# Predictive Modeling of risk factors in slaughterhouses using Low-cost inertial sensors

Thesis Defense

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#### Outline

- 1 Work-Related Muscukuloskeletal Disorders in Slaughterhouses
- 2 Thesis Hypothesis and Objectives
- 3 Methodology, Prototype & Experiments
- 4 Predictive Modeling & Decision Making
- 5 Conclusions
- 6 Energy, Infrastructure, City
- 7 Sustainable model use
- 8 Mathematical modelling, optimisation, case study

#### Section 1

# Work-Related Muscukuloskeletal Disorders in Slaughterhouses

## State of the Slaughterhouse Industry

(BMUB 2015/16)

- 1. Labour is a high pertcentage of the costs.
- 2. Fatigue and bad practices lead to injuries, absenteesim and costs.

#### Literature Review

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# Section 2

# Thesis Hypothesis and Objectives

# Section 3 Methodology, Prototype & Experiments

# Section 4 **Predictive Modeling & Decision Making**

# Section 5 Conclusions

Section 6 **Energy, Infrastructure, City** 

## Questions about Germany's Climate Action Plan 2050

(BMUB 2015/16)

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- 4. Which role do decentralised energy supply concepts play?
- 5. How can the electricity and heating/cooling markets be more closely integrated [...]?

## Perspective

Disciplines for investigating energy topics

#### **Physics**

Theoretical feasibility (Natural laws)

#### **Engineering**

Technical feasibility (Technologies)

#### **Economy**

Economic feasibility (Funding)

#### Society

Social feasibility (Decision space)

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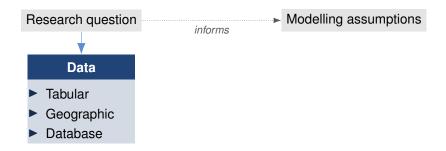
#### **Techno-economic modelling**

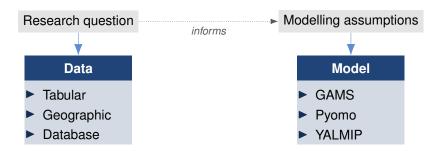
How much energy? For how much?

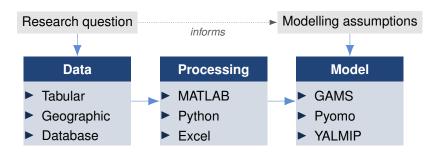
# Section 7 Sustainable model use

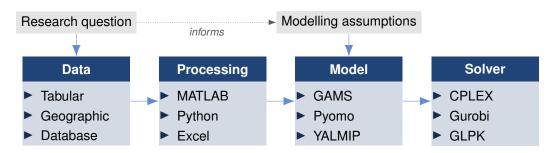
Research question

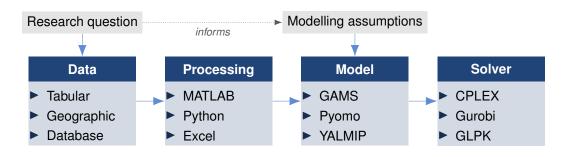
Research question Modelling assumptions

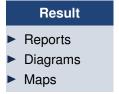


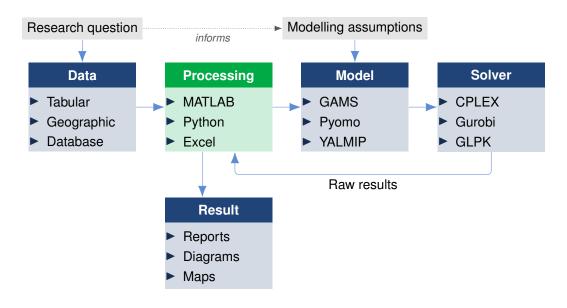


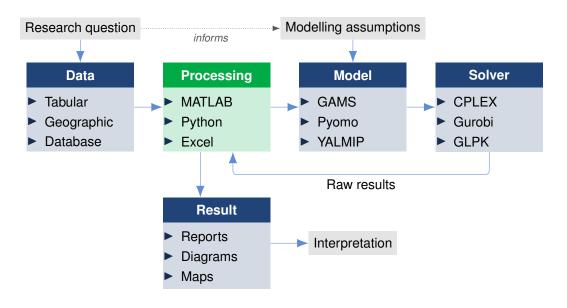


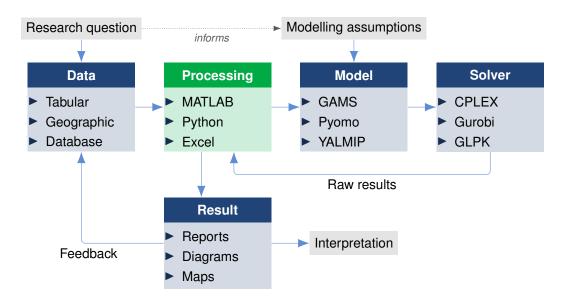








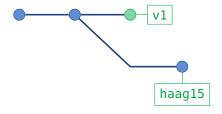


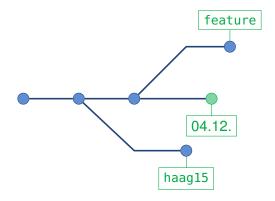


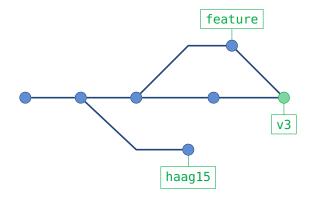


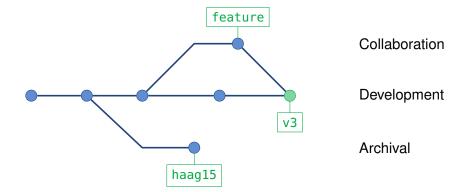


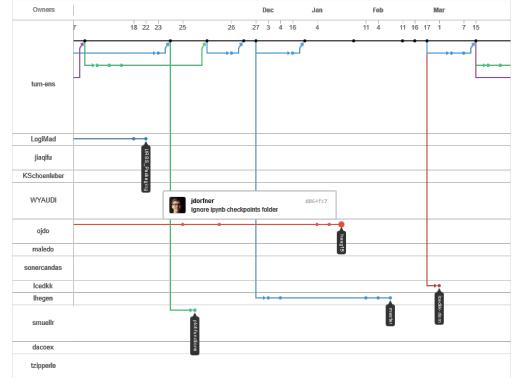




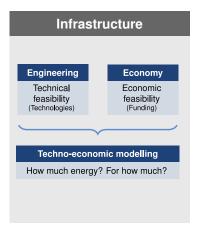


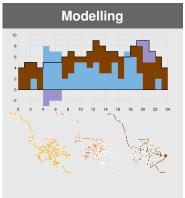


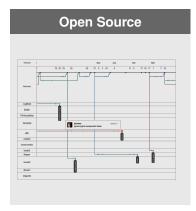




#### Conclusion



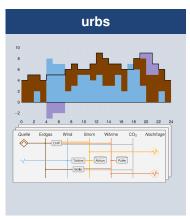




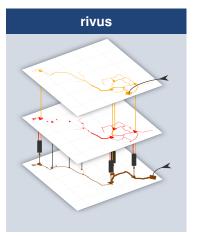
# Section 8

## Mathematical modelling, optimisation, case study

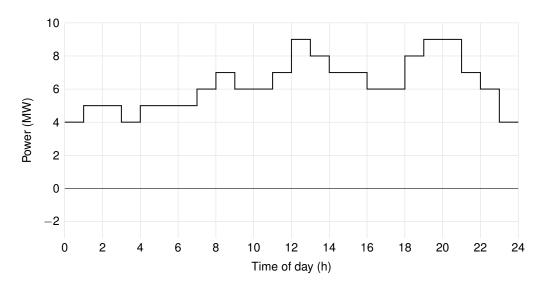
#### Model overview

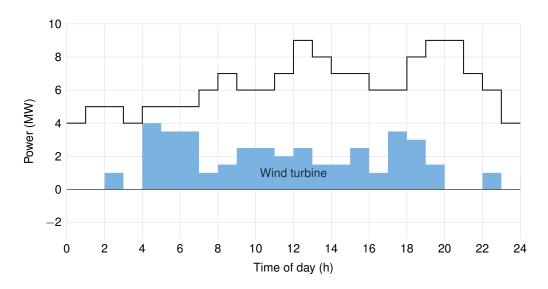


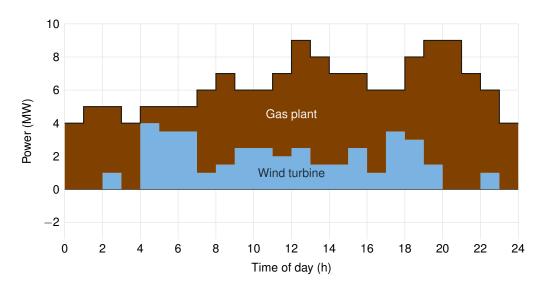
https://github.com/tum-ens/urbs

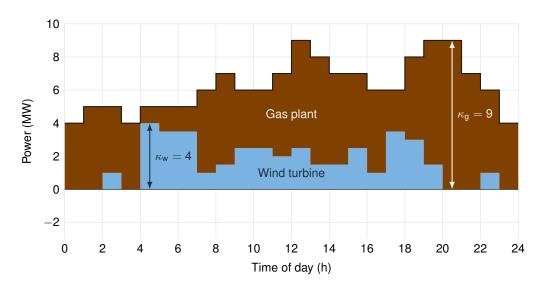


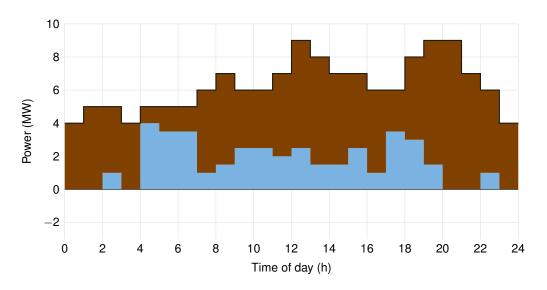
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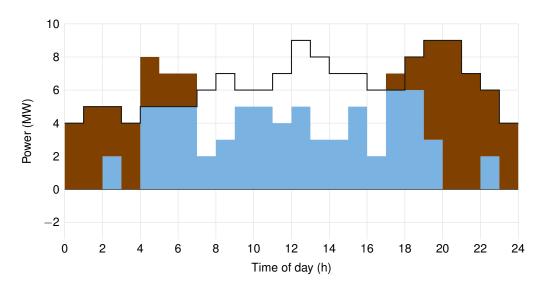


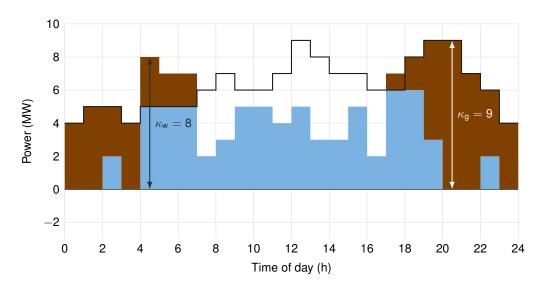


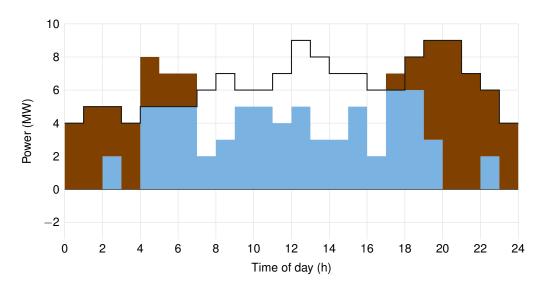


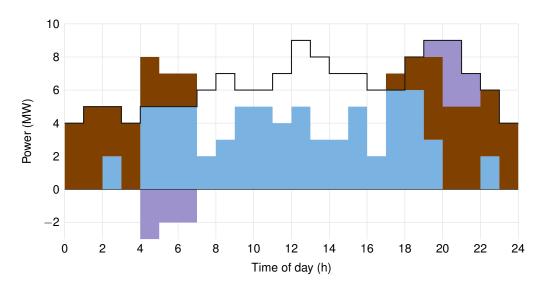


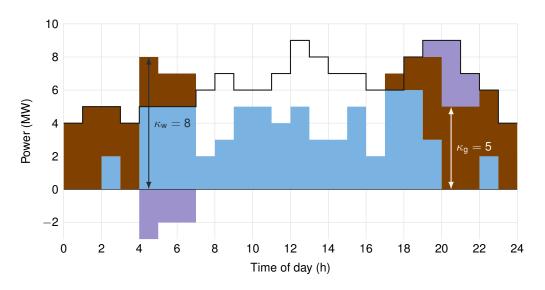


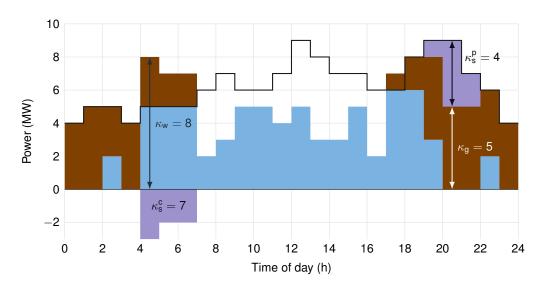












Sets 
$$t \in T, p \in P, s \in S, ...$$
  
Parameters  $d_t$ 

Sets 
$$t \in T$$
,  $p \in P$ ,  $s \in S$ , ...  
Parameters  $d_t$ ,  $k_p^{fix}$ ,  $k_s^{fix,c}$ ,  $k_s^{fix,p}$ 

Sets 
$$t \in T$$
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```
\begin{array}{ll} \text{Sets} & t \in \mathcal{T}, \ p \in P, \ s \in \mathcal{S}, \ \dots \\ \\ \text{Parameters} & d_t, \ k_p^{\text{fix}}, \ k_s^{\text{fix,c}}, \ k_s^{\text{fix,p}}, \ k_p^{\text{var}}, \ k_s^{\text{var}}, \ \dots \\ \\ \text{Variables} & \kappa_p, \ \kappa_s^c, \ \kappa_s^p \end{array}
```

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$$t \in \mathcal{T}, \ p \in P, \ s \in S, \ldots$$

Parameters  $d_t, \ k_p^{\text{fix}}, \ k_s^{\text{fix,c}}, \ k_s^{\text{fix,p}}, \ k_p^{\text{var}}, \ k_s^{\text{var}}, \ldots$ 

Variables  $\kappa_p, \ \kappa_s^c, \ \kappa_s^p, \ \epsilon_{pt}, \ \epsilon_{st}^{\text{in}}, \ \epsilon_{st}^{\text{out}}, \ \epsilon_{st}^{\text{con}}, \ldots$ 

Objective  $\min \sum_{p \in P} \left( k_p^{\text{fix}} \kappa_p + \sum_{t \in \mathcal{T}} k^{\text{var}} \epsilon_{pt} \right) + \sum_{s \in S} \left( k_s^{\text{fix,c}} \kappa_s^c + k_s^{\text{fix,p}} \kappa_s^p + \sum_{t \in \mathcal{T}} k_s^{\text{var}} \left( \epsilon_{st}^{\text{in}} + \epsilon_{st}^{\text{out}} \right) \right)$ 

Constraints s.t.  $\forall t \in \mathcal{T}$ :  $\sum_{p \in P} \epsilon_{pt} + \sum_{s \in S} \left( \epsilon_{st}^{\text{out}} - \epsilon_{st}^{\text{in}} \right) = d_t$ 

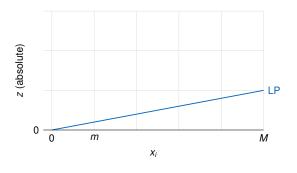
## Standard form of linear optimisation problems (LP)

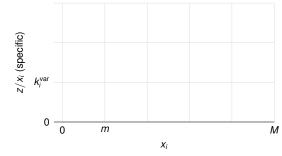
#### Generic form

$$\min_{\boldsymbol{x}} z = \boldsymbol{c}^\mathsf{T} \boldsymbol{x}$$

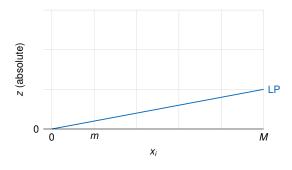
s.t. 
$${\it Ax} \le {\it b}$$

with 
$$\mathbf{x} \in \mathbb{R}^n$$
,  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{b} \in \mathbb{R}^m$ ,  $\mathbf{c} \in \mathbb{R}^n$ .

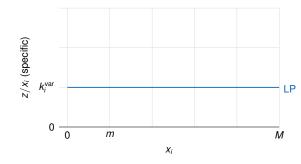




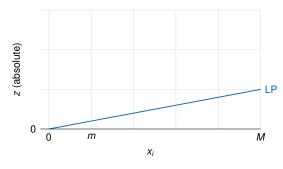
$$LP \quad z = k_i^{\text{var}} x_i \\
x_i \le M$$



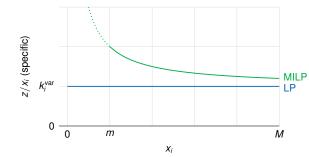
LP 
$$z = k_i^{\text{var}} x_i$$
  
 $x_i \leq M$ 



LP 
$$\frac{z}{x_i} = k_i^{\text{var}} \equiv \text{const}$$



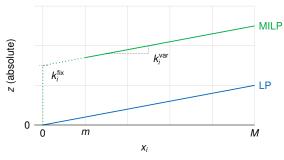
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LP 
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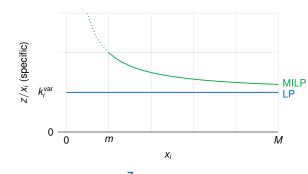
MILP  $\frac{z}{x_i} = k_i^{\text{var}} + \frac{k_i^{\text{fix}}}{x_i}$ 

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LP  $z = k_i^{\text{var}} x_i$  $\chi_i < M$ 

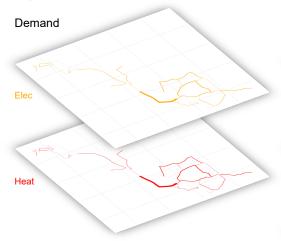
MILP  $z = k_i^{fix} y_i + k_i^{var} x_i$  $y_i \in \{0, 1\}$  $m y_i < x_i < M y_i$ 



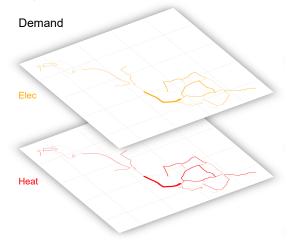
LP 
$$\frac{z}{x_i} = k_i^{\text{var}} \equiv \text{const}$$
  
MILP  $\frac{z}{x_i} = k_i^{\text{var}} + \frac{k_i^{\text{fix}}}{x_i}$ 

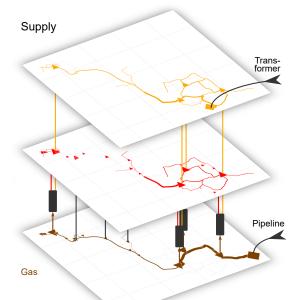
$$MILP \quad \frac{2}{x_i} = k_i^{\text{var}} + \frac{\kappa_i}{x_i}$$

### rivus



### rivus

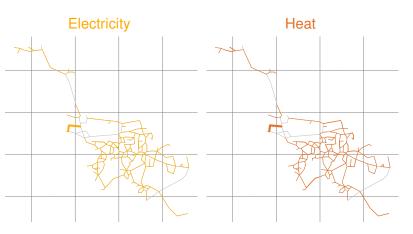








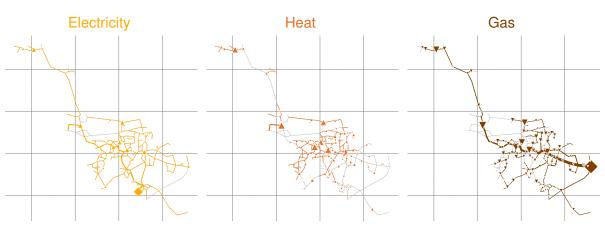
## Input data rivus



Light industry (Schletter) biggest single consumer

https://github.com/tum-ens/rivus/data/haag15

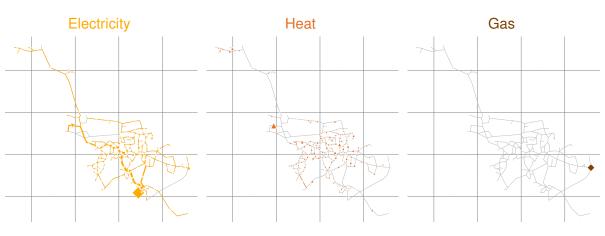
## Result **rivus** – Capacities in scenario base



Full networks for electricity and gas, several local heating networks

https://github.com/tum-ens/rivus/runhg15.py:scenario\_no\_electric\_heating()

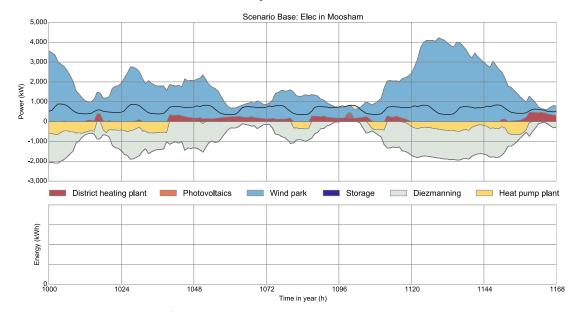
## Result **rivus** – Capacities in scenario future



Strong electricity grid, no gas network, only heat pumps

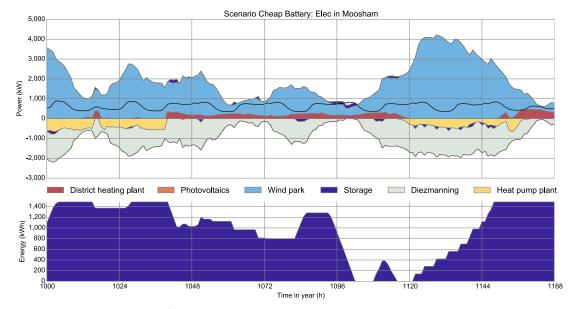
https://github.com/tum-ens/rivus/runhg15.py:scenario\_renovation()

## Result **urbs** – 1 week electricity in scenarios base



https://github.com/ojdo/urbs/tree/haag15/rivhg15.py:scenario\_base()

## Result **urbs** – 1 week electricity in scenario cheap battery



https://github.com/ojdo/urbs/tree/haag15/rivhg15.py:scenario\_cheap\_battery()