data:
    - Wavelets / Splines
    - Principal components
    - Account for autocorrelation...
    - Weather

Audience input...