

Diagnostic Visualisation of *Building Management System* Energy Data

Paul Shabajee, Dan Schien & Chris Preist

Dept. of Computer Science, University of Bristol

John Brenton & Chris Jones

University of Bristol Estates/Sustainability Office



Background



- Local Energy Systems management – campus scale
 - e.g. Overall energy & peak demand reduction and demand shifting/balancing with renewable generation and storage
- University of Bristol campus
 - Large estate of buildings, ranging from offices and student halls to state-of-the-art research labs and light industrial buildings, ...
- Large volumes of rich Building Management System (BMS) data
- Difficult for facilities/energy managers to make practical use of ...
 - High-level aggregated data (e.g. half hourly meter, aggregate BMS state)
 - Low level sensor or BMS ‘node’ level

Working with Facilities/Energy Managers

- Key Needs Identified
 - Obtaining an overview of energy use and energy using behaviours across various timescales, device types, site & building structural scales, etc.
 - Patterns and variations in behaviours of buildings and residents
 - Evaluation of existing and pilot energy management initiatives
 - Spotting anomalies, such as areas of high or low energy use that might require further investigation, sensor errors, etc.

BMS Data – Typical Example

- *Prefect* BMS over electrical heating systems in student halls
- BMS has 100s of ‘nodes’ (combined sensor/actuator)
 - Each node is sampled for state data approx. every 300 seconds
 - **Sensor data per sample:** node ID, date-time, temperature, thermostat set-point, power relay status (on-off), energy management program in operation, temperature manually adjusted, ...
 - **Metadata:** power rating of node device (Watts), type of node (water heater, corridor heater, room heater, etc.), building, building story, etc.
- Sense of “... drowning in data”

Prototype Analytical Interface

- Sensor and metadata combined

- Calculate/estimate energy use (relay time on * power rating)
- Choose interesting 'dimensions'
- Sum energy for dimensions
- Interactively filter data
- Open source libraries*

Number and % of data points selected

Energy use (kWh) of selection by

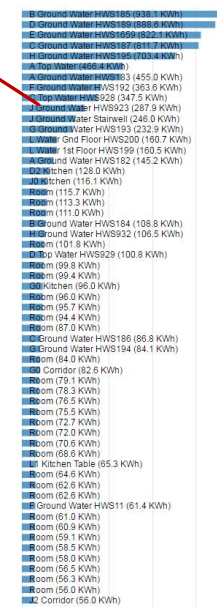
- operational 'program'
- node temperature
- node type

Energy use (kWh) of each 'node'* for given filters

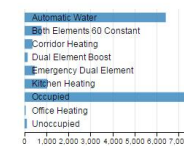
Analysis based on data from Prefect BMS

641,898 (100% of data points) selected out of 641,898 objects | [Reset All](#)

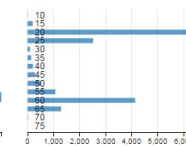
Space Name (Energy KWh)



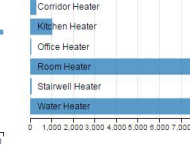
Program Name (Energy KWh)



Room/Node Temperature (Energy KWh)



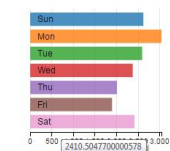
NodeType (Energy KWh)



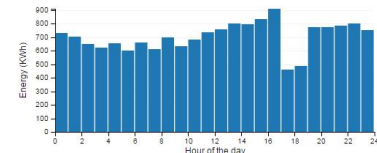
Date (Energy KWh)



Day (Energy KWh)



Hour (Energy KWh)



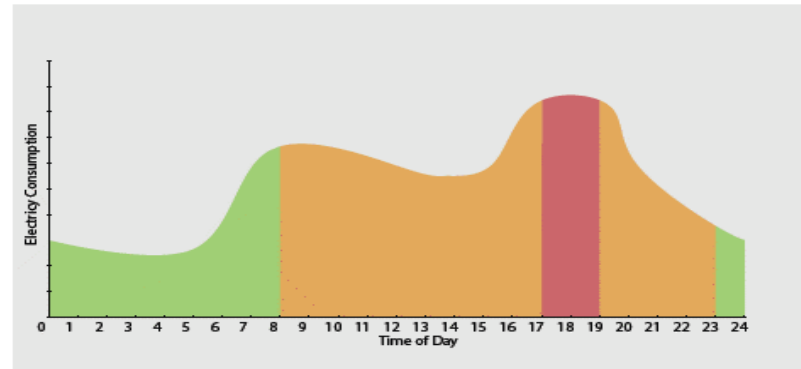
Energy use (kWh) of selected data by:

- Date
- Day of the week
- Hour of the day

Demo

- IODiCUS → Evaluate effectiveness of BMS energy use reduction programmes to:

‘avoid red-DUoS (peak) electricity cost charges between 5-7pm on weekdays’



Enhancements - examples

- Configure for other 'dimensions' e.g. temperature adjusted by resident, building structure (building, storey), room orientation (study solar gain), ...
- Other core metrics kWh → count of samples, £ and kgCO₂(e)
- Other types of chart (d3.js javascript library very rich in options)
- Other kinds of linked interactive 'drill down' and analytical interfaces

Limitations – examples

- Slow to load and sluggish on date-time dimensions with large datasets > 500k rows
 - 700,000 to 1,400,000 samples practical limit
 - Means one week whole site *or* months for a block, ...
 - Possible solutions
 - aggregating data-points, dynamic pre-filtering of data, exploring ways to support more data points, ...
 - BMS sensor and configuration errors – leads to under and over estimation of energy use (values indicative only)
 - Uses BMS state data but some BMSs supply event (state change) data
-

Questions?

Thank you