

# Energy Management for Building Climate Comfort in Smart Thermal Grids with Seasonal Storage Systems

Vahab Rostampour

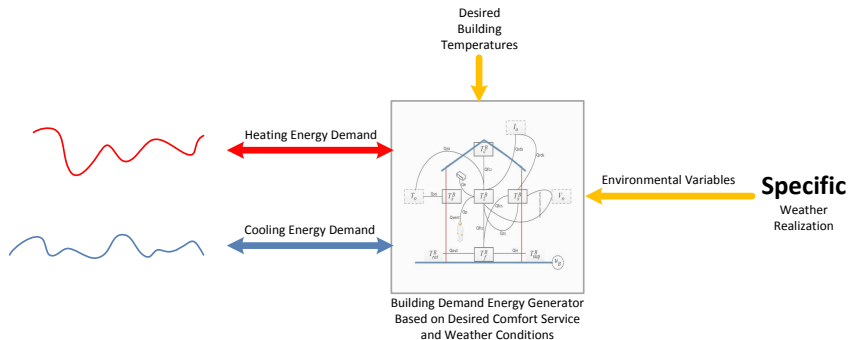
Delft University of Technology  
Delft Center of Systems and Control

Users' Group Meeting

November 10, 2016



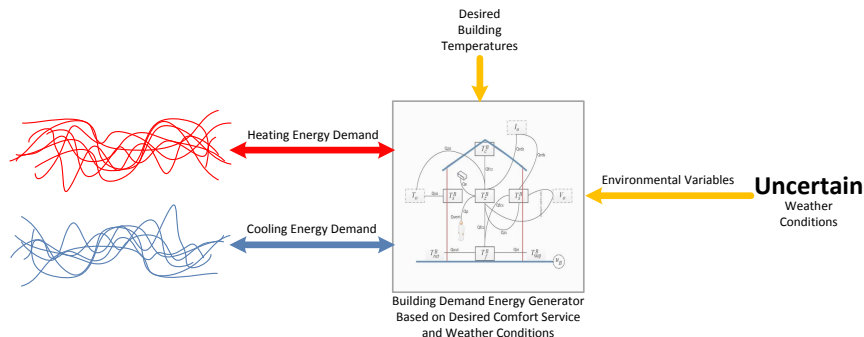
# Recap: Building Thermal Energy Demand



## Thermal Energy Demand Profile:

- Complete and detailed building dynamical model
- Desired building temperatures (local controller unit)
- In specific weather realization, certain demand profiles are generated

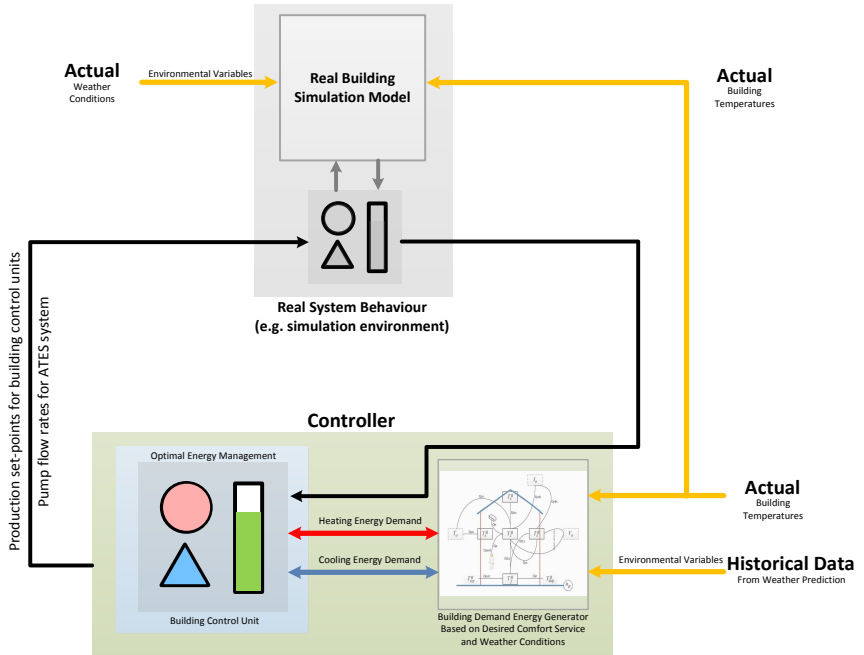
# Recap: Building Thermal Energy Demand



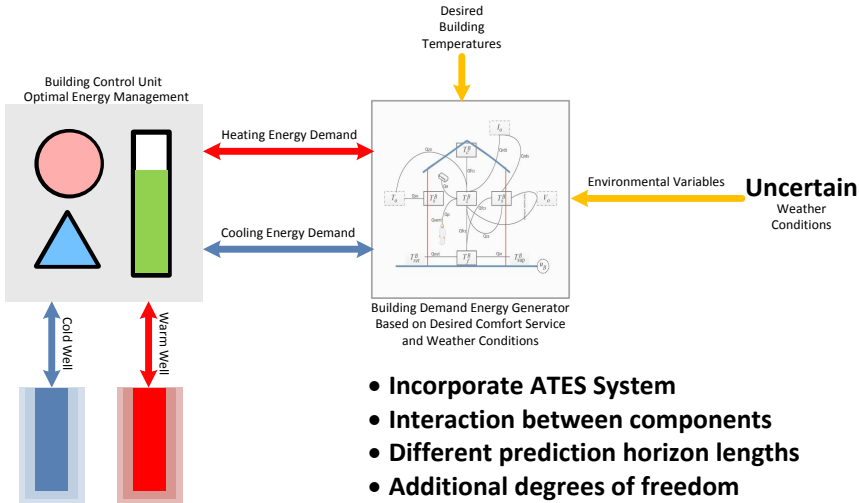
## Thermal Energy Demand Generator:

- Complete and detailed building dynamical model
- Desired building temperature (local controller unit)
- In uncertain weather conditions, uncertain demand profiles are generated



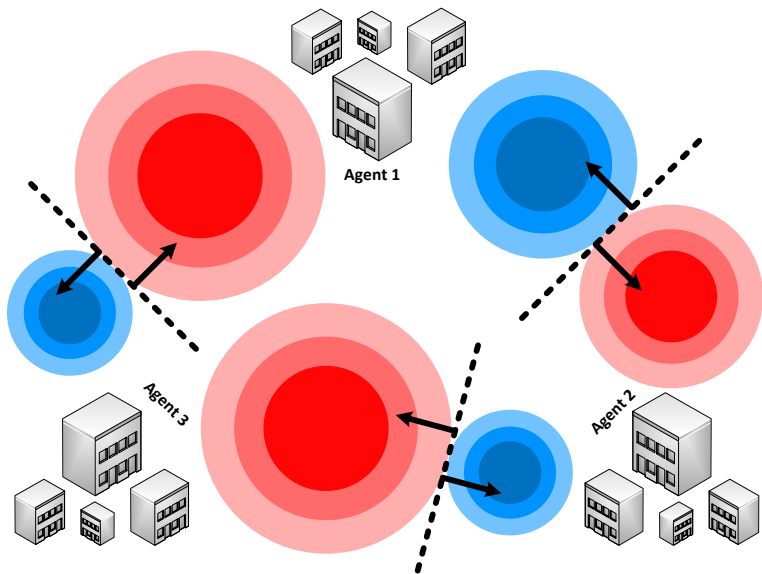


# Building Climate Comfort and ATES Systems <sup>1</sup>



<sup>1</sup>[Rostampour, Keviczky, Submitted to IFAC World Congress, 2017]

# ATES Systems in Smart Thermal Grids <sup>1</sup>



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# Control Frameworks and Solutions

## ① Control formulations:

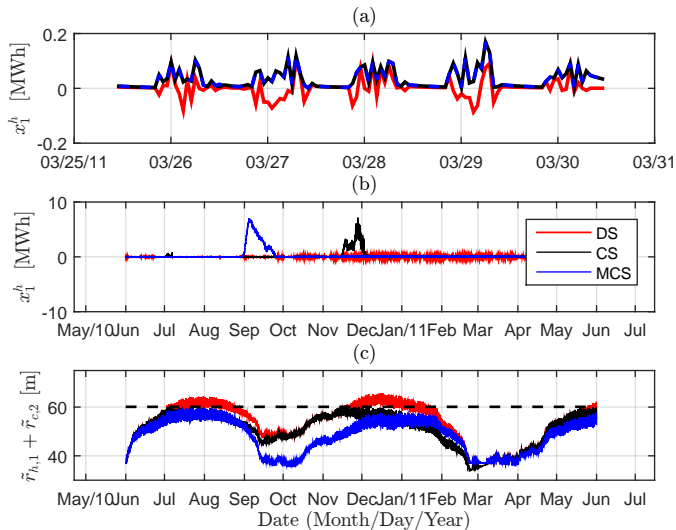
- Decoupled Solution (DS): each agent problem w/o common constraint
- Centralized Solution (CS): a complete problem with coupling constraint
- Move-Blocking Centralized Solution (MCS): CS + multi-rate actions
- Deterministic DS: we fixed the uncertain elements ( $w_i$ )

## ② Simulation Setup:

- Actual registered weather data; realistic buildings parameters
- Three-agent example in city center of Utrecht, the Netherlands
- simulation time is one year with hourly-based sampling time
- DS and CS with a day-ahead (24 hours) prediction horizon
- MS with a season long (3 month) prediction horizon
- MCS with multi-rate actions:
  - hourly-based during first day
  - daily-based in the first week
  - weekly-based within the first month, and monthly-based in the rest of season



# Simulations Results: A-Posteriori Feasibility Validation



# Conclusions and Future Works

- Remarks:

- ① ATES System Modeling + Energy Balance Problem
- ② Integrated ATES System into Building Climate Comfort System
- ③ ATES in STGs + Preventing Unwanted Mutual Interactions
- ④ DS, CS, and MCS Control Problem Formulations
- ⑤ Simulation Results Showed Expected Results

- Next Steps:

- ① Refinement of the ATES System Model
- ② Develop/Formulate A Hierarchical Framework to Predict longer horizon

# Scientific Output

- Building Climate Energy Management in Smart Thermal Grids via Aquifer Thermal Energy Storage Systems  
Journal of Energy Procedia, Elsevier. (2016, to appear)
- A Model Predictive Framework of GSHP coupled with ATES System in Heating and Cooling Networks of a Building  
Accepted for the IEA Conference on Heat Pump, Rotterdam, The Netherlands. (2017, May)
- Chance Constrained Model Predictive Controller Synthesis for Stochastic Max-Plus Linear Systems  
IEEE International Conference on Systems, Man, and Cybernetics, Budapest, Hungary. (2016, October)
- A Control-Oriented Model for Combined Building Climate Comfort and Aquifer Thermal Energy Storage System  
European Geothermal Congress, Strasbourg, France. (2016, September)
- Robust Randomized Model Predictive Control for Energy Balance in Smart Thermal Grids  
European Control Conference, Aalborg, Denmark. (2016, June)
- Distributed Energy Management in Smart Thermal Grids with Uncertain Demands  
MSc thesis ( Ananduta, W.W.), DCSC, TU Delft. (2016, August)
- A Set Based Probabilistic Approach to Threshold Design for Optimal Fault Detection  
Submitted to American Control Conference 2017
- Probabilistic Energy Management for Building Climate Comfort in Smart Thermal Grids with Seasonal Storage Systems  
Submitted to IFAC World Congress 2017

# Current Research Status

## Under Progress:

- Stochastic Distributed MPC for Chance Constrained Linear Systems
- Computationally Distributed Algorithm for Chance Constrained Optimization
- On the Road Between Convex and Nonconvex Scenario Program for Chance Constrained Optimization Problems
- Privacy Preserving Scheme in Distributed Constrained Optimization

## Master Students' Thesis:

- Distributed Stochastic Production Scheduling in Smart Grids  
Ole ter Haar (Expected Graduation February 2017)

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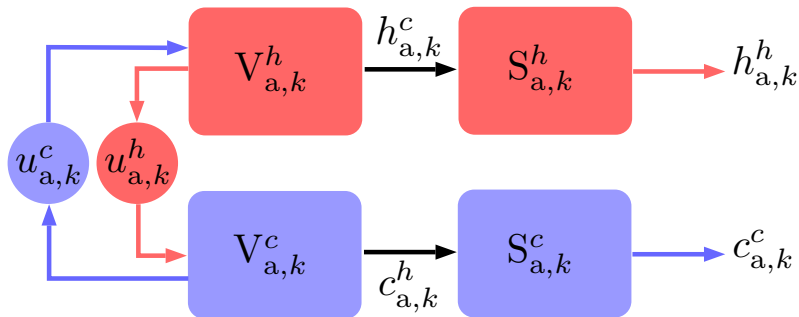
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# Outline

- ① Single Agent System
- ② Multi-Agent Network: ATES-SGs
- ③ Frameworks and Simulation Results

# ATES System: Conceptual Block Diagram



# ATES System: Model Dynamics — Heating Mode

$$V_{a,k+1}^h = V_{a,k}^h - (u_{a,k}^h - u_{a,k}^c)$$

$$V_{a,k+1}^c = V_{a,k}^c + (u_{a,k}^h - u_{a,k}^c)$$

$$S_{a,k+1}^h = \eta_a S_{a,k}^h - (h_{a,k}^h - h_{a,k}^c)$$

$$S_{a,k+1}^c = \eta_a S_{a,k}^c + (c_{a,k}^h - c_{a,k}^c)$$

$$h_{a,k} = \alpha u_{a,k}^h \quad , \quad h_{a,k}^h = \alpha_h u_{a,k}^h \quad , \quad c_{a,k}^h = \alpha_c u_{a,k}^h$$

$$\alpha = \alpha_h + \alpha_c \quad , \quad \begin{cases} \alpha_h &= \rho_w c_{pw} (T_{a,k}^h - T_{a,k}^{\text{amb}}) \\ \alpha_c &= \rho_w c_{pw} (T_{a,k}^{\text{amb}} - T_{a,k}^c) \end{cases} ,$$



# ATES System: Model Dynamics — Cooling Mode

$$V_{a,k+1}^h = V_{a,k}^h - (u_{a,k}^h - u_{a,k}^c)$$

$$V_{a,k+1}^c = V_{a,k}^c + (u_{a,k}^h - u_{a,k}^c)$$

$$S_{a,k+1}^h = \eta_a S_{a,k}^h - (h_{a,k}^h - h_{a,k}^c)$$

$$S_{a,k+1}^c = \eta_a S_{a,k}^c + (h_{a,k}^h - h_{a,k}^c)$$

$$c_{a,k} = \alpha u_{a,k}^c \quad , \quad c_{a,k}^c = \alpha_c u_{a,k}^c \quad , \quad h_{a,k}^c = \alpha_h u_{a,k}^c$$

$$\alpha = \alpha_h + \alpha_c \quad , \quad \begin{cases} \alpha_h &= \rho_w c_{pw} (T_{a,k}^h - T_{a,k}^{\text{amb}}) \\ \alpha_c &= \rho_w c_{pw} (T_{a,k}^{\text{amb}} - T_{a,k}^c) \end{cases} ,$$

# ATES System: Energy Balance between Wells

An important operational limitation of ATES system is that the sum of injected and extracted thermal energy over a period of time has to be zero:

$$\left. \begin{aligned} \sum_{t=k}^{N_y} (h_{a,t}^h - h_{a,t}^c) &= 0 \\ \sum_{t=k}^{N_y} (c_{a,t}^c - c_{a,t}^h) &= 0 \end{aligned} \right\} \longrightarrow S_{a,t}^h + S_{a,t}^c = \bar{S}_a$$

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$$S_{a,t}^h + S_{a,t}^c \leq \bar{S}_a + e_{i,t}$$

$$S_{a,t}^h + S_{a,t}^c \geq \bar{S}_a - e_{i,t}$$

$e_{i,t}$  is an auxiliary control variable to soften the formulated constraint

# Energy Management in Single Agent System

At each sampling time, a finite-horizon chance-constrained mixed-integer quadratic optimization problem is formulated for each agent  $i$ :

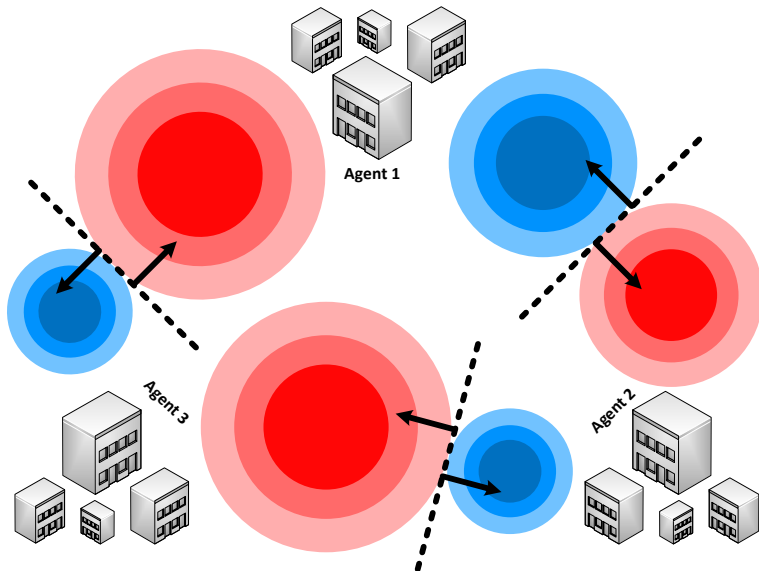
$$\begin{aligned} & \underset{\mathbf{u}_i, \mathbf{v}_i}{\text{minimize}} && \mathcal{V}_i(\mathbf{x}_i, \mathbf{u}_i) = \mathbb{E}_{\mathbf{w}_i}[\mathcal{J}_i(\mathbf{x}_i, \mathbf{u}_i)] \\ & \text{subject to} && E_i \mathbf{u}_i + F_i \mathbf{v}_i + P_i \leq 0, \\ & && \mathbb{P}_{\mathbf{w}_i}[A_i x_{i,k} + B_i \mathbf{u}_i + C_i \mathbf{w}_i \geq 0] \geq 1 - \varepsilon_i \\ & && \mathbf{w}_i \in \mathcal{W}_i \end{aligned}$$

The index of  $\mathbb{E}_{\mathbf{w}_i}, \mathbb{P}_{\mathbf{w}_i}$  denotes the dependency of the state trajectory on the string of random scenarios  $\mathbf{w}_i$

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# ATES Systems in STGs



# ATES Systems in STGs: Mutual Interactions

Thermal radius can be determined using the water volume in each well:

$$r_{a,t}^l = \sqrt{\frac{c_{pw} V_{a,t}^l}{c_{aq} \pi \ell}} \quad , \quad l \in \{h, c\}$$

where  $c_{aq} = (1 - n_p)c_{sand} + n_p c_{pw}$ , and  $c_{sand}, c_{pw}, n_p$ , are parameters.

We restrict the thermal radius between the neighboring agents:

$$(r_{a,t}^h)_i + (r_{a,t}^c)_j \leq d_{ij}$$

where  $d_{ij}$  is a given distance between the neighboring agents.

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where  $c_{aq} = (1 - n_p)c_{sand} + n_p c_{pw}$ , and  $c_{sand}, c_{pw}, n_p$ , are parameters.

We restrict the stored volume of water between the neighboring agents:

$$(V_{a,t}^h)_i + (V_{a,t}^c)_j \leq V_{ij} - \delta_{ij,t}$$

where  $V_{ij}$  is an upper-bound on the capacity of common resource pool and  $\delta_{ij,t} = 2c_{aq}\pi\ell (\bar{r}_{a,t}^h)_i (\bar{r}_{a,t}^c)_j / c_{pw}$  is a common uncertainty source.



# Energy Management in ATES-SGs

At each sampling time, the energy management problem for ATES systems in STGs is as follows:

$$\begin{aligned} & \underset{\{\mathbf{u}_i, \mathbf{v}_i\}_{i=1}^N}{\text{minimize}} && \sum_{i=1}^N \mathcal{V}_i(\mathbf{x}_i, \mathbf{u}_i) \\ & \text{subject to} && E_i \mathbf{u}_i + F_i \mathbf{v}_i + P_i \leq 0 \\ & && \mathbb{P}_{\mathbf{w}_i} \left[ A_i x_{i,k} + B_i \mathbf{u}_i + C_i \mathbf{w}_i \geq 0 \right] \geq 1 - \varepsilon_i \\ & && \mathbb{P}_{\boldsymbol{\delta}_{ij}} \left[ H_i \mathbf{x}_i + H_j \mathbf{x}_j \leq \bar{V}_{ij} - \boldsymbol{\delta}_{ij} \right] \geq 1 - \bar{\varepsilon}_{ij} \\ & && \mathbf{w}_i \in \mathcal{W}_i, \boldsymbol{\delta}_{ij} \in \Delta_{ij}, \forall j \in N_i \\ & && \forall i \in \{1, 2, \dots, N\} \end{aligned}$$

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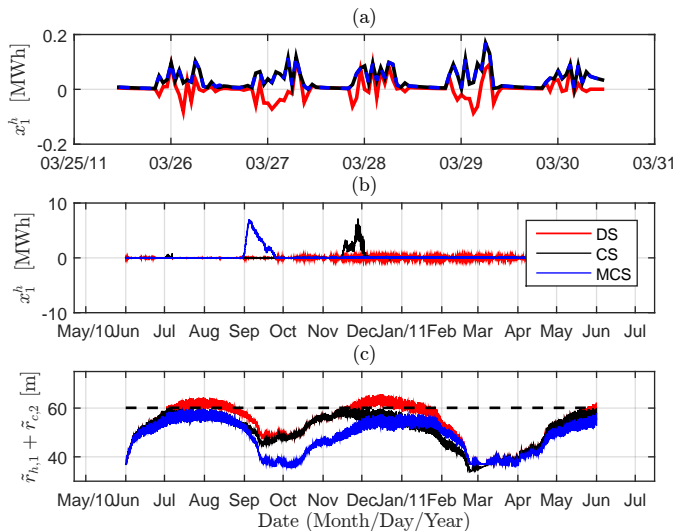
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# Simulations Results: A-Posteriori Feasibility Validation



# MODFLOW Results: Average Thermal Efficiency

