



A Platform Architecture for Sensor Data Processing and Verification in Buildings

Jorge Ortiz

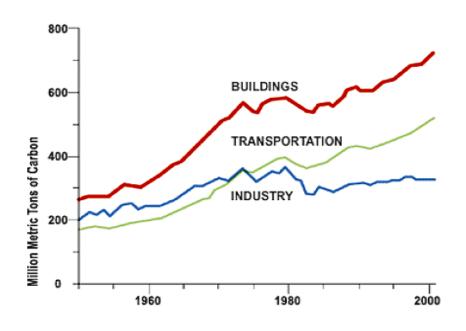
University of California Berkeley

http://www.cs.berkeley.edu/~jortiz

Buildings As Prime Target

Buildings

- 72% of electrical consumption
- 40-50% of total consumption
- 42% of GHG footprint
- US commercial building consumption doubled 1980-2000, 1.5x more by 2025 [NREL]
- 4.8 M Commercial &
 350,000 industrial buildings



Renewable energy consumption

Represent wider body of computational and information challenges of complex physical systems

Huge Sensor Networks

151 Temperature Sensors





312 Light Relays



> 6,000 Sense and Control Points

50 Electrical Sub-meters



12 Variable Speed Fans



138 Air Dampers





Sutardja Dai Hall Built in 2009 140k sq. ft.

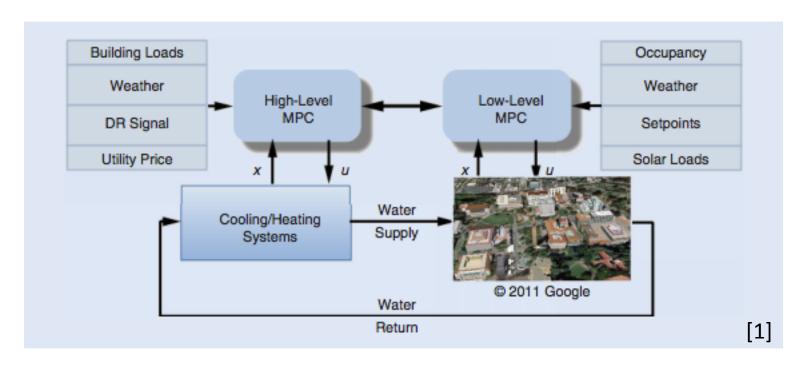
6 Variable Speed Pumps



121 Controllable Valves



Holistic Building Optimization



- Aggregate load decisions trickle down
- Local control based on local conditions

[1] Yudong Ma; Kelman, A.; Daly, A.; Borrelli, F., "Predictive Control for Energy Efficient Buildings with Thermal Storage: Modeling, Stimulation, and Experiments," Control Systems, IEEE, vol.32, no.1, pp.44,64, Feb. 2012

Diagnostics and Control with Mobiles Phones

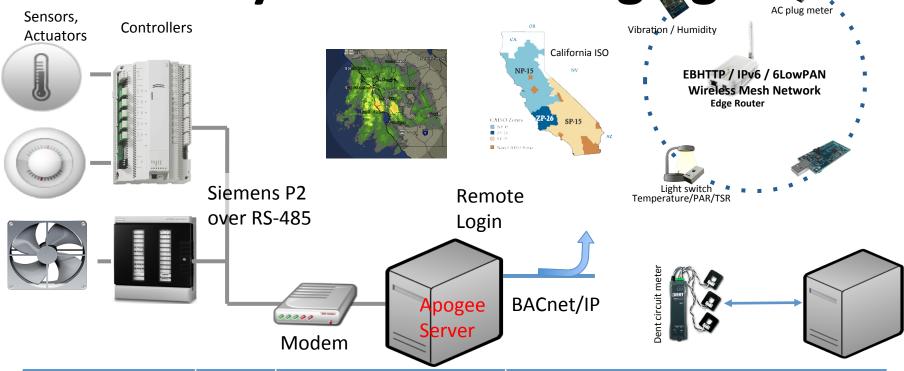






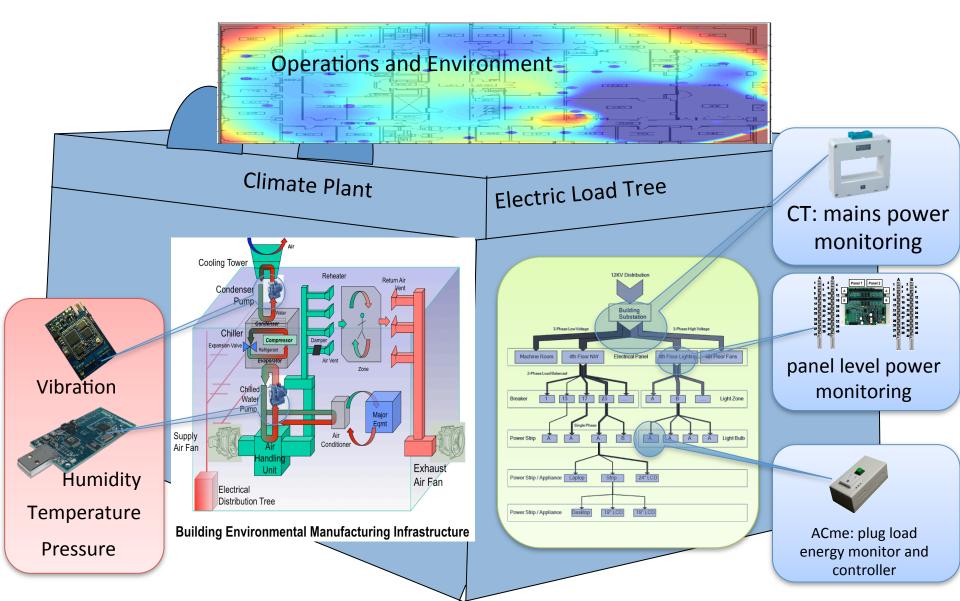
We need to manage building sensors and actuators in a unified fashion





Protocol	Year	Network	Example Applications
Modbus	1979	RS-485, TCP/IP	Panel monitoring, alarms
Fieldbus/HART	1988	various	Industrial Control
BACnet	1995	ARCNET, Ethernet, IP, RS-232, etc.	HVAC, Lighting, Fire
WirelessHART	2007	802.15.4e	Industrial control, wire replacement
Zigbee SEP 2.0	2011?	802.15.4	Plug-load monitoring

Why is this challenging?



Complications

- Thousands of sensors in separate silos
 - No holistic view, information sharing or extensibility
- Many legacy/emerging standards and interfaces
 - No integration, portability, scale
- Buildings are living entities, constantly evolving
 - Difficult to manage

Solutions

- Systemic view through OS Abstractions
 - Focus on filesystem abstraction, data services

- Continuous verification of physical relationships
 - Office re-arranged, walls moved

Solutions

- Systemic view through OS Abstractions
 - Focus on filesystem abstraction, data services

- Continuous verification of physical relationships
 - Office re-arranged, walls moved

Sensors/Actuators As Resources

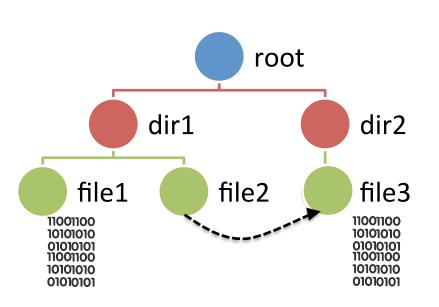
Data producing, data consuming

Uniquely identifiable

User-driven multi-categorization

Direct, uniform access

Where have we seen this before?



Unix filesystems!

 Resource exposed via namespace

FS controls access to resources and collections

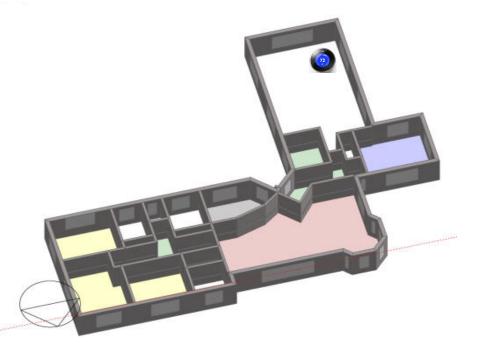
Sensors as Files



uid:d73a3040e095

temp

owner:jorge installed:06/28/10



/room/410/temp

/room/420/temp

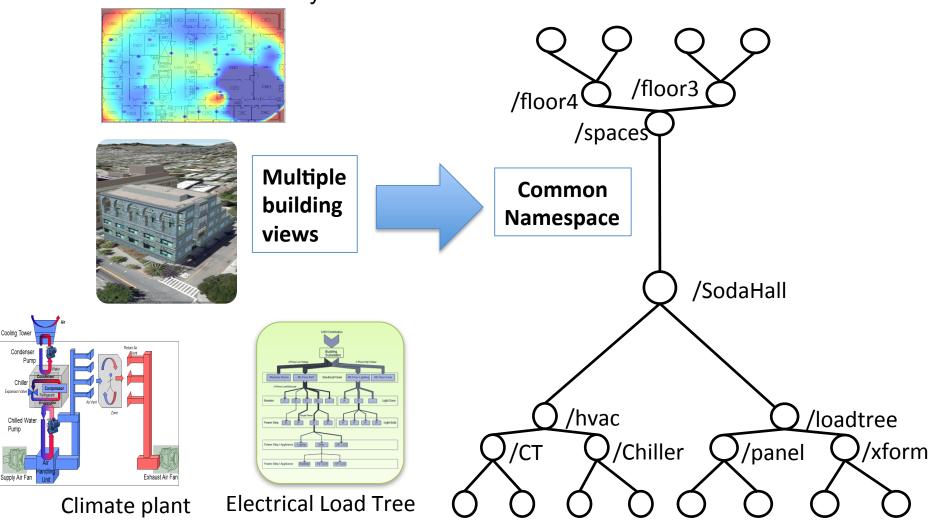
*

*

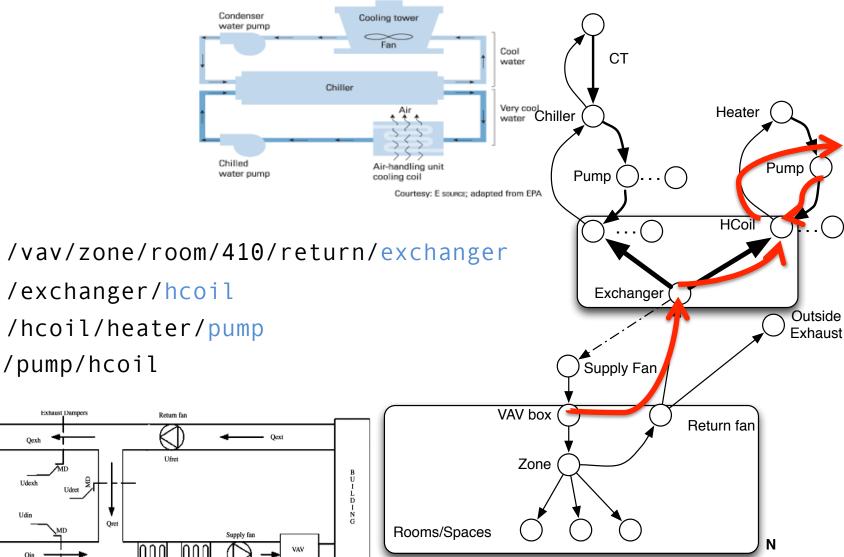
*

Views Captured in Namespace

Environment and Activity



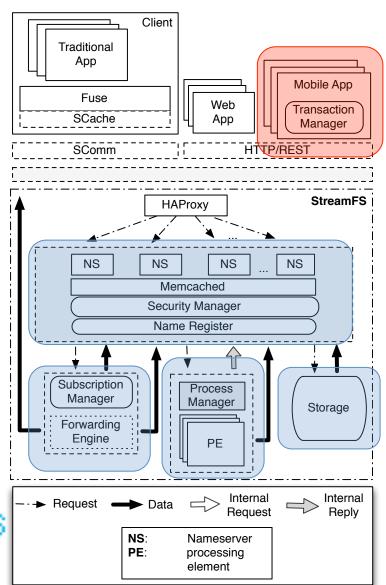
Capturing Inter-relationships



Implementation

- ~20k lines of code (Java)
- 4 core components
- Web service architecture





Application: The Mobile Energy Lens

Energy Lens

Home / About / Grapher / QRC Generator / Download

Track Your Energy. on Android



Generate QR Codes to place on your personal items. Tagging items allows you to create a virtual representation of your device that enables the coupling of virtual services on your devices, such as the device energy viewer or the personal energy view or control it remotely with the remote control app.

Register

Register your tagged items using the Android-based mobile application. Using the mobile application you can register your item, input information about the item and attach virtual services on it as well. Make sure to set the deployment configuration on the phone before getting started.

Swipe

Swipe the QR codes in the world arond you! By swiping QR codes in the physical world, you can pull up the virtual services associated with that item, If an item moves, make sure to reswipe the item in the new location, that way we can provide you with context-aware virtual services.





Energy Lens Challenges

- Tracking people and things
 - Active Aggregate Statistics

Consistency Management

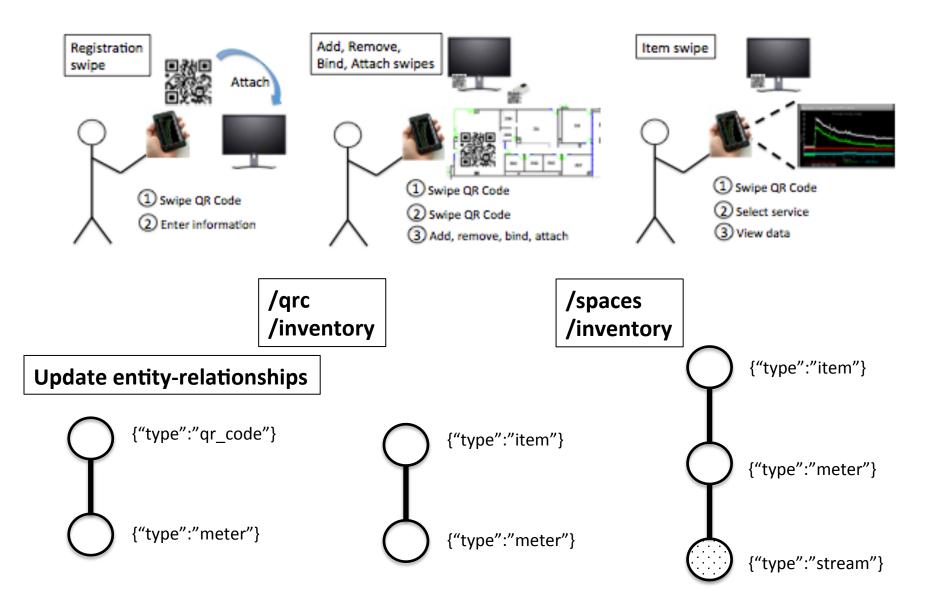
Disconnect Operation

File Abstraction

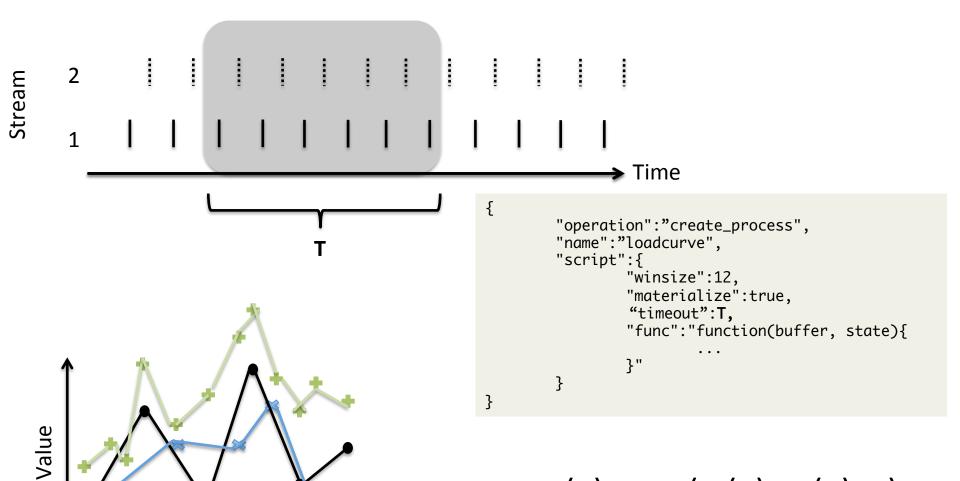
- 4 Types
 - container, stream, ctrl, special

- Canonical container names
 - /dev/ctrl
 - /dev/strm
 - -/proc

Swipes Capture Links



Stream Resampling



Time

$$s_3(t_i) = sum(s_1(t_i), s_2(t_i),...)$$

Process Management Through FS

/proc/loadcurve/

```
{
    "status": "success",
    "type": "PROCESS_CODE",
    "properties": {
        "operation": "save_proc",
        "name": "slope_intercept_tester",
        "script": {
            "winsize": 10,
            "materialize": "true",
            "timeout": 80000,
            "func": "
                function(buffer, state){
                 loadcurve.push({'ts':cleaned
                [thissubid][ts_idx].ts,
                 'value':sum});
                outObj.data = loadcurve;
                 return outObj;
    "children": [
        "eda76b7a3b23"
```

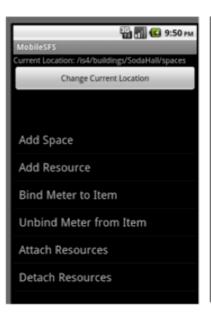
/proc/loadcurve/eda76b7a3b23/

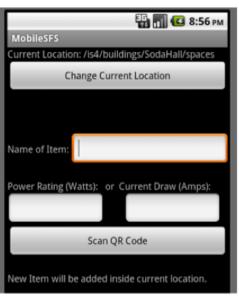
```
{
    "status": "success",
    "pubid":
"85e23cce-2378-4c85-9395-09be857e40ff",
    "head": {
        "value": 3271.6666672825813,
        "ts": 1345573470
    },
    "properties": {}
}
```

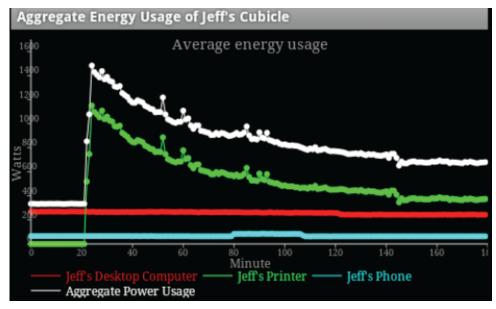
/sub/1b524cb/

```
{
    "status": "success",
    "pubid":
"85e23cce-2378-4c85-9395-09be857e40ff",
    "head": {
        "value": 3271.6666672825813,
        "ts": 1345573470
    },
    "properties": {}
}
```

Collection & Live Aggregation

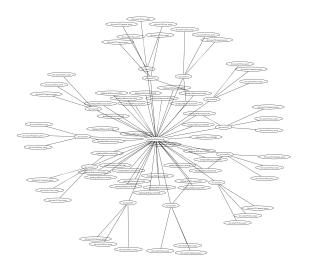


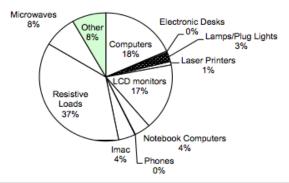


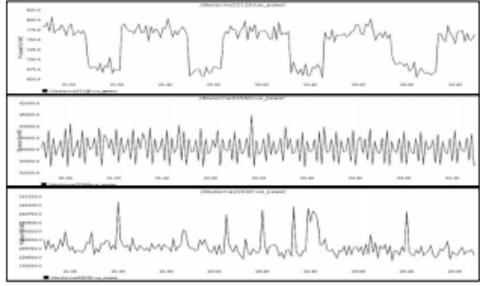


Live Statistics On Building Inventory

- SDH deployment
- Registered and classified
 - 20 meters
 - 371 loads
 - 139 rooms
 - 7 floors







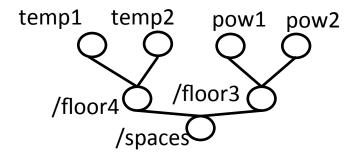
Solutions

- Systemic view through OS Abstractions
 - Focus on driver interface, filesystem abstraction

- Continuous verification of physical relationships
 - Office re-arranged, walls moved

Verification

Inter-relationships manually specified



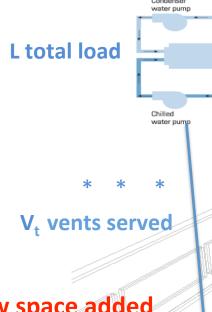
Geometric Relationship: relative spatial associations

Can we use the sensor data to determine *geometric* relationships?

Models Depend on Geometry

$$C^*\Delta T = u + P_d + (T_{oa} - T)/R$$

- C thermal capacitance
- T temperature in the space
- u heating/cooling power input
- P_d occupant/computer load
- T_{oa} outside air temperature
- R is the resistivity of the walls



Very cool

V_s vents in

space

y: E source; adapted from EPA

New space added

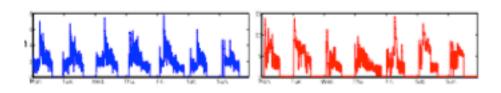
V_s changes per space

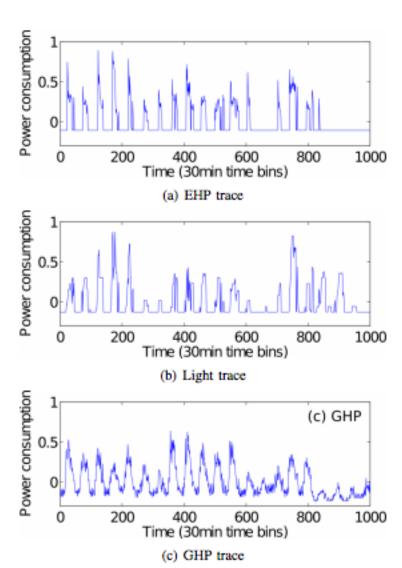
Mapping vents to zones/spaces must be accurate

How can we automatically discover changes made to the physical layout?

Similar Trend in Data Streams

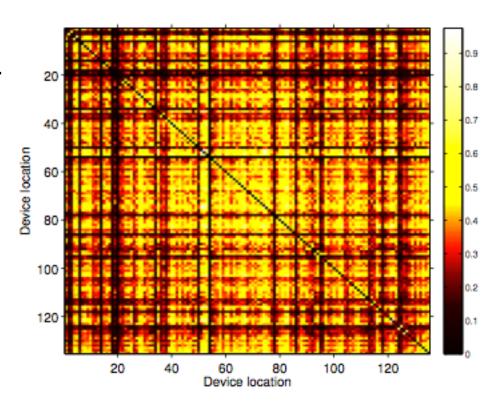
- Sensor streams driven by same phenomena
- Common trend ineffective at uncovering relationships





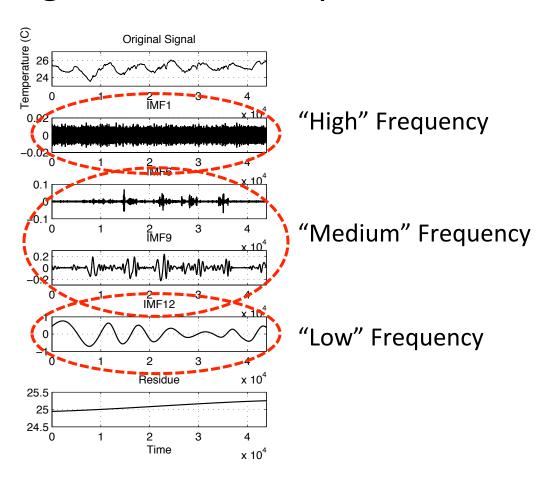
No Discernible Correlation Pattern in Raw Traces

- Each row/column is a location in the building
 - Each location has one or more sensors
- Cell (i,j) is the average device pairwise correlation between sensors at locations i and j



An example...

Applying EMD on a temperature trace



Experimental Setting

University of Tokyo

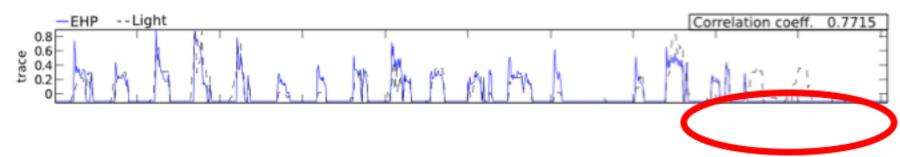


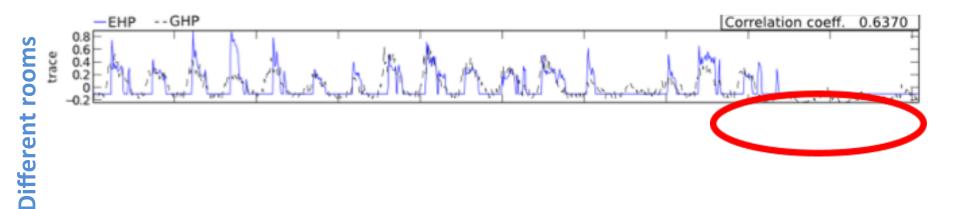
Engineering building 2

- 12-story building
 - Built in 2005
- >600 sensor stream dataset
- Lighting and HVAC for 231 rooms
- 10 weeks of summer 2011
 - Post-Fukushima Daiichi nuclear disaster

Analysis with EMD

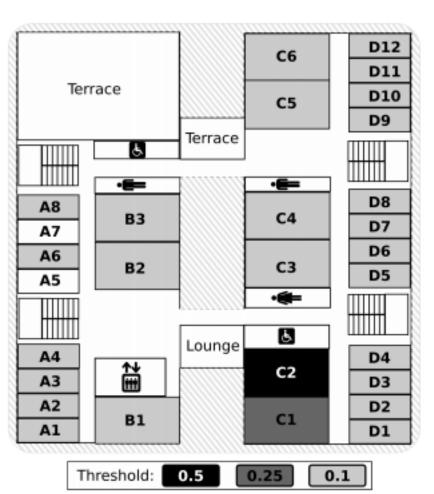




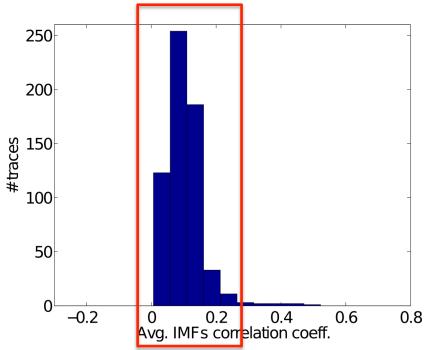


Broader validation

Run across 674 sensors



Geometric Verification Promising



Many sensors intrinsically unrelated, Few related

Data set

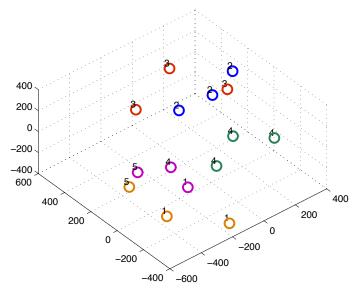
- 5 rooms, 3 sensors/room
- Sensor type: temperature, humidity, CO₂
- Over a one-month period



Results

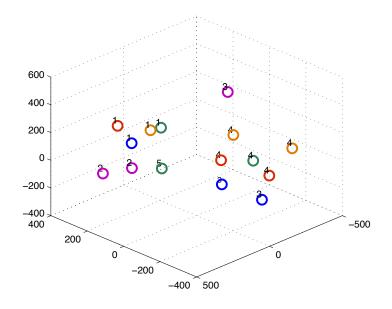
Clustering results (MDS + k-means)

-MDS: Multidimensional Scaling



On corrcoef from re-aggr.

12/15 correct = 80%



On corrcoef from raw traces

8/15 correct = 53.3%

Results

Clustering results (thresholding based)

	\boldsymbol{A}	B	C	D	\boldsymbol{E}	
$Sensor A_1$	1	0	0	0	0	\checkmark
A_{2}	1	0	0	0	0	\checkmark
A_3	1	0	0	0	0	\checkmark
B_1	0	1	0	0	0	\checkmark
B_2	0	1	0	0	0	\checkmark
$egin{array}{c} B_3 \ C_1 \ C_2 \ C_3 \end{array}$	0	1	0	0	0	\checkmark
C_1	0	0	1	0	0	\checkmark
C_2	0	0	1	0	0	\checkmark
C_3	0	0	1	0	0	\checkmark
D_1	0	0	0	1	0	\checkmark
$_D_2$	0	0	0	1	0	\checkmark
D_3	0	0	1	0	0	×
E_1	U	U	Ü	U	Ι	√
E_{2}	0	0	0	0	1	\checkmark
E_3	0	0	0	0	1	\checkmark

14/15 correct = 93.3%

Waste in Buildings

Lights/HVAC left on





Exhaust fan hard-wired Always "on"



Retro-commissioning

Table 2. Top faults causing energy inefficiencies in commercial buildings (top 13 of 100+)

	National Energy Waste	Electricity	
	(Quads,	equivalent	Cost
	primary/year)	(BkWh/year)	(\$billion/year)
Duct leakage	0.3	28.6	2.9
HVAC left on when space unoccupied	0.2	19.0	1.9
Lights left on when space unoccupied	0.18	17.1	1.7
Airflow not balanced	0.07	6.7	0.7
Improper refrigerant charge	0.07	6.7	0.7
Dampers not working properly	0.055	5.2	0.5
Insufficient evaporator airflow	0.035	3.3	0.3
Improper controls setup / commissioning	0.023	2.2	0.2
Control component failure or degradation	0.023	2.2	0.2
Software programming errors	0.012	1.1	0.1
Improper controls hardware installation	0.01	1.0	0.1
Air-cooled condenser fouling	0.008	0.8	0.1
Valve leakage	0.007	0.7	0.1
Total (central estimate)	1.0	94.6	9.6
Total (range)	0.34-1.8	32.4-171.4	3.3-17.3

Adapted from Roth et al. (2005) assuming 10,500 BTU/kWh, and \$0.10/kWh

Design Goals

1. Scalable

- Any building, any data source, everywhere
- Simple tuning

2. Range-based accuracy

Identify segment in data to focus on and search around

Tradeoff accuracy for scalability

Expanded w/Anomaly Detection

Learn typical correlations patterns

Discover changes

Return associated change points in trace

Experimental Setting

UC Berkeley

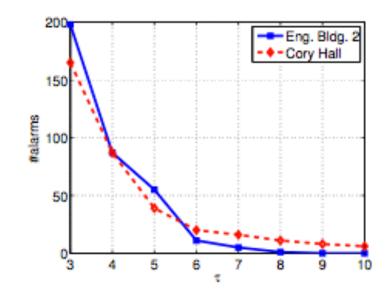


Cory Hall

- 5-story building
 - Built in 1950
- 70 sensors stream dataset
- Lighting, HVAC, elevator, electrical receptacles and panels, etc.
- 15 different HVAC systems
- 8 weeks in Spring 2011

Results

- High power usage
 - Alarms corresponding to electricity waste
- Lower power usage
 - Alarms representing abnormal low electricity consumption
- Punctual
 - Short increase/decrease in electricity consumption
- Missing data
 - Possible sensor failure
- Other
 - unknown

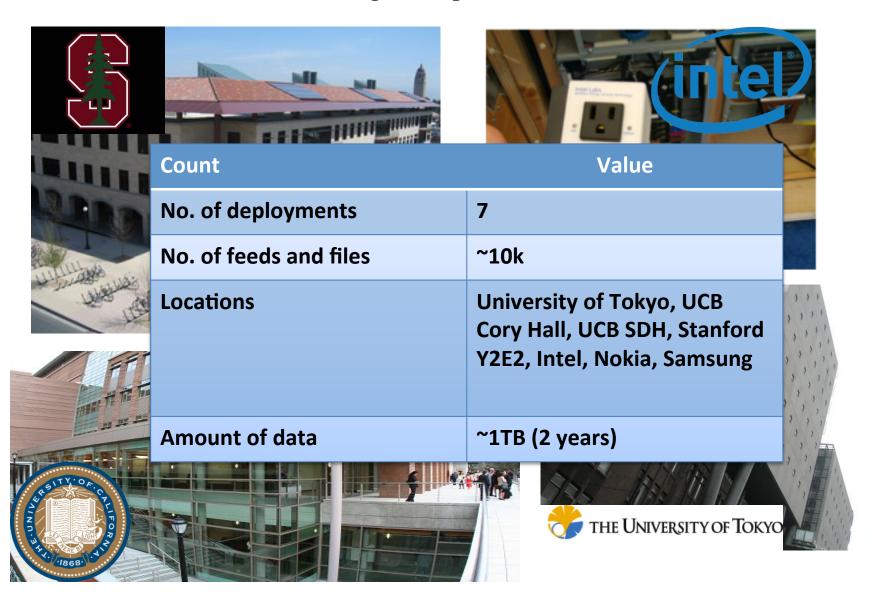


	High	Low	Punc.	Missing	Other
Eng. Bldg 2	9 (5)	6 (5)	1(1)	36 (1)	3 (3)
Cory Hall	25 (7)	7(3)	4 (4)	0 (0)	3 (3)

Alarms in Eng. Bldg 2 @Todai

Alarms in Cory Hall

Deployments



Future Work

- Automate verification further
 - Examine value and spatial verification
- More sophisticated control applications
 - MPC applications on building
- Version control over StreamFS
 - Provenance checking for streams, jobs, and control settings

Conclusion

- Device drivers
 - Standard Interfaces

- File system
 - common namespace
- Verification
 - statistical

