

#### Data Driven Investigation of Faults in HVAC Systems

#### Comparative data mining in energy systems

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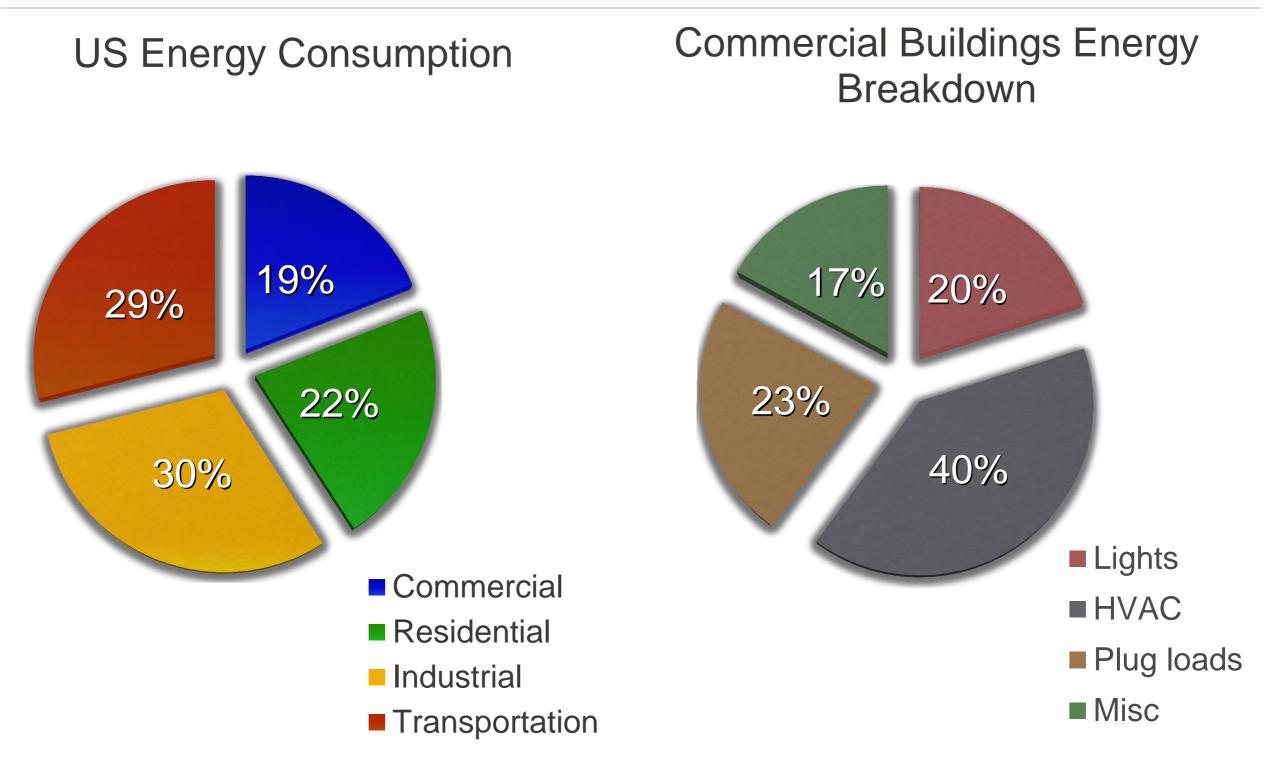


Microelectronic Embedded Systems Laboratory





#### Building HVAC is important!





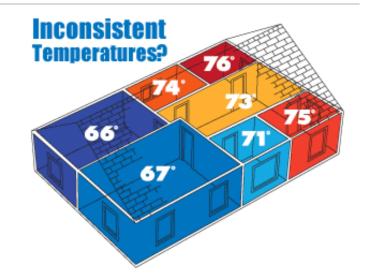


#### Buildings are already complex



Many thousands of lights





Energy hogging elevators

Strong requirements



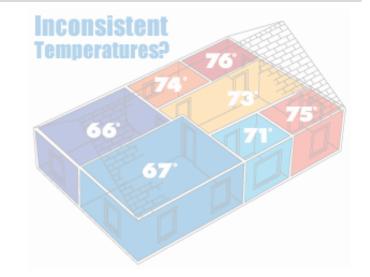


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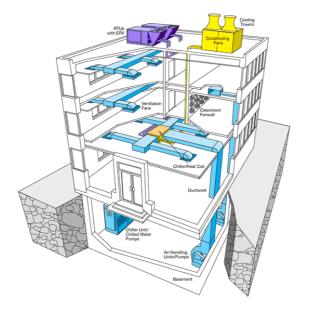
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Energy hogging elevators

Strong requirements



Complex HVAC system with many moving parts



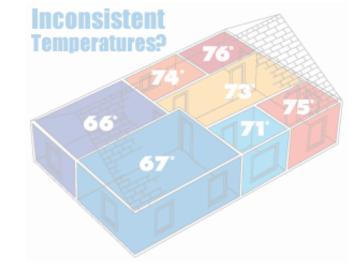


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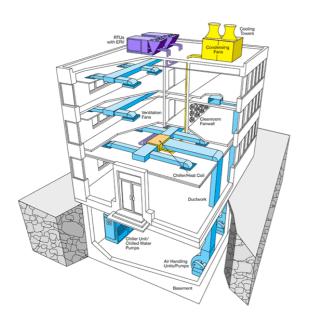


Many thousands of lights





Strong requirements



Complex HVAC system with many moving parts

#### Energy hogging elevators



#### many 10s of large fans



many 100s of dampers



10s of large pumps





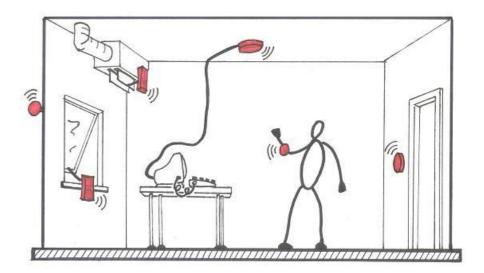
#### Buildings are becoming smarter



Better sensed



More controllable



Smart and personalized





#### Buildings are becoming more complex





Built in 2004, 145,000 sq ft, 5 floors HVAC : VAV with reheat coil, 237 zones Occupants : Faculty, staff and students 17+ sensors per thermal zone 100s of Air Handler (AHU) sensors more than 1 year data, every ~5 minutes new

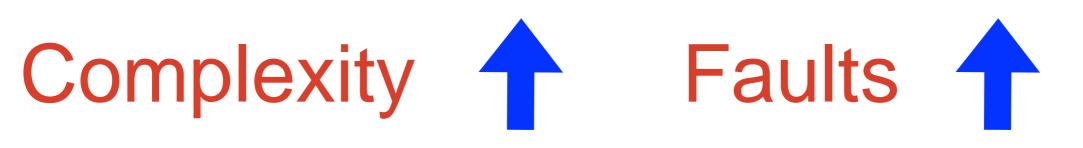
not very big

~5000 sensors More Complexity!

DATA!!

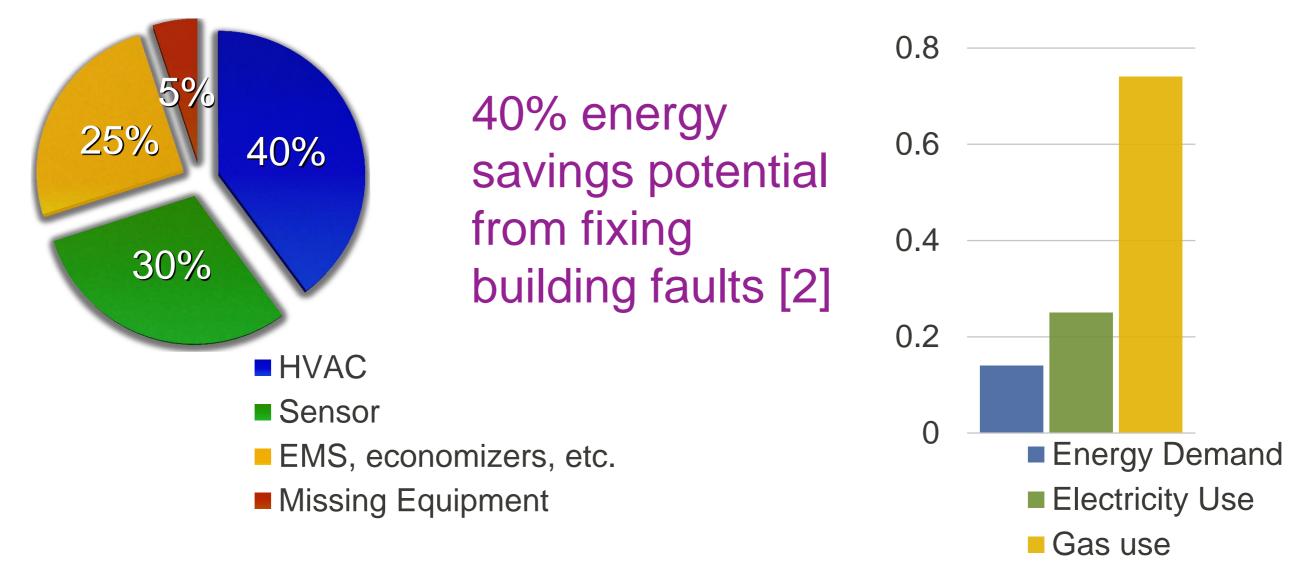






More than 50% of buildings have control faults [1]

#### Verified savings [3]



[1] M.A.Piette, et.al. Quantifying energy savings from commissioning: preliminary results from the pacific northwest. In Proc. of National Conference on Building Commissioning, 1994.

[2] J. Schein, et. al.. A rule-based fault detection method for air handling units. Energy and Buildings, 38(12):1485–1492, 2006.

[3] Selch, M. and J. Bradford (2005). Recommissioning Energy Savings Persistence. National conference on building commissioning.





#### Solution : Data aware energy use!



#### **Building Management Systems**

law Action lissen Icola Querr Help	No.				VIVI-2	
Oventew All Berns - Expert	ACCOUNTS ON THE OWNER.				Online Normal	
Overnew All terms - Expert	Summary L	Fotos	Hardware Diagnostic Snapst	Not Facure	Source and the	
User Wews	Status		tem	Value	Description	
NAE-SUPERVISOR	100000	_	VXV1-2 Zone Temperature	73.8 deg F	Zone Temperature	
BACnet Protocol Eng			VAV1-2 Network Temp Occ.	Not fiet	CS Input (Enum)	
Eth IP Datalink			VXV1-2 Supply Air Flow	274.7 cfm	Artfow Rate	
Energy			VAVT-2 Discharge Air Velocity Pressure	0.3540 in wc	Discharge Air Velocity Pressure	
Programming		n.	VAV1-2 Effective Heating Setpoint	68.5 deg F	Calculated Heating Setpoint	
Equipment List		2	VRV1-2 Effective Cooling Setpoint	72.5 deg F	Calculated Cooling Setpoint	
WW1-1_OVR		2	WKV1-2.2one Conimon Setbornt	\$20 dig F	Common Setport	
WW3-2			VXV1-2.0cc Heating Setpoint	68.0 deg F	Use this to make permanent changes	
WW1-2 Zone Temperature			VAV1-2 Occ Cooling Selpoint	72.0 deg F	Use this to make permanent changes	
VRV1-2.Network Temp Occ			VAV1-2. Occupancy Status	Occupied	Occupancy Status Display	
VAV1-2 Supply Air Flow			VXV1-2.Effective Occupancy	Occupied	Current Occupancy Command	
VRV1-2 Discharge Air Velocity Pressure			VXV1-2 Warmer-Cooler Adjust	-1.5 deg F	Zone Setpoint Adjustment (+)-2F)	
VXV1-2 Effective Heating Setpoint			VXV1-2 Supply Air Damper Output	55.2%	Supply Air Damper Output	
VWV1-2 Effective Cooling Setpoint			VXV1-2.Occupancy Schedule	Occupied	Used in Weekly Schedule	
VXV1-2 Zone Common Setpoint			VAVI-2. Occupancy Bigass Time VAVI-2. Zone Temporary Occupancy	3,600.0 seconds Inactive	Length of Temp Occupancy Zone Temporary Occupancy	
VXV1-2.Occ Heating Setpoint			VAV1-2.2018 Temporary Occupancy VAV1-2.Auto-Calibrate Command	False	Use this to calibrate box flow readings	
VXV1-2 Occ Cooling Setpoint VXV1-2 Occupancy Status			VAV1-2 Auto-Calibrate State	Normal	Current Calibration Status	NAL-1550HARBOR:NAL-1550HARBOR/TC-1.VMA12.Zone X
VXV1-2 Effective Occupancy			VAV1-2 Supply Air Flow Setpoint	275.0 cfm	Current Box Arriow Setpoint	
VXV1-2 Warmer-Cooler Adjust			VW1-2 Heating Min Flow	75.0 cfm	Use this to make permanent changes	Select the command to issue, then click Send. (Specify command
VKV1-2 Supply Air Damper Output			VAV1-2 Cooling Min Flow	75.0 cfm	Use this to make permanent changes	parameters, Krequired)
VXV1-2 Occupancy Schedule			VXV1-2 Cooling Max Flow	275.0 cfm	Use this to make permanent changes	
			VAV1-2 Unocc Heating Setpoint	E1.0 deg F	Use this to make permanent changes	terreterreterreterreterreterreterreter
VRV1-2 Zone Temporary Occupancy			VAV1-2.Unocc Cooling Setpoint	82.0 deg F	Use this to make permanent changes	Show Command Set All Commands
54 WW1-2 Auto-Calibrate Command			VXVT-2 System Mode	Auto	Current Box Operating Mode	
WW1-2 Occupancy Bypacs Time WW1-2 Zone Temporary Occupancy WW1-2 Auto-Calibrate Command WW1-2 Auto-Calibrate State		-			1.2 C C C C C C C C C C C C C C C C C C C	Adjust
VXV1-2 Supply Air Flow Setpoint						Value 72.0 deg F
VXV1-2 Heating Min Flow						Operator Override
VRV1-2 Cooling Min Flow						Operator Override
VXVI-2.Cooling Max Flow						Release Operator Override
VAVI-2 Unocc Heating Setpoint						C Release
VIIIV1-2 Unocc Cooling Setpoint						🕞 Release
VRVT-2.System Mode						Release All
WV1-3						
VAV1-4						FORMATING PROPERTY.
C VAV1-5						TRANSPORT OF COMMENT
WV1-6						And the second sec
VW1-7_0VR						
WV1-8						
WV1-9						
WV1-10						
WV1-11						
WV1-23_0VR						
RTU-2     VAV1-12						
Boler						

#### More than 10,000 alarms per day on the UC San Diego campus.

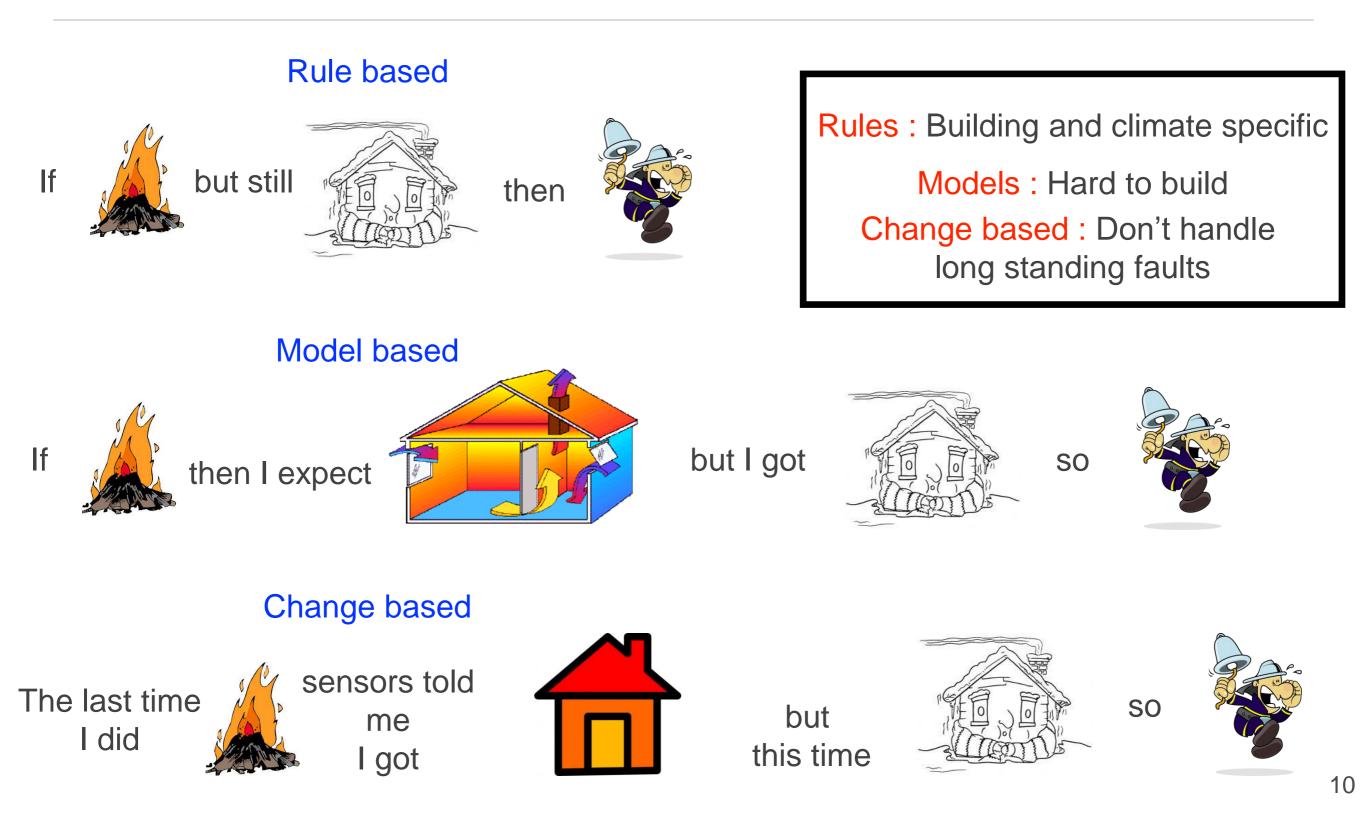


How do we bridge this gap between theory and practice?





# Prior work

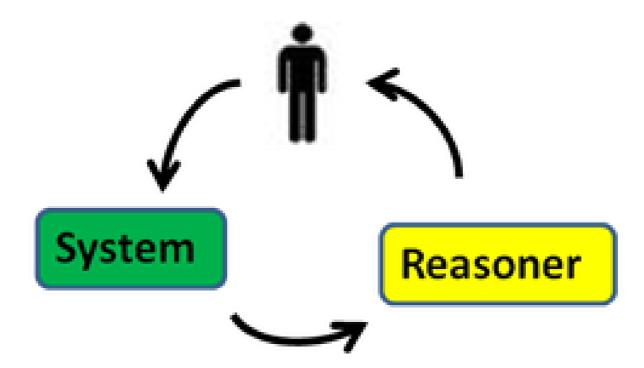






#### What this talk is about

- Qualitative evaluation of different FDD methods for Variable Air Volume (VAV) boxes
- \* To locate long standing faults we use comparative data mining
  - Comparing black-box zone models across zones can find anomalous zones -MCC
- Intelligent Rules based on data driven fault exemplars can get the best of both worlds (data and model driven)







#### Outline

- Need for building FDD
- Background on building HVAC
- Analysis of what a solution would look like
- Our Approach (Model Cluster Compare MCC)
- Evaluation
- Research Questions



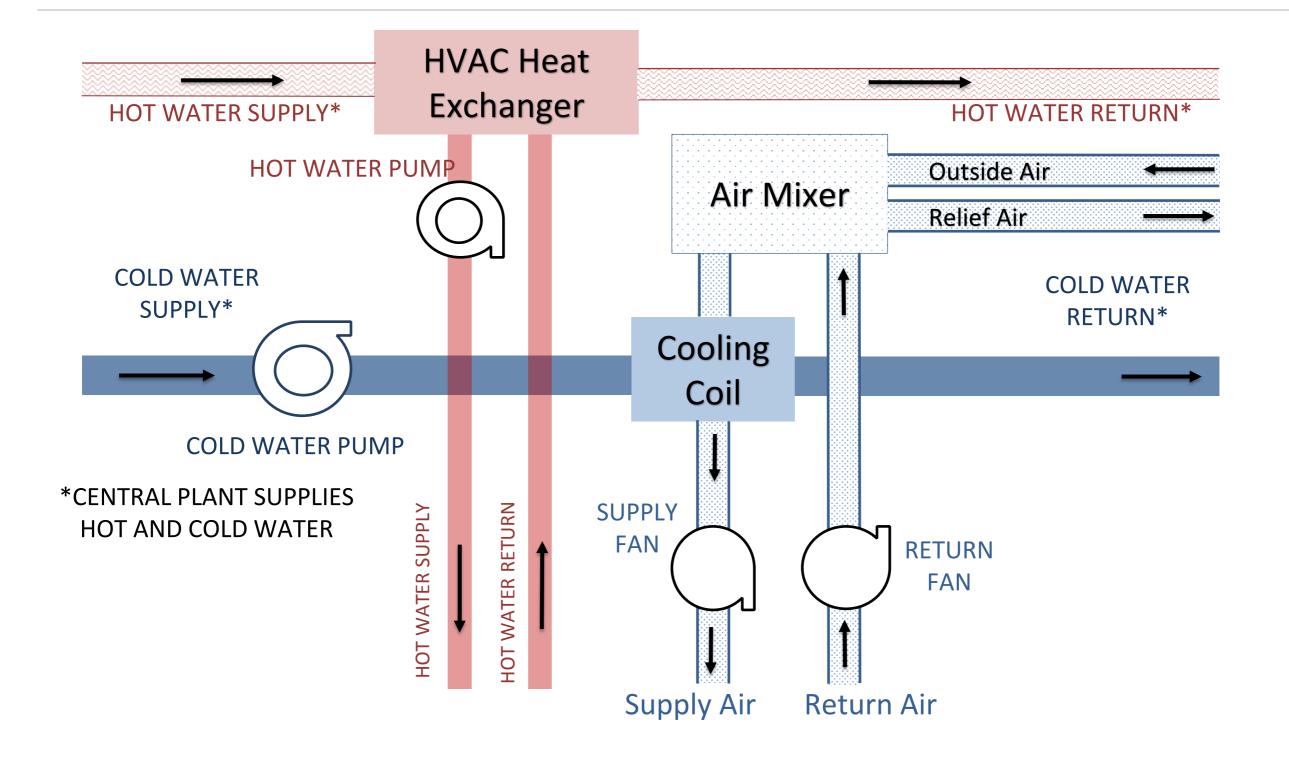


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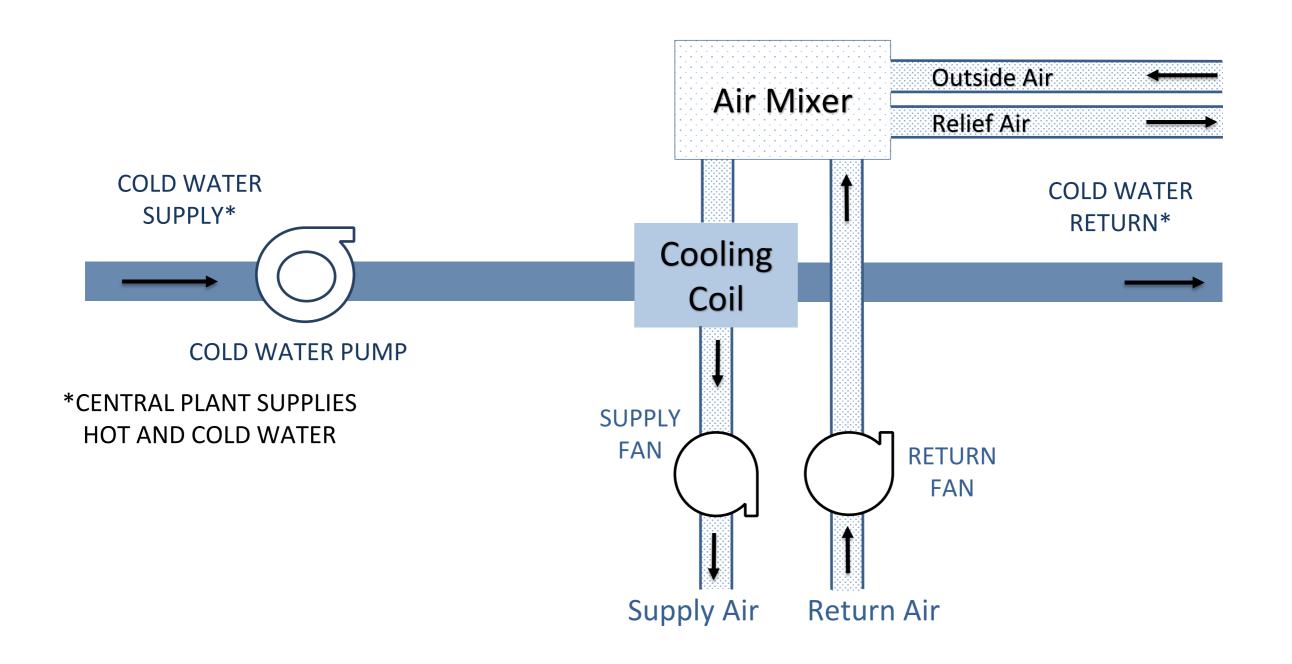






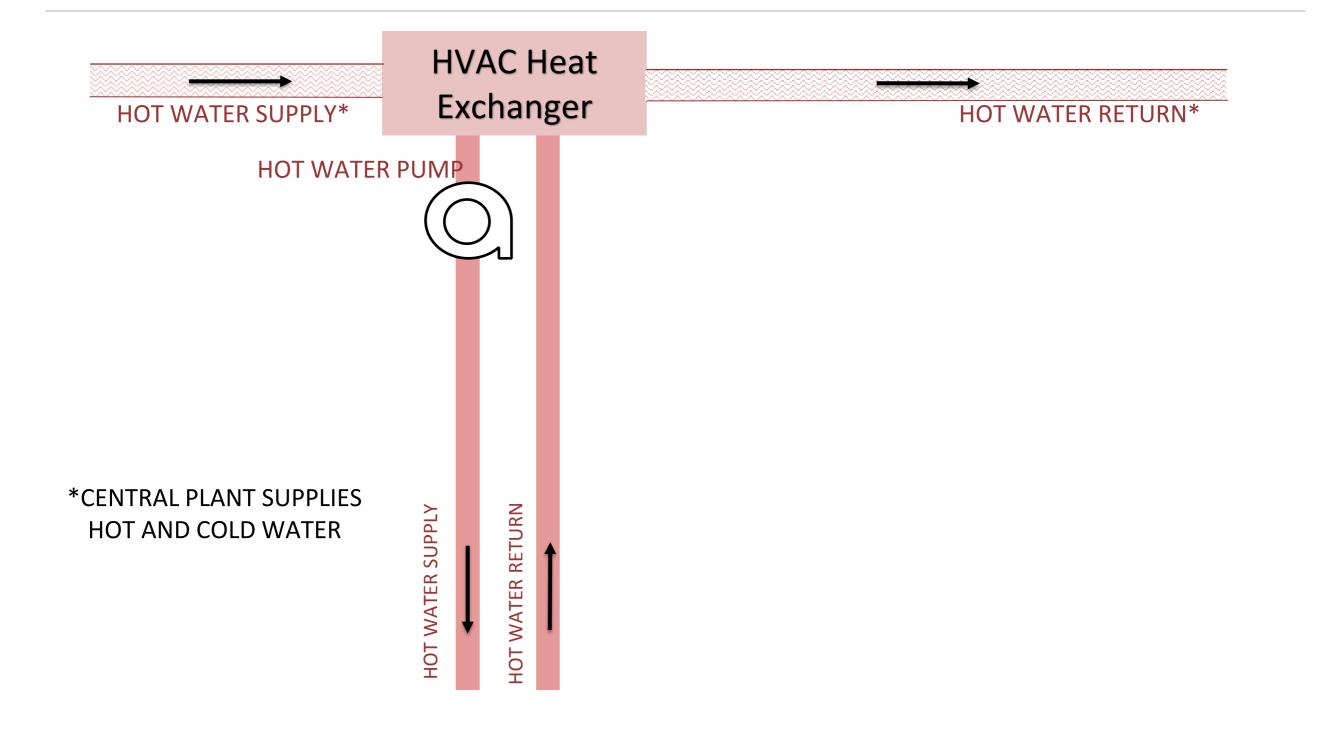








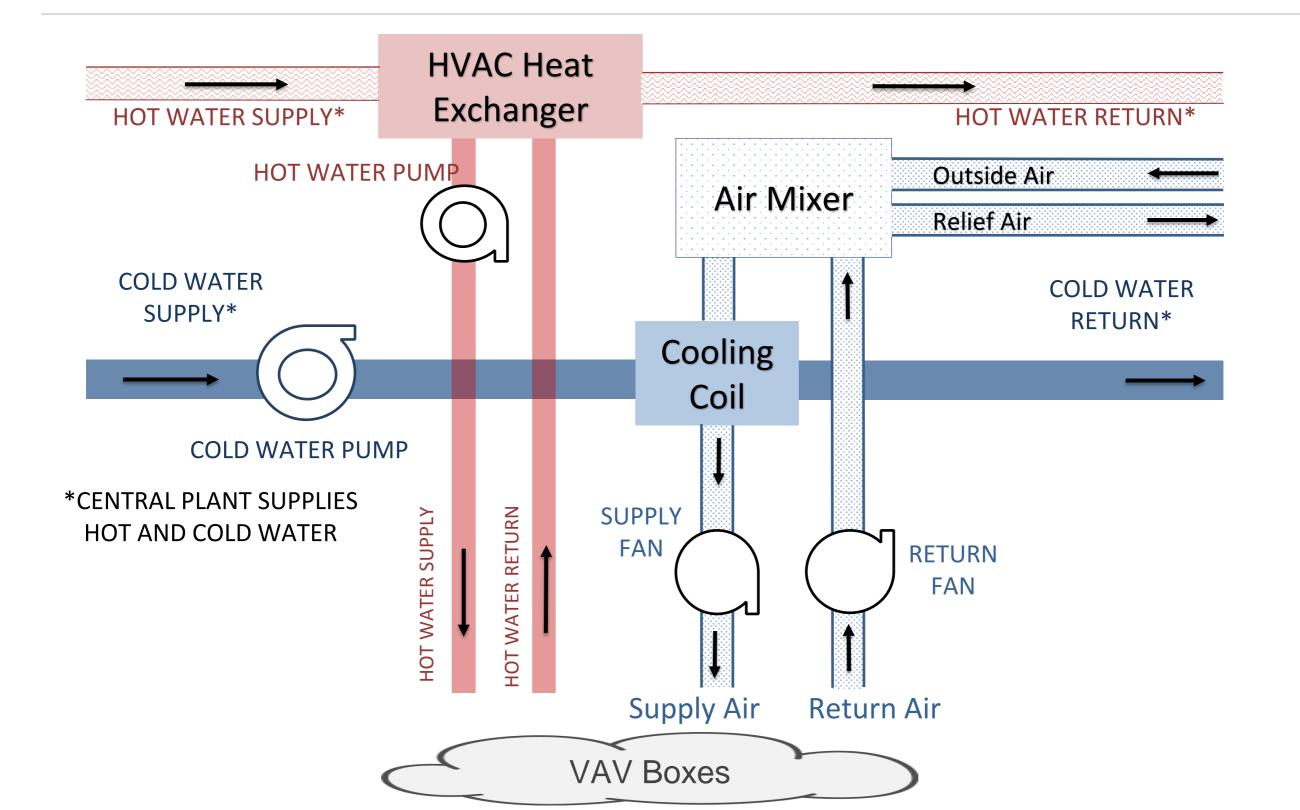






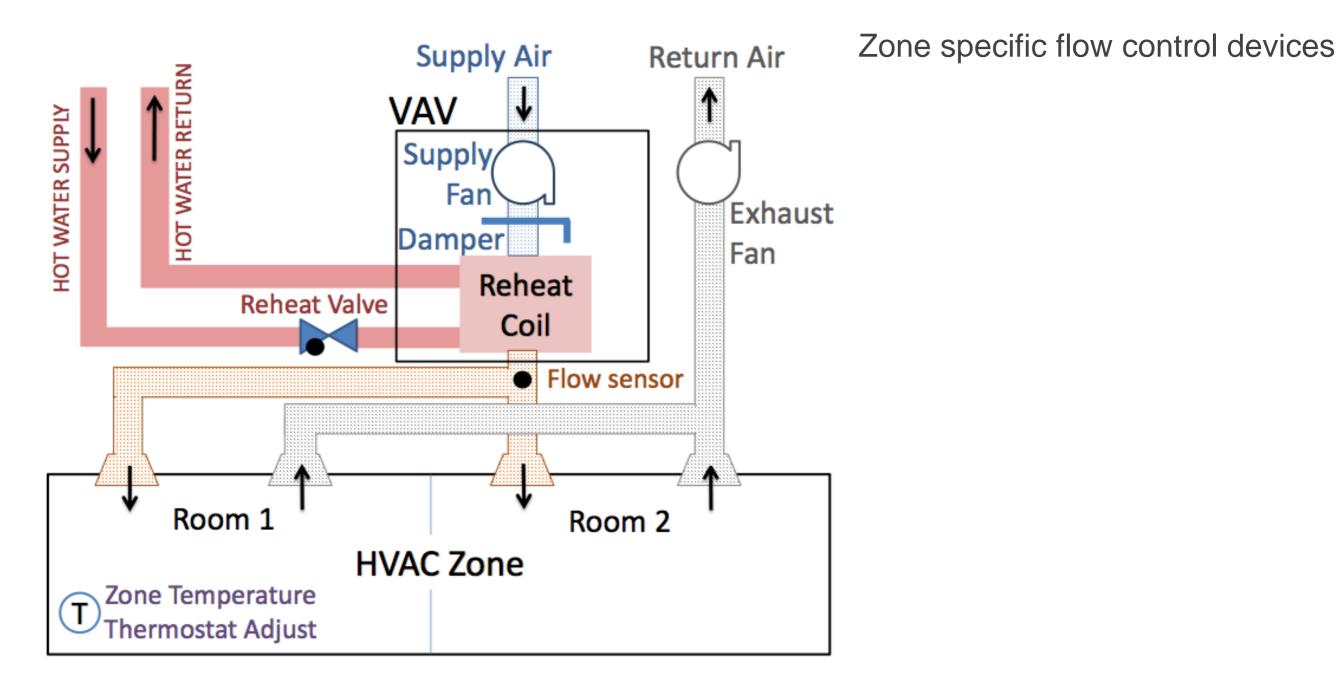


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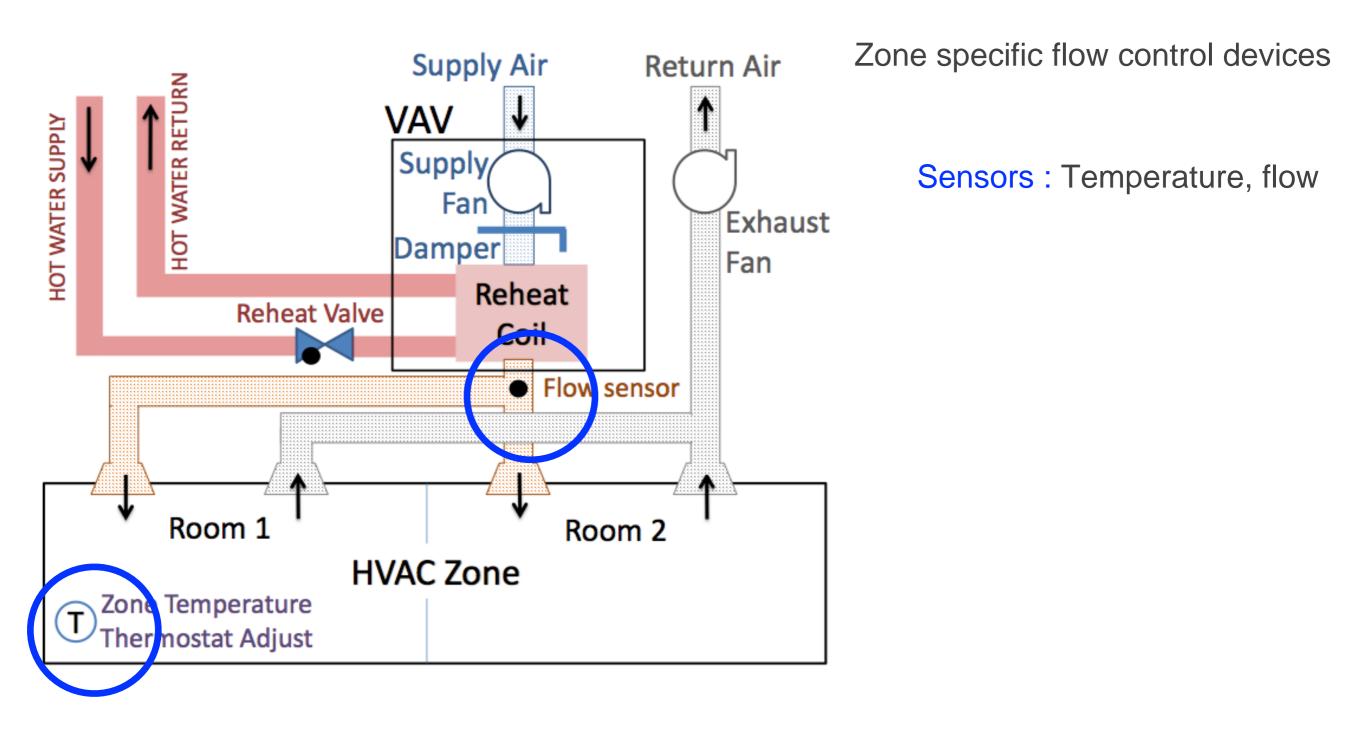






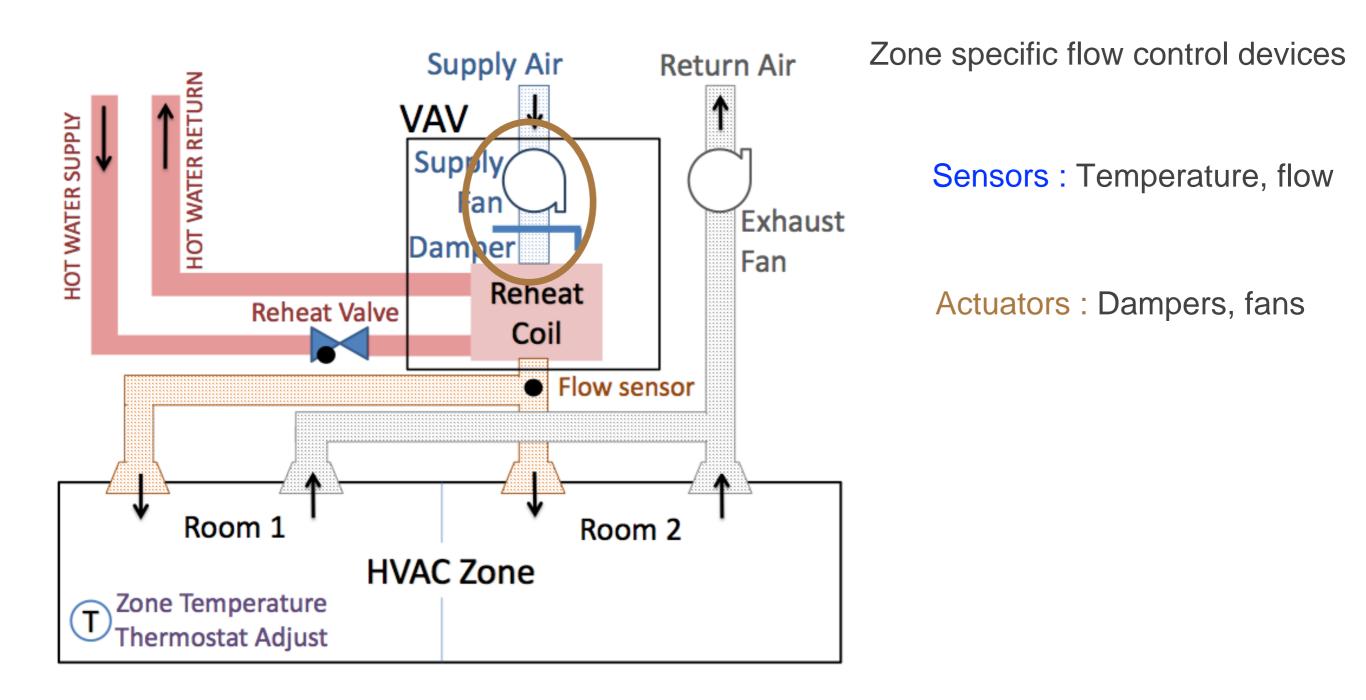






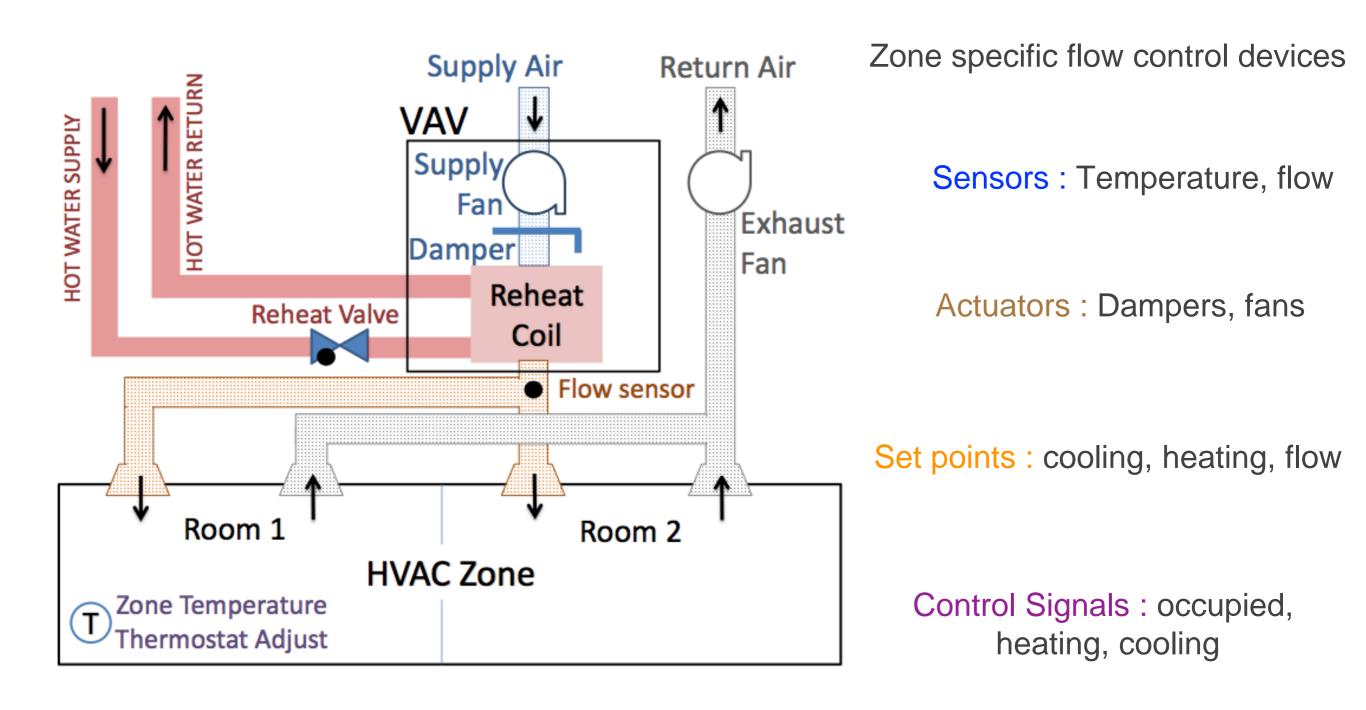








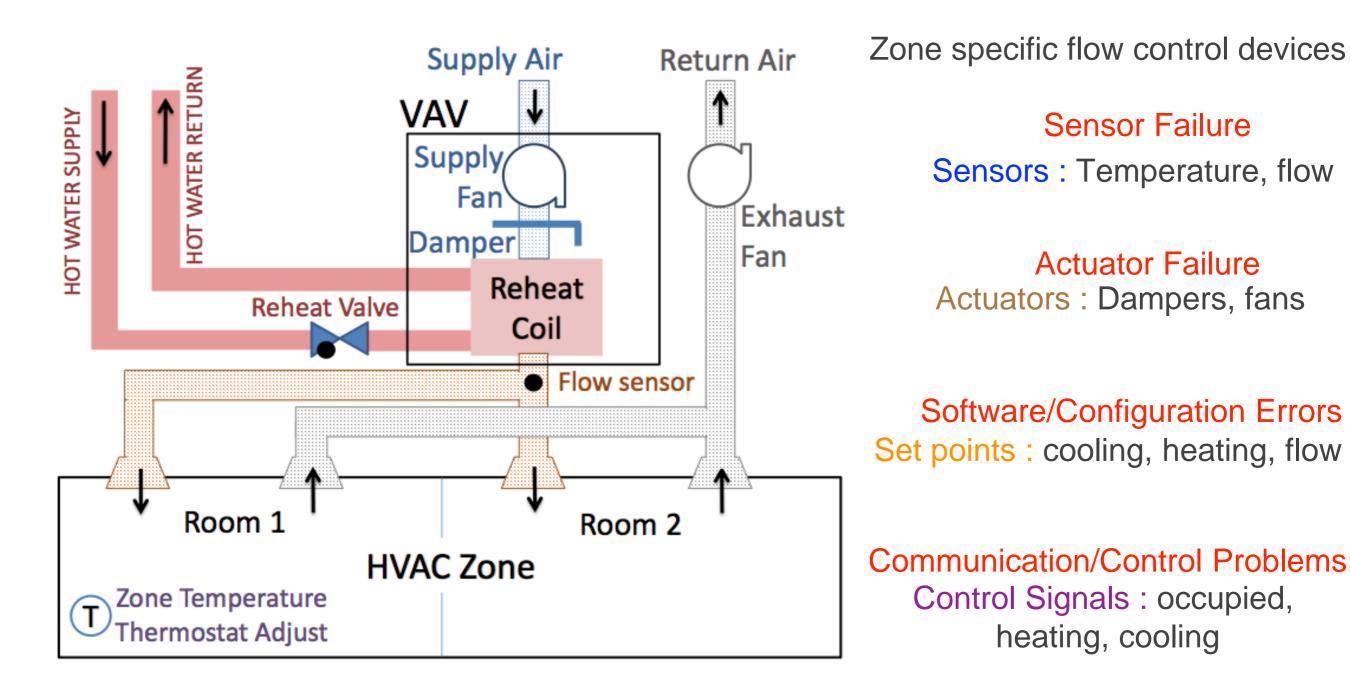








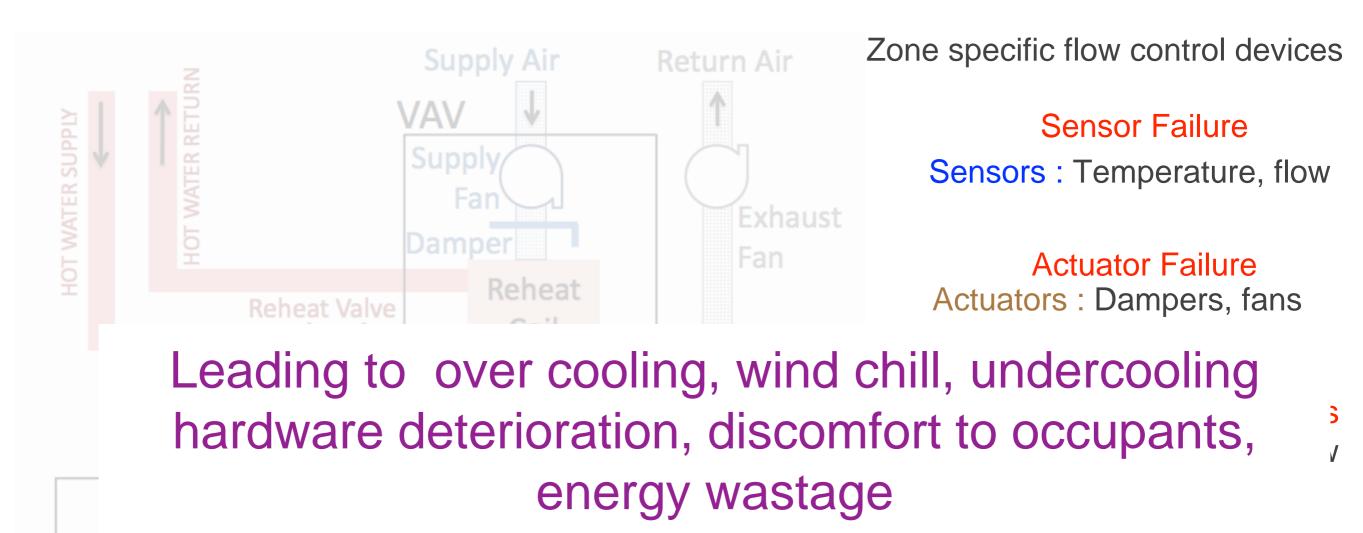
#### Background : Faults in VAV units







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**HVAC** Zone

**D** Thermostat Adjust Communication/Control Problems Control Signals : occupied, heating, cooling





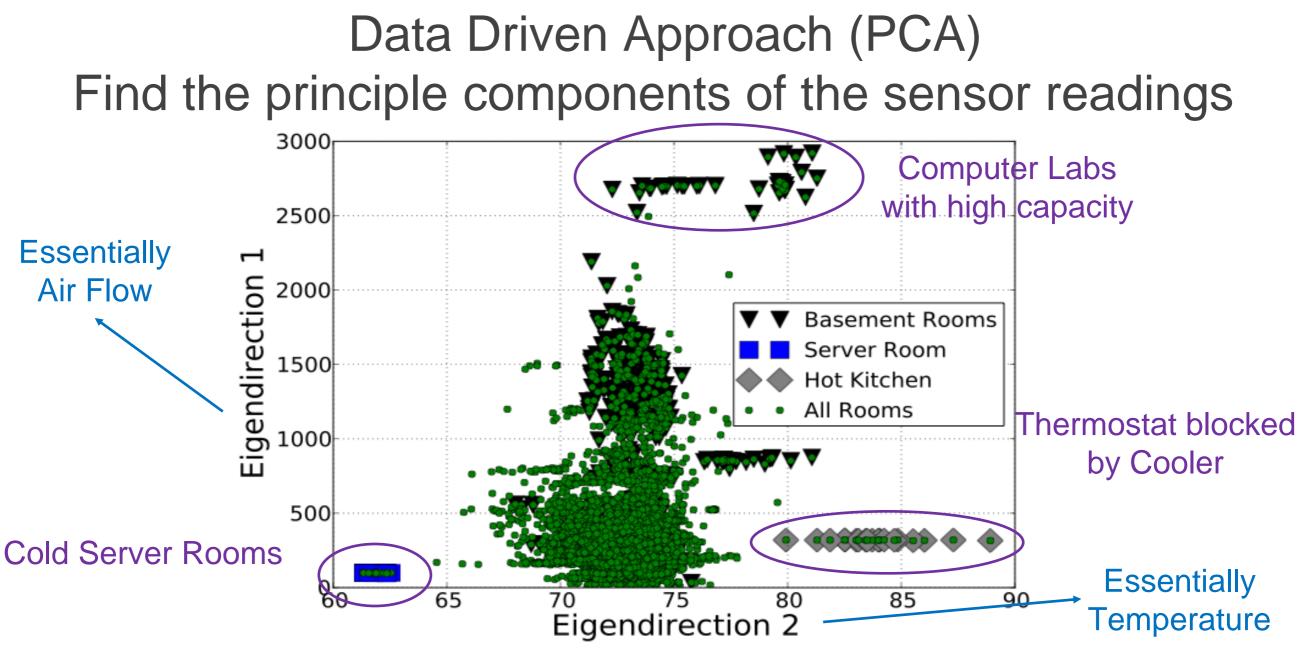
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#### Analysis : Working with sensor readings

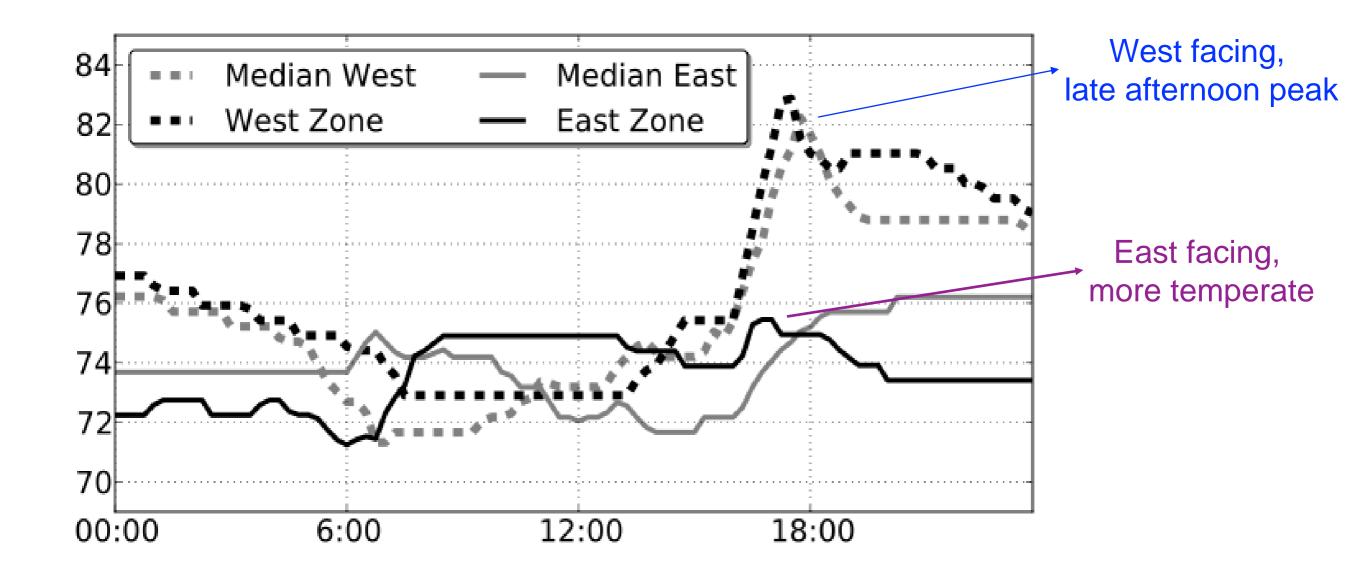


Working directly with sensor readings tends to find extreme measurements





#### Analysis : Generic rules and models



Large differences between zones even on the same day



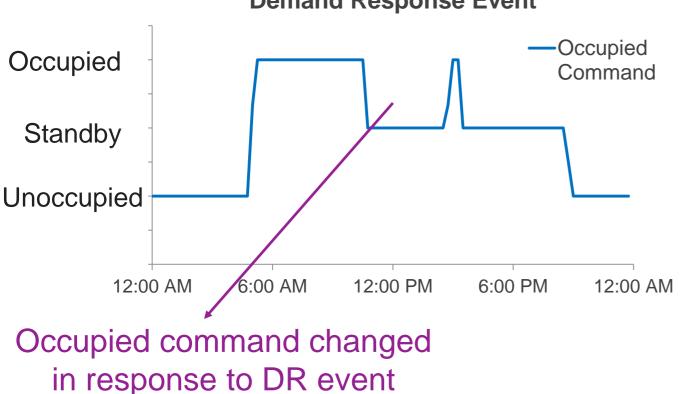


Change detection -Anomalies represent changes in characteristics of process history





Change detection -Anomalies represent changes in characteristics of process history

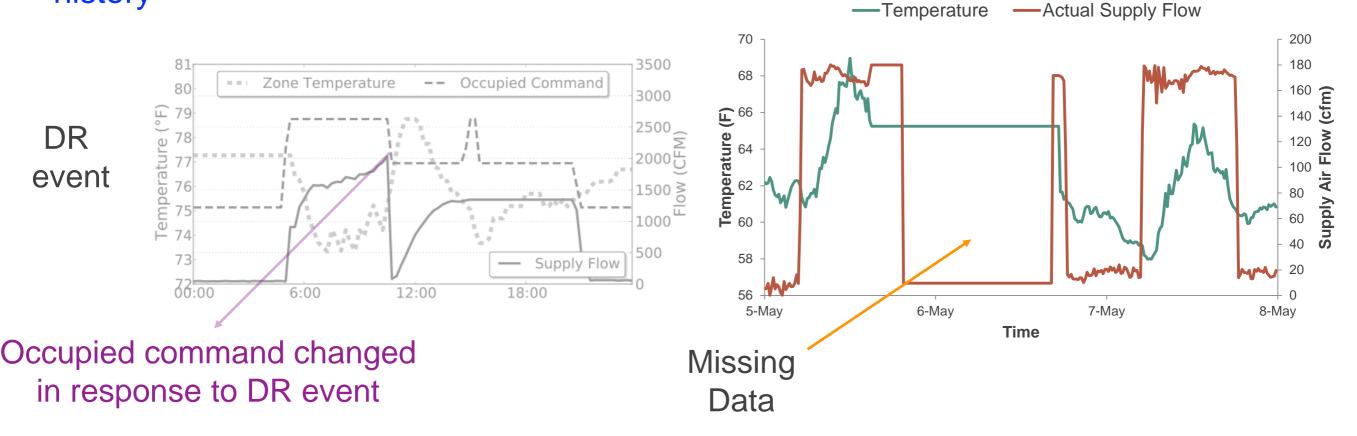


**Demand Response Event** 





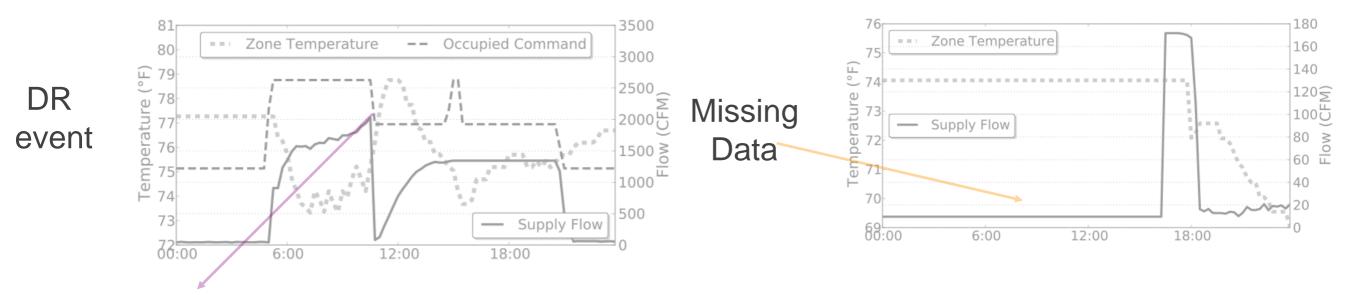
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#### Change detection -Anomalies represent changes in characteristics of process history



Occupied command changed in response to DR event

#### Sensitive to confounding parameters, like external events





# Analysis : Putting it all together

- Working directly with sensor readings tends to find extremes in sensor readings
  - Models that capture on the inter-relationships between sensors and parameters of interest
- Large differences between zones even on the same day
  - Cluster rooms with the same characteristics
- Sensitivity to confounding parameters
  - Compare rooms that have the same confounding parameters





#### Outline

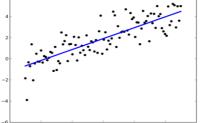
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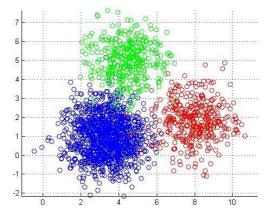




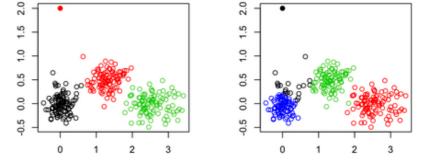
### MCC : Model, Cluster, Compare

- Build black-box statistical models of the inter-relationship of sensors in a VAV box
  - We use linear models
- Cluster these model parameters
  - We use k-means





- Compare them across time and space to identify anomalies
  - We say far away from any cluster
- Many, many, many alternatives!







#### MCC : Model

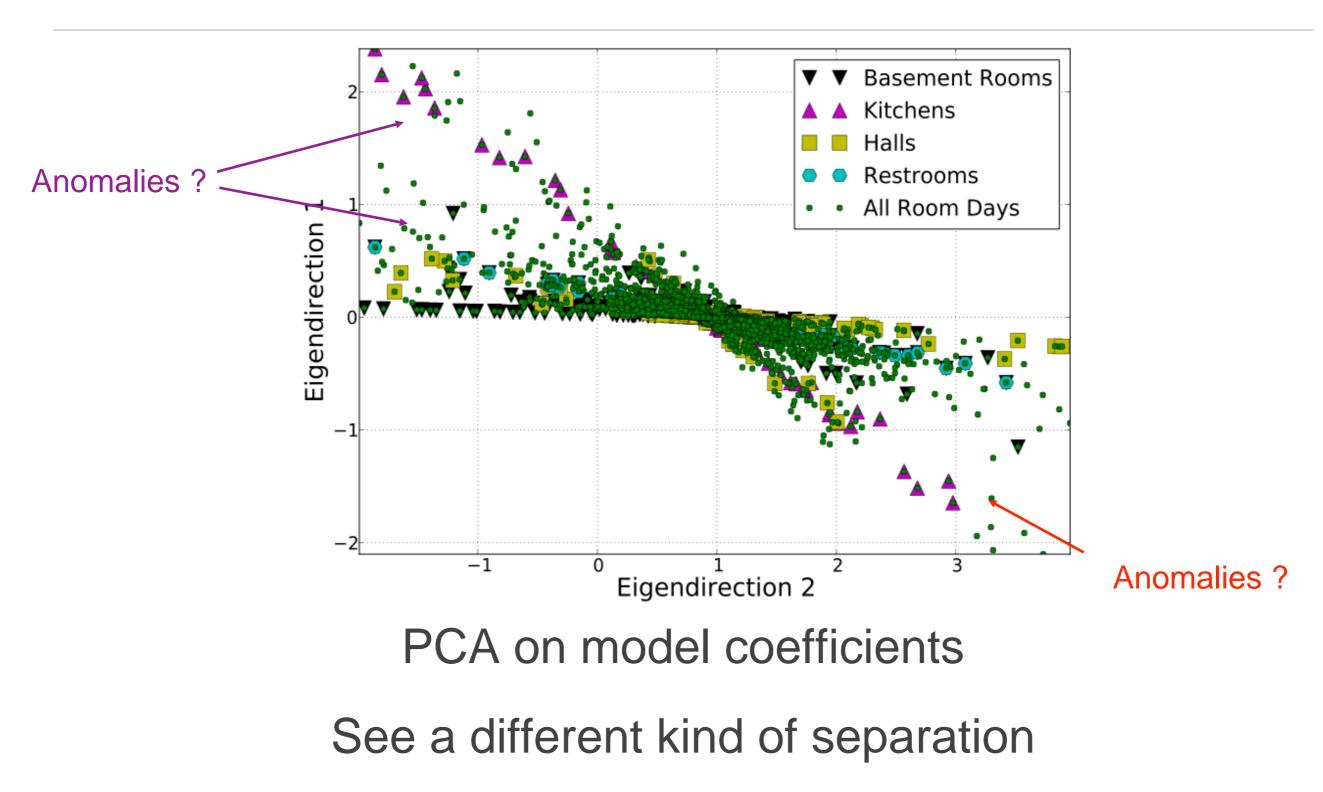
- A zone has multiple sensors and actuators
  - (temperature, air-flow, damper position etc.)
- Sense and actuate the same spatio-temporal space
- Results in analytical redundancy
  - captured by model coefficients : both structure and value

$$\begin{array}{c} \text{Model} \\ \downarrow \\ \text{Energy} \\ \text{consumption} \end{array} \xrightarrow{} E^{i}_{t+1} = \begin{array}{c} \Theta \vec{x}^{i}_{t} + n^{i}_{t} \\ \swarrow \end{array} \\ Final \\ \text{Sensor + Actuator} \\ \text{Readings} \end{array}$$





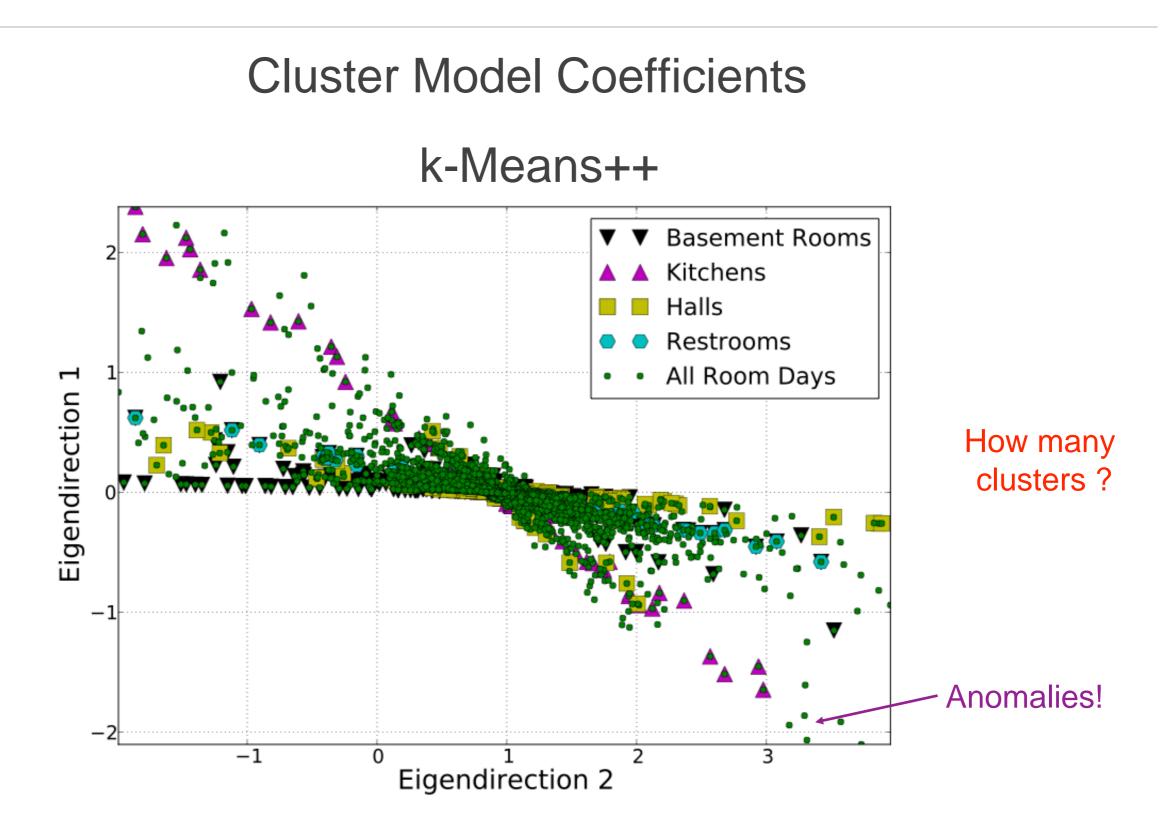
#### MCC : Model







#### MCC : Cluster and Compare







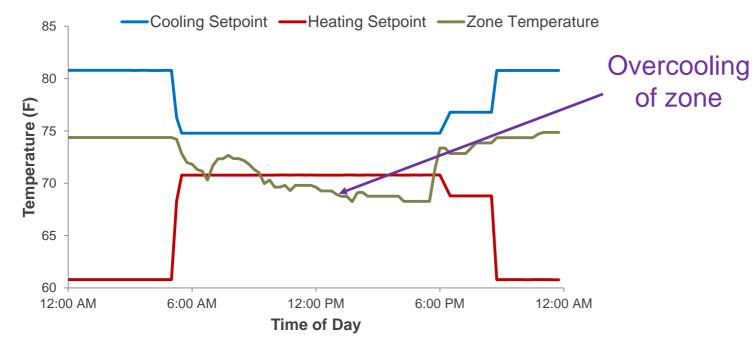
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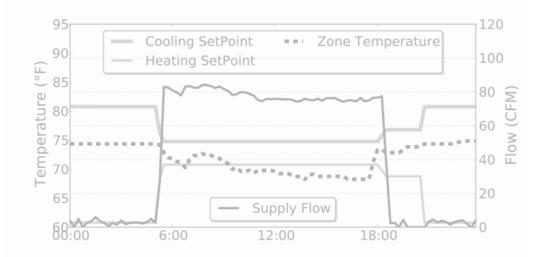
#### **Evaluation : Quantitative**



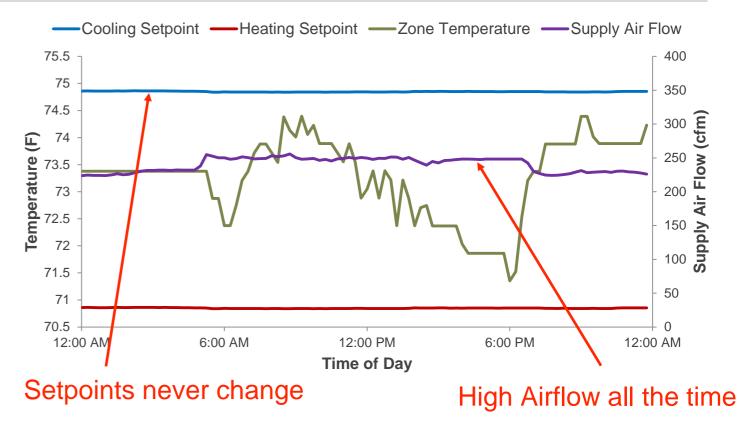
Zone temperature not within limits







Zone temperature not within limits

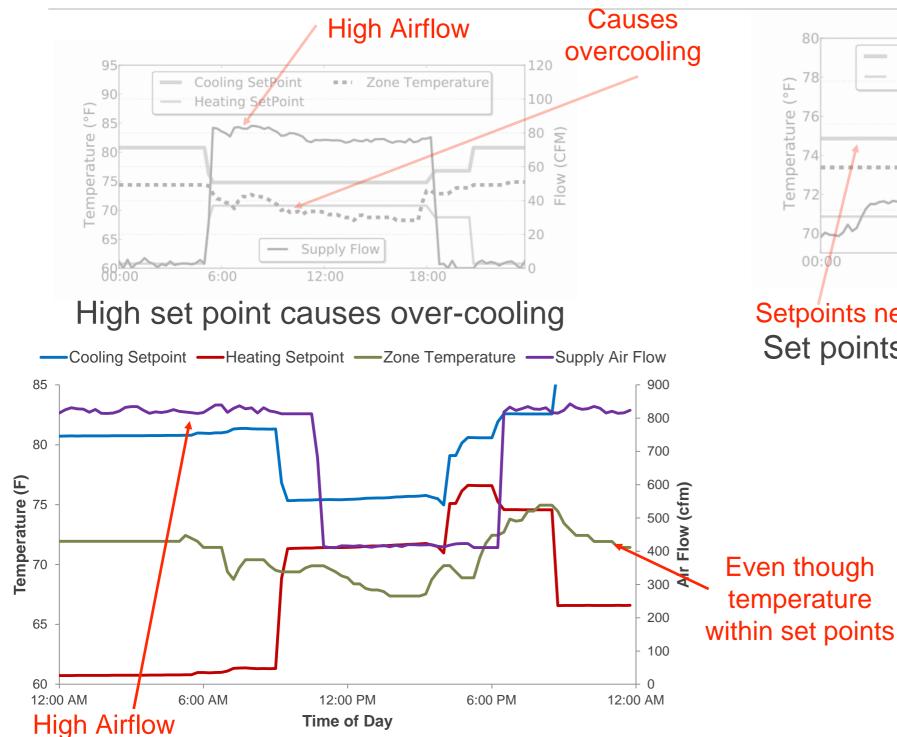


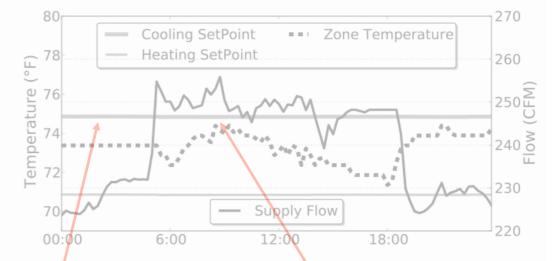
Set points and actuation not in sync





#### **Evaluation : Quantitative**





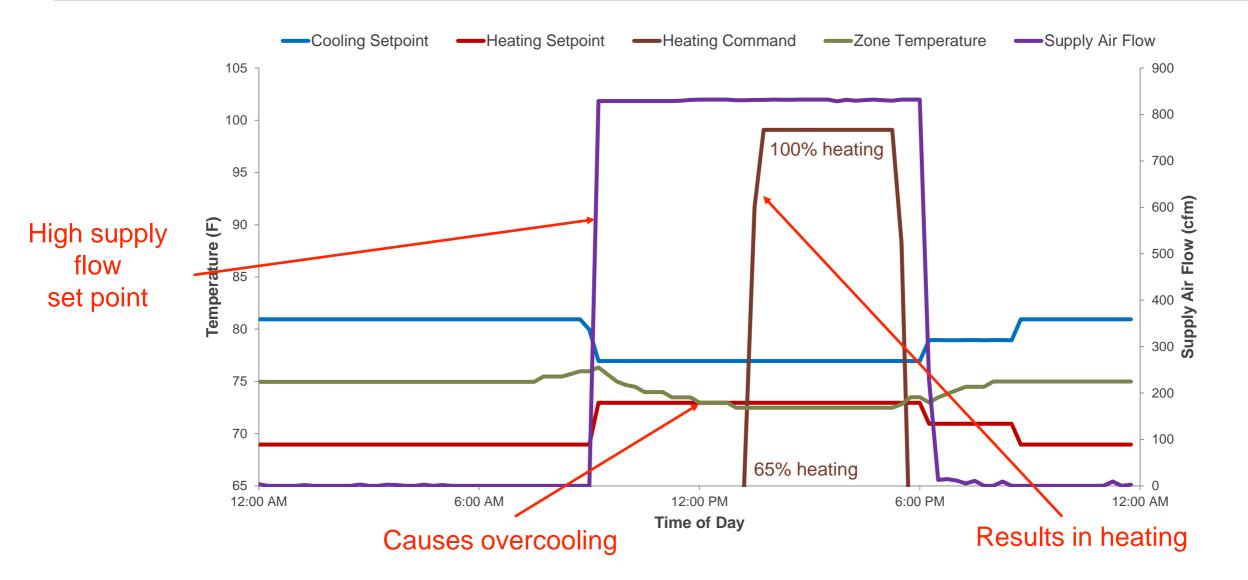
Setpoints never change High Airflow all the time Set points and actuation not in sync

(Unnecessary) high flow when within set points





### **Evaluation : Dynamic Anomaly**



'Dynamic' Anomaly - Not anomalous at any one point

**Overcooling - Heating Cycles** 

Big Energy Waste! [but hard to fix without occupancy]





# Evaluation : A problem with data driven methods

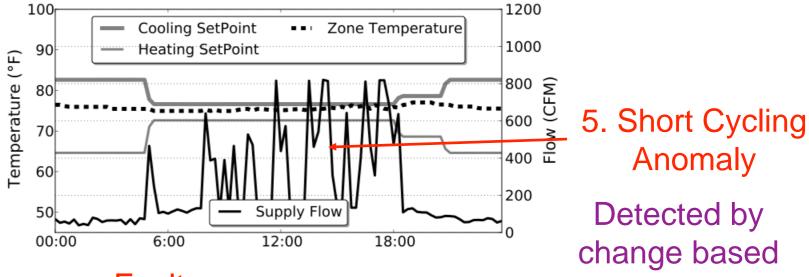
- Problem :
  - A data driven algorithm (PCA, SBS, MCC) does find 'interesting' anomalies
    - Different kinds of anomalies all useful ?
  - However, they don't find all instances of every anomalies
- Solution :
  - Design rules based on fault exemplars
  - \* iRules (don't sue us) Intelligent Rules





#### **Evaluation : Qualitative**

- 1. Heating or Cooling is Ineffective
- 2. Excess Flow
- 3. Heating-Cooling Anomaly
- 4. Temperature and Air Flow not within set points



#### **Faults**

Anomaly
Detected by
change based
methods

Method	Data	Ineffective	Excess Flow	Heating-Cooling	Set-Point	Cycling	False Alarms
Manual	Fourth floor, 30 days	7	1	3	2	1	0
Rules	Fourth floor, 30 days	4	1	2	2	0	2
Correlation	Fourth floor, 30 days	1	1	0	0	1	8
PCA	Fourth floor, 30 days	1	0	1	2	1	5
MCC	Fourth floor, 30 days	5	1	3	1	0	2
Rules (top 10 verified)	All Floors, 1 year	0	0	0	10	0	0
Correlation (top 10 verified)	All Floors, 1 year	0	1	0	1	0	8
PCA (top 10 verified)	All Floors, 1 year	1	0	0	1	0	8
MCC (top 10 verified)	All Floors, 1 year	0	2	6	1	0	1
iRules	All Floors, 1 year	18	12	14	27	7	1

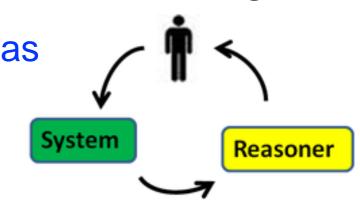
#### Performance





#### Conclusions

- Buildings are a big source of data
  - \* Too big ? Data mining, visualization, HCI are critical
- Despite data- Many faults, many long-standing faults, much energy waste
  - Data driven anomaly detection is useful
- Generalization to new buildings, climates is hard
  - Within building comparative data mining is useful
- We find common areas [kitchens, conference rooms] often mis-configured
  - Value of occupant count based control in these areas
- Data mining good, human in the loop better



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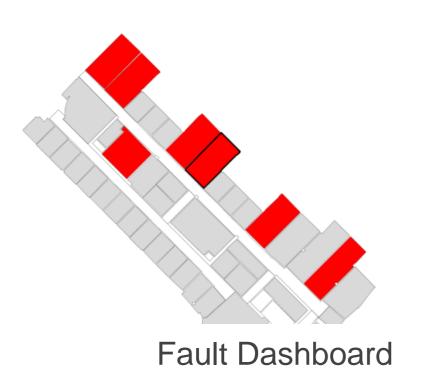




#### BuildingSherlock: Fault Management

Framework to detect faults and prioritize fixing them

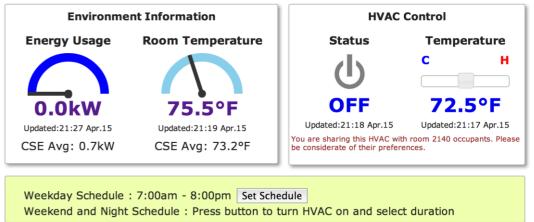
- Make sensor data and metadata available
- Allow complex fault detection algorithms
- Provide contextual information on faults



Abnormal	En
Normal	Energy U
4242 HVAC Zone:4242 Temperature:74.57F Cooling Power:2.72kW Heating Power:0kW Electrical Power:0.07kW	0.0k Updated:21:27 CSE Avg:
	Weekday S Weekend a



EBU3B - 2140 (Research Laboratory/Studio)





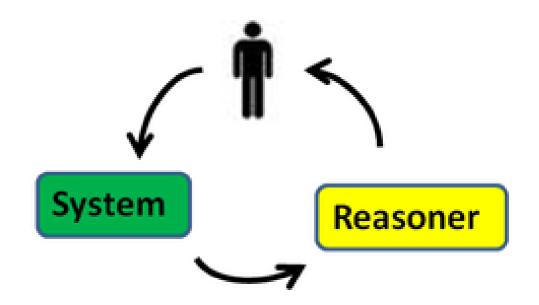




#### Future Work

- Specific Comparative data mining
  - \* How many clusters ? Other clustering methods ?
  - Fault Diagnosis
  - Better (non-linear) room models
  - Other buildings, climates
- More general How much human in the loop ?
  - Automatic rule generation
  - What is anomalous ?
  - Automatic fixing of software errors ?
  - Automatic fixing of hardware errors ?









#### Thank You!



Acknowledgements: Anna Levitt, Rizhen Zhang, Anthony Nwokafor, UC San Diego Facilities Management