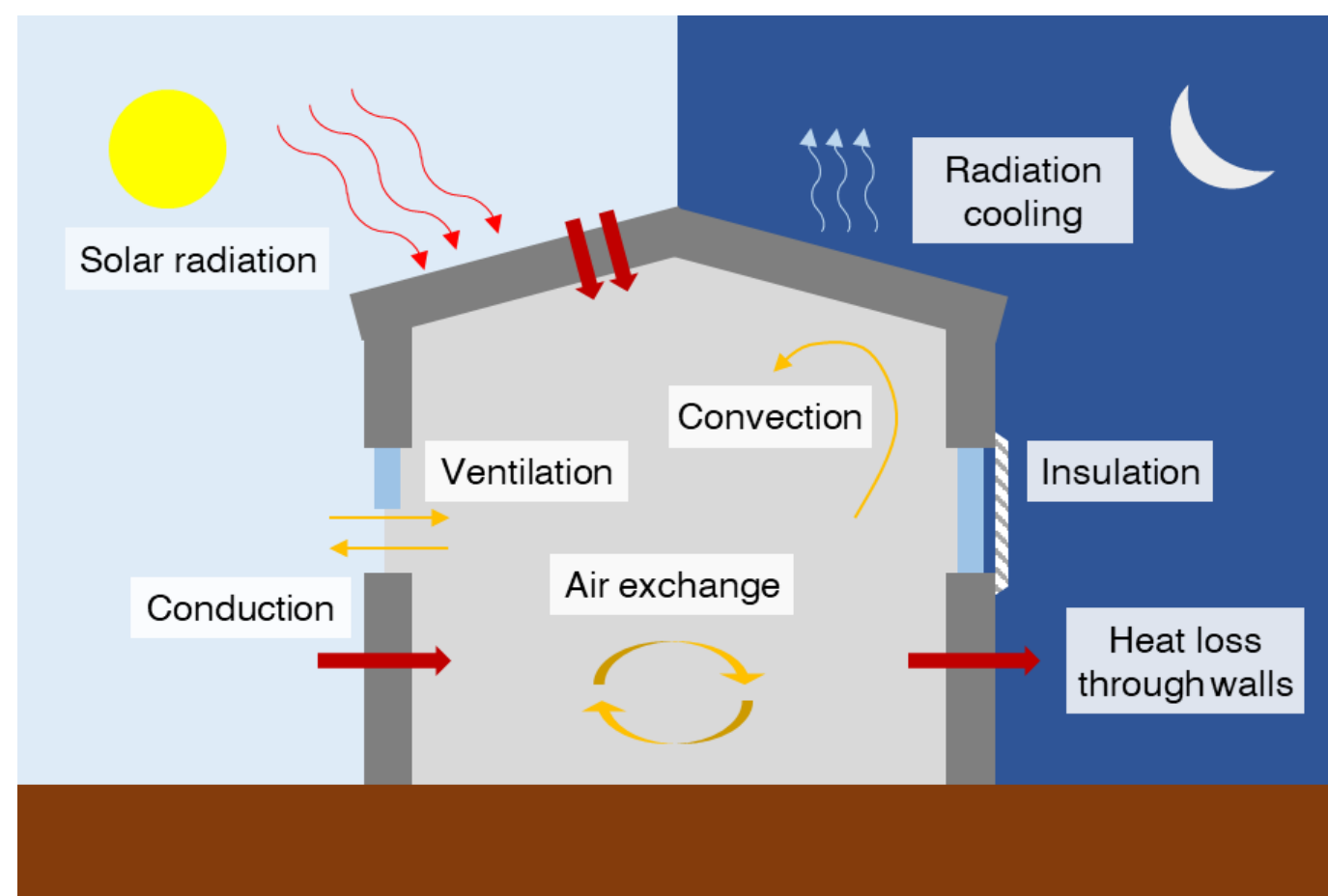


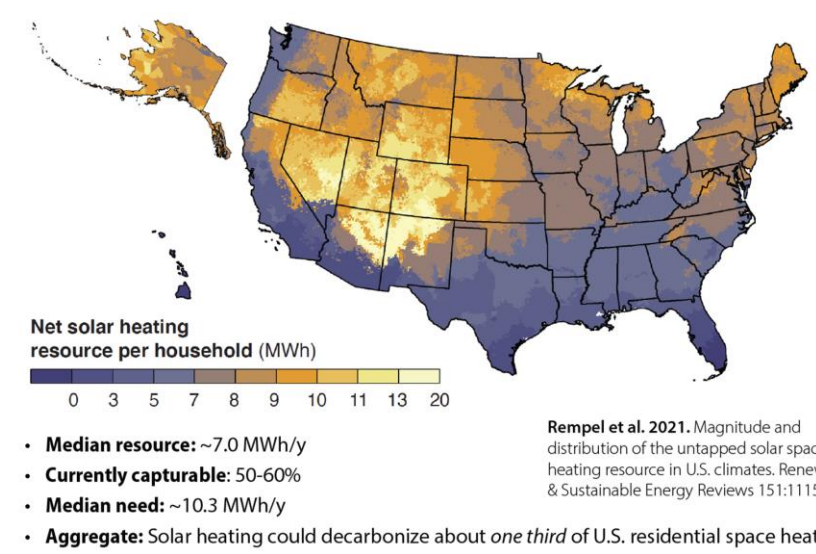
# CPS Medium: Physics-informed Learning and Control of Passive and Hybrid Conditioning Systems in Buildings

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**Goal:** Design model-based and data-driven strategies for controlling passive and hybrid (active + passive) building conditioning systems



**Motivation:** Harnessing passive climatic resources for heating and cooling in buildings will provide significant energy savings



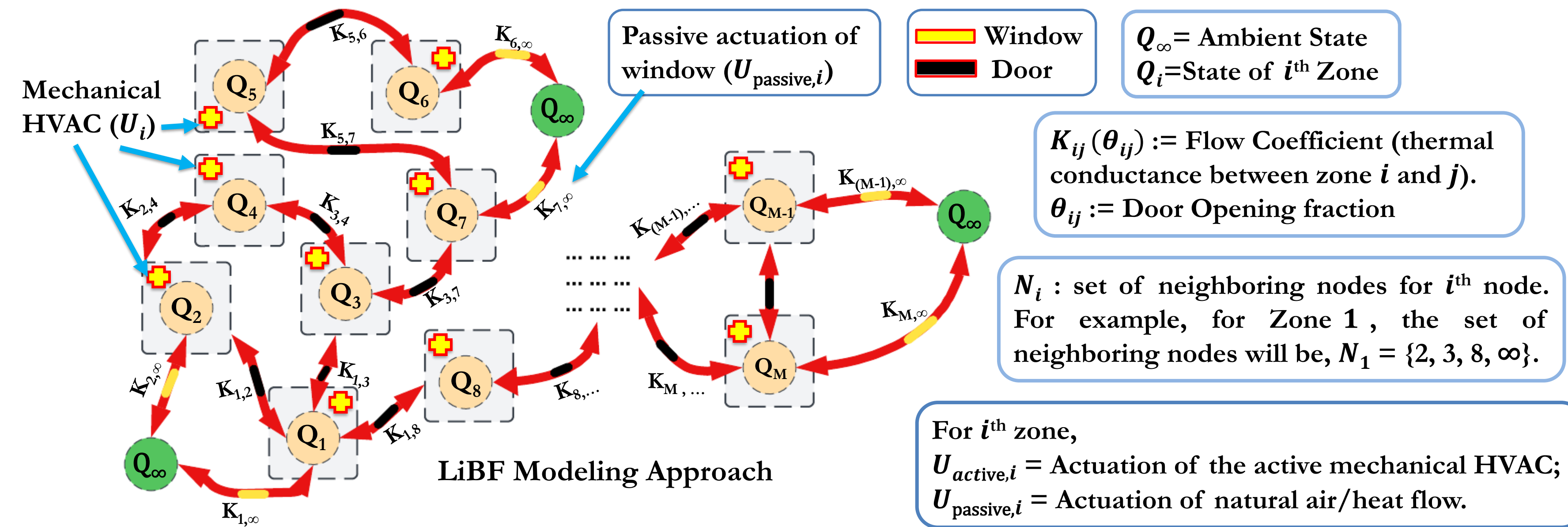
## Challenges:

- Passive climatic resources such as available solar radiation, outside air (temperature and quality), etc. are time-varying and unpredictable: system must respond and adapt quickly.
- Control strategies must be climate-specific, portable across building types and layouts, and easily deployable for wide adoption.

## Key innovation and approach:

- Novel graph-based model for these systems: combines physics-based knowledge (e.g., conservation laws) with data-driven approaches
- Exploit this unique model structure for both optimal control design as well as for providing certificates (e.g., stability, passivity, robustness to parameter variation)
- Incorporate learning, forecasts, and human intervention into control strategy

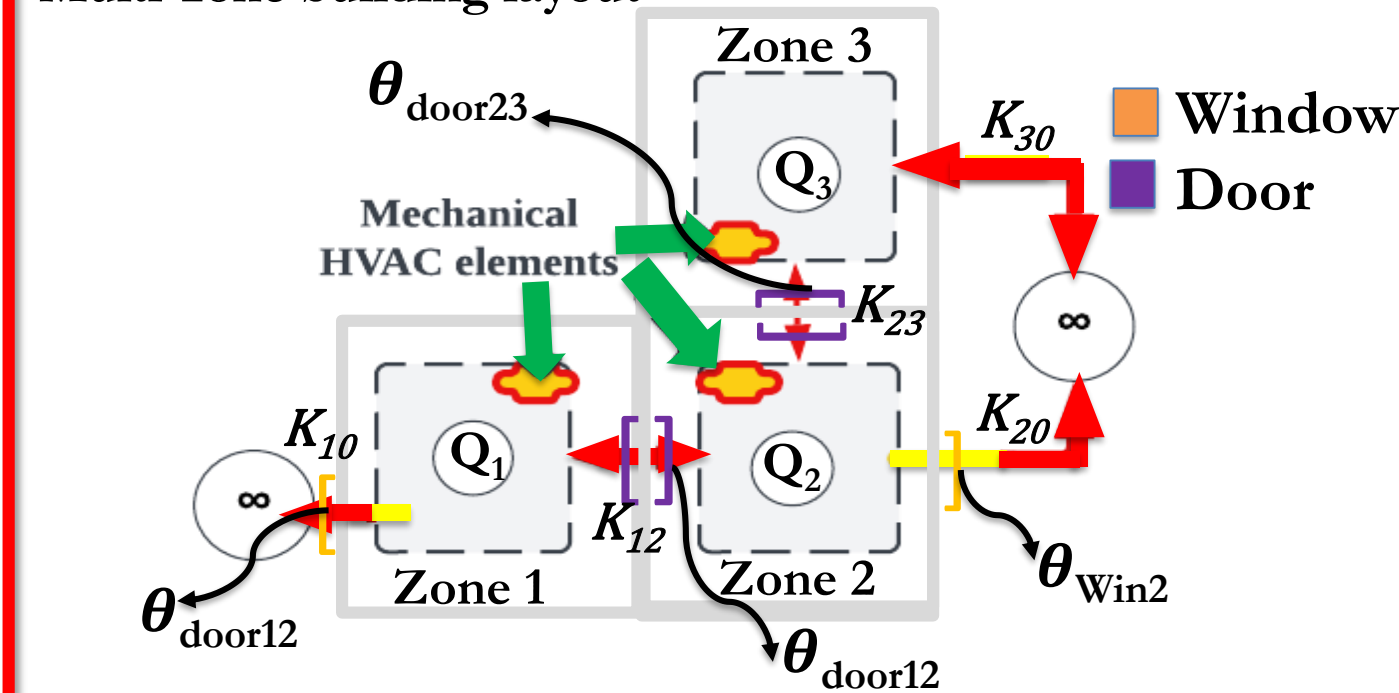
## Locally-interactive Bilinear Flow(LiBF) Models for Hybrid Conditioning Systems



Physics-informed Graph-based Model

$$C_i \dot{Q}_i = \sum_{j \in N_i} K_{ij}(\theta_{ij})(Q_j - Q_i) + B_i(U_{active,i})Q_i + B_{i\infty}(U_{passive,i})Q_\infty$$

## Benchmark Study for Validating LiBF Modeling Approach

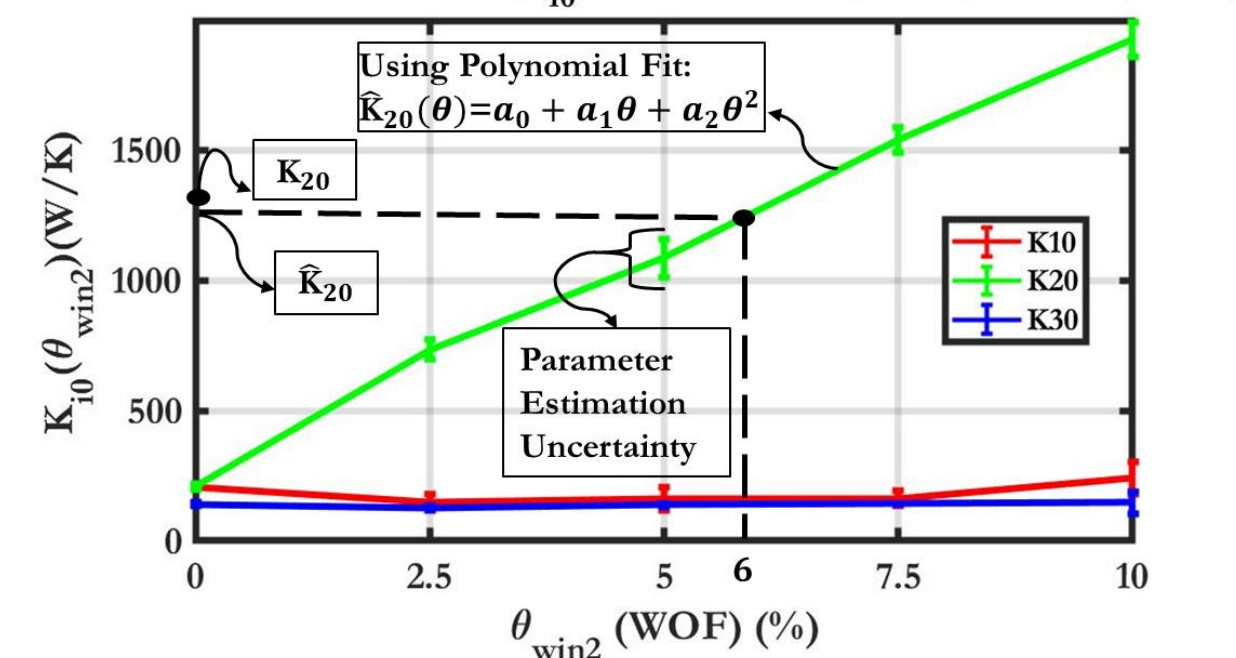


### Model for the Multizone building

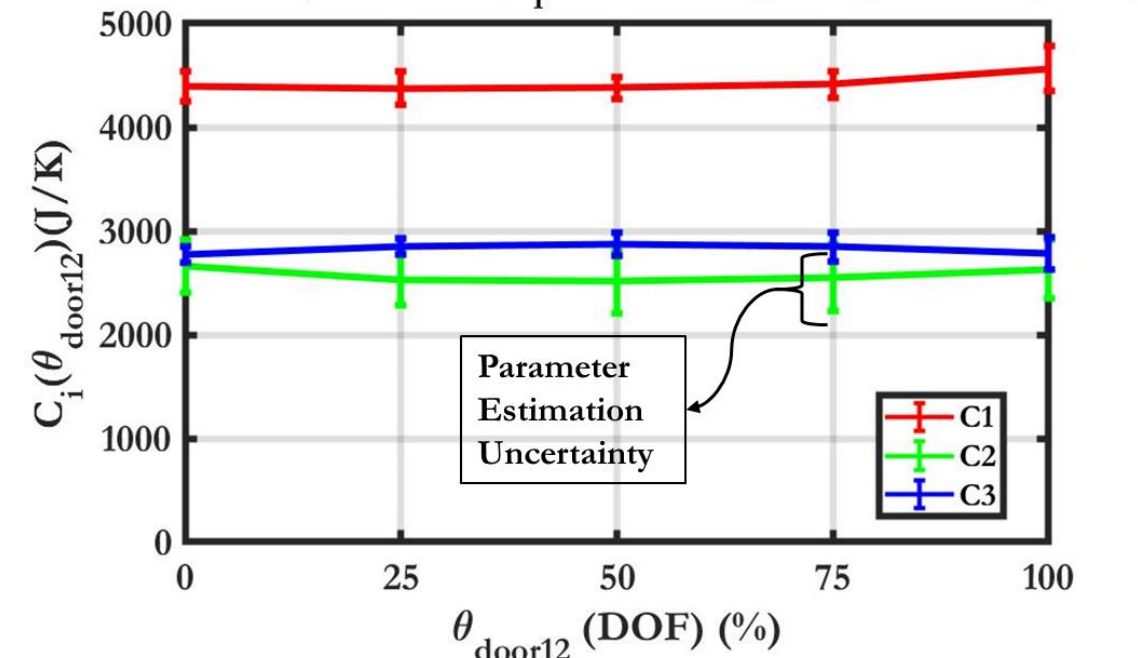
$$\begin{aligned} C_1 \dot{T}_1 &= K_{12}(\theta_{door12})(T_2 - T_1) + K_{10}(T_\infty - T_1) + U_1 \\ C_2 \dot{T}_2 &= K_{12}(\theta_{door12})(T_1 - T_2) + K_{23}(T_3 - T_2) + K_{20}(\theta_{win2})(T_\infty - T_2) + U_2 \\ C_3 \dot{T}_3 &= K_{23}(\theta_{door23})(T_2 - T_3) + K_{30}(T_\infty - T_3) + U_3 \end{aligned}$$

$T_i$  = Temperature of Zone  $i$ ,  $T_\infty$  = Outdoor Temperature,  $U_i$  =  $i^{\text{th}}$  zone heat input

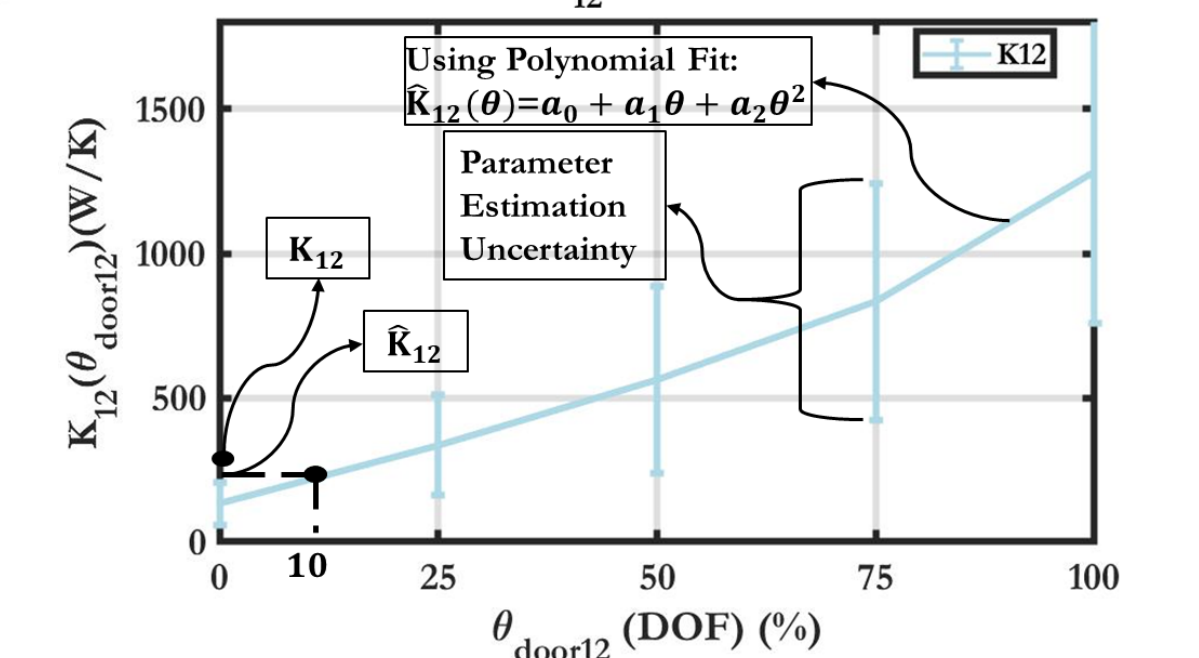
Thermal Conductance ( $K_{ij}$ ) vs Window Opening Factor (WOF)



Thermal Capacitance ( $C_i$ ) vs Door Opening Factor (DOF)

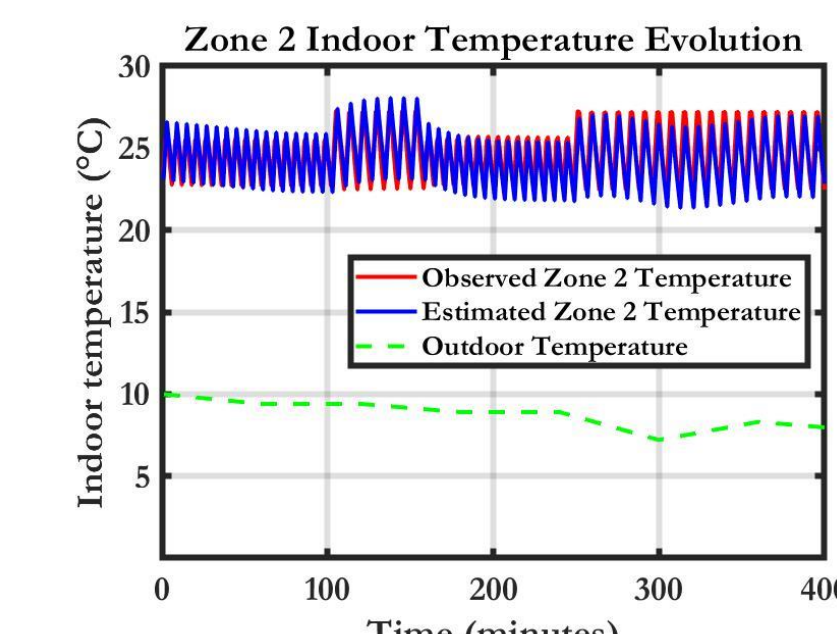
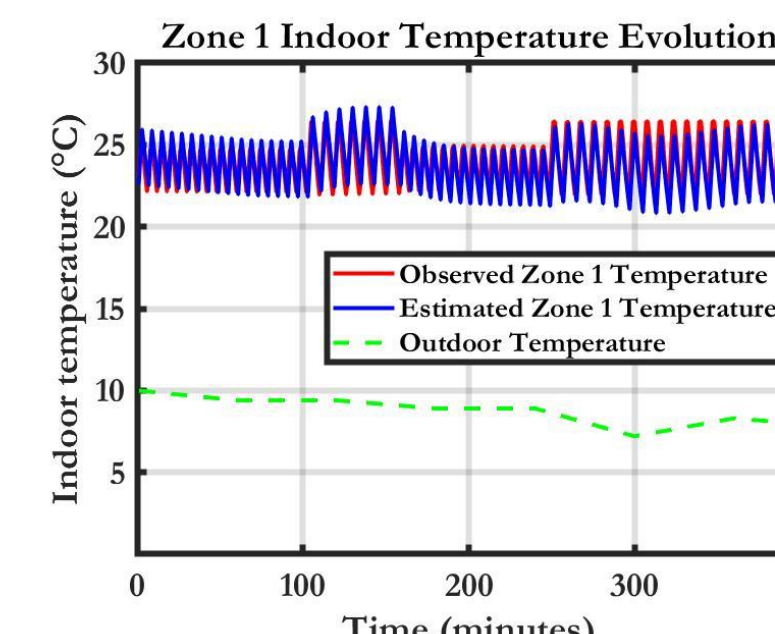


Thermal Conductance ( $K_{ij}$ ) vs Door Opening Factor (DOF)



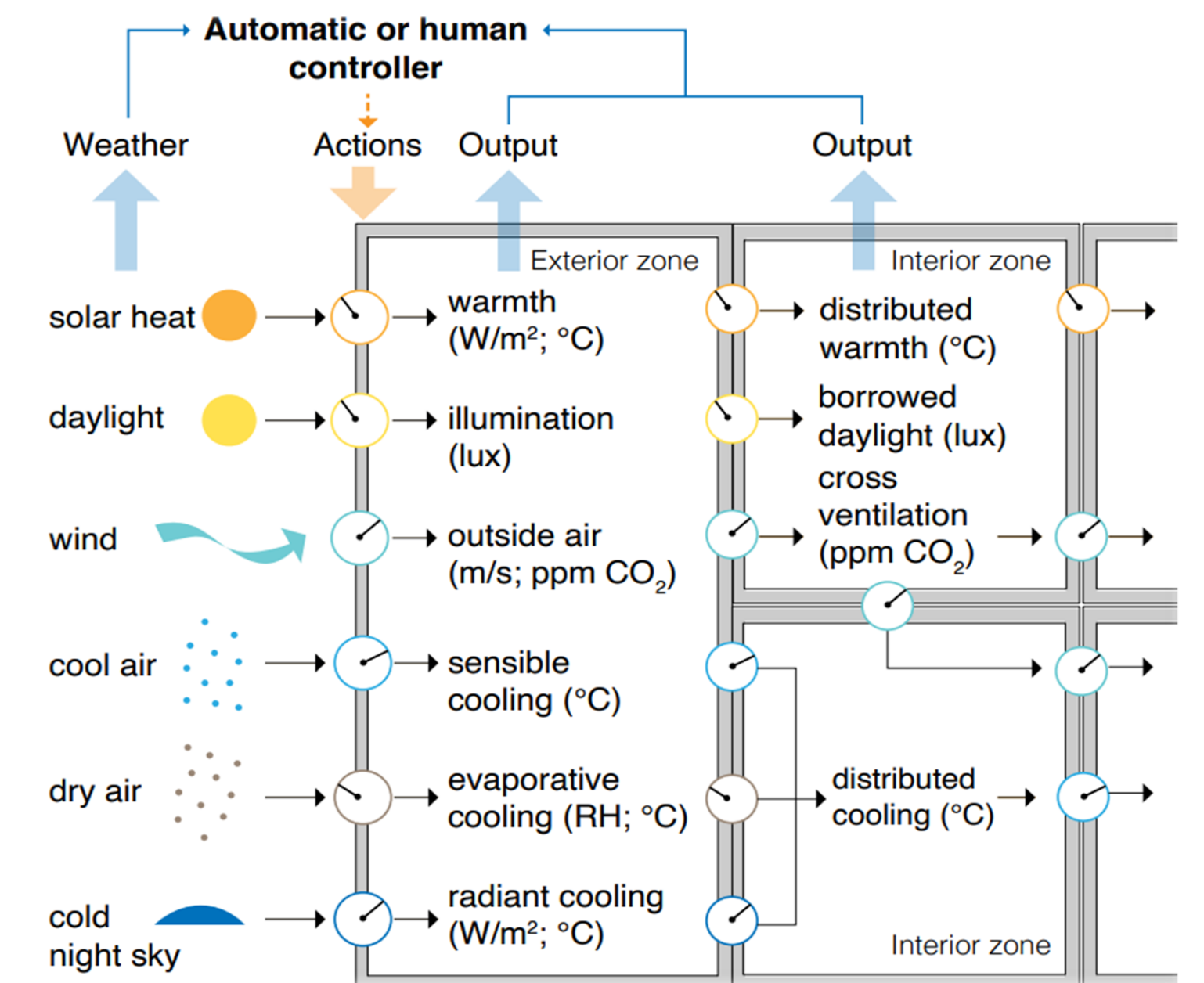
### Key observations from study:

- LiBF model is able to capture the effect of passive controls (door and window opening factors)
- $K_{20}$  increases as  $\theta_{win2}$  increases (more passive resource harvesting),  $K_{10}$  and  $K_{30}$  remain unchanged with increasing  $\theta_{win2}$
- $C_1, C_2, C_3$  also unaffected by increasing  $\theta_{door12}$
- $K_{12}$  increases as  $\theta_{door12}$  increases (increased inter-zone coupling)

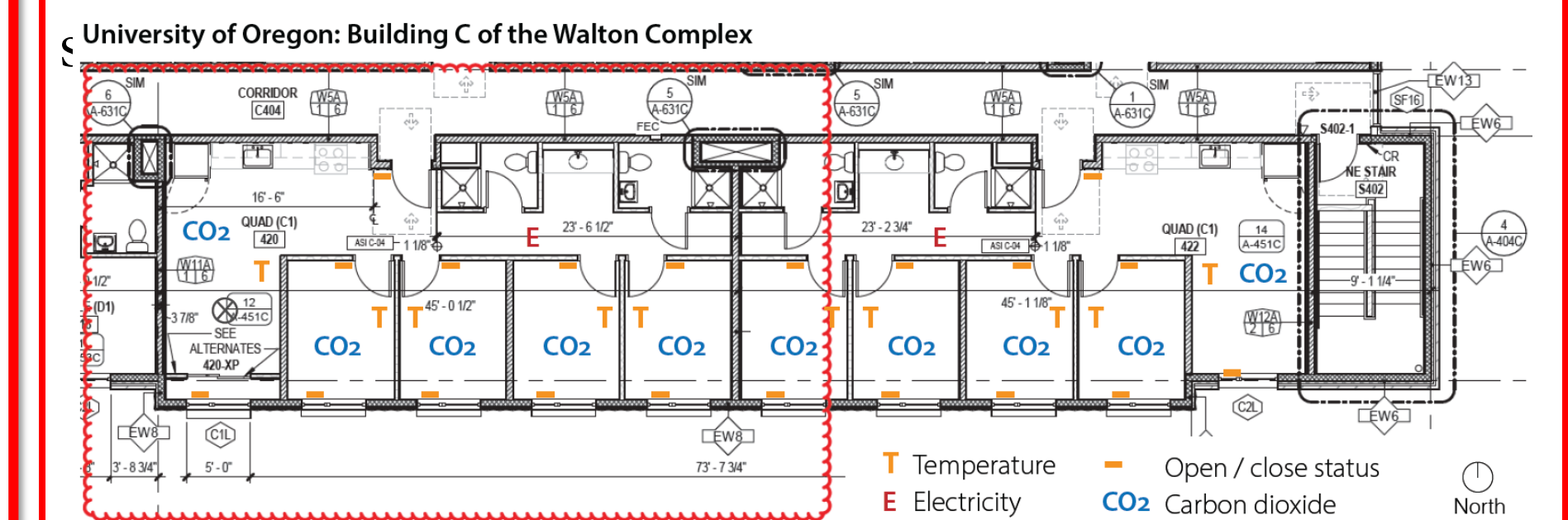


Validation: Zone 1, Zone 2 temperatures (compared against predictions from model)

## Control of Passive & Hybrid Conditioning:



**Validation Testbed (at Univ. of Oregon):** Fully instrumented building with a digital twin (calibrated Energy Plus model) for modeling/control design, energy savings, and occupant response: data collection



## Scientific Impact:

- New modeling paradigm for a broad class of CPS systems with coupled mass and heat transfer using graph-based locally interactive bilinear structure.
- Novel approaches for analysis and design of controllers that exploit this structure.

## Broader Impacts:

- Adoption of passive mechanisms for heating, cooling and ventilation: increased energy savings
- Public perceptions about passive strategies influenced through outreach, awareness and education; and dissemination of research findings



*Table 1 : Change in conductance for different door/window opening fraction*

Door/Wind. Opening Fraction (%)	10	20	30	40
10	1	2	3	4