

Parsimonious models in monitoring the environment and wireless sensor networks

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Research group

- Group leader:
 - Jaakko Hollmén
- Post-doc:
 - Mika Sulkava
- Graduate students:
 - Mikko Korpela
 - Janne Toivola
 - Prem Raj Adhikari
- Visitors 2010-2011: Miguel Angel Prada, Serafín Alonso Castro, Antonio Morán, and Andrés Sanz García

Methodological goal: parsimony

- Learning parsimonious models from large and high-dimensional noisy data sets
- Parsimonious = compact, sparse, economical
- *Mission: model representation* should be as simple as possible or minimal
- Computationally intensive algorithms
- Good basis for interpretation

Current research projects

- Environmental Informatics for Analyzing the Role of Forests in the Global Carbon Cycle (2009-2011)
- Environmental proxy selection in temperature reconstruction
- Intelligent Structural Health Monitoring System (2008-2011)
- Gene selection in time-series gene expression data

Recent activity



- Organization of the 14th International Conference on Discovery Science (DS 2011)
- Local organization of DS 2011 and the 22nd International Conference on Algorithmic Learning Theory (ALT 2011), 5-7.10.2011

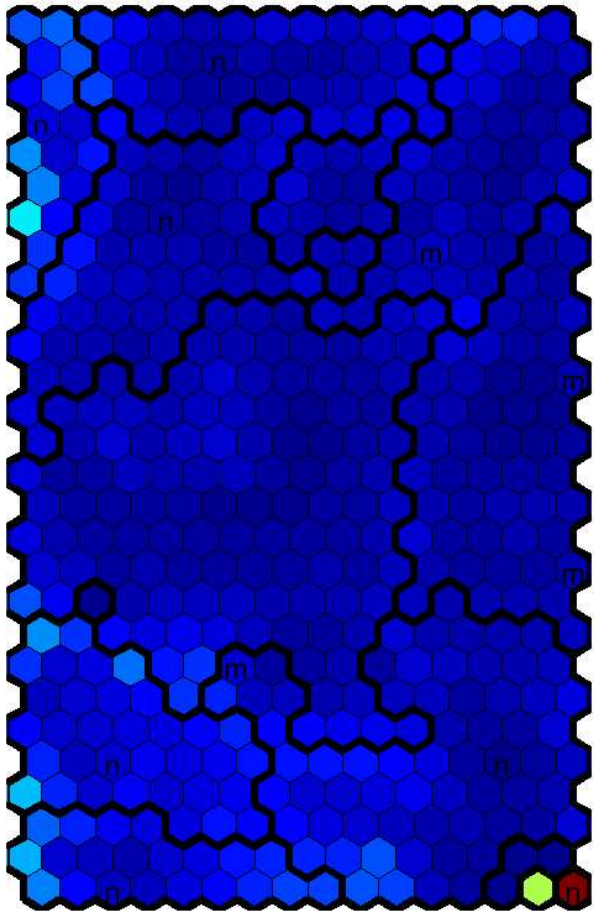
Examples of current research

- Design of ecosystem monitoring networks (Mika Sulkava)
- Environmental proxy selection problems in temperature reconstruction (Mikko Korpela)
- Collaborative filtering for coordinated monitoring in wireless sensor networks (Janne Toivola)

Monitoring network design

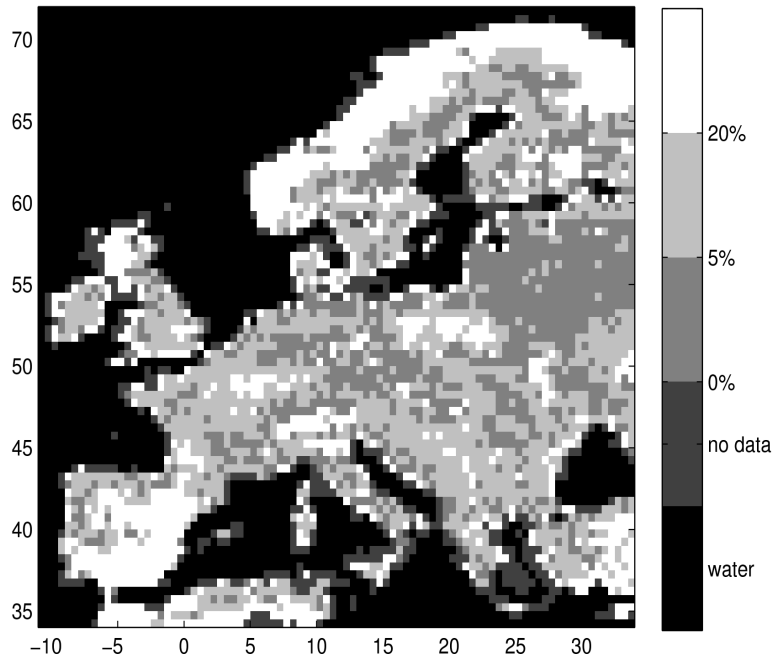
- Understand the complex variability of the global flow of carbon and the climate, the growth of forests and the interconnections between them
- Design a representative network of measurement towers
- Carbon exchange between ecosystems and atmosphere
- Collaboration: Laboratoire des Sciences du Climat et de l'Environnement, France, University of Tuscia, Italy

Monitoring network design



- CO₂ exchange of ecosystems is measured in towers around the world
- How many towers are needed for good representation?
- Where should new towers be located?
- Multidimensional climatic space
- Stratify study area using SOM and clustering

Monitoring network design

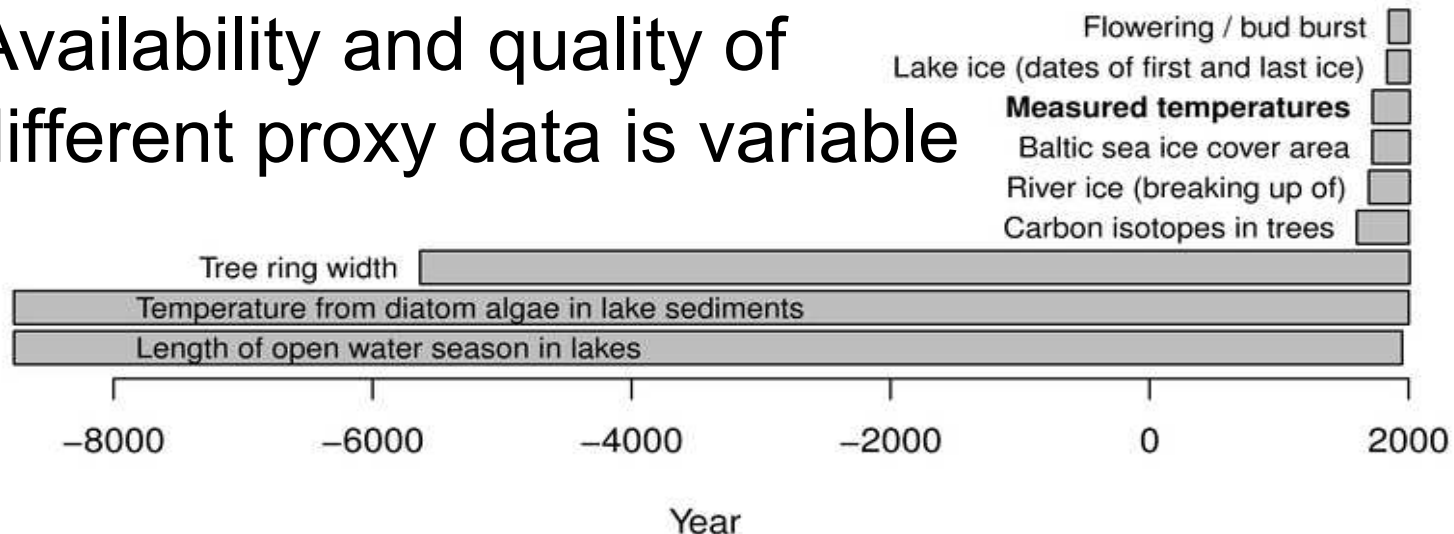


Upscaling error of a tower network

- Select locations of towers based on clustering
- Use an ecosystem model to estimate carbon exchange
- Upscale carbon exchange to the whole study area using the selected tower locations
- Use upscaling results to compare different networks
- Current research: importance of extreme climatic conditions

Environmental proxy selection problems in temperature reconstruction

- Direct temperature measurements (partly) available from the past few hundred years
- Proxies must be used for estimating longer temperature records
- Availability and quality of different proxy data is variable



Comparison of published temperature records

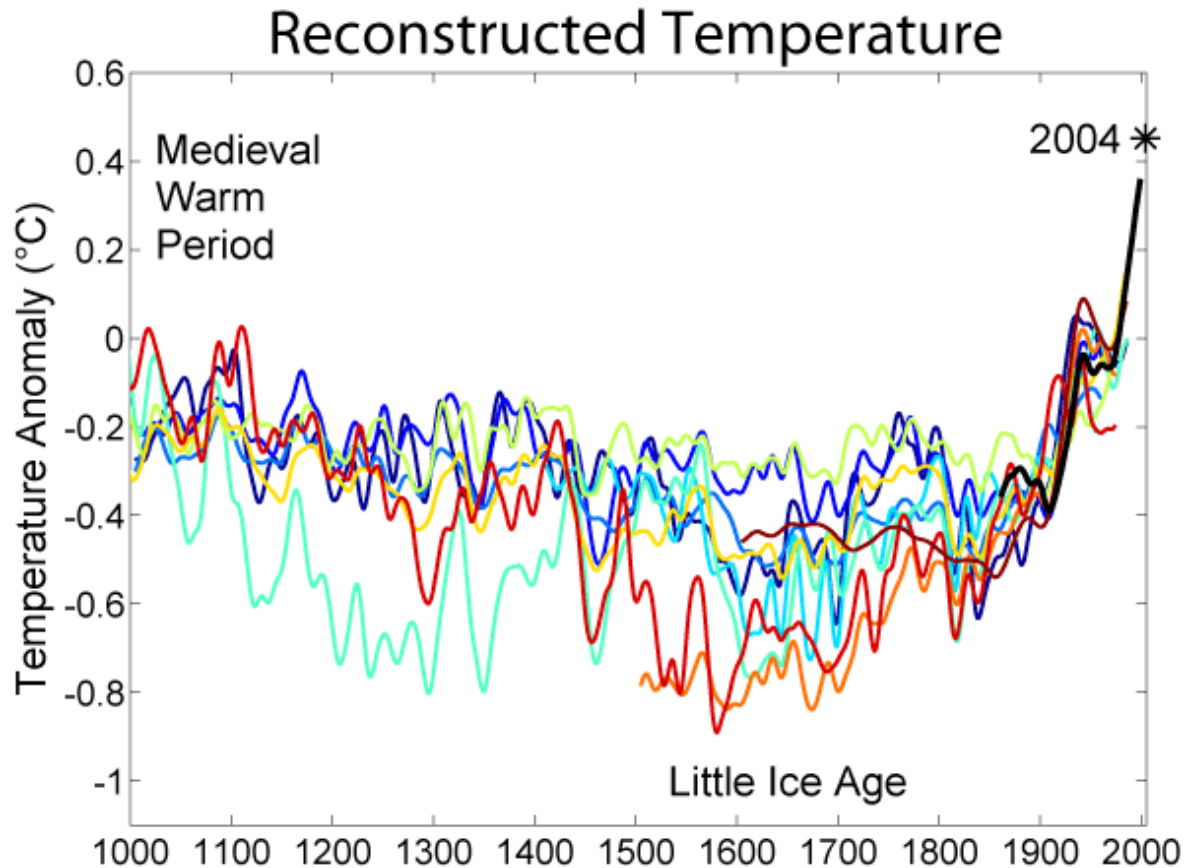
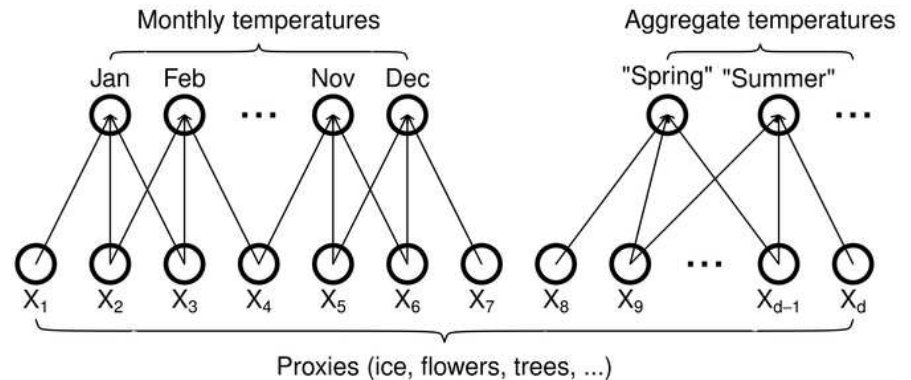
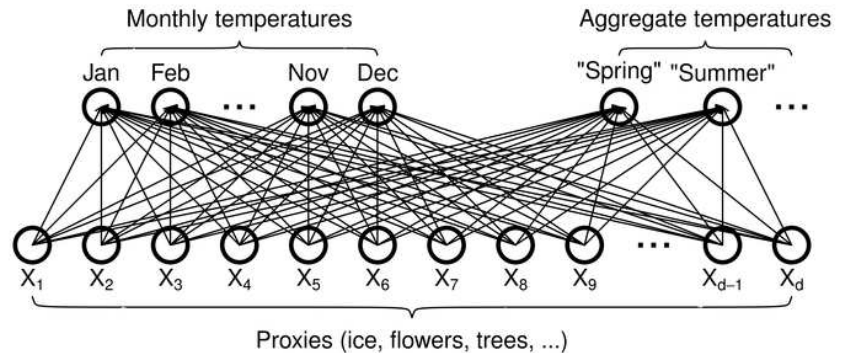


Image created by Robert A. Rohde / Global Warming Art

http://www.globalwarmingart.com/wiki/File:1000_Year_Temperature_Comparison_png

Variable selection problem

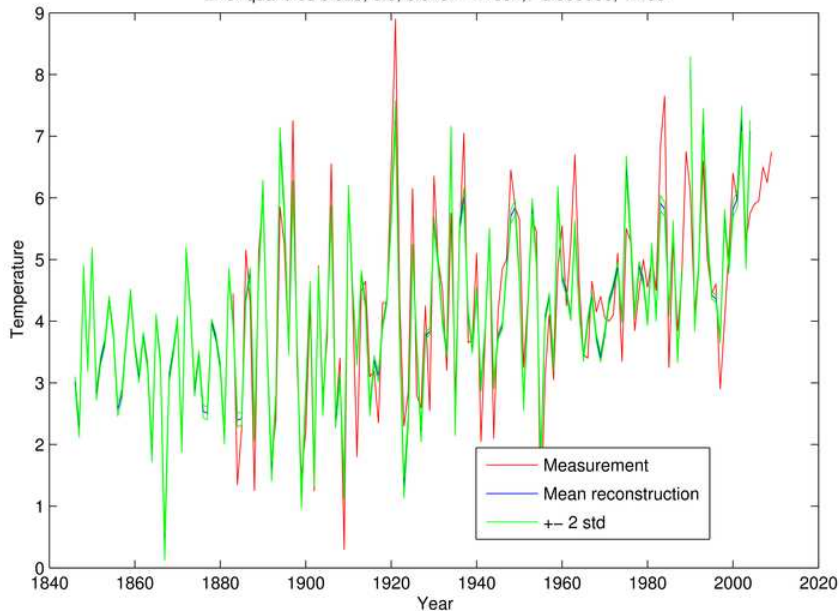
- Full model is not plausible
- Find relevant proxies
 - Dependent on month / season
- Good temperature reconstruction possible for all seasons?



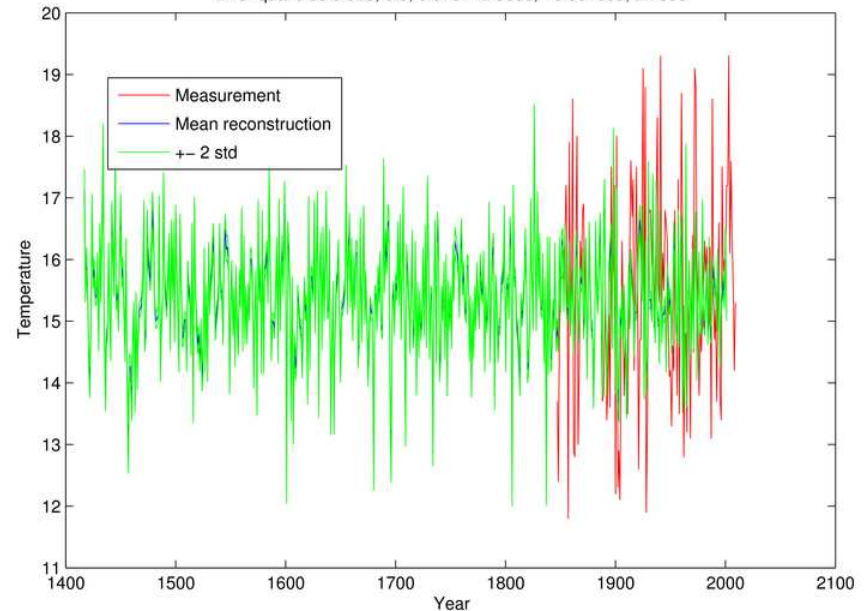
Temperature reconstruction experiments (early results)

- With models using one proxy, there are large differences in the quality of the reconstructions

Cross-validated temperature reconstruction with one proxy
Mean of Jyvaskyla April and Jyvaskyla May vs. Bird cherry
Mean correlation 0.89505, overlap of 121 years in 1883–2004
RMSE 0.68943, mean abs error 0.54265
Error quantiles 0.025, 0.5, 0.975: -1.1834, -0.035986, 1.483

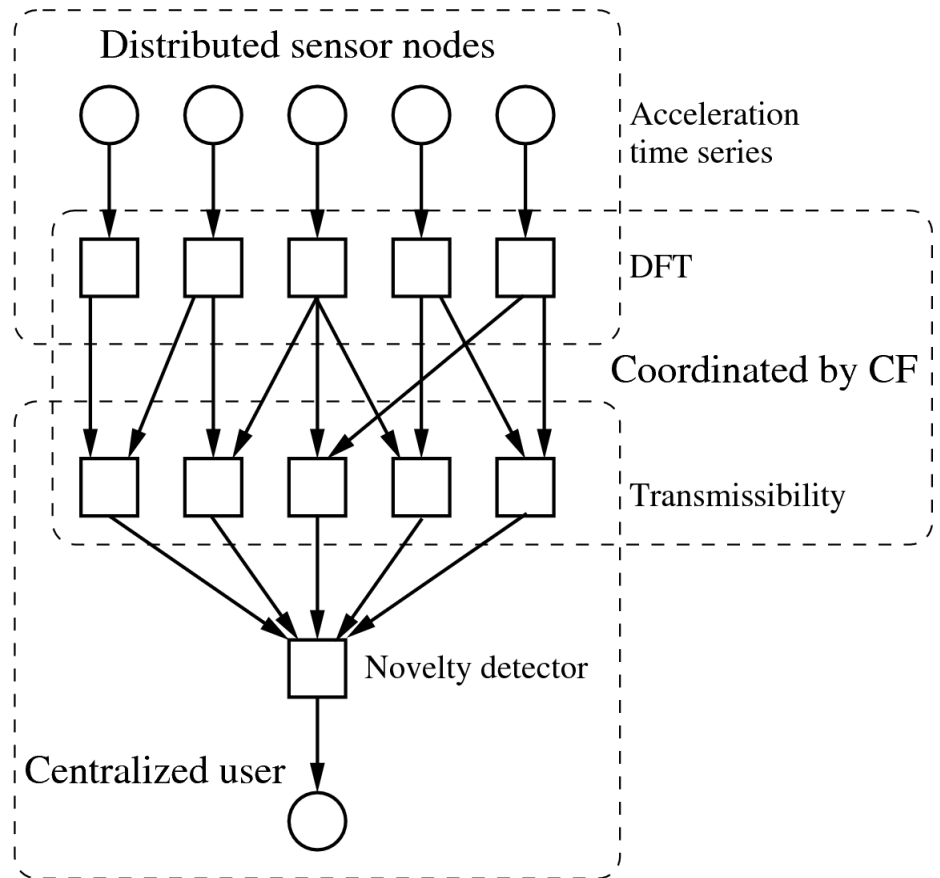


Cross-validated temperature reconstruction with one proxy
Kajaani July vs. Pine north
Mean correlation 0.45609, overlap of 141 years in 1847–2001
RMSE 1.4764, mean abs error 1.1965
Error quantiles 0.025, 0.5, 0.975: -2.8803, -0.031605, 2.7635



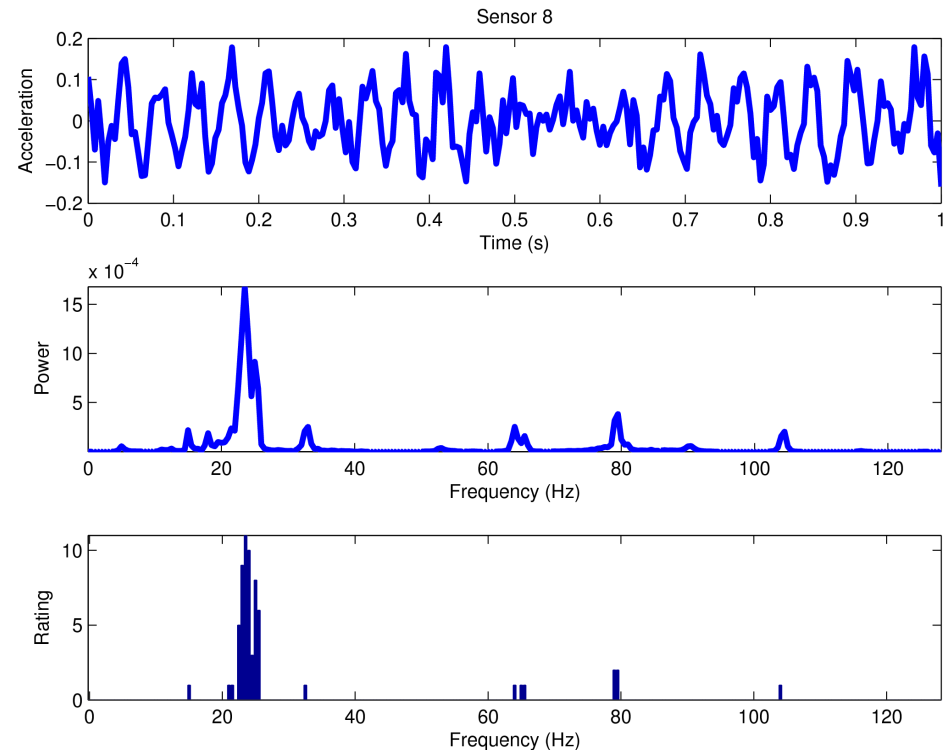
Collaborative filtering in WSN

- Set of sensors
 - Not detectors...
- Local features
 - Max. D / node
- Combined features
 - Depend on distributed nodes
- Novelty detection
 - Wrapper-based feature selection not possible

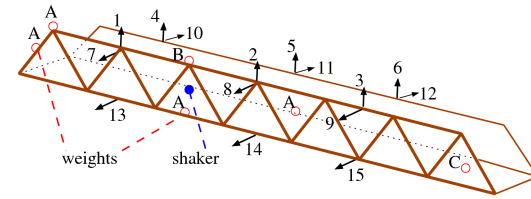


SHM: vibration power spectrum ratings

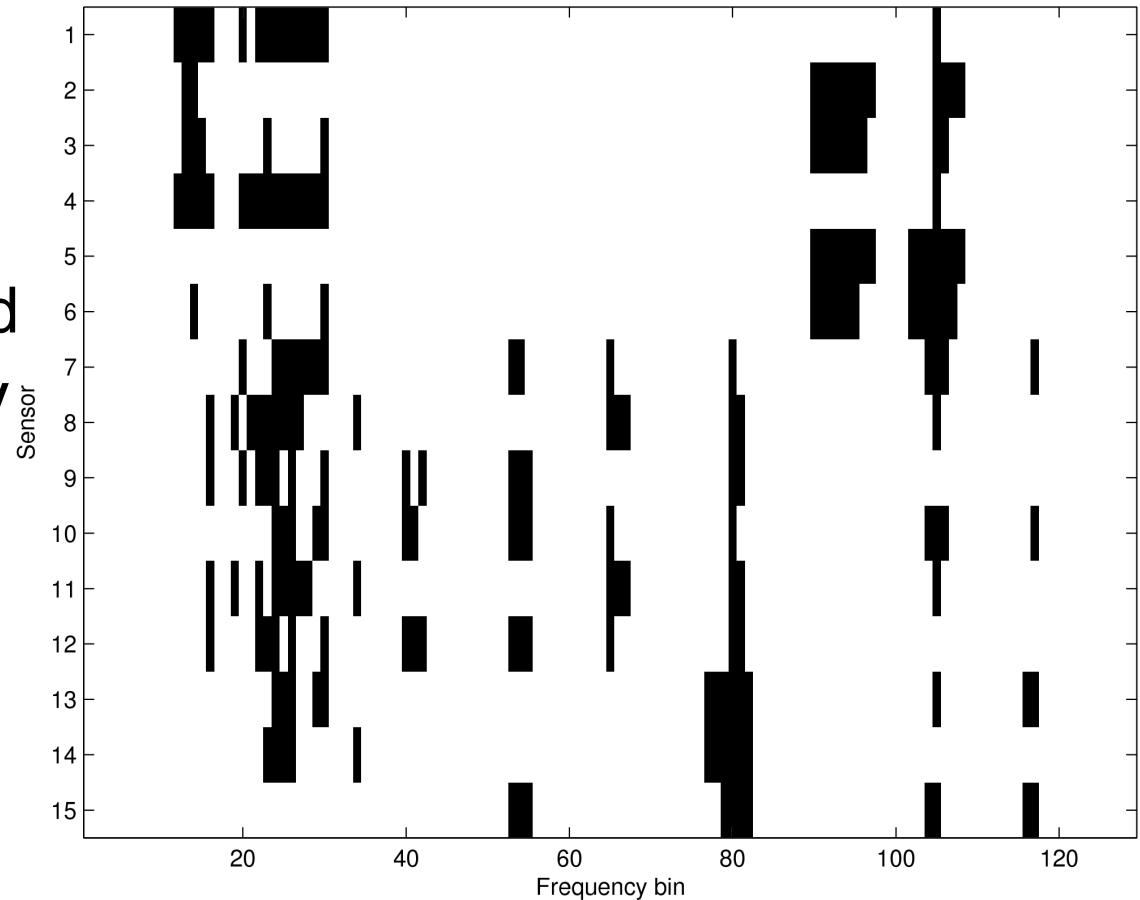
- Accelerometers
- Local features
 - Power spectrum: $X[k]$
- Local ratings
 - Accumulated votes for D “best” bins
- Combined features
 - Transmissibility: $X^1[k] / X^2[k]$
- Collaborative filtering
 - Select best features



SHM: filtering example



- 15 sensors
- 128 features
 - Max. 20 / node to be monitored
- Transmissibility
 - Pairs of black bars
- Sensors both collaborate and specialize



SHM: detection accuracy (AUC)

- Area under ROC curve
 - 10 iterations
 - min, med, max
- $D = 2 \dots 20$
- Collaborative filtering
- Majority vote
- Random selection

