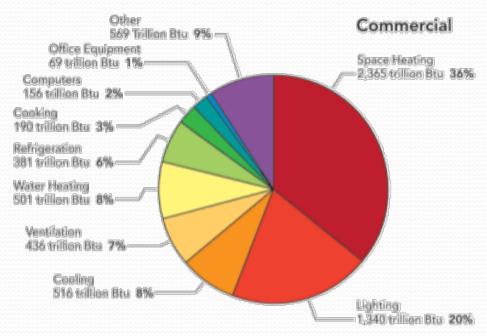
# SPOT\* A Smart Personalized Office Thermal Control System

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## **HVAC** Energy use

- Buildings use 1/3 of all energy
- 30-50% of building energy is for HVAC



Percentage of energy use for commercial facilities (EIA)

# Can we improve efficiency?

- Yes!
- Change temperature setpoint:
  - 1°C higher when cooling ≈ 10% saving
  - 1°C lower when heating ≈ 2-3% saving
- But this can reduce comfort

## Focus of this work

How to keep office workers comfortable while reducing energy use?

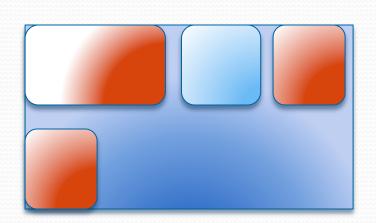


# Option 1: Tweak Status Quo

- Problem: centralized control and management
  - ignores workers
    - or has minimal input from them
  - invades privacy



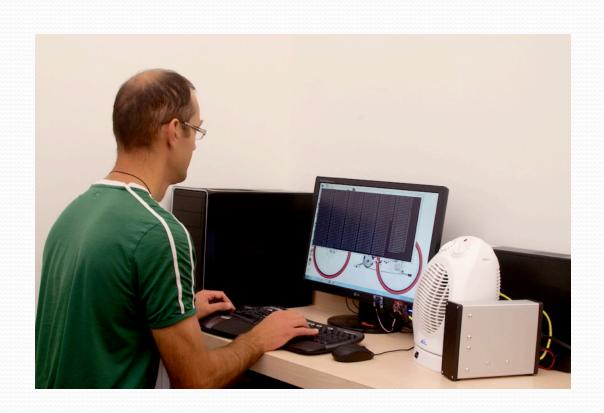




Our insight



# Option 2: Decentralize!



# Temperature or comfort?

Most current systems maintain temperature Why not control comfort instead?



# Comfort according to ASHRAE

Cold	Cool	Slightly Cool	Neutral	Slightly Warm	Warm	Hot
-3	-2	-1	O	1	2	3

# ...can be predicted!

- Predicted Mean Vote (PMV) model (ISO 7730)
- Six input parameters
  - Air Temperature
  - Background Radiation
  - Air Velocity
  - Humidity
  - Metabolic Rate
  - Clothing Level



- Developed by P.O. Fanger in 1970
  - Extensively validated

## ... and personalized

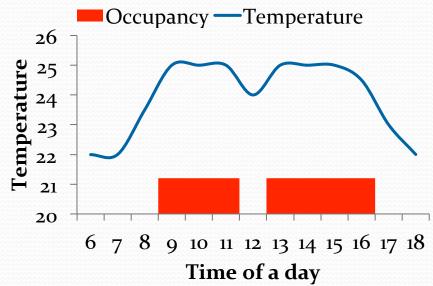
- PMV model represents the average
  - for a single office, only the occupant's vote matters
- Predicted Personal Vote (PPV) Model

$$ppv = f_{ppv} (pmv)$$

where  $f_{ppv}(\cdot)$  is a linear function

### Our idea in a nutshell

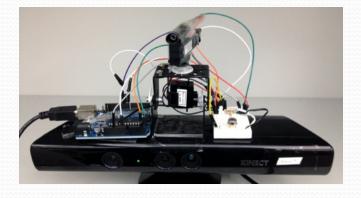
- Measure occupancy and comfort
- When occupied, keep office comfort at the minimum acceptable level
- When vacant, turn heating/fan off
  - but pre-heat if needed



## Two systems

#### **SPOT** (2011-2013)

- Extreme sensing
- In office only
- Expensive
- Reactive or pro-active



#### **SPOT\*** (2014)

- Minimal sensing
- Flexible location of functionality
- Low-cost
- Reactive



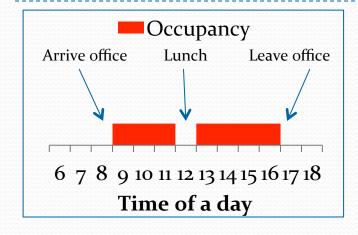
## Mathematical basis

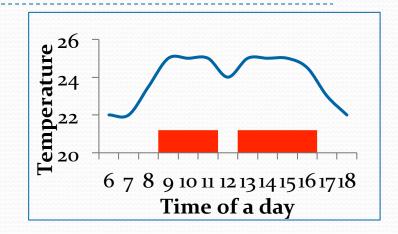
Personal Thermal Comfort Evaluation

Learning-Based Modeling

Cold	Cool	Slightly Cool	Neutral	Slightly Warm	War m	Hot
-3	-2	-1	0	1	2	3







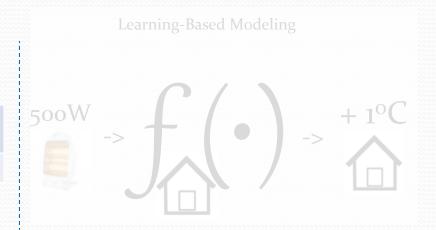
**Occupancy Prediction** 

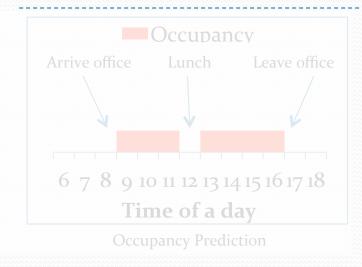
Optimal control

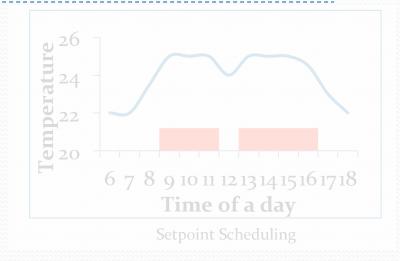
## Mathematical basis

#### Personal Thermal Comfort Evaluation

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# **Monitoring PMV**

Air Temperature	Measured by sensor
Background Infrared Radiation	Measured by sensor
Air Velocity	Measured by sensor
Humidity	Measured by sensor
Metabolic Rate	Constant for indoor activity
Clothing Level	Estimated



## SPOT+: extreme sensing

#### 5° infrared sensor:

 Detects users' clothing surface temperature

#### Microsoft Kinect:

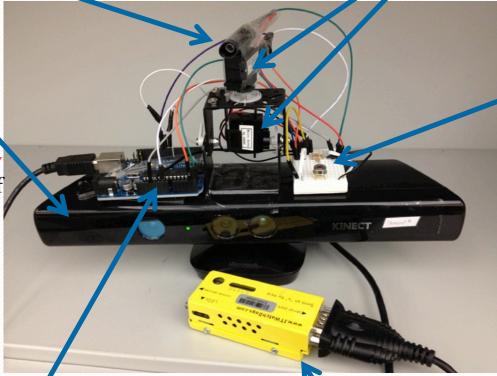
- Detects occupancy
- Detects location of the user

#### Servos:

 Controls the direction of the 5° infrared sensor

#### 90° infrared sensor:

 Detects background radiant temperature



#### Microcontroller:

- Pull data from the sensors
- Control the rotation angle of the servos

#### Weatherduck sensor:

 Detects air temperature, humidity, air velocity

# SPOT\*: minimal sensing

#### Occupancy/temperature sensor

Passive IR



## Clothing level estimation (SPOT+)

- Estimate clothing by measuring emitted infrared
  - More clothing => lower infrared reading

$$Clo = k * (t_{clothing} - t_{background}) + b$$

- $t_{clothing}$  is the infrared measured from human body
- $t_{background}$  is the background infrared radiation
- *k* and *b* are estimated by regression

# Learning PPV

- Training phase
  - SPOT knows PMV
  - Occupant votes periodically = PPV
  - PMV -> PPV relation learnt by linear regression



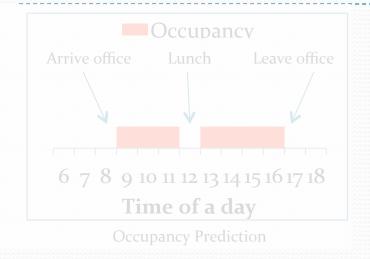
## Mathematical basis

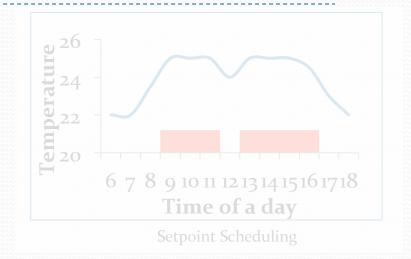
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# Forecasting comfort

- Heating: Learning-Based Predictive Control (LBMPC)
  predicts the temperature given heating minutes
  - plug into PMV equation
- Cooling: fan speed lowers perceived temperature
  - plug into PMV equation

## Thermal model

rate of thermal loss

net heat input 
$$P_{loss} = k(T_{in} - T_{out})$$

$$P = eP_{hvac} - P_{loss} = eP_{hvac} - k(T_{in} - T_{out})$$

$$P = \frac{dQ}{dt} = C\frac{dT_{in}}{dt}$$

$$\frac{dT_{in}}{dt} = \frac{eP_{hvac} - k(T_{in} - T_{out})}{C}$$

### Discrete time model

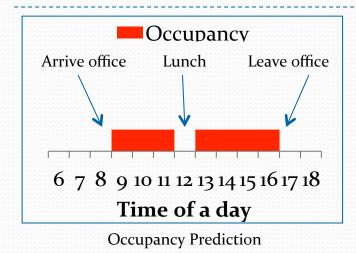
$$T_{in}(s+1) = T_{in}(s) + \frac{eP_{hvac}(s) - k(T_{in}(s) - T_{out}(s))}{C}$$

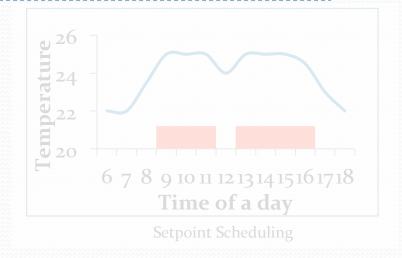
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Personal Thermal Comfort Evaluation

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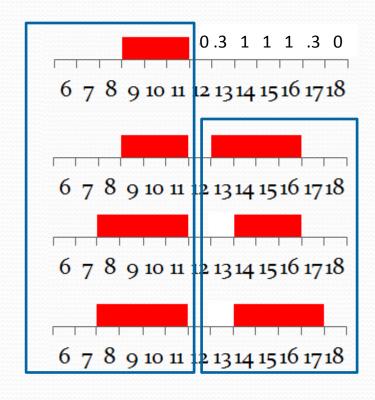




# **Occupancy Prediction**

- Predict occupancy using historical data
  - (only needed for pre-heating)

Match Previous similar history

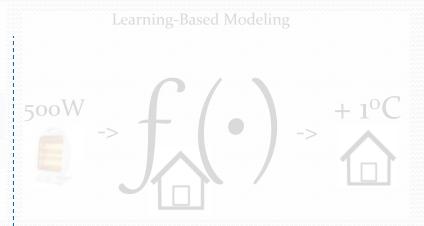


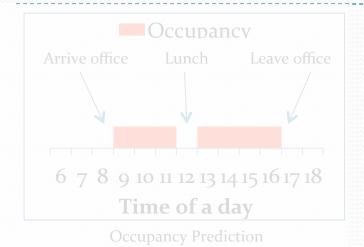
Predict using matched records

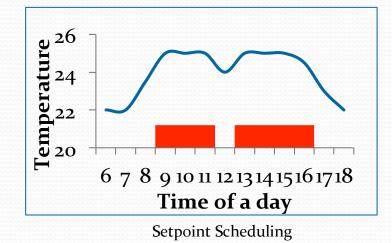
## Mathematical basis

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#### Reactive control

- When occupant is present
  - if comfort is not in [-0.5, 0.5]
  - either heater + max fan
  - or fan speed control

## **Optimal Control**

Forecasts occupancy as well as effect of heating

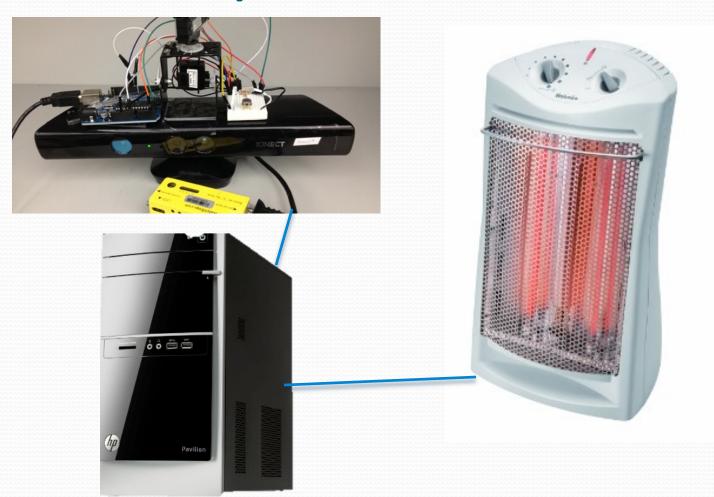
$$\min \sum_{s=1}^{S} P_{hvac}(s) + \lambda \sum_{s=1}^{S} m(s)(\beta_c(s) + \beta_h(s))$$

Overall energy consumption in the optimization horizon S

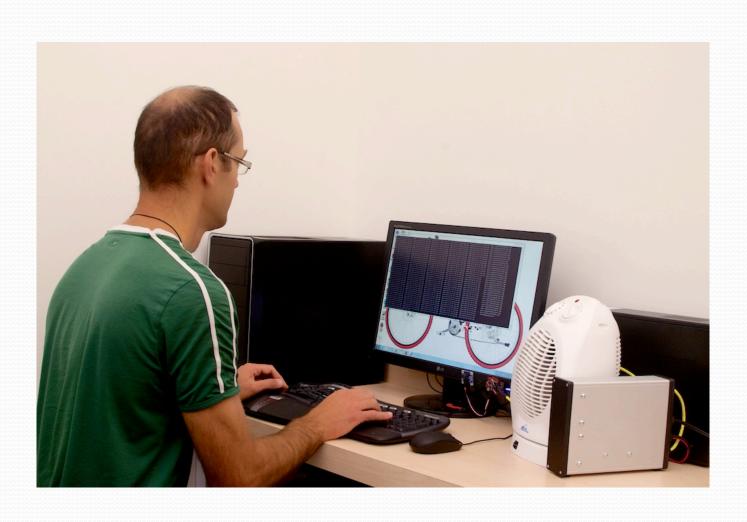
Weight of comfort, set to large value to guarantee comfort first Predicted occupancy, we only guarantee comfort when occupied i.e., when m(s) = 1

Thermal comfort penalty. Both term equal o when the user feels comfortable

# SPOT+ system



# SPOT\* system



# SPOT\* control

#### Display

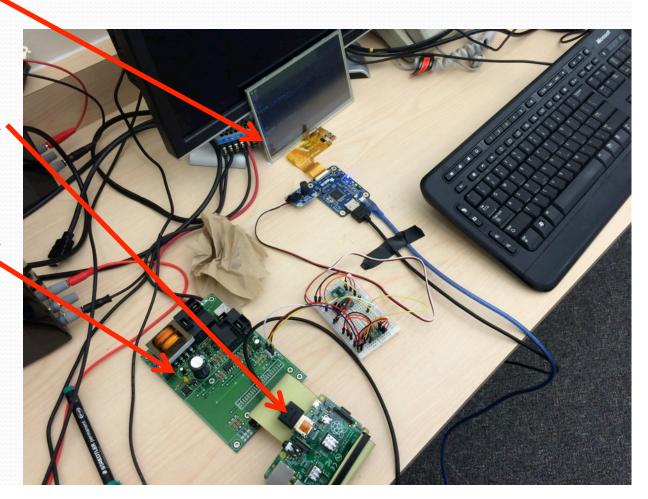
For local control

#### Raspberry Pi Microcontroller

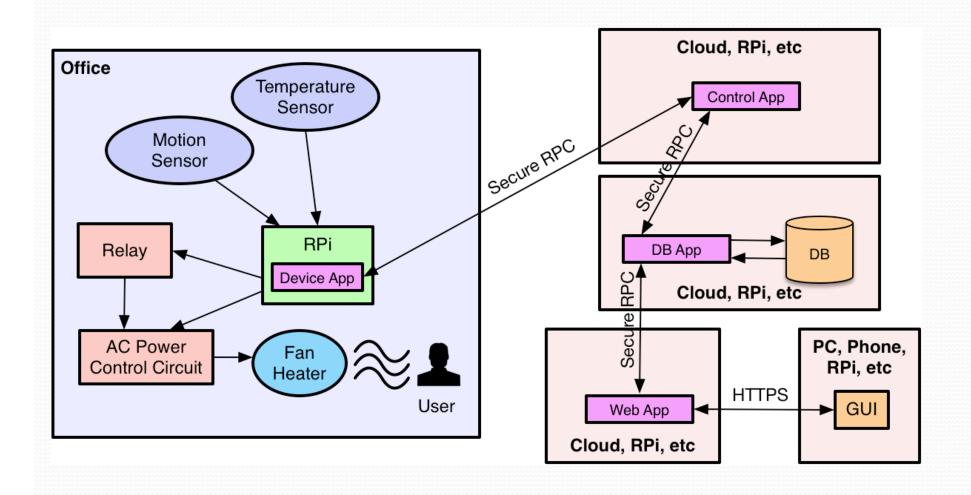
- Pulls data from the sensors
- Controls heat and fan speed

#### Fan speed controller

- Pulls data from the sensors
- Controls heat and fan speed



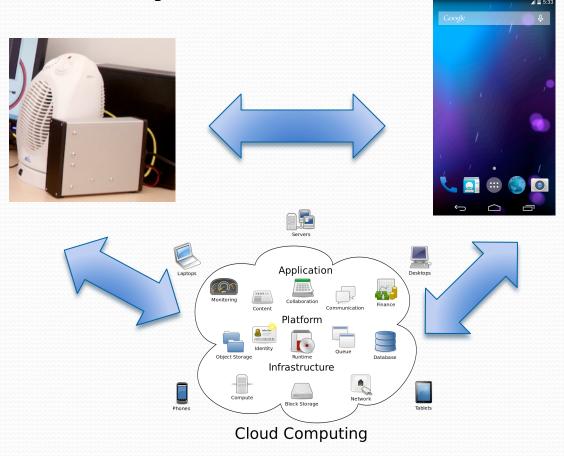
# SPOT\* system



# SPOT\* flexibility

Flexible location of functionality:

- sensing
- actuation
- user input
- data storage
- control logic

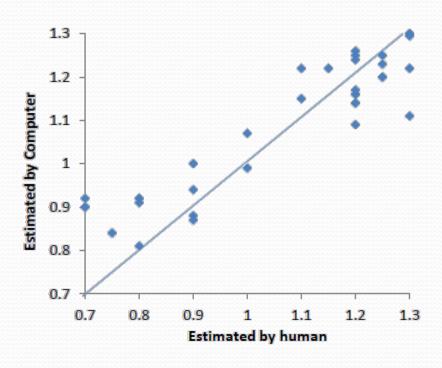


# Results



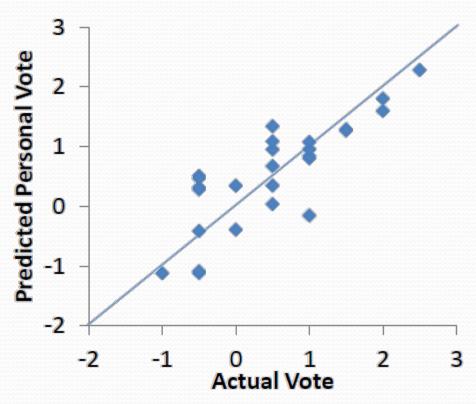
# Sensing accuracy

#### Clothing level estimation



- Root mean square error (RMSE) = 0.0918
- Linear correlation = 0.92

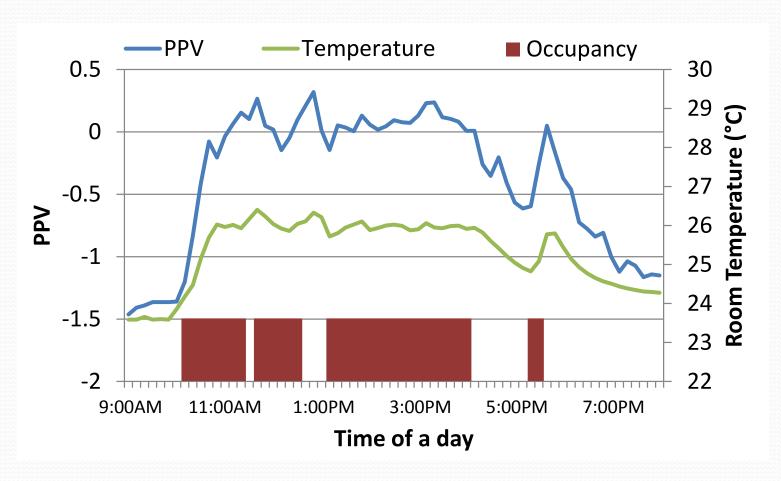
#### **PMV** Estimation



- Root mean square error (RMSE) = 0.5377
- Linear correlation = 0.8182

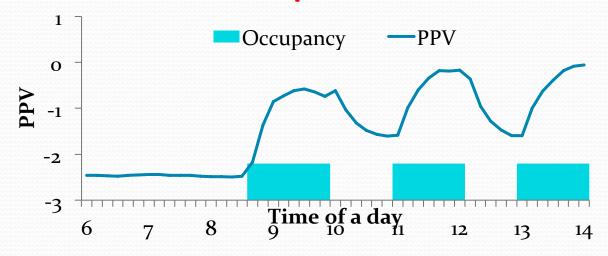
#### **Reactive Control**

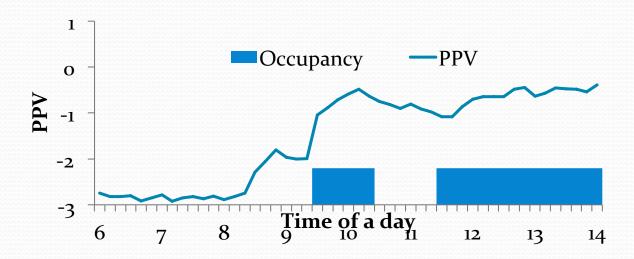
# Room temperature and PPV



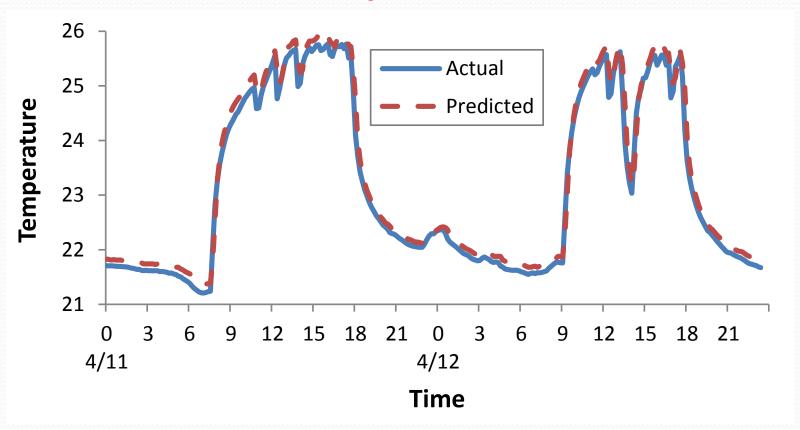
# **Optimal Control**

#### Reactive Control vs. Optimal Control



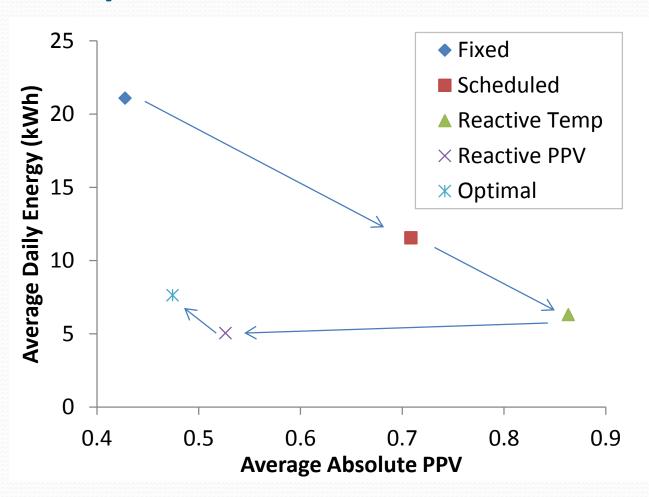


# **Accuracy of LBMPC**



• The RMSE over a day is 0.17C.

# Comparision of schemes



#### Conclusions

- We extended PMV model for personalized thermal control
- We design and implement SPOT+ and SPOT\*
- SPOT can accurately maintain personal comfort despite environmental fluctuations
- Legacy and privacy compatible