



Thermal Modeling for Buildings

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Motivation

- Buildings account for approximately 40% of world energy use -> 21% of greenhouse gas emissions
- In the U.S., buildings contribute 1 billion metric tons of greenhouse gas emissions.
- On an annual basis, buildings Consume
 - 39% of total U.S. Energy
 - 68% of U.S Electricity
 - 38% of U.S. carbon dioxide (primary greenhouse gas associated with climate change)
 - 25% of the nitrogen oxides found in the air
- The only energy end-use sector showing growth in energy intensity

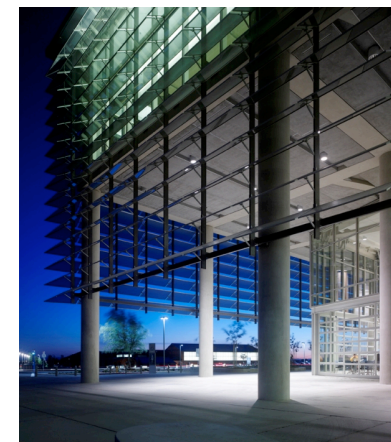
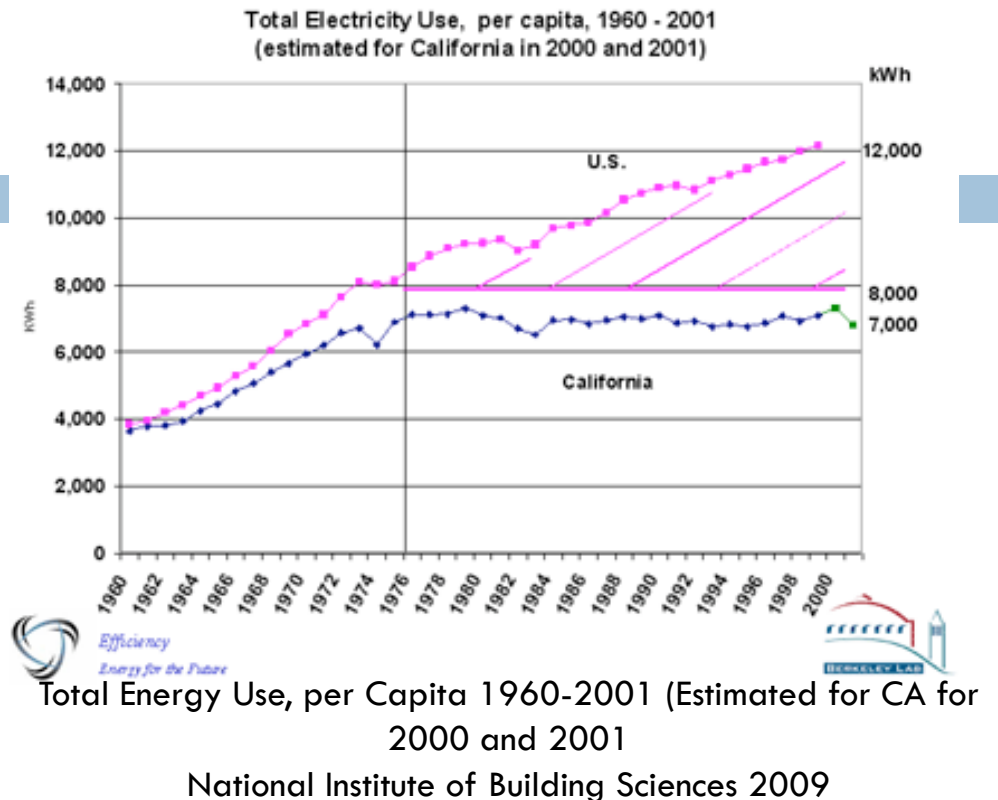
Motivation



- According to the National Institute of Building Sciences, an Intelligent Buildings must seek to:
 - ▣ Reduce heating, cooling and lighting loads through climate-responsive design and conservation practices
 - ▣ Employ renewable energy sources (daylighting, passive solar heating, photovoltaics, geothermal, and groundwater cooling)
 - ▣ Specify efficient HVAC and lighting systems that consider part-load conditions and utility interface requirements
 - ▣ Optimize building performance by employing energy modeling programs and optimize system control strategies
 - ▣ Monitor project performance
- Energy codes provide minimum building requirements that are cost-effective in saving energy

Motivation

- The UC Merced campus: full scale test-bed for advanced energy management in campus buildings and systems
- Test-bed facility involves design, implementation and evaluation of a sensor network prototype
- Sensors include:
 - Light
 - Temperature
 - Humidity
 - Light intensity
 - Mobility patterns of humans inside building
- New measurements can help improve and design control Energy Management and Control Systems (EMCS)



UC Merced Campus

Problem Statement



- In order to investigate how new technologies can help improve energy usage at the UC-Merced buildings, a scalable thermal model of the building needs to be created, verified and validated.
- Scalability is important when analyzing control systems for large buildings. Relevant to hybrid control system models
- Objectives
 - ▣ Create a scalable model (in Simulink) of building thermodynamics
 - ▣ Model UC Merced building
 - ▣ Validate model using reference data of UC Merced

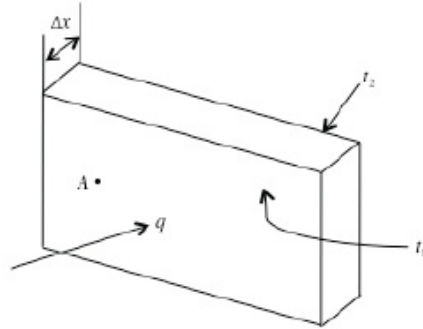
Related Work

- Maasoumy, Holcomb [Fall 2008]. High Performance Building. Simulink modeling and control. More emphasis on control systems
- Schijndel [2003]. FemLab (Physical system simulator) and Simulink are evaluated as solvers for HVAC problems based on PDEs. Proposes the use in combination of both system, but doesn't address scalability.
- Mendes, et al [2001]. Mathematical model applied to both building thermal analysis and control system design using MatLab and Simulink. Approach incorporates multilayer walls, but does not address scalability.
- Felgner, et al [2002]. Modelica is used. Modelica has libraries for building thermal characteristics. Paper makes a good argument about the power of Modelica. However, feedback from Control Engineering at the ME department reveals that it is not an easy tool to use.
- Others: Address in detail HVAC systems, but do not address plant modeling or control systems.

Heat Transfer Basics

□ Conduction

$$q = kA \frac{t_1 - t_2}{\delta x}$$



□ Convection

$$\frac{q}{A} = h(t_s - t_f)$$

□ Radiation

$$E = \epsilon\sigma T^4$$

Nomenclature

C_p	Specific Heat
ρ	Density
t	Temperature, K
τ	Time, s
m	Mass, kg
q	Heat flow
h	Heat transfer coefficient of the material
k	Conduction coefficient
L	Length of conductor, m
A	Area, m^2

□ Heat Storage

- Heat capacity describes how much energy is required to increase the temperature by a specified amount. Proportional to the mass of the object

$$mC_p \frac{dt}{d\tau} = q$$

Heat Transfer Basics: Thermal Circuits

- Useful concept in the representation of thermal transfer
- Thermal circuit is a representation of the resistance to heat flow as though it were a resistor
- Heat storage elements can be represented as capacitors
- Temperature can be represented as potential

- Network nodal analysis, the following must be satisfied at each node

$$\sum_j \frac{t_j - t_i}{R_{ij}} + q_i = 0$$

- Where q_i represents heat added to the node by means other than surface convection. If internal heat is present, the q_i 's are known

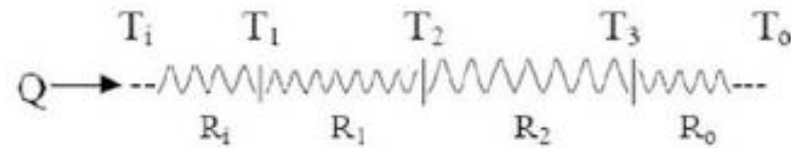
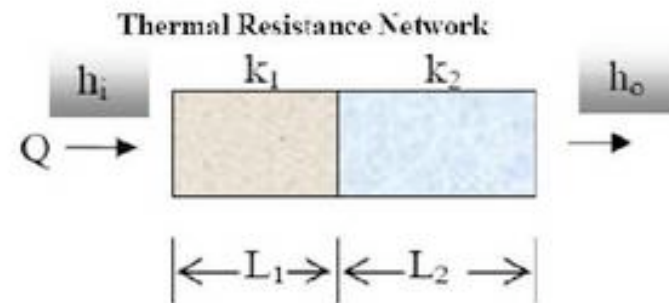
Heat Transfer Basics: Thermal Circuits

- For conduction the resistance is:

$$R = \frac{\delta x_{ij}}{kA_{k_{ij}}}$$

- For convection it is:

$$R = \frac{1}{h_{ij}A_{c_{ij}}}$$



Proposed Approach



- Simscape from MathWorks
 - ▣ Extends the Simulink with tools for modeling systems spanning mechanical, electrical, hydraulic and other physical domains and physical networks
 - ▣ For our approach, use the electrical physical components to simulate a room and investigate its scalability
- CHALLENGES
 - ▣ No UC Merced model data for validation
 - ▣ Mitigation: model using other tool (ComSol, Modelica or PDE Matlab)
 - ▣ Other tools have higher learning curve. Verified the Model against Matlab PDEs

Model: Analytical



- Since building is a complex system, a complete theoretical approach is impractical.
- Assumptions:
 - ▣ Air in the zone is fully mixed. Temperature distribution is uniform and the dynamics can be expressed in a lump capacity model
 - ▣ Effect of each wall is the same
 - ▣ Ground and roof have no effect on the zone temperature
 - ▣ The density of the air is assumed to be constant and is not influenced by changing the temperature and humidity ratio of the zone

Model: Description



- State variables:
 - ▣ Zone temperature and wall temperatures
- People, lights and extreme weather conditions are uncontrolled inputs.
- Input variables:
 - ▣ Air flow rate for each zone

Model: Analytical Model

$$\frac{dt_i}{d\tau} = \frac{1}{C_w} \left[q_{rad} + \frac{t_o - t_i}{R_{iw_i}} + \frac{t_a - t_i}{R_{ow_i}} \right] \quad (1)$$

Where

$$C_w = LA\rho_w C_{pw}$$

$$R_{iw_i} = \frac{R_{w_i}}{2} + R_{in} \quad (3)$$

$$R_{w_i} = \frac{L}{KA} \quad (4)$$

$$R_{in} = \frac{1}{h_{in}A} \quad (5)$$

$$R_{ow_i} = \frac{R_{w_i}}{2} + R_{out} \quad (6)$$

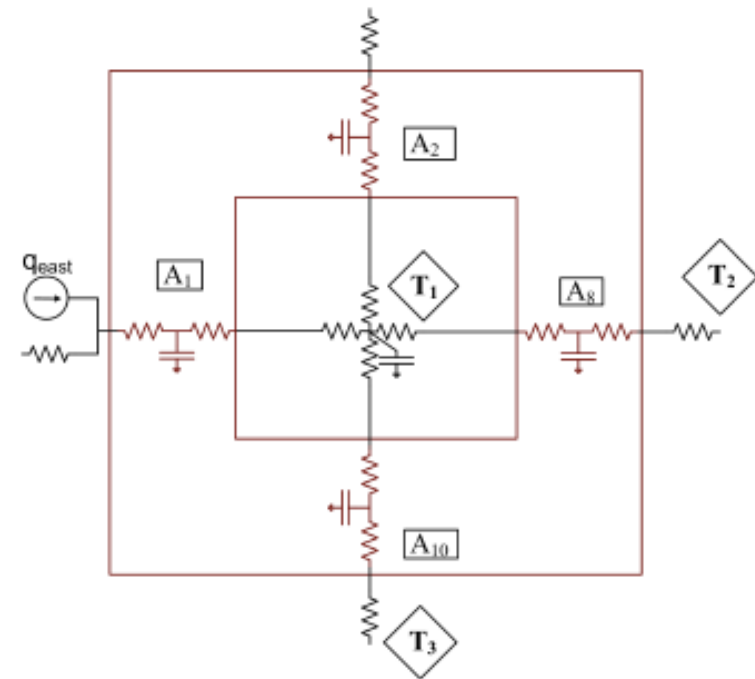
$$R_{out} = \frac{1}{h_{out}A} \quad (7)$$

$$\frac{dt_a}{d\tau} = \frac{1}{C_a} \left[\sum_{i=1}^n \frac{t_i - t_a}{R_{iw_i}} + q_{in} + q_{others} \right] \quad (8)$$

$$q_{in} = \dot{m}C_{pa} (t_{input} - t_a) \quad (9)$$

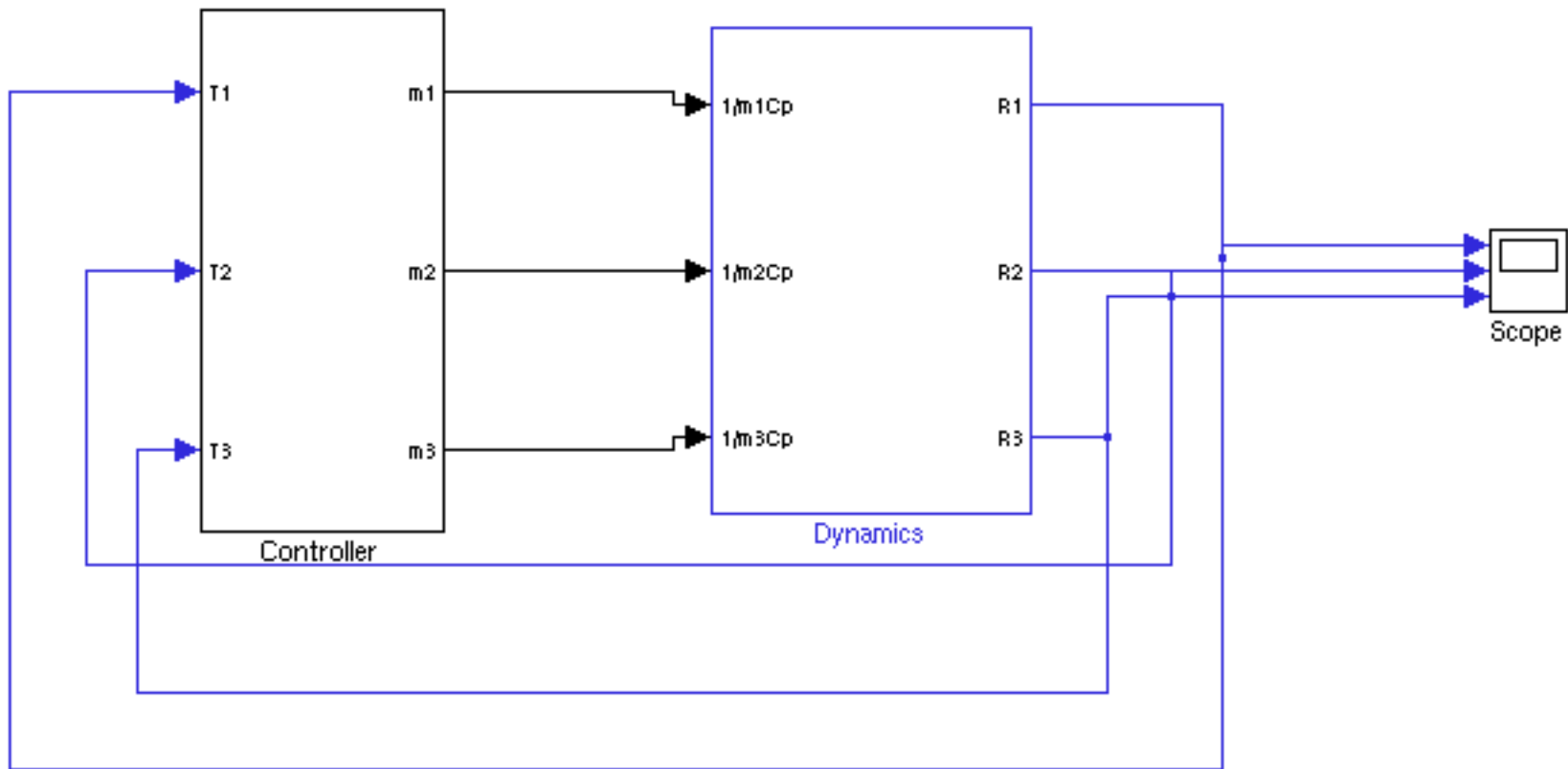
Other heat gains can be set to zero.

$$D(\tau) = \sum_{j=1}^m \frac{t_{eq}(\tau) - t_A(\tau)}{R_j} + q_p + q_l + q_o$$



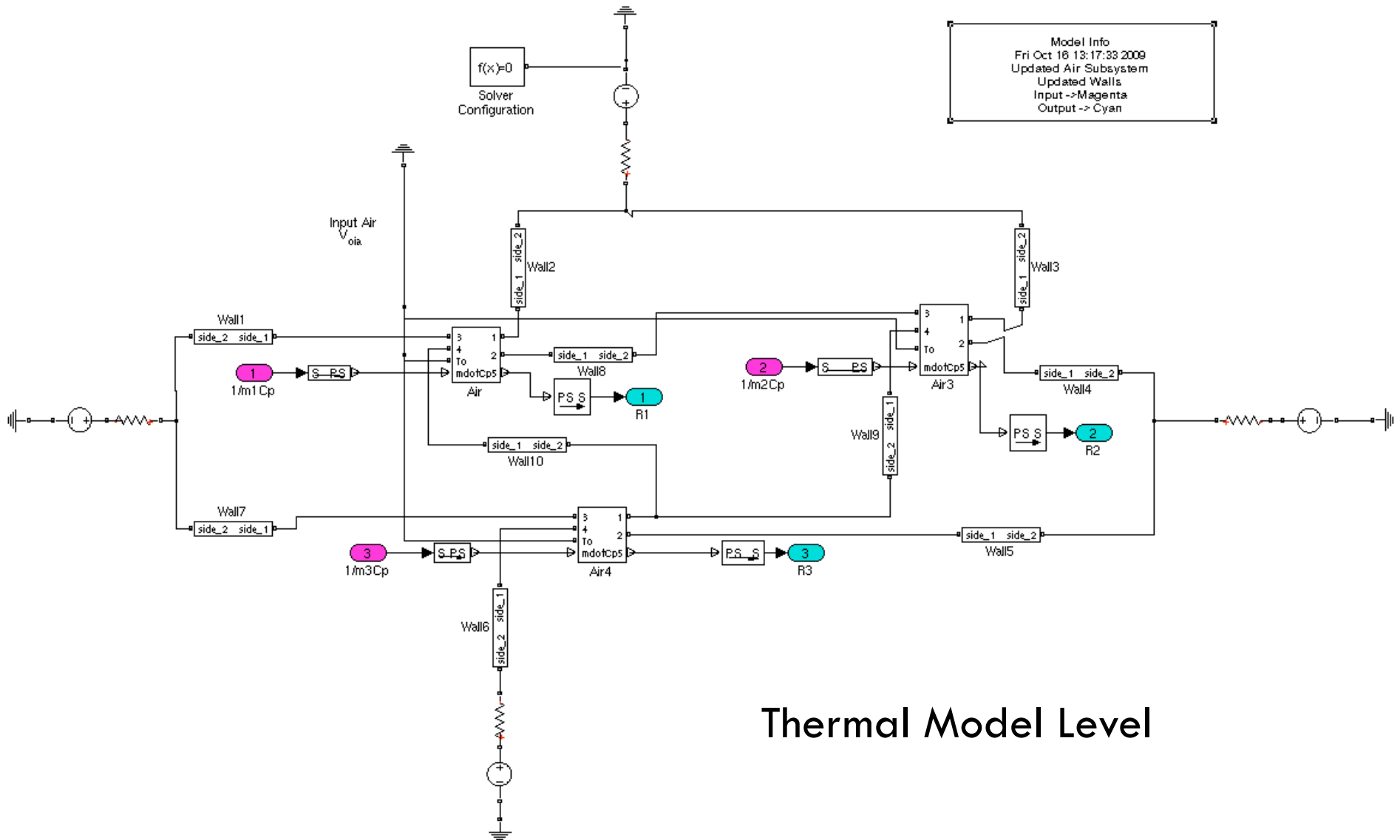
Fall 2008 Room Model

Model: Simulink (1 and 3 Rooms)



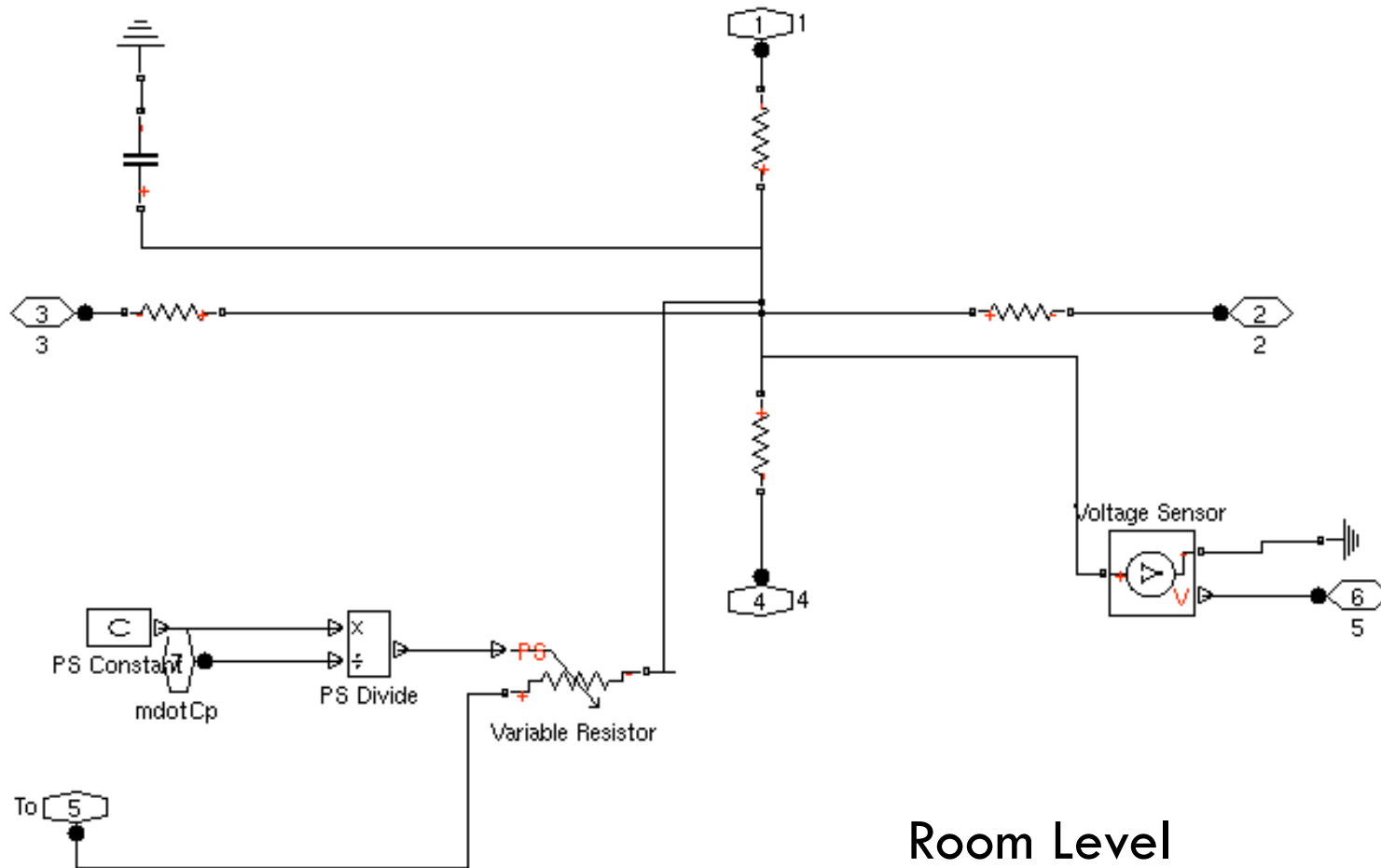
Top Level on Simulink

Model: Simulink (1 and 3 Rooms)

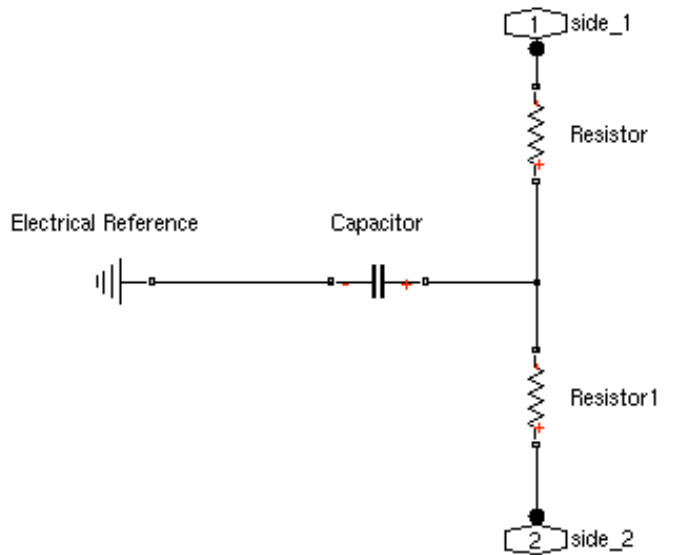


Thermal Model Level

Model: Simulink (1 and 3 Rooms)

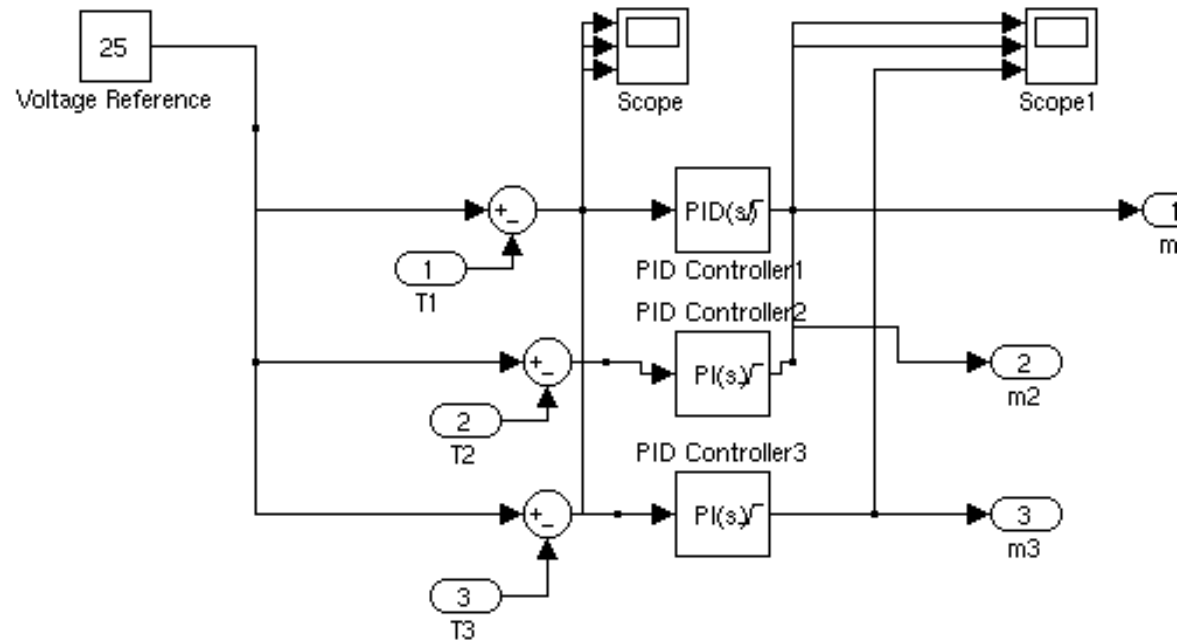


Model: Simulink (1 and 3 Rooms)

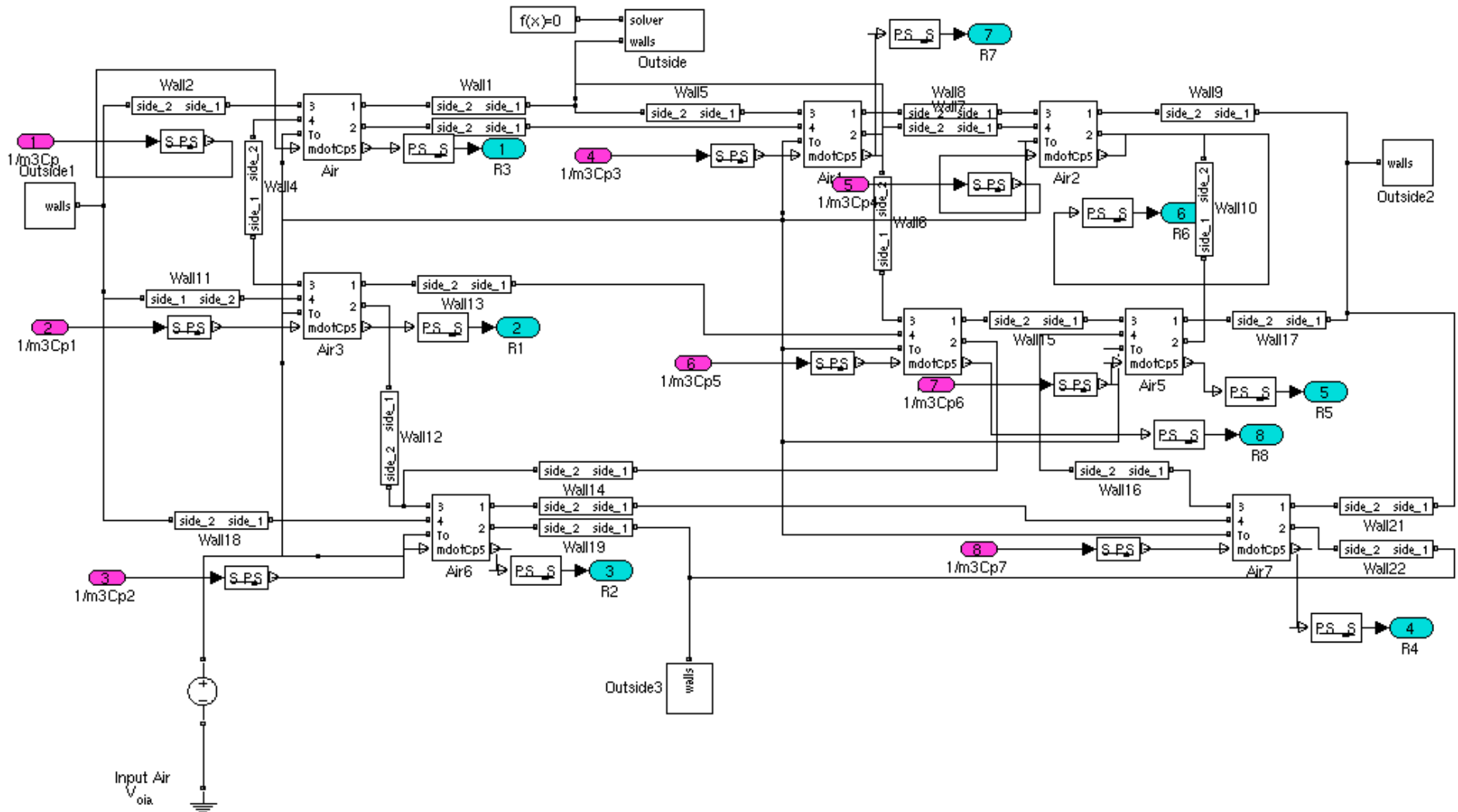


Wall Level

Control Block

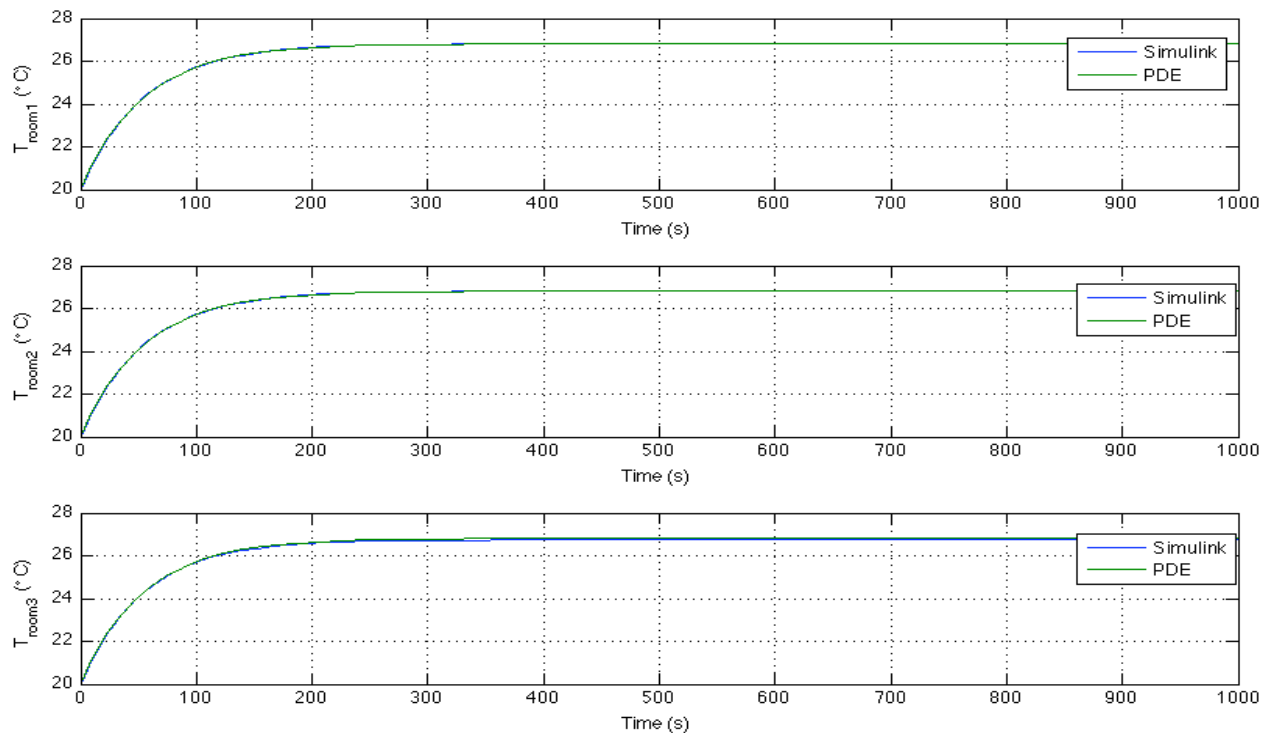


Model: Simulink 8 rooms



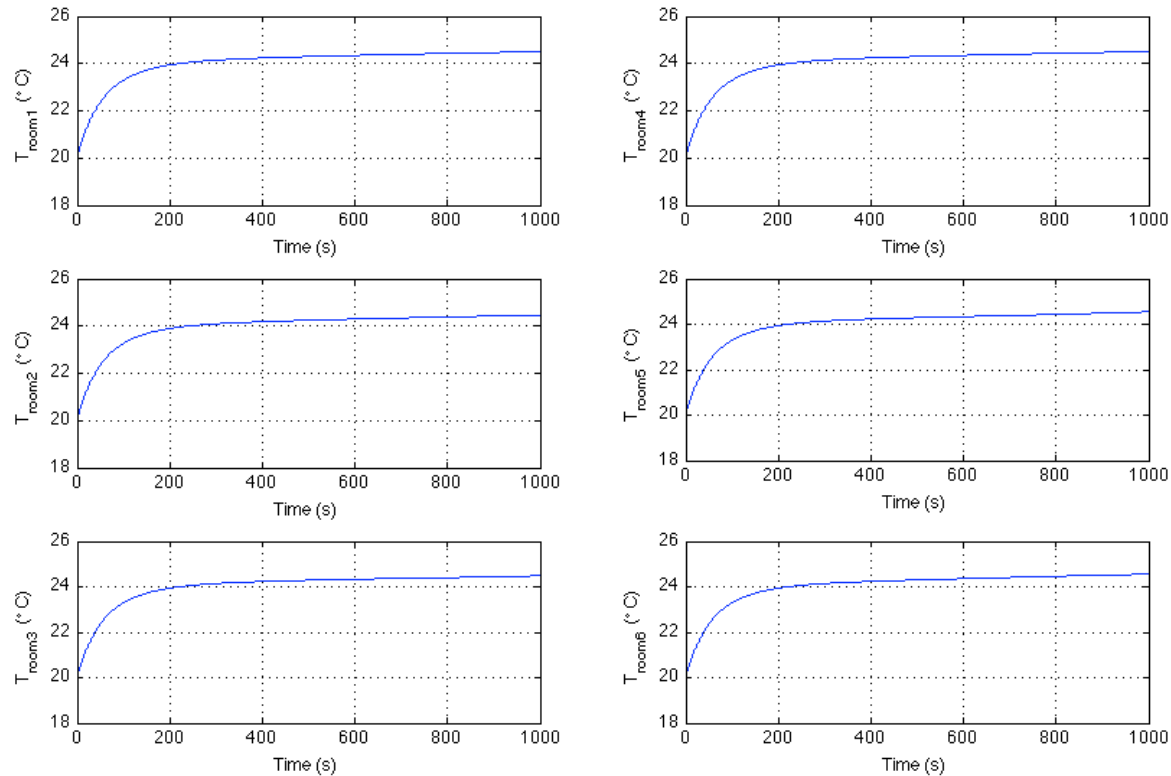
Model: Results PDE vs. Simulink (constant input)

- Error is in the order of $e-9$, probably due to numerical integration errors.



Model Results: Scalability-PID 8 room model

- Set point is 25 degrees C. Compute time increases, but model is easily scalable on Simulink. If wall properties are similar, Matlab model is easily scalable as well



Control Effort



- PID Model for room uses the desired temperature errors as input. Very easy to implement on Simulink.
- Other advance control techniques (LQR, CLQR, MPC) require state space representation. Scalability can be a problem and state-space representation is not easily implementable.
- No direct state-space extension from circuit analysis

Control Effort

- For the 3-room model system we can define the state space as:
- Where x is the state and u is the input vector. This non-linear system can be linearized about a nominal trajectory (x_n, u_n) for implementation of LQR or CLQR. The constraint is the input since the amount of airflow should be $\dot{x} = Ax + \text{diag}(x^T R_1) R_2 u + Bu$

Control Effort



- The matrices A , R_1 , R_2 and B are defined by the dynamics of the system. R_1 and R_2 are necessary to represent the non-linearity in the model.
- Linearization is straight forward. Selection of nominal trajectory is not.
- Future work should focus on providing the correct system dynamics to the controller.

Summary



- A scalable model of a thermal system was presented. The model was verified using another modeling tool. Behavior of the systems to same inputs produce same outputs. Errors are only due to numerical integration
- Validation and model tuning can be done once UC Merced data is available
- Advanced control techniques require a different definition of model (state-space). A simple representation was shown, which can be linearized about a nominal trajectory
- Future work should include the proper representation of the non-linear model for advanced control techniques