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# Analysis and Visualization of Large-Scale Power Quality Monitoring Campaigns

ICHQP 2024 Panel

Turning data into information: Recent developments and future needs in Power Quality data analysis  
October 17, Chengdu, China

# Agenda

## Power Quality data analysis

### Example use case: Correlation analysis

- Methodology and dataset
- Correlations between PQ parameters
- Correlations between measurement sites

## Conclusions

# Power Quality data analysis

## Data amounts

- Measurement of PQ parameters (current and voltage) for one measurement site

Measured	every 10 min	per day	per week	per year
up to 2.5 kHz	324	46.656	326.592	ca. 17 Mill.
up to 150 kHz	4767	686.016	ca. 5 Mill.	ca. 250 Mill.

- Increasing number of large PQ monitoring campaigns in all voltage levels
- Resulting data amounts can often no longer be evaluated manually

# Power Quality data analysis

## Use cases / analysis goals

- Compliance with limits
- Anomaly detection
- Emission profiles
- Correlation and propagation
- Seasonal variations
- Trend identification
- Trend forecasting
- ...

## General analysis steps

1. Define use case / analysis goal
2. Data acquisition
3. Data pre-processing
4. Analysis methods
5. Visualization and knowledge extraction

# Example use case: Correlation analysis

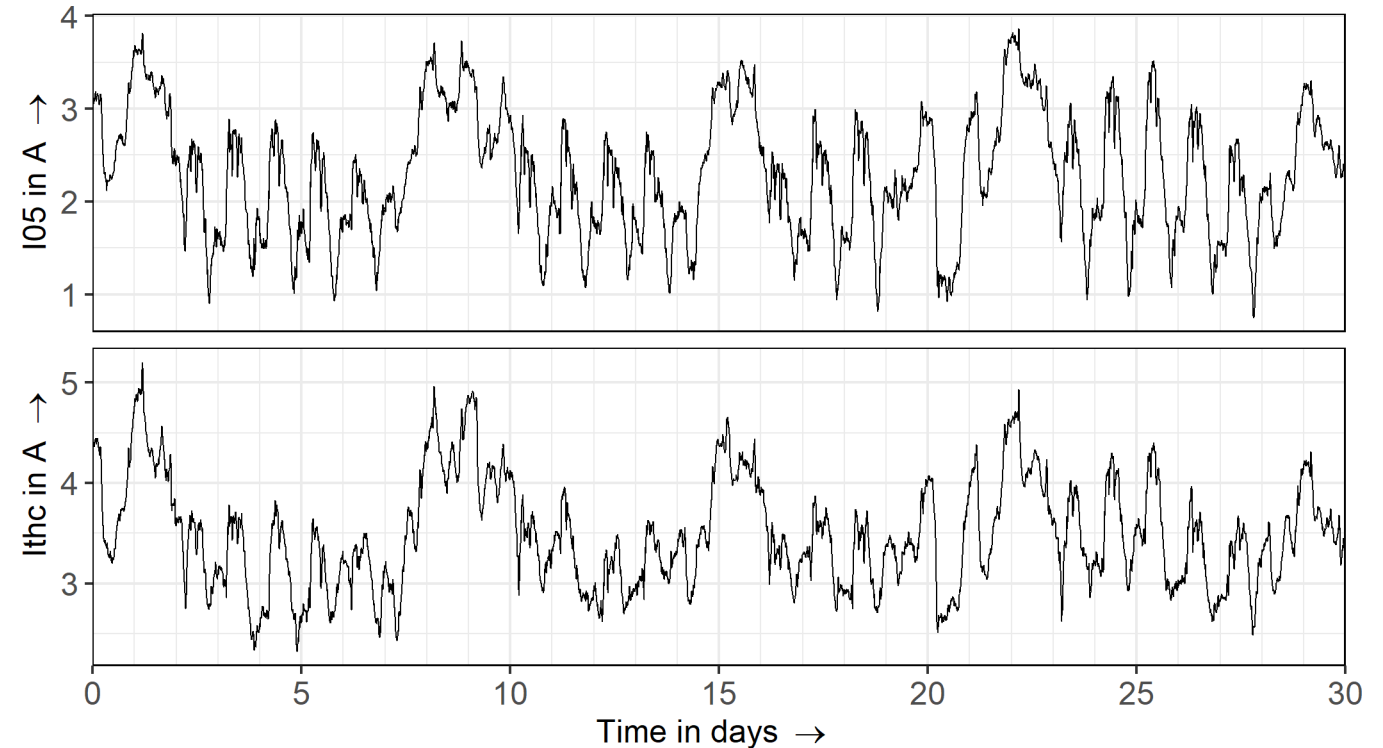
## Correlation analysis

- Key for analysing relationships and dependencies in measurement data
- Can help to identify similarities in observed behaviour or assess propagation within the network

## Types of correlations

- Auto-Correlation  
between different time intervals  
(e.g. days) within one time series
- Cross-Correlation  
between PQ parameters of one site or  
between sites for one PQ parameter

*PQ parameters of one measurement site*



# Example use case: Correlation analysis

## Methodology

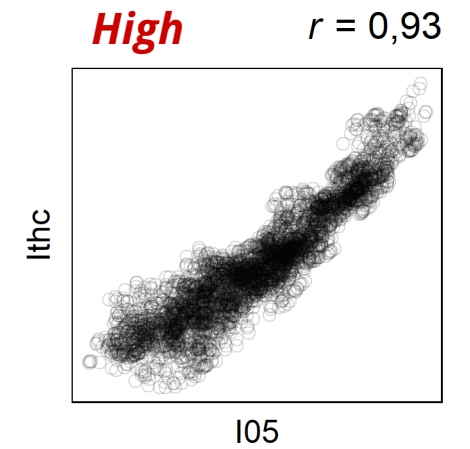
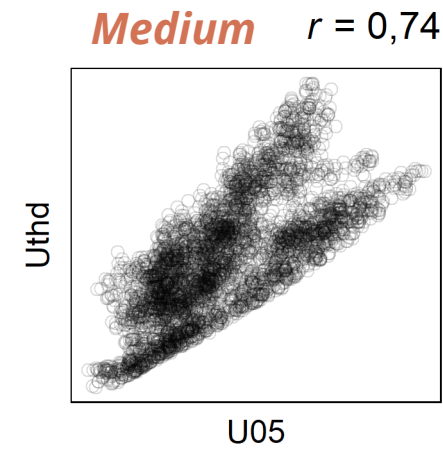
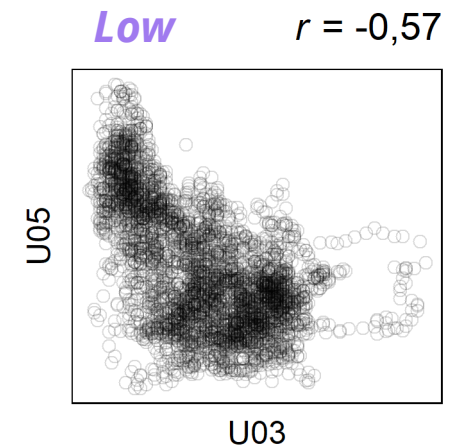
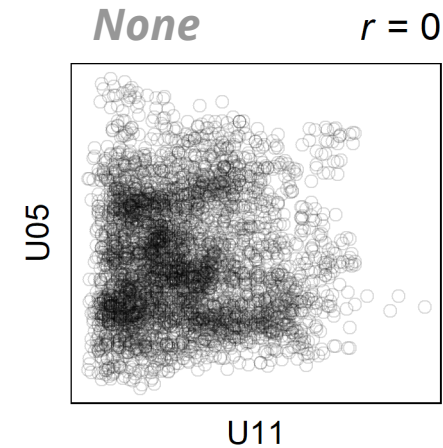
### Spearman's correlation coefficient

- Measure of monotonic behaviour based on ranks
- Not limited to linear relationships

$$r = \frac{\sum_i (R(x_i) - \bar{R}_x) (R(y_i) - \bar{R}_y)}{\sqrt{\sum_i (R(x_i) - \bar{R}_x)^2} \sqrt{\sum_i (R(y_i) - \bar{R}_y)^2}}$$

### Levels of correlation

- None  $0 \leq |r| < 0.5$
- Low  $0.5 \leq |r| < 0.7$
- Medium  $0.7 \leq |r| < 0.9$
- High  $0.9 \leq |r| \leq 1$



# Example use case: Correlation analysis

## Dataset

### Measurement data

- Monitoring of German transmission system
- **85 measurement sites** in HV and EHV
  - 110 kV with 38 sites
  - 220 kV with 21 sites
  - 380 kV with 26 sites
- **20 PQ parameters**
  - RMS of voltage ( $U_{rms}$ ) and current ( $I_{rms}$ )
  - Voltage unbalance (UNB)
  - Short-term flicker ( $U_{pst}$ )
  - Distortion of voltage ( $U_{thd}$ ) and current ( $I_{thc}$ )
  - Harmonics of voltage (U03-U15) and current (I03-I15)
  - In total 58 parameter-phase combinations
- 10 min average values for a duration of 30 days

### Correlations between PQ parameters

- 58 parameters = 1.653 unique correlations
- $1.653 \times 85 \text{ sites} = 140.505 \text{ correlations}$

### Correlations between measurement sites

- 85 parameters = 3.570 unique correlations
- $3.570 \times 58 \text{ parameters} = 207.060 \text{ correlations}$

**Analysis of a total of 347.565 correlations**

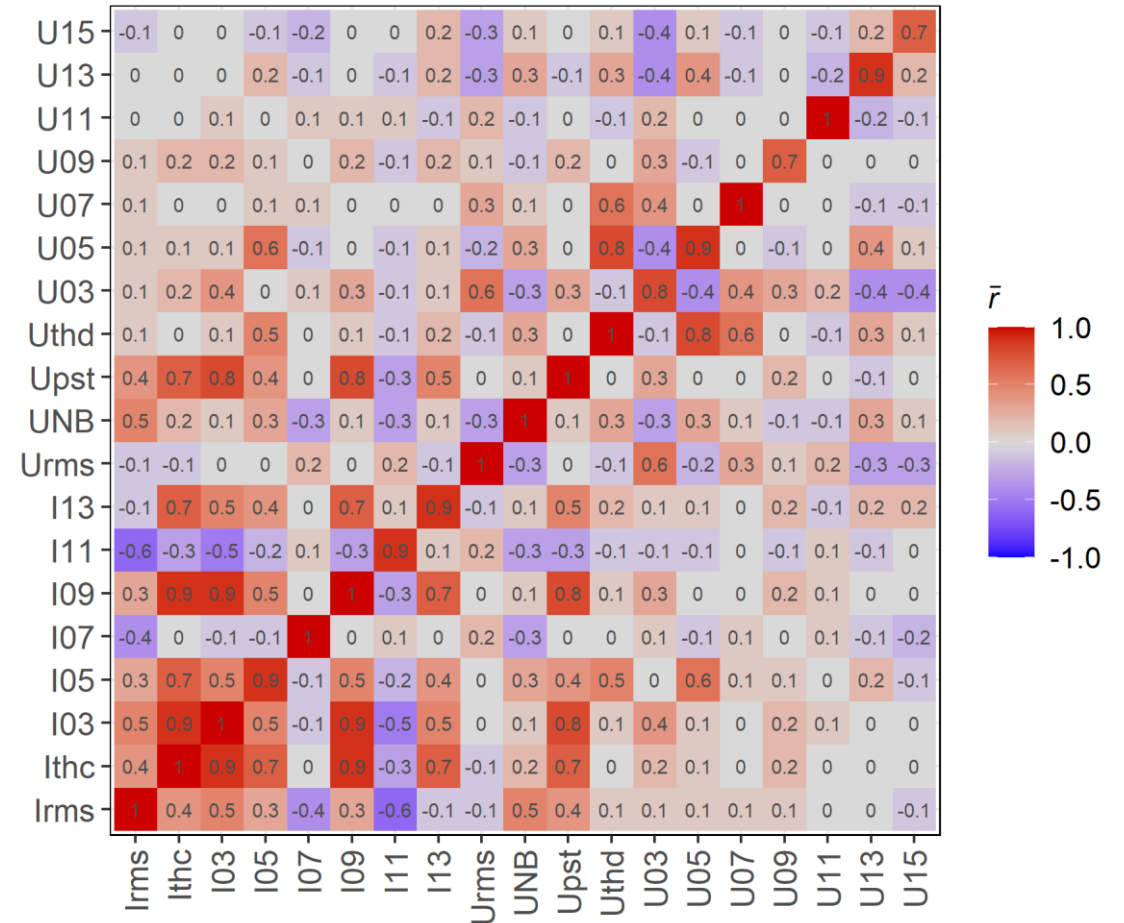
# Correlations between PQ parameters

## Results aggregation (1)

### Averaging of correlation coefficients

- Up to 9 phase combinations between two parameters (e.g. Ithc L1/L2/L3 ~ I05 L1/L2/L3)
  - Averaging reduces matrix by nearly a factor of 9 (from  $58^2 = 3.364$  to  $20^2 = 400$  elements)
  - Averaging using Fisher Z-transformation (otherwise too low averages due to value range  $[-1, 1]$ )
- Further aggregation of results necessary

*Correlations between PQ parameters for one measurement site in 380 kV*





# Correlations between PQ parameters

## Results aggregation (2)

### Aggregation of significant correlations

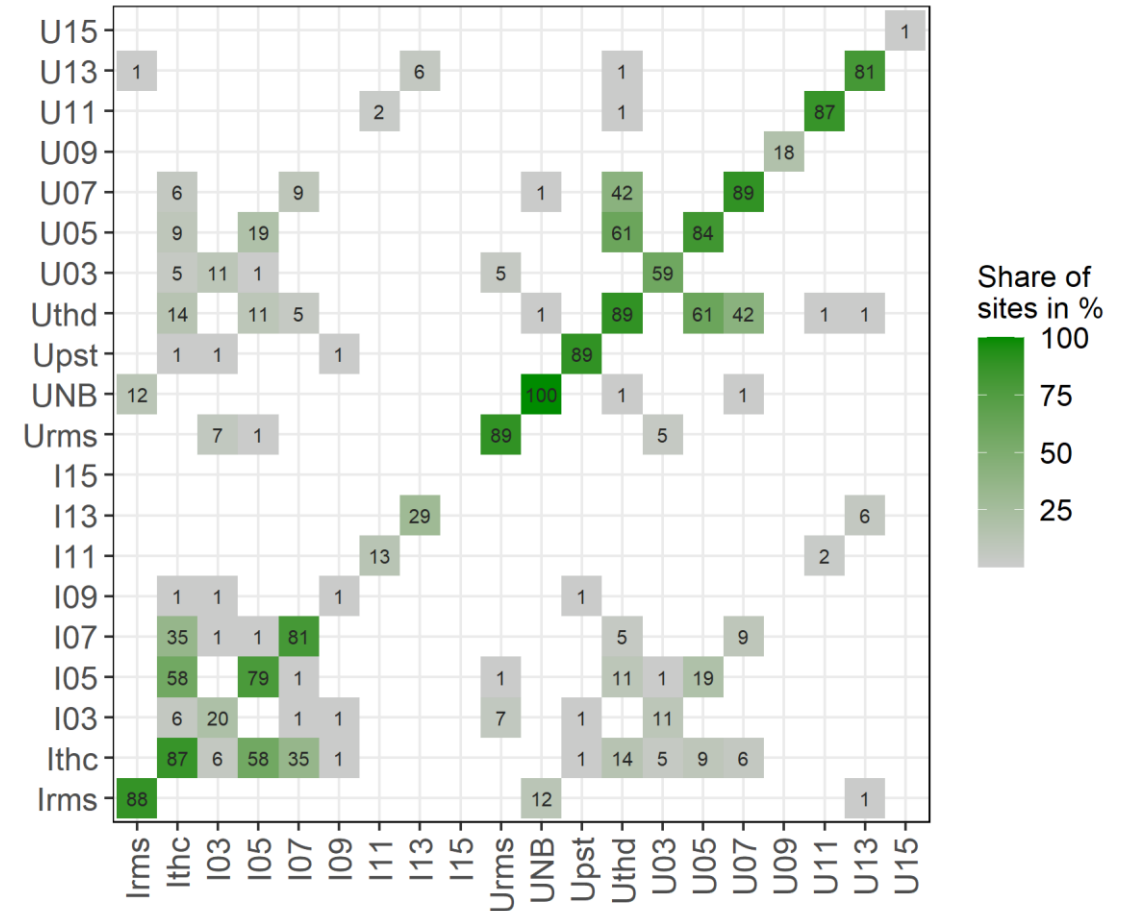
- Filtering of significant correlations (e.g.  $|r| \geq 0.7$ )
- Calculation of share across all matrices

### Visualization of matrices

- Very large matrices may be difficult to interpret
  - Various visualizations available for distance matrices
- Calculation of distance matrix from the matrix of significant correlation shares:

$$D_{ij} = \begin{cases} 100 - S_{ij}, & \text{for } i \neq j \\ 0, & \text{for } i = j \end{cases}$$

*Share of significant correlations  
between PQ parameters across all sites*



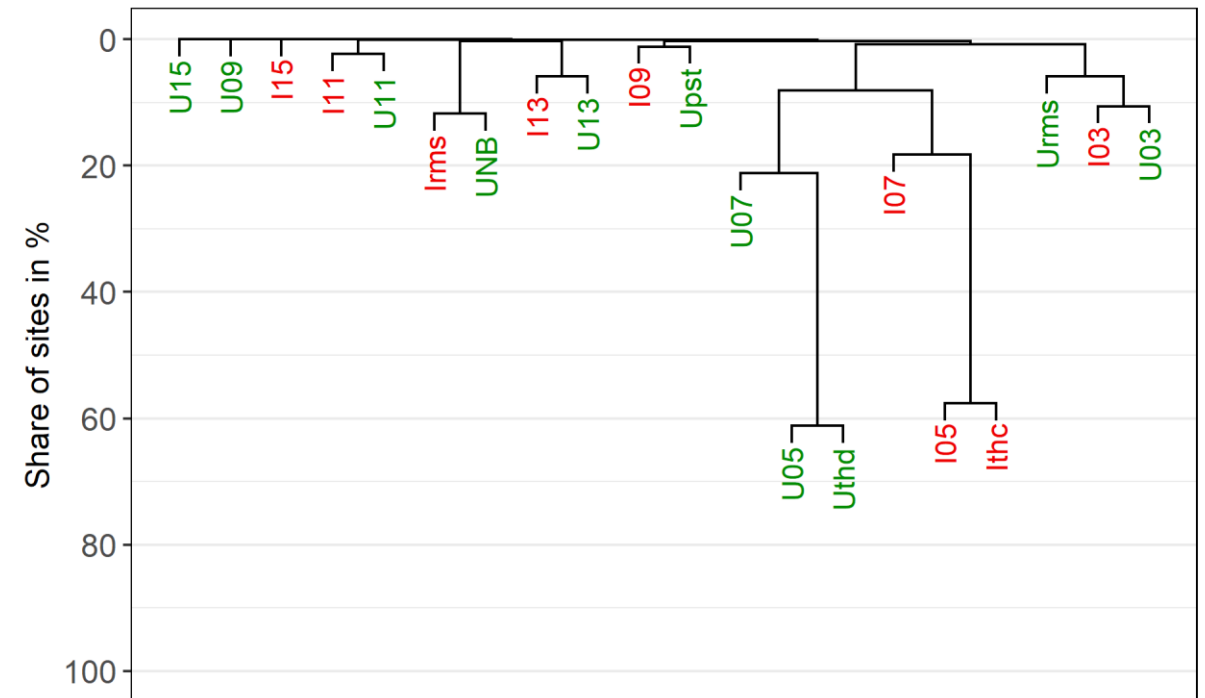
# Correlations between PQ parameters

## Visualization (1)

### Dendrogram of clustering

- Clusters merged step-by-step based on similarity/distance
- Tree-like diagram showing hierarchical relationships between elements
- Height of branch indicates the level of similarity

*Dendrogram of clustering based on the share of significant correlations*



# Correlations between PQ parameters

## Visualization (2)

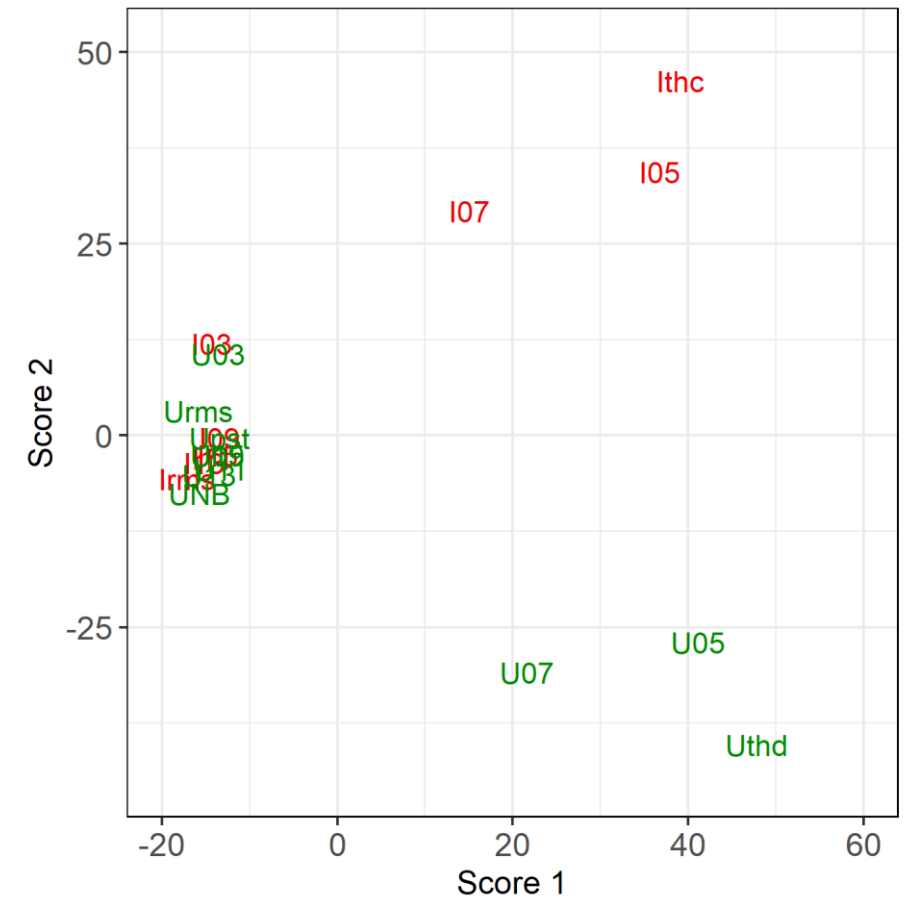
### Dendrogram of clustering

- Clusters merged step-by-step based on similarity/distance
- Tree-like diagram showing hierarchical relationships between elements
- Height of branch indicates the level of similarity

### Multidimensional scaling (MDS)

- Technique for dimension reduction
- Visualizes distances between elements in lower-dimensional space
- Preserves relative distances to reflect similarities

*2D scatterplot using MDS based on the share of significant correlations*

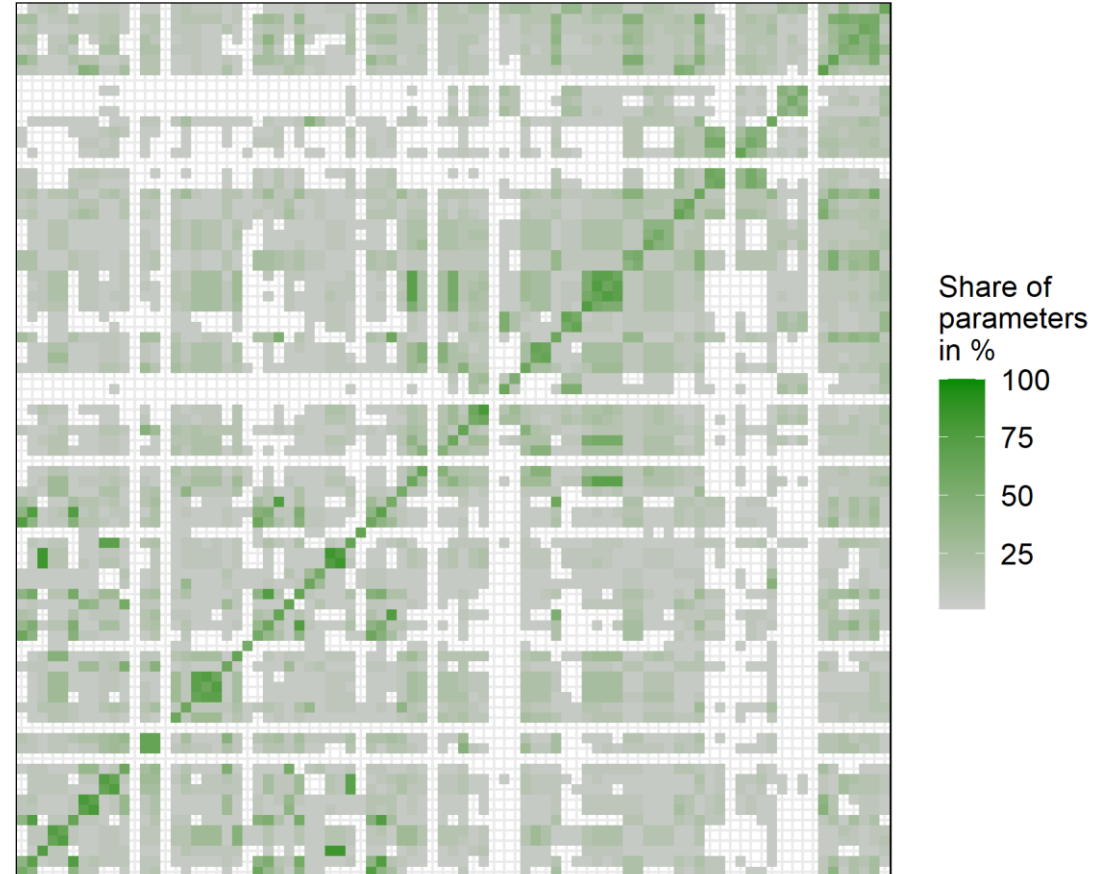


# Correlations between measurement sites

## Results aggregation

- Difficult to interpret holistically due to large number of individual elements
- Still, some observations:
  - Some sites with no correlations
  - Many sites with few correlations (ca.10 %)
  - Few sites with many correlations (up to 85 %)

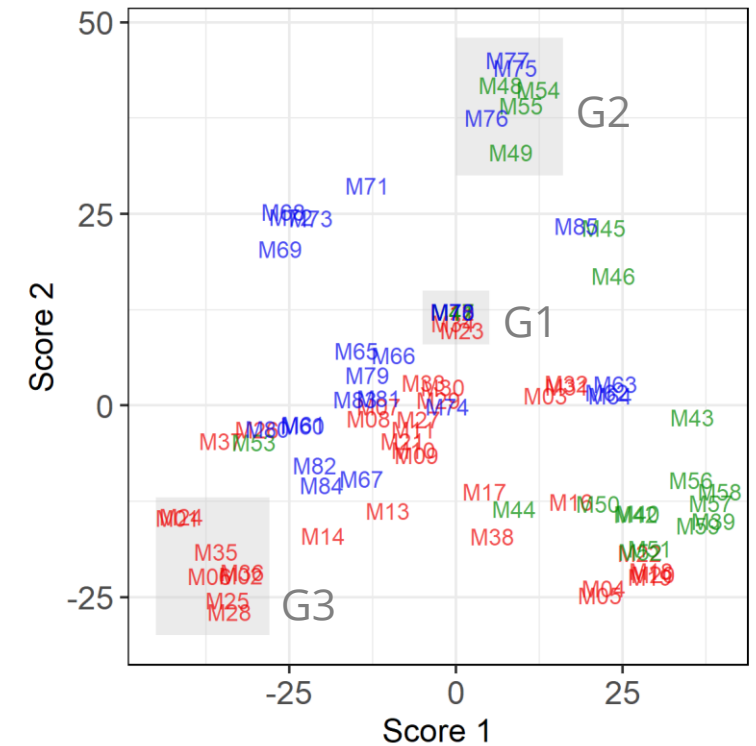
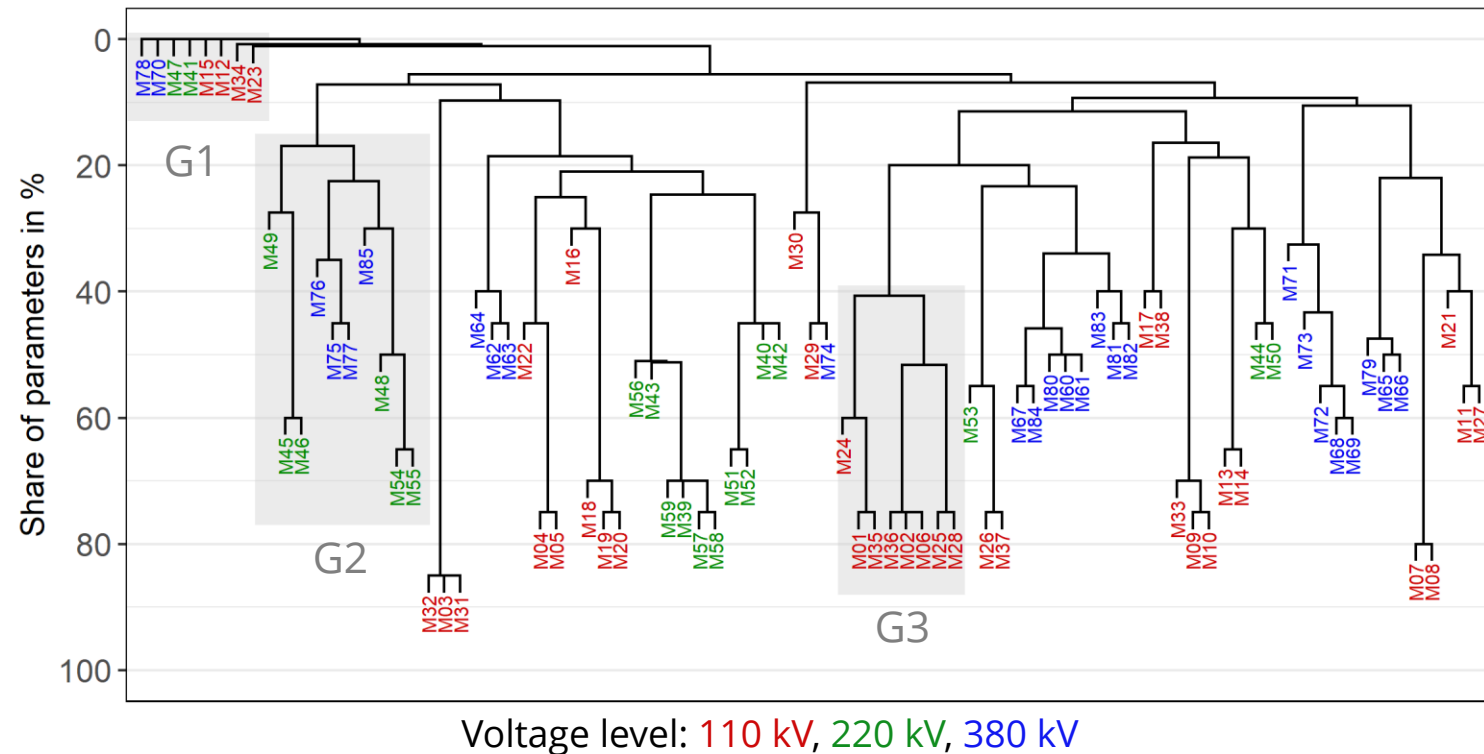
*Share of significant correlations  
between sites across all parameters*



# Correlations between measurement sites

## Visualization (1)

- Combination of dendrogram and 2D scatterplot allows to identify different groups
- G1 = Sites that do not or barely correlate
- G2 = EHV sites partially correlated (ca. 20% - 60%)
- G3 = HV sites highly correlated (ca. 40% - 80%)

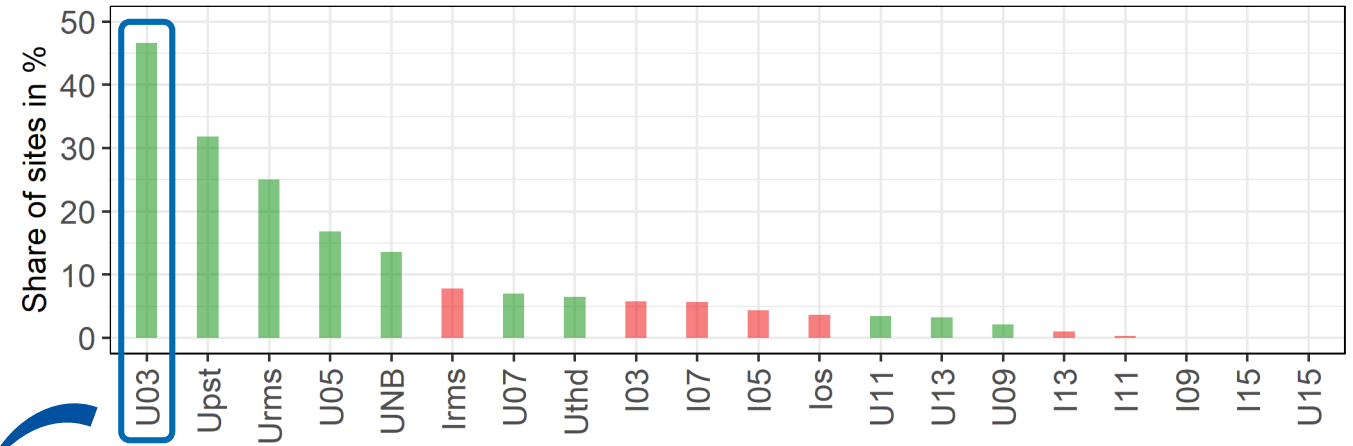


# Correlations between measurement sites

## Visualization (2)

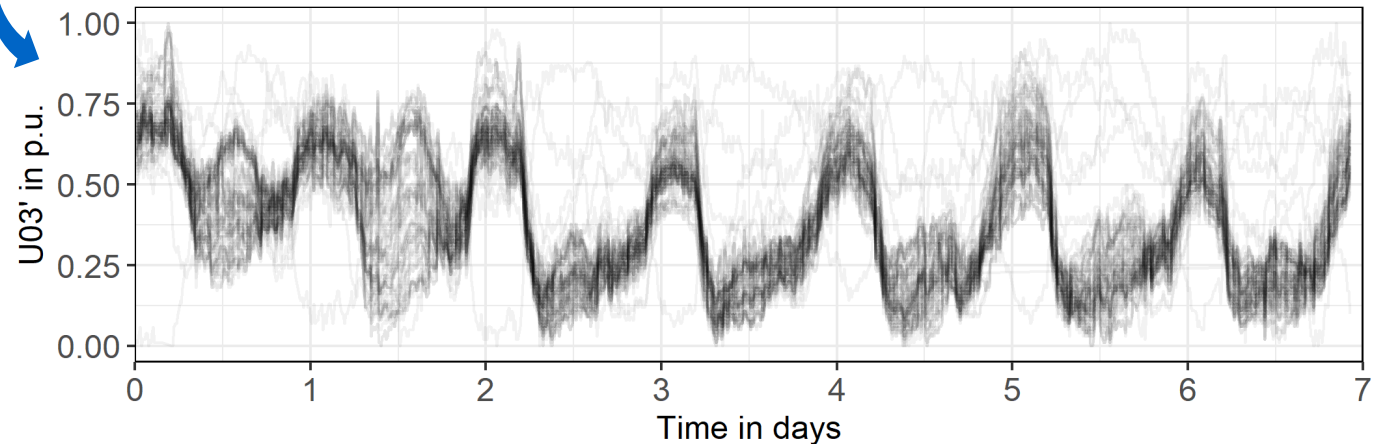
- Most correlations between sites for 1 or 2 parameters
- Highest share of correlations between all 85 sites for:
  - 3<sup>rd</sup> voltage harmonic (U03) and
  - Short-term flicker (Upst)

*Share of correlations between sites*



- 3<sup>rd</sup> voltage harmonic with distinctive daily pattern across all voltage levels
- Possible harmonic sources:
  - Converters
  - Transformer magnetization
  - Corona discharges

*Scaled time series for all 85 sites*



# Conclusions

## Power Quality data analysis

- Robust and reliable methods are essential for an automated application
- Results aggregation and visualizations are key for:
  - Extracting actionable insights and
  - Making data accessible to non-experts

## Future needs

- Advanced multivariate analysis methods for large-scale PQ monitoring campaigns (e.g. clustering approaches and dimension reduction)
- Automated data pre-processing:
  - Data validation (e.g. measurement errors, measurement uncertainty)
  - Data imputation (e.g. handling of missing values)

# Thank you for your attention!



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