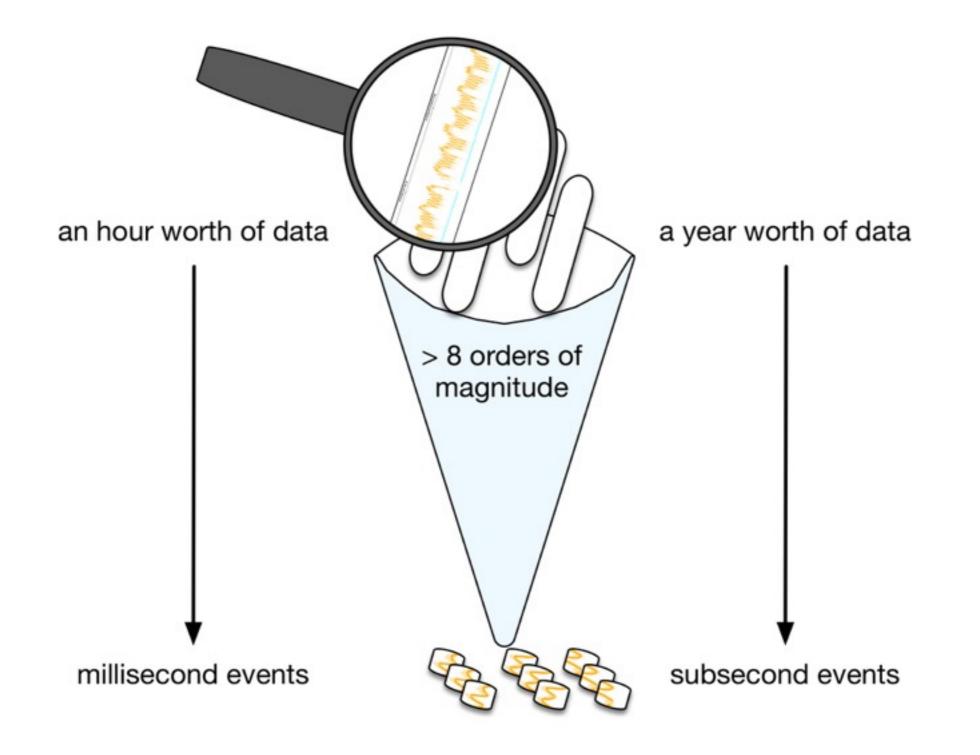
Efficient Analytics with BTrDB: From Grids to Buildings

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SDB Winter Retreat





Introduction

High-precision high-sample-rate data from distributed high fidelity
high bandwidth sensors

* timescale and size

- Finding anomalies in these systems is the holy grail
 - failing to identify and react to critical events in a timely manner may cost millions of dollars
- Energy data analytics (both real-time and historical) is critical yet computationally expensive
 - * the ability to detect, analyze, and control with a limited time budget

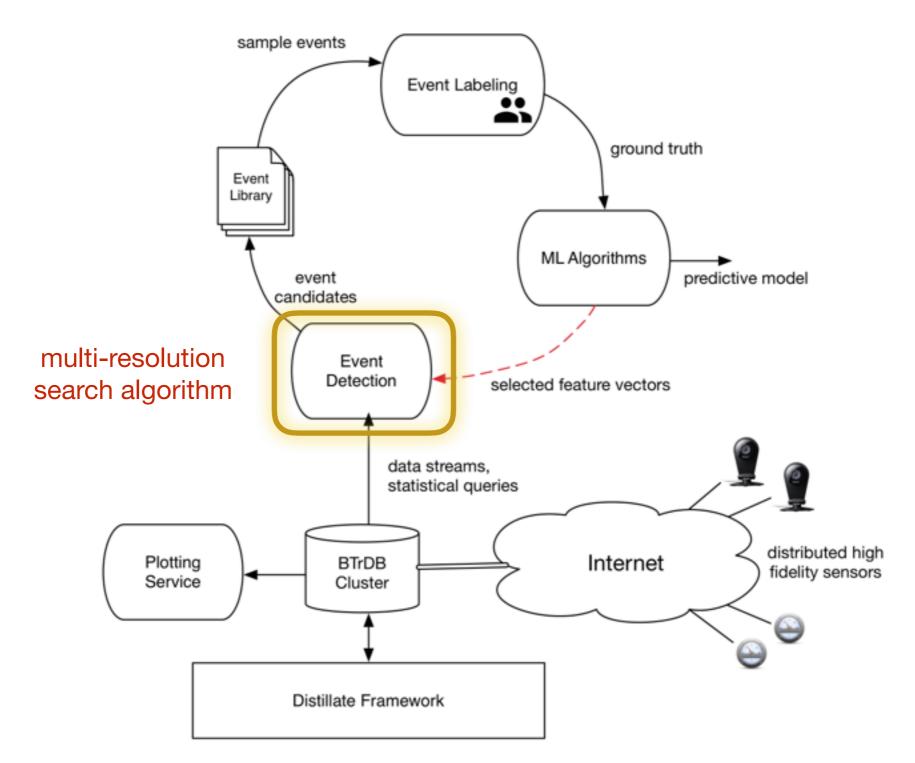
Objectives

• Detect: identify rare events

 * using an efficient search algorithm that is logarithmic in the size of the data set and linear in the number of events that are found

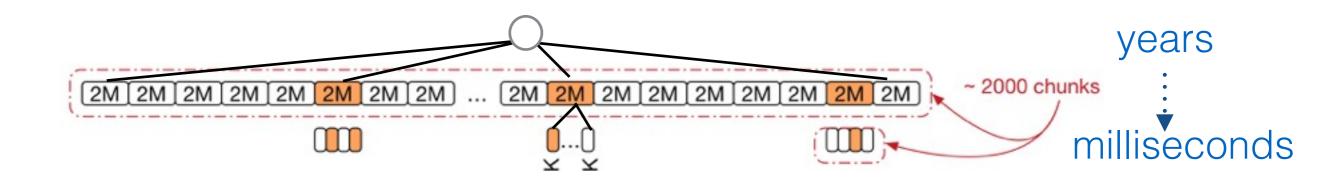
- Analyze: run compute intensive tasks on smaller chunks of data
- Control: take corrective/preventive actions (in real-time applications)

System Architecture



Statistical Summaries

 statistical summaries (max, min, average, and count) are stored at different temporal resolutions



Example Query

Find 5-second intervals that contain at least one value greater than a threshold

Example Query

Find 5-second intervals that contain a value greater than a threshold

- Query **max** at the given temporal resolution
- Dive down if max_{resolution} > threshold
- Repeat for the next temporal resolution until the desired resolution is reached

Multi-Resolution Search

- Query statistical summaries of data at a given temporal resolution
- Compare a function of these statistical summaries against a threshold
- Dive down if the condition is satisfied
- Query raw data when the desired resolution is reached and run your algorithm on a small chunk of data

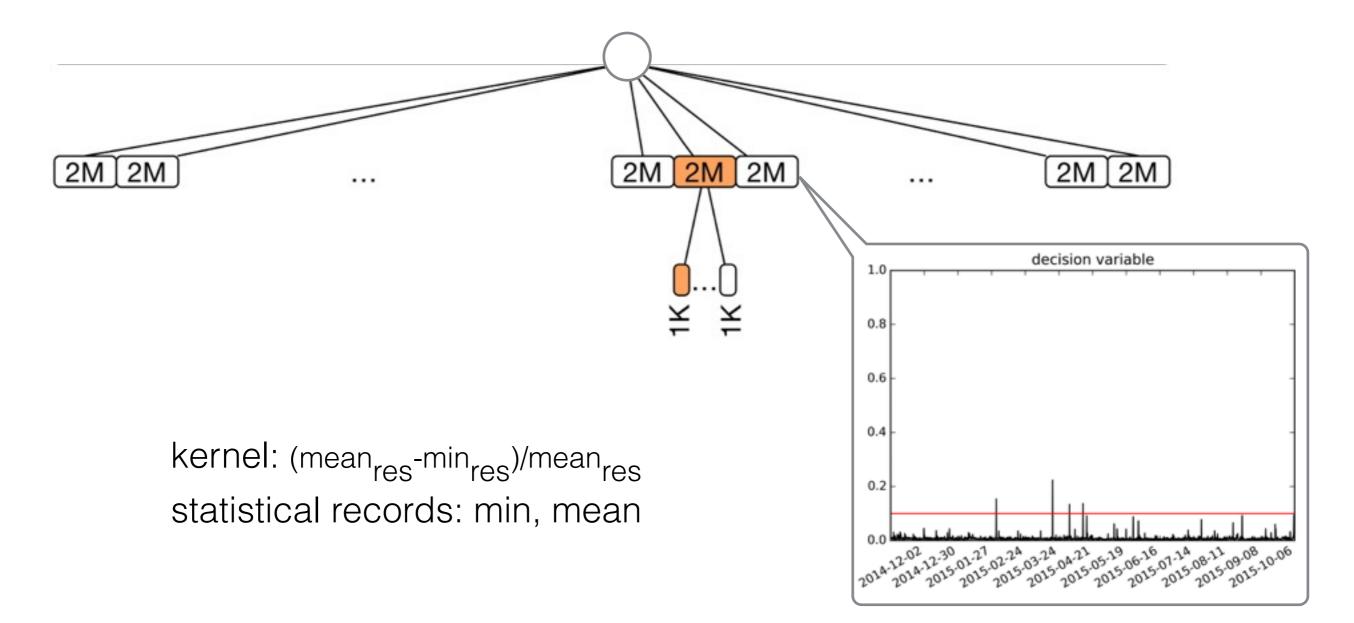
Grid Energy Analytics

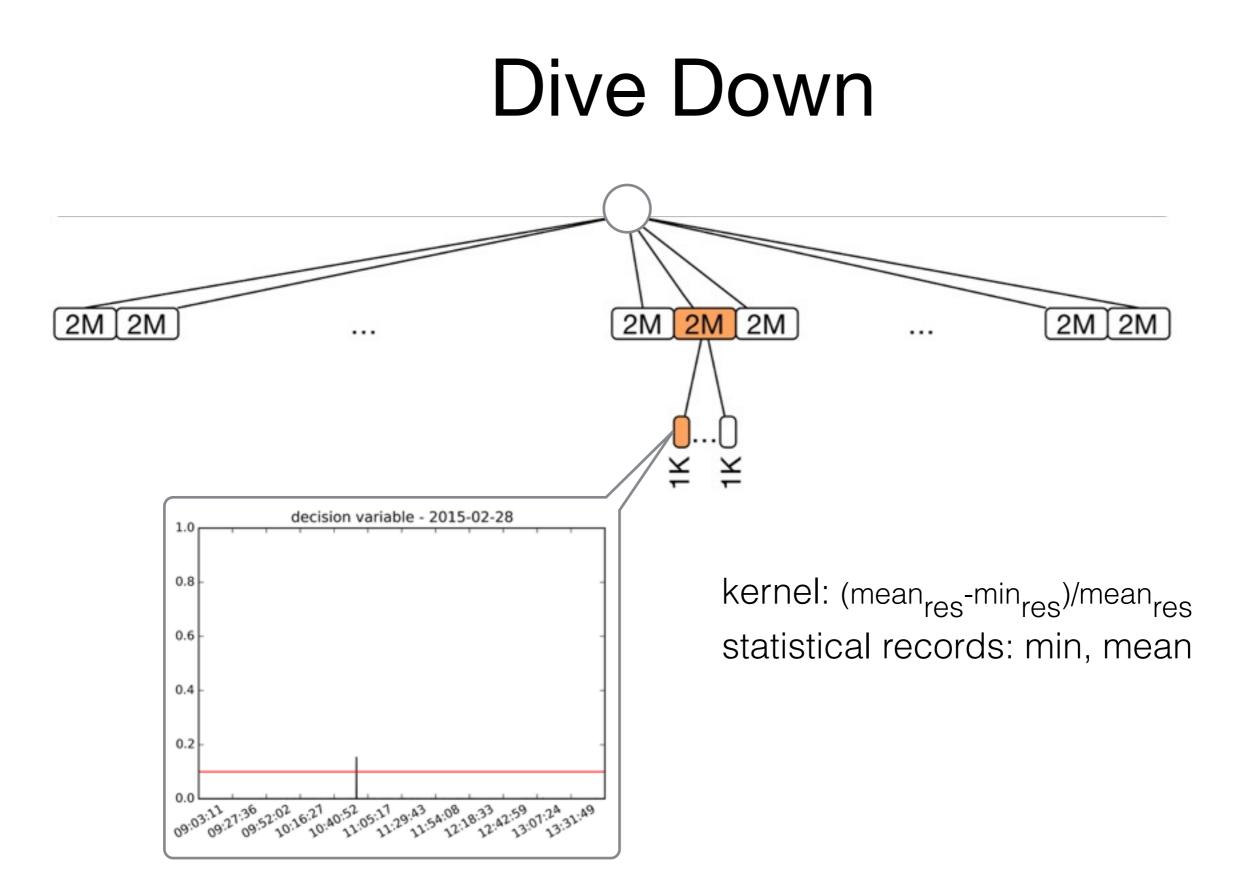


Critical Events

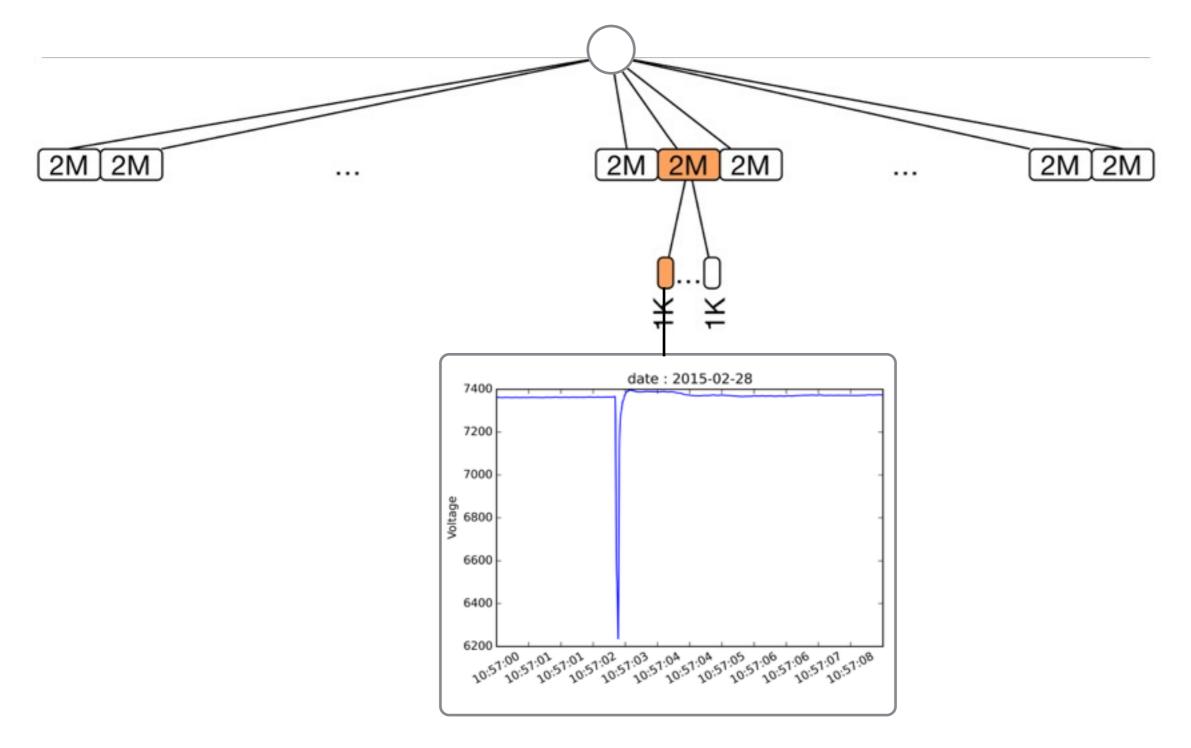
- voltage sags
 - voltage magnitude (raw stream)
- tap changing events
 - angle difference (distillate)
- reverse flows
 - real power or displacement power factor (distillate)
- switching events

Case Study: Voltage Sag Detection

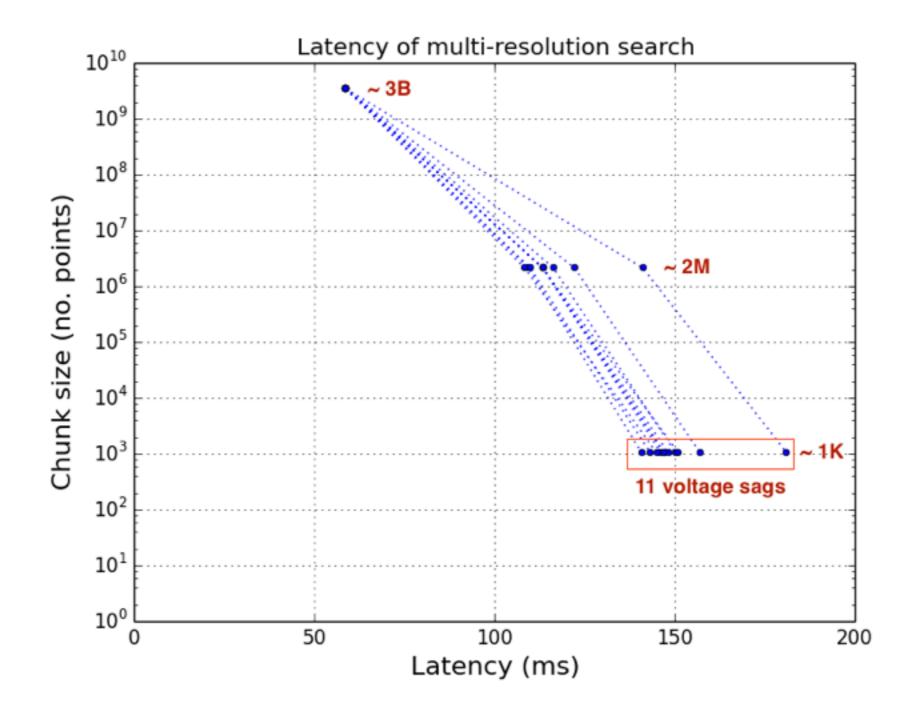




Query Raw Data



Evaluation



Example Result

	no. events (0.05)	runtime (ms)	no. events (0.1)	runtime (ms)	no. events (0.15)	runtime (ms)	no. events (0.2)	runtime (ms)	da	ays
/clean/GP_BUS1/L1MAG	9	431.77	4	237.13	0	76.78	0	88.41	13	35
/clean/GP_BUS1/L2MAG	10	394.39	4	226.85	1	115.30	0	70.55	13	35
/clean/GP_BUS1/L3MAG	5	309.07	2	163.25	1	118.95	0	77.08	13	35
/clean/switch_a6/L1MAG	14	666.59	6	273.01	3	194.95	1	132.75	33	30
/clean/switch_a6/L2MAG	21	947.24	11	523.78	4	235.44	3	190.83	33	30
/clean/switch_a6/L3MAG	11	608.94	4	318.44	2	213.57	0	90.06	33	30
/clean/RPU/CE_CERT_Bld_1200/L1MAG	8	312.53	2	68.41	1	64.93	1	66.55	86	3
/clean/RPU/CE_CERT_Bld_1200/L2MAG	12	379.19	4	163.71	3	119.51	2	112.95	86	5
/clean/RPU/CE_CERT_Bld_1200/L3MAG	12	627.72	4	228.18	2	111.41	2	133.00	86	5

10% drop

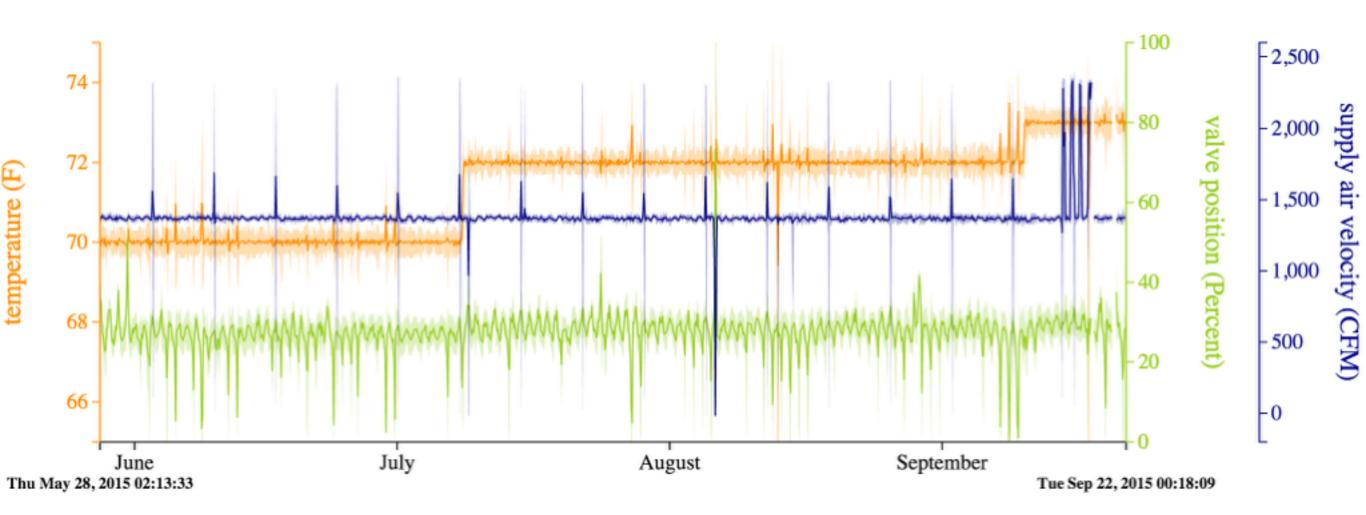
logarithmic in the size of the data set and linear in the number of events that are found

Building Energy Analytics

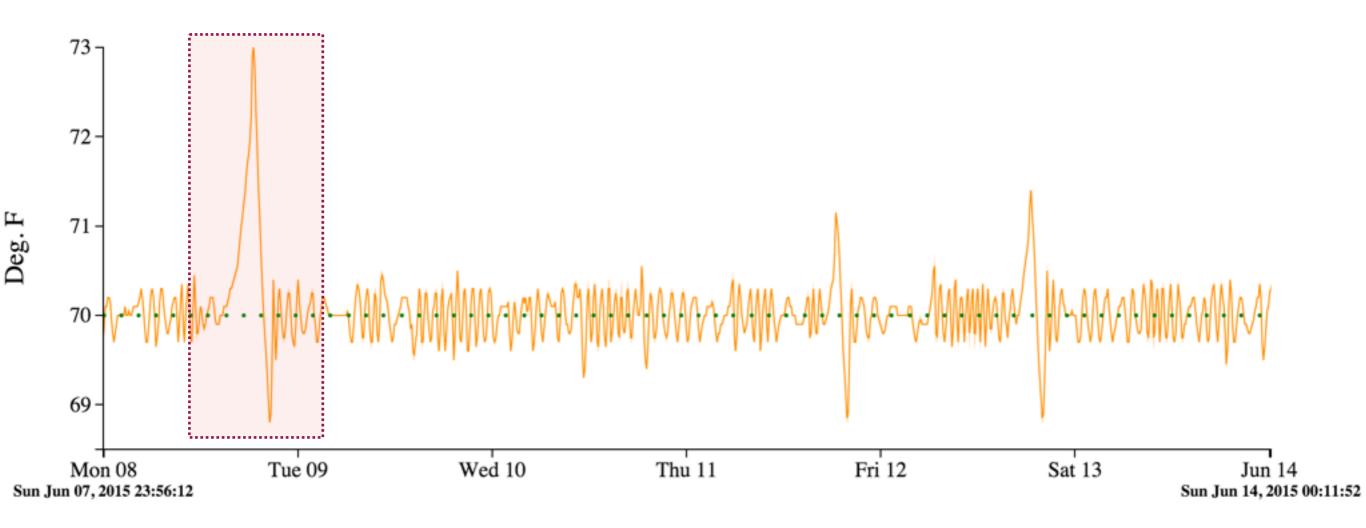
Identified Events

- HVAC anomalies
- Water leakage events
- Correlated use of large reactive loads

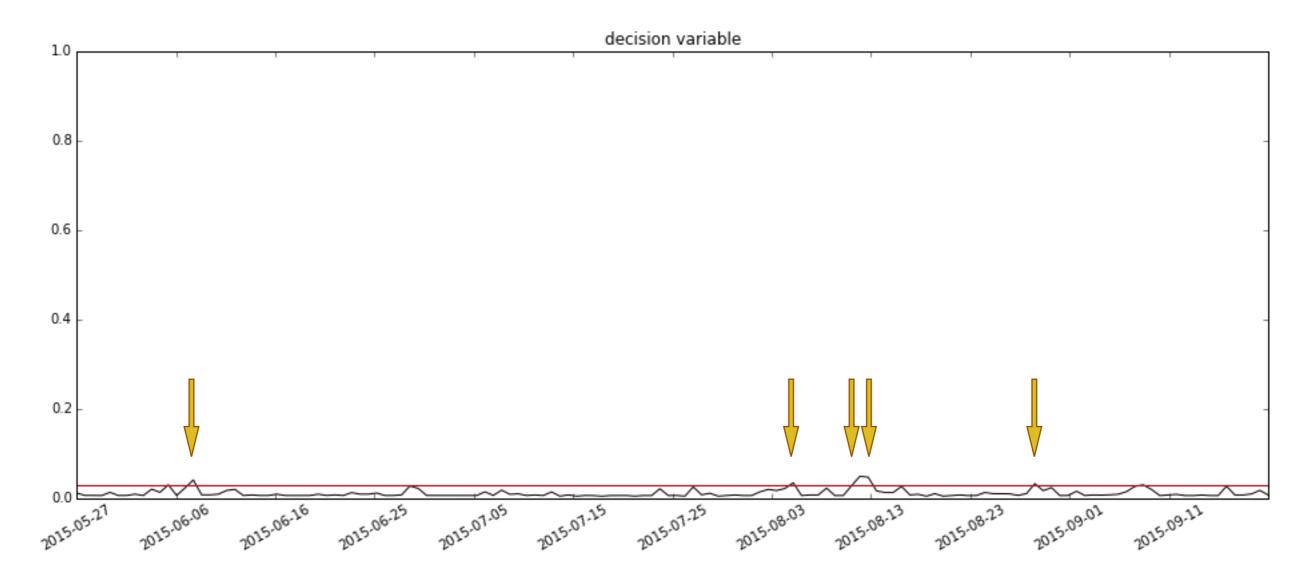
Anomalies in Buildings





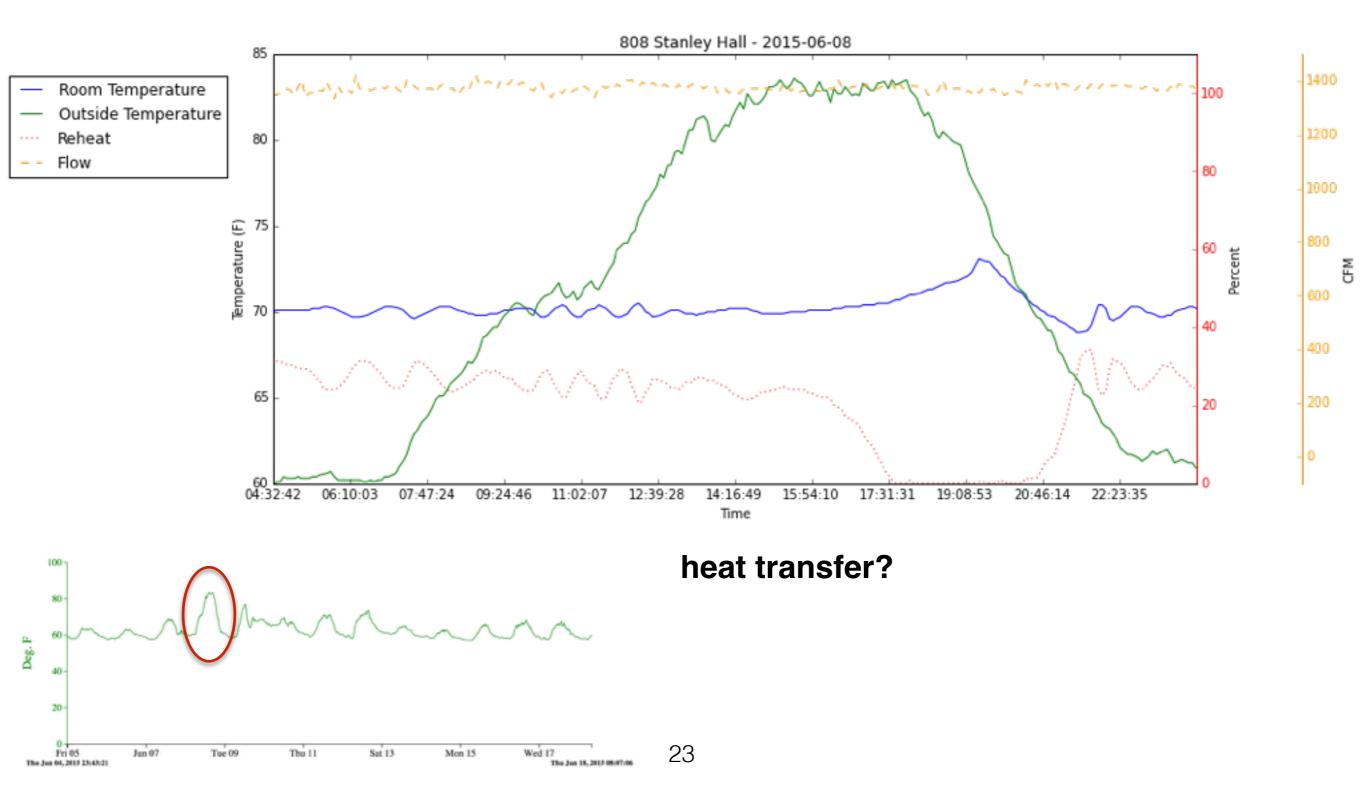


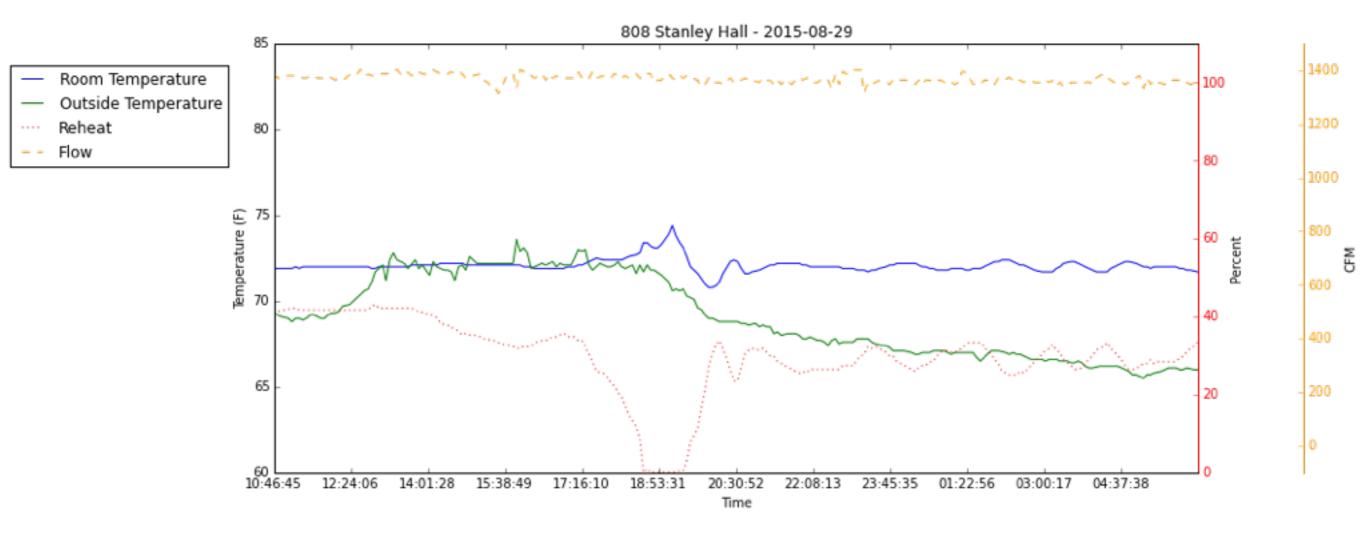
find intervals in which zone temperature increases by more than X%



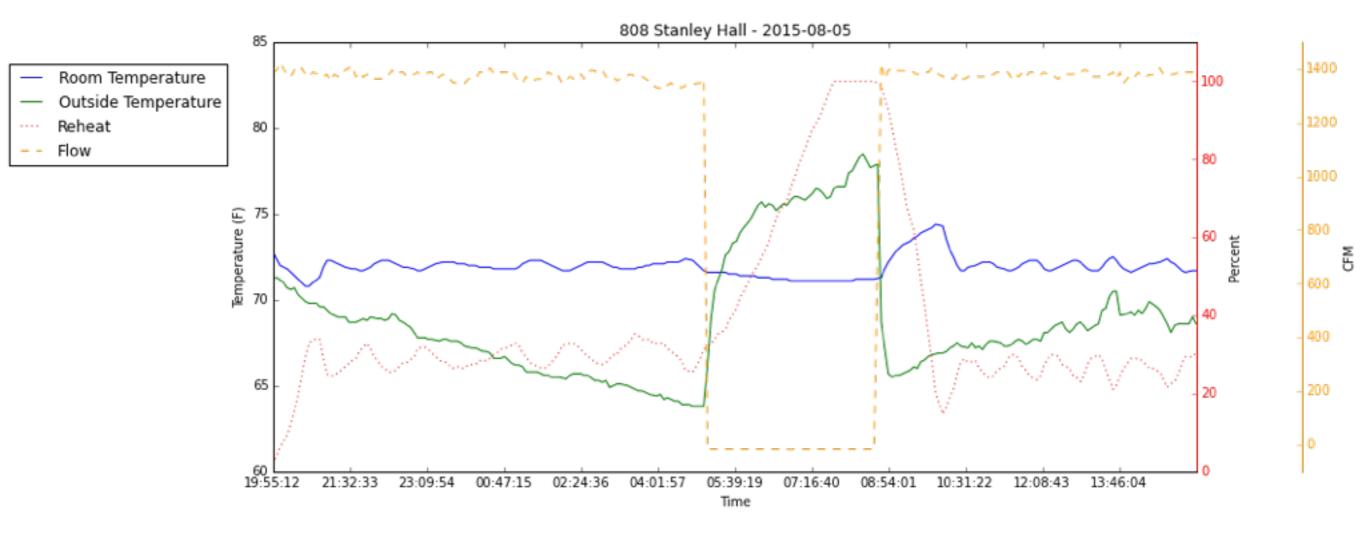
find intervals in which zone temperature increases by more than 3%

kernel: (max_{res}-mean_{res})/mean_{res} statistical records: max, mean

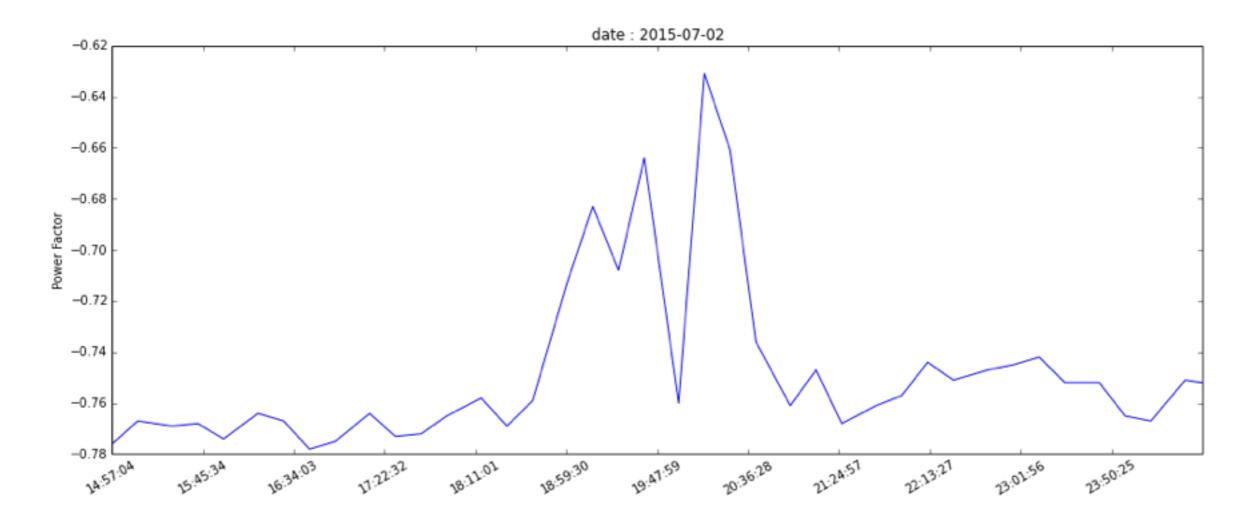




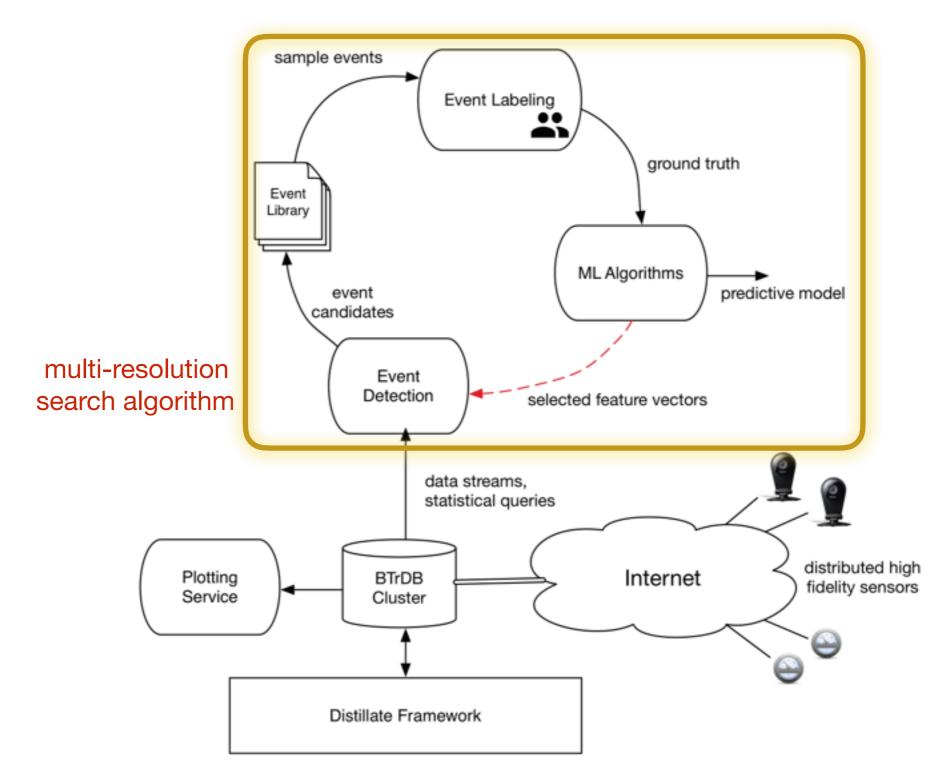
heat load due to occupancy, equipment load, direct sunlight?



Case Study: Power Flow Reduction Detector



kernel: max_{res} statistical records: max



Takeaways

- Complexity of the search algorithm is O(nLog(L))
- Writing a kernel function can be quite challenging in some cases
- Machine learning techniques can be used to develop sophisticated detectors