

Energy Flow Modeling of Night Storage Heaters and Its Usage for Sector Coupling

Marc Gennat, Raschad Damati, Arne Graßmann

Hochschule Niederrhein – University of Applied Science

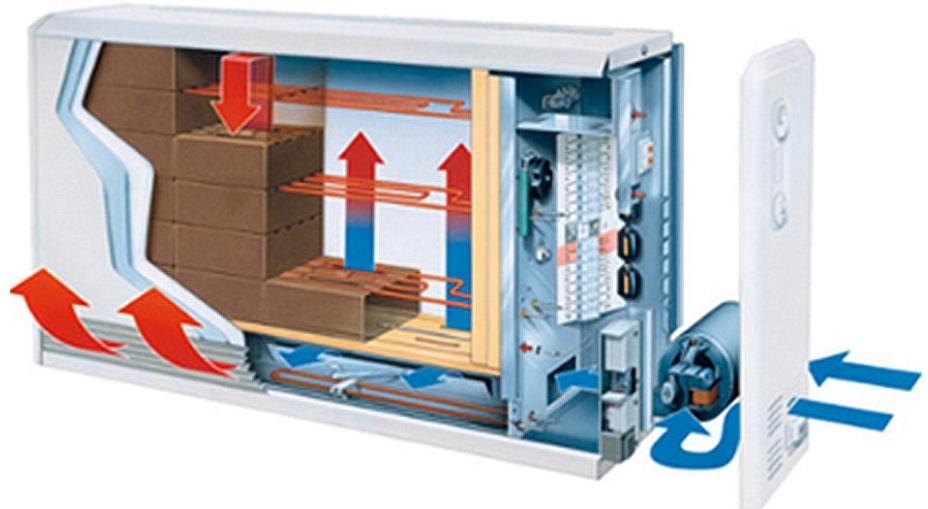
8th June 2021 | 1st Online-IAEE Conference

Outline

- **Introduction – Night Storage Heaters**
- **Challenge – Modeling the Room Temperature**
- **Test Setup Room**
- **Energy Transport**
- **Parameter Estimation**
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- **Outlook**

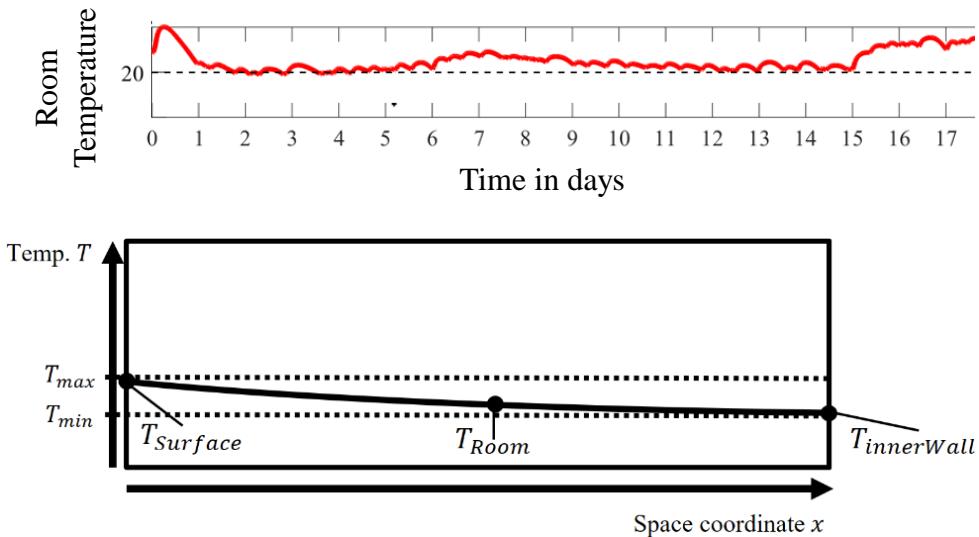
Introduction – Night Storage Heaters

- Widely build in Germany in the 1950th and 1960th
- Cheap electric energy was available during the night
- Today still in use in Germany
- Approx. 1.5 million apartments
- Quite expensive compared to heating with fossil fuels due to taxes and fees
- If electric energy has a significant share of coal and gas, NSH will pollute much more CO₂ than natural gas

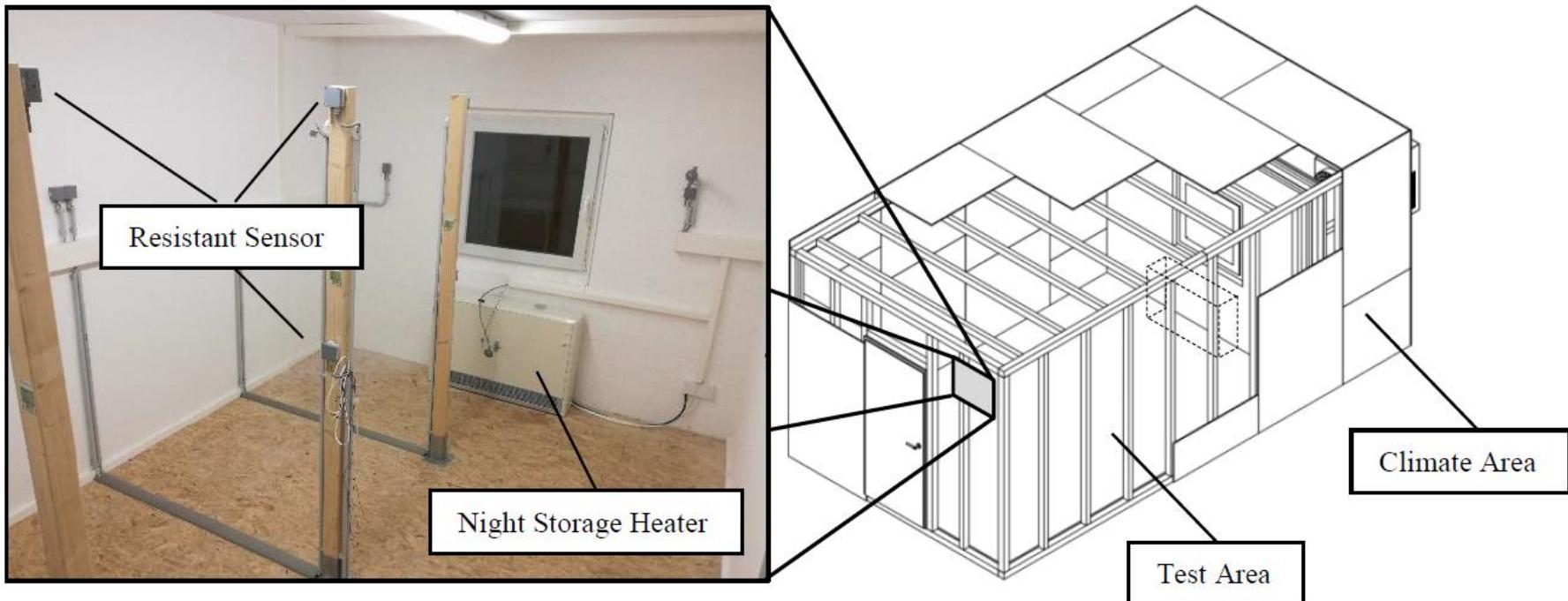


Challenge – Modeling the Room Temperature

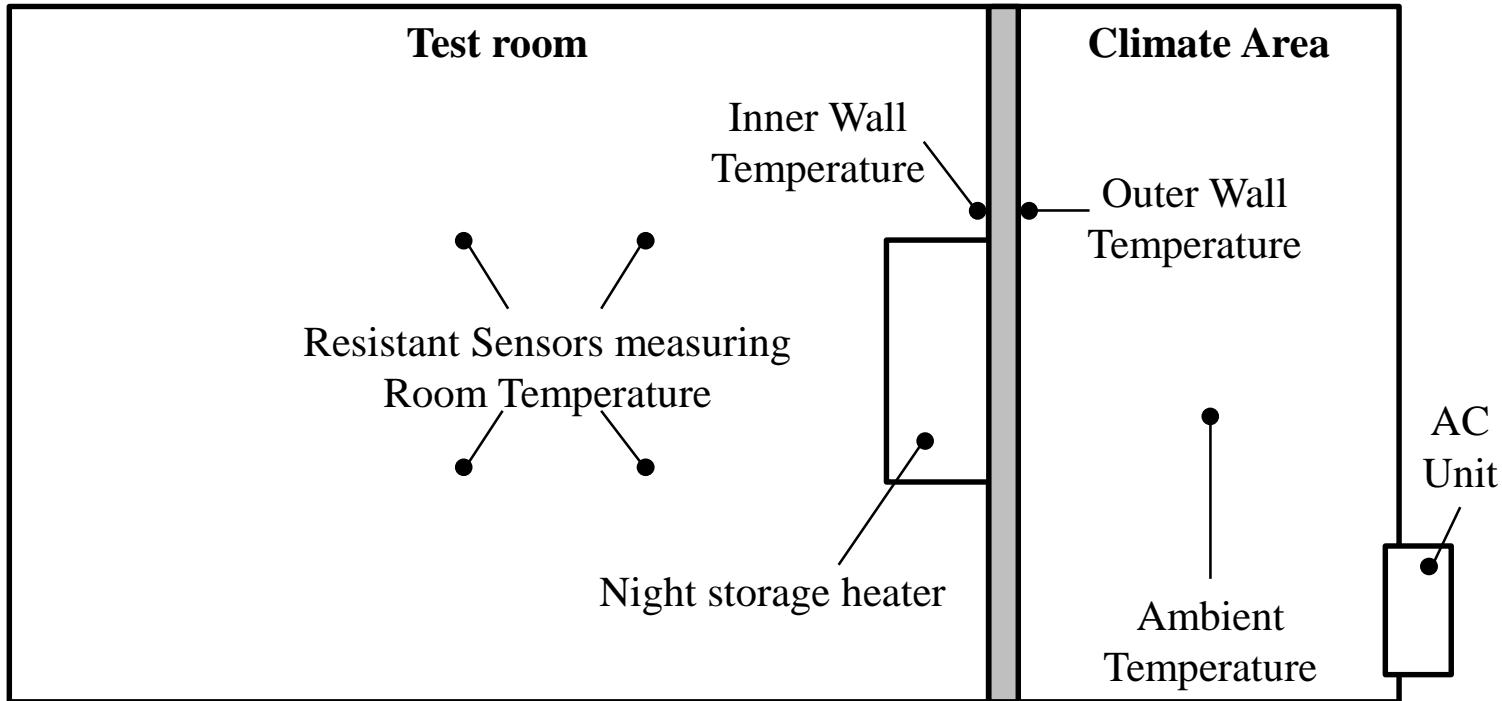
- Volatile Electric Energy Prices could help to
 - improve the home comforts temperatures with NSH, and
 - reduce heating costs.
- With a room temperature model with two input parameters
 - Electric Power Supply
 - Ambient temperaturethe room temperature could be simulated over a long period of time.



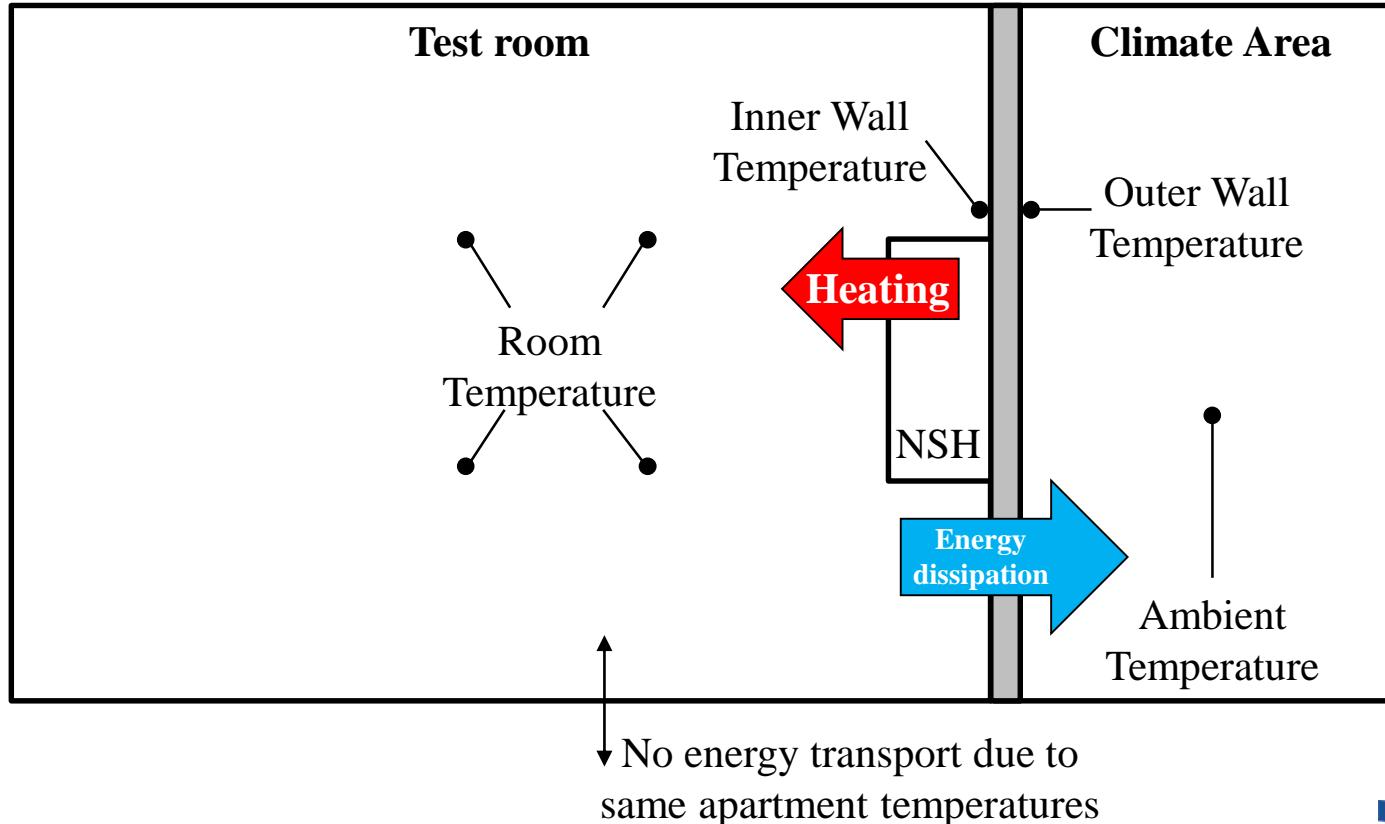
Test Room Setup



Test Setup



Test Setup



Energy Transport

Fourier's heat conduction law

$$\frac{\partial T}{\partial t} = \alpha \cdot \nabla^2 T = \alpha \cdot \operatorname{div}(\operatorname{grad} T)$$

Thermal conductivity equation

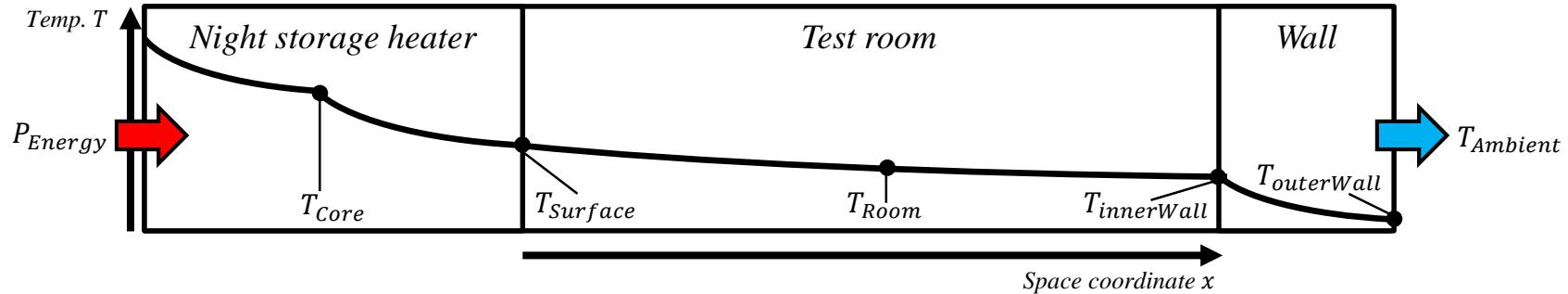
$$\dot{q} = -k \nabla T = -k (T_1 - T_2) \text{ with } T_1 > T_2,$$

$$m c \frac{dq}{dt} = U_1 A_1 T_1 - U_2 A_2 T_2.$$

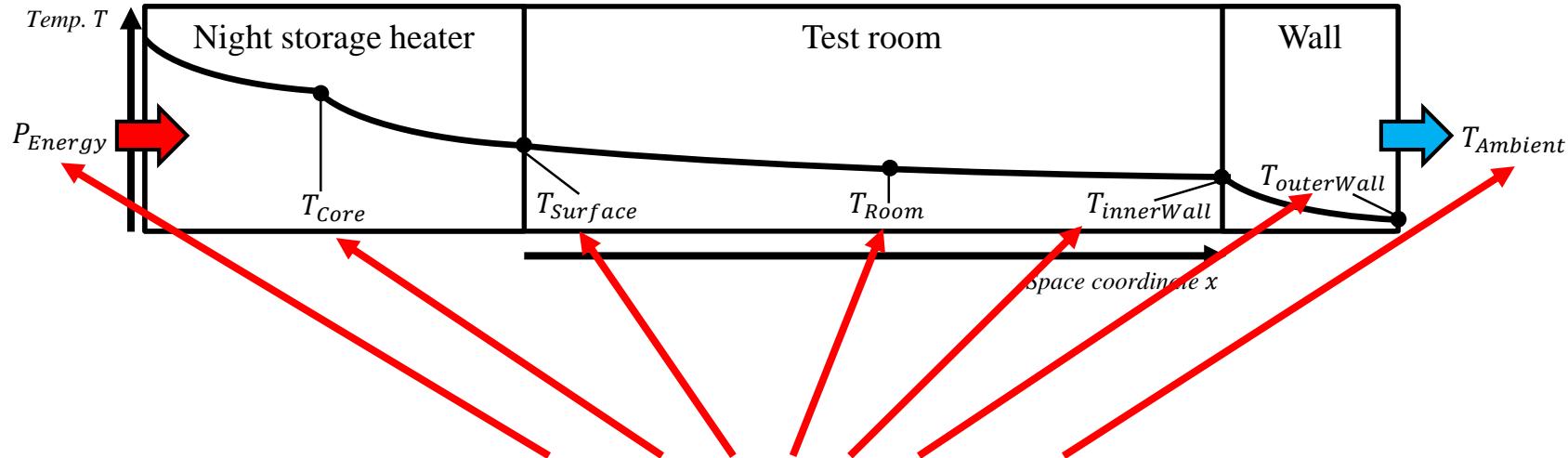
Relaxation leads to a parametrized equation

$$\frac{dT}{dt} = \frac{U_1 A_1}{m c \Delta S} T_1 - \frac{U_2 A_2}{m c \Delta S} T_2 \Leftrightarrow \frac{dT}{dt} = p_1 \cdot T_1 - p_2 \cdot T_2 \text{ with } T_1 > T > T_2,$$

Energy Transport

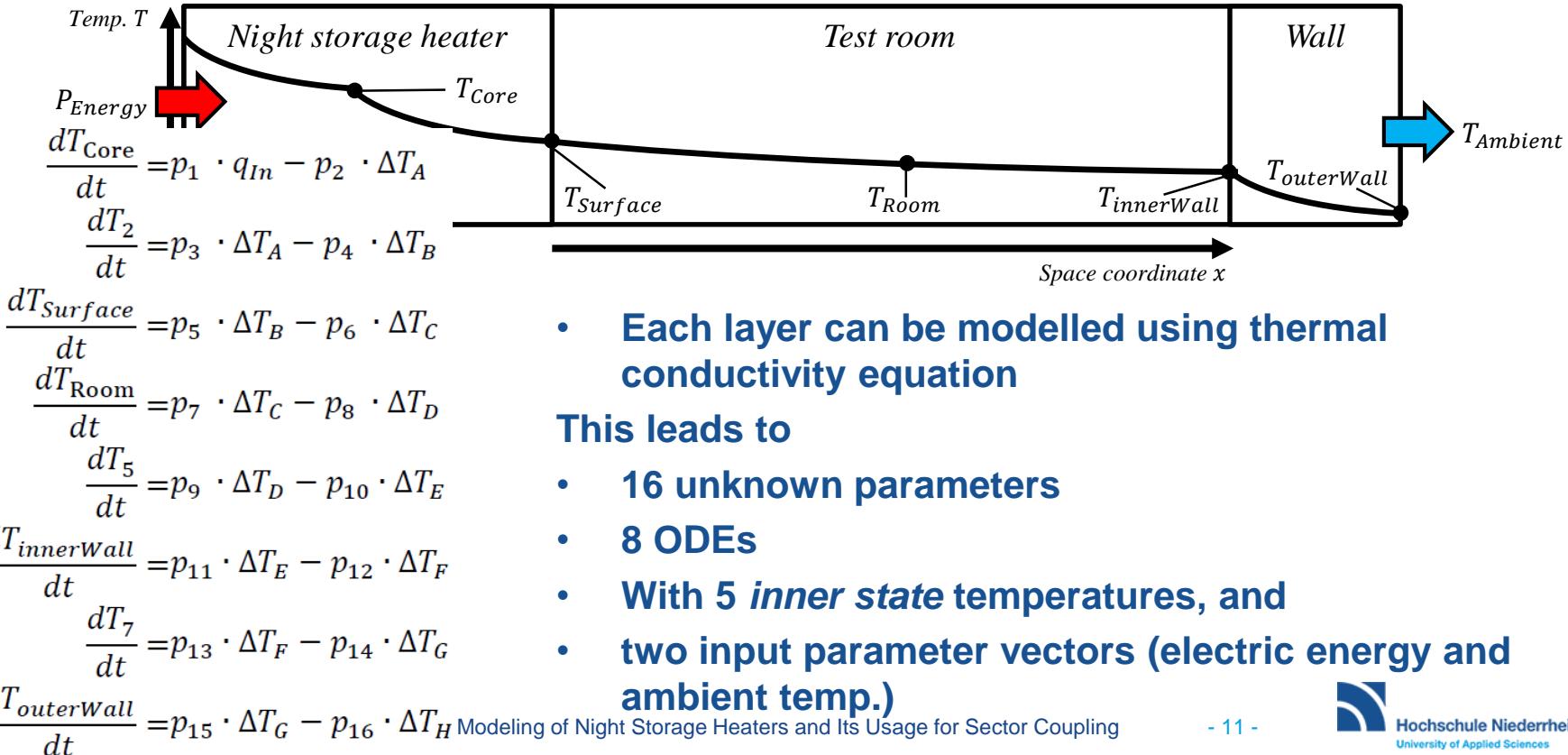


Energy Transport



Measurable and known temperatures and
input parameters for parameter estimation

Energy Transport

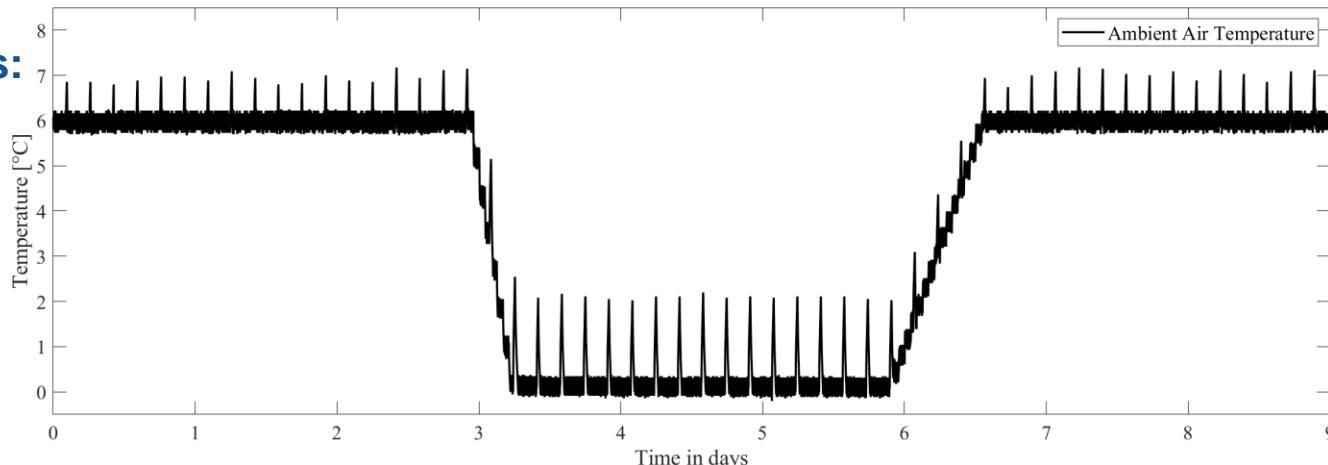


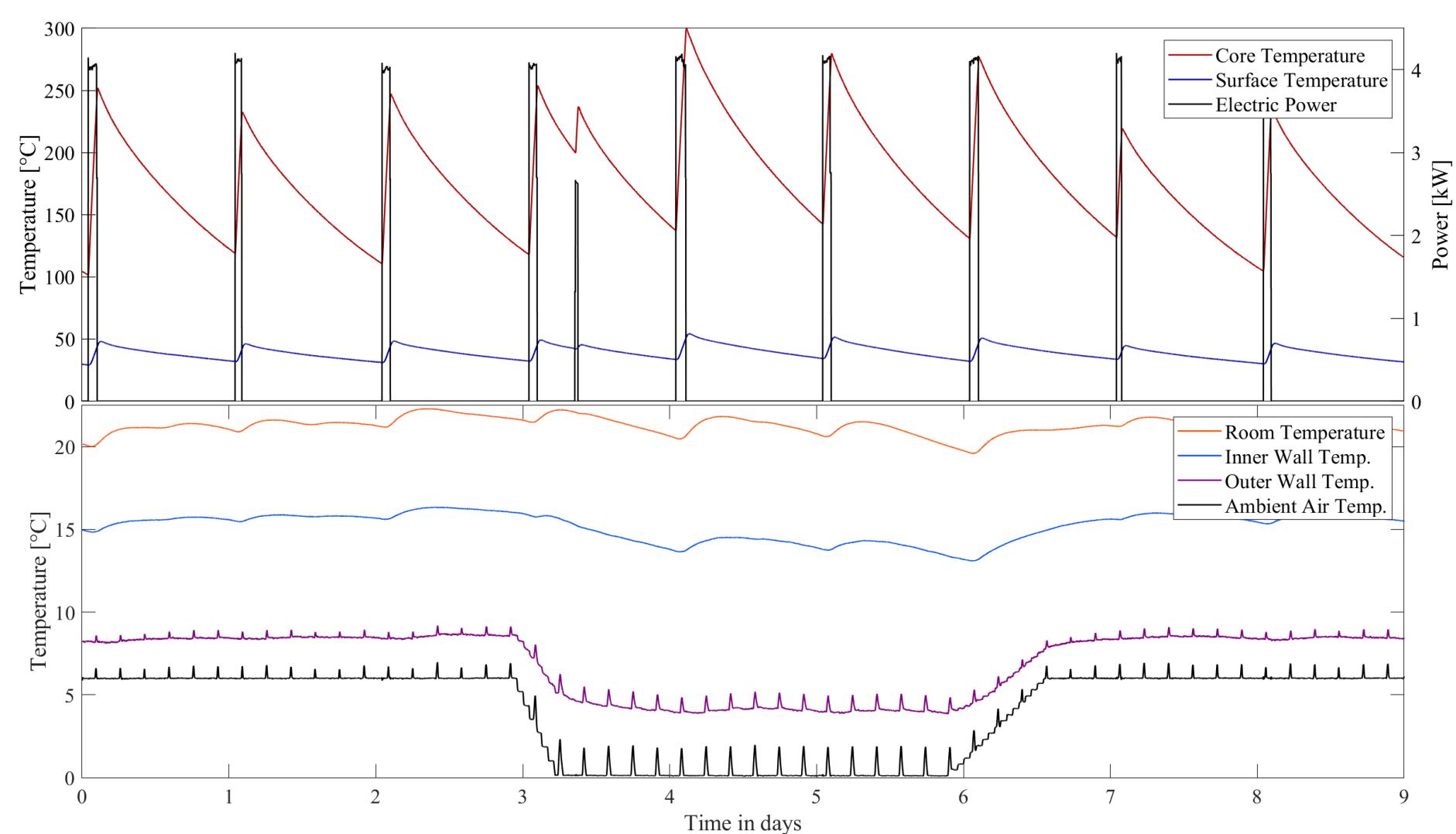
Reference Scenario for Parameter Estimation

- Nine day test run
- Ambient temperature was predefined
- Standard heating program was applied:
Electric Heating was allowed during night times

Measured temperatures:

- Core
- Surface
- Room
- Inner Wall
- Outer Wall





Parameter Estimation

State space representation

$$\dot{\underline{x}}(t) = \underline{A} \underline{x}(t) + \underline{B} \underline{u}(t), \underline{x}(0) = \underline{x}_0, \underline{x} \in \mathbb{R}^8, \underline{A} \in \mathbb{R}^{8 \times 8}, \underline{B} \in \mathbb{R}^{8 \times 2}, \underline{u} \in \mathbb{R}^2.$$

Algebraic solution

$$\underline{x}(t) = e^{\underline{A}t} \underline{x}_0 + e^{\underline{A}t} \int_0^t e^{-\underline{A}\tau} \underline{B} \underline{u}(\tau) d\tau$$

Expression size was not computable, thus, Matlab's ODE45 solvers were used.

Parameter Estimation

ODEs to be solved

$$\dot{x}_1(t) = p_1 \cdot P_{Energy}(t) - p_2(x_{Core}(t) - x_2(t))$$

$$\dot{x}_2(t) = p_3(x_{Core}(t) - x_2(t)) - p_4(x_2(t) - T_{Surface}(t))$$

$$\dot{x}_{Surface}(t) = \dot{x}_3(t) = p_5(x_2(t) - x_{Surface}(t)) - p_6(x_{Surface}(t) - T_{Room}(t))$$

$$\dot{x}_{Room}(t) = \dot{x}_4(t) = p_7(T_{Surface}(t) - x_{Room}(t)) - p_8(x_{Room}(t) - x_5(t))$$

$$\dot{x}_5(t) = p_9(x_{Room}(t) - x_5(t)) - p_{10}(x_5(t) - T_{innerWall}(t))$$

$$\dot{x}_{innerWall}(t) = \dot{x}_6(t) = p_{11}(x_5(t) - x_{innerWall}(t)) - p_{12}(x_{innerWall}(t) - x_7(t))$$

$$\dot{x}_7(t) = p_{13}(x_{innerWall}(t) - x_7(t)) - p_{14}(x_7(t) - T_{outerWall}(t))$$

$$\dot{x}_{outerWall}(t) = \dot{x}_8(t) = p_{15}(x_7(t) - x_{outerWall}(t)) - p_{16}(x_{outerWall}(t) - T_{Ambient}(t))$$

Parameter Estimation

Unknown parameter vector \underline{p}^* is computable by least squares estimation

$$\underline{p}^* = \underset{\underline{p}}{\operatorname{argmin}} \sum_{k=1}^{k_{max}} \left(x_{Measure}(k) - x_{Model}(k; \underline{p}) \right)^2,$$

whereby

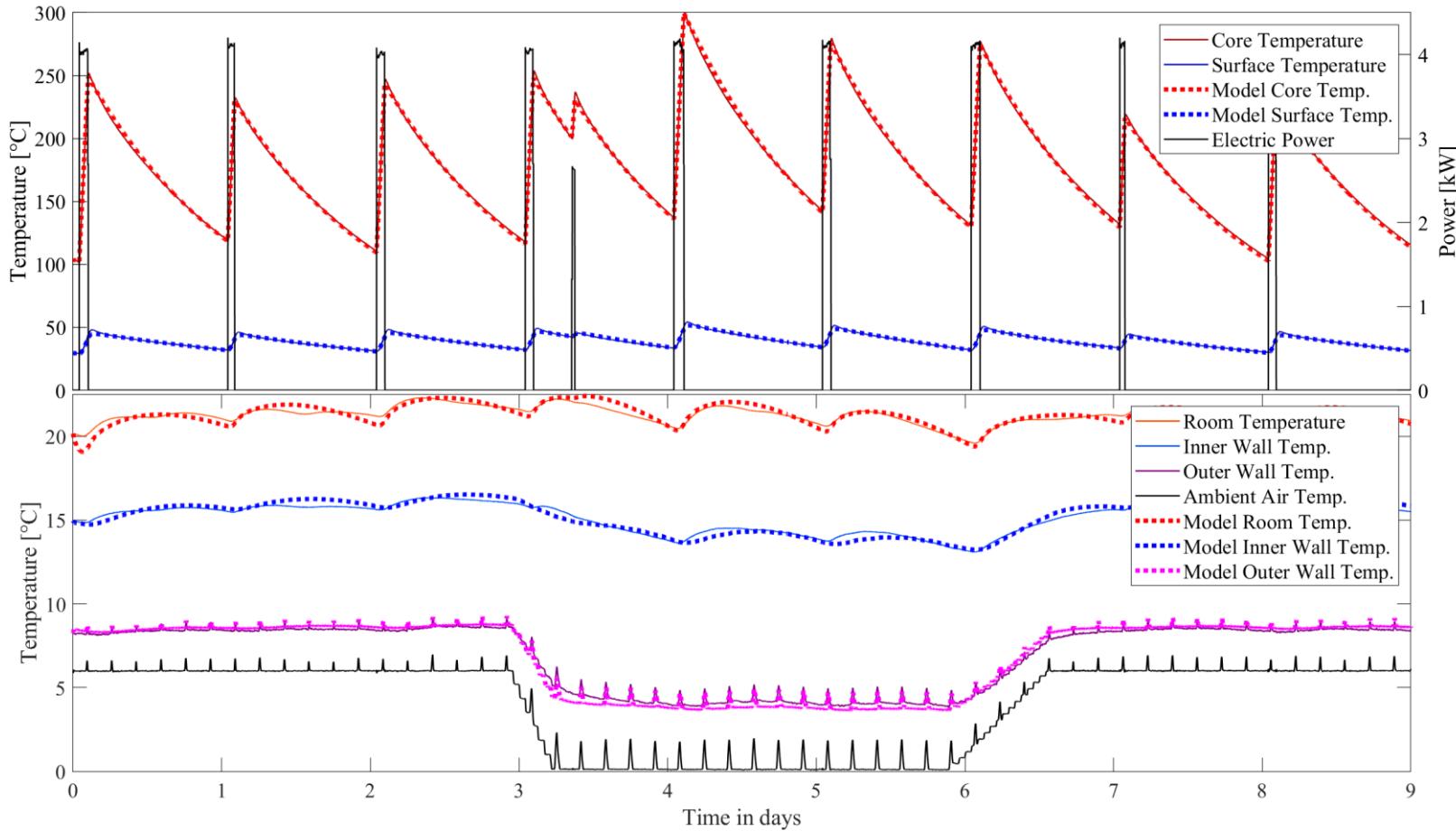
$x_{Measure}(k)$ represents temperature measurements of five metering points, and $x_{Model}(k; \underline{p})$ is computed by solving the above mentioned ODEs. k counts the second wise timestamp over nine days from 1 to 777,600.

In general, nonlinear programming does not provide the global minimum, thus, the given result represents a local minimum.

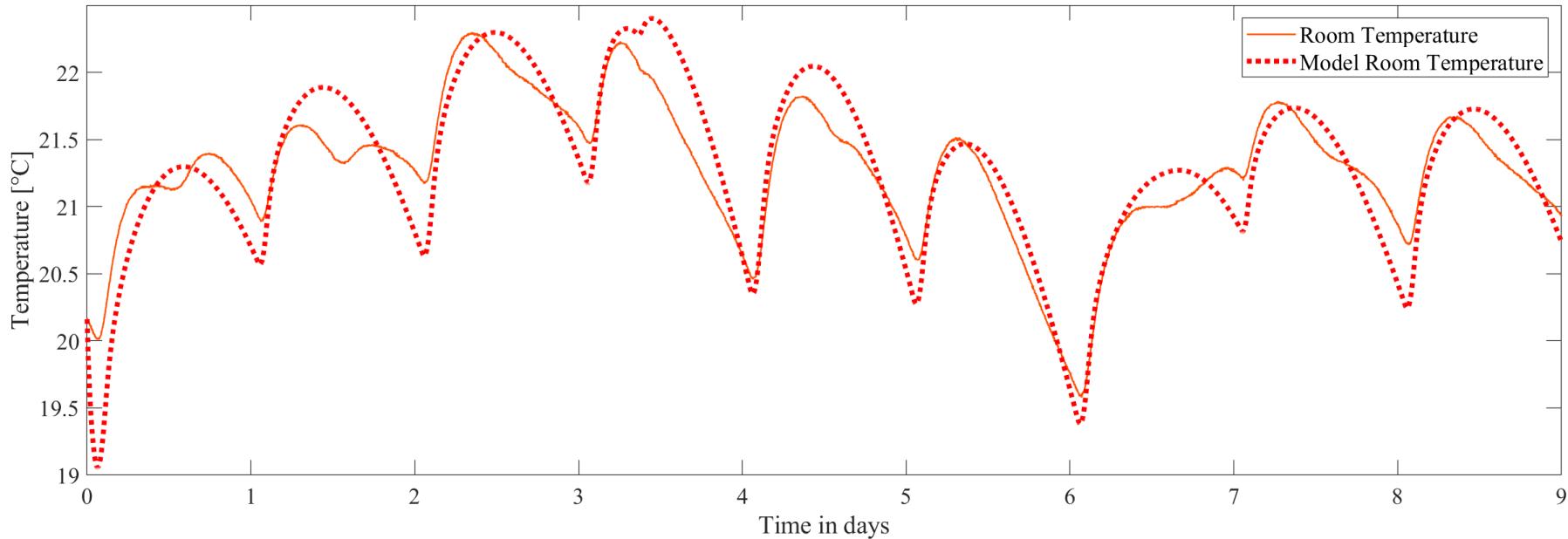
Parameter Estimation

$$\begin{aligned}\underline{p}^* = & \underset{\underline{p}}{\operatorname{argmin}} \sum_{k=1}^{k_{\max}} \left(x_{Measure,Core}(t_k) - x_{Core} \left(t_k; \underline{p}, P_{Energy}, T_{Ambient} \right) \right)^2 \\ & + \sum_{k=1}^{k_{\max}} \left(x_{Measure,Surface}(t_k) - x_{Surface} \left(t_k; \underline{p}, P_{Energy}, T_{Ambient} \right) \right)^2 \\ & + 2 \cdot \sum_{k=1}^{k_{\max}} \left(x_{Measure,Room}(t_k) - x_{Room} \left(t_k; \underline{p}, P_{Energy}, T_{Ambient} \right) \right)^2 \\ & + \sum_{k=1}^{k_{\max}} \left(x_{Measure,innerWall}(t_k) - x_{innerWall} \left(t_k; \underline{p}, P_{Energy}, T_{Ambient} \right) \right)^2 \\ & + \sum_{k=1}^{k_{\max}} \left(x_{Measure,outerWall}(t_k) - x_{outerWall} \left(t_k; \underline{p}, P_{Energy}, T_{Ambient} \right) \right)^2\end{aligned}$$

Results



Results



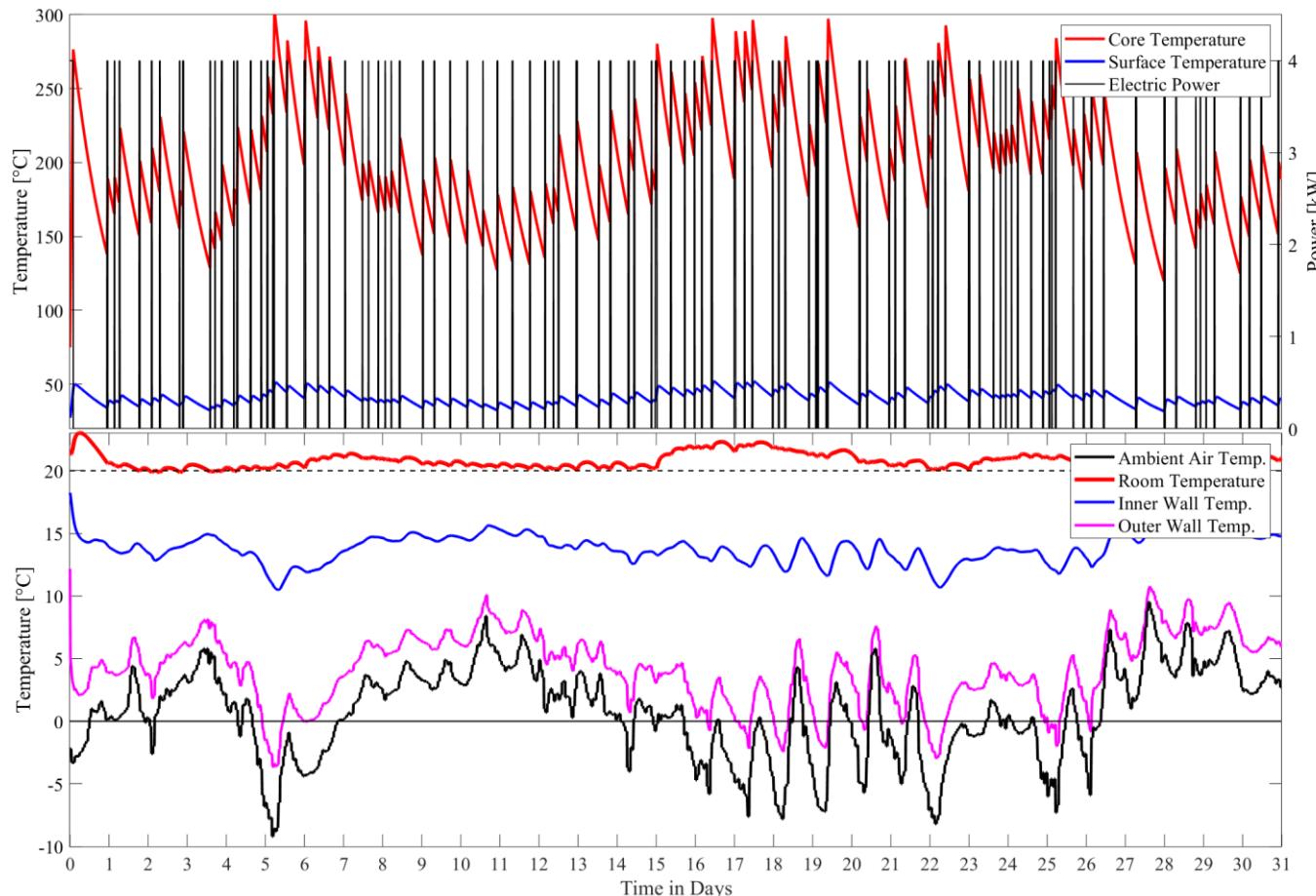
Maximum temperature deviation

0.96 Kelvin

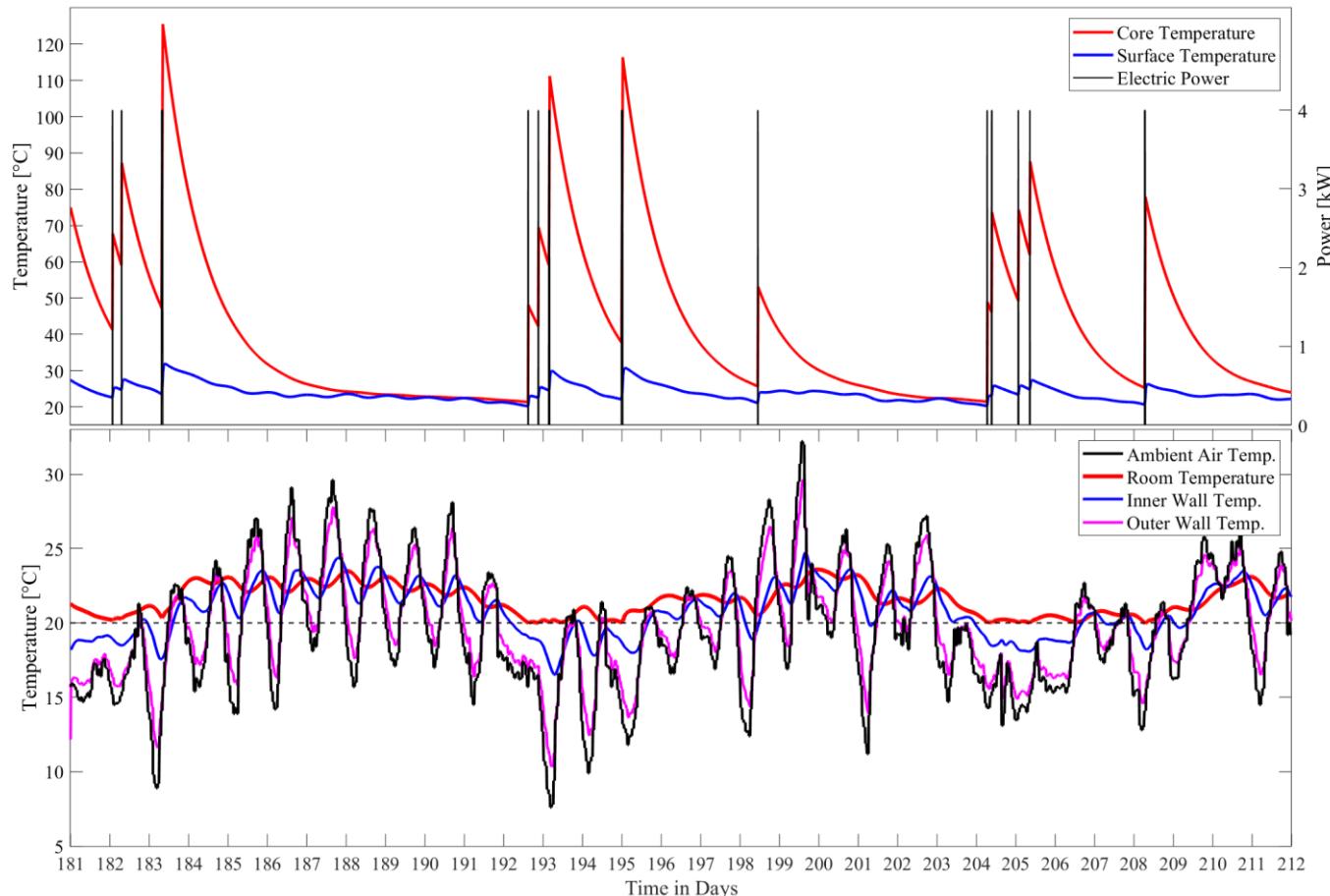
Average temperature deviation

0.21 Kelvin

Results - January Ambient Temperature Test Simulation



Results - July Ambient Temperature Test Simulation



Outlook

- Apartment comfort temperature is within 20 to 22°C
- It is controllable with a reliable temperature model
- Energy expenses could be minimized using fluctuating market prices
- Moreover, net-grid-stability could be improved
- Modeling real apartment's energy dissipation has to be done to apply this method to real world applications

Thank you for your attention!



50 Jahre Hochschule Niederrhein



Hochschule Niederrhein

University of Applied Sciences

SWK E²

Institut für Energietechnik und
Energiemanagement

Institute of Energy Technology and
Energy Management