



## **Some Ideas on Data Rate Reduction for Measurement Data Transmissions in Hierarchical Smart Grid Environments**

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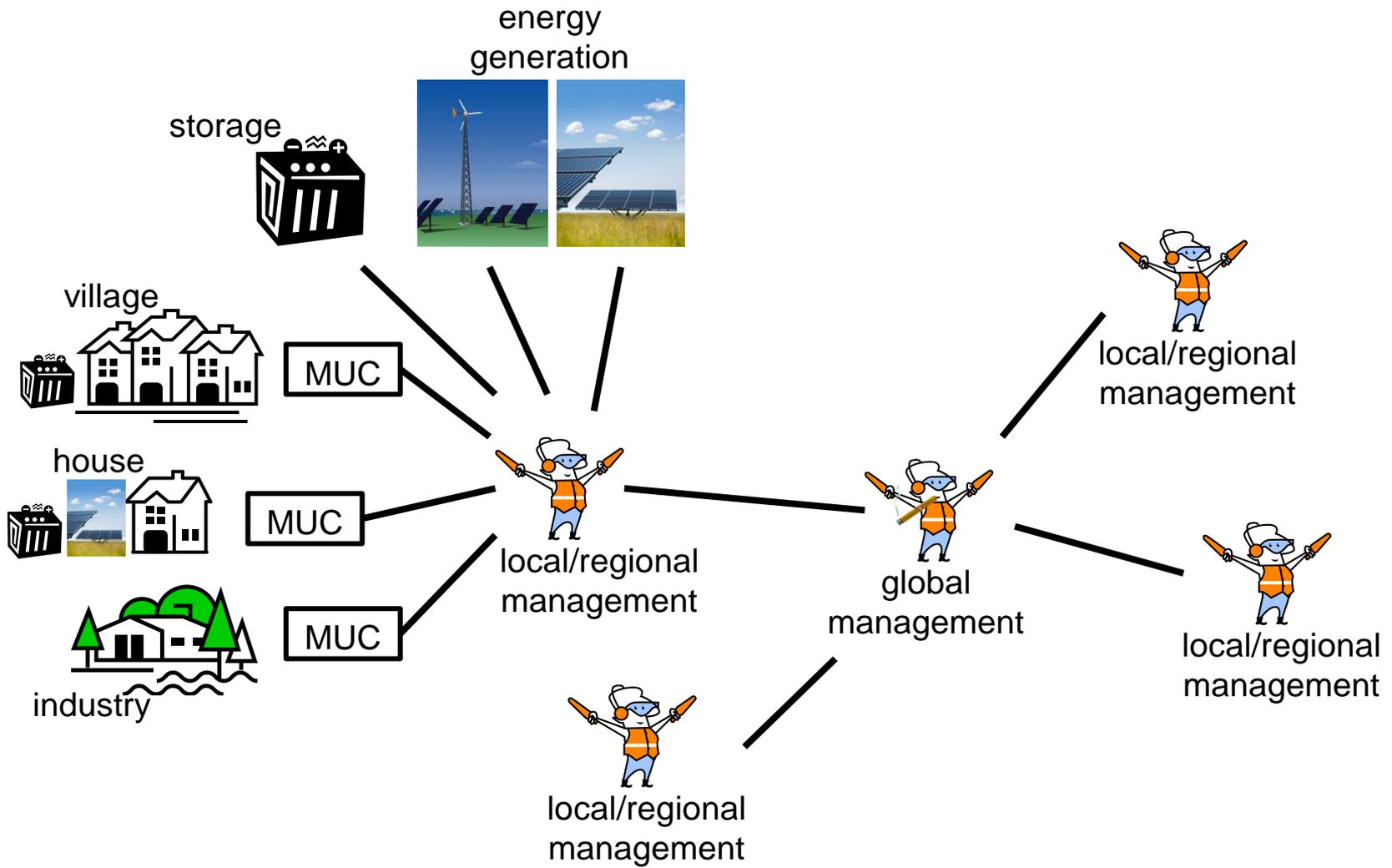
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## Aim

- Usage of regression models for reducing communication demands of measurement data transmissions
- Original time series of measured data is represented by a regression model
  - Lossy representation of time series!
  - Increased level of uncertainty!
- Only model parameters and a few additional parameters are transmitted
- Model is used to forecast behavior until next model parameter transmission

# Environment



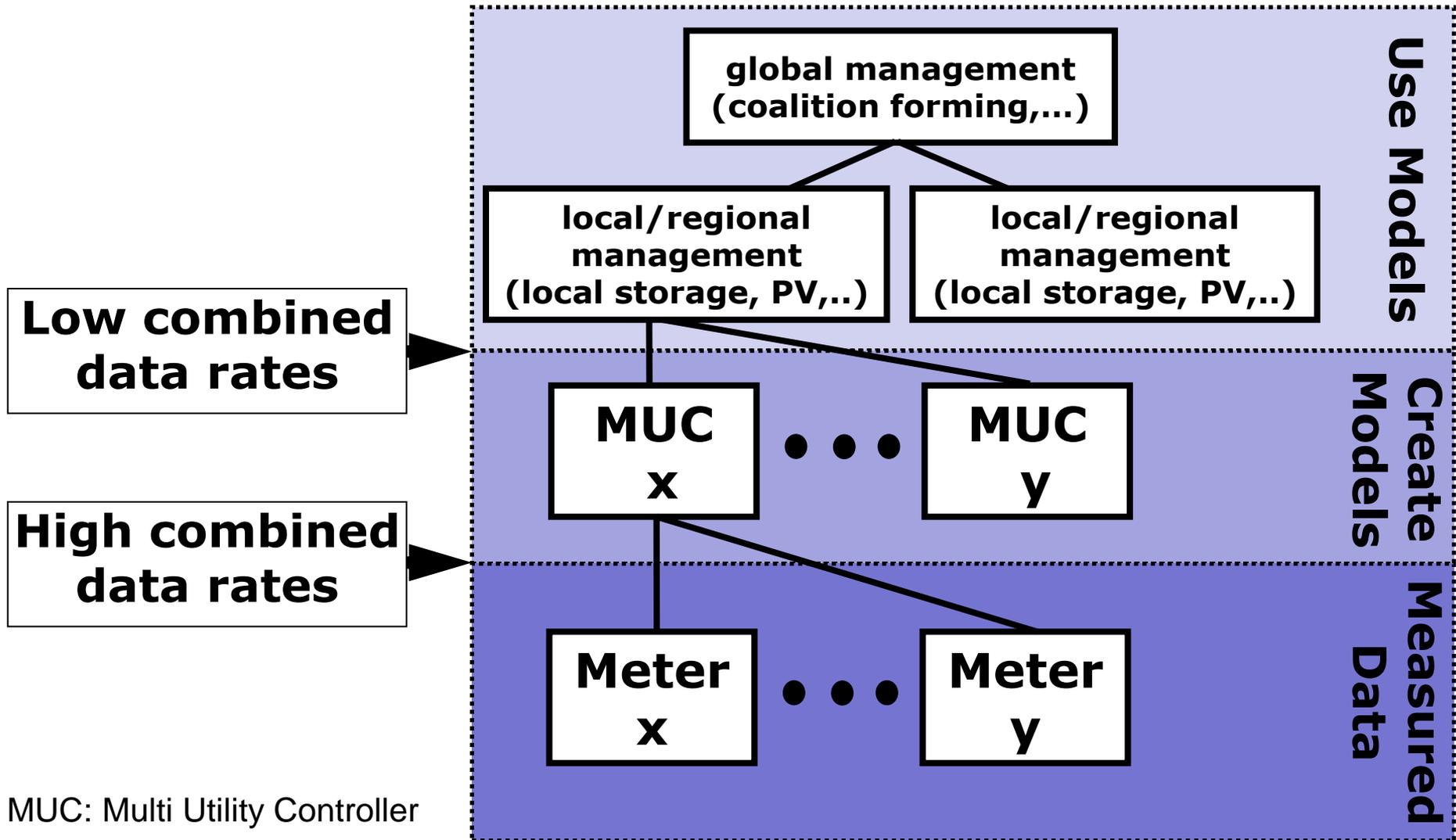
# Outline

- System Model Description
- Example
- Open Questions

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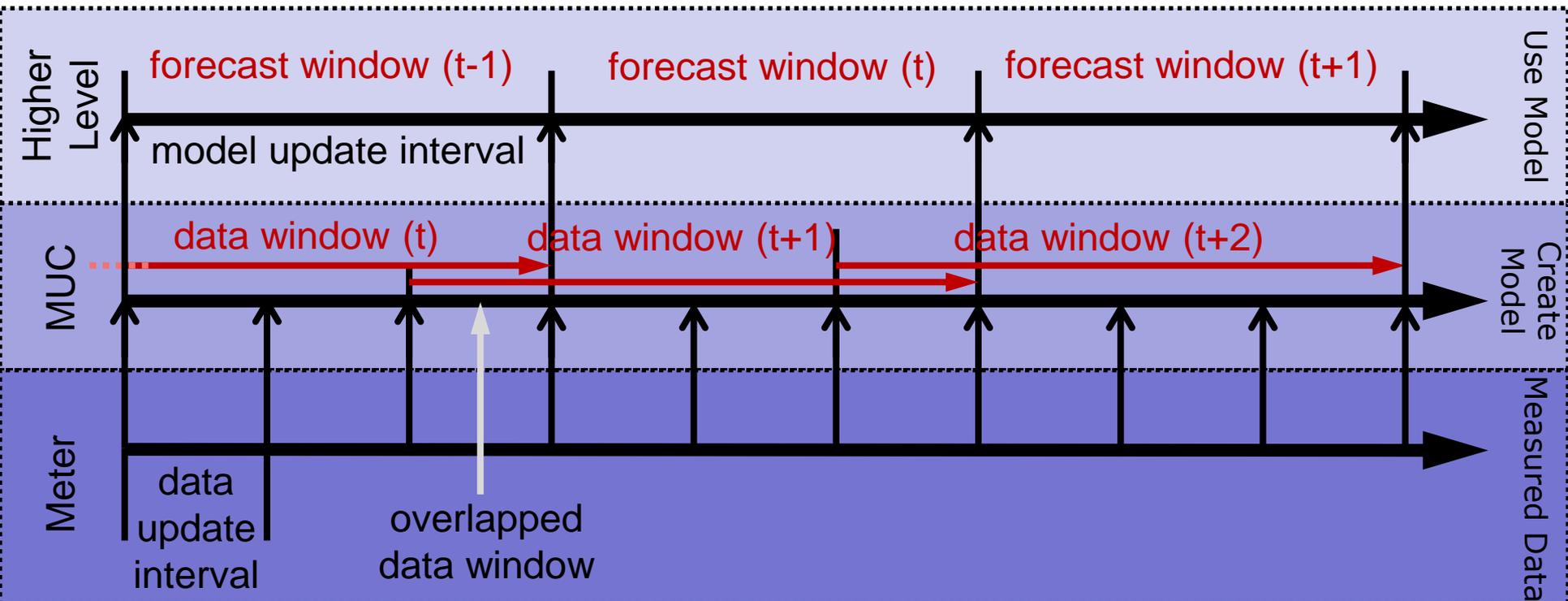
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# General System Model



MUC: Multi Utility Controller

# Model Update Process



intended time horizon: sec up to hour

# Model Update Process - Commentary

- Data updates from meter represent time series at MUC
- Data update interval is not necessarily constant! (e.g. due to intelligent recording)
- Data updates within one data window are used to estimate regression model parameters and variance
- Model parameters are sent to higher level according to the model update interval (interval is not necessarily constant!)
- Model parameters are used to forecast behavior until next model update event
  - Short term forecasts! Intended time horizon is seconds up to hours

# Implications Derived from Update Process

- Size of model parameters has to be smaller than size of time series information within one data window
  - Model update interval has to be large enough
  - Lossy model -> increased uncertainty
- Data window should cover enough data updates to obtain a sufficient time series for model parameter estimation
- Time series information can be reused for model parameter estimation (overlapping data windows)
  - Higher storage requirements in MUC
  - Model is less responsive to short term changes if overlapping window increases
- Forecast window should be shorter than data window to achieve sufficiently accurate forecasts
  - Short enough to be accurate and long enough to reduce data transfers sufficiently

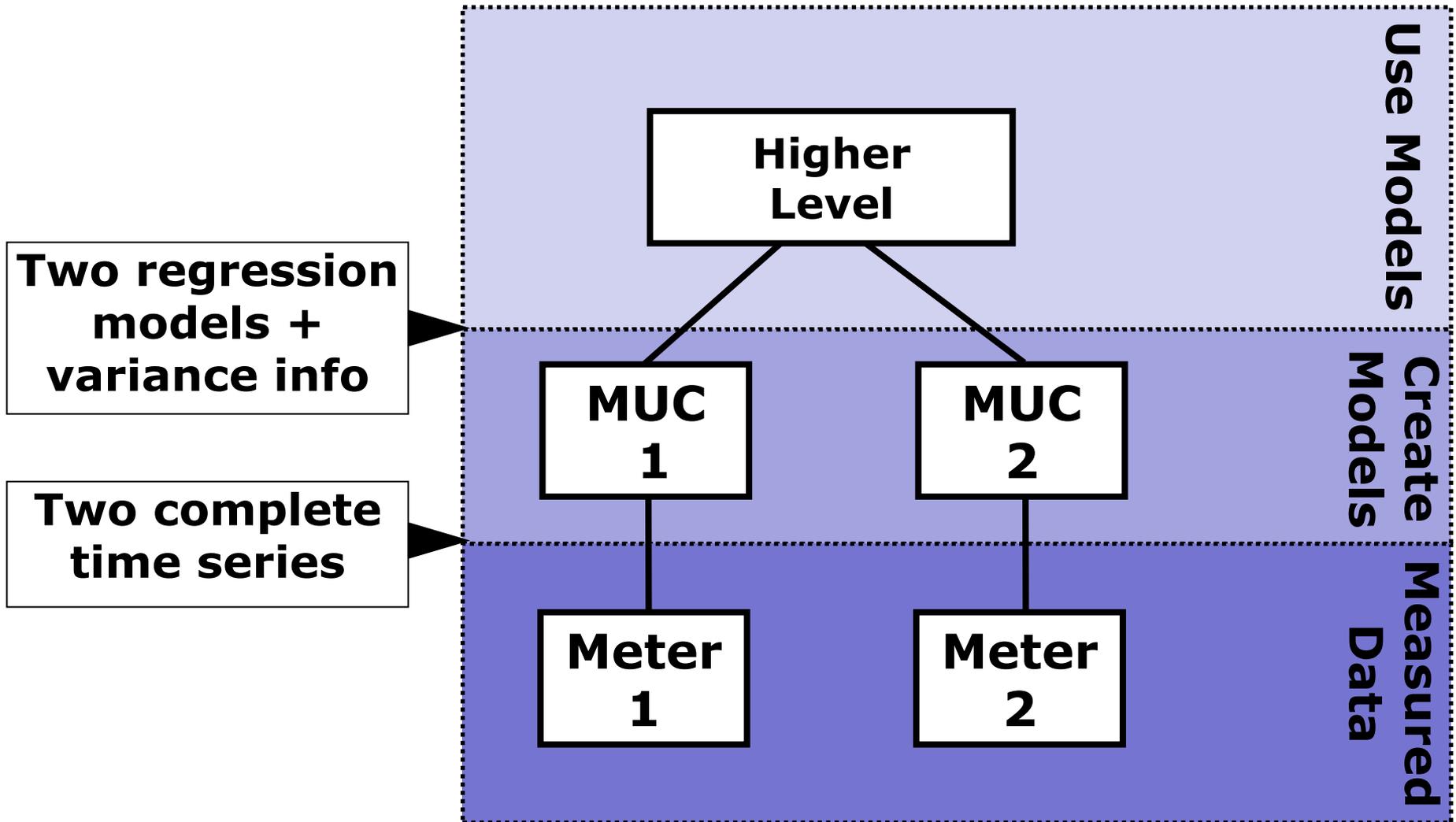
# Model Parameter Estimation

- Regression model has to be defined in advance
  - MUCs and higher levels have to know the model
  - Necessary for merging parameters
- Straightforward combination of regression model parameters
  - Parameters merge like original data
  - Good for hierarchical system models (Smart Grid)
  - Low computational effort for merging
- Estimation of merged noise (variance) is more difficult
  - Solved by heuristic estimation of cross correlation based on model cross correlation and variance information

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## Example Model



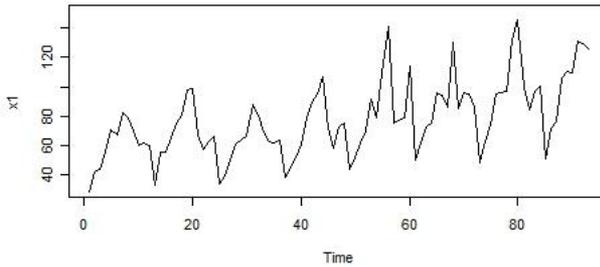
# Regression Model

- Linear model with additive seasonal variables
  - Just as an example!
  - $b_i$ ,  $s_i$ ,  $c_i$  (slopes) &  $f$  (period frequency) are model parameters
  - Time is the only modeled influence
    - ▶ Other predictor variables can be integrated
  - $w_t$  noise information

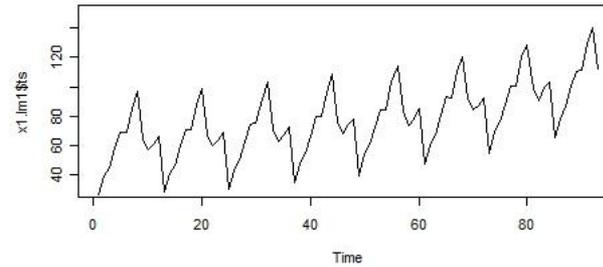
$$x_t = b_0 + b_1 t + b_2 t^2 + \sum_{i=1}^{\lfloor f/2 \rfloor} \left\{ s_i \sin\left(\frac{2\pi i t}{f}\right) + c_i \cos\left(\frac{2\pi i t}{f}\right) \right\} + w_t$$

# Merging Process of Example

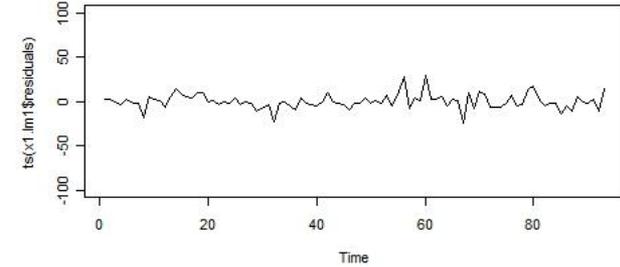
1. meter data



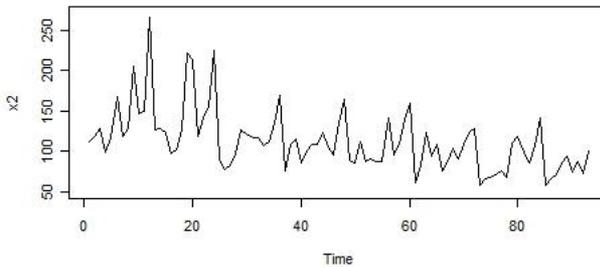
1. model (by MUC 1)



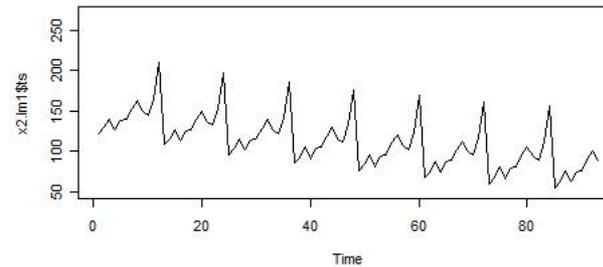
1. model norm noise



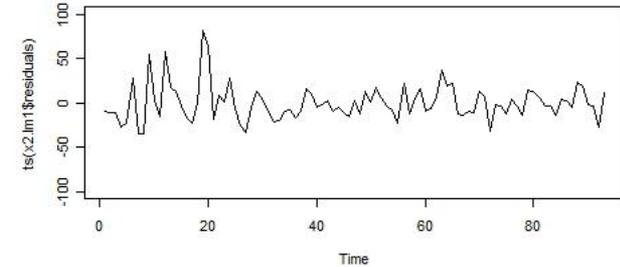
2. meter data



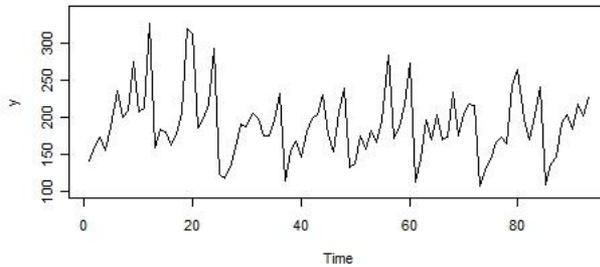
2. model (by MUC 2)



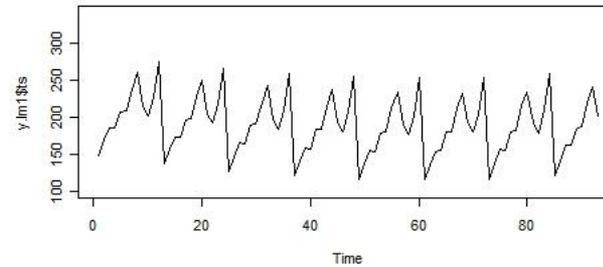
2. model norm noise



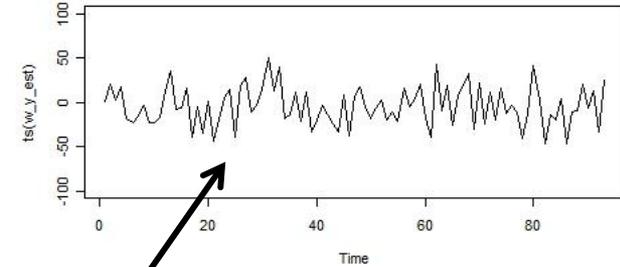
merged orig. data



merged model (by higher level)



merged model est. norm noise



random process with distribution type and variance estimated based on model cross correlation and variance of 1st and 2nd model

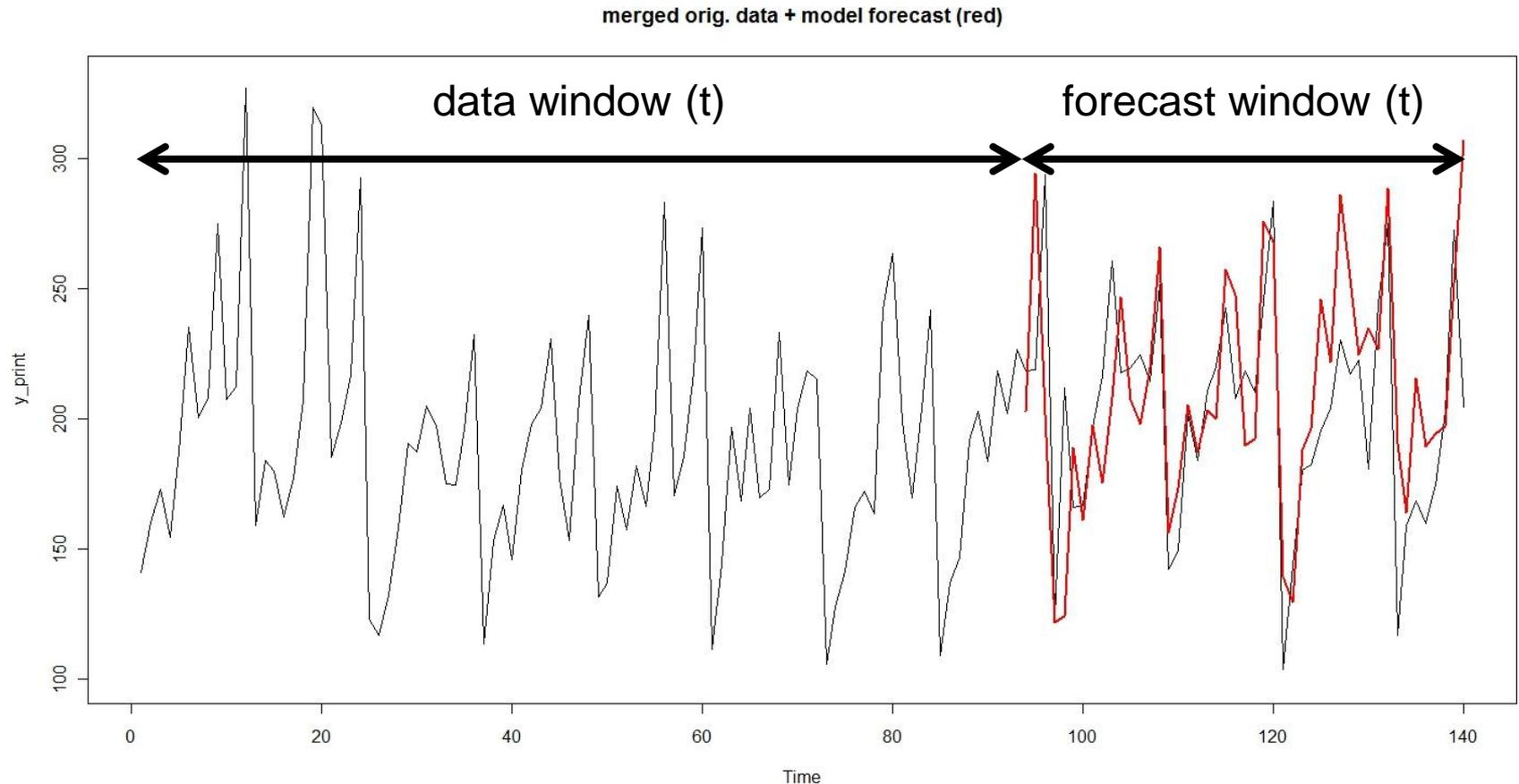
## Comparison of Models for Merged Data

- Parameter match of merged model and model of original merged data

Parameter	Difference
b0	-2.842171e-14
b1	-2.220446e-16
b2	1.734723e-18
c1	7.105427e-15
s1	-3.552714e-15
c2	7.105427e-15
s2	-4.440892e-16
c3	7.105427e-15

Parameter	Difference
s3	-1.776357e-15
c4	1.776357e-15
s4	-1.776357e-15
c5	0.000000e+00
s5	2.331468e-15
c6	1.776357e-15
s6	1.562500e-02

# Model Forecast for Example Data at Higher Level



- This forecast window has half the size of the data window
  - Forecast window should be shorter than data window
  - Relation depends on allowed uncertainty (to be investigated)

## Evaluation of Example

- Comparison of mean and variance in data window (t)

Parameter	Merged original data	Merged Model
Mean	189.7	189.7
Variance (incl. noise)	2075.4	1949.6
Variance of noise	591.6*	465.82 <sup>+</sup>

(\*) based on regression model residuals    (+) estimated

- Comparison of mean and variance in forecast window (t)

Parameter	Merged original data	Model forecast
Mean	203.3	211.2
Variance (incl. noise)	1810.7	1949.6
Variance of noise	n/a	465.82

## Evaluation of Example (II)

- Merged model fits well in data window
- Forecast is reasonable but has to be improved
  - Different regression model
    - ▶ Additional/other influencing variables (predictors)
  - Shorter forecast window
  - Different heuristic for variance estimation
- Original load time series needed to continue this work

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# Open Questions

- What kind of short term forecast models are already in use?
  - Time horizon: seconds, minutes or hours
  - Level: per household or per village
- What are the important parameters?
  - How important is variance information?
  - What are the important influences (e.g. sunshine level, wind, temperature...)?
- Is it possible to get original load time series?
  - Measured data at meters, at MUCs and at higher levels to derive a well fitting regression model