

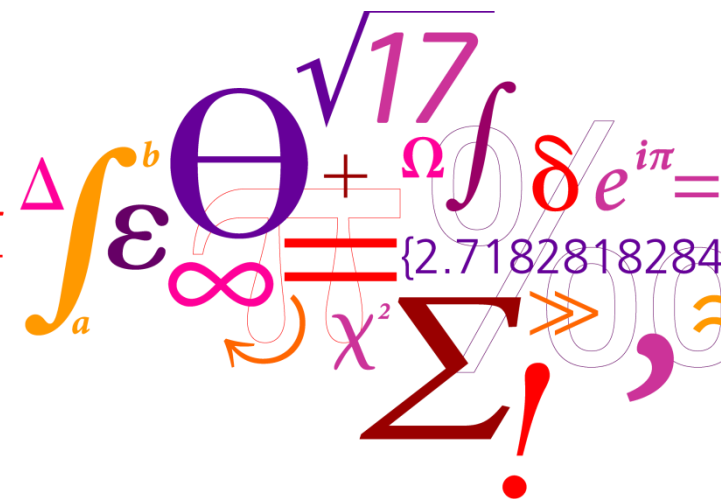
WP3: Intelligent Energy System Integration

Panagiota Gianniou
PhD student

Supervisors:

Prof. Carsten Rode
Sen. Res. Per Sieverts Nielsen
Assoc. Prof. Alfred Heller

$$\frac{\partial T}{\partial t} = \frac{\lambda}{\rho c_p} \frac{\partial^2 T}{\partial x^2}$$



Agenda

- Introduction

- Theoretical Background
- Problem Statement
- Aggregating Building Energy Demands
- Main Objective
- Implementation of City Energy Model
- Results-Conclusions

- Next steps-Future work
- CITIES project-structure
- Similar projects worldwide



Introduction

- **Diploma** in Mechanical Engineering, Aristotle University of Thessaloniki, Greece (Energy Production and Utilization specialization)
- **MSc** in Sustainable Energy, DTU, Denmark (Energy Savings study line)
- **MSc Thesis:** 'Development of Aggregation Models for Building Energy Demands applied to Smart Cities', DTU Byg
- **PhD** project (September 2014): 'Buildings for Smart Energy Cities', DTU Byg

MSc Project

Theoretical Background

➤ Building stock energy modelling approaches

- Bottom-up → Calculate energy demand based on **individual** buildings and **extrapolate** the results to a whole region/country → data intensive
- Top-down → Refer to **aggregate** level, while typically represent a historical time series of national energy consumption data → Require up-to-date well-distributed data for **large** groups of buildings

➤ Parametric modelling

- Creates a plethora of outcomes and design alternatives in a **dynamic** manner
- Inputs/**parameters** control the design
- **Automates** city modelling and helps generate bottom-up urban structures

➤ Load modelling of buildings

- ✓ Building energy simulation methods
- ✓ Statistical analyses
- ✓ Intelligent computer systems

Problem Statement

- ❖ **How can buildings contribute to the development of Smart Cities?**
 - Integration of smart energy solutions
 - Increase of building-energy efficiency → high energy & CO₂ savings
- Energy demand of building sector plays an important role on national energy balances



- Many efforts to estimate energy demand of neighborhoods, districts or cities



Aggregation



Aggregating building energy demands

❖ First aggregation way

- Energy estimates of **individual** buildings are added up to calculate the total energy use of the building stock

$$Y = \sum_{i=1}^n X(i)$$

Y = total energy demand of the examined building stock [kWh]

n = number of individual buildings

X = energy demand per building [kWh]

❖ Second aggregation way

- **Reference** buildings are used → representative for the whole stock and weighting factors are used proportionally for every category

$$Y = \sum_{j=1}^N EUI(j) A(j)$$

EUI = energy demand per floor area [kWh/m²] for each building type

A = total floor area [m²] of all buildings included in the respective type

Main objective

- Investigate the existing aggregation ways by **implementing** them on a real case-study
- Propose a **methodology** for estimating realistic energy demand models for districts or cities



Implementation of city energy model

➤ Description of the case study

- **16** one-floor **single-family houses**
- Located in Sønderborg, Denmark
- Constructed mainly in 1960s
- Floor areas: 85 - 175 m²
- Connected to local district heating network
- No solar heating panels
- No mechanical cooling



Figure 1. Typical design of the single-family house



Implementation of city energy model

➤ Sources of information

- Danish Building Register (BBR)
- TABULA project (WebTool)
- Google Maps-StreetView
- Questionnaires
- Measurements


➤ Simulation tool

- Termite
- Developed by PhD student **Kristoffer Negendahl**, DTU Civil Eng.
- Based on **Rhinceros®** (CAD design environment) and **Grasshopper™** (visual parametric programming interface)
- Uses **Be10** for energy simulations



Implementation of city energy model

➤ Information levels

- A. Simple typological data (TABULA)
 - B. Online public databases (BBR)
 - C. Google Maps, Street View
 - D. Personal visits, on site measurements
or distributed questionnaires to the occupants
- 
 Number of occupants
 Energy refurbishments

- 6 different scenarios were created based on these information levels

Information level	Scenario 1	Scenario 2	Scenario 3	Scenario 4	Scenario 5	Scenario 6	
Scenario 1 A	BBR						Design based on generic means
Scenario 2 B	BBR & Google Maps/Street View	X	X	X	X	X	
Scenario 3 C	BBR & Refurbishment	X	X	X	X	X	
Scenario 4 Di	BBR & Google Maps/Street View & Occup. Loads	X	X	X	X	X	
Scenario 5 Dii	BBR & People Loads & Refurbishments	X	X	X	X	X	
Scenario 6	BBR & Google Maps/Street View & Occup. Loads & Refurbishments	X	X	X	X	X	

Implementation of city energy model

➤ Information levels

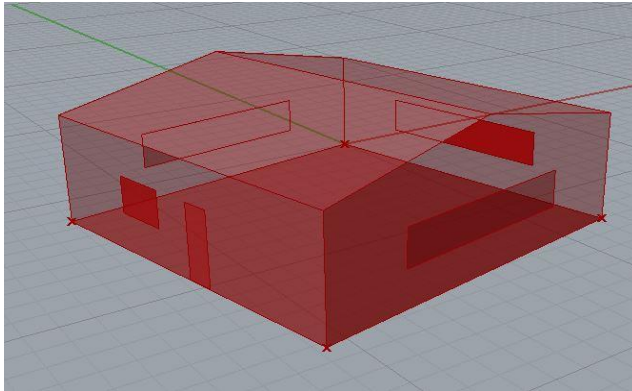


Figure 6. Design of House no.1 based on Scenario 1

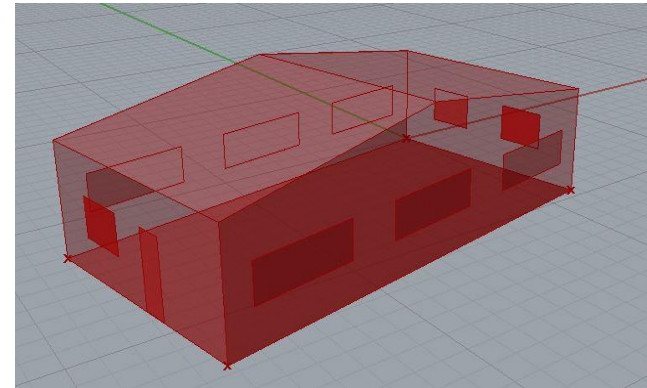


Figure 7. Design of House no.1 based on Scenario 6

- Initially aggregation of building energy demands was made based on the individual summing up (first way)

☐ **Aggregation based on reference buildings**

	Building type	Construction period	No. of incl. buildings	Total floor area [m ²]
Example buildings →	A	1931-1950	2	238
	B	1951-1960	2	180
	C	1961-1972	10	1,530
	D	1973-1978	1	117
	E	1979-1998	1	122

Results - Conclusions

➤ First aggregation way

- Knowing the energy **refurbishment** state → improved the model's results by **24%** compared to the model including information only from BBR database
- Benefits of energy renovations **depend** highly on the building's age → usual renovation measures can have impressive reductions of energy demand in **older** buildings
- Most representative: model with **highest** information level (Scenario 6) → 12% deviation from measured values (annual aggregate heat demand)

➤ Second aggregation way

- Example buildings represented quite well the **individual** houses' performance, but not so well the **real** heat demands as measured
- Error increased when the houses had undergone energy **renovations**
- Example buildings did not include improved U-values → different classification **differentiating** between renovated and not renovated buildings may be more effective

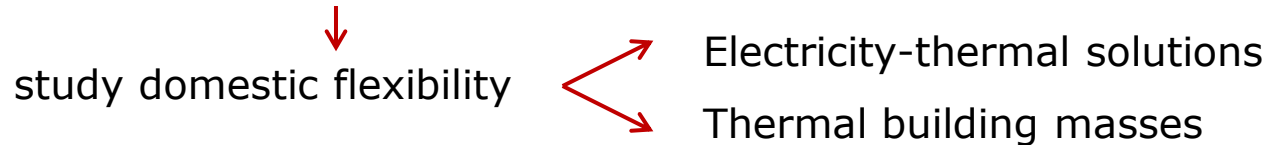
Proposal: Include information about current energy **refurbishment** state of buildings in most national estate databases

PhD project

Next steps – Future work

- Expand the sample to a **larger** building population → collect case studies

- Implement a **dynamic** energy simulation tool



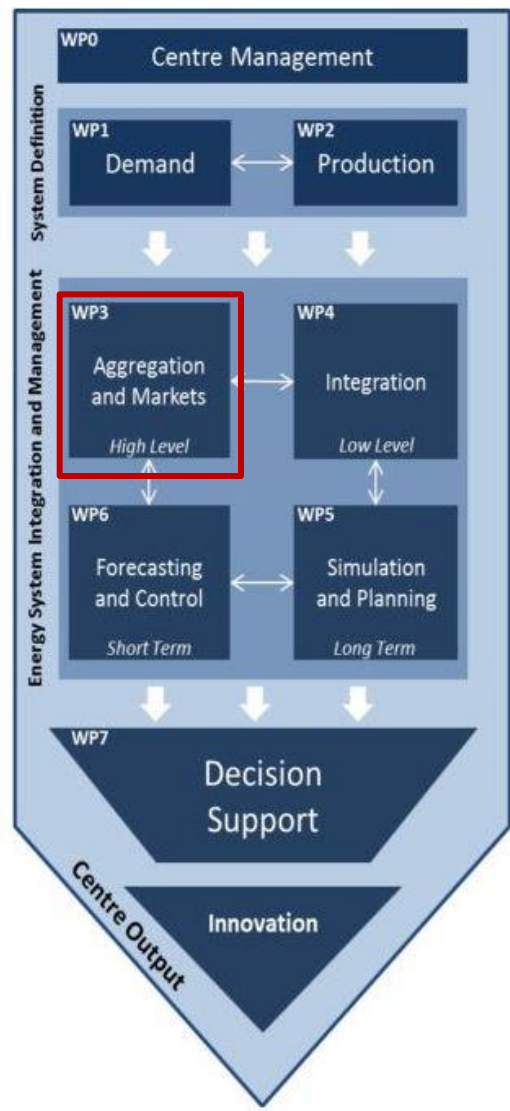
- **Challenges** to be met at **national** scale:

- ✓ Data availability: critical
- ✓ **Parametric** modelling tools
- ✓ Vast computation times → use '**autotuning**' methods
- ✓ Combine building energy simulation tools with **statistical** models (regression analysis)
- ✓ City-scale models: **less** homogeneous → use 'system-types' categorization




CITIES project - structure

WP3 →



Work Package 3

- **WP3.1 :**
 - ✓ Investigate novel methods for **aggregate** modelling and simulation techniques
 - ✓ Address any **interoperability** issues between different energy modelling and optimisation tools
 - ✓ Investigate the capabilities of individual **tools** for modelling energy systems with multiple energy flows.

 - **WP3.2:**
 - ✓ Study low level aggregation techniques
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- Collect evidence and experiences from real world cases of **demand modelling and demand flexibility** demonstration projects and state of the art literature, methods and tools for aggregation models.

Similar projects around the world

- **IssyGrid**: the 1st smart district energy network in France
- **Kalasadama**: a model district of smart urban development with the help of open data in Helsinki
- **Celsius**: Combined Efficient Large Scale Integrated Urban Systems (smart district heating & cooling) in Gothenburg
- **FEED&D**: Future Energy Efficient Buildings & Districts project by ETH Zurich
- **Zernez** (Switzerland): independent of fossil fuels by 2020



Thank You!

Panagiota Gianniou

PhD student

DTU Civil Engineering Department

pagian@byg.dtu.dk