

Being SMART The Role of Timely Analytics

Krithi Ramamritham
IIT Bombay

Center
for
Urban Sc. & Engg.

Smart
Energy
Informatics
Lab

Towards Sustainable Energy

Reduce

- Consumption
- Peaks
- Reliance on unsustainable energy sources

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Reuse resources

Towards Sustainable Energy

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Reuse resources

Recycle resources

Towards Sustainable Energy

Reduce

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- Peaks
- Reliance on unsustainable energy sources

Reuse resources

Recycle resources

The 3 R's

Towards Sustainable Energy

Be SMART

Reduce

- Consumption
- Peaks
- Reliance on unsustainable energy sources

Reuse resources

Recycle resources

The 3 R's

Being Smart

*clever, bright, intelligent, able,
shrewd, astute, perceptive, savvy*

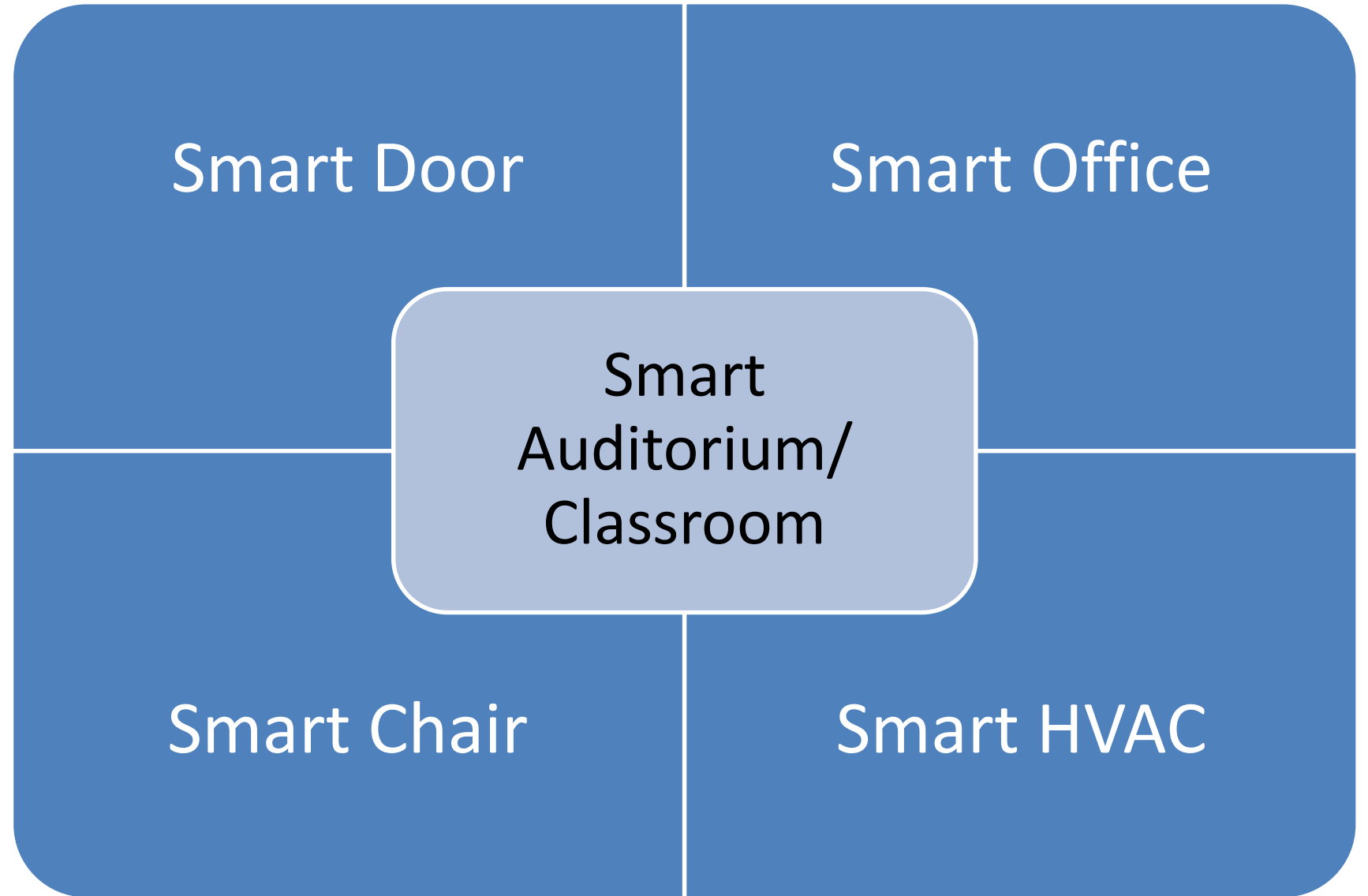
brisk, energetic, vigorous

fast, rapid, swift, lively

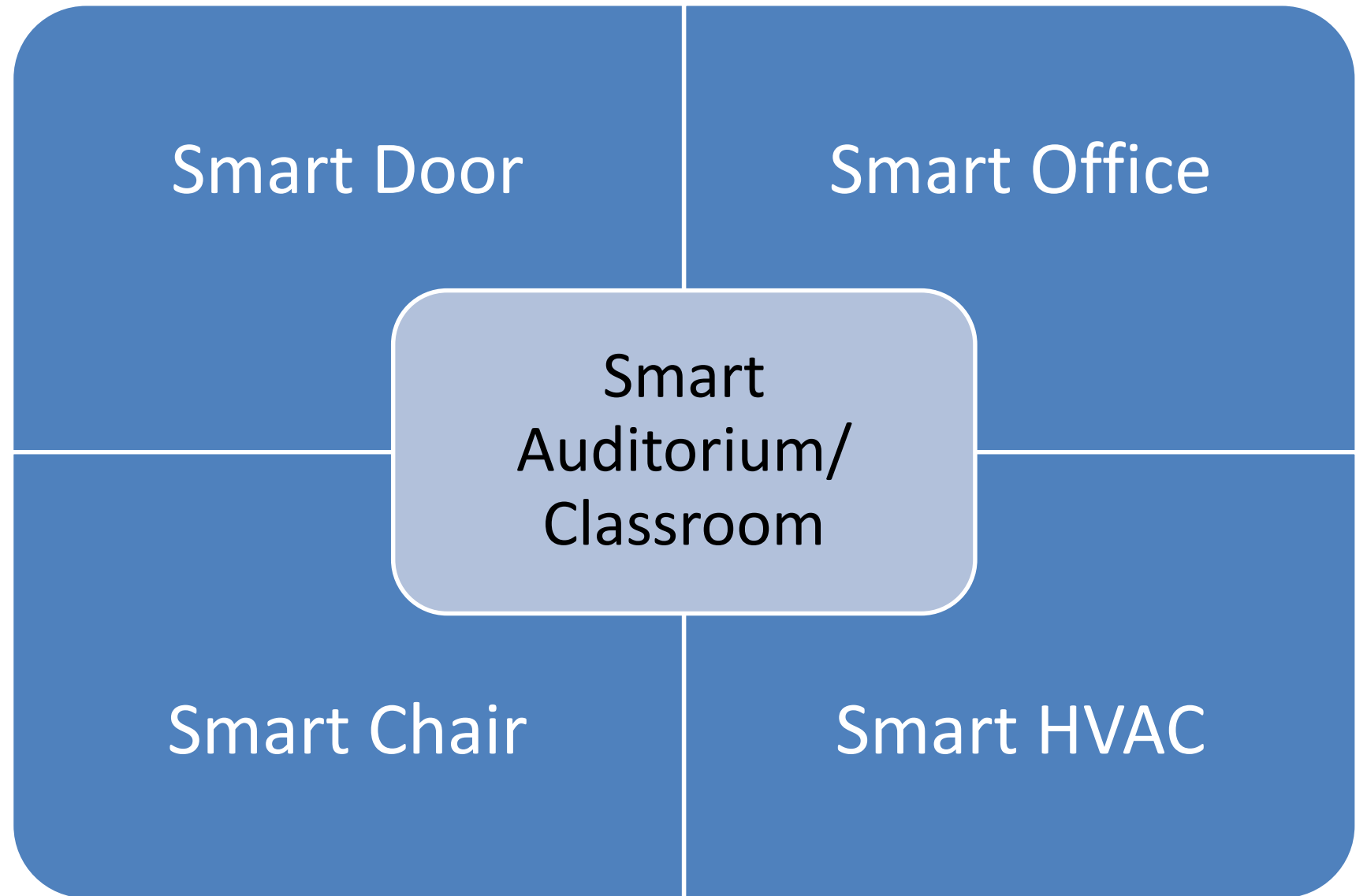
*sharp-witted, quick-witted
quick on the uptake*

SMART Energy

Smart buildings
Smart campuses
Smart grids
Smart cities



Smart buildings
Smart campuses
Smart grids
Smart cities



Spotlight:
Smart
Classroom



The Smart Classroom Complex

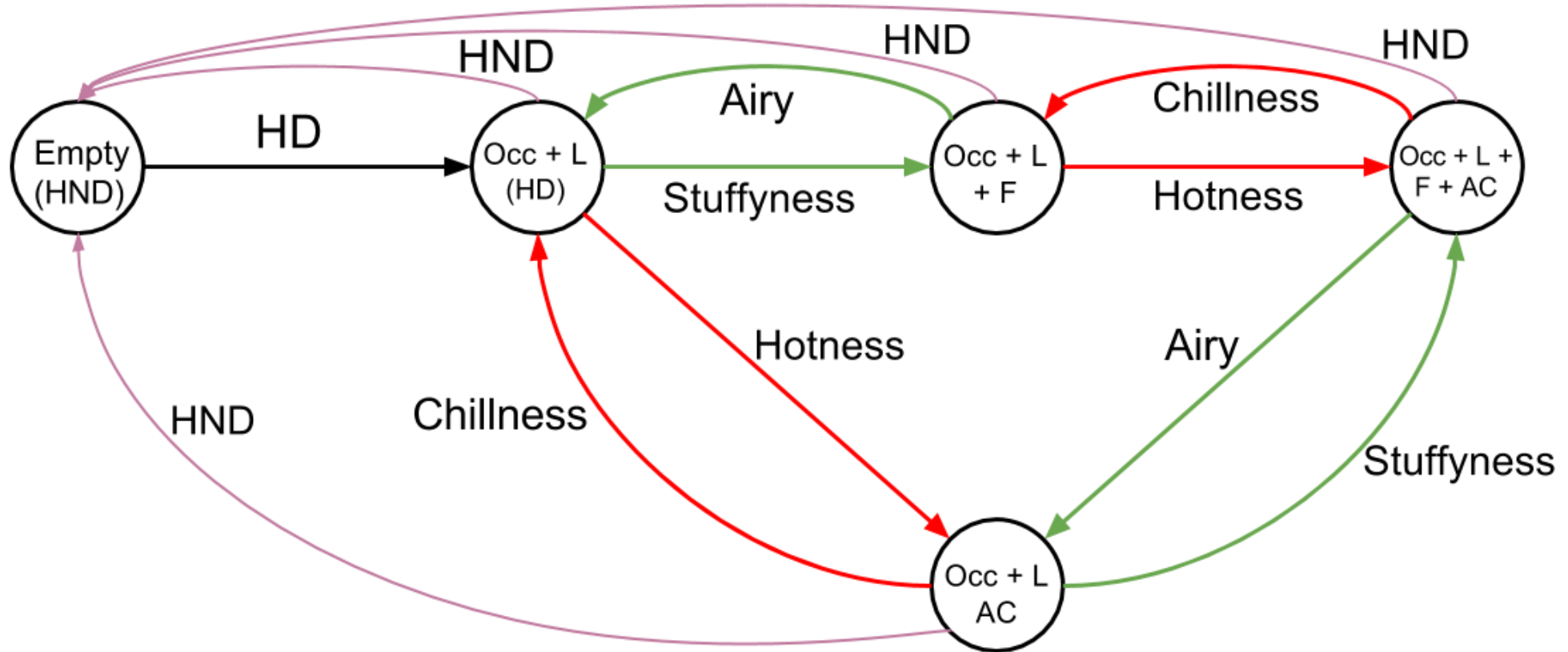
Spotlight:
Smart
Classroom



The Smart Classroom Complex

Energy Consumed ONLY IF Occupied

The Smart Classroom Complex



Occ - Occupied

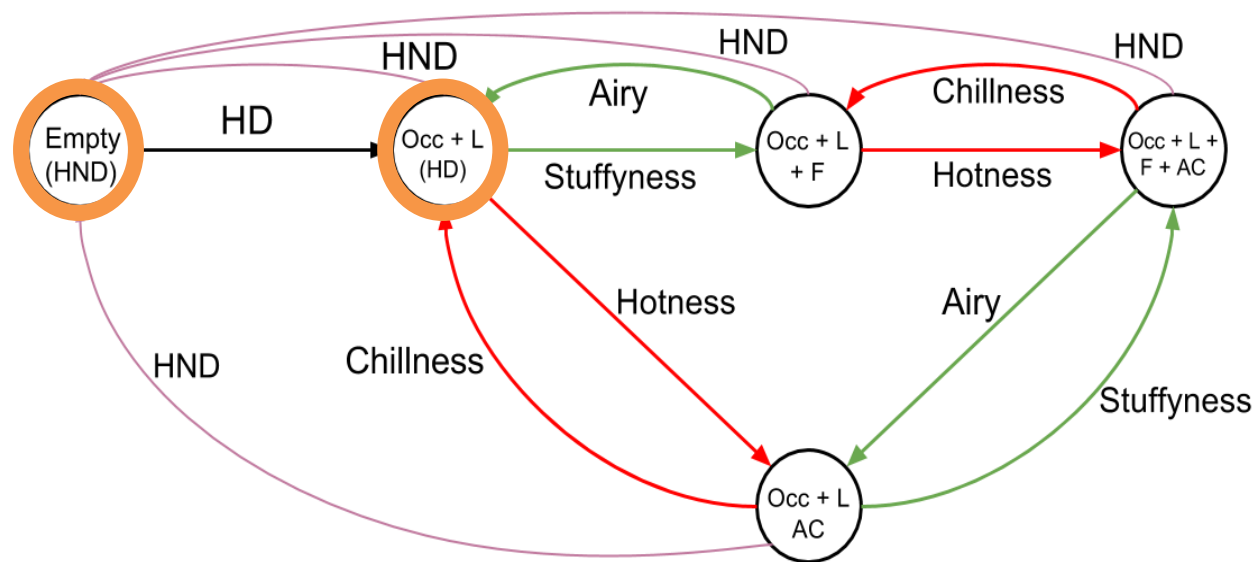
HD - Human Detected

HND - Human Not Detected

L- Lights

F - Fans

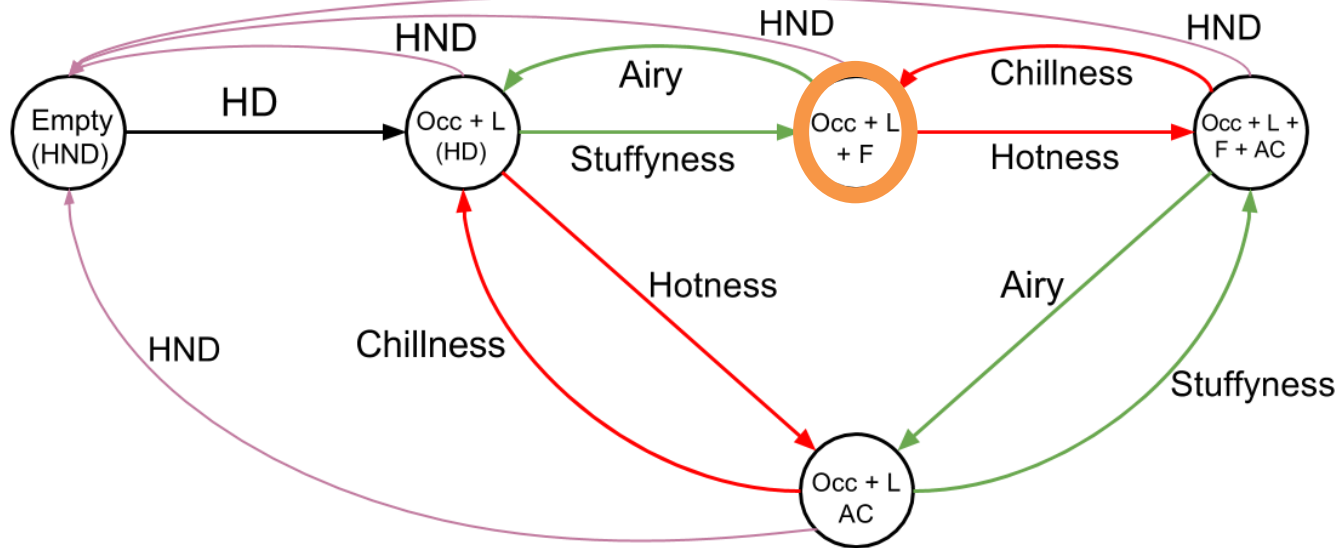
AC - Air Conditioners



Sense Temperature & Humidity

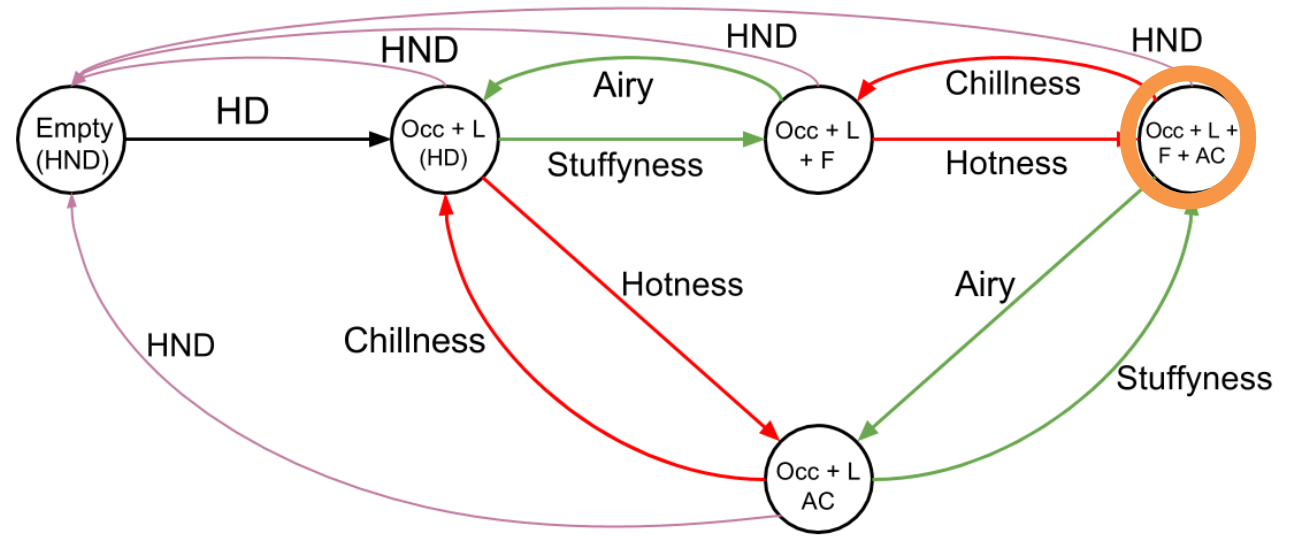
Analyze Temperature & Humidity > Threshold of Fan

Respond Turn ON Fans





Sense Temperature
Analyze Temperature > Upper Threshold of AC
Respond Turn ON AC

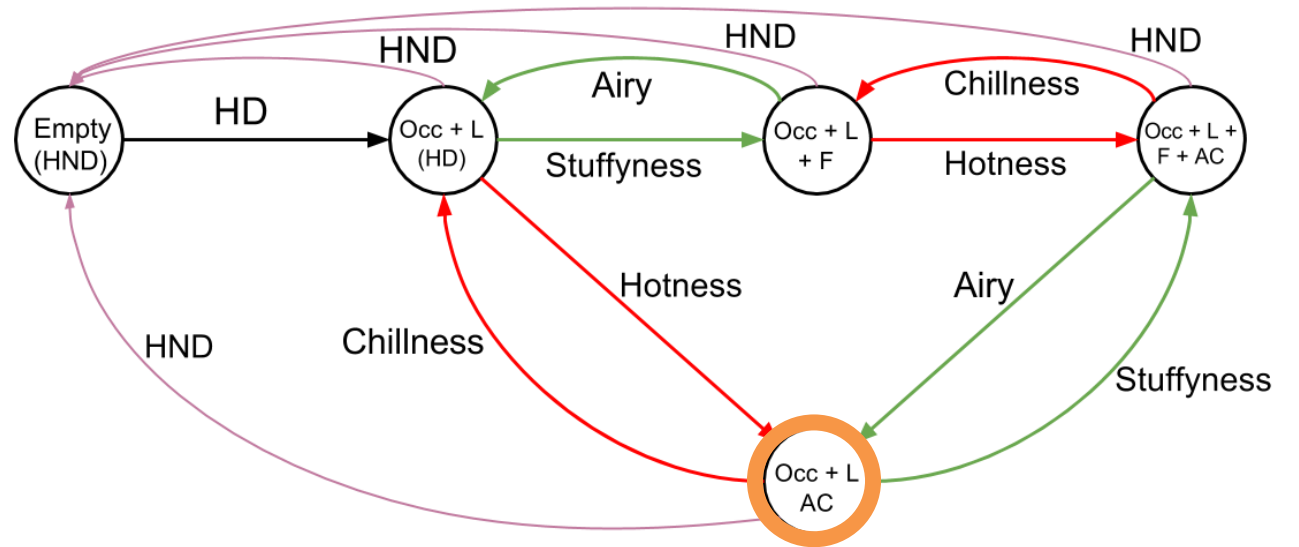


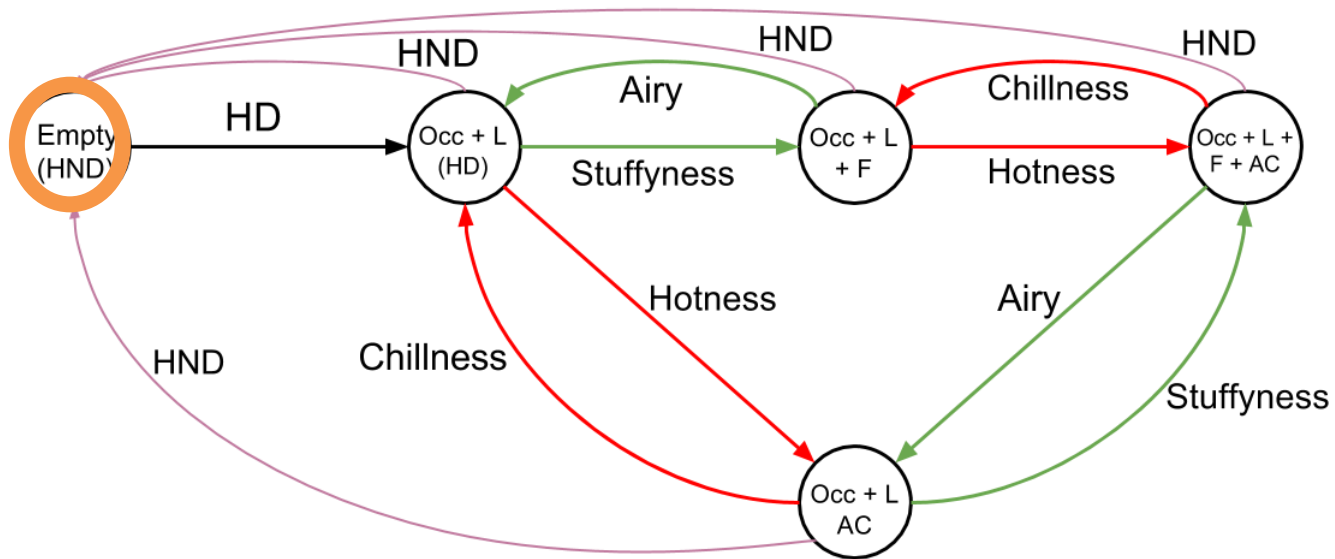


Sense Temperature & Humidity

Analyze Temperature > Upper Threshold of AC
 Temperature & Humidity < Threshold of Fan

Respond Turn OFF Fans

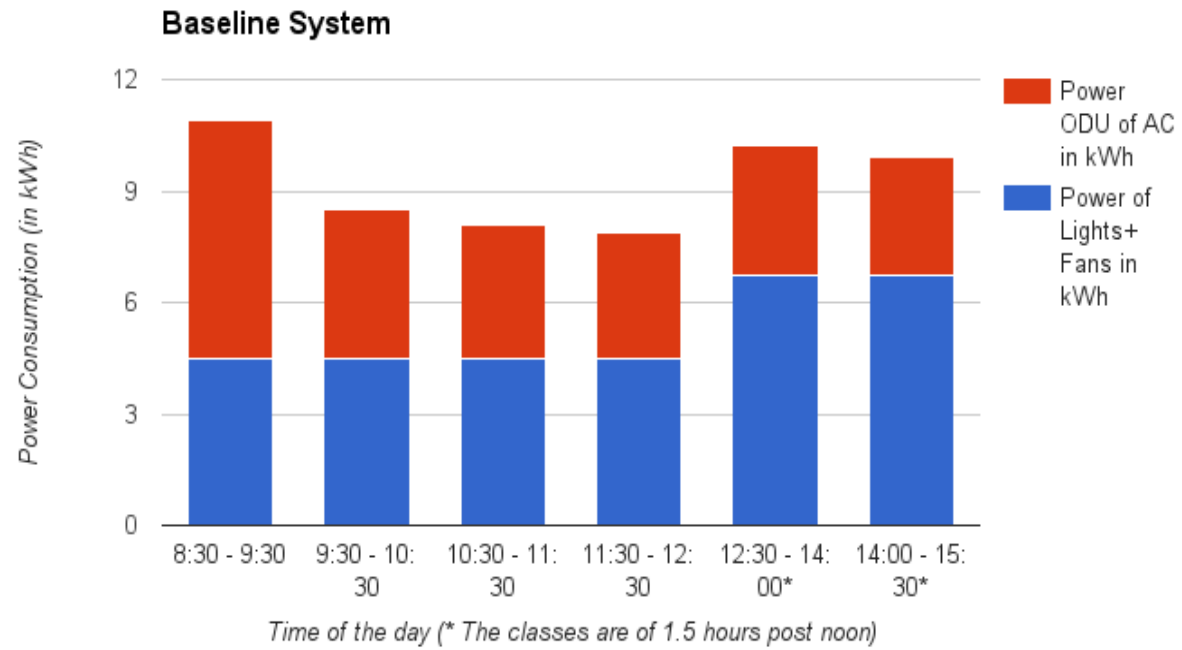




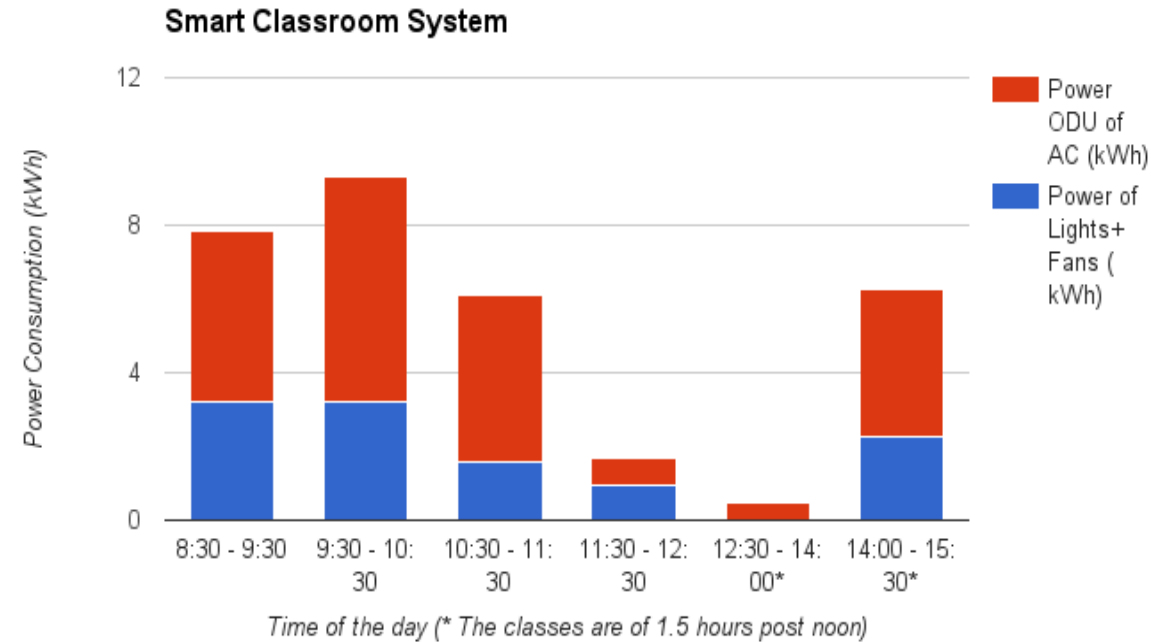
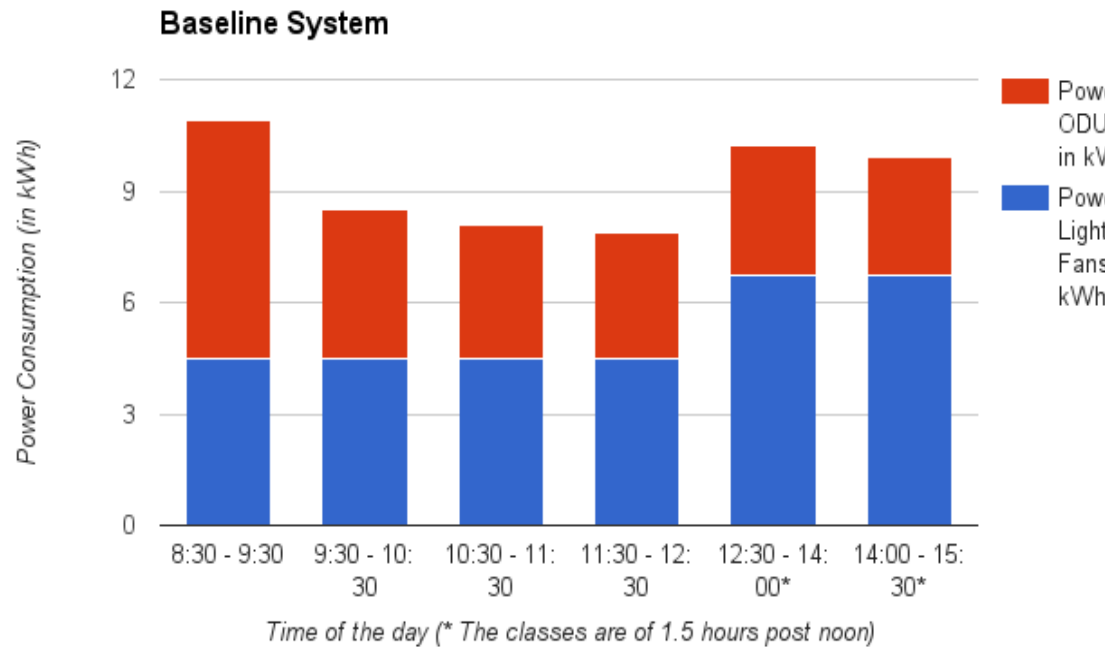
**Zero
Occupancy**

**Zero
Consumption**

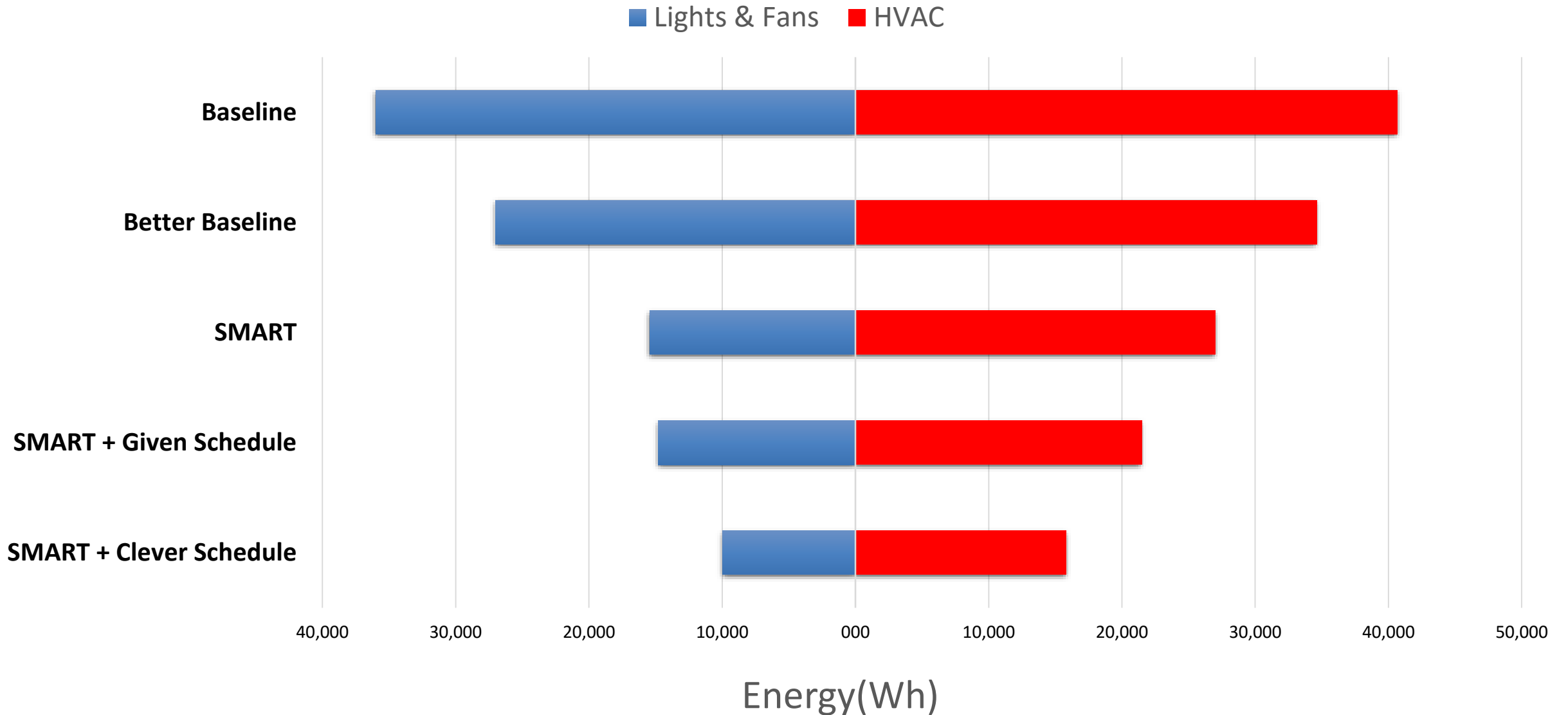
Energy savings



Energy savings

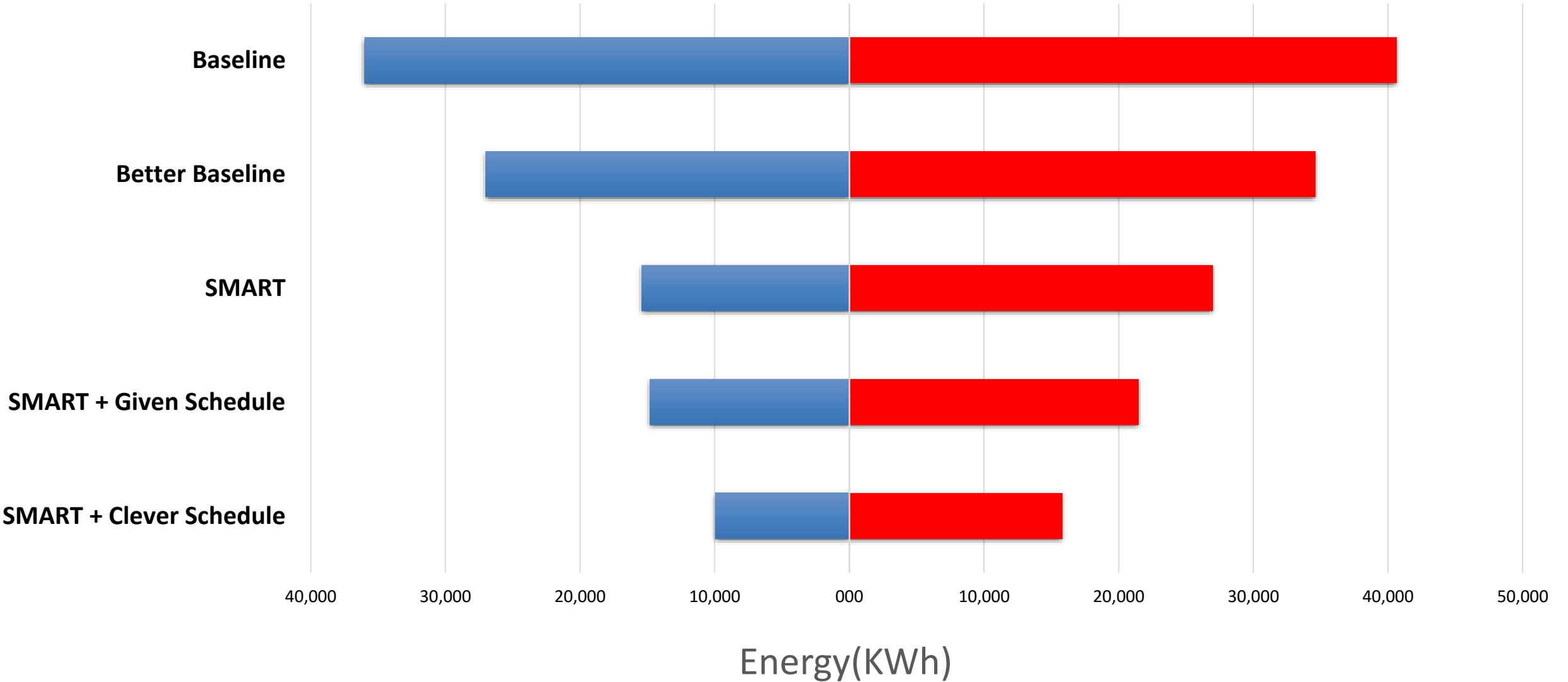


Energy Saavings from different Variations of Smart Classroom Complex



Energy Savings: 58%
Payback period:
6 academic months

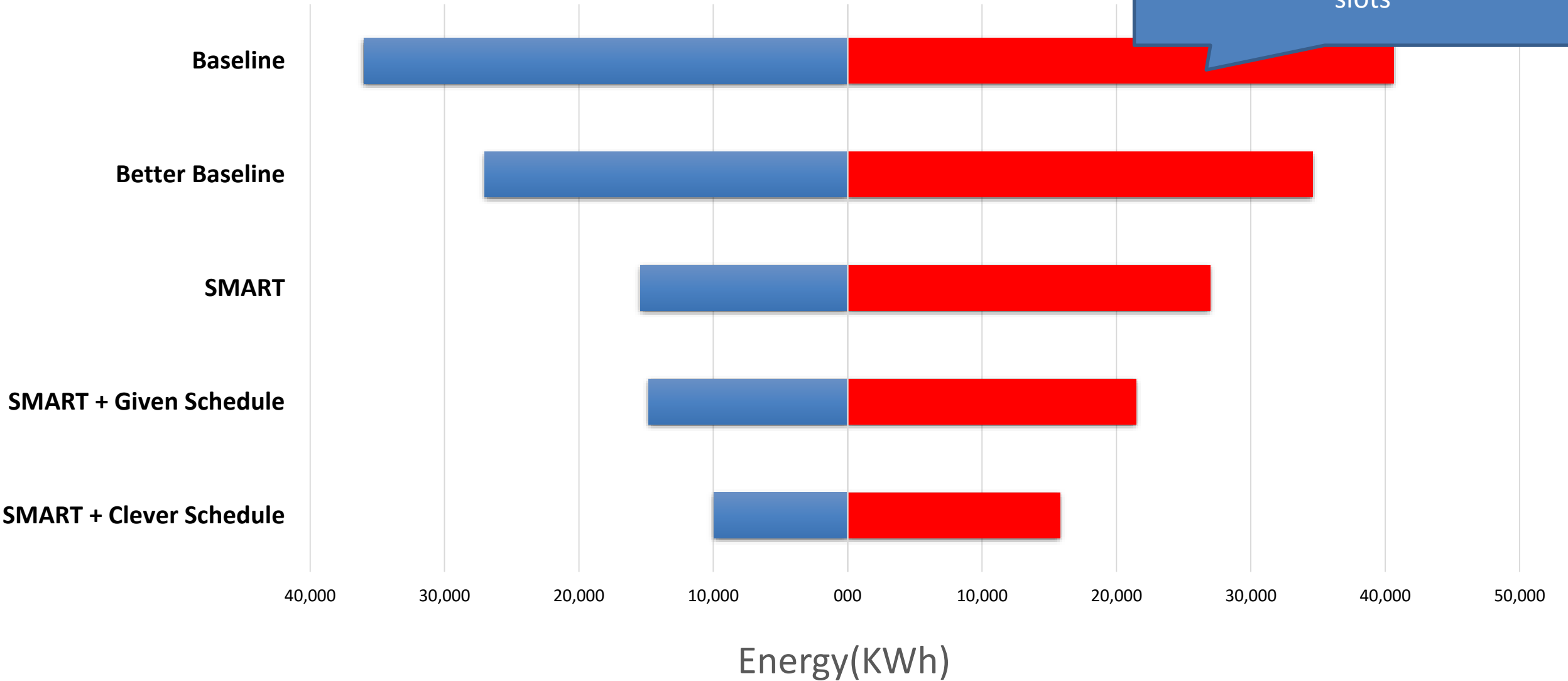
Lights & Fans HVAC



Energy Savings: 58%
Payback period:
6 academic months

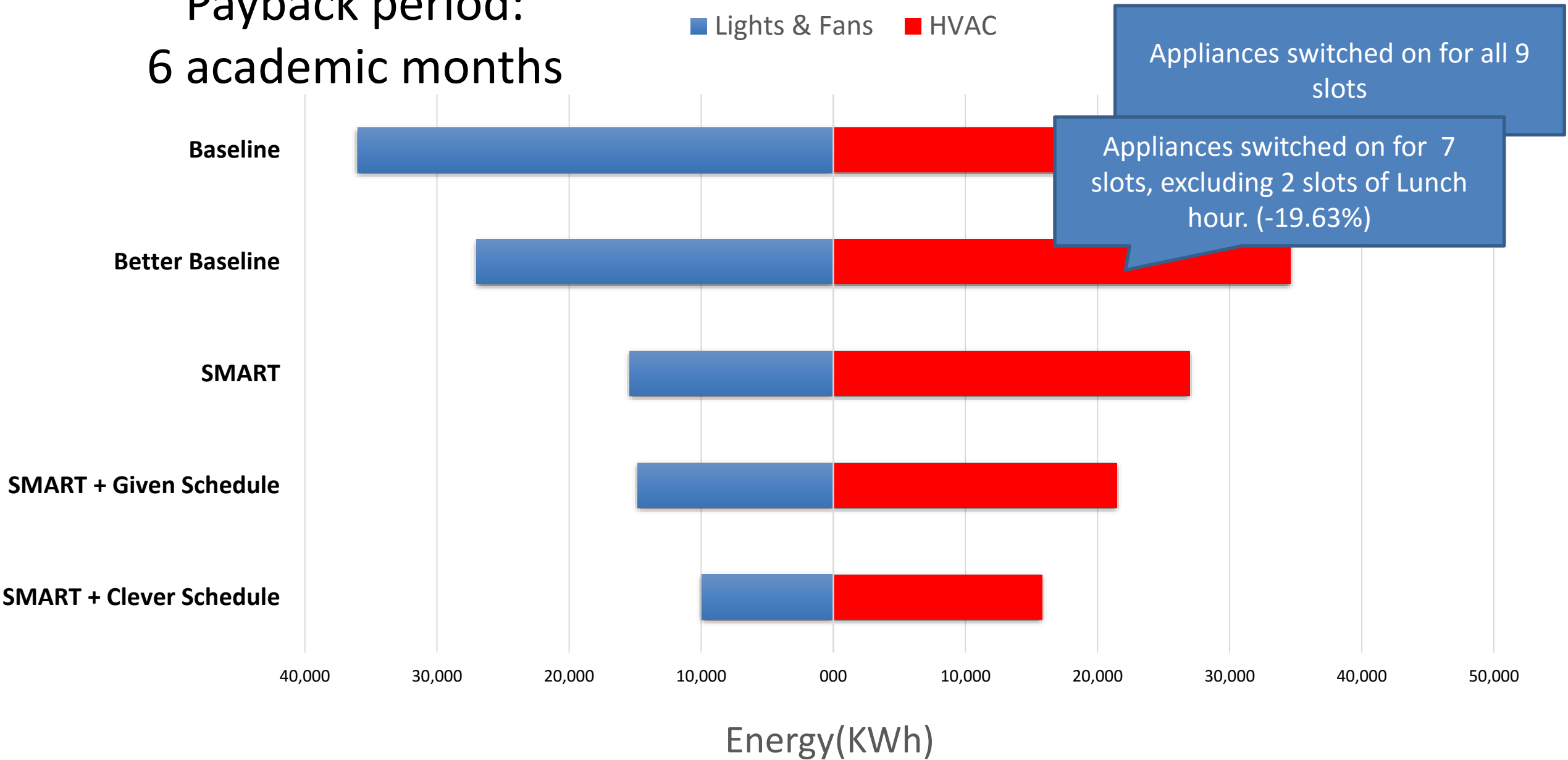
■ Lights & Fans ■ HVAC

Appliances switched on for all 9 slots



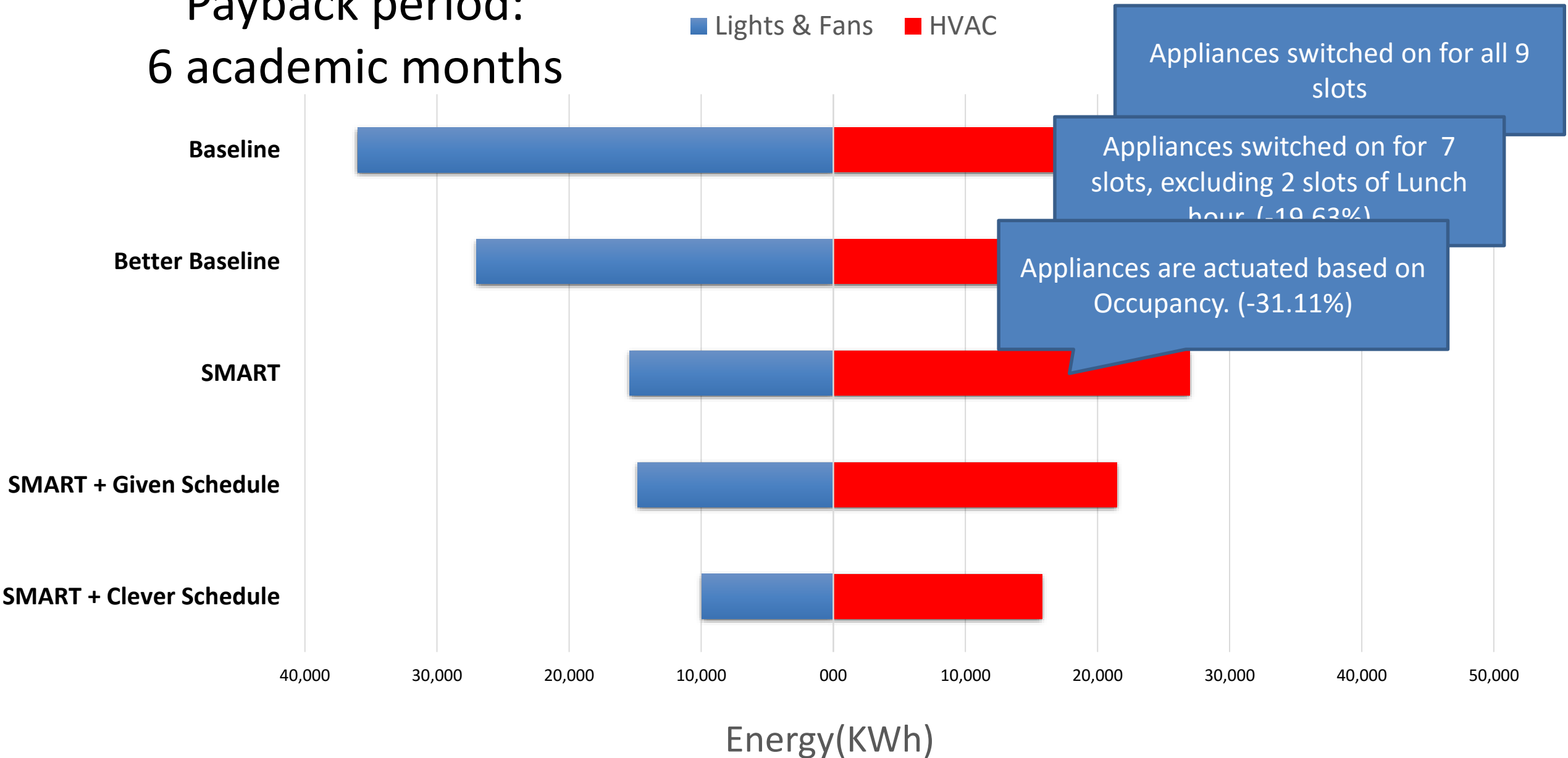
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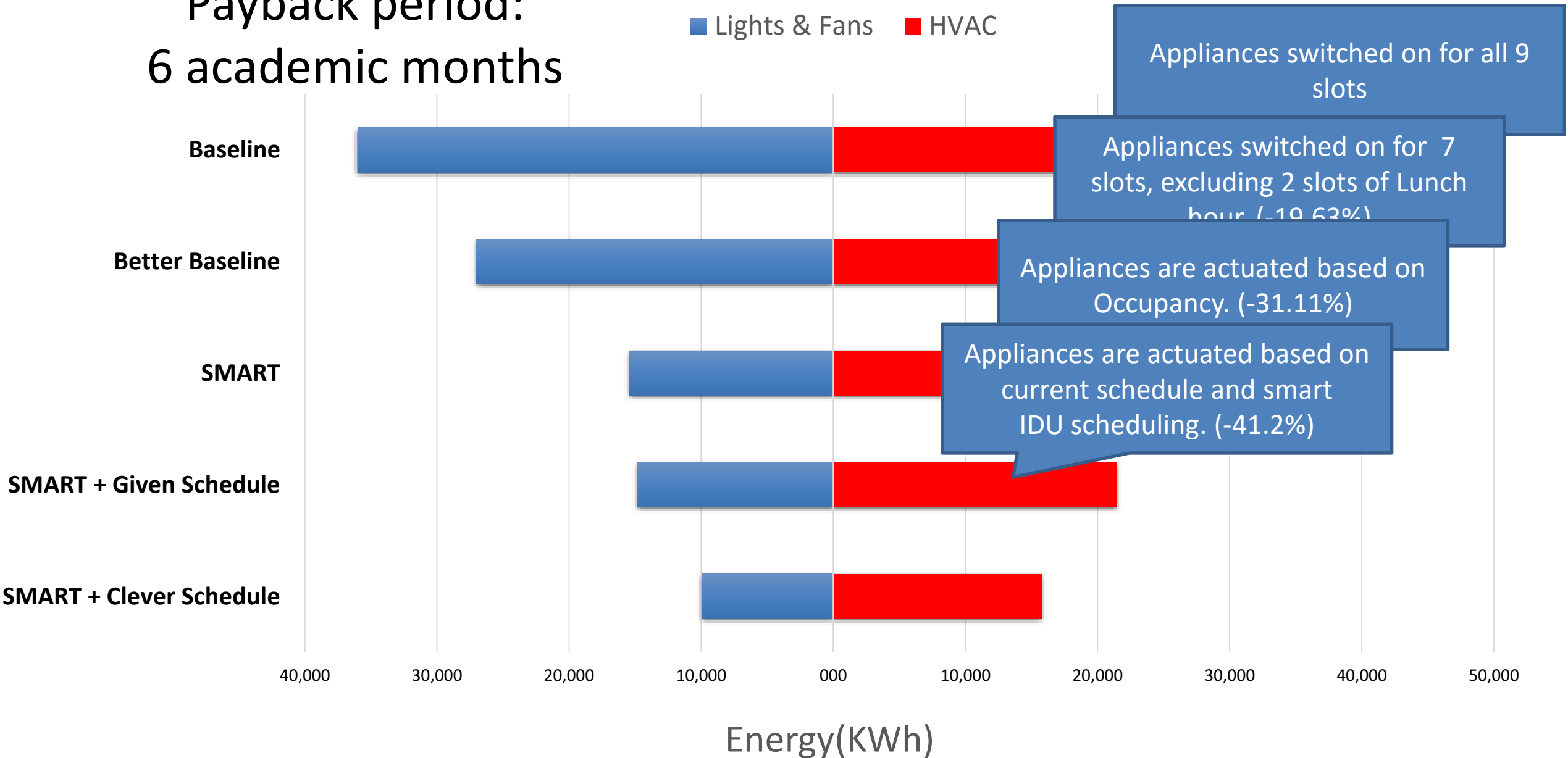
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Payback period:
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■ Lights & Fans ■ HVAC



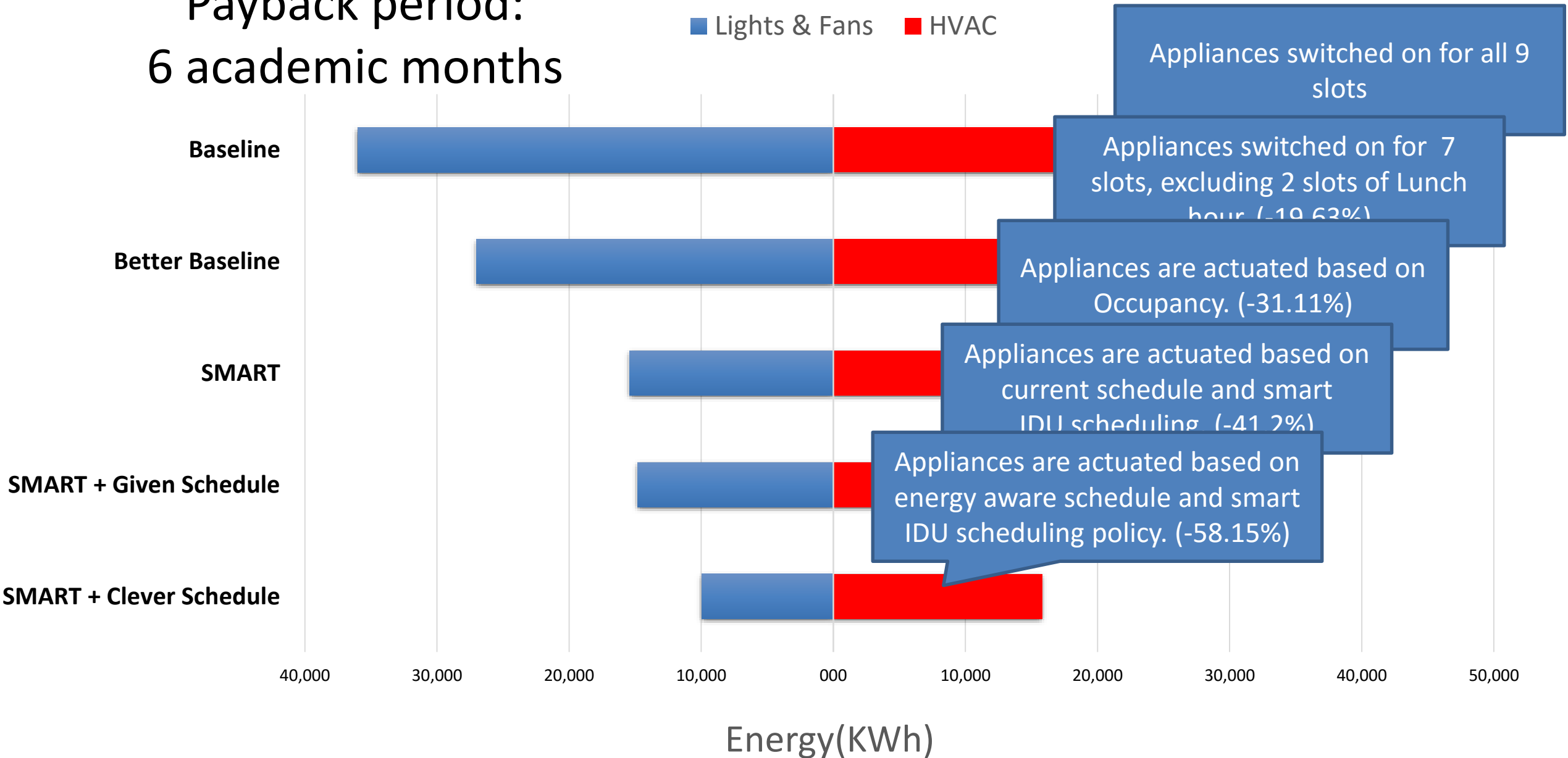
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Energy Savings: 58%
Payback period:
6 academic months

■ Lights & Fans ■ HVAC



Room Scheduling to Reduce Consumption

Old Schedule

Slot	SIC-205	SIC-305	SIC-201	SIC-301
1				
2	HS 490 (12)		CS 681 (38)	CS 435 (68)
3			CS 735 (53)	CS 718 (38)
4		CS 632 (10)	CS 743 (31)	
5				
6				
7	CS 735 (14)		CS 736 (36)	CS 741 (54)
8				
9				CS 775 (20)

New Schedule

Slot	SIC-205	SIC-305	SIC-201	SIC-301
1				
2	HS 490 (12)	CS 681 (38)		CS 435 (68)
3	CS 718 (38)			CS 735 (53)
4	CS 743 (31)	CS 632 (10)		
5				
6				
7	CS 735 (14)	CS 736 (36)		CS 741 (54)
8				
9		CS 775 (20)		

SIC 201 and sic 301 are large classrooms

9



Deploying Sensors:

What to sense? How to sense?

Observability:

A Principled Approach to Deploying Sensors

Is space occupied?



PIR sensors



Camera sensors

Blind spots

False positives/negatives

Sensing Occupancy (Ranking)

Alternative s	Correctness	Prompt- ness	Data	Resilience to Network Problems	Cost Efficiency
<i>PIR</i>	5	2	1	1	1
<i>PIR Array</i>	4	3	2	3	2
<i>Camera</i>	3	1	5	5	3
<i>Camera + PIR</i>	2	5	4	4	4
<i>Camera + PIR Array</i>	1	4	3	3	5

Sensing Occupancy (Properties)

Alternatives	False Positive	False Negative	Network Usage	Zonal Occupancy
<i>PIR</i>	High	High	None	No
<i>PIR Array</i>	High	Low	None	Yes
<i>Camera</i>	Medium	Medium	High	No
<i>Camera + PIR</i>	Low	Low	High	No
<i>Camera + PIR Array</i>	Low	Low	Low	Yes

Factors affecting the choice of sensors

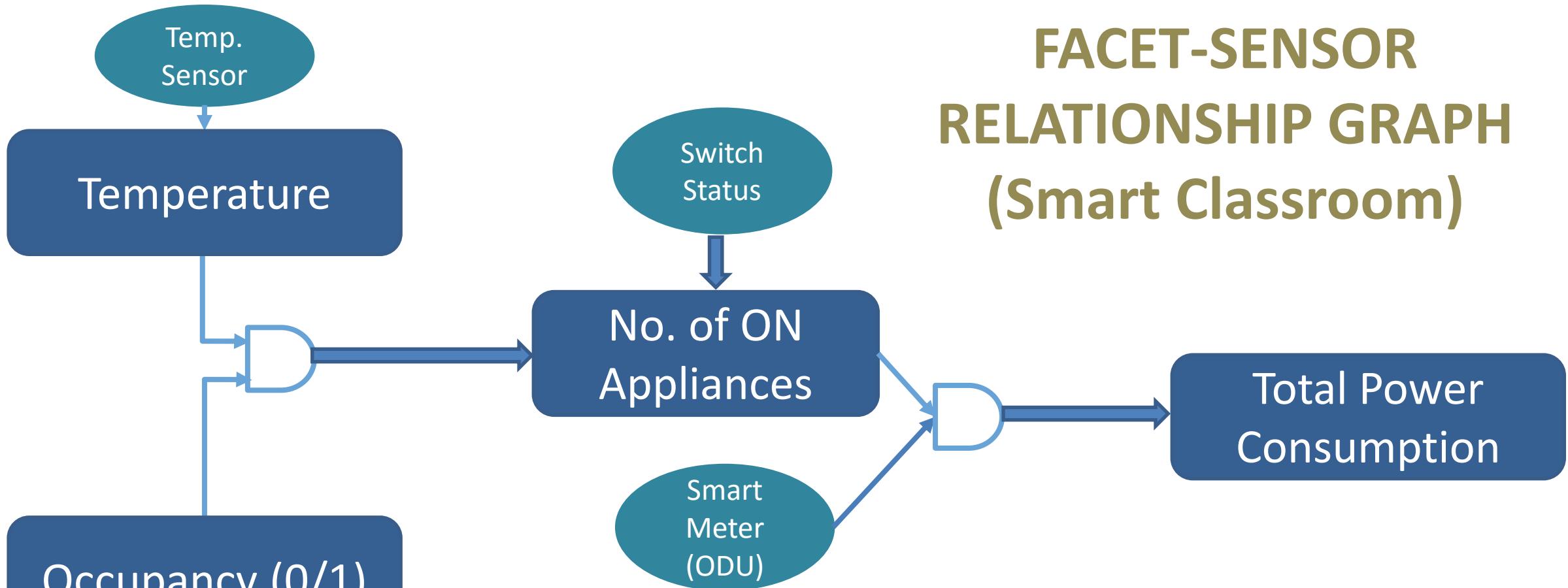
Choosing a combination of sensors:
based on

- what is already available
- accuracy, maintainability, granularity, reliability

If the accuracy from current sources does not meet
accuracy needs do we need a dedicated physical sensor?

If the inference requires more than one input,
for example occupancy and temperature,
will reliability be affected?

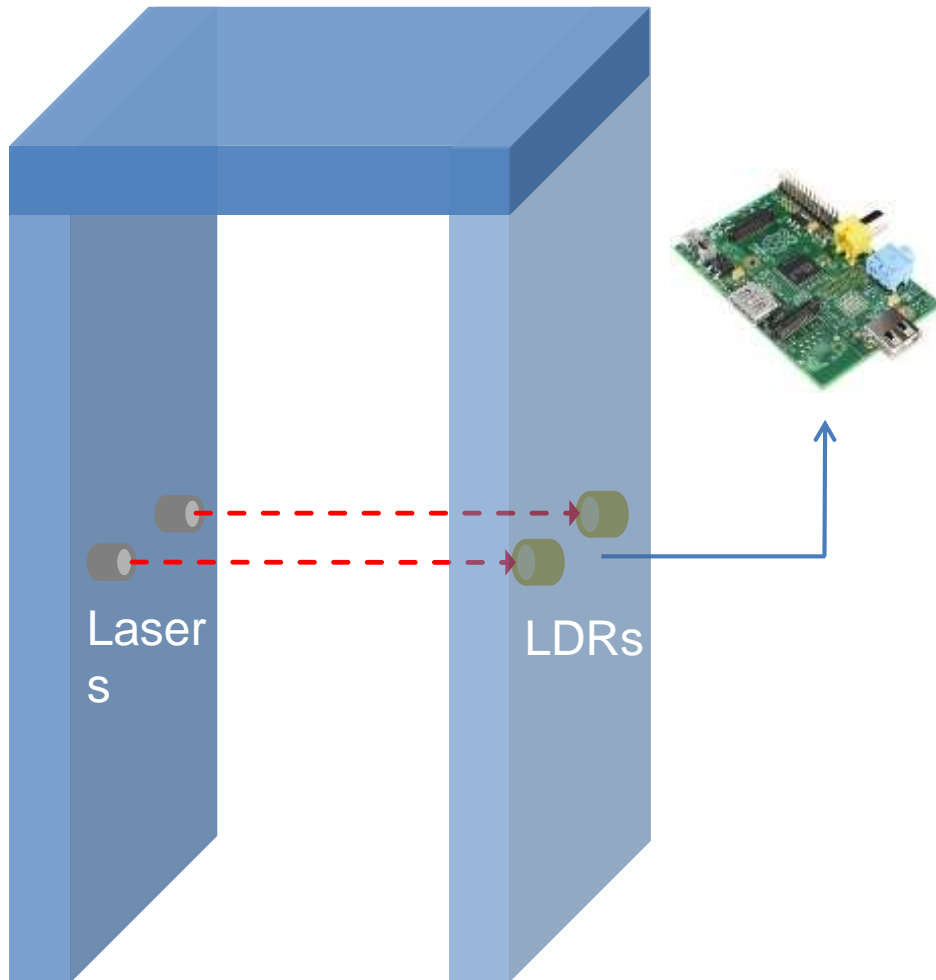
FACET-SENSOR RELATIONSHIP GRAPH (Smart Classroom)



EDGE'S LEGEND		NODE'S LEGEND	
	Data through Inference Engine		Observability Facet
	Data flow		Hard Sensor
			Soft Sensor

#Occupants

Spotlight
SMART
Door



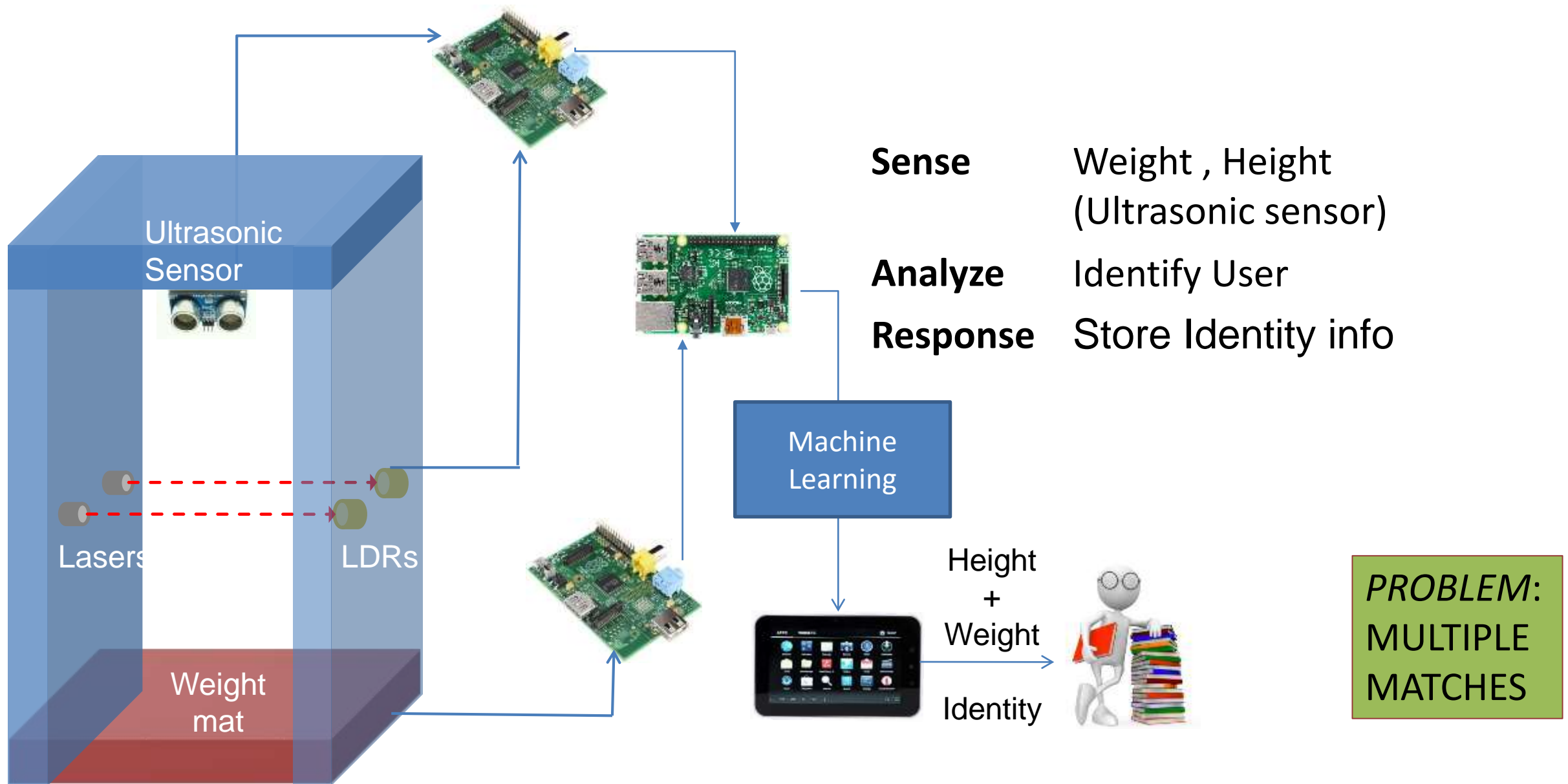
Goal: Be counted,
while walking in/out
at normal pace

Sense	Entry/Exit
Analyze	Hand or Torso? <ul style="list-style-type: none">• Intra Laser Threshold
Response	Update Occupancy Count

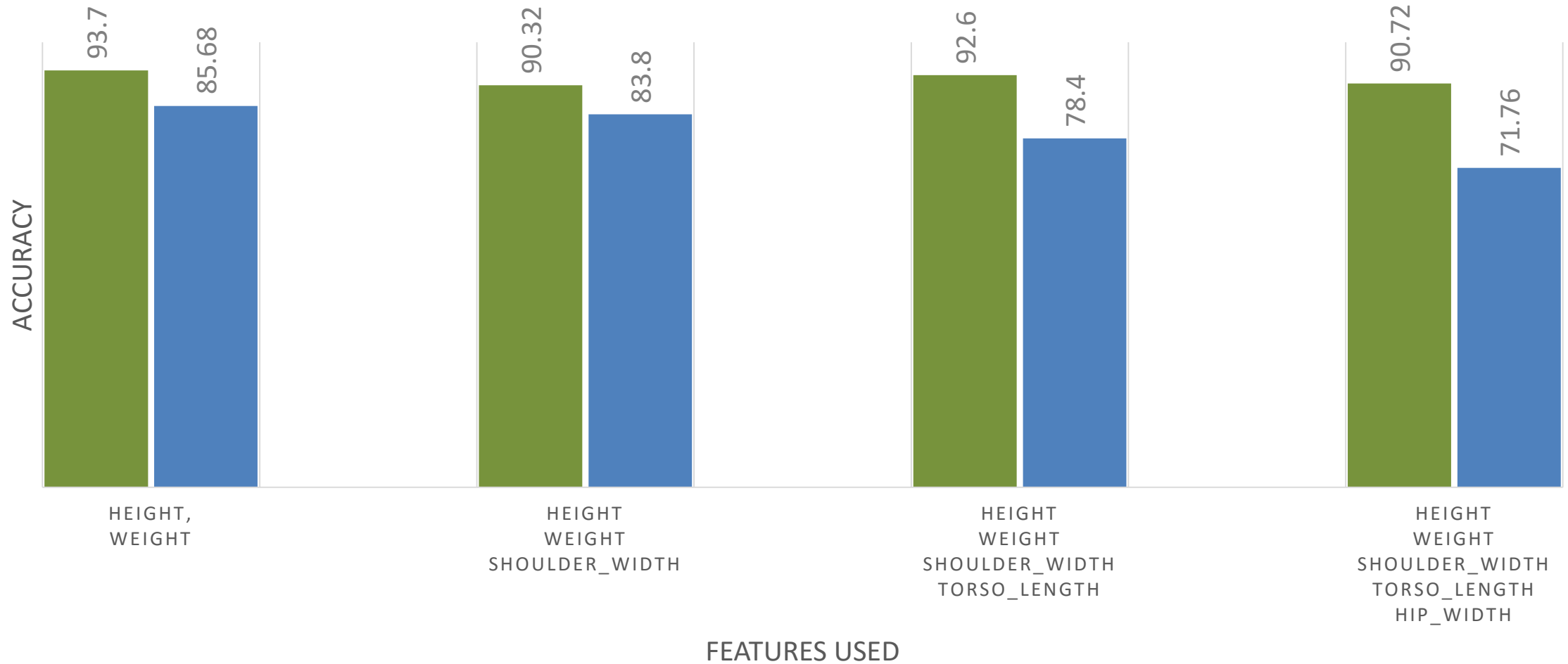
Entering/Leaving?



Who is in the room?



CLASSIFICATION USING DIFFERENT FEATURES



■ Random Forest

■ SVM

Disambiguation through Sensor Fusion

From *height + weight* based (*hard sensor*)
information

IDENTIFY the set of possible occupants given
occupants list

-- {A,B,C}

Applicable to spaces with limited number of users, e.g., offices, class rooms

Disambiguation through Sensor Fusion

From *height + weight* based (*hard sensor*)
information

IDENTIFY the set of possible occupants given
occupants list

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PRUNE set using *soft sensors*

Applicable to spaces with limited number of users, e.g., offices, class rooms

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(people's/rooms' schedules/occupancy, etc.)

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-- C is in a class

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PRUNE set using *soft sensors*

(people's/rooms' schedules/occupancy, etc.)

-- C is in a class

-- B usually comes after lunch

Applicable to spaces with limited number of users, e.g., offices, class rooms

Disambiguation through Sensor Fusion

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IDENTIFY the set of possible occupants given
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PRUNE set using *soft sensors*

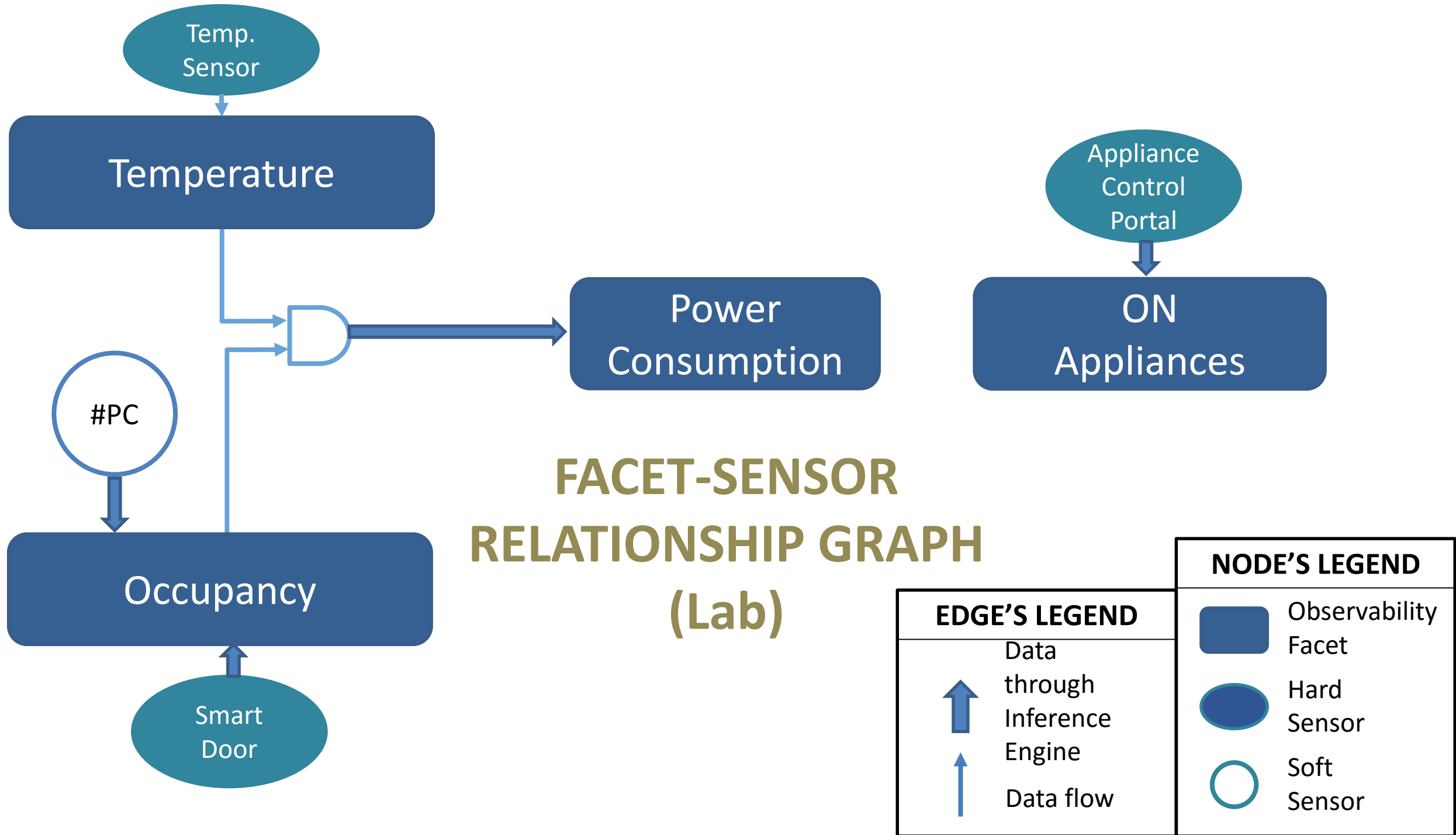
(people's/rooms' schedules/occupancy, etc.)

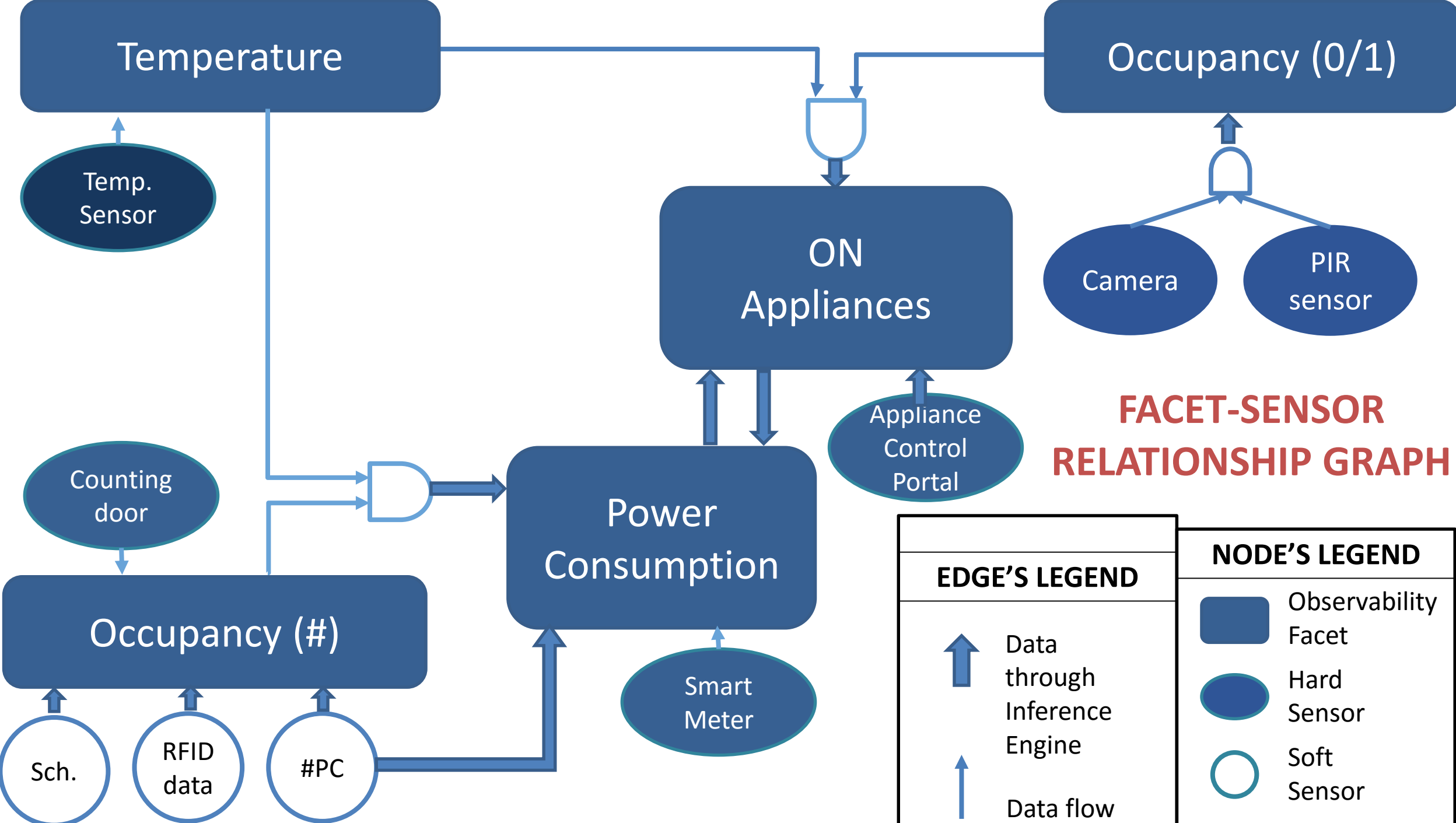
-- C is in a class

-- B usually comes after lunch

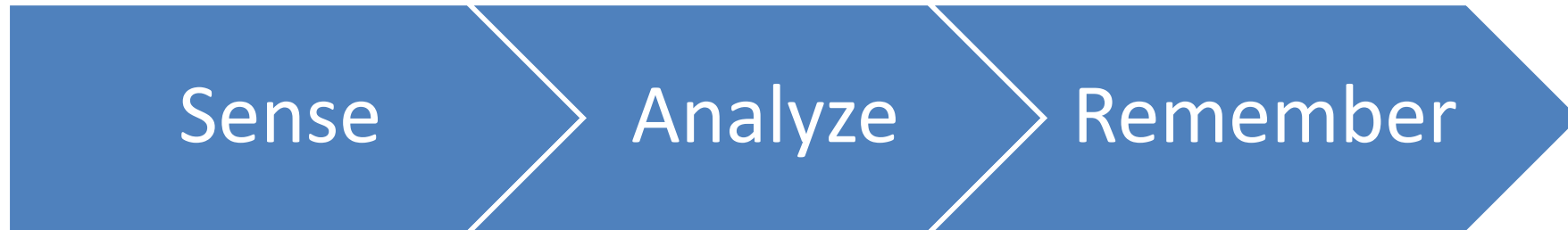
=> A is entering

Applicable to spaces with limited number of users, e.g., offices, class rooms





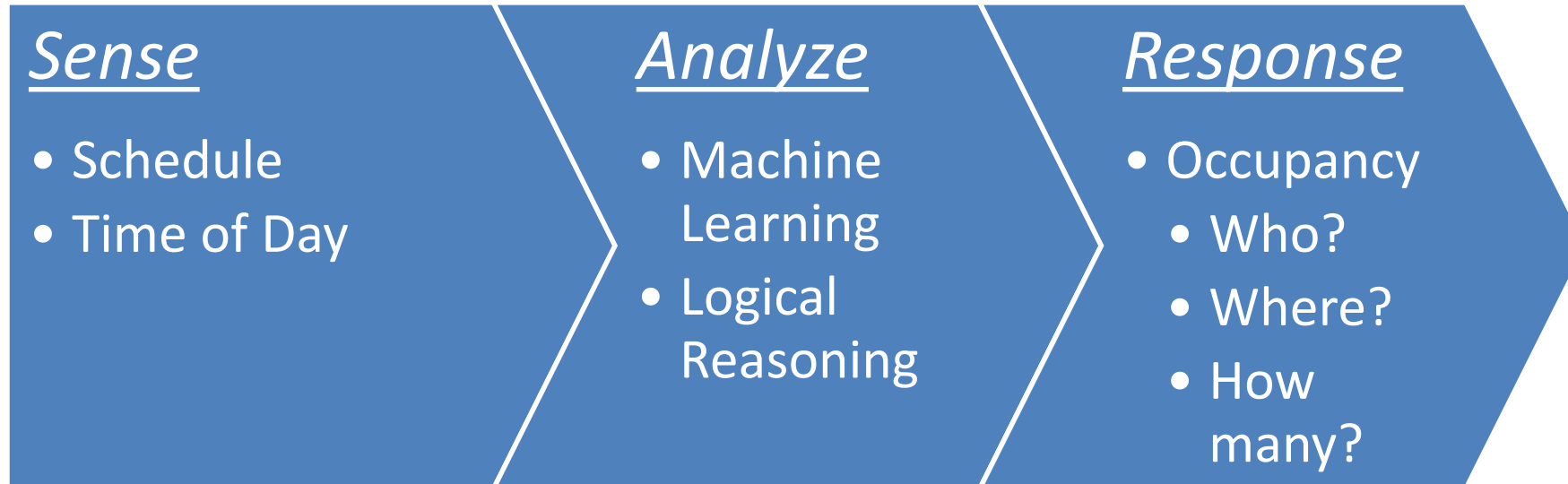
**FACET-SENSOR
RELATIONSHIP GRAPH**



Meaningfully

Timely

Information can be sensed/inferred

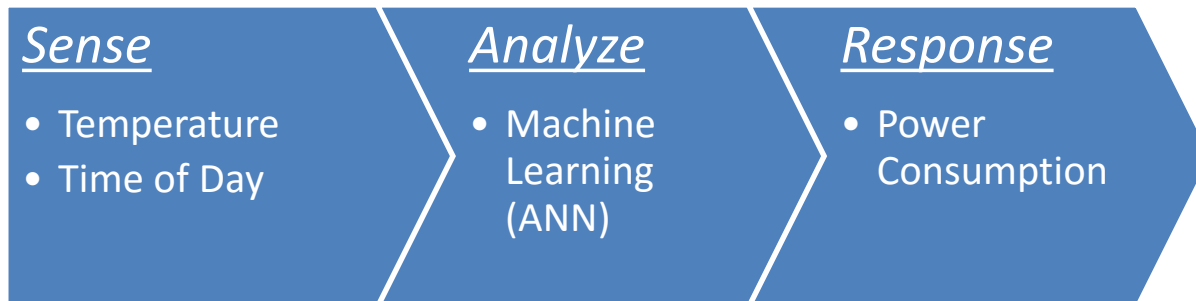
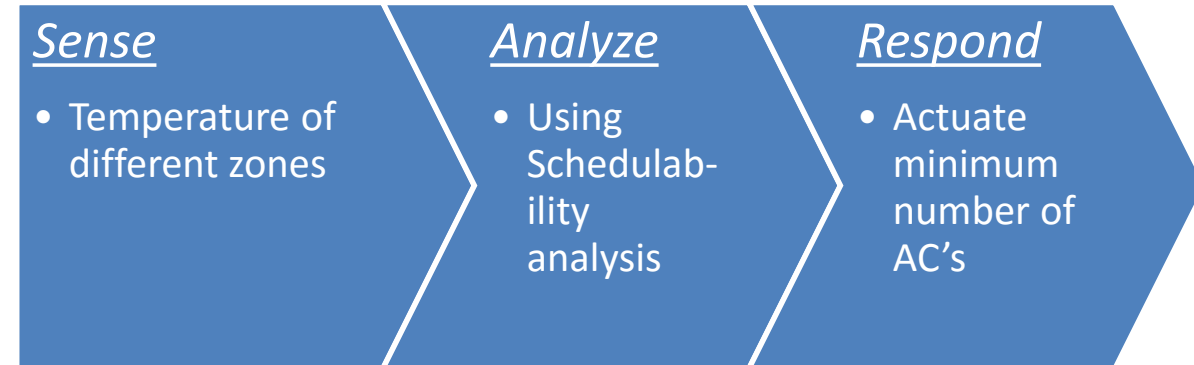
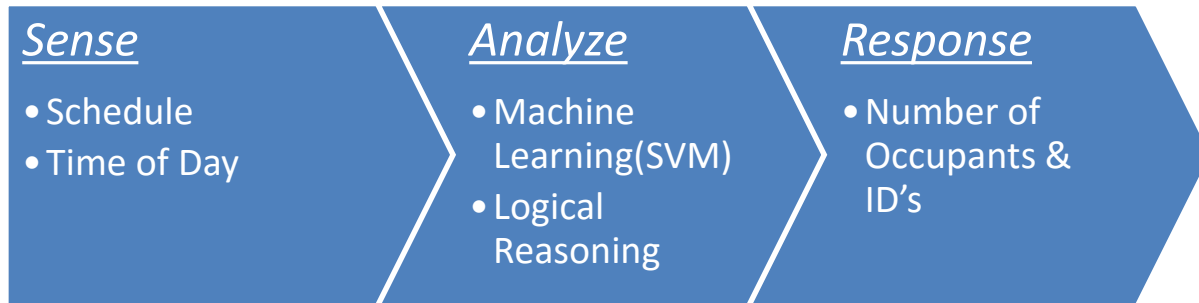


Goal:
Achieve needed
Observability of
phenomena

*Key Contributor
to Smartness:
Analysis*

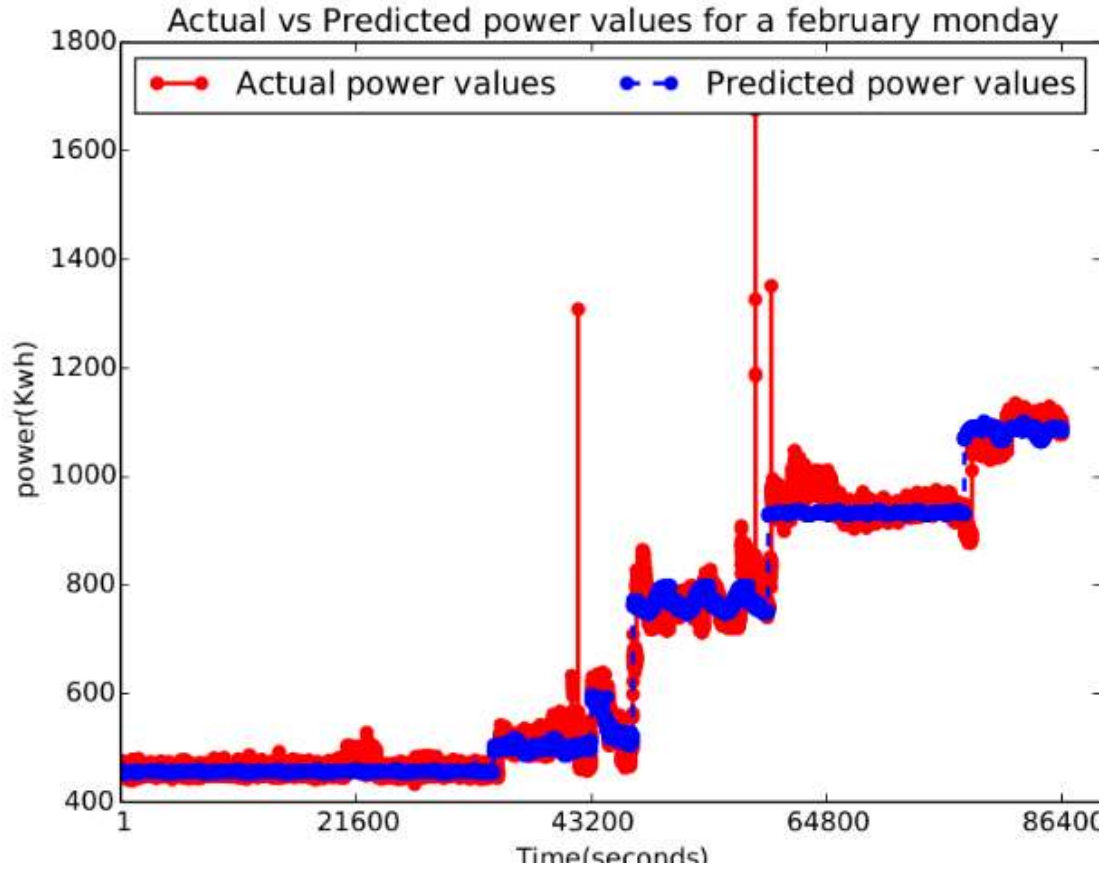
Location
of sensors
Types of sensors
– hard vs soft

Types of Analysis

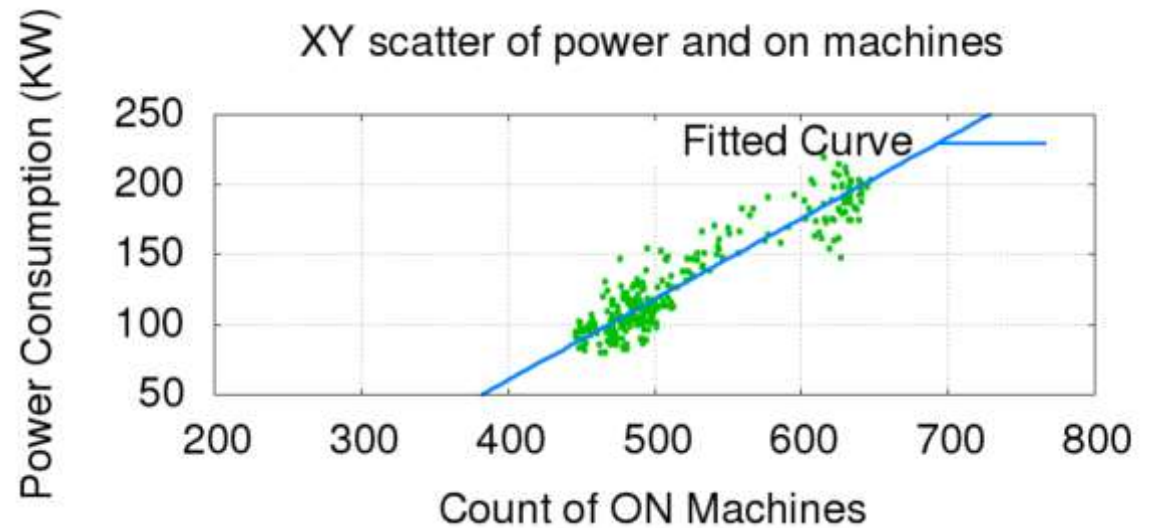


Analysis:

- Model-based reasoning
- Exploiting correlated phenomena
- Regression

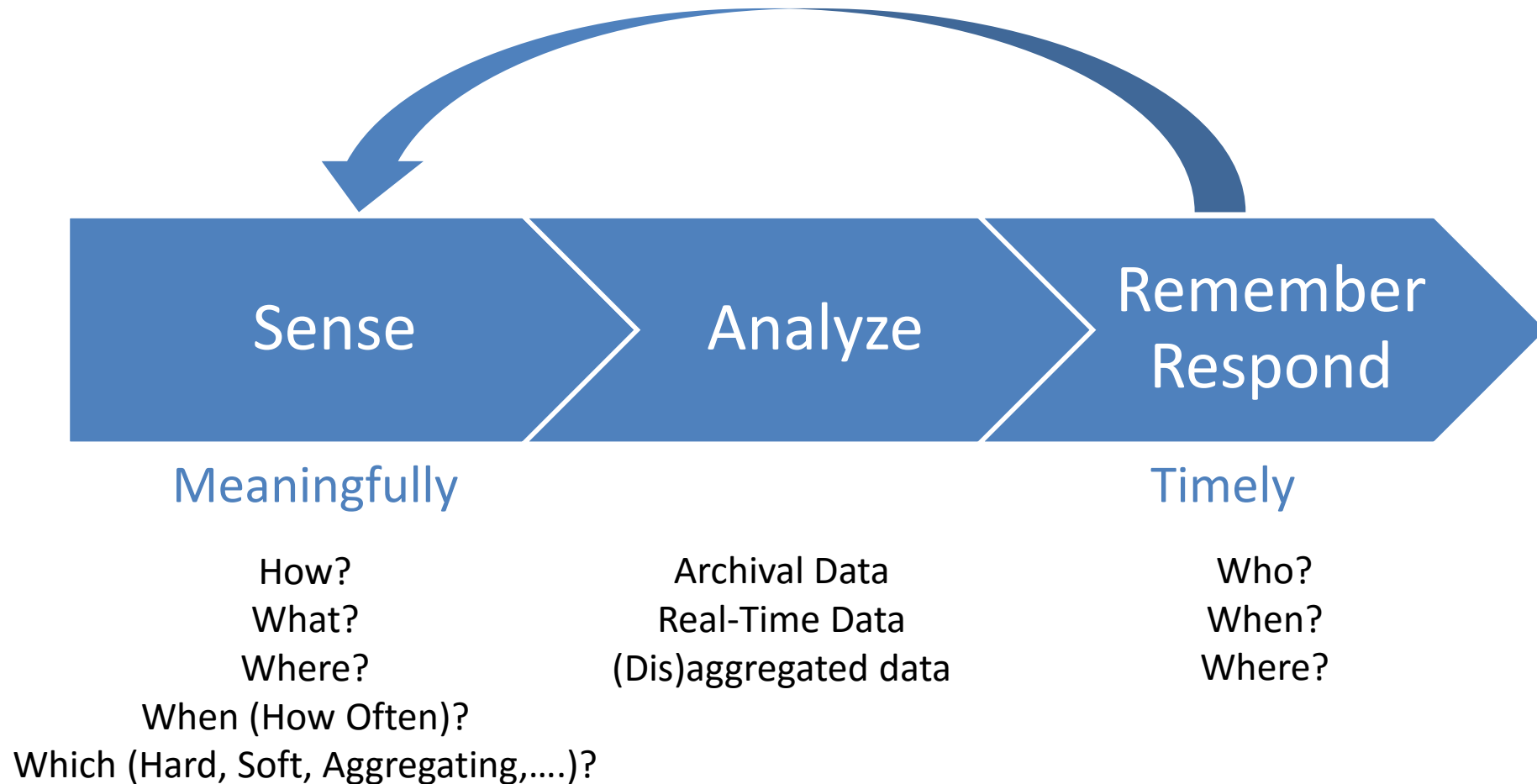


Inference Engines for predicting power consumption



SMART Design Philosophy

It is a continuous feedback-driven process



Being

S M A

R T

SENSE

What
When (how often)
How
Where
Which (hard, soft, aggregating...)

• MEANINGFULLY

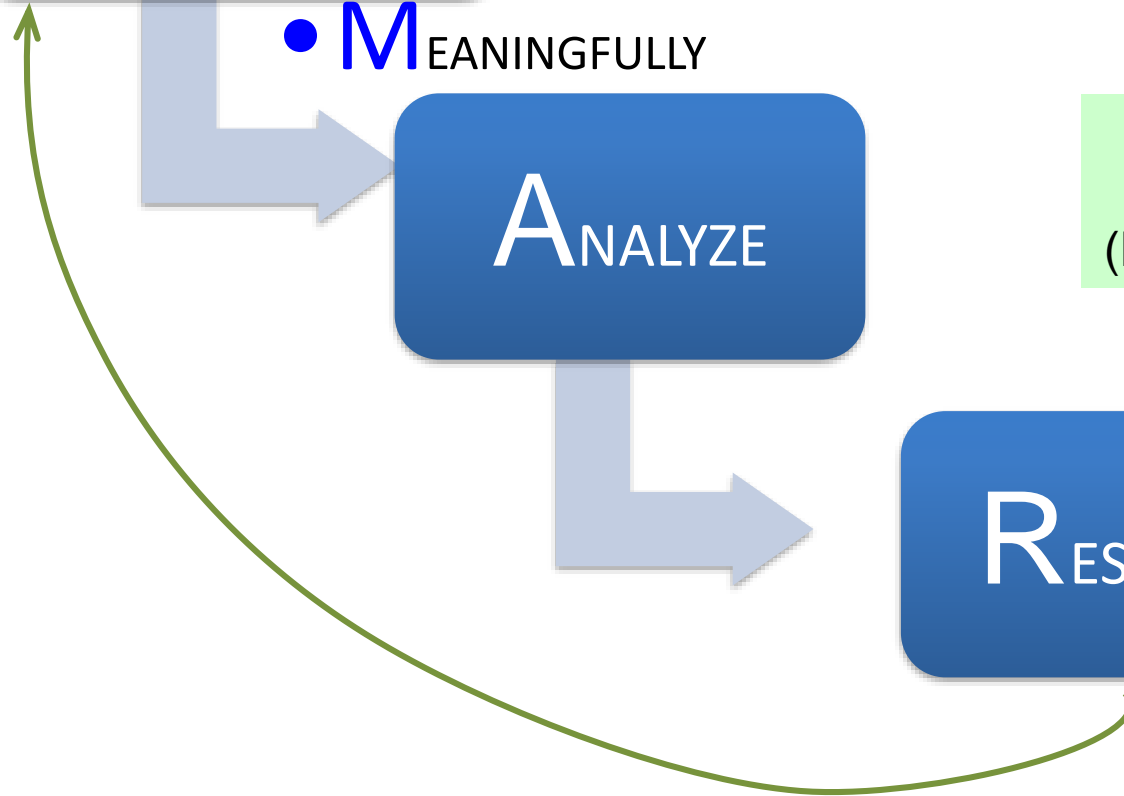
ANALYZE

Archival data
Real-Time data
(Dis)aggregated data

REMEMBER
RESPOND

Who
When
Where

• TIMELY



Being

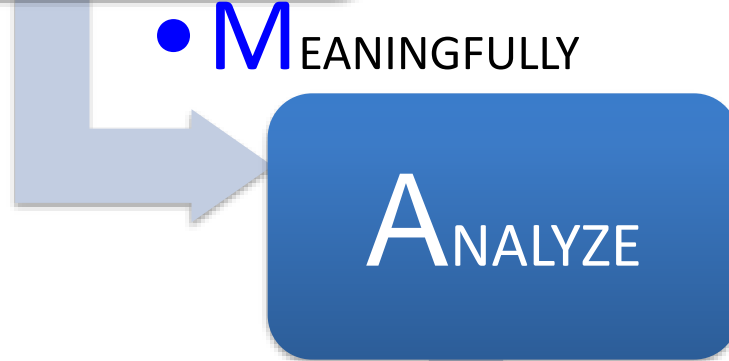
S M A

R T



What
When (how often)
How
Where
Which (hard, soft, aggregating...)

• M EANINGFULLY



Archival data
Real-Time data
(Dis)aggregated data

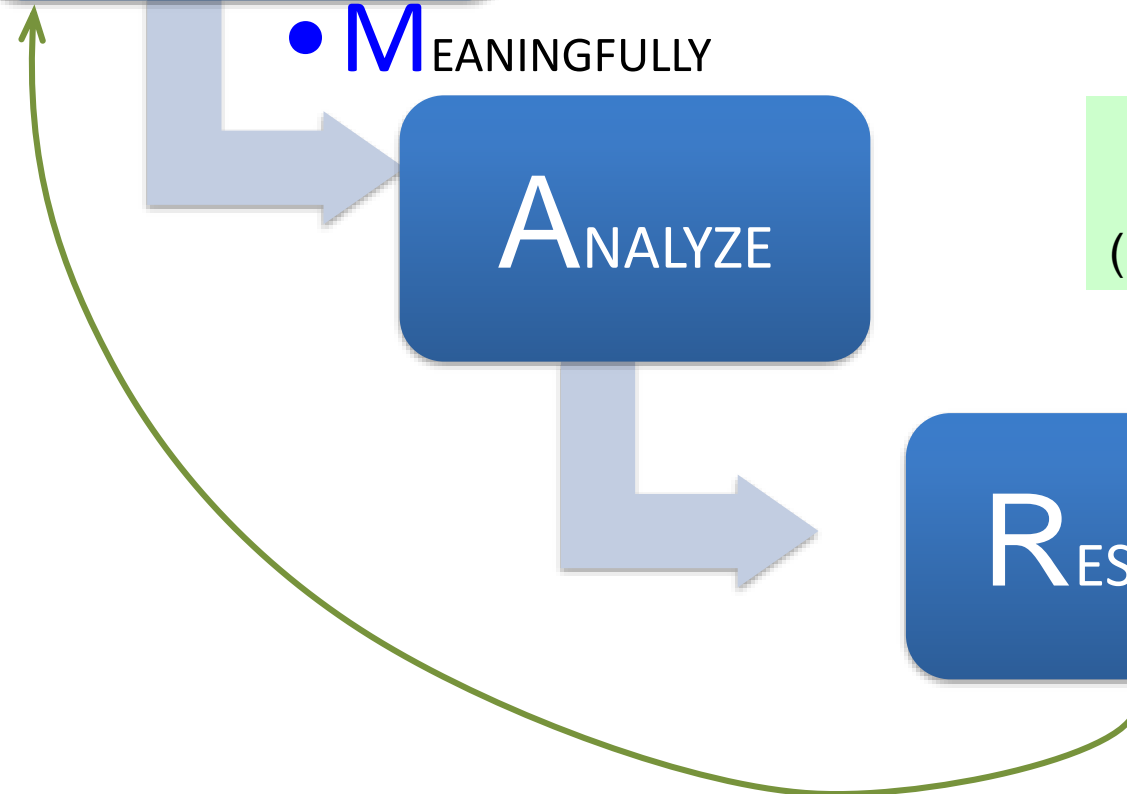
Applicable at

- *Design*
- *Deployment*
- *Initialization*
- *Run Time*



Who
When
Where

• T IMELY



Smart Lecture Hall

Requesting people in the back rows
to move to the front



Smart Lecture Hall

Requesting people in the back rows
to move to the front

During this talk
we will save



Smart Lecture Hall

Requesting people in the back rows
to move to the front

During this talk
we will save

30% energy



Smart Lecture Hall

Requesting people in the back rows
to move to the front

During this talk
we will save

30% energy

\$ 60



Smart Lecture Hall

Requesting people in the back rows
to move to the front

During this talk
we will save

30% energy

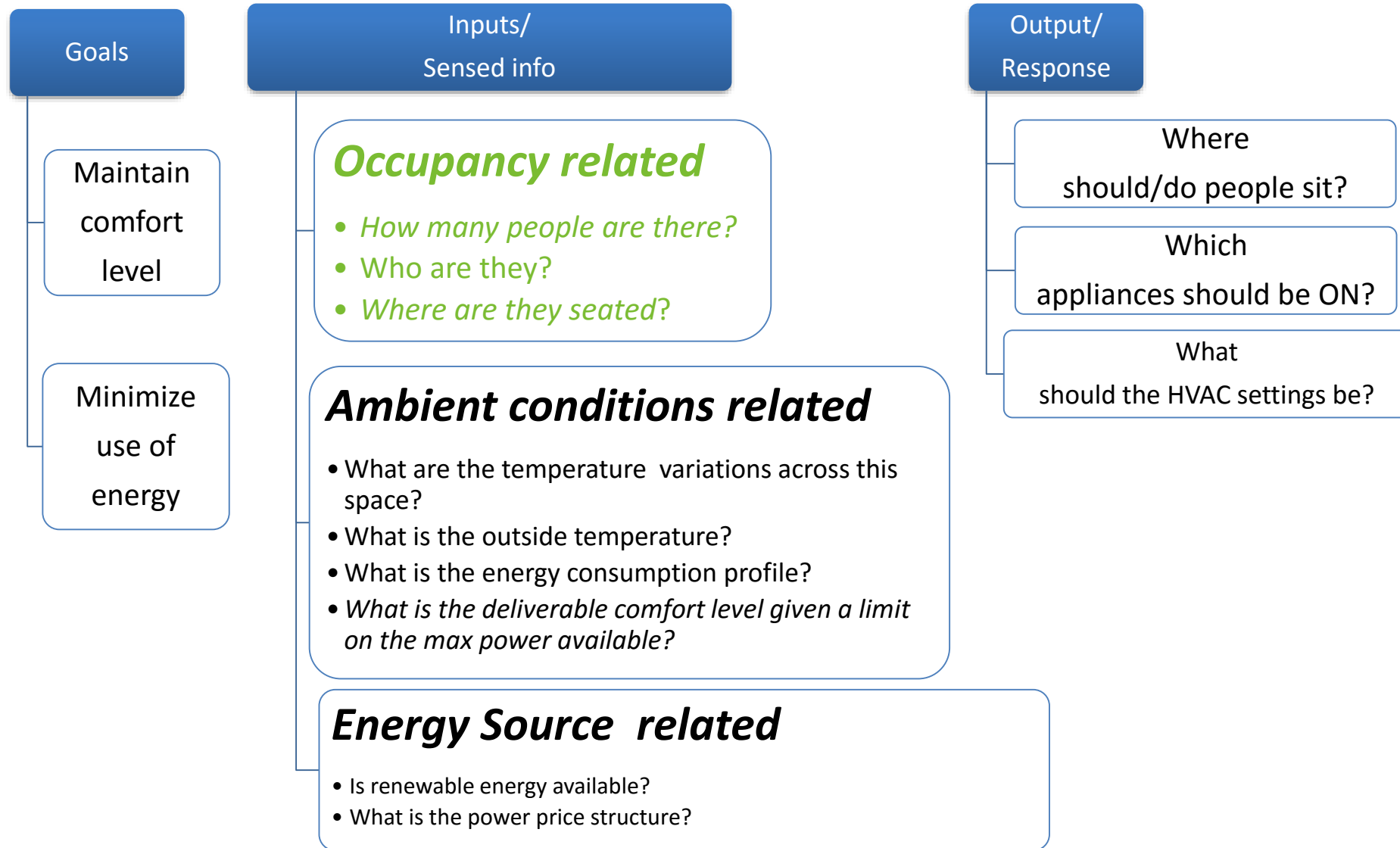
\$ 60

2.5 trees



What does the system do?

What does the system do?





Thermostatically Controlled Electrical Devices

**Exhibit Periodic
ON-OFF
Operation**



Air-Conditioners



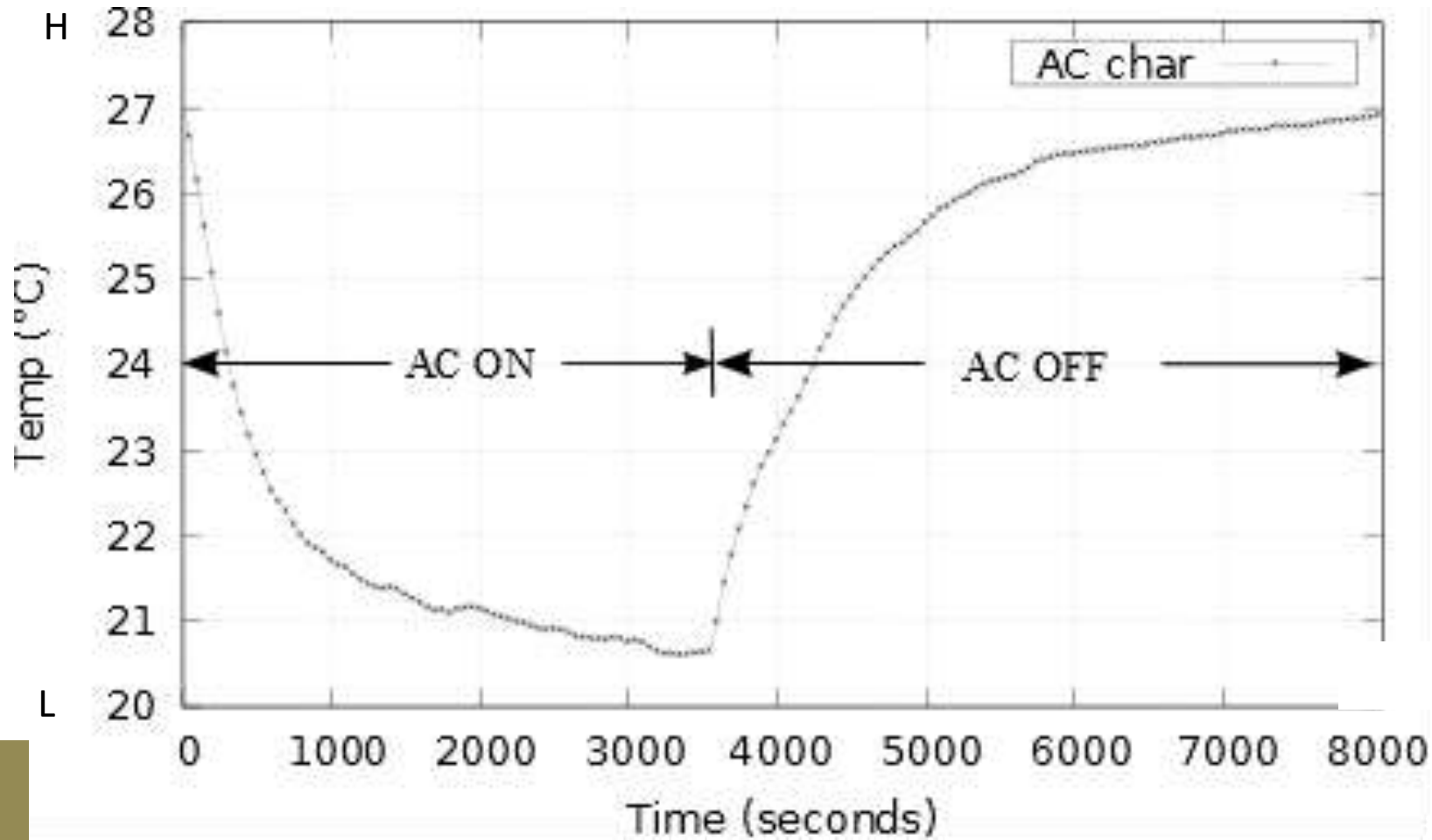
Refrigerators



Room-heaters

**Maintain
Temp within
[L, H]**

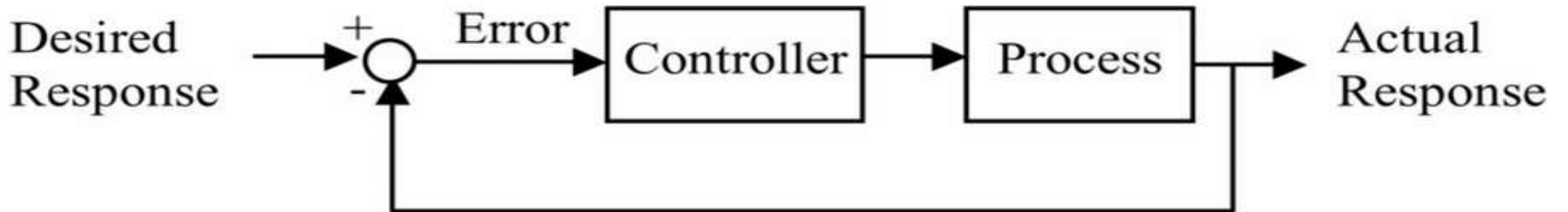
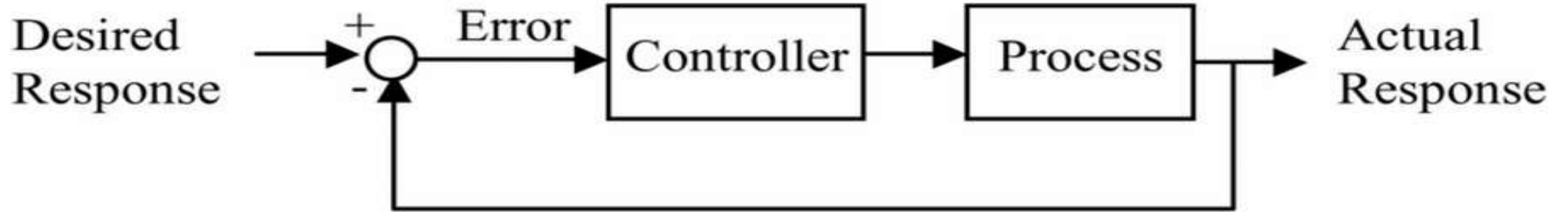
Behavior of a Cooling Unit



Run Time

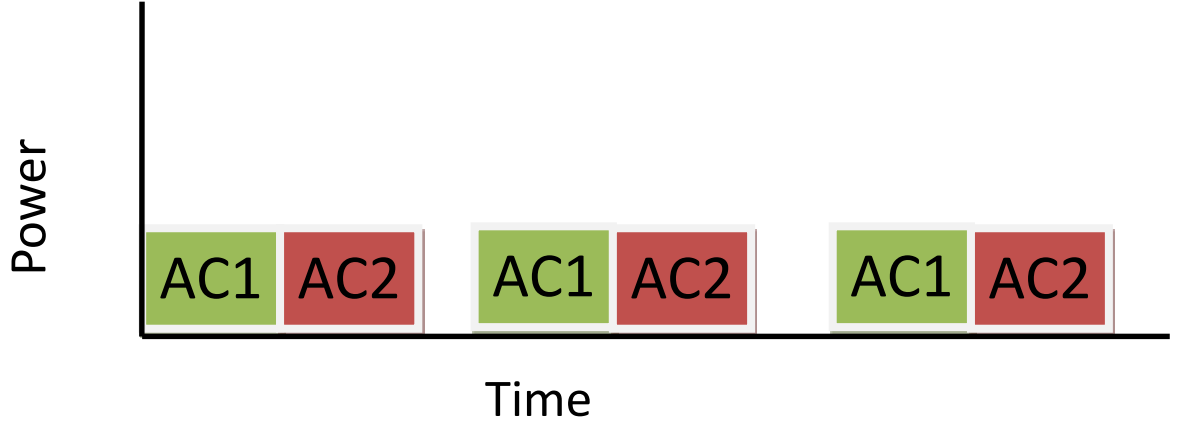
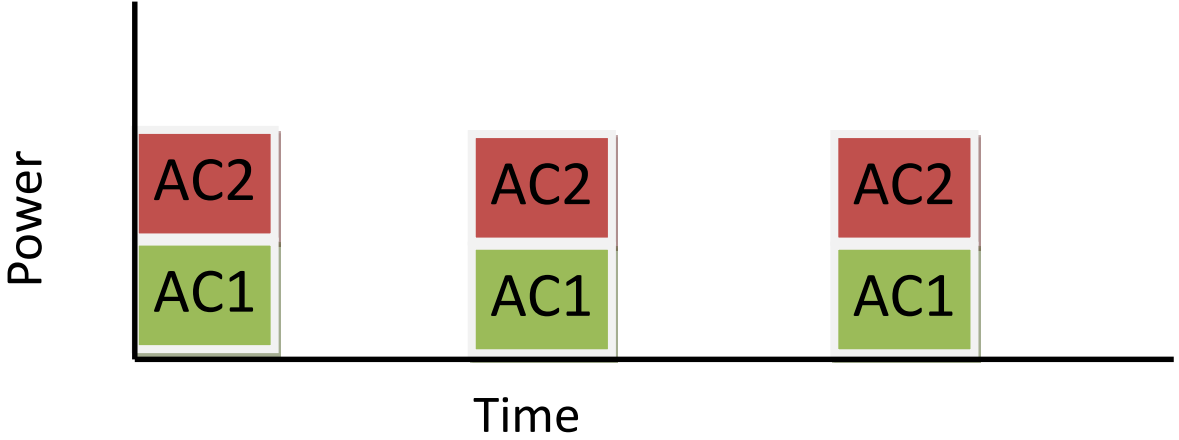
```
Sense Temp
If Temp > H
    Cooler_ON
If Temp < L
    Cooler_OFF
```

Feedback



Have to learn from the past

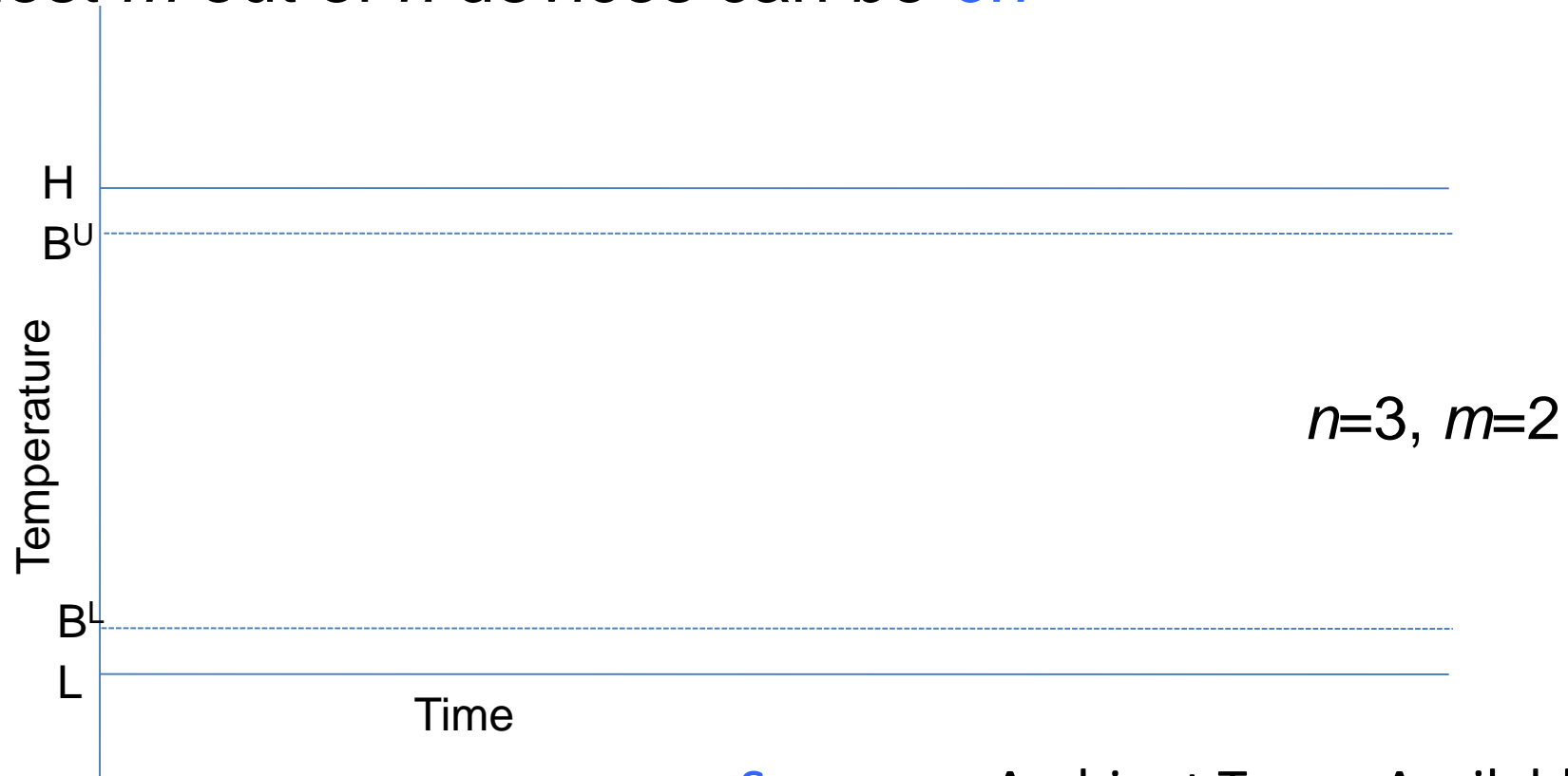
Smart Scheduling: Peak Demand Reduction



Thermal Comfort Band Maintenance Algorithm

Peak energy constraint:

At most m out of n devices can be *on*

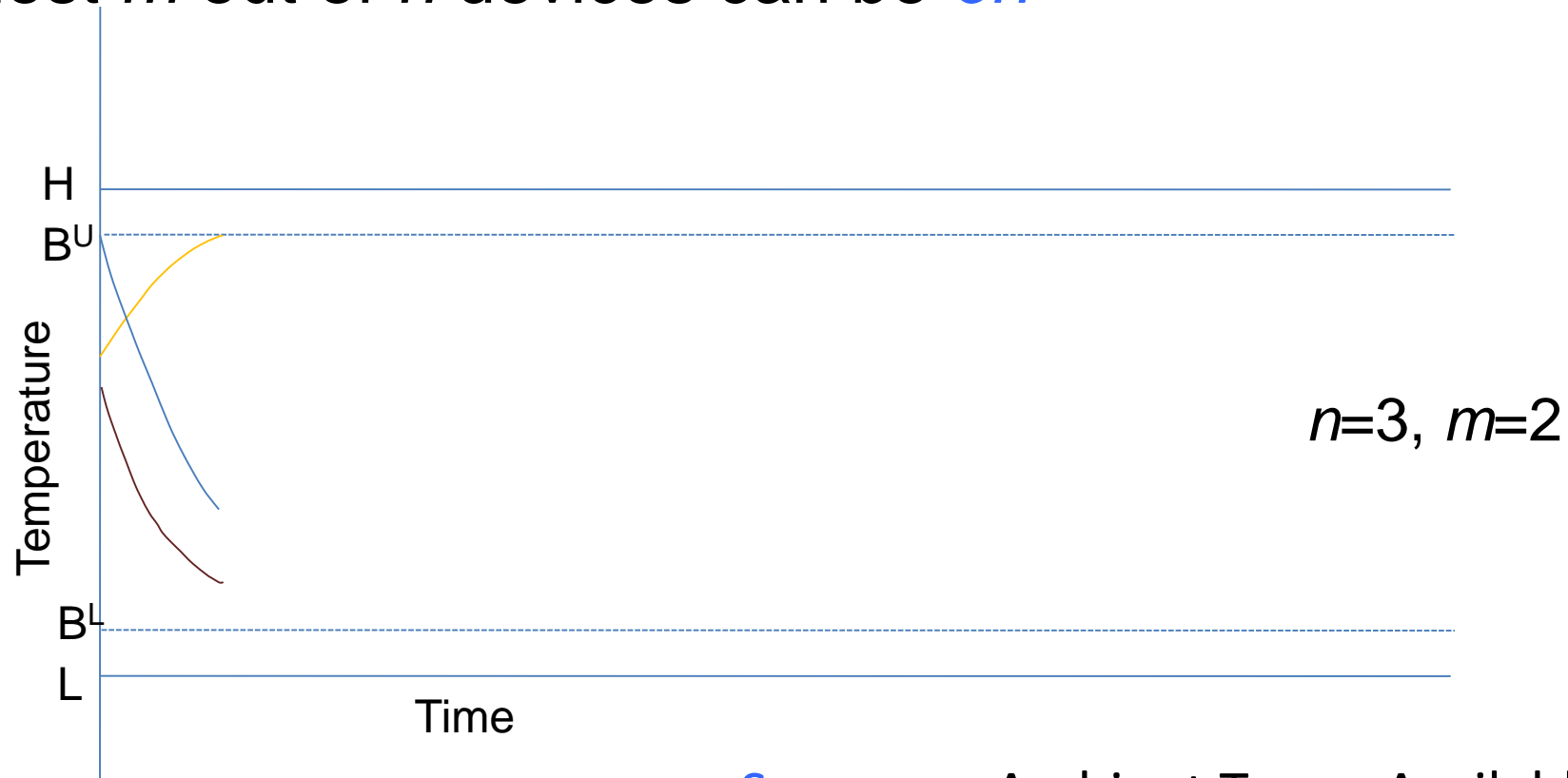


- | | |
|---------|------------------------------------|
| Sense | Ambient Temp, Available Peak Power |
| Analyze | Schedulability |
| Respond | Set m , Temp Band |

Thermal Comfort Band Maintenance Algorithm

Peak energy constraint:

At most m out of n devices can be *on*

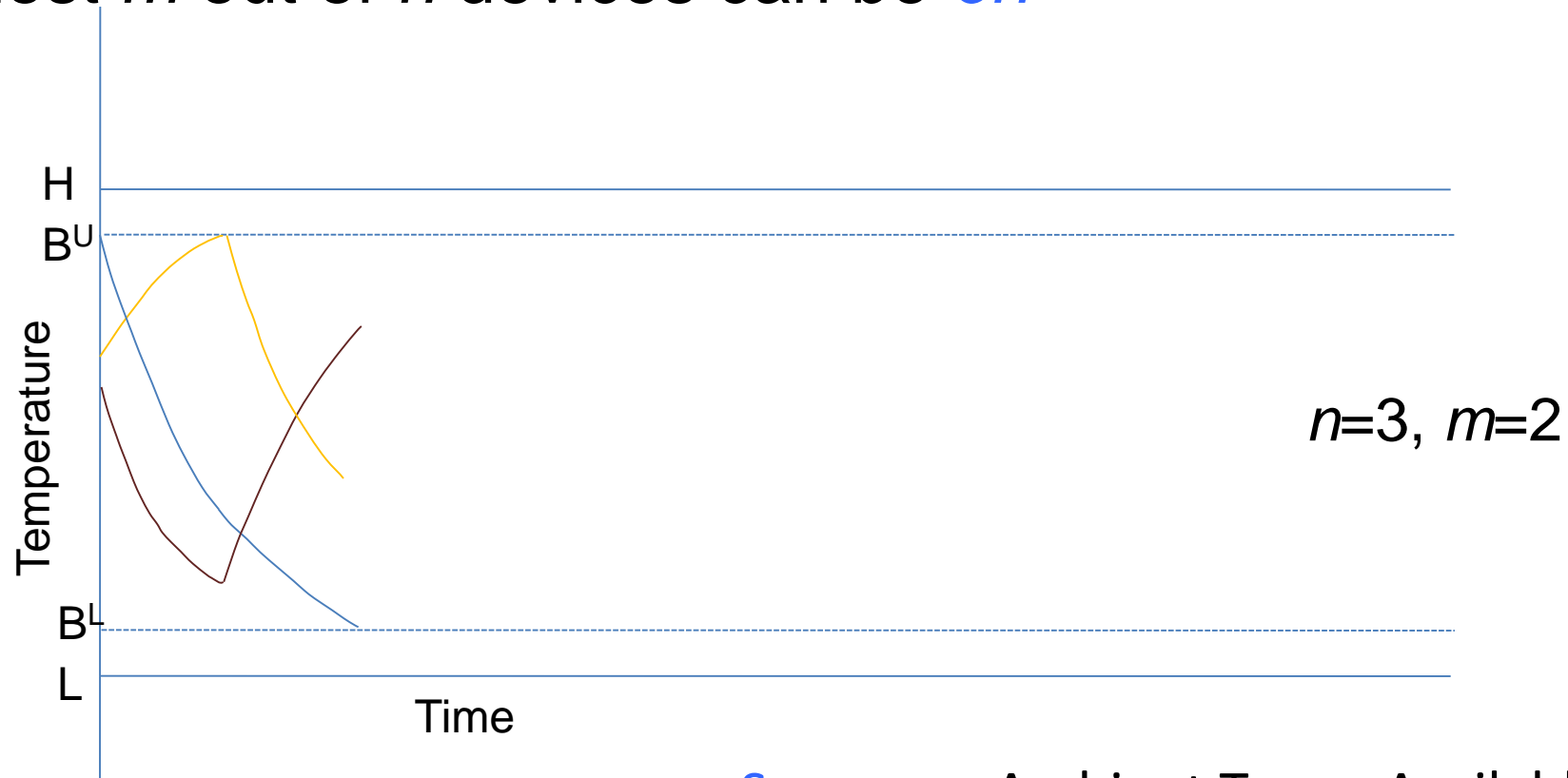


Sense	Ambient Temp, Available Peak Power
Analyze	Schedulability
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Thermal Comfort Band Maintenance Algorithm

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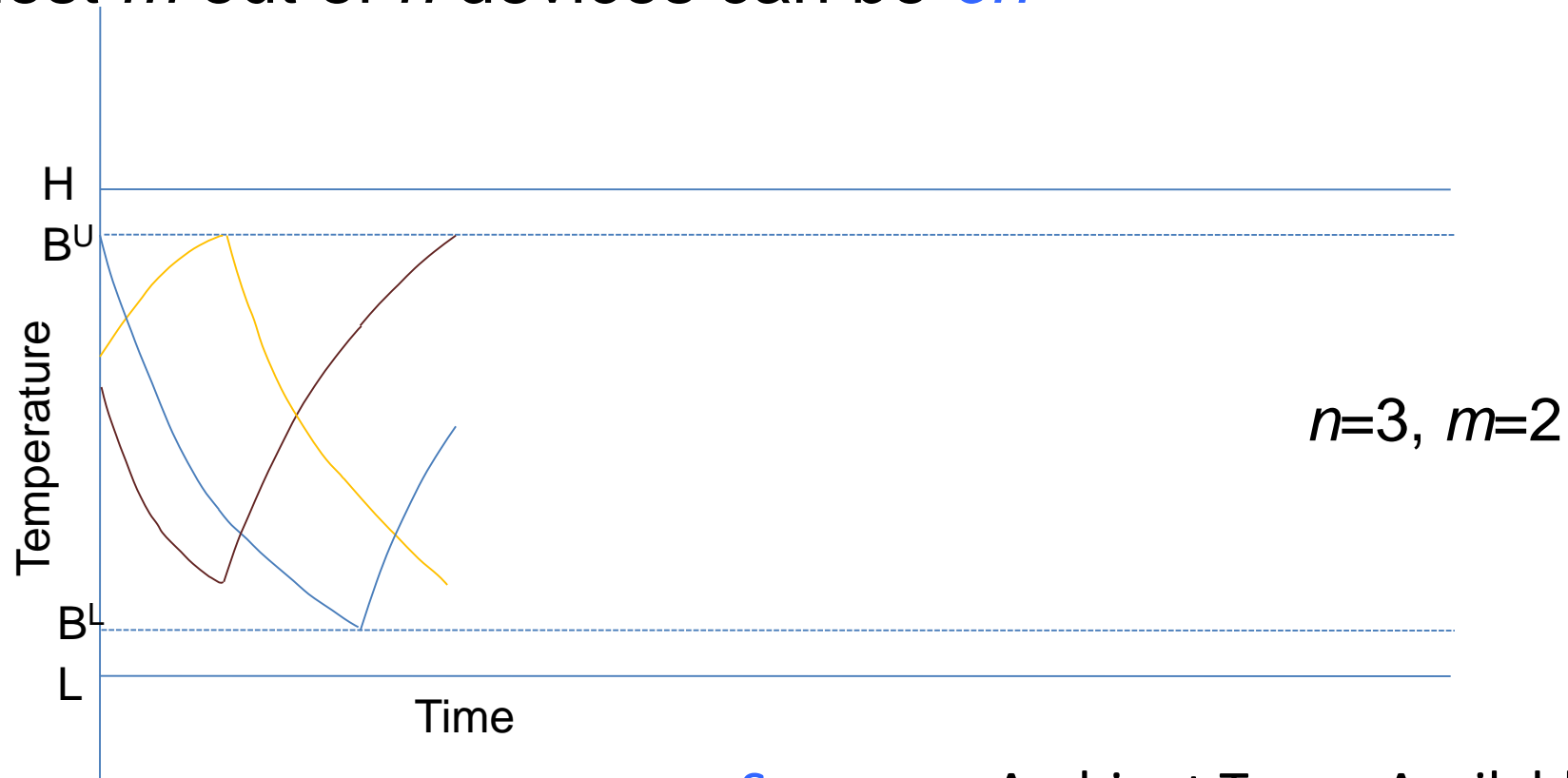


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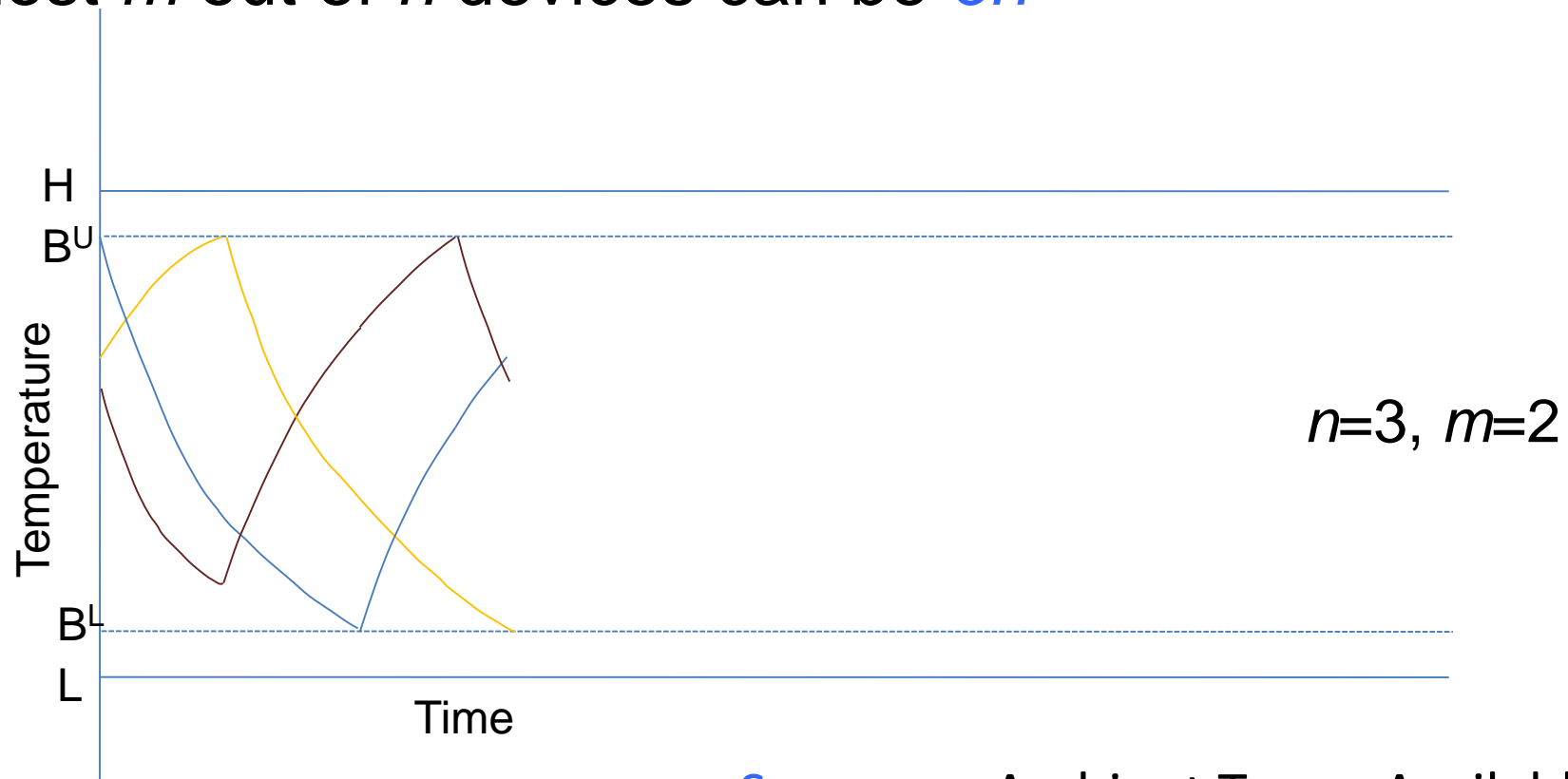


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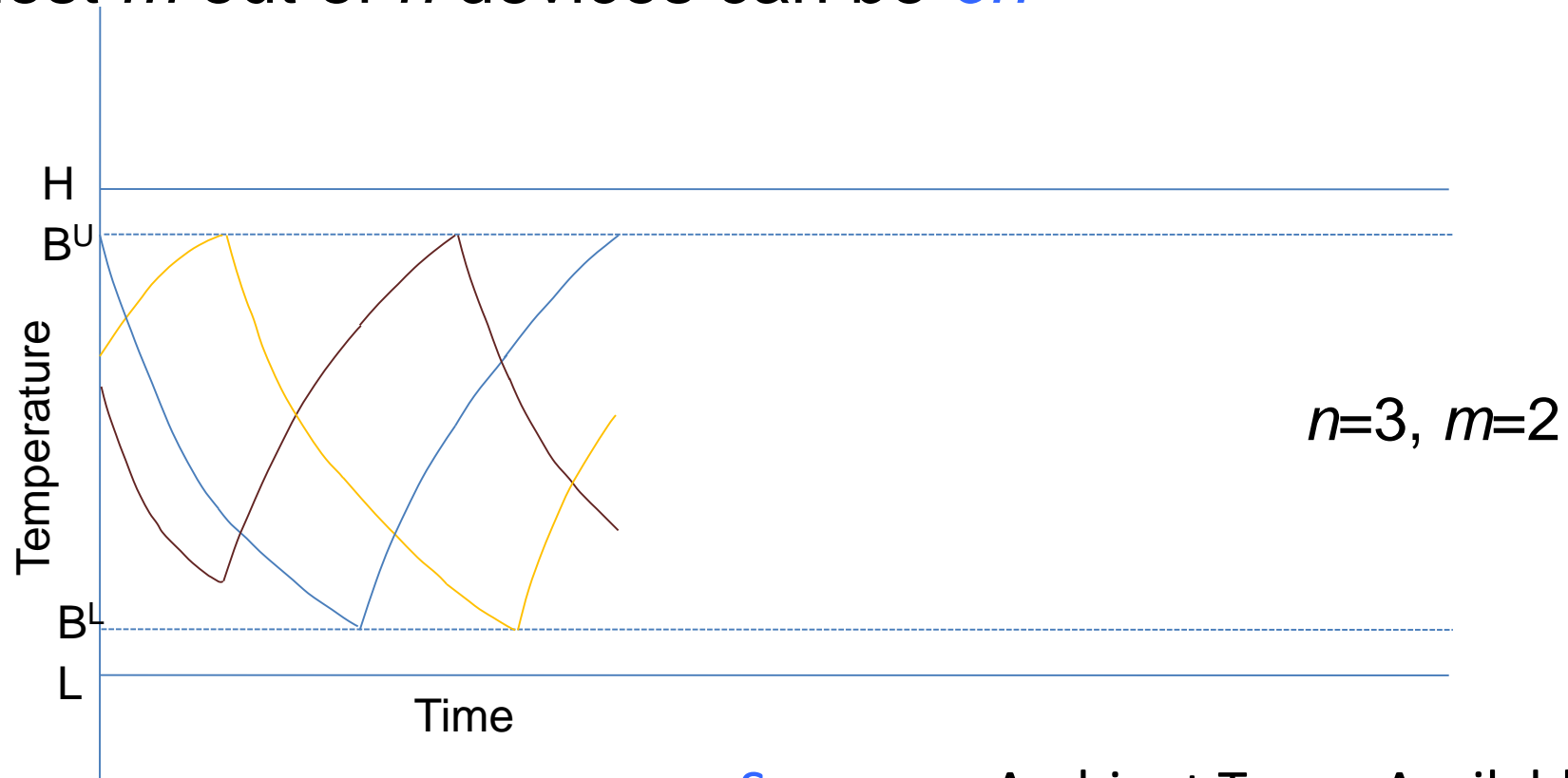


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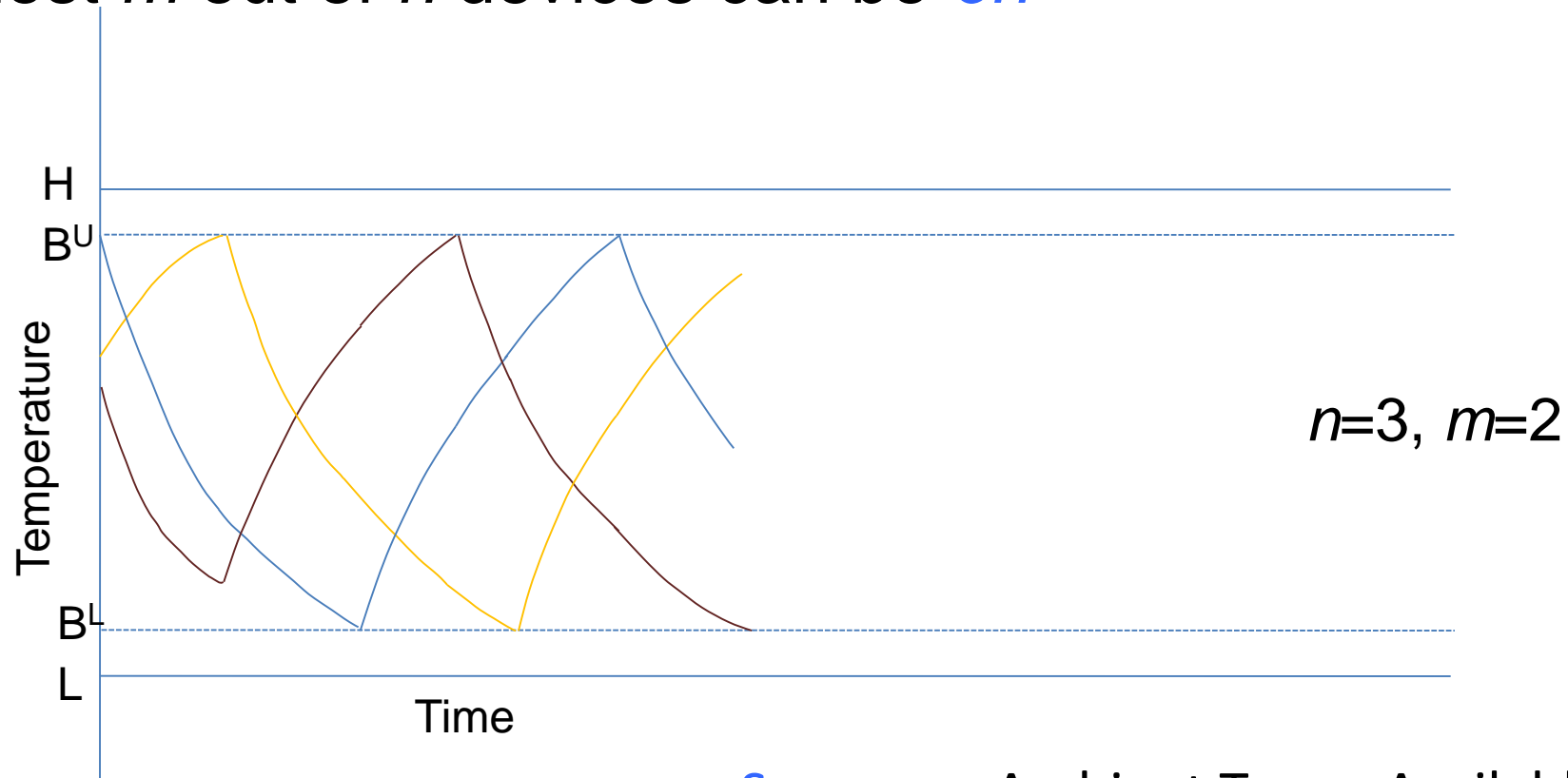


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Thermal Comfort Band Maintenance Algorithm

Peak energy constraint:

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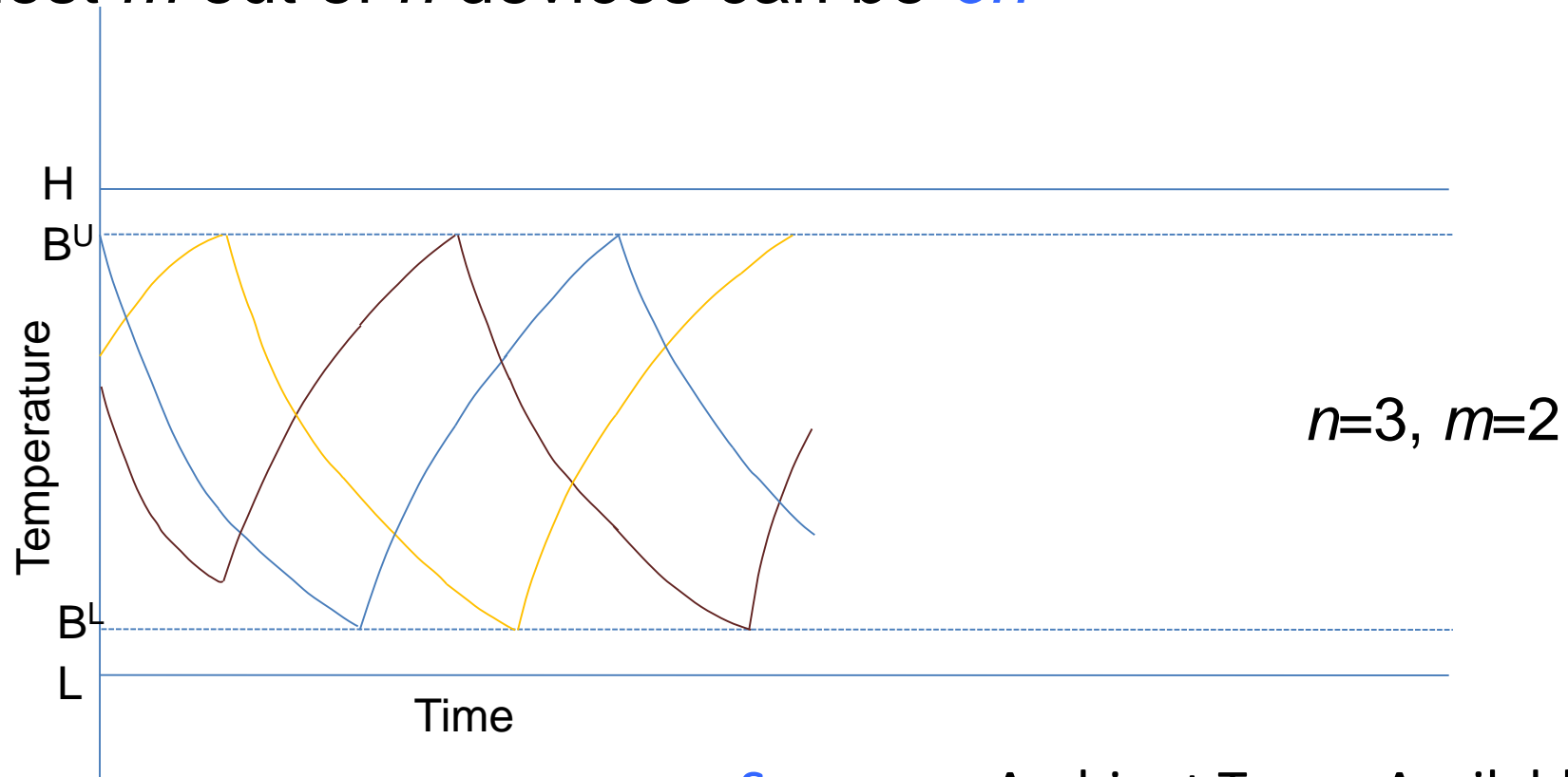


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Thermal Comfort Band Maintenance Algorithm

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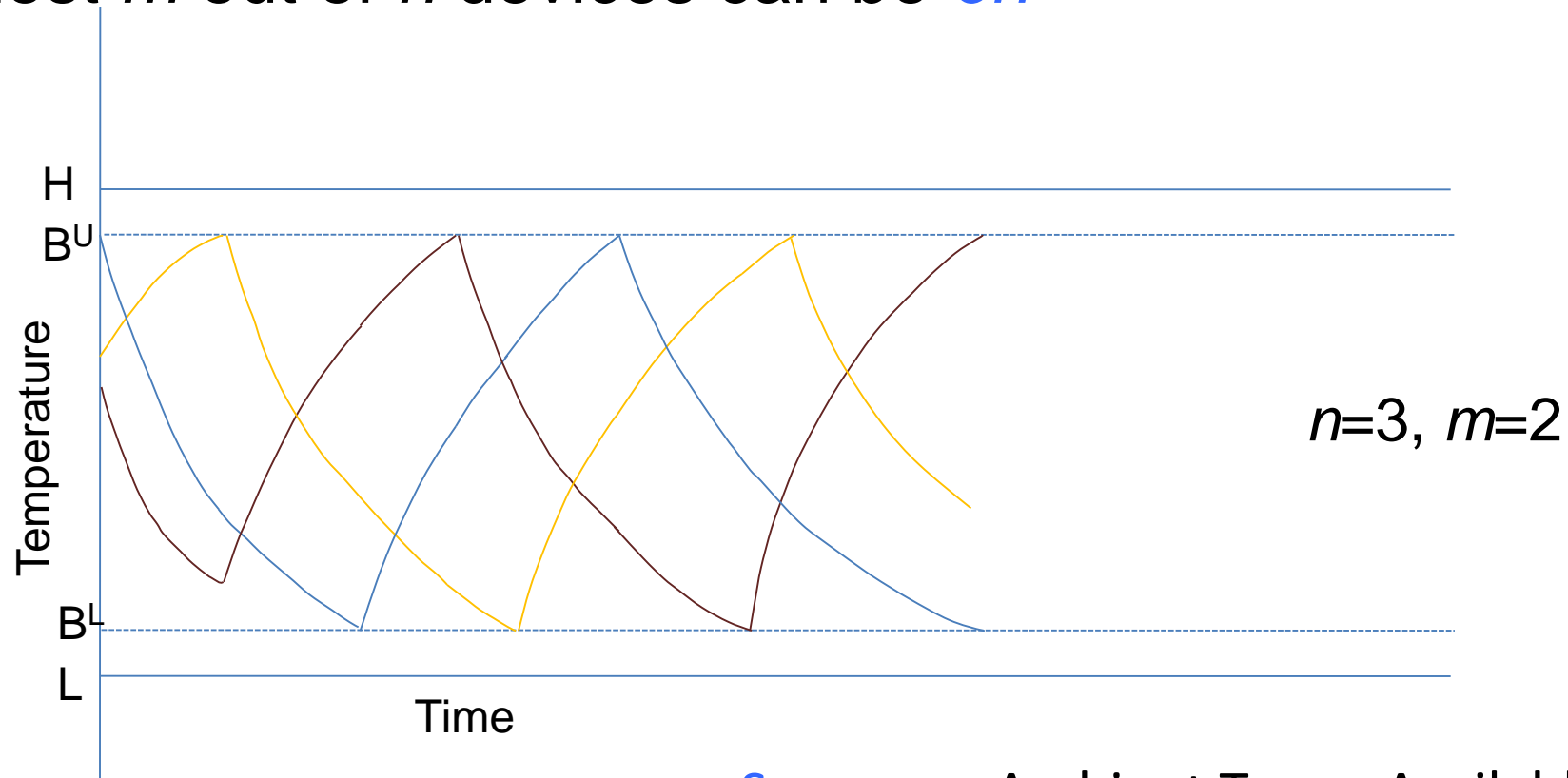


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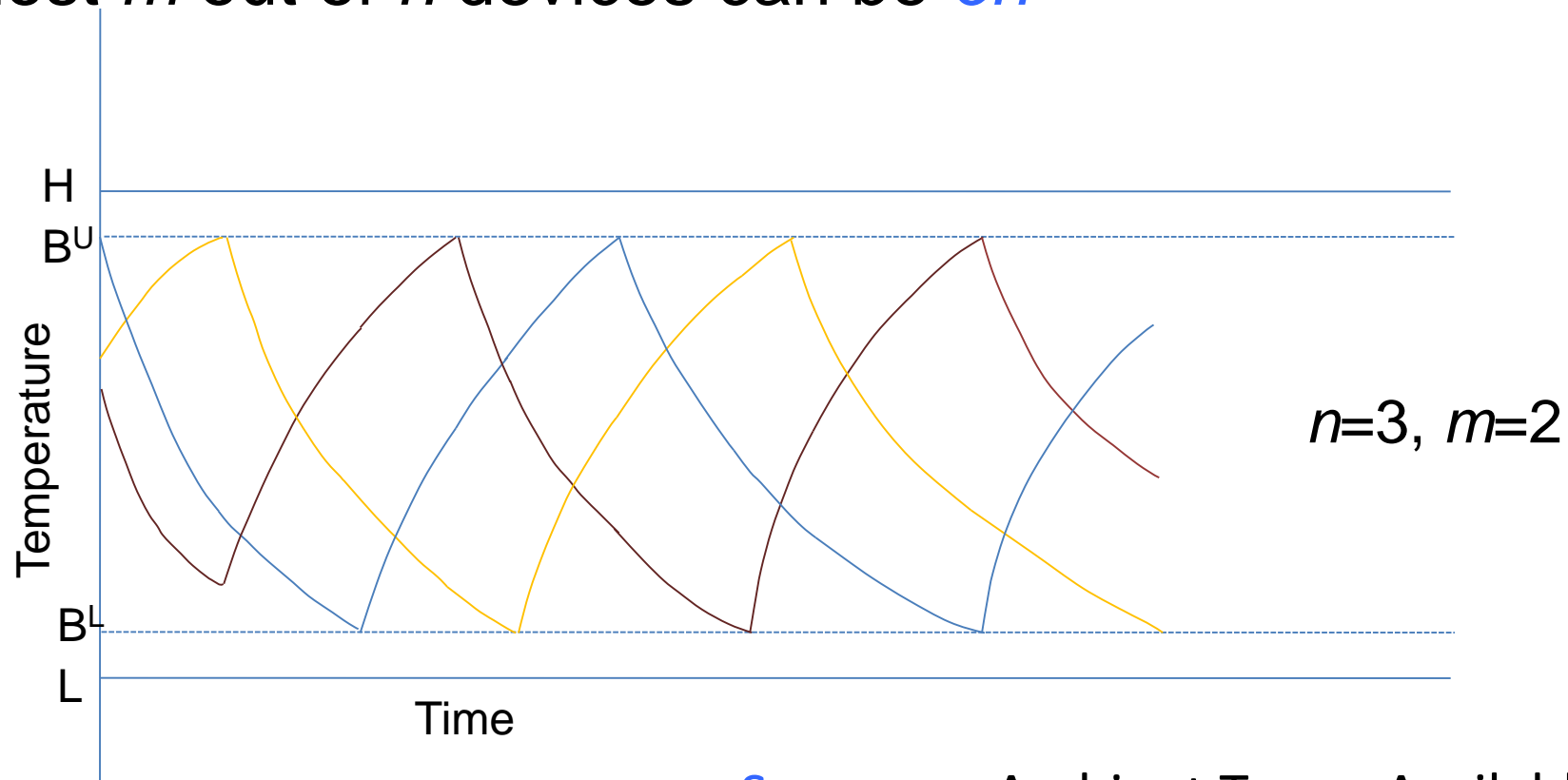


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Thermal Comfort Band Maintenance Algorithm

Peak energy constraint:

At most m out of n devices can be *on*



Sense	Ambient Temp, Available Peak Power
Analyze	Schedulability
Respond	Set m , Temp Band

Feasibility Analysis

Can the given band [L, H] to be maintained
by running at most ' m ' ACs at a time

-- using the characteristics of each of the ACs.

$$m \times \text{abs}(S_f) \geq (n - m) \times S_r$$

$$S_f = \min(S_f^i \mid i = 1, 2 \dots n)$$

$$S_r = \max(S_r^i \mid i = 1, 2 \dots n) \text{ at } T_i = B^U$$

Analysis:

Is Aggregate cooling
greater than
aggregate warming

S_f - slope of cooling curve

S_r - slope of warming curve

Demand-Response Management

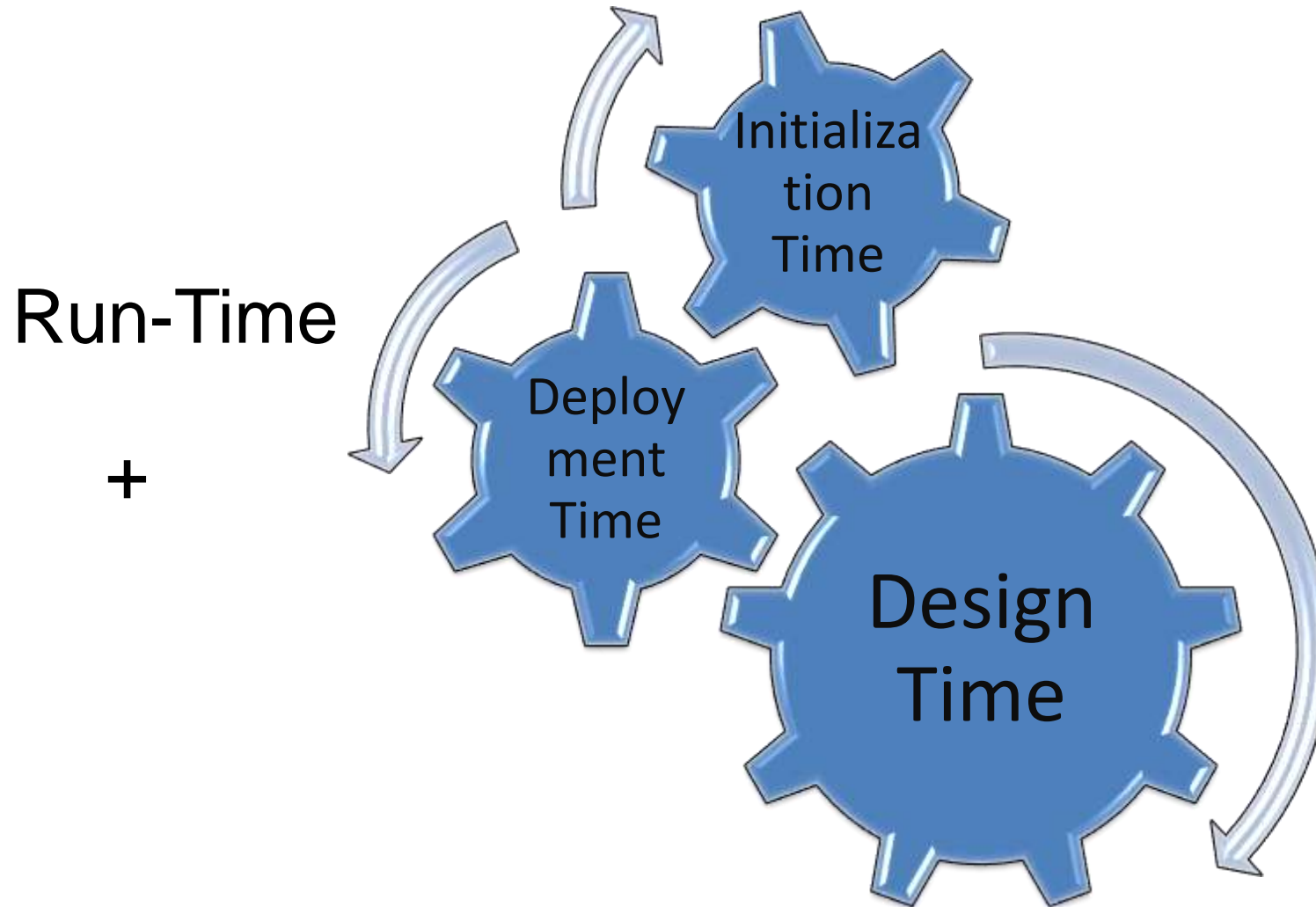
<i>Comfort band</i>	<i>m</i>	<i>Feasible?</i>
23-25	3	Yes
23-25	2	No
24-26	2	Yes

Improved
Schedulability
by shifting
comfort-band
(5 ACs)

		Ambient 27 ⁰ C			Ambient 30 ⁰ C		
Comfort-band	m	ΣS_f	ΣS_r	Feasibility	ΣS_f	ΣS_r	Feasibility
23 – 25	2	0.149	0.140	Yes	0.35	0.53	No
24 – 26	2	0.218	0.098	Yes	0.46	0.35	Yes

Reinitialization Time

Smartness at all “TIMES”



Towards Sustainable Energy

Reduce need for unsustainable energy sources

- Reduce unnecessary consumption
- Flatten peaks in consumption

Increase reliance on sustainable energy sources

- Exploit renewables
- Store excess energy

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Need
SMART
Approaches

Using Stored/Renewable Energy

- batteries in home settings:
peak flattening, reduced costs,
smoothed intermittent sources,
surviving grid outages

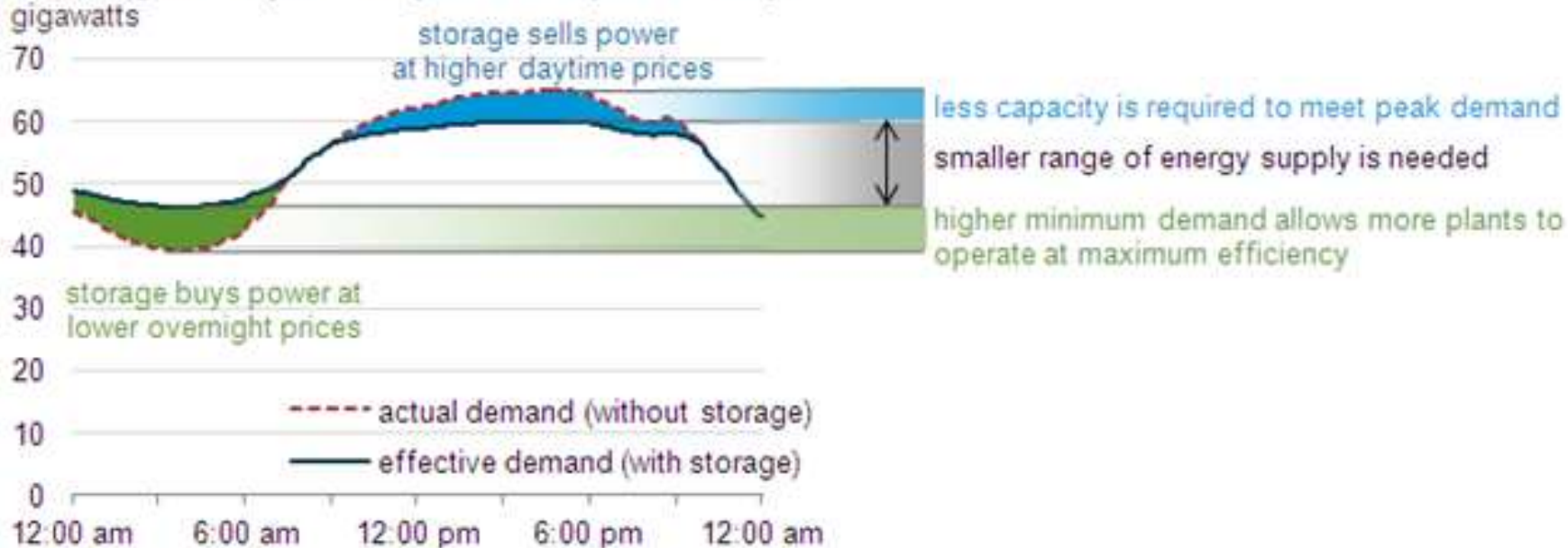


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Flattening the daily load shape, 24-hour period example

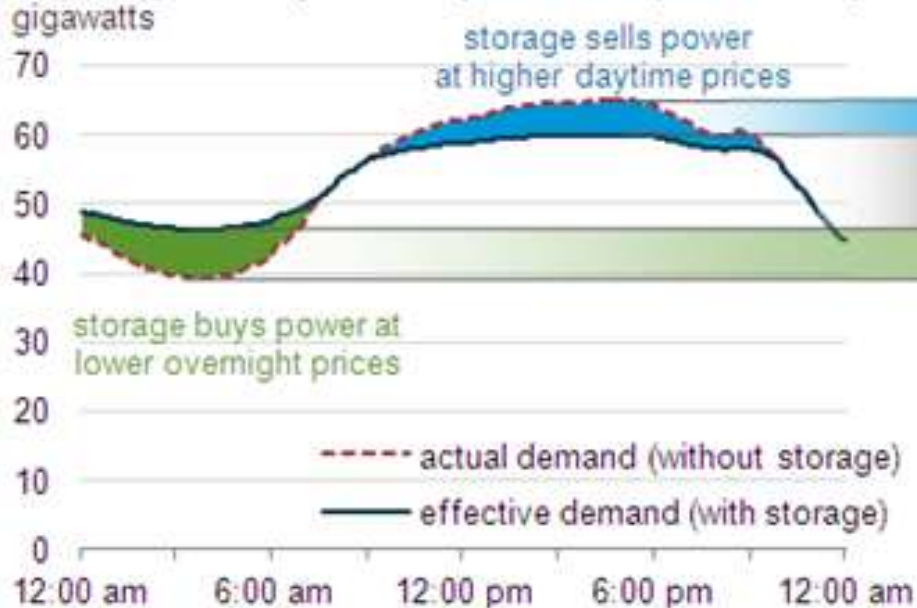


Using Stored/Renewable Energy

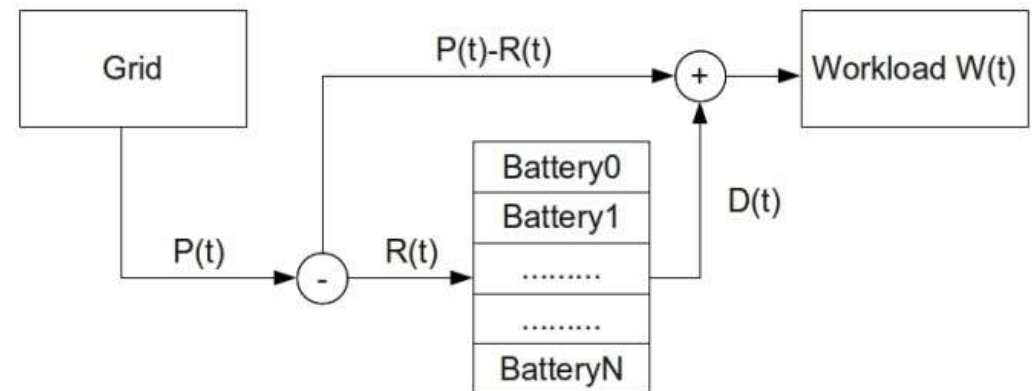
- batteries in home settings:
peak flattening, reduced costs,
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Flattening the daily load shape, 24-hour period example



less capacity is required to meet peak demand
smaller range of energy supply is needed
higher minimum demand allows more plants to operate at maximum efficiency



Towards Sustainable Energy

Reduce need for unsustainable energy sources

- Reduce unnecessary consumption
- Flatten peaks

Use energy efficient appliances

Turn off devices when not needed

Set comfort levels smartly

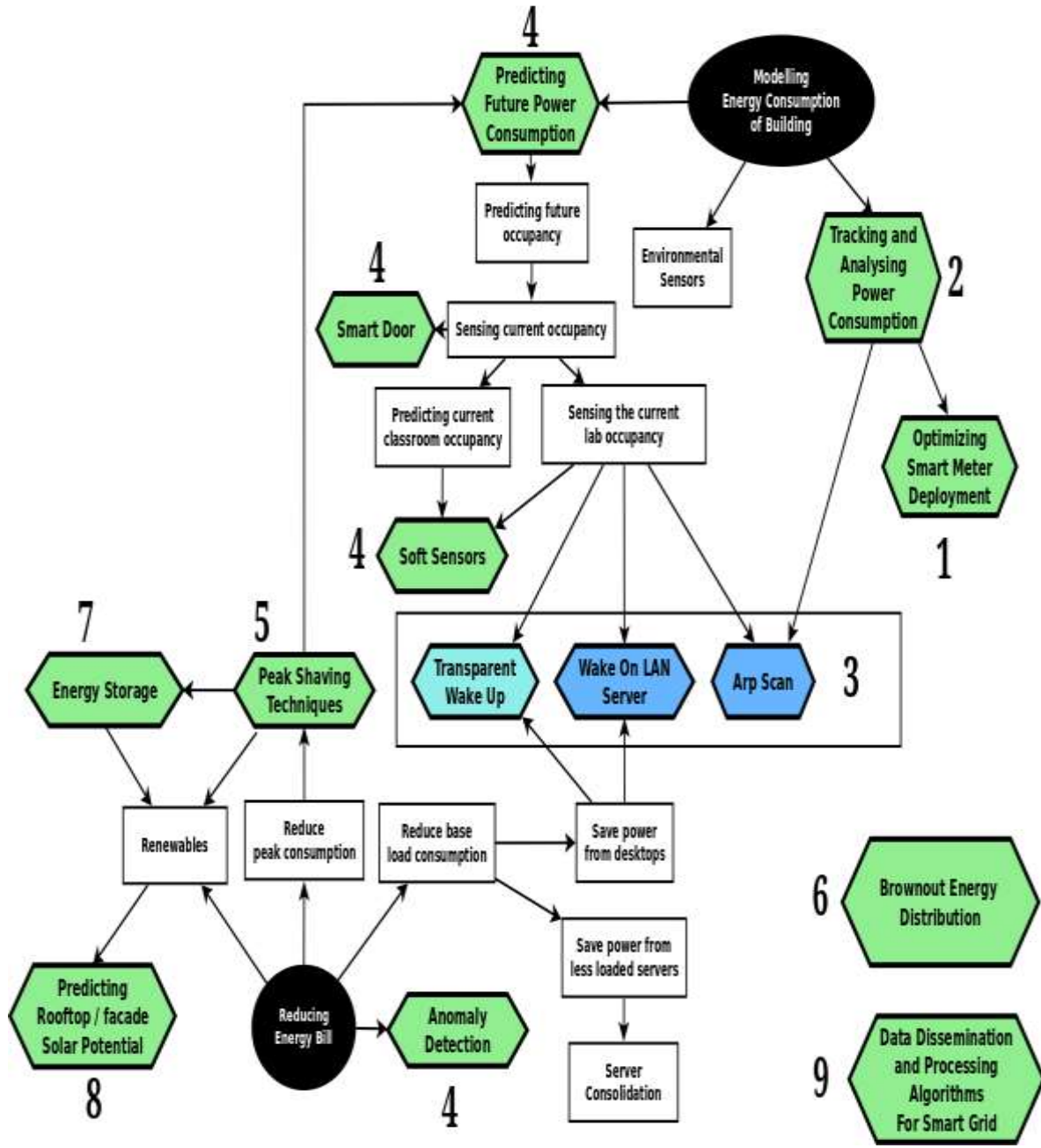
Increase reliance on sustainable energy sources

- Exploit renewables
- Store excess energy during off-peak – avoid blackouts

Exploit elasticity of appliance operations

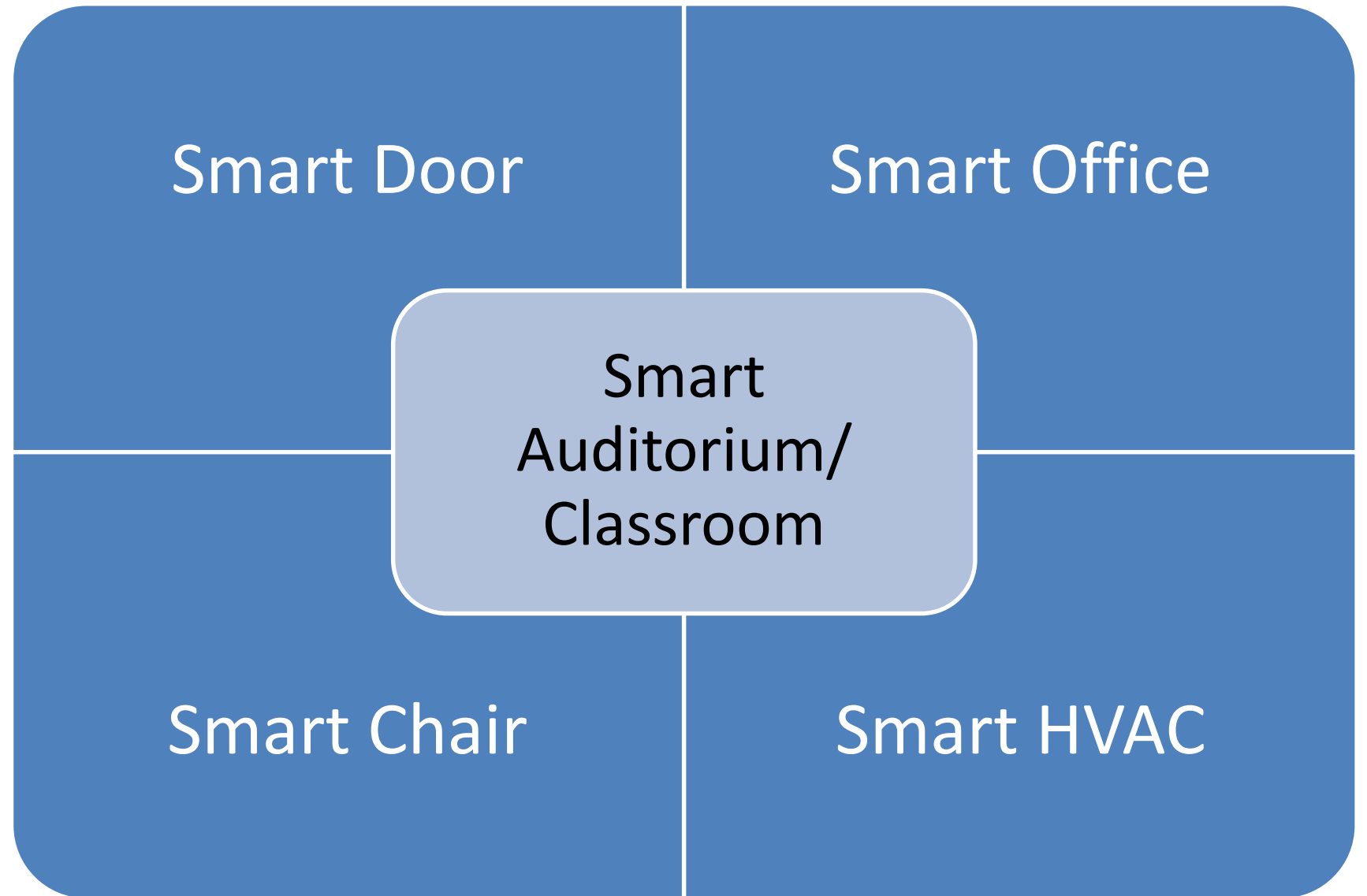
Store excess energy

Our Work In Perspective

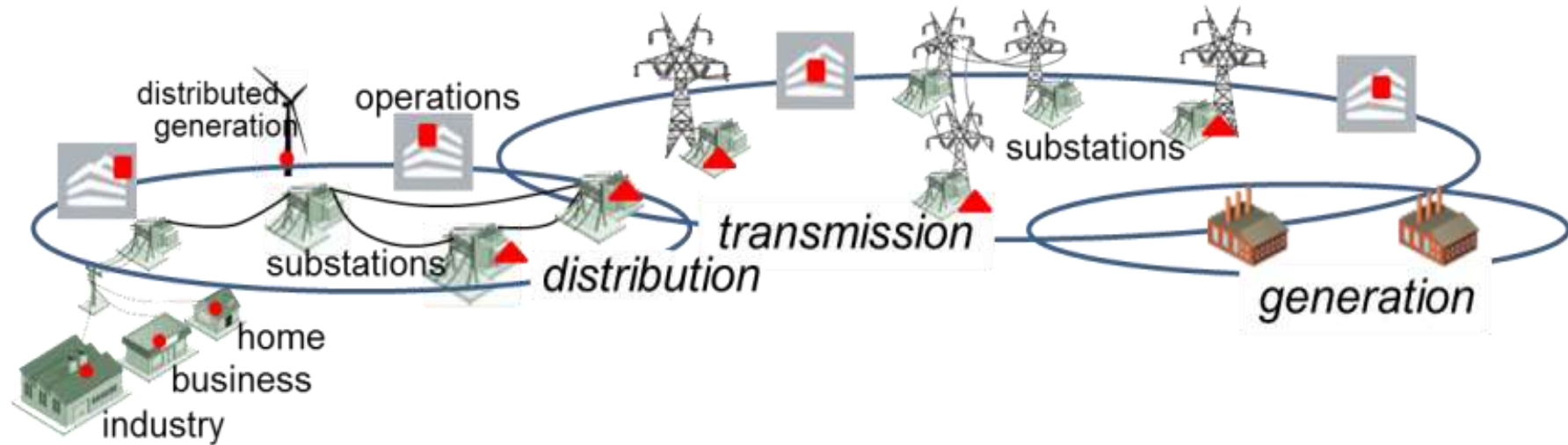


1.	Locating and Sizing Smart Meter Deployment in Buildings
2.	Smart Meter Data Analysis System
3.	Increasing Predictability and Minimizing Power Consumption
4.	Demand Forecasting using Occupancy Patterns
5.	Thermal Comfort-Band Maintenance and Adaptive Demand-Response Control
6.	Brownout Energy Distribution: Graceful Degradation of QoS in Smart Grids
7.	Energy Cost Minimization with SmartStore under Demand Based Static Pricing Schemes
8.	Solar Intensity Prediction
9.	Scaling up the capability of Smart Grids: Distributed Stream Computing
10.	Robust Strategies for PMU Placement

Smart buildings
Smart campuses
Smart grids
Smart cities

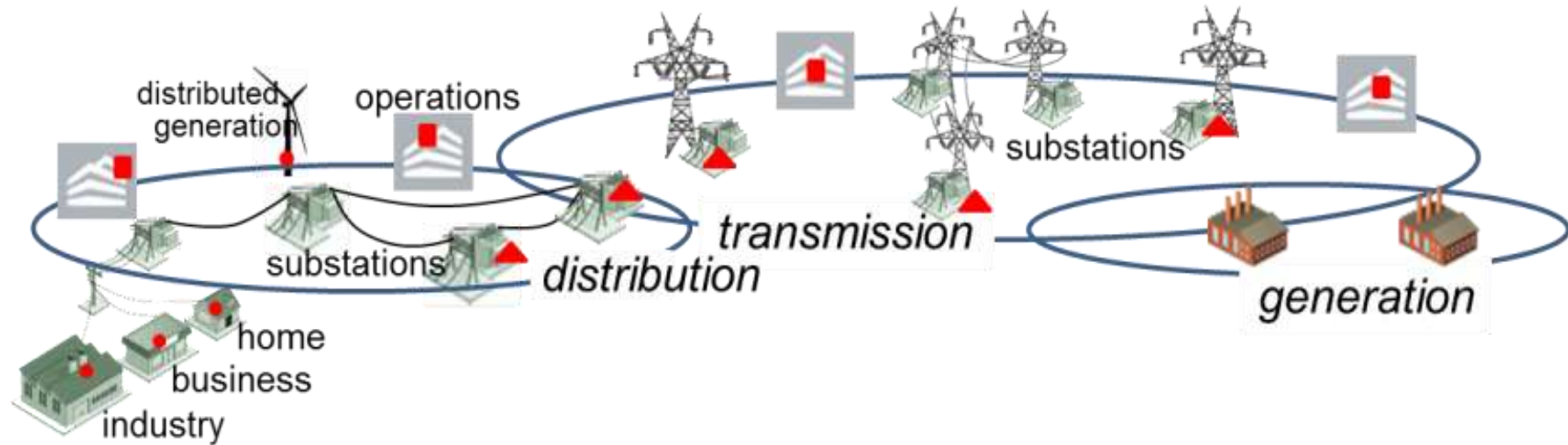


Smart Building/Home: Part of Smart Grid



Smart Building/Home: Part of Smart Grid

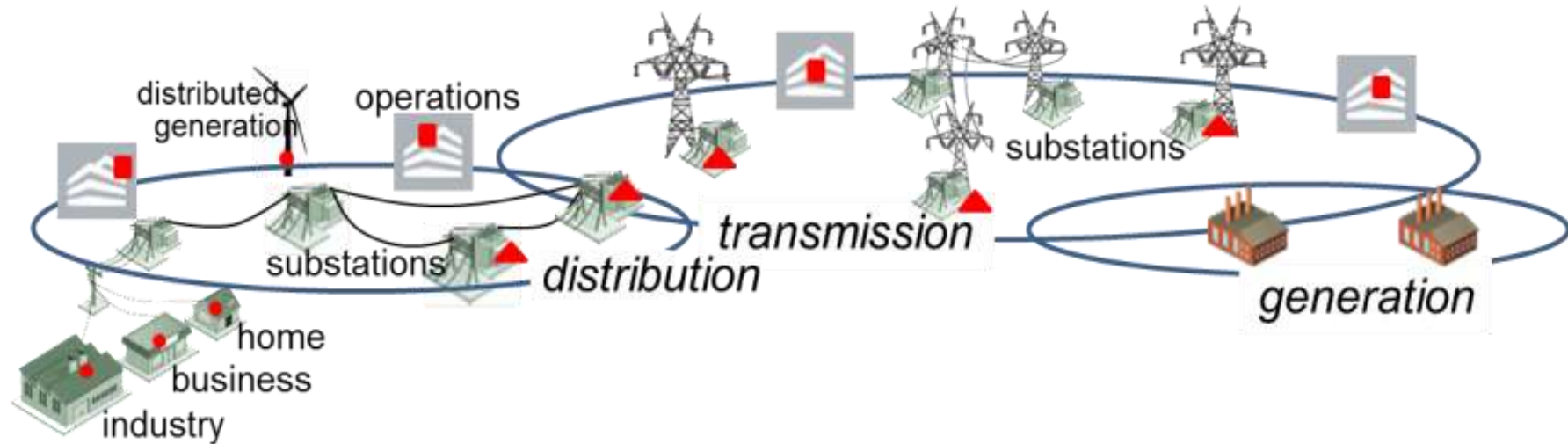
Grid monitoring: PMU placement, data distribution, event detection, state estimation



Smart Building/Home: Part of Smart Grid

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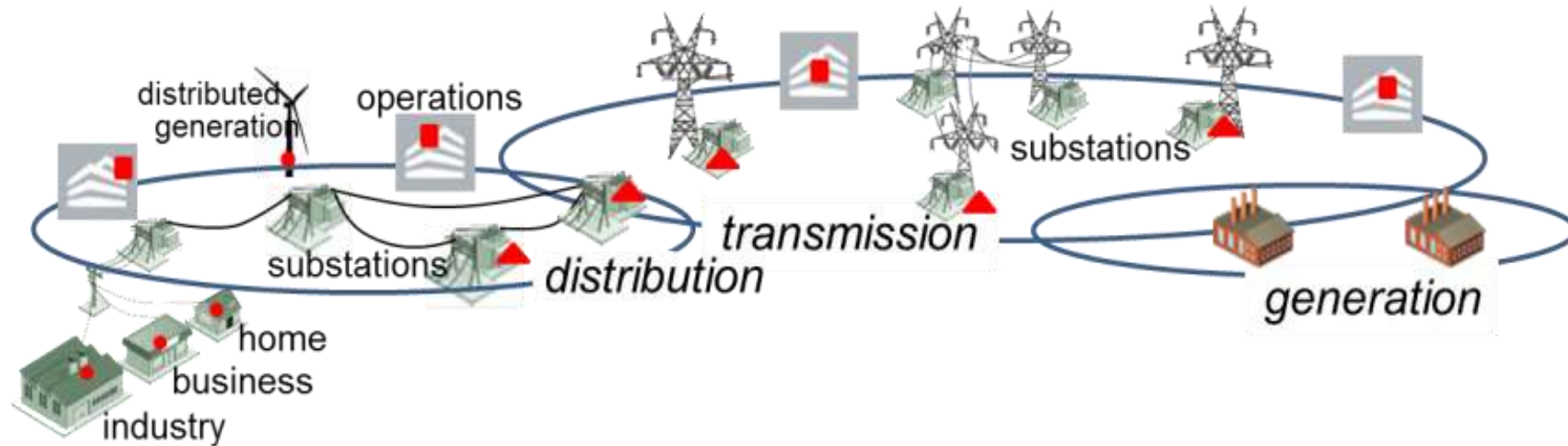
Prediction, modeling: intermittent, renewable source prediction, consumption models



Smart Building/Home: Part of Smart Grid

Grid monitoring: PMU placement, data distribution, event detection, state estimation

Prediction, modeling: intermittent, renewable source prediction, consumption models

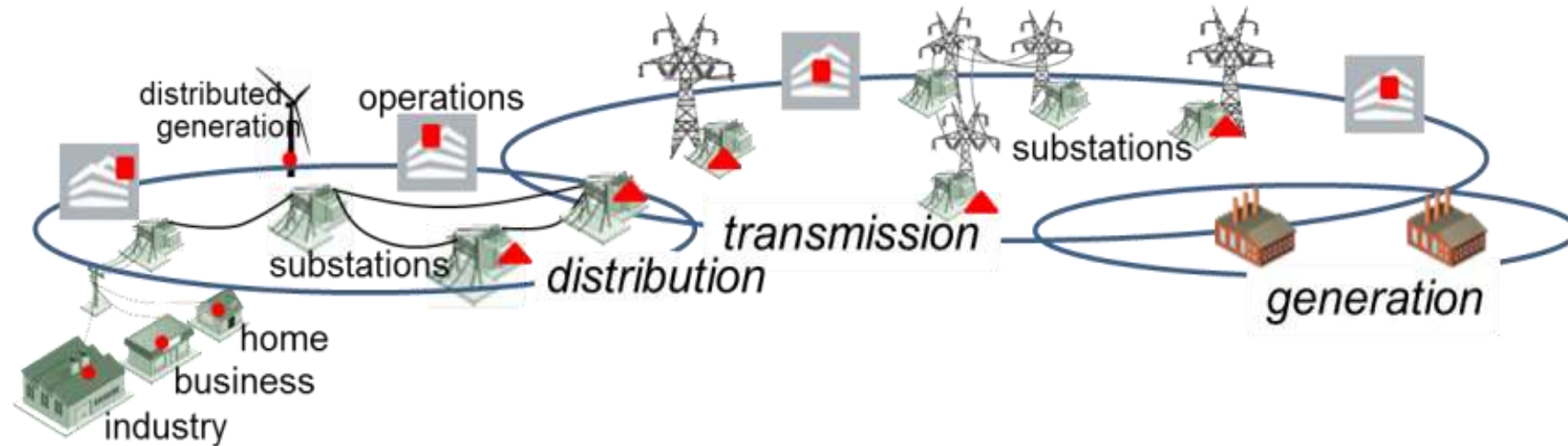


Energy Storage: battery use for peak smoothing, intermittent source buffering

Smart Building/Home: Part of Smart Grid

Grid monitoring: PMU placement, data distribution, event detection, state estimation

Prediction, modeling: intermittent, renewable source prediction, consumption models



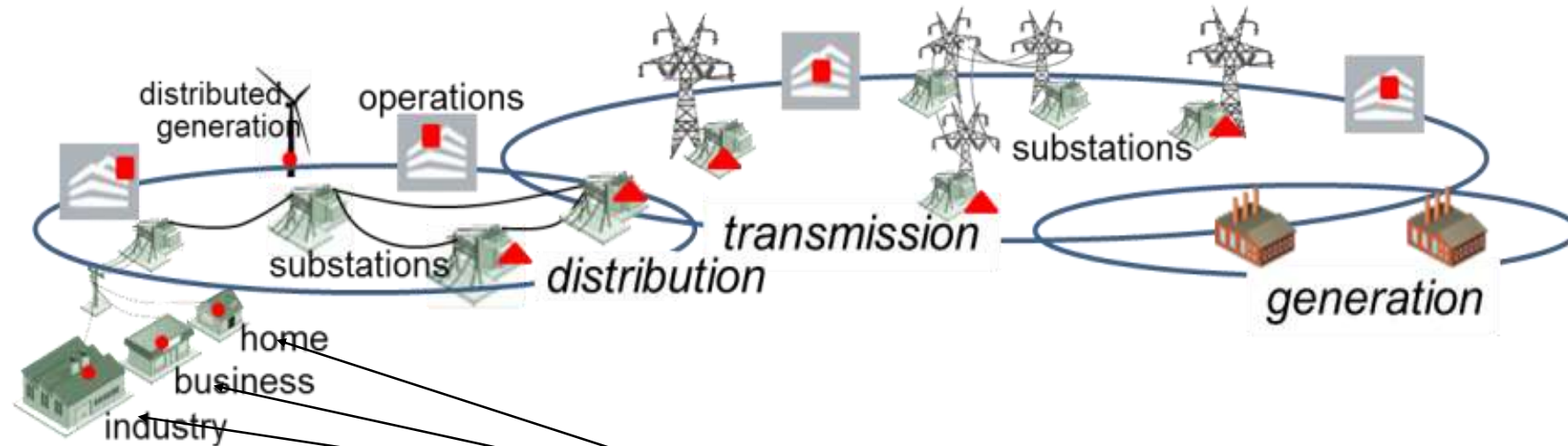
Energy Storage: battery use for peak smoothing, intermittent source buffering

Demand-side energy management: deferring elastic loads, **coordinating background loads**

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Energy Storage: battery use for peak smoothing, intermittent source buffering

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Some Closing Thoughts...

Any time we answer a question or make a decision,
our response reflects our smartness

Smartness at all stages

--*Design Time, Deployment Time, Initialization Time, Run Time*

-- it is continuous feedback-driven, resource and time constrained

There is always a smarter way of doing something...

=> Today's smarter than yesterday's

=> Tomorrow's will be smarter than today's

Thorny questions -- Intrusion, privacy, security

Cost-effectiveness, payback period

High Tech should be used to achieve Low Cost

All Science and Engineering disciplines have a major role

Different topics in CS/EE contributing to smart things

- **AI** – reasoning, analytics, machine learning, data fusion, forecasting, planning, reinforcement learning
- **Communication and networking, distributed systems** -- light weight protocols, timely data delivery
- **Embedded systems** -- low energy / footprint devices, sensors
- **Autonomic computing**
- **Algorithms** -- optimization, data analysis, state estimation
- **Software engineering** -- v & v, embedded sw
- **Databases** -- *storage, storage and retrieval query processing/optimization, stream processing*
- **Operating systems** – caching prefetching, scheduling
-

Thanks a lot

Thanks to my collaborators