

# HEAT

*Sustainable Energy Futures, Annual Conference 2016*

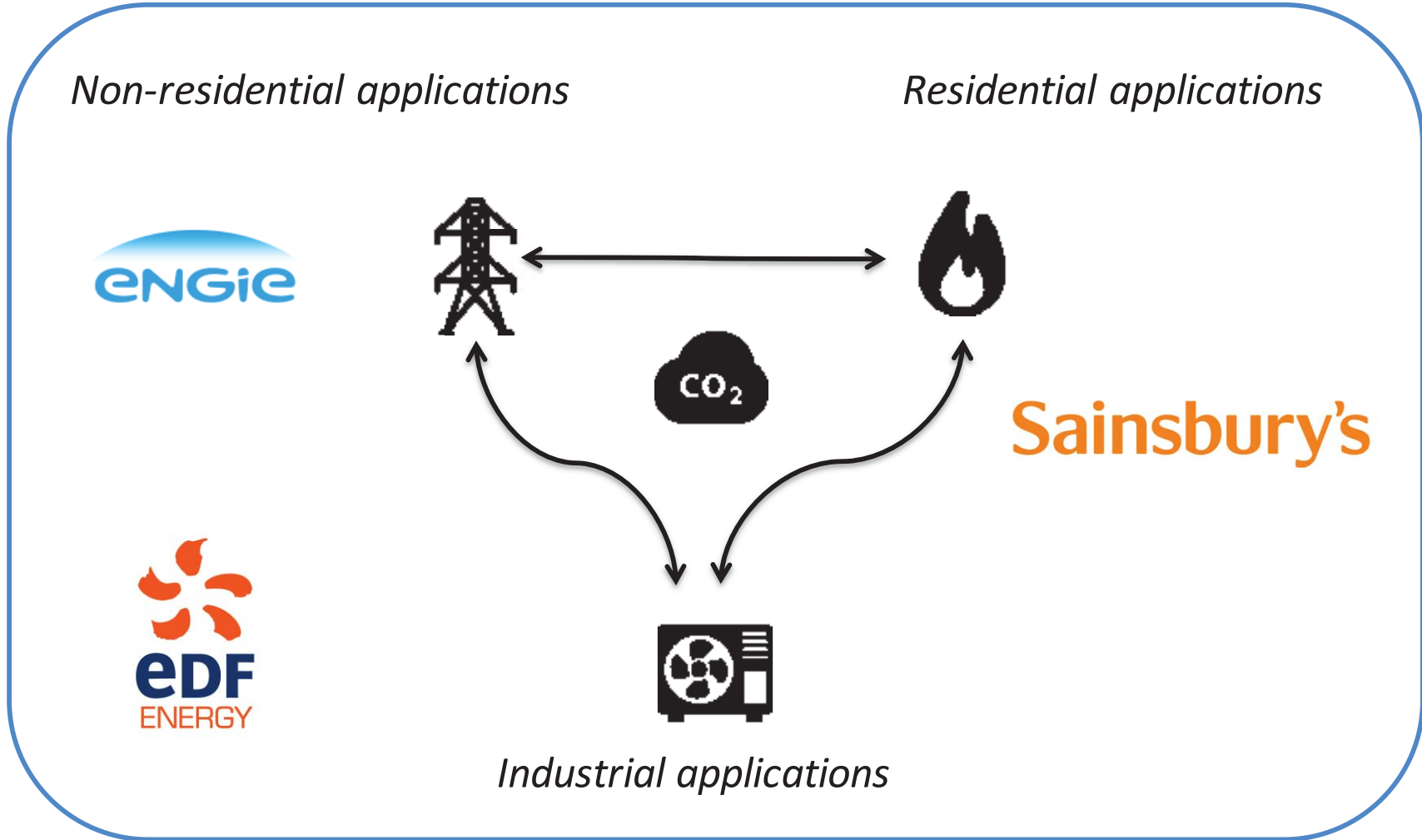
Maria Briola  
Axelle Delangle  
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Aspasia Georgakopoulou  
Panagiotis Ladas  
Adrian Regueira-Lopez

**Group 7**

**Poster No: 09-14**

# Introduction





## Simulating Heat and Electricity Demand in Urban Areas

Panagiotis LADAS



**Supervisors:**

Dr. Koen H. van Dam  
Gonzalo Bustos-Turu, PhD  
Dr. Salvador Acha

# Aim & Research Motivation



## Background

- 54% of the world's population currently lives in urban areas
- Cities account for around 75% of global energy demand

**Challenge:** Understanding of the dynamics behind the energy demand in urban areas



*City of London*

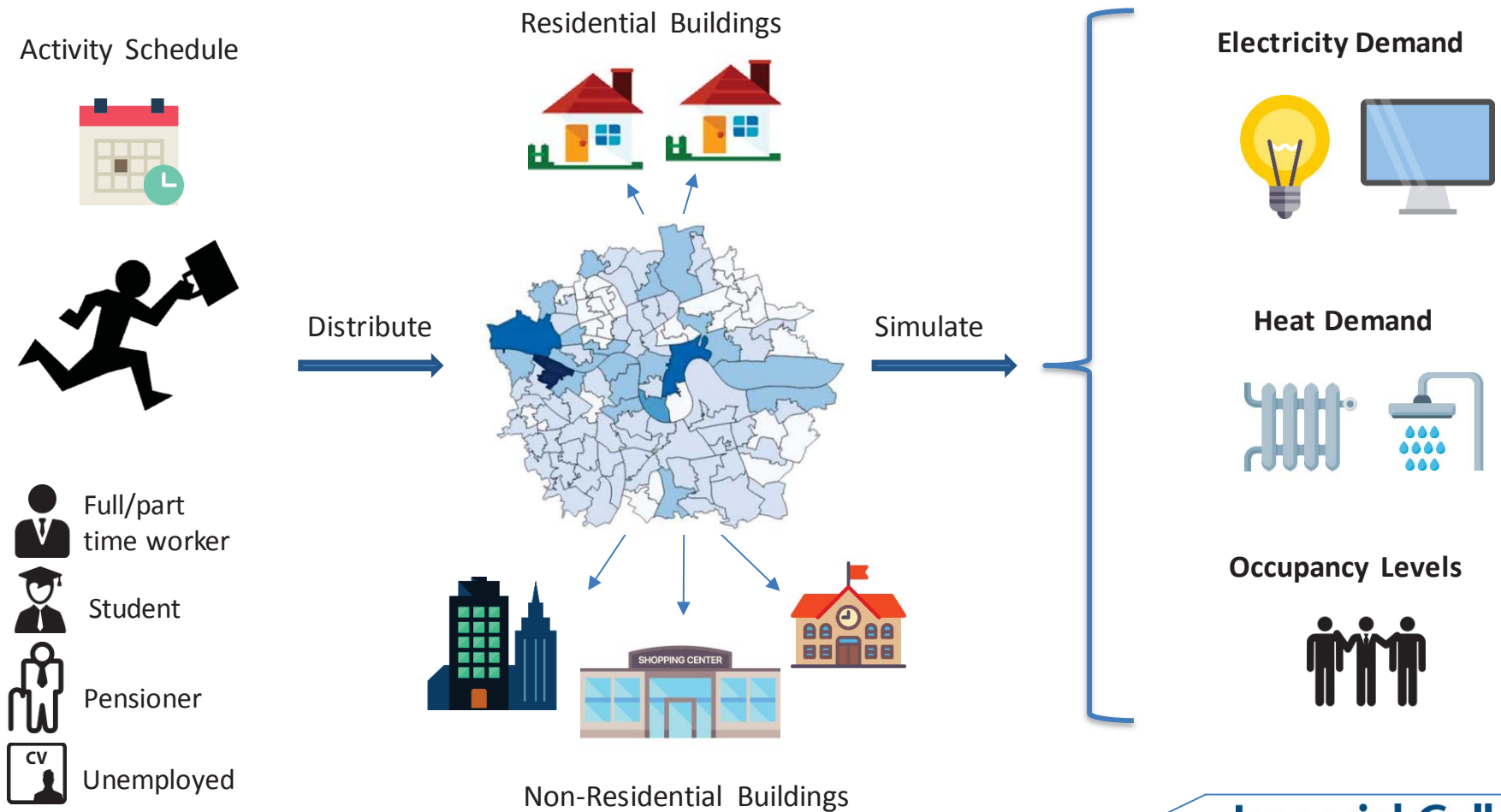
**Aim:** Simulate the spatial and temporal energy loads in

- Residential buildings
- Non-residential buildings

# Methodology



## Agent-Based Model (ABM) following a bottom-up approach



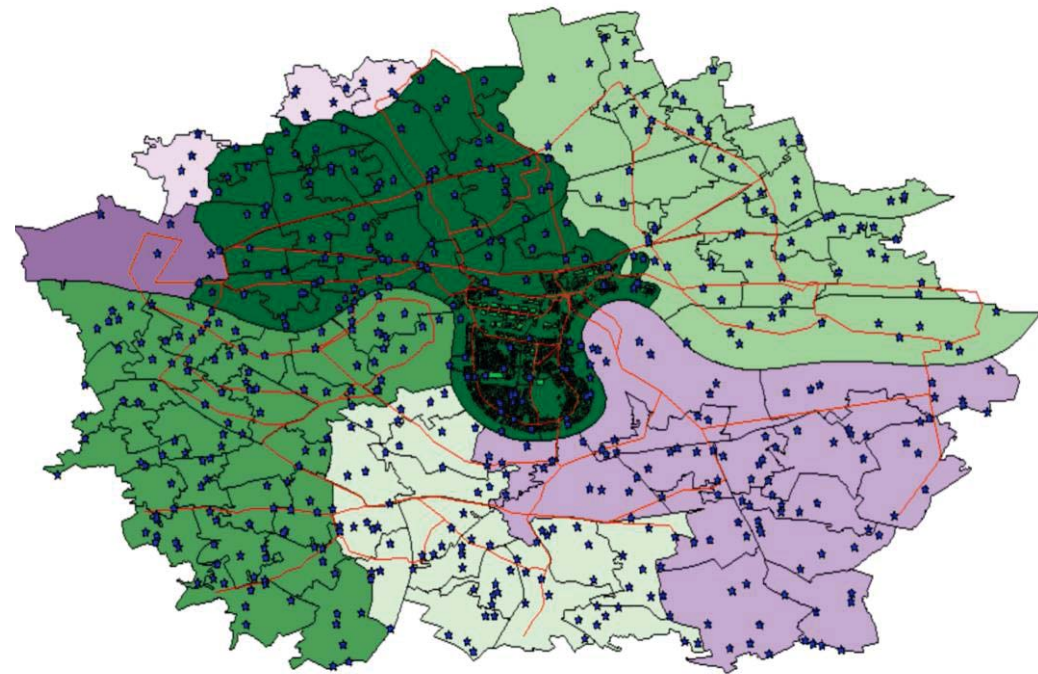
# Case Study



## Isle of Dogs area (East London)



- London Borough of Tower Hamlets
- 42,000 residents
- 93,000 people work in the area

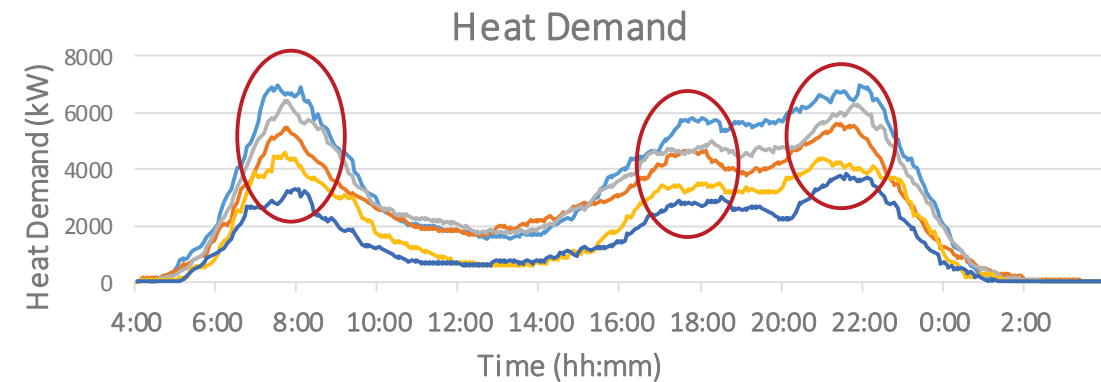
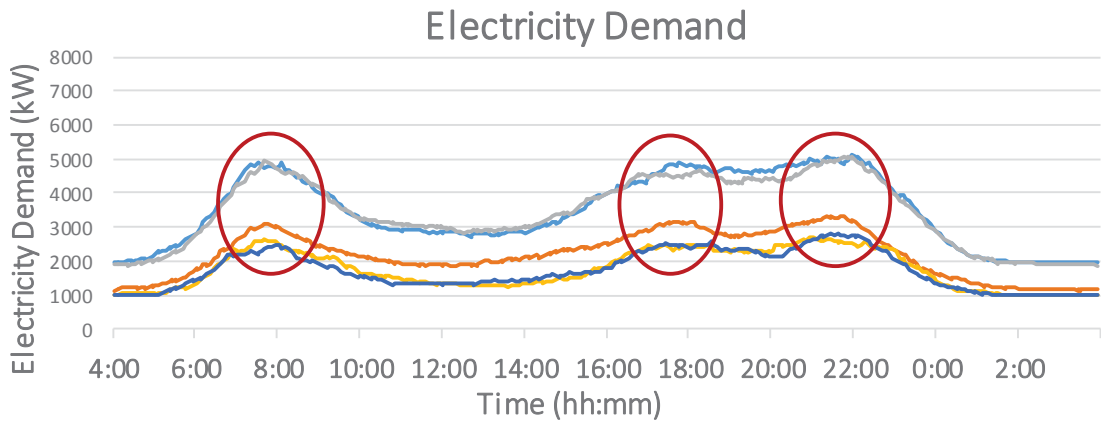


*Simulation of the greater area of London from SmartCity Model*

# Results: Daily Profiles Residential Buildings

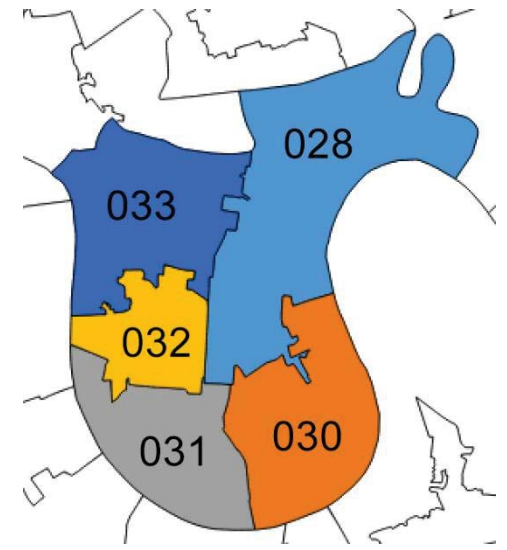


## Winter - Weekday



— Tower Hamlets 028 — Tower Hamlets 030 — Tower Hamlets 031  
— Tower Hamlets 032 — Tower Hamlets 033

## Tower Hamlets



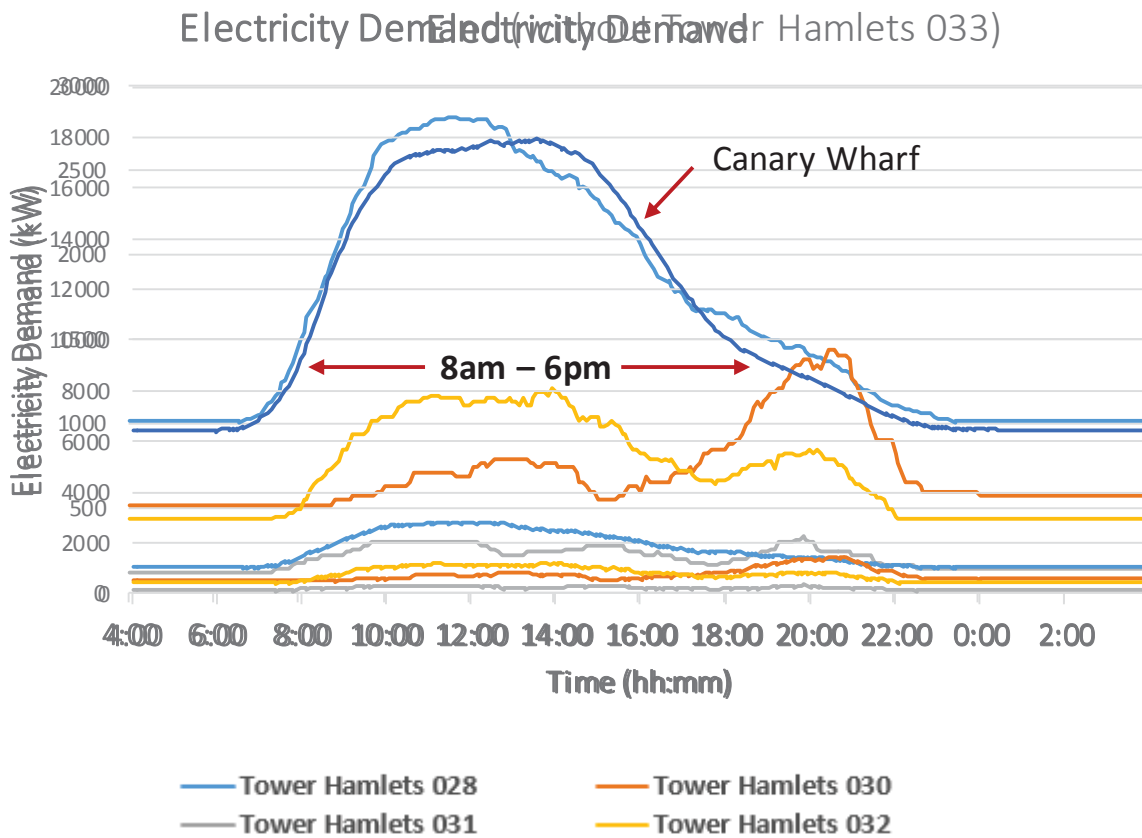
### Peak Demand

- **7-8am**: wake-up
- **5-6pm**: back from work/school
- **10pm**: all agents at residences

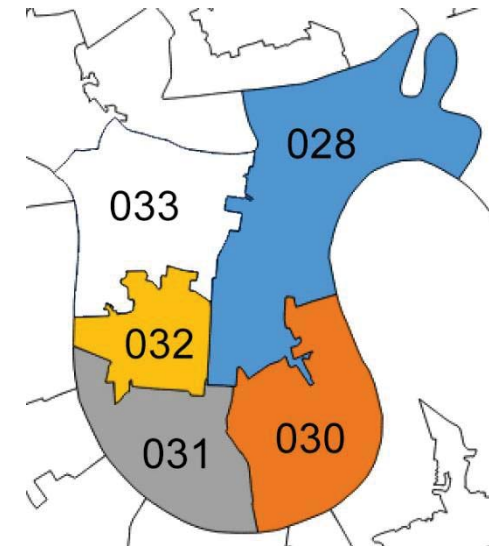
# Results: Daily Profiles Non-Residential Buildings



## Winter - Weekday



## Tower Hamlets



Peak Demand  
8am-6pm: working hours

- Schools
- Offices
- Commercial Stores

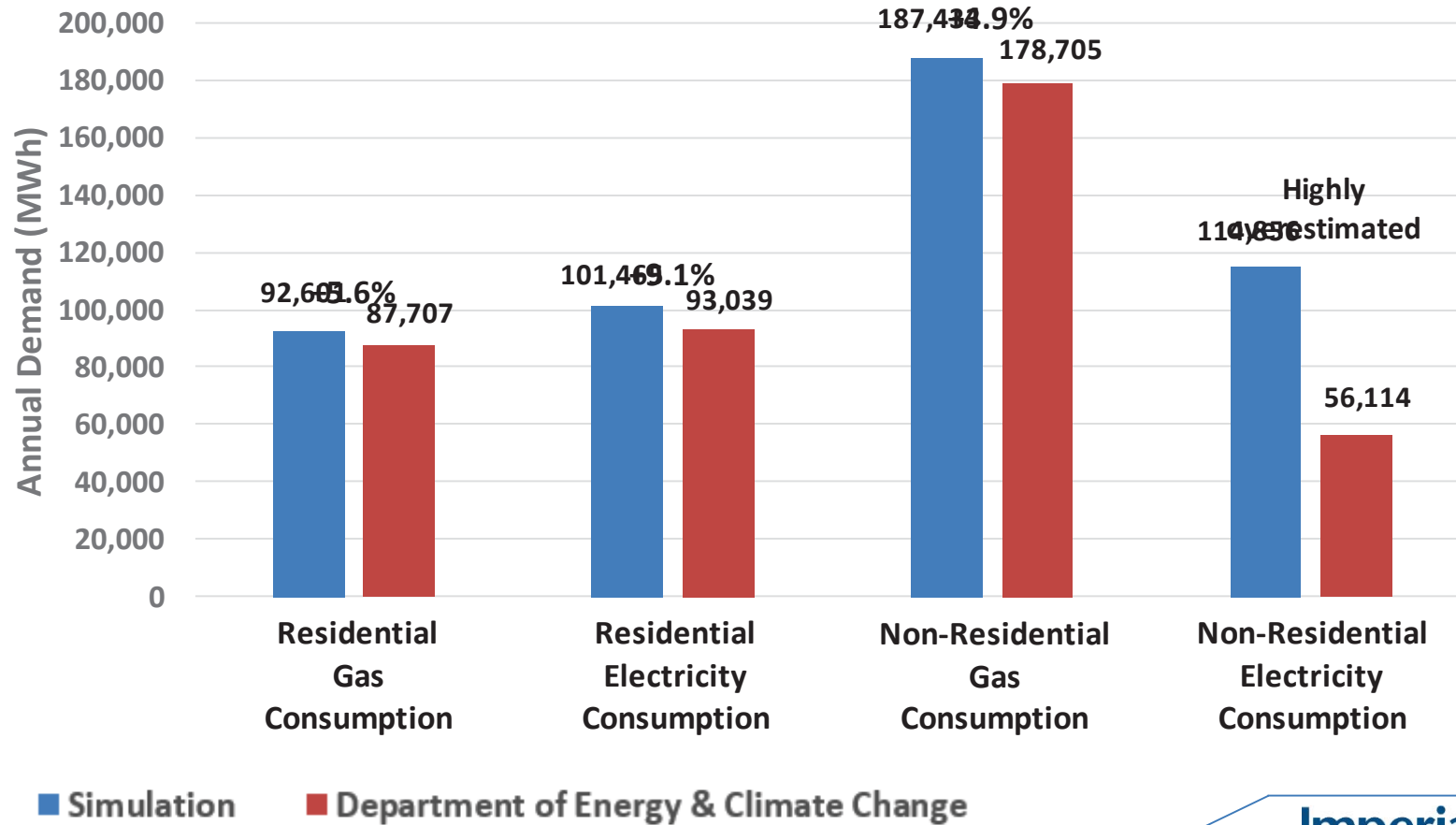


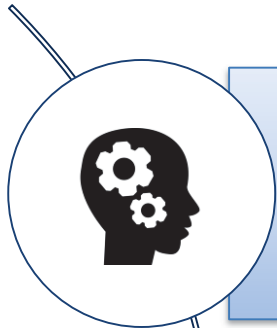
# Results: Annual Demand



## Validation

### Annual Demand VS DECC data





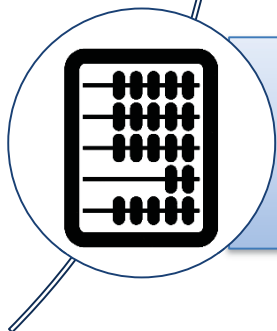
## Process

- Reproduces trends of real demand
- Test different scenarios
- Re-usable



## Non-Residential Demand

- Lack of adequate data
- Calibration of the results



## Computational Tool

- Decision support tool for key stakeholders
- Incorporates end-user behaviour



## Agent-Based Modelling of Residential Heat Demand in a District Heating Network

Maria BRIOLA



**Supervisors:**

Dr. Koen H. van Dam

Dr. Christoph Mazur



## District Heating Networks (DHN)

- Heat for an area is produced centrally
- Heat is distributed through pipes

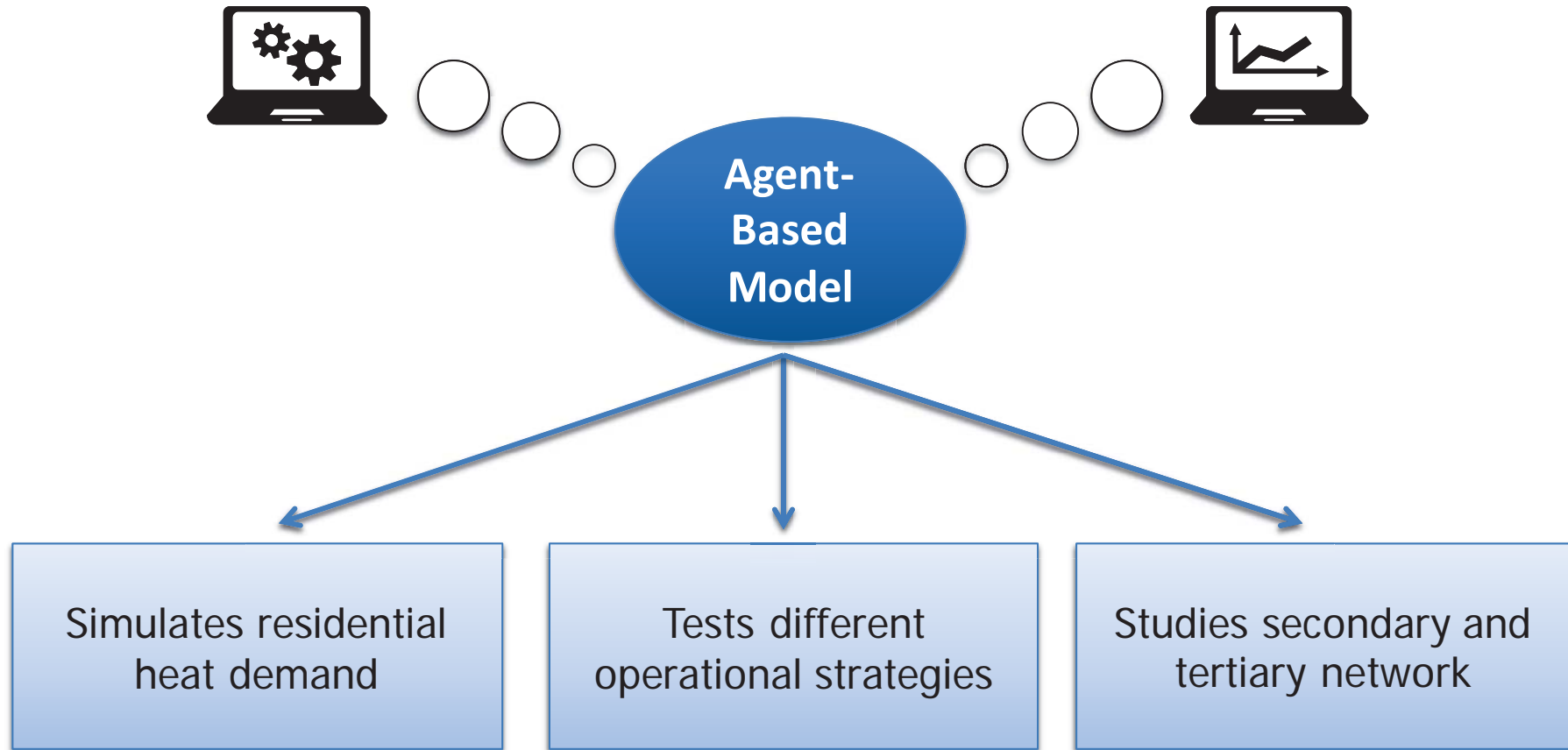
## Challenges of DHN

- Demand side barriers
- Residents' mistakes when using thermostats
- Faults in secondary consumer systems



## Aim

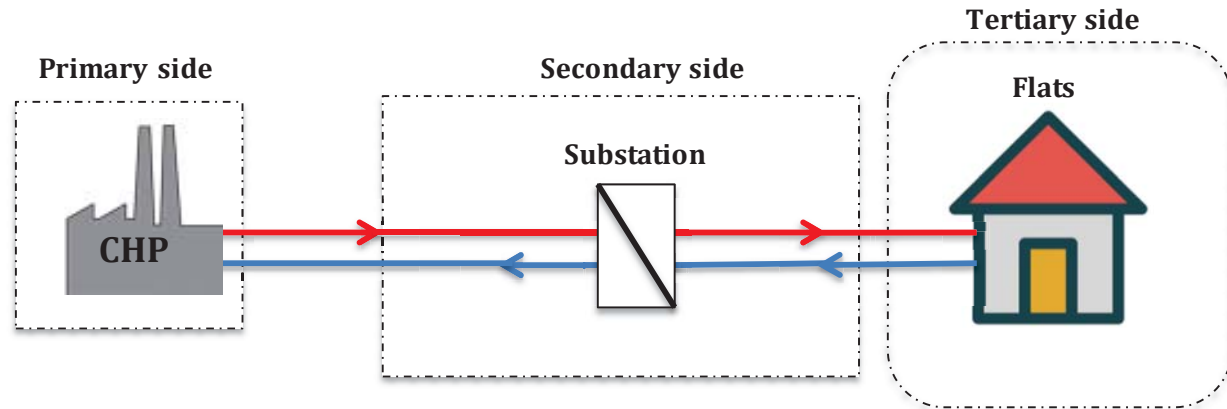
Investigate different strategies that could be applied to achieve optimal operation and maximum efficiency of DHNs.



# Layout of the DHN



Layout of the DH block examined in this model



## Tertiary or demand side of DHN

- Under-floor space heating



- Domestic Hot Water system





## Queen Elizabeth Olympic Park



## East Village (residential area)



### Problem with the DHN

- High return temperatures
- High heat losses

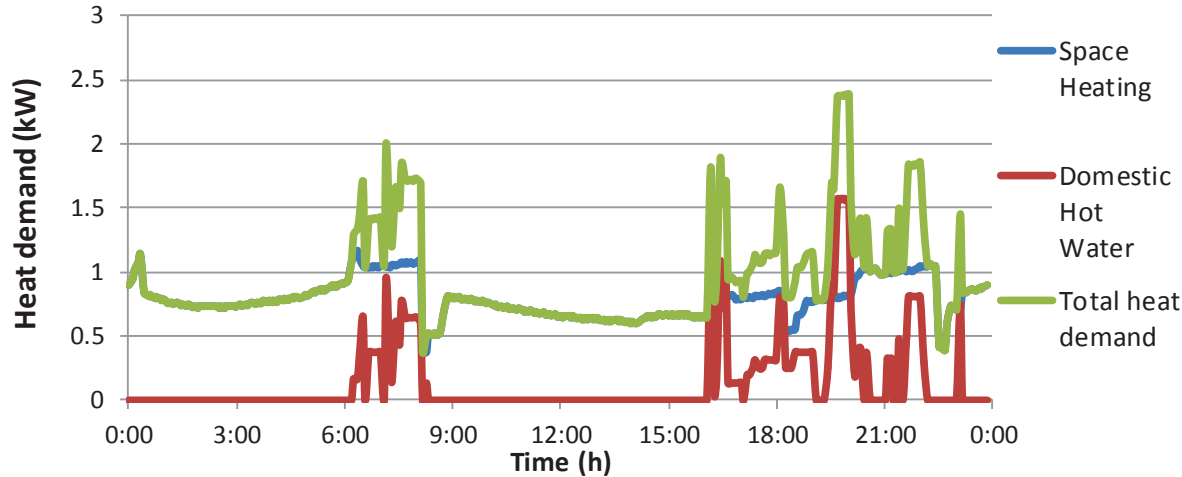
### Solution

- Apply the ABM
- Test three scenarios
  - Scenario 1: Tank resizing scenario
  - Scenario 2: DSM scenario (improved and ideal case)
  - Scenario 3: Alteration of boosting periods scenario

# Baseline scenario results



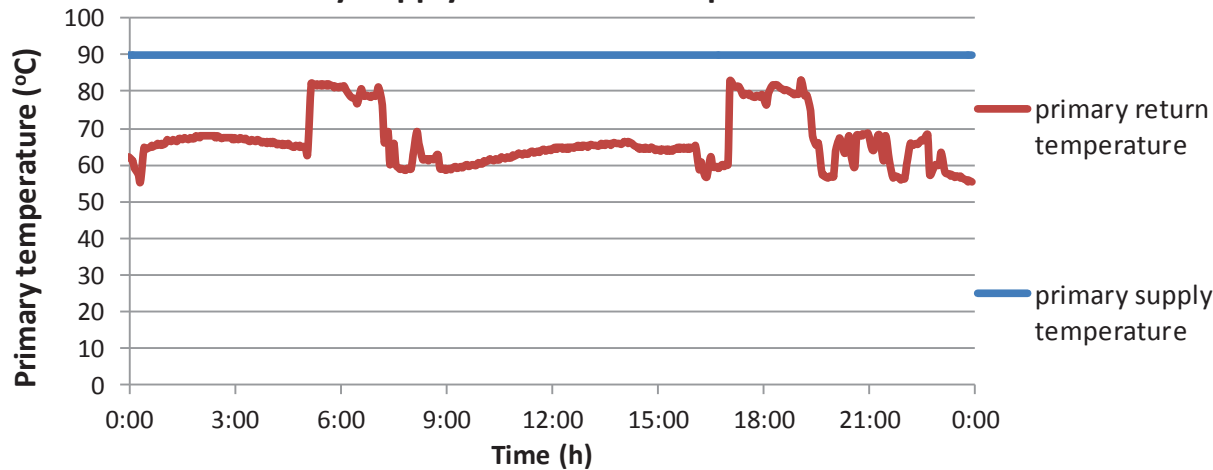
### Residential heat demand



**Morning peaks:**  
residents prepare for work

**Evening peaks:**  
residents return from work

### Primary supply and return temperature



Average difference  
between primary supply  
and return temperature:  
**25°C**



# Comparison of scenarios



	Change in heat losses	Change in difference between supply and return temperature	Change in fuel consumption	Change in primary return temperature
Scenario 1	-0.7%	+0.96%	-0.07%	-0.32%
Scenario 2 (improved case)	-3.16%	+3.3%	-0.76%	-1.3%
Scenario 2 (ideal case)	-1.8%	+13%	-1.38%	-4.6%
Scenario 3	-12.3%	-	-2.7%	-

- **Scenario 1:** lowest improvement
- **Scenario 2:** lowest return temperatures and higher difference between supply and return temperatures
- **Scenario 3** lowest fuel consumption and heat losses

# Conclusions



## Case study

- Combination of scenario 2 and 3 could be the best solution



## Stakeholders

- Effective collaboration must be achieved



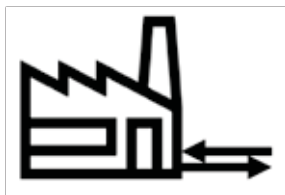
## Model

- Could be used as a decision support tool for DHN operators



## Modelling and optimisation of a district energy centre

**Adrian REGUEIRA-LOPEZ**



**Supervisors:**

Prof. Nilay Shah

Dr. Romain Lambert

# Overview



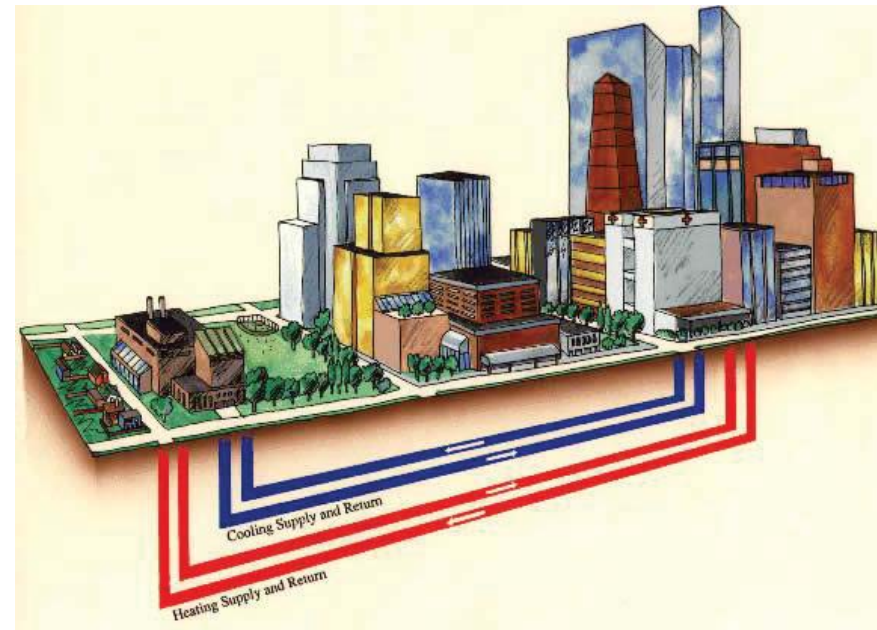
## District energy schemes:

- Centralized energy production
- Economies of scale

## Energy centers:

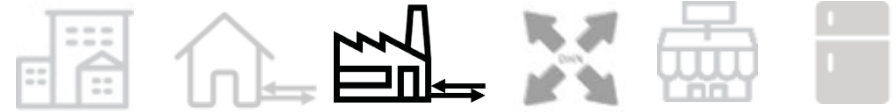
“Big” equipment  
+ Several technologies

Efficiency + CO<sub>2</sub> savings

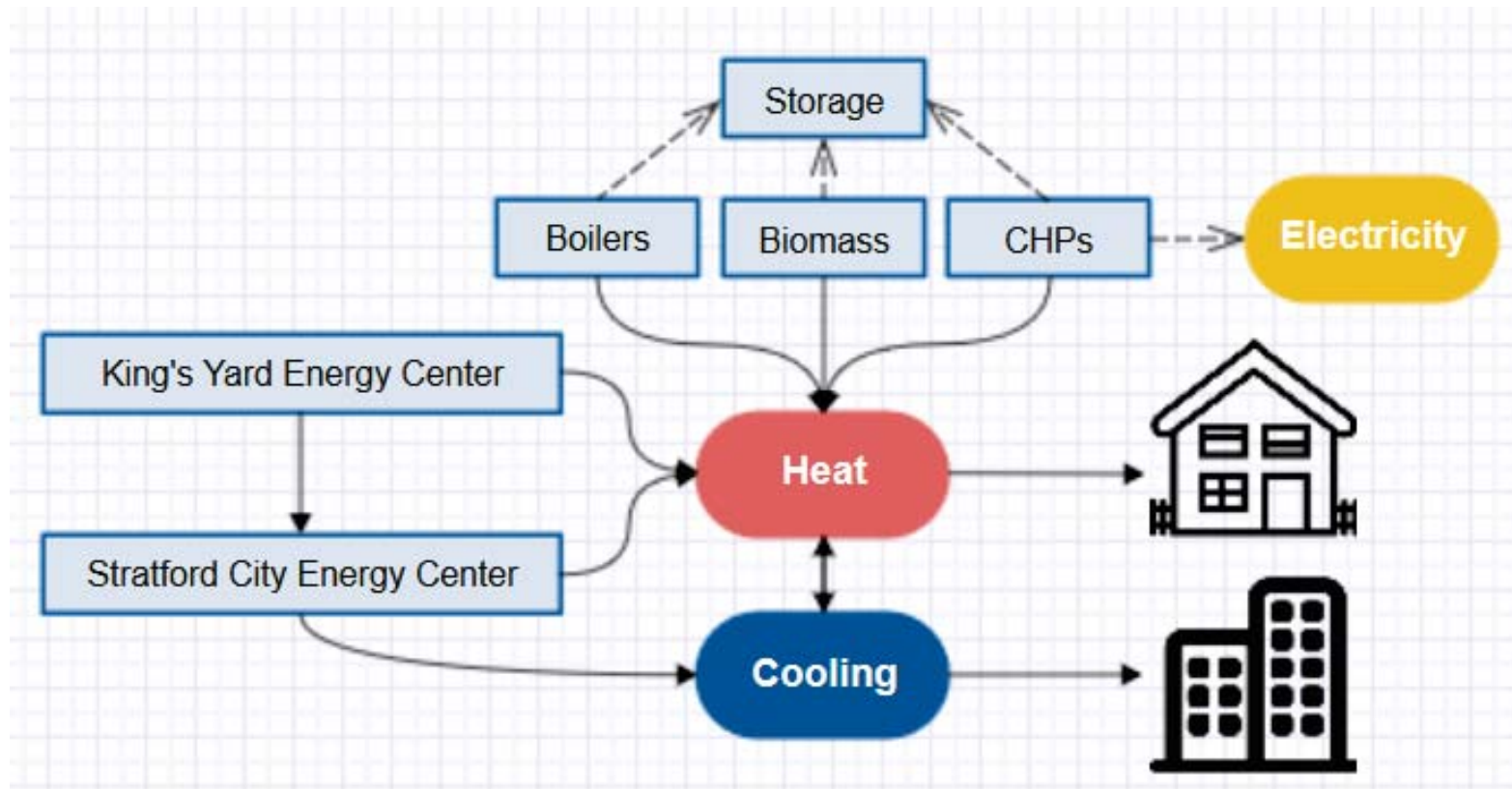


➔ System more complex than a domestic boiler

# Need for an optimisation?

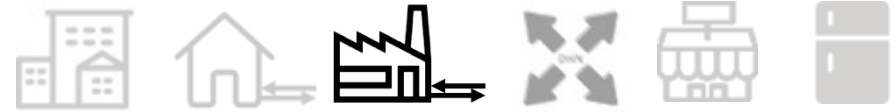


## Queen Elizabeth Olympic Park

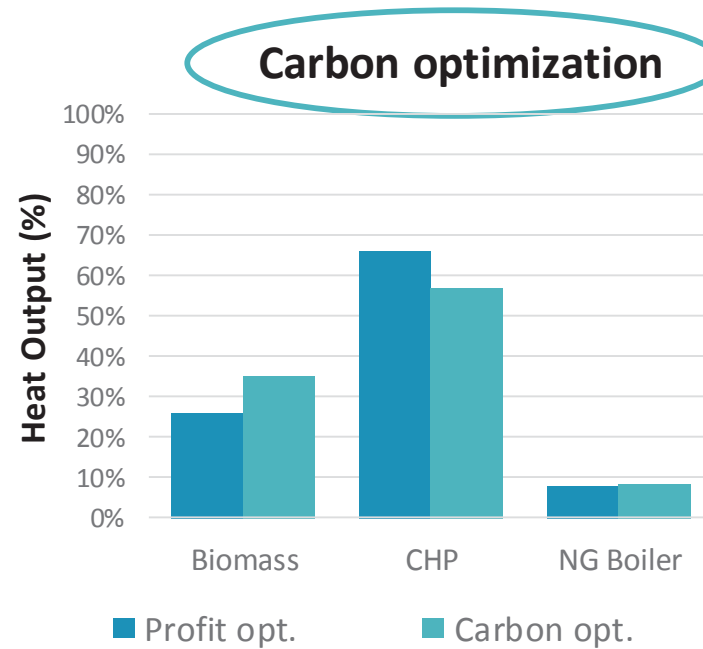
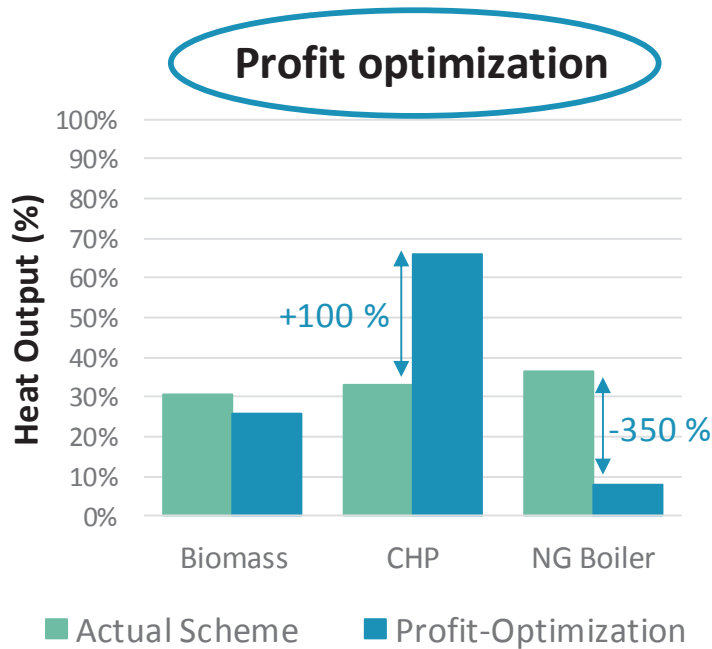


→ Too many factors to operate “manually” = Model

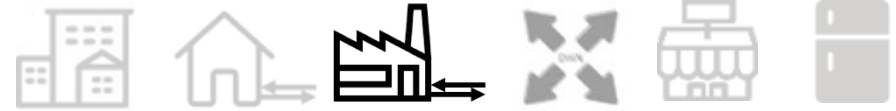
# Results: annual



	Heat carbon int. (gCO <sub>2</sub> /kWh)	CO <sub>2</sub> emissions (tons)	Profit (£)
Profit opt.	90.5	12.7k	5.73m
Carbon opt.	83.3	11.7k	5.36m
Actual scheme	103	-	-

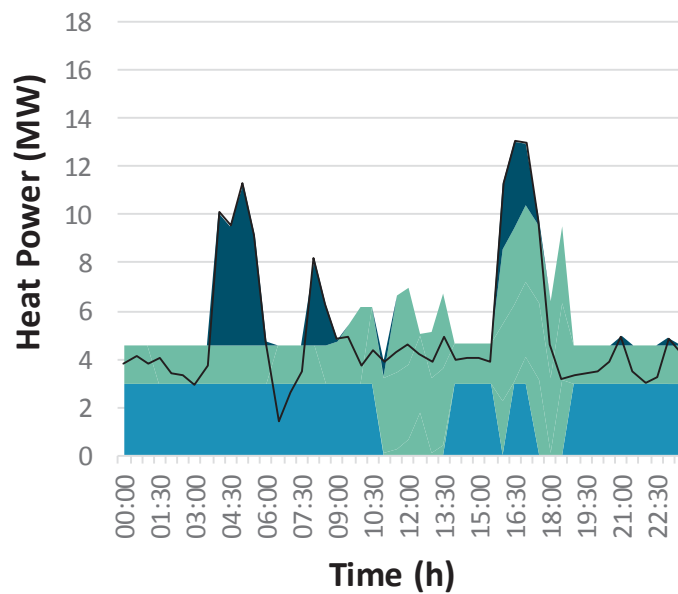


# Results: daily profiles

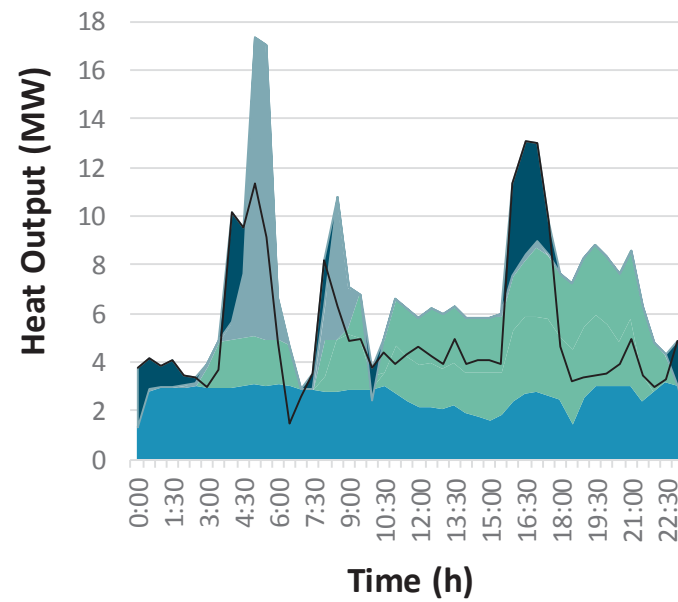


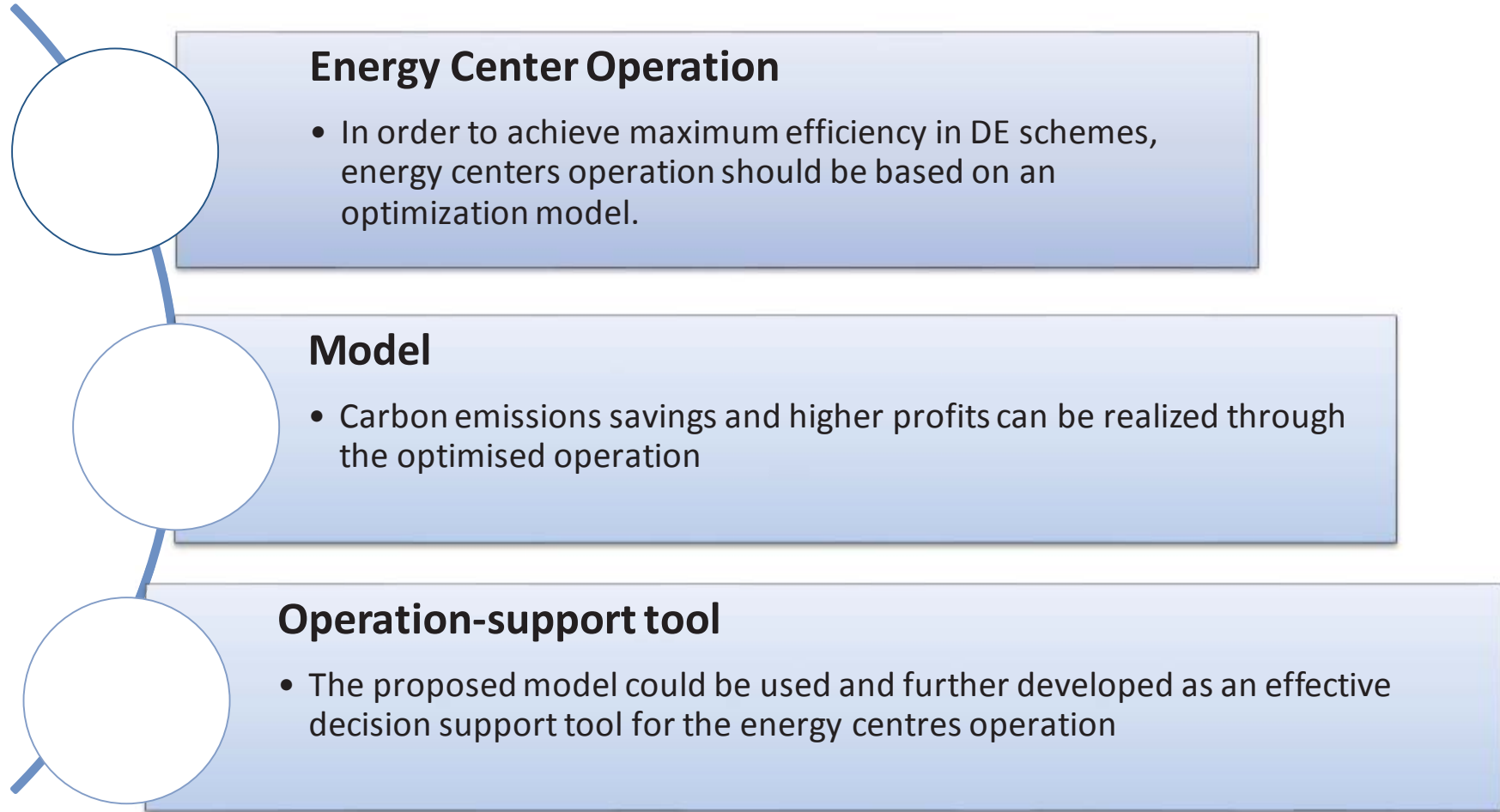
 Biomass boiler  CHP  NG Boiler  Storage

### Modeled operation



### Actual operation









## Modelling and optimisation of a district heating network's marginal extension

Axelle DELANGLE



**Supervisors:**

Prof. Nilay Shah

Dr. Romain Lambert

Dr. Salvador Acha

# Aim & Research Motivation



## District heating networks (DHN):

- Well known & mature technologies
- Expansion already considered in several DHNs

## → Limited research on DHN expansion

### Main objectives:

- Modelling approach
- Strategy to design & operate the energy centre
- Connection scenario to select



# Case study: Barkantine



## Existing network:

- 22 buildings connected
- 2.4 km of pipes
- One existing energy centre

## Extension considered:

- 31 buildings to connect
- 3.5 km of pipes
- 12 years horizon time



# Methodology



Heat demand analysis

- Scenario 1
- Scenario 2

## Design of the energy centre



£ Maximise net present value or minimise GHG emissions under given constraints.  
CO<sub>2</sub>



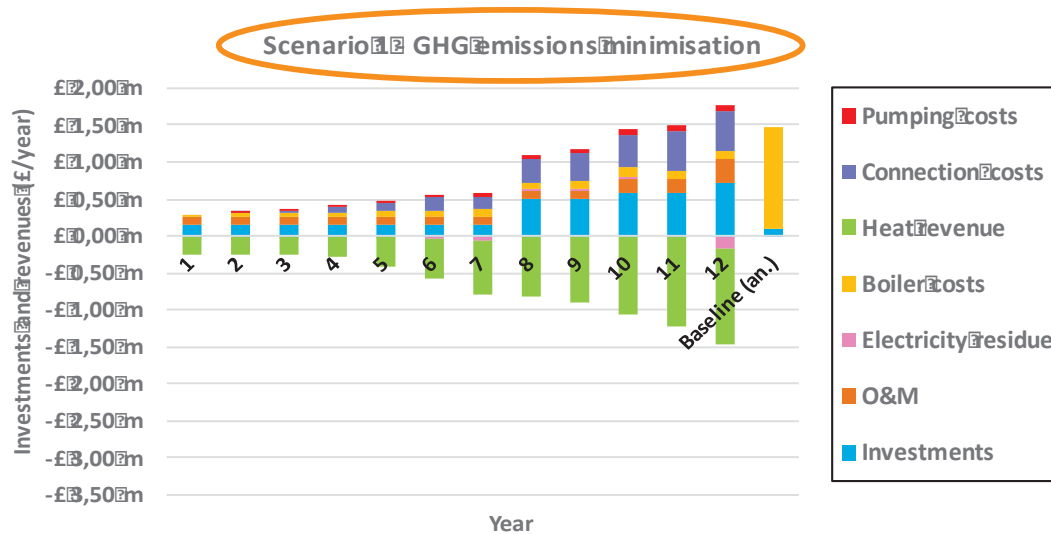
Spatial network extension costs

# Main results



Optimisation performed	Objective function	Scenario 1 (slow connections first)	Scenario 2 (quick connections first)
Profit maximisation	Net Present Value (£)	£10,101,800	£13,173,700
GHG emissions minimisation (DUKES)	GHG emissions (tCO <sub>2</sub> eq)	1,670	1,920

Table 1: Design of the energy centre



→ Not financially viable



→ Financially viable

# Conclusions

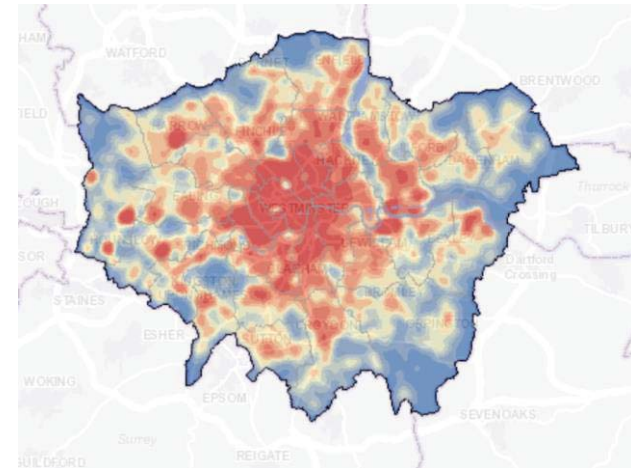


## DHN expansion:

- Financial & environmental advantages
- Planning strategies are essential
- Using an optimisation approach is crucial

## Barkantine case study:

- **Costs** optimisation performed: **Scenario 2.**  
→ **Financially viable**
- **GHG emissions** minimisation: **Scenario 1.**  
→ **Additional subsidies required**

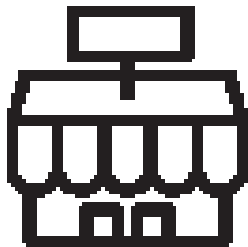


**The model developed can be applied to other DHNs.**



## Modelling and Optimisation of Distributed Energy Resources in Food Retail Buildings: An Investment and Management Approach

Aspasia GEORGAKOPOULOU



**Supervisors:**

Dr. Salvador Acha

Dr. Christos N. Markides

Prof. Nilay Shah

# Research motivation and aim of the project



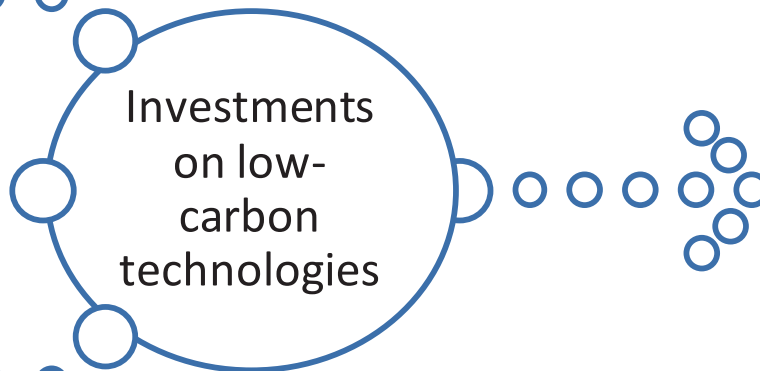
6000 buildings



1% of total UK emissions



High energy demand



## Combined Heat and Power (CHP)



### Aim of the project:

“Optimal implementation and operation of gas-fired internal combustion CHP units, when integrated with heat recovery and conversion technologies.”



# Method: Optimisation Model

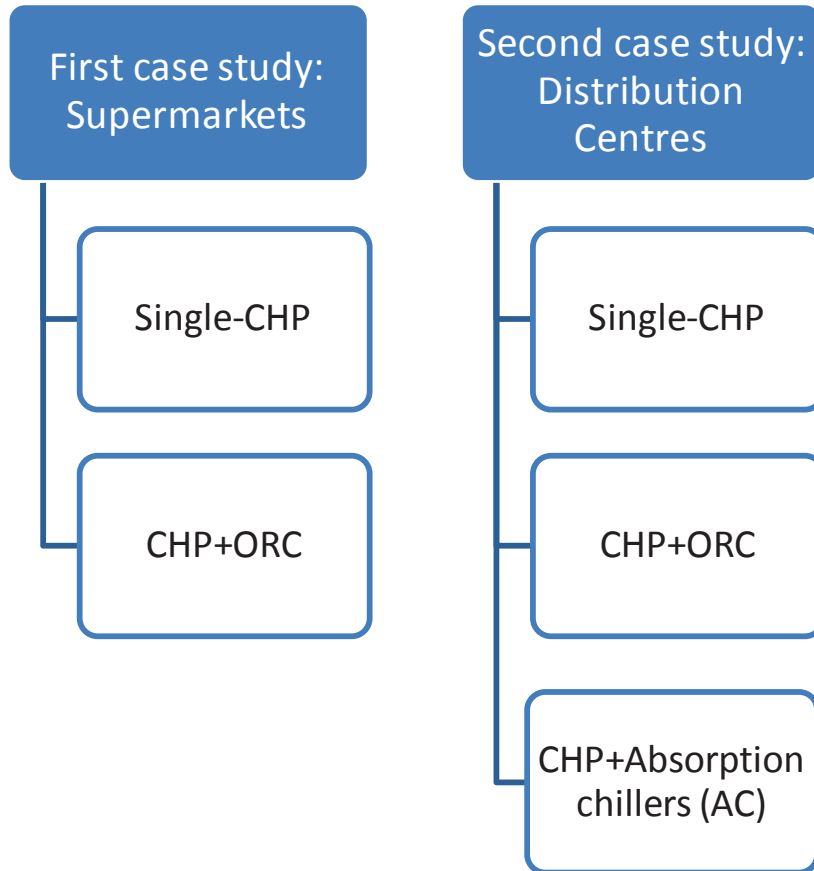


- ✓ *Gas & Electricity prices*
- ✓ *Technology options*
- ✓ *Energy demand*
- ✓ *Fuel employed*

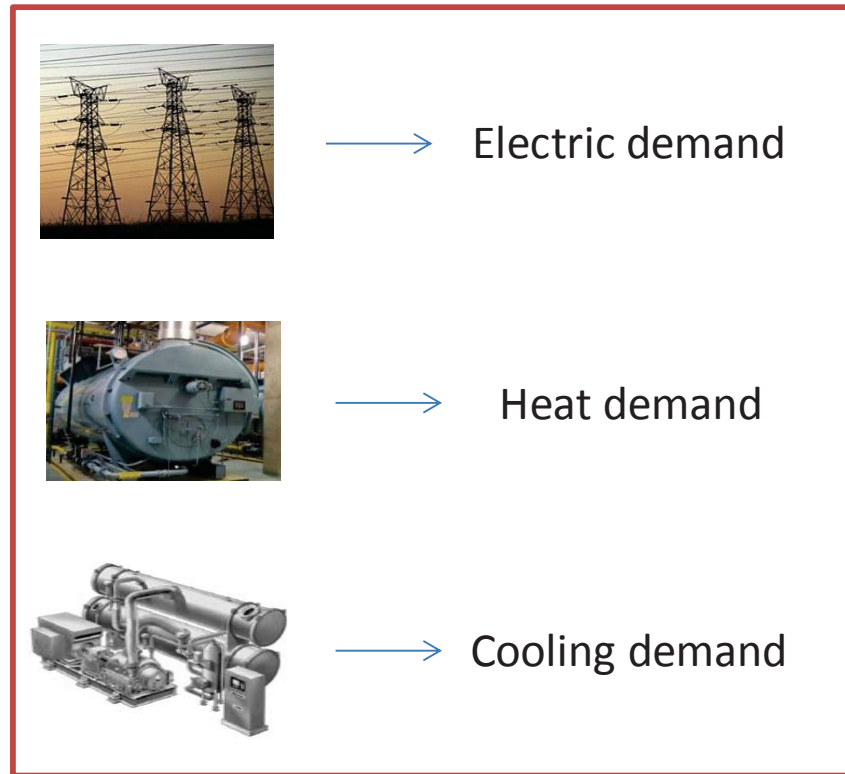
- ✓ *Optimal technology*
- ✓ *Operational strategy*
- ✓ *Total cost*
- ✓ *Total emissions*



# Method: Case studies



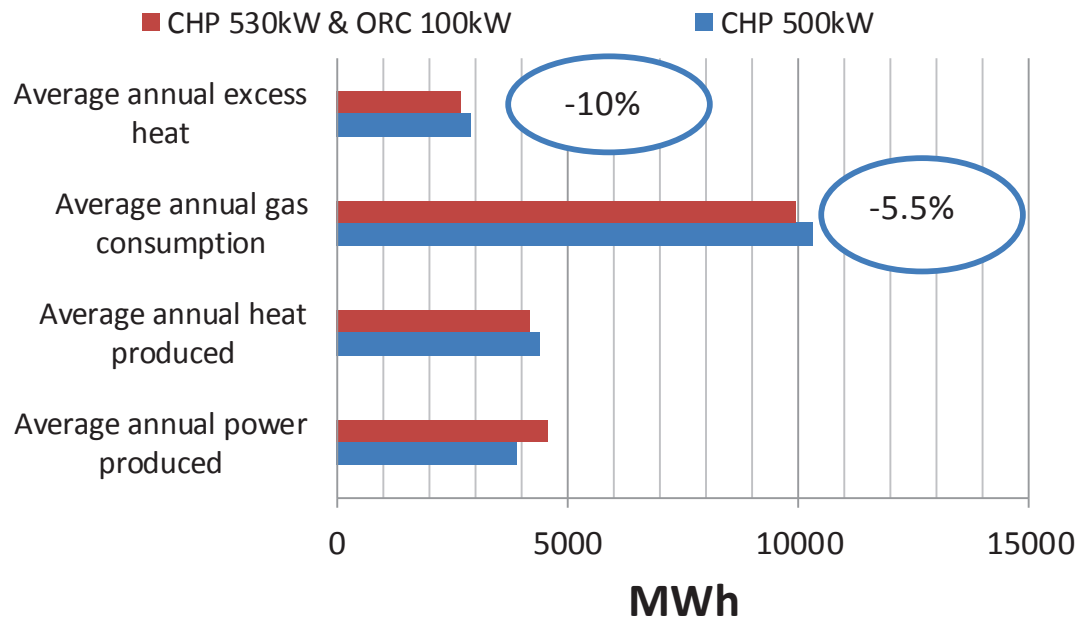
## *Business-as-usual scenario*



# Results: first case study



<b>CHP 530kW ORC 100kW</b>	Average savings per year (£)	Savings (%)	Average carbon emissions reduction per year (tn CO <sub>2</sub> eq)	Reduction (%)
Minimum cost NG vs Baseline	£193,500	32.5%	430	18.8%
Minimum cost BM vs Baseline	£227,400	38.2%	2260	99.3%



✓ Payback period: 4 years  
✓ ROI: 321%

# Results: second case study

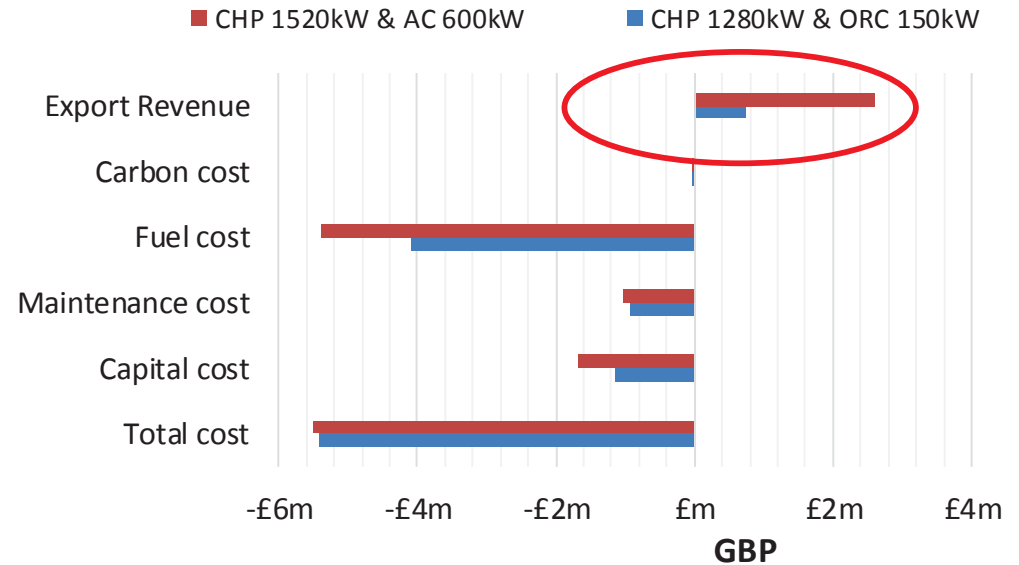
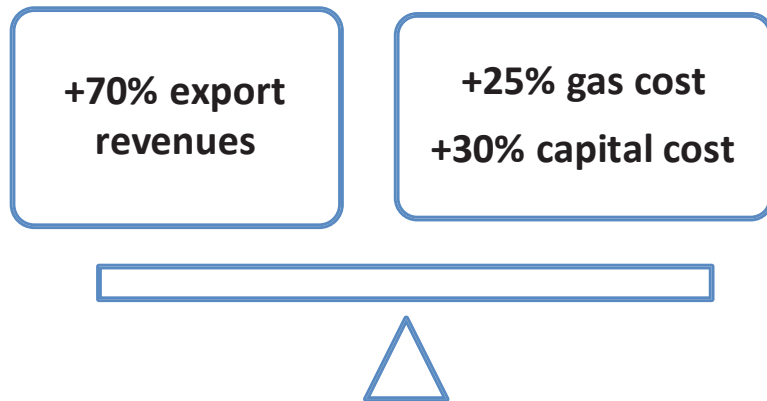


Technology	CHP 1280kWe & ORC 150kWe	CHP 1520kWe & AC 600kWth
Fuel	Biomethane	Biomethane
Average annual cost savings	37.8%	37.1%
Average annual carbon savings	99.8%	99.8%
IRR	37.8%	29.3%
ROI	277%	186%
Payback period	4 years	5 years



Total cost for CHP+AC  
**1.25%** higher than for  
CHP+ORC

# Results: second case study



## CHP+ORC Operational strategy:

- No grid-imported electricity;
- **10% of electricity is exported;**
- Electric chiller covering cooling demand
- Boiler not contributing to heat generation;

## CHP+AC Operational strategy:

- No grid-imported electricity;
- **40% of electricity is exported;**
- AC covers 40% of cooling demand;
- Boiler provides 15% of heat required.



**Optimisation model:  
Generic tool for selection of technology portfolios in commercial buildings**



Waste heat conversion technologies:

- ✓ Reduced gas consumption
- ✓ Increased CAPEX

Fuel employed:

- ✓ Biomethane for full decarbonisation
- ✓ Natural gas CHP+ORC: 19% emissions reduction

CHP coupled with ORC or AC:

- ✓ Comparable benefits
- ✓ Applications in buildings with high cooling demand



## Water-Cooled Refrigeration Systems in the Food Retail Industry

Maria-Aliki EFSTRATIADI



**Supervisors:**

Dr. Christos N. Markides

Dr. Salvador Acha

Prof. Nilay Shah

# Aim & Research Motivation



Supermarket refrigeration: **30%-60%** of total energy consumed in the stores

- High amounts of low-grade heat rejected by the air-cooled condensers to the ambient air
- Current status: Air-cooled condensers situated on the rooftop of the stores

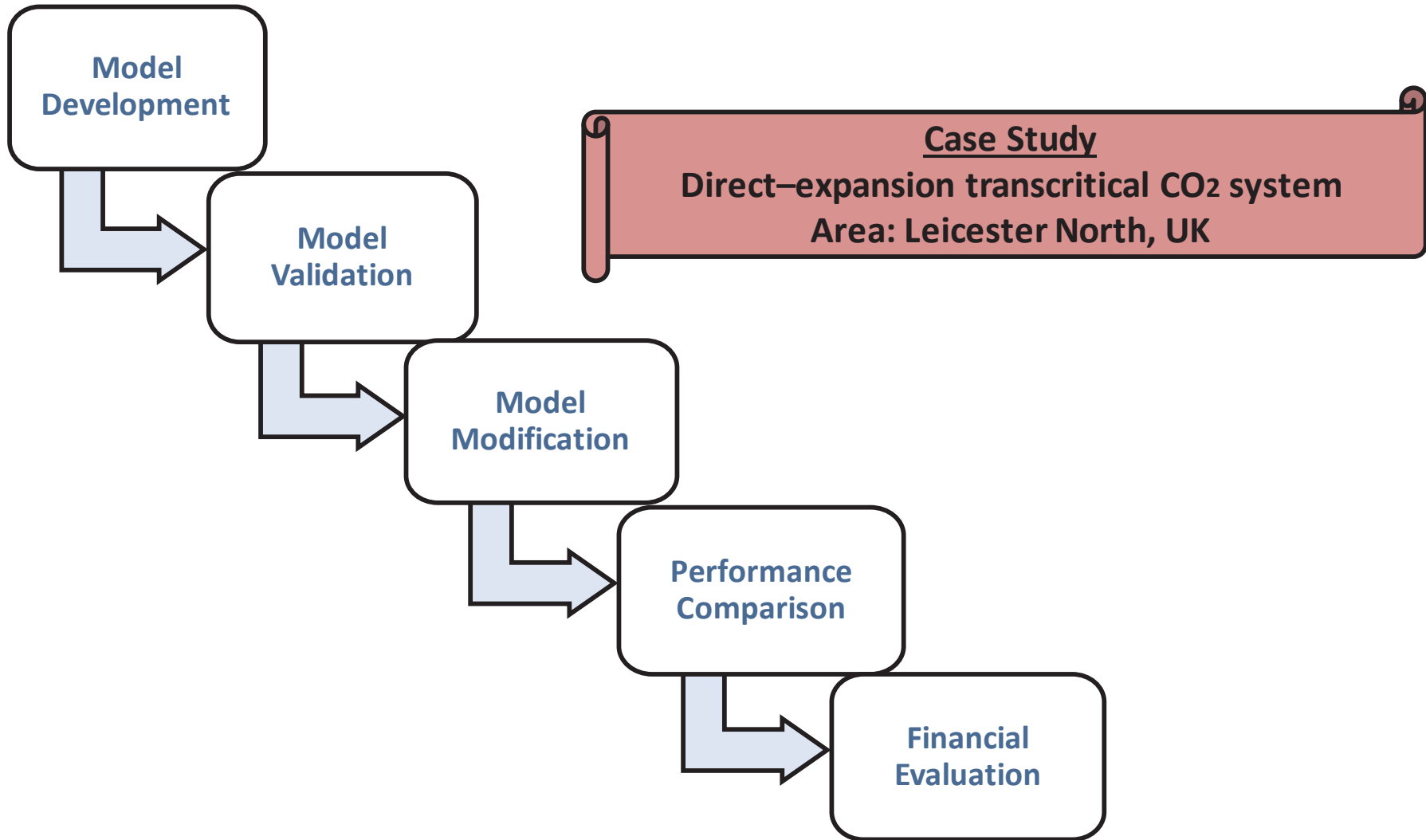


*Typical commercial air-cooled condenser*

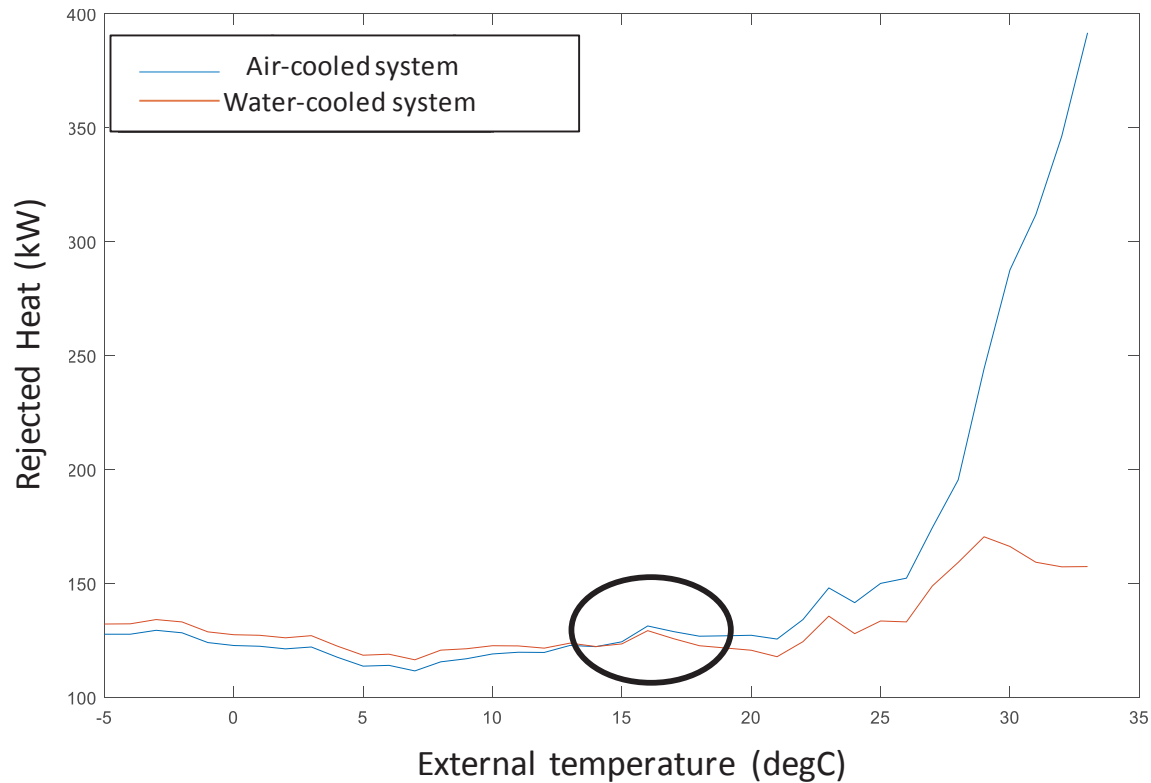
**Could a water-cooled condenser rejecting heat to the soil via an intermediate closed water-circuit address this problem?**



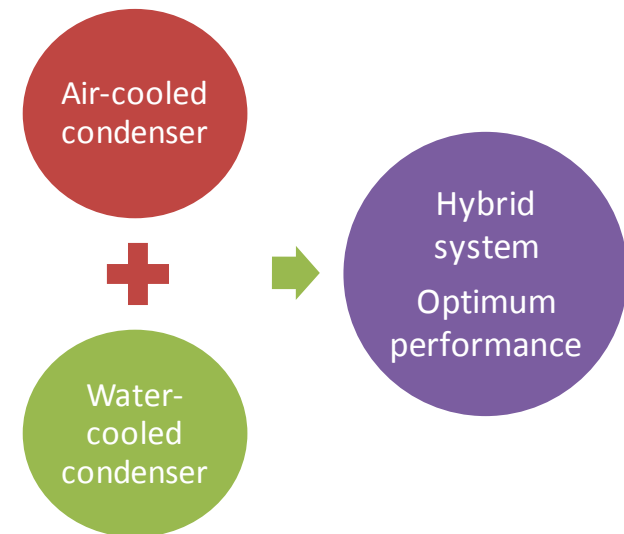
# Methodology



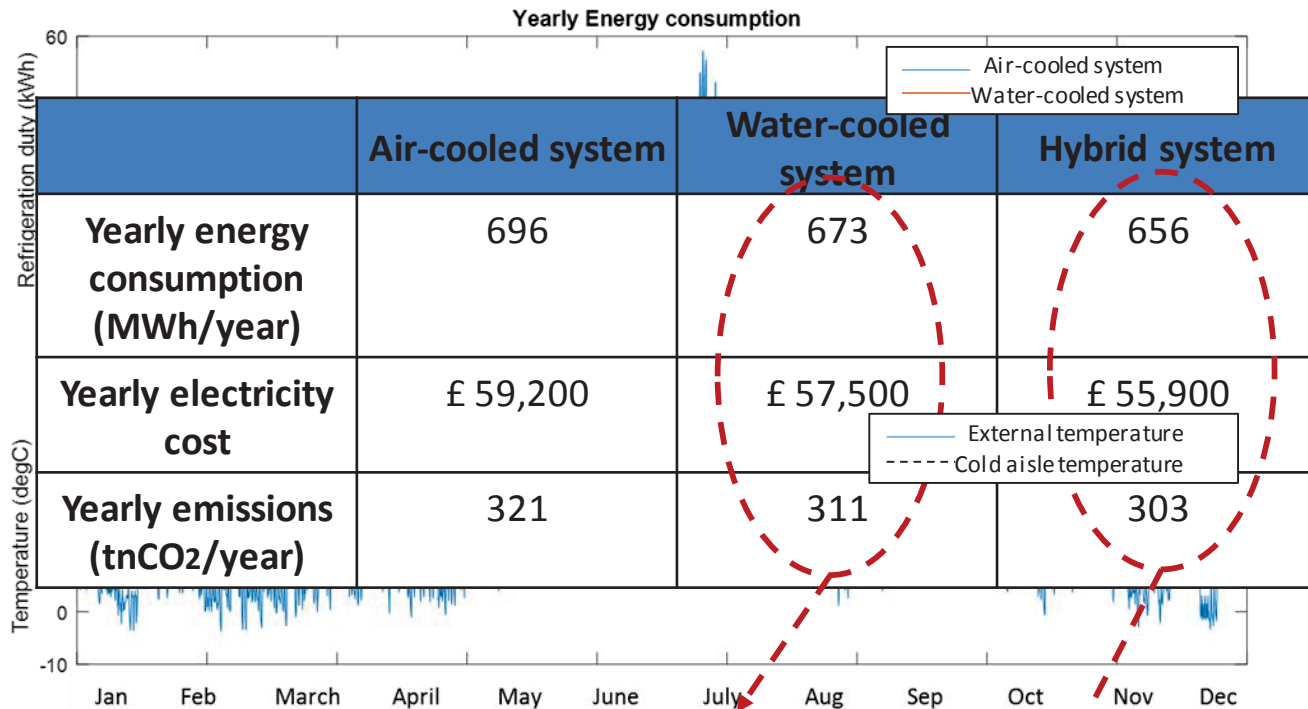
# Modelling Results(1/2)



- ✓ **High external temperature:** Up to 5 times less heat rejected by the water-cooled condenser
- ✓ **Low external temperature:** Marginally higher performance of the air-cooled system



# Modelling Results (2/2)



➤ **65% of the year** the air-cooled system consumes less energy than the water-cooled system

➤ **Main reason:** Low average temperature throughout the year

-3.5%

-5.5%

**Marginal energy savings**

# Financial Evaluation



	Retrofit with a Hybrid System	Water-cooled System in a new store	Hybrid system in a new store
CAPEX	£ 83,000	<b>-£ 40,000</b> relative to BAU	<b>+£ 7,000</b> relative to BAU
OPEX relative to BAU	+200%	+150%	+200%
Energy costs per year relative to BAU	-£ 3,500	-£ 1,800	-£ 3,500
Total annual savings relative to BAU	£ 1,800	£ 1,100	£ 1,800
Payback Period	>10 years	Immediate payback	4.8 years



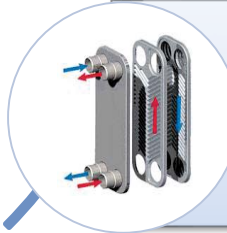
**Systems downsized upon design**

# Conclusions



## Air-cooled systems

- Highly dependent on external conditions
- Show higher performance levels in cold ambient conditions



## Water-cooled systems

- Less sensitive to external temperature variations
- Systems downsized
- Attractive economics for new stores applications

**Model: Reliable but Case specific**

*Needs to be applied in other systems for a general understanding*

# Acknowledgments



Dr. Salvador Acha

Gonzalo Bustus-Turu, PhD

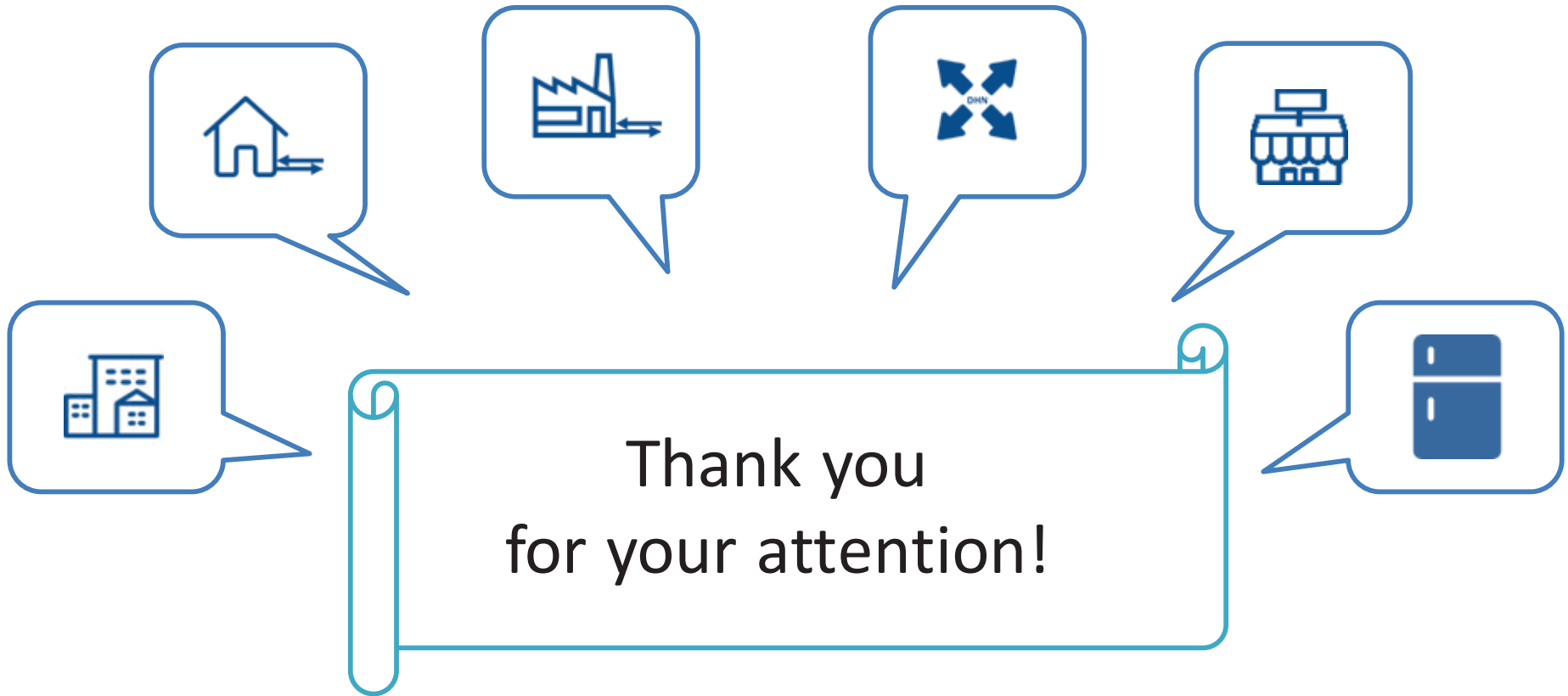
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Dr. Christos N. Markides

Dr. Christoph Mazur

Dr. Koen H. van Dam

Prof. Nilay Shah





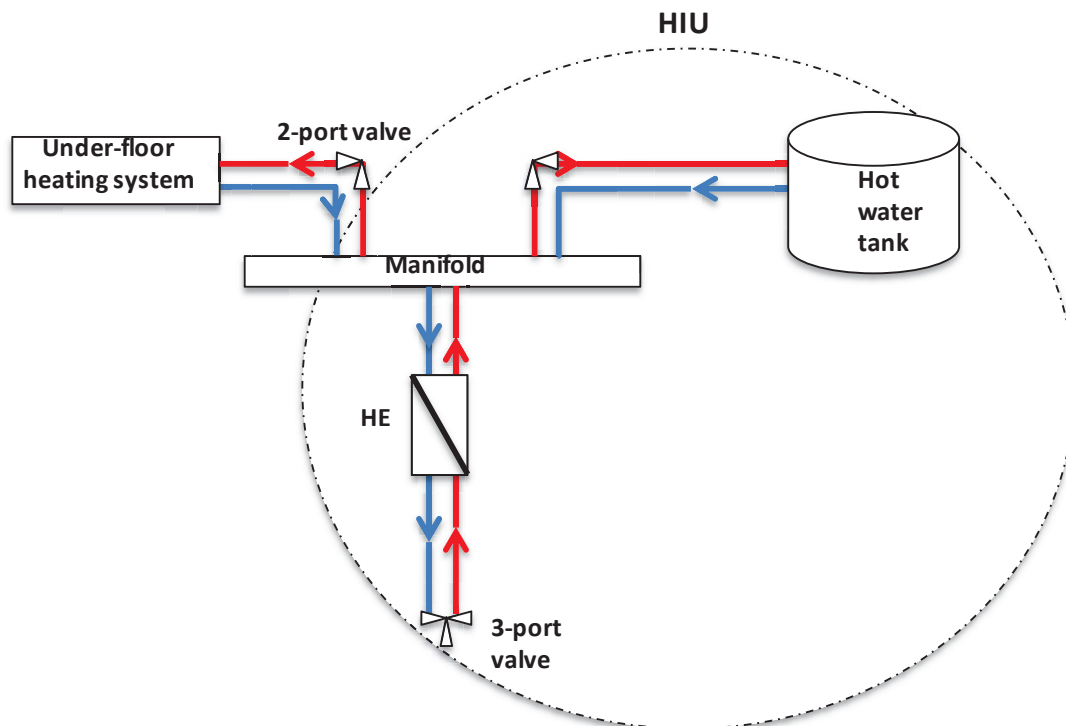
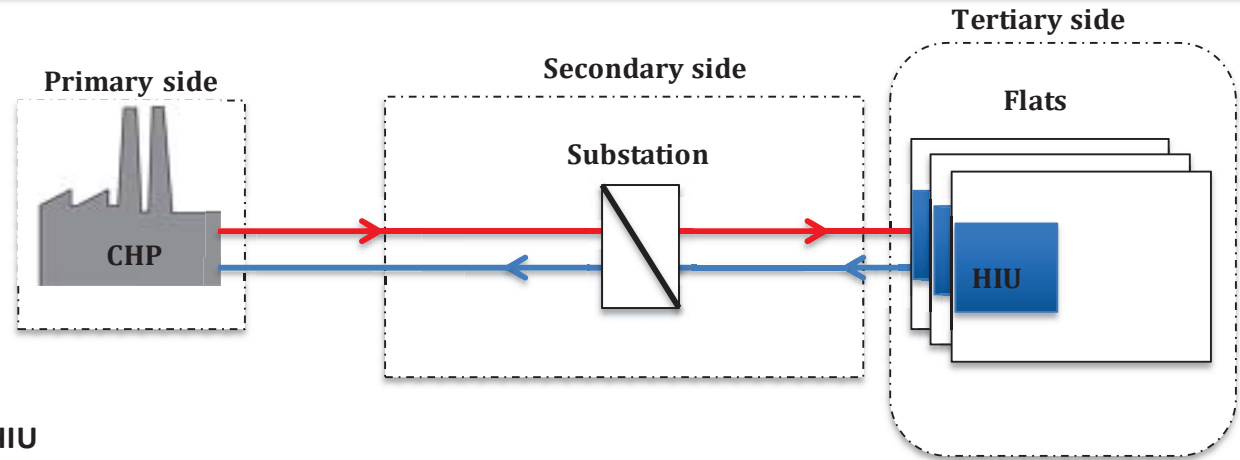
## APPENDICES



# Layout of the DHN

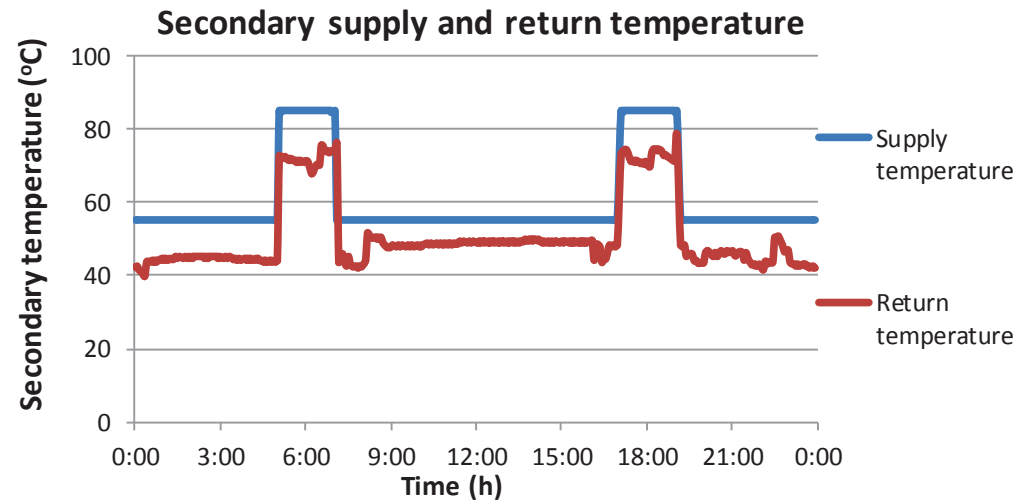
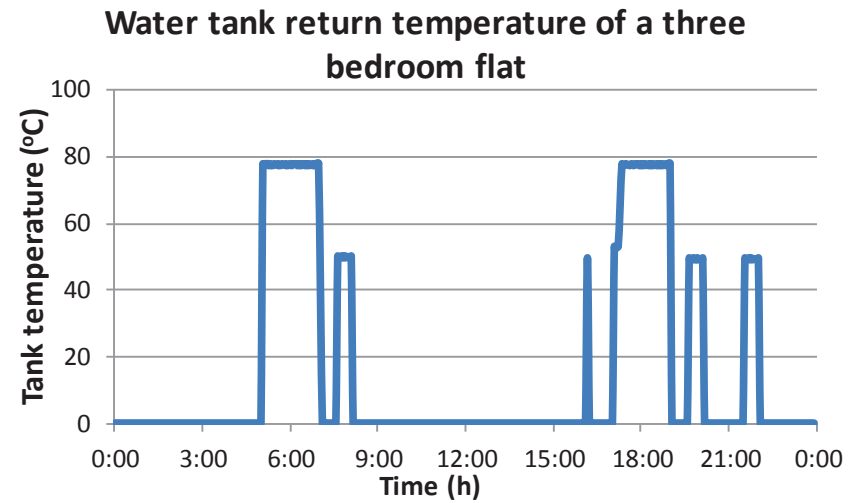
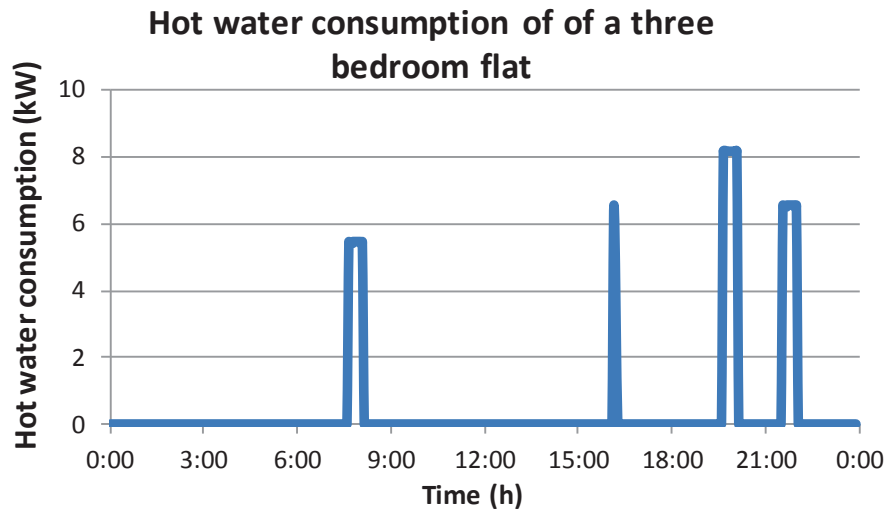


Layout of the DH block examined in this model



Simplified representation of the tertiary side studied in this model

# Baseline scenario results

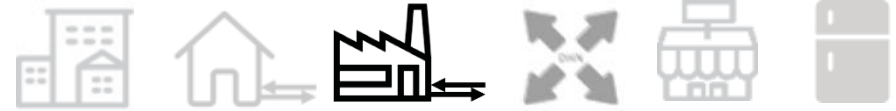


# Comparison of scenarios

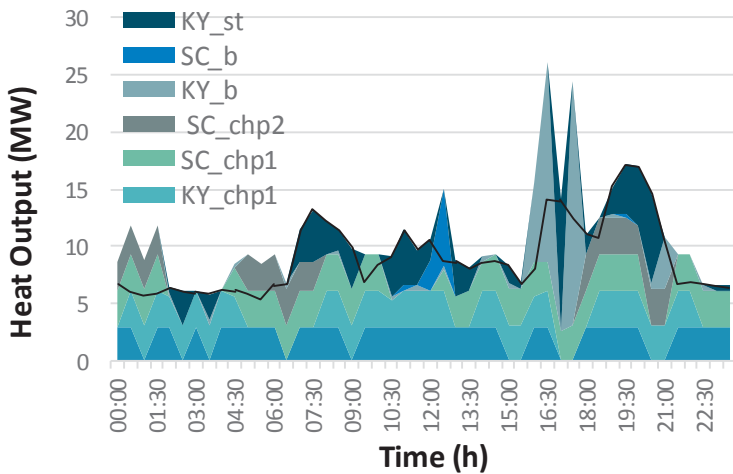
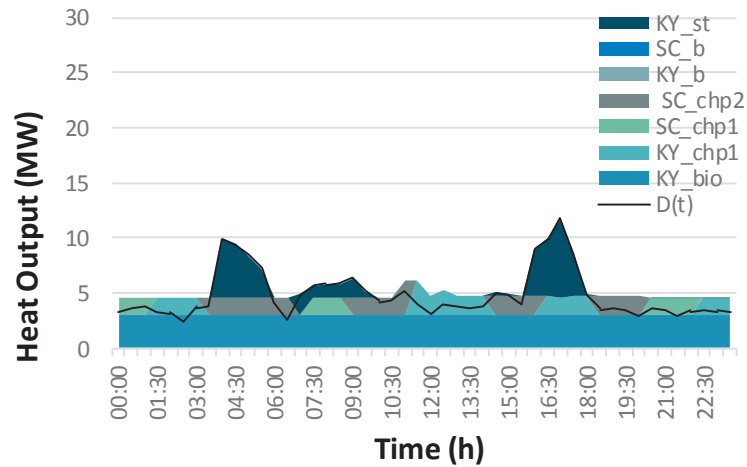


	Change in heat losses (%)	Fuel consumption change (%)	Change in cost of fuel consumption (%)	Change in carbon emissions (%)	Change in temperature difference between supply and return (%)	Water tank return temperature change (%)	Primary return temperature change (%)	Secondary temperature change (%)
<b>Scenario 1</b>	-0.7	-0.07	0	-0.086	+ 0.96	-1.8 (for non zero values)	-0.32	- 0.16
<b>Scenario 2 (improved case)</b>	-3.16	-0.76	-0.88	-0.76	+ 3.3	No change	-1.3	-0.5
<b>Scenario 2 (ideal case)</b>	-1.8	-1.38	-1.55	-1.38	+ 13	No change	-4.6	-2.4
<b>Scenario 3</b>	-12.3	-2.7	-2.7	-2.7	Not consistent during the day	Negligible	Not consistent during the day	Not consistent during the day

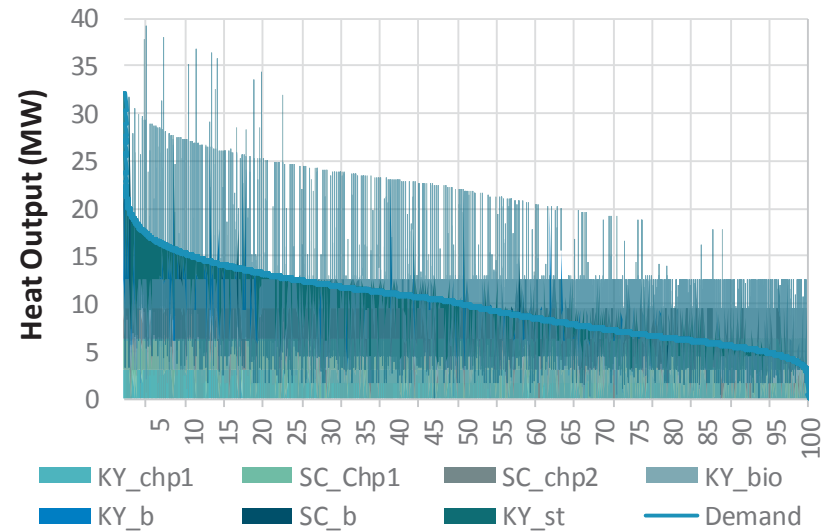
# Results: heat



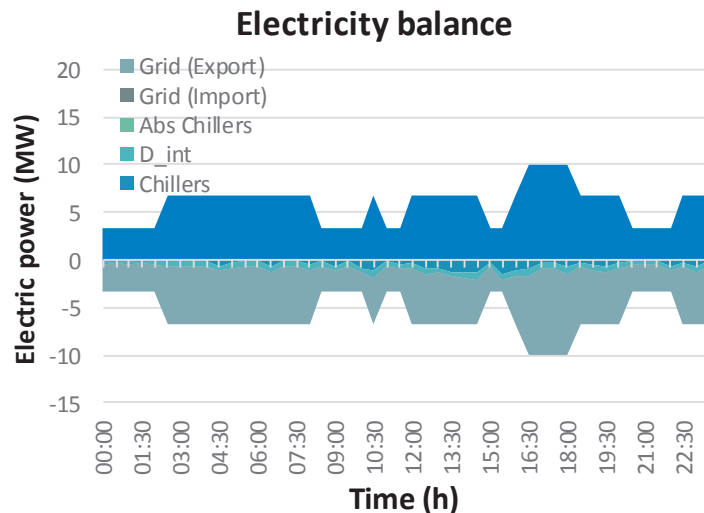
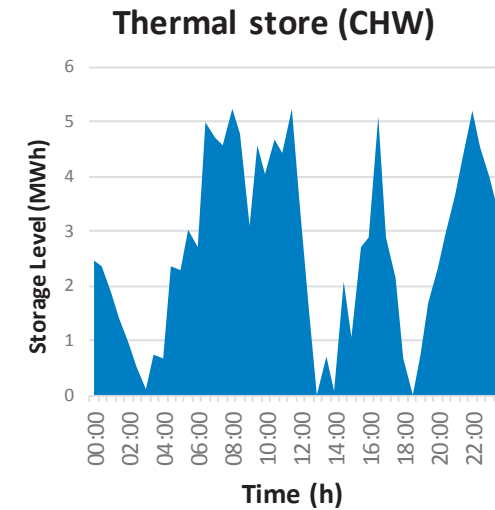
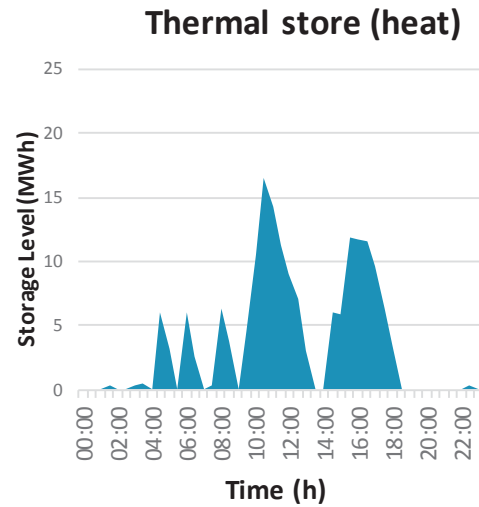
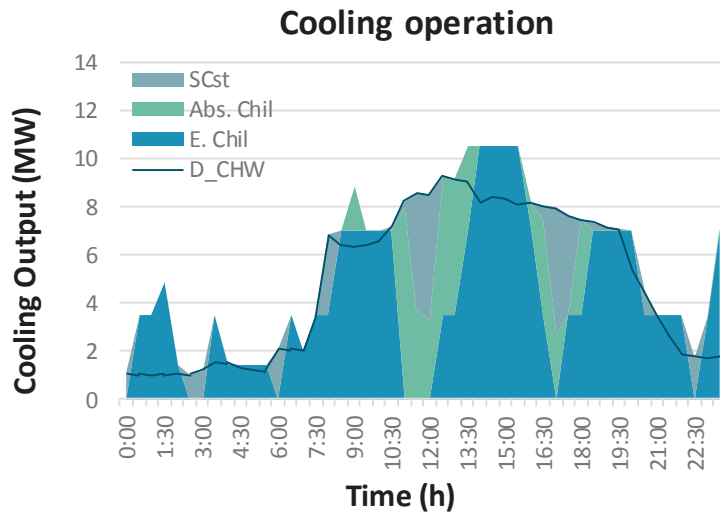
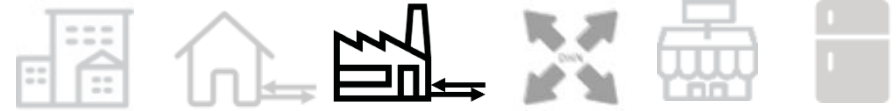
### Daily half-hourly profiles



### Annual load duration curve



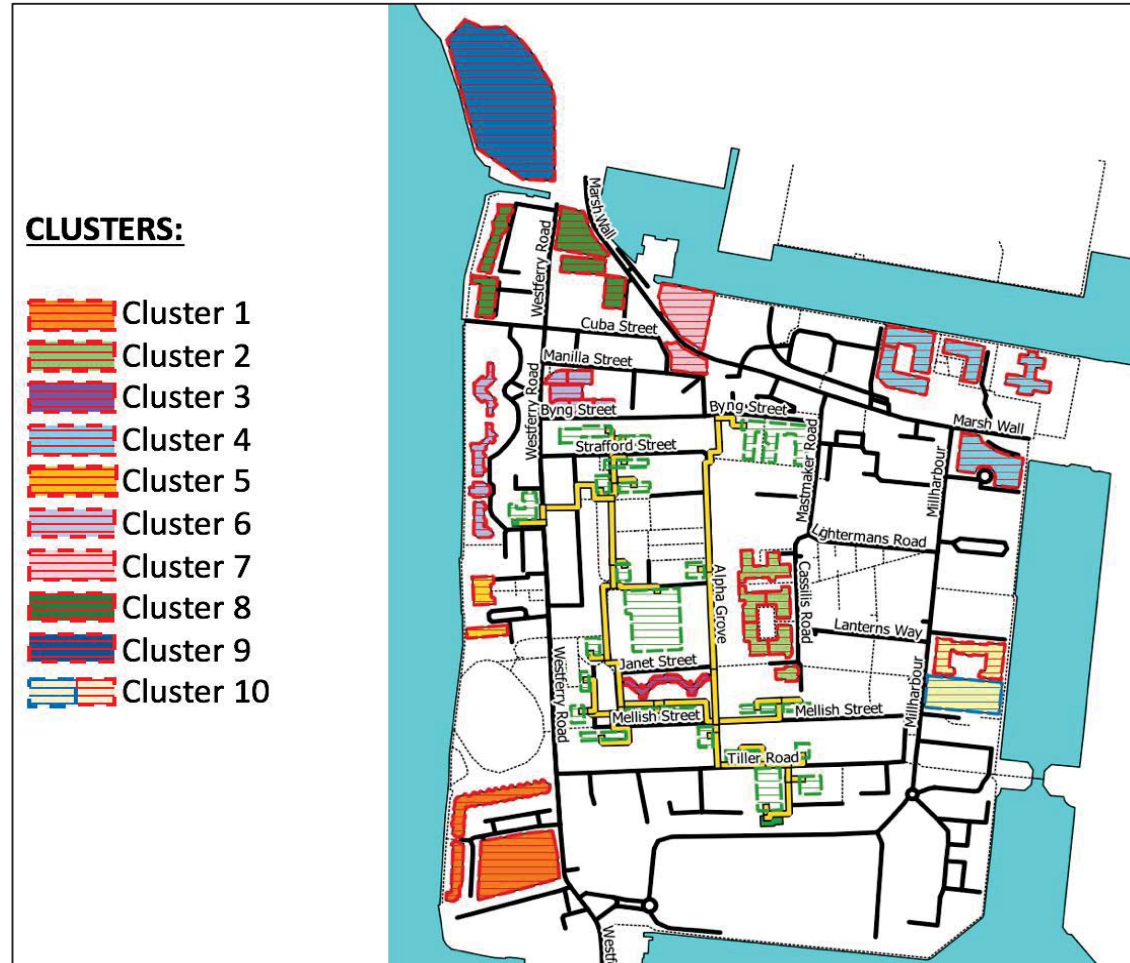
# Results: other outputs



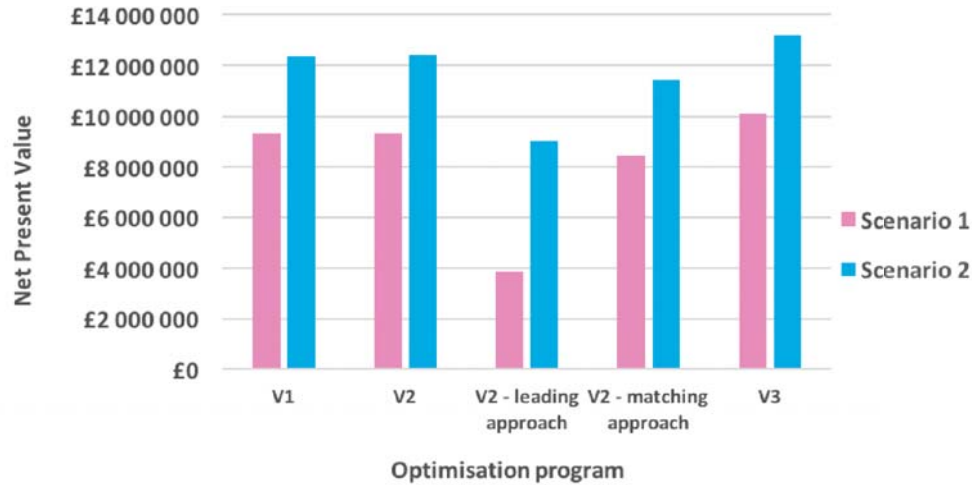
### Annual Results

	Heat carbon int. (gCO <sub>2</sub> /kWh)	CO <sub>2</sub> emissions (tons)	Profit (£)
Profit opt.	90.5	12.7k	5.73m
Carbon opt.	83.3	11.7k	5.36m
Actual scheme	103	-	-

# Gathering of loads into clusters

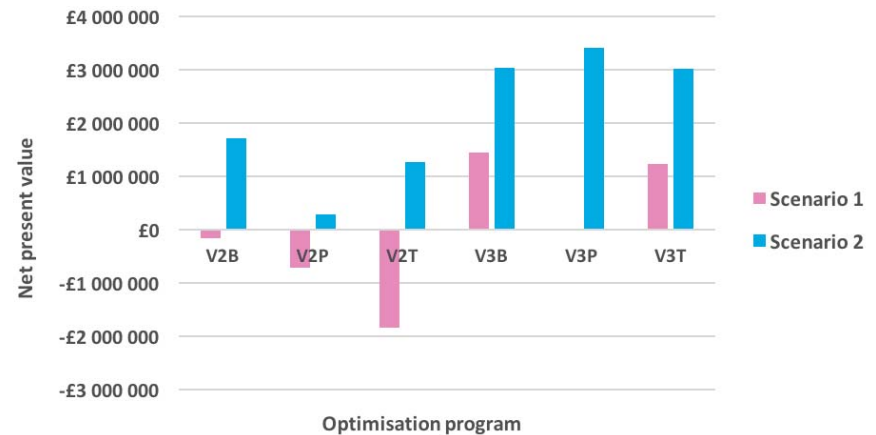


# Comparison of the NPV obtained in each model

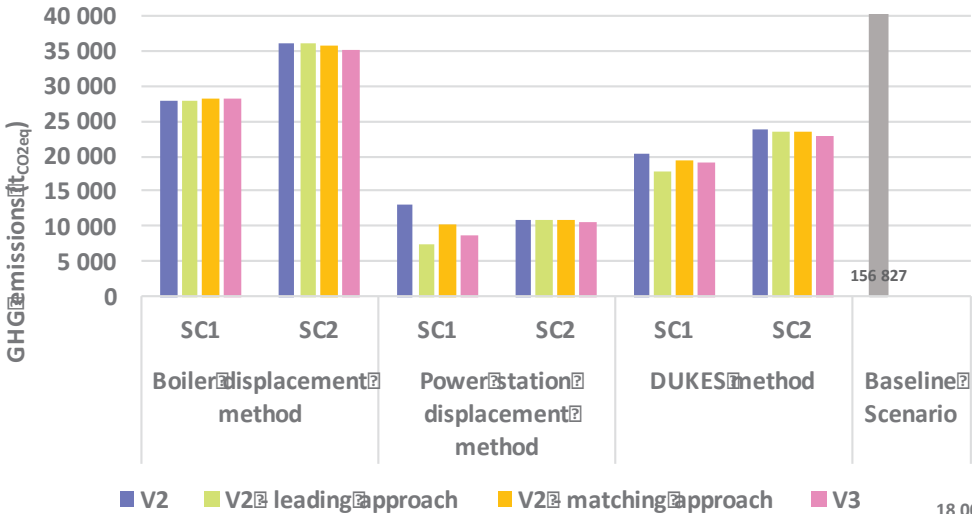


## Costs optimisation approaches

## GHG emissions minimisation approaches

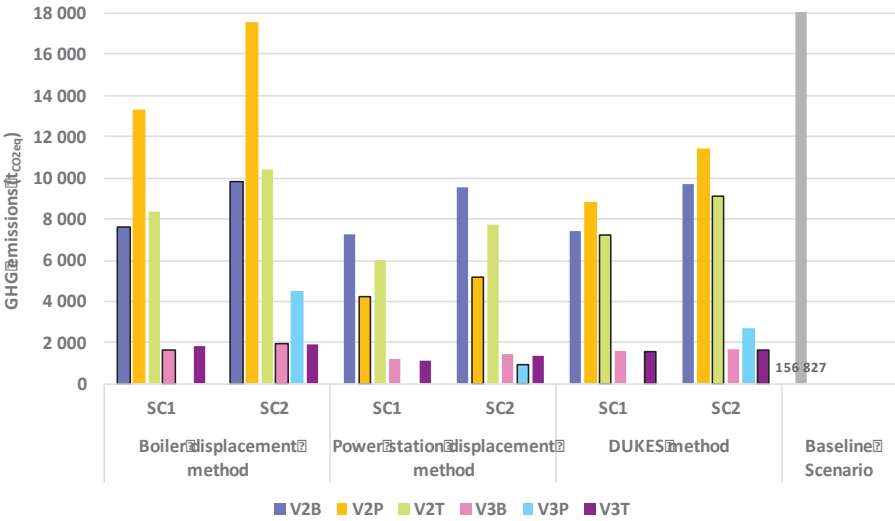


# Comparison of the emissions obtained in each model



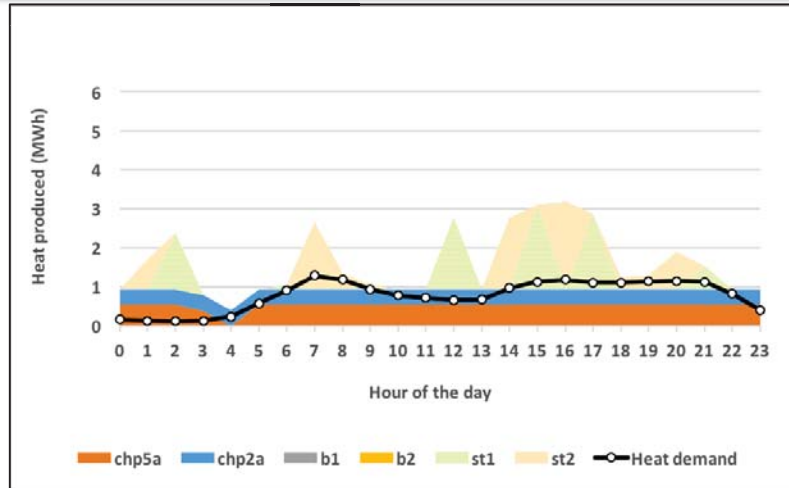
## Costs optimisation approaches

## GHG emissions minimisation approaches

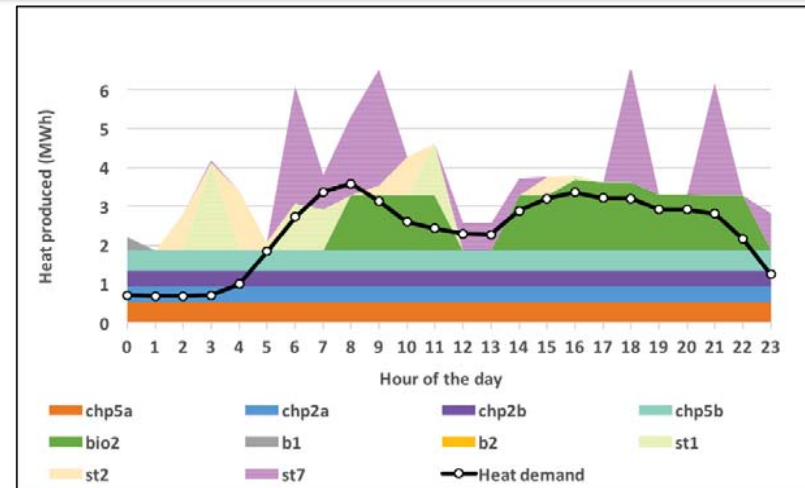




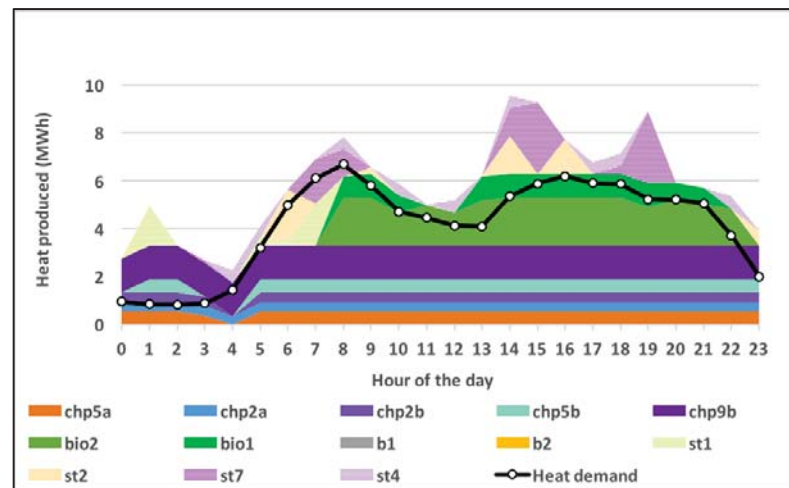
# Heat production profiles, costs optimisation model (V3), day 1



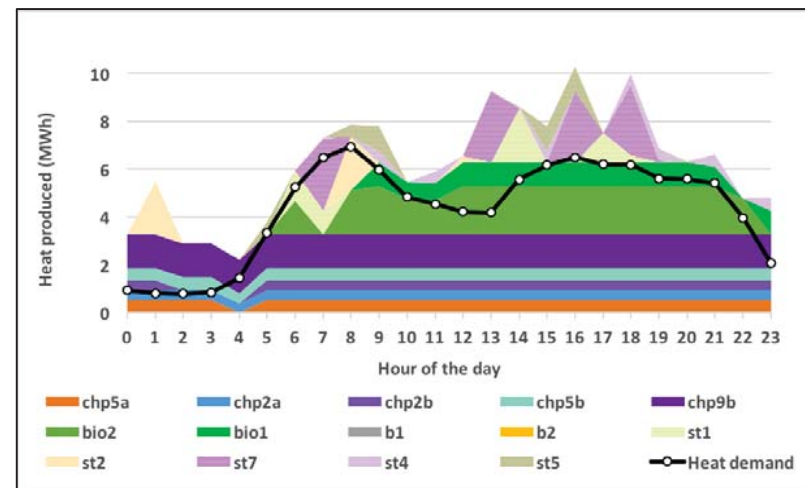
Year 2



Year 5

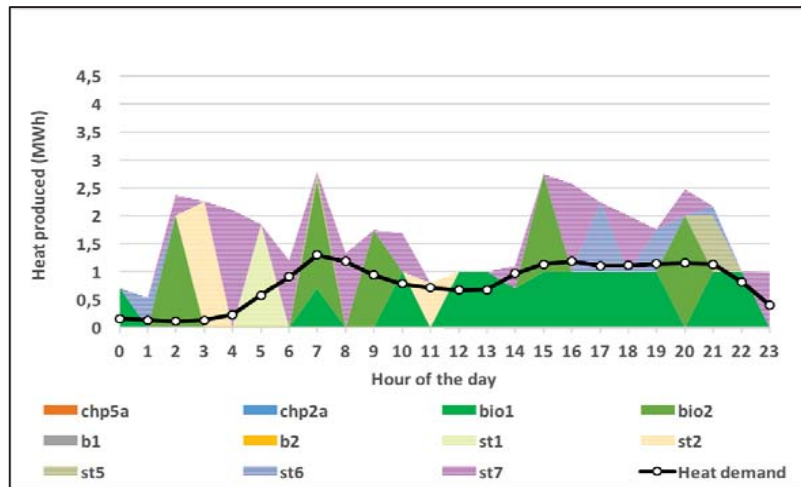


Year 10

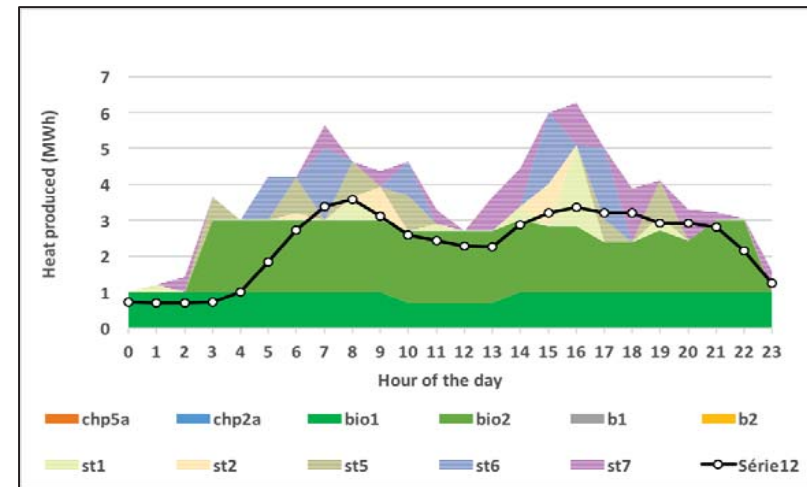


Year 12

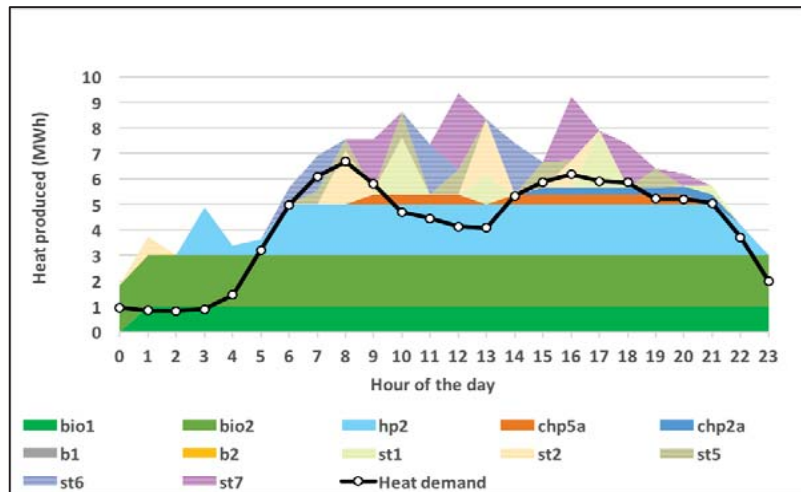
# Heat production profiles, emissions minimisation model (V3T), day 1



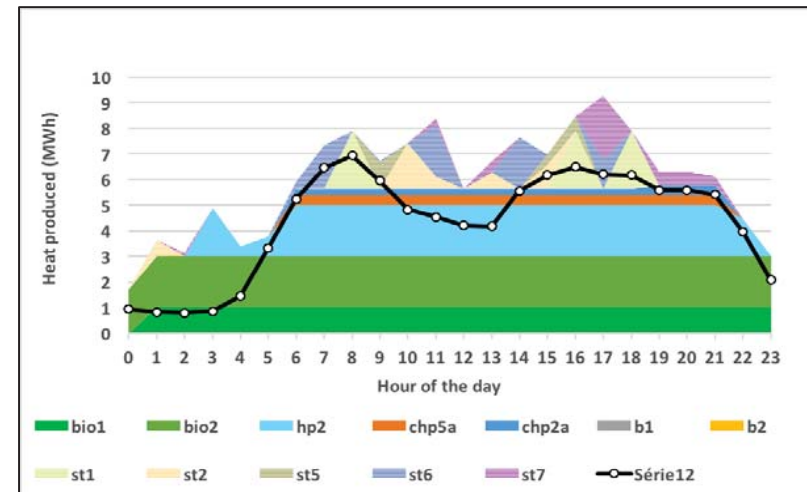
Year 2



Year 5

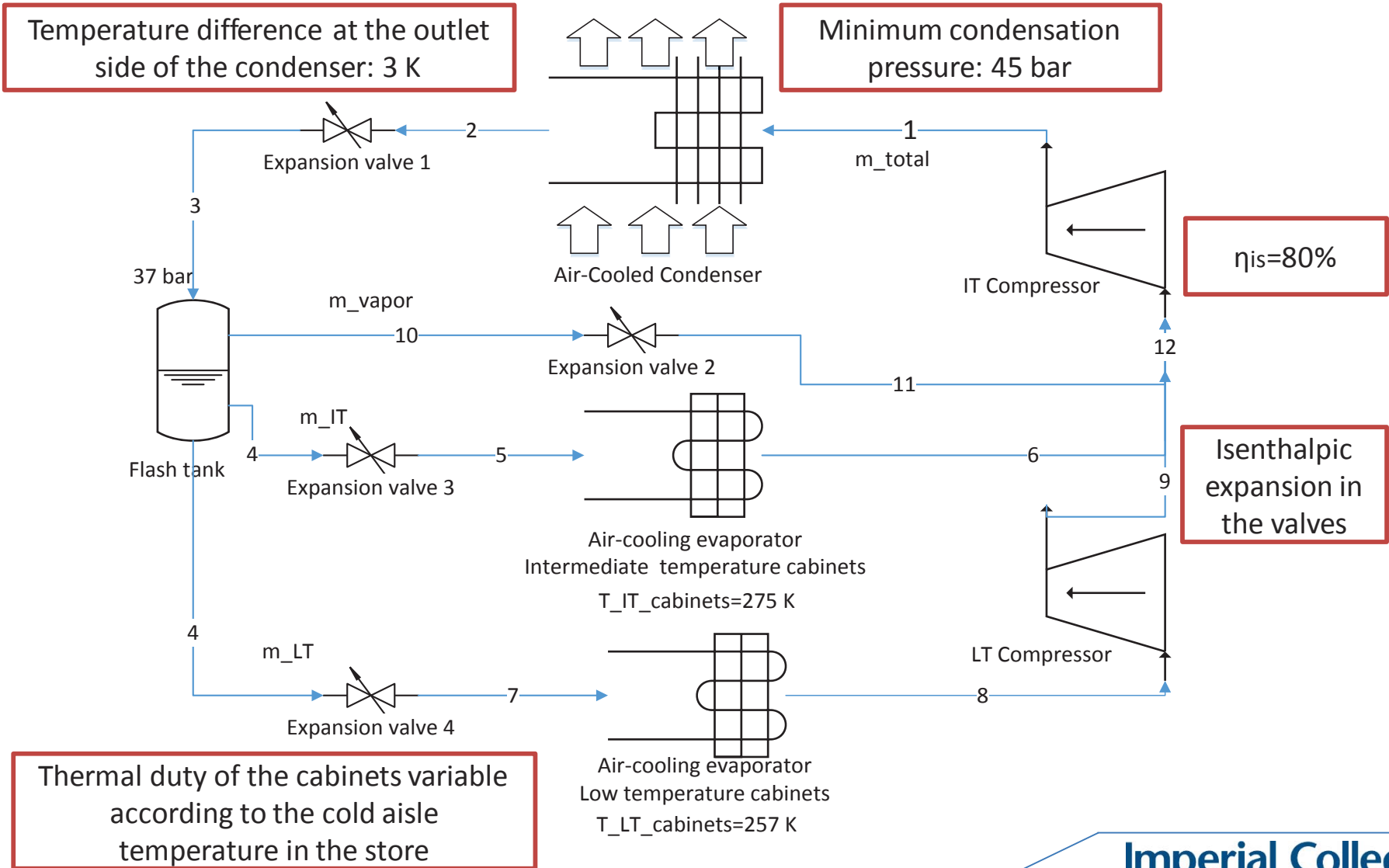


Year 10



Year 12

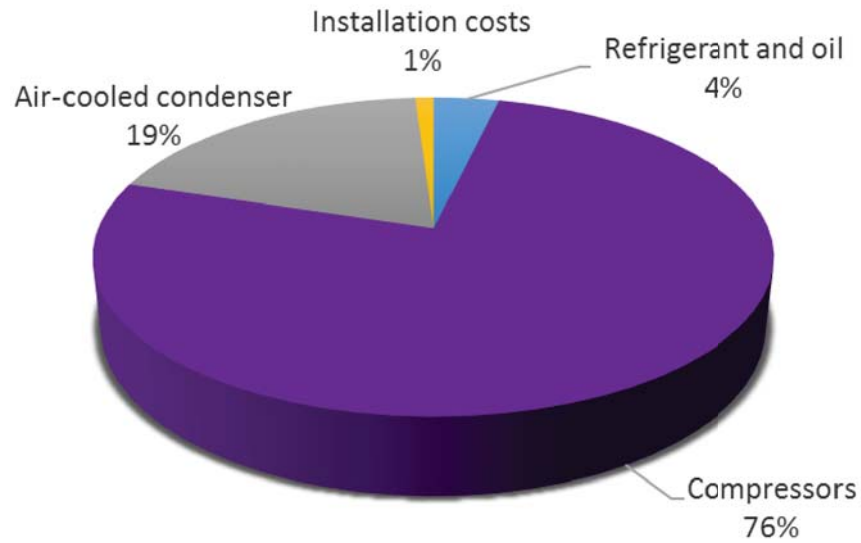
# Case Study: Leicester North



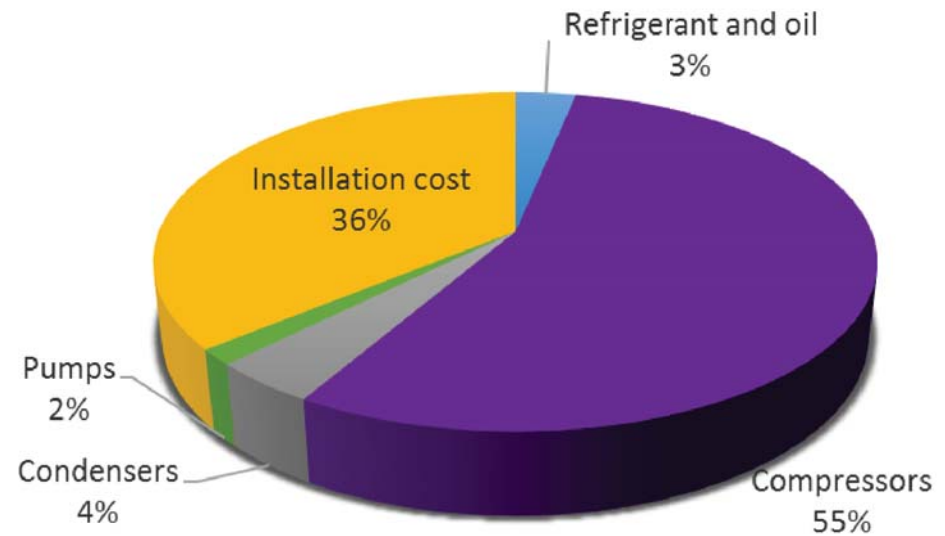
# CAPEX of air-cooled and water-cooled systems



Air-cooled system  
Total cost: £ 193,000



Water-cooled system  
Total cost: £ 200,000



**Costs of the heat rejection system increase the CAPEX**