

Gesture Recognition Using Ambient Light

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Motivation

- Gesture Recognition enables various interactive applications.



Gaming



Health Care



Smart Homes



AR

- Multiple Modalities



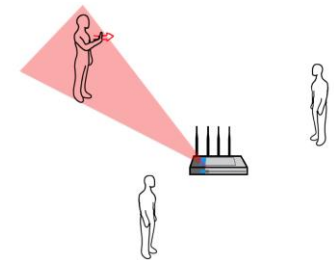
Wearables



Sound



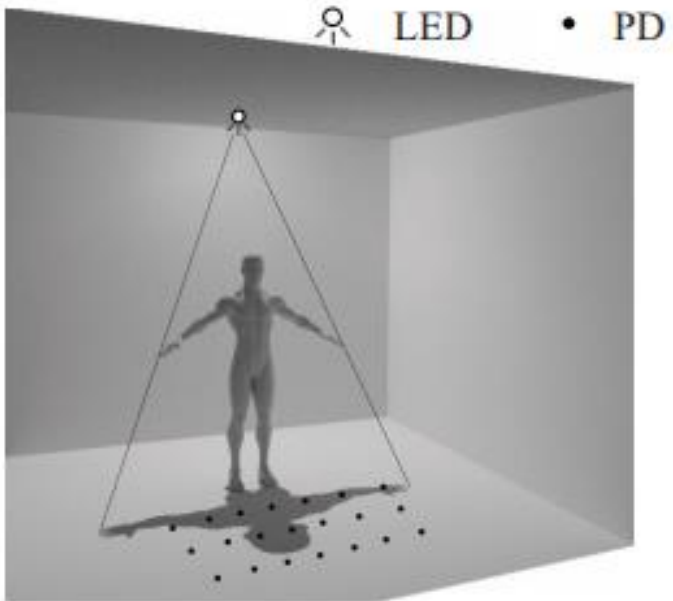
Vision/IR



RF

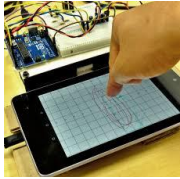
Motivation

- Gesture Recognition using Ambient Light Signals

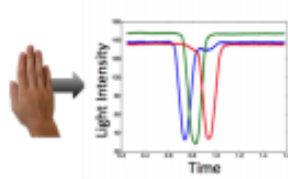


- **Ubiquitous:**
Light Sources are available everywhere
- **Non-invasive:**
Movements can be sensed from shadows
- **Preserve Privacy:**
Signals do not leak through walls

Existing Approaches

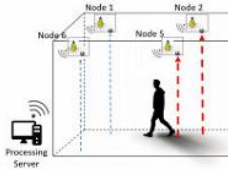


Okuli
(MobiCom '15)

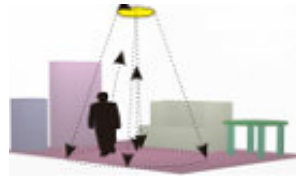


GestureLite
(DTR '16)

Limited Range
($< 30\text{cm}$)



VLAS
(VLCS '16)



CeilingSee
(PerCom '16)

Limited Resolution
(Room-Level Semantics)



LiSense
(MobiCom '15)



StarLight
(MobiSys '16)

Active Sensing :

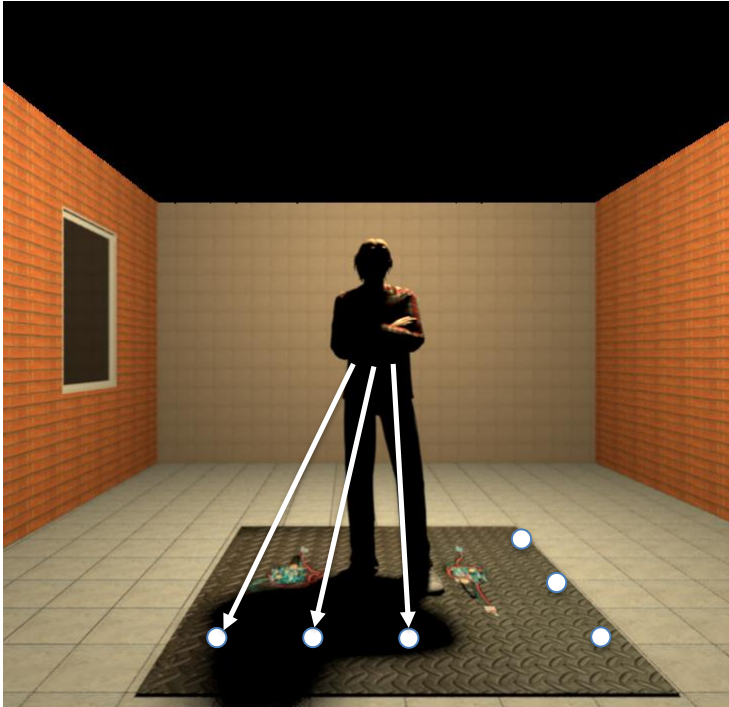
Controlled lighting infrastructure
(Modulated LED lights)

Problem Statement

Design a passive, *ambient* light based gesture recognition system

- Unmodulated Light Sources
- Agnostic to lighting conditions
- Agnostic to user's position and orientation
- Recognize gestures of any given user

Approach

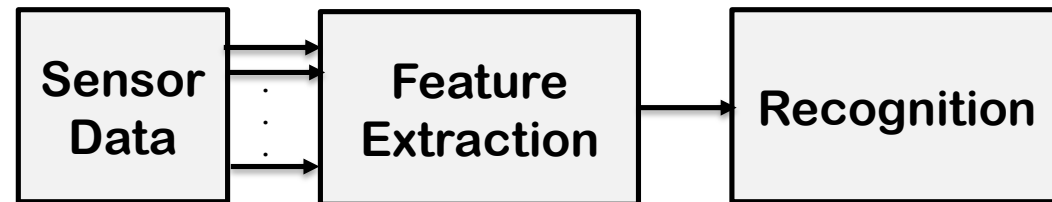


Observation:

- Shadows follow movements
- Different gestures create distinct shadow patterns on the floor

Idea:

- Instrument floor to learn shadow patterns using ML models and infer gestures.

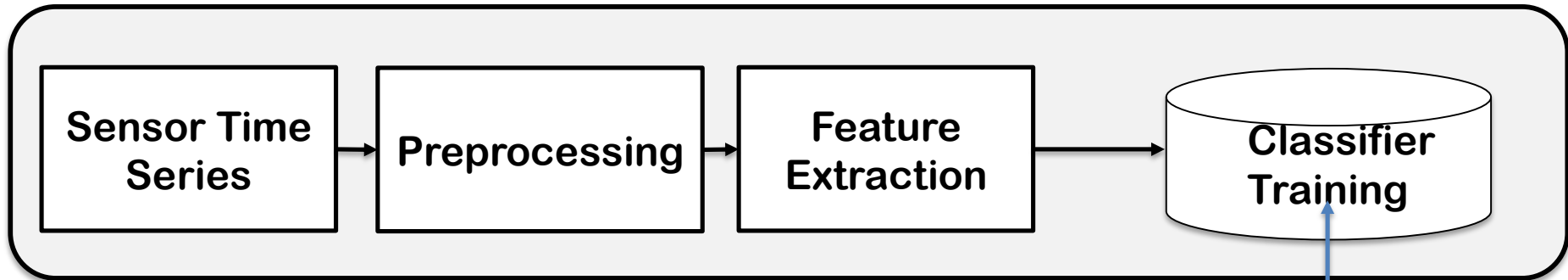


Challenges

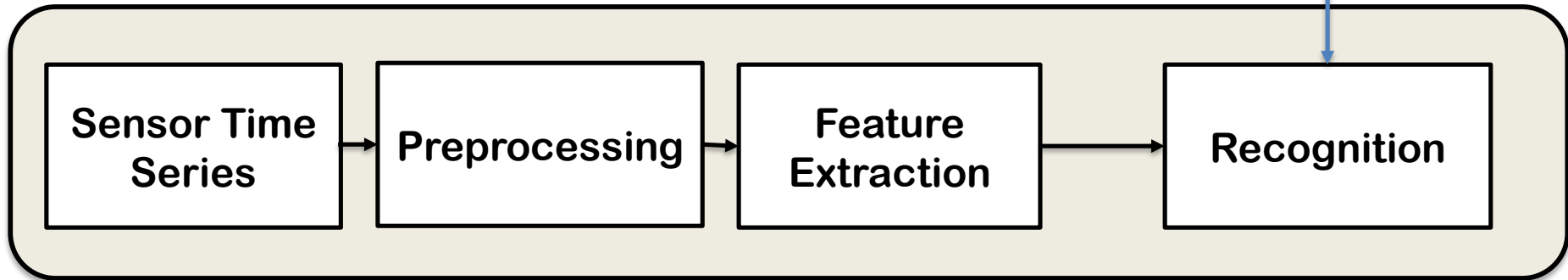
- Capturing features agnostic to different lighting conditions, user positions and orientations

Overview

Training



Runtime



I. Preprocessing

1. Denoising:

Challenge 1 : Separating signal from the noise

(i) Stray Shadows and Reflectors :

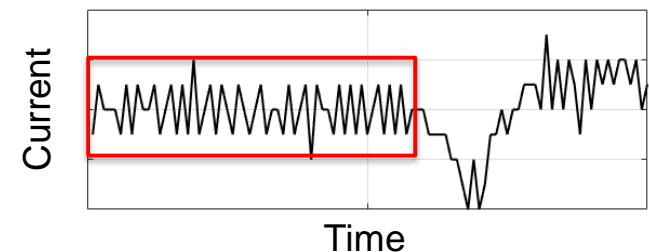
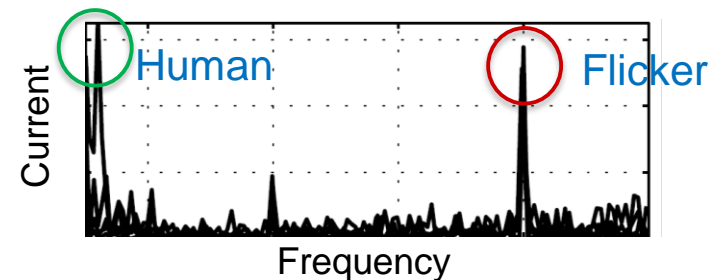
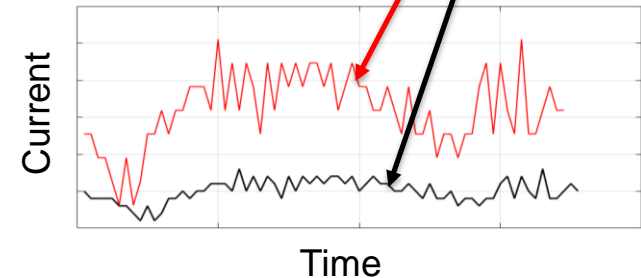
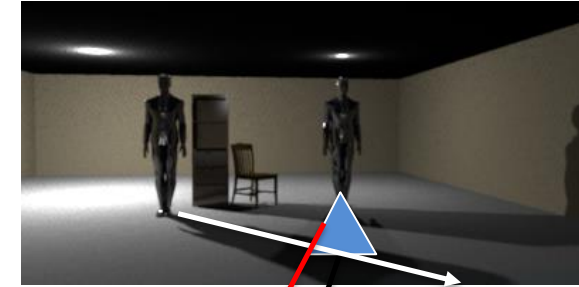
Varying photocurrent offsets with environmental changes.

(ii) Light Source Flicker (AC Powered):

Fluctuations of comparable magnitude
-Well localized in Frequency Domain

(iii) Shot noise

Spurious burst noises
-Well localized in Time Domain



I. Preprocessing

1. Denoising:

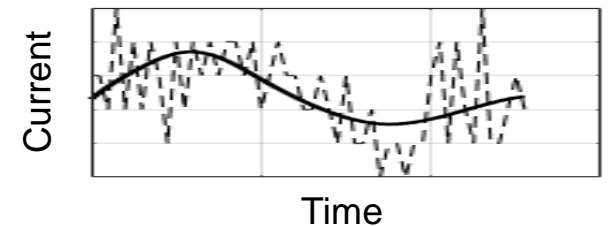
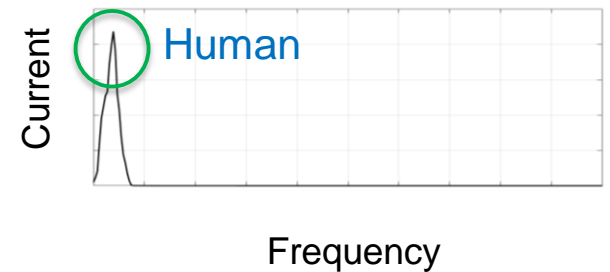
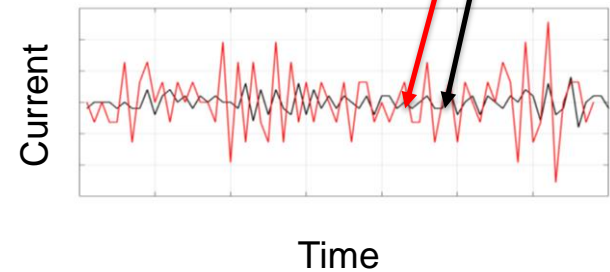
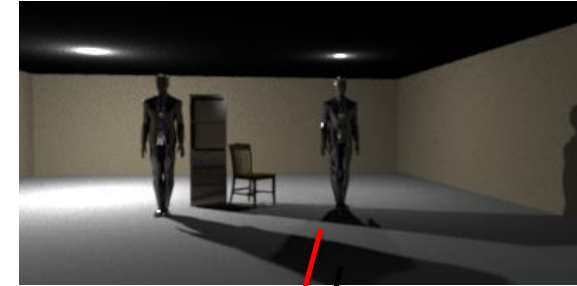
Challenge 1 : Separating signal from the noise

(i) Stray Shadows and Reflectors :

Solution: Differential

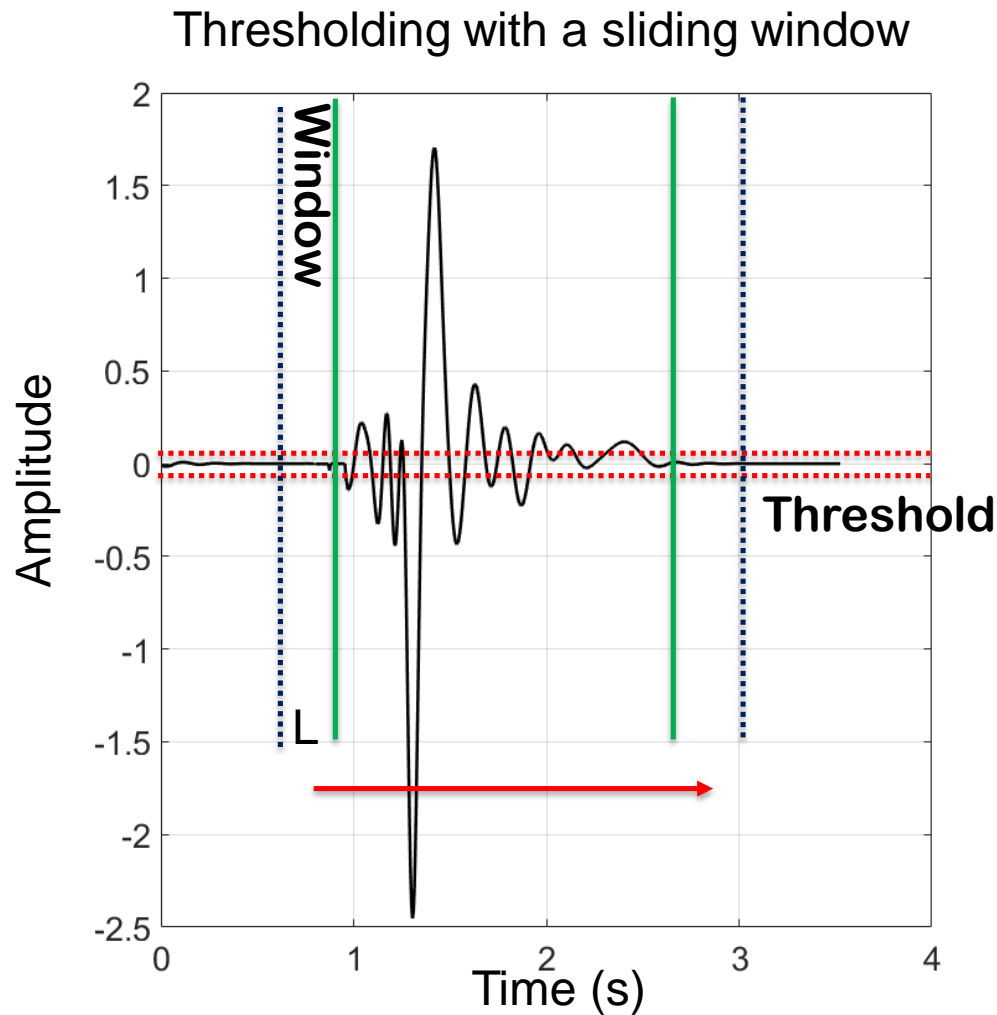
(ii) Light Source Flicker (AC Powered): (iii) Shot noise

Solution : Time-Frequency based denoising :
Hard wavelet thresholding using DWT



I. Preprocessing

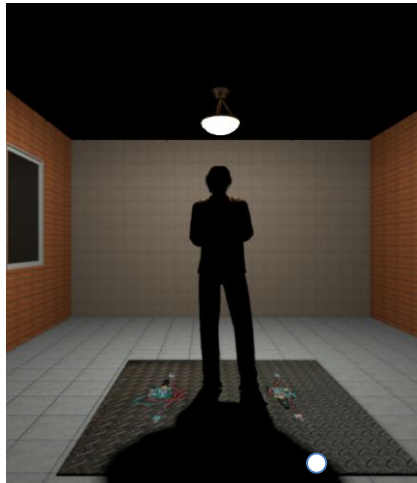
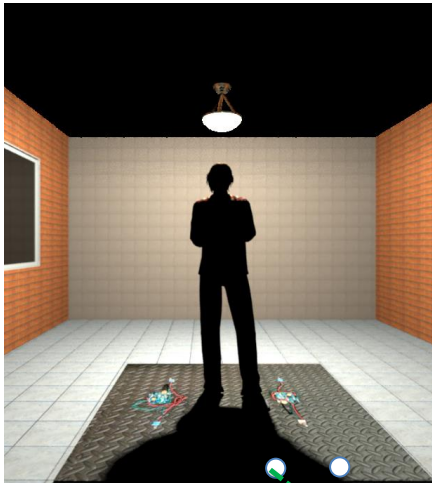
2. Gesture Detection:



I. Preprocessing

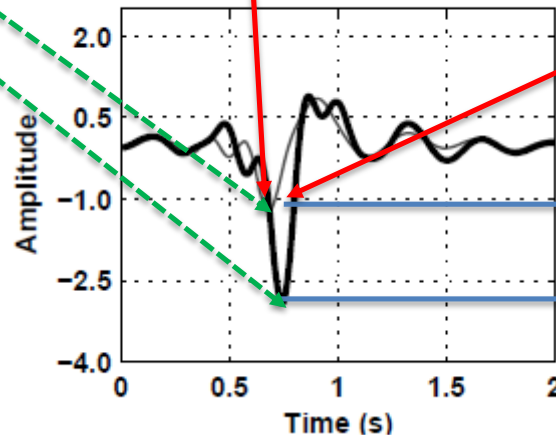
3. Standardization

Challenge 2 : Handling changes in intensity across sensors



Dimming

Adding multiple lights



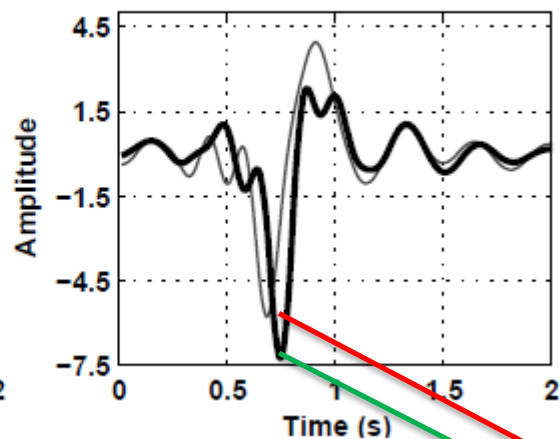
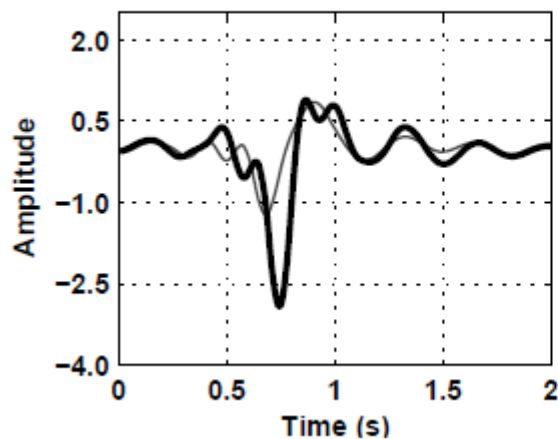
Significant difference in darkness

I. Preprocessing

3. Standardization

Challenge 2 : Handling changes in intensity across sensors

Solution : Scale each sensor time series by deviation



Similar
darkness

II. Feature Extraction



II. Feature Extraction

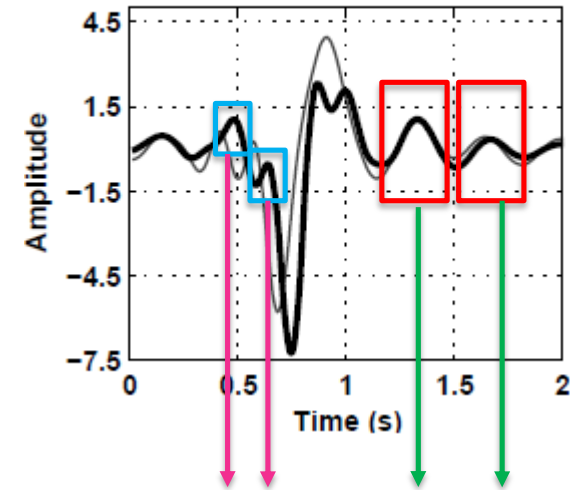
1. Wavelet Transformation:

Objective:

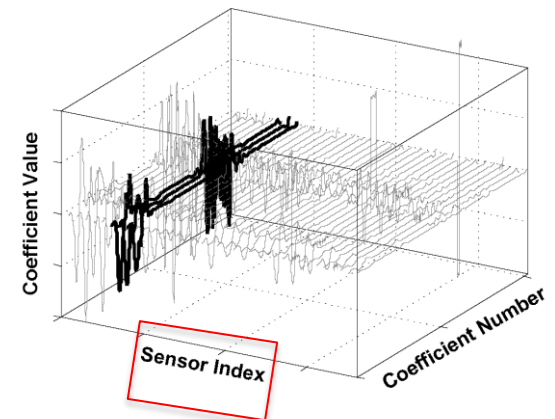
Characterize shape of the signal.

Approach:

Extract a joint signature in time and frequency domains using Discrete Wavelet Transform



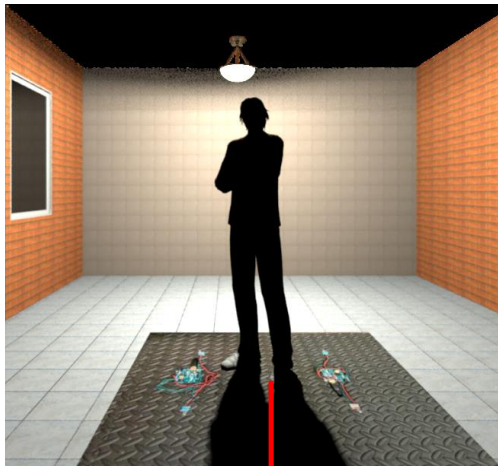
Varying Frequency over time



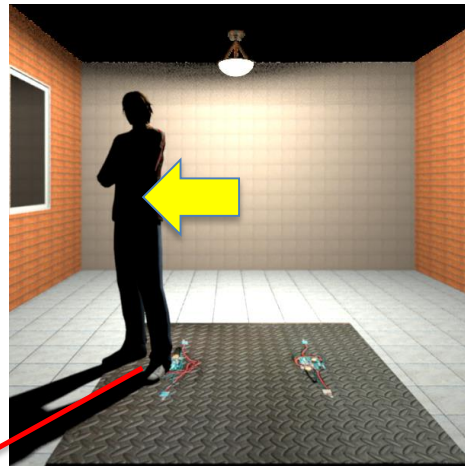
II. Feature Extraction

2. Rasterization:

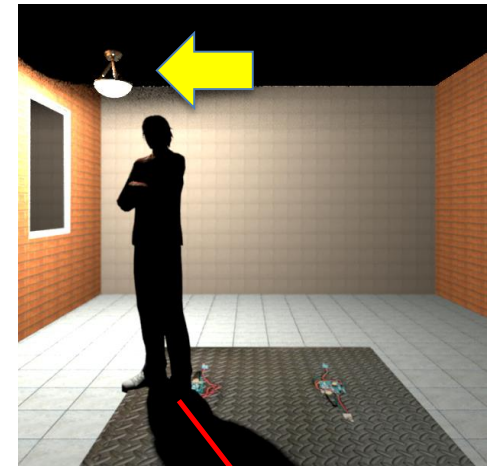
Challenge 3 : Handling changes in features caused by shifts in position of light sources or position of users



Example Position



Shift in User Position



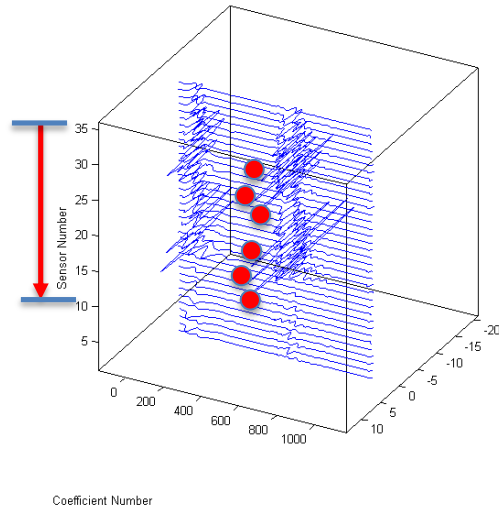
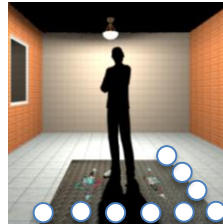
Shift in Light Source

Effect : Changing Direction and length of shadows across samples

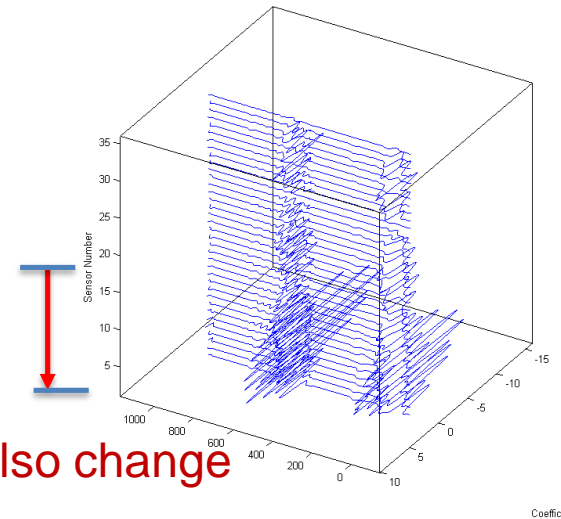
II. Feature Extraction

2. Rasterization:

Example:



6 sensors see variations



Sensors also change

10 sensors see variations

Need a way to negate the effects of change in shadow length/direction

II. Feature Extraction

2. Rasterization:

Existing Approaches:

- Identify blockage of individual light sources using Frequency Modulation to localize shadows
- Shadows can then be scaled, translated or rotated



- **Cannot be applied to unmodulated / unknown light sources**

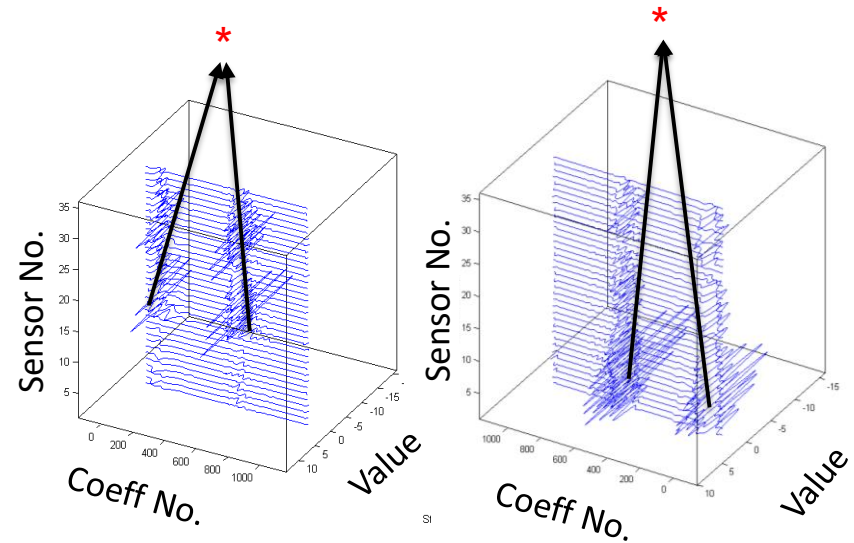
II. Feature Extraction

2. Rasterization:

How to handle variations in position of light sources or position of users with unmodulated light sources?

Observations:

- 1) Sensor values still have similar patterns* due to same blocking source
- 2) More light sources =>
Multiple redundant shadows
- 3) Change in shadow length =>
Change in No. of sensors
- 4) Change in shadow direction =>
Change in index of sensors



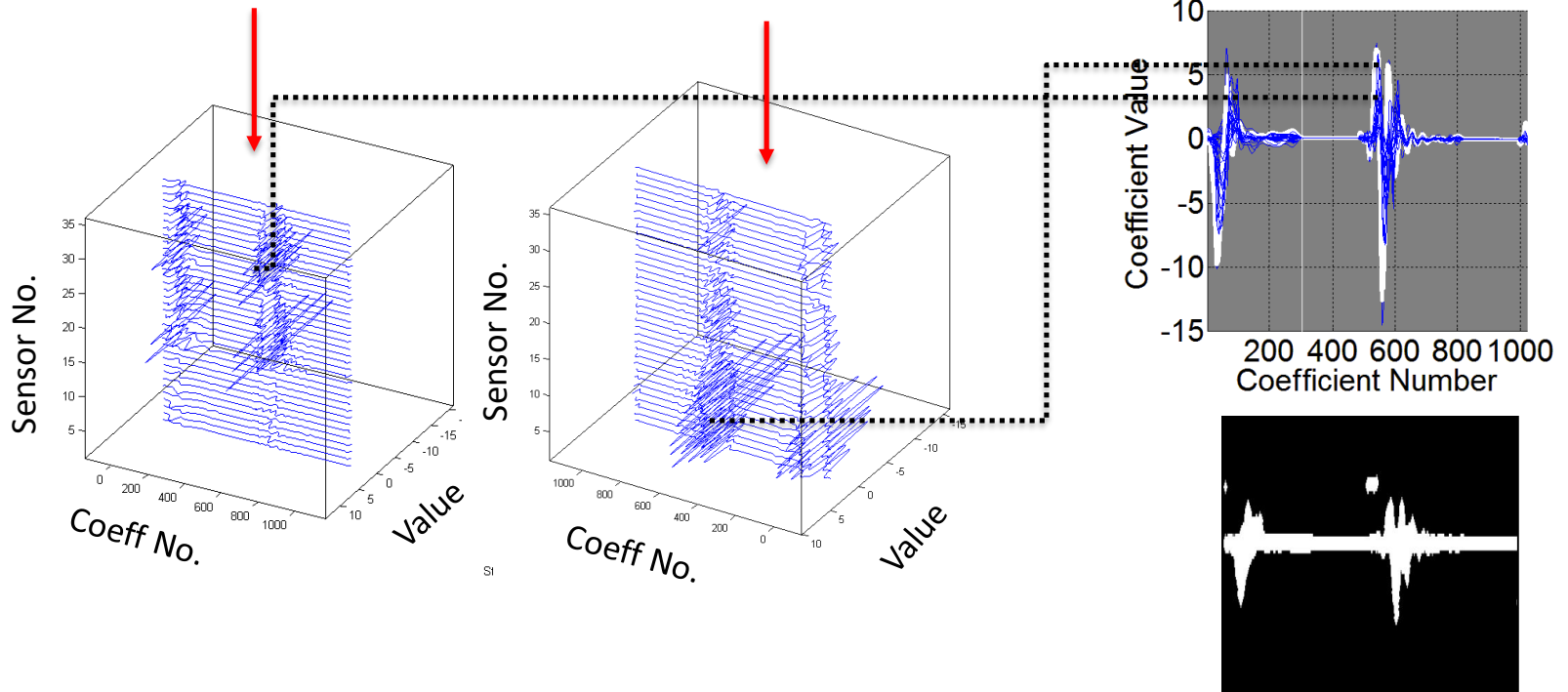
II. Feature Extraction

2. Rasterization:

How to handle variations in position of light sources or position of users with unmodulated light sources?

Solution: Map all sensor coefficients into a 2D image

Redundant / Similar patterns merge



II. Feature Extraction

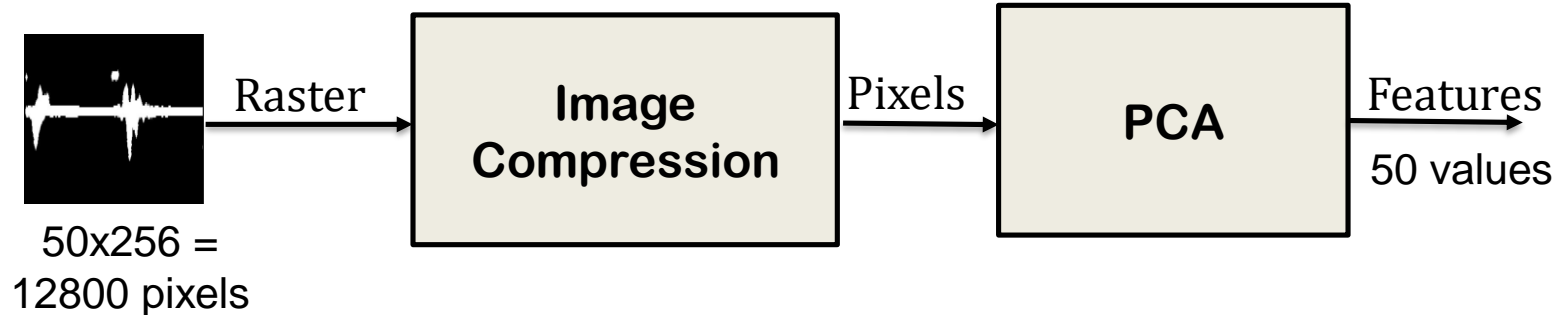
3. Feature Reduction:

Objective:

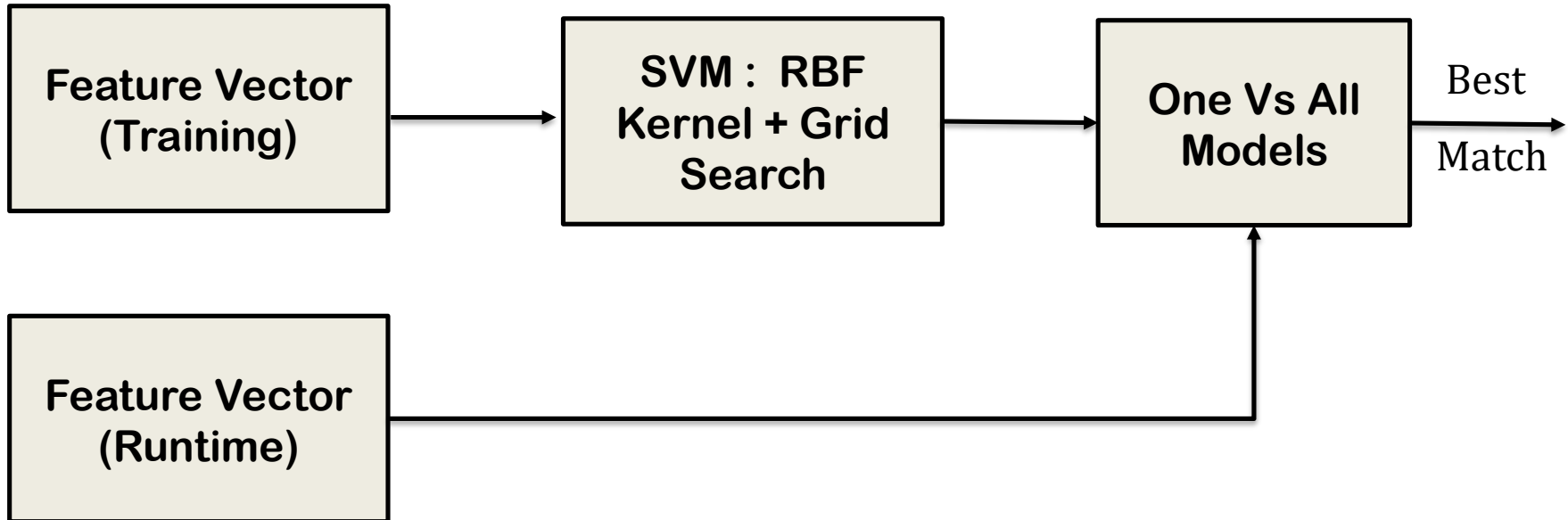
Extract only features of high classification potential

Approach:

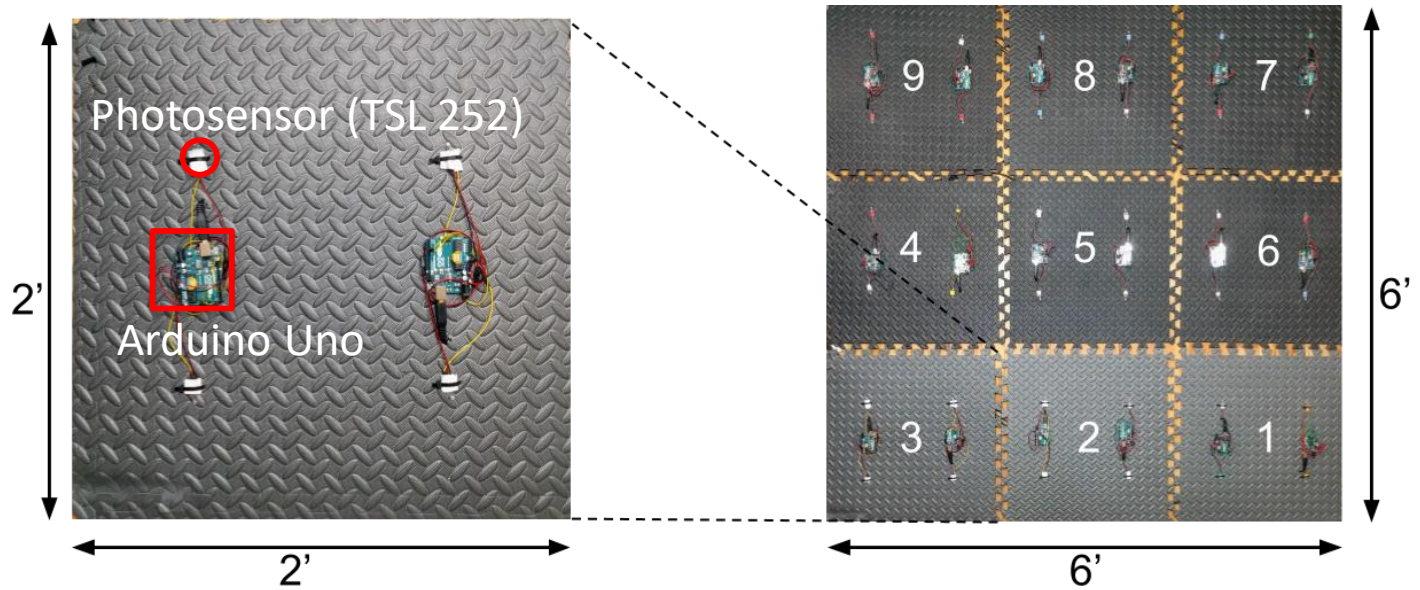
Dimensionality Reduction



III. Classifier Training / Recognition



Implementation



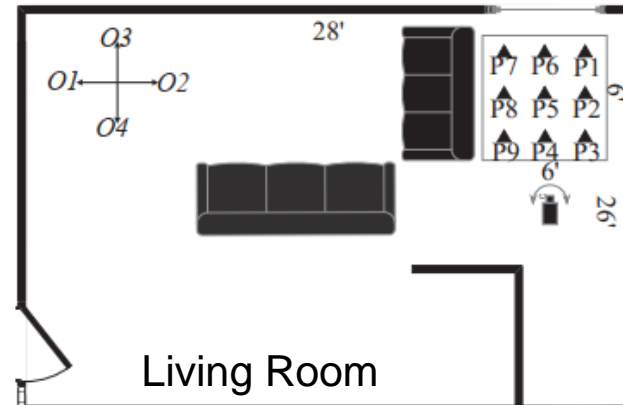
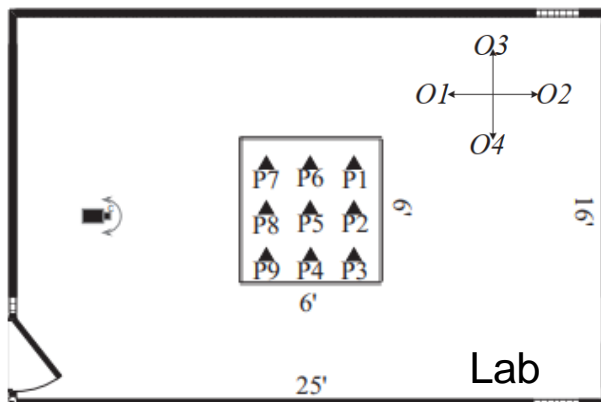
Sensor Density	1/sq.ft
Production System Cost	\$ 0.2 /sq.ft
Avg Cost of Carpeting	\$ 3 /sq.ft

Evaluation

- **Volunteers** : 20
Height : 150cm – 180cm
- **Gestures** : 5
Clap, Hug, Jump, Punch, Step

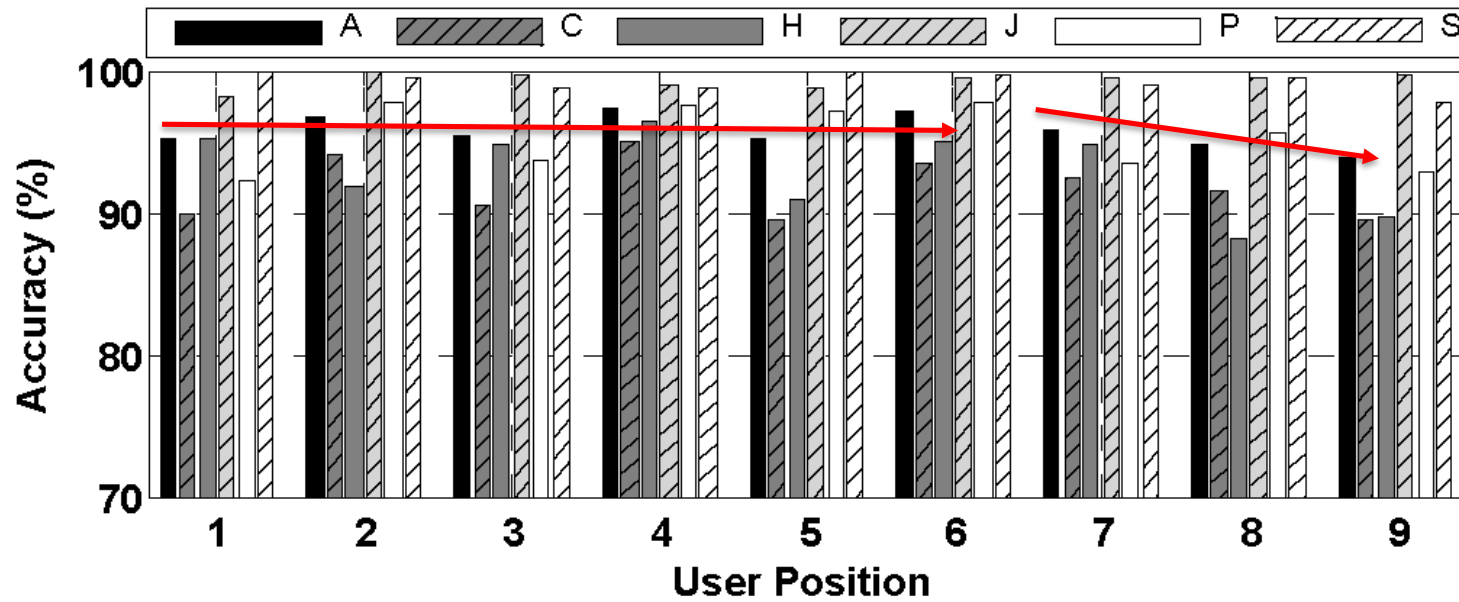


- **Positions** : 9
- **Orientations** : 4
- **Lighting Conditions** : 11
- **Environments** : 2 (15175 Samples)



Evaluation

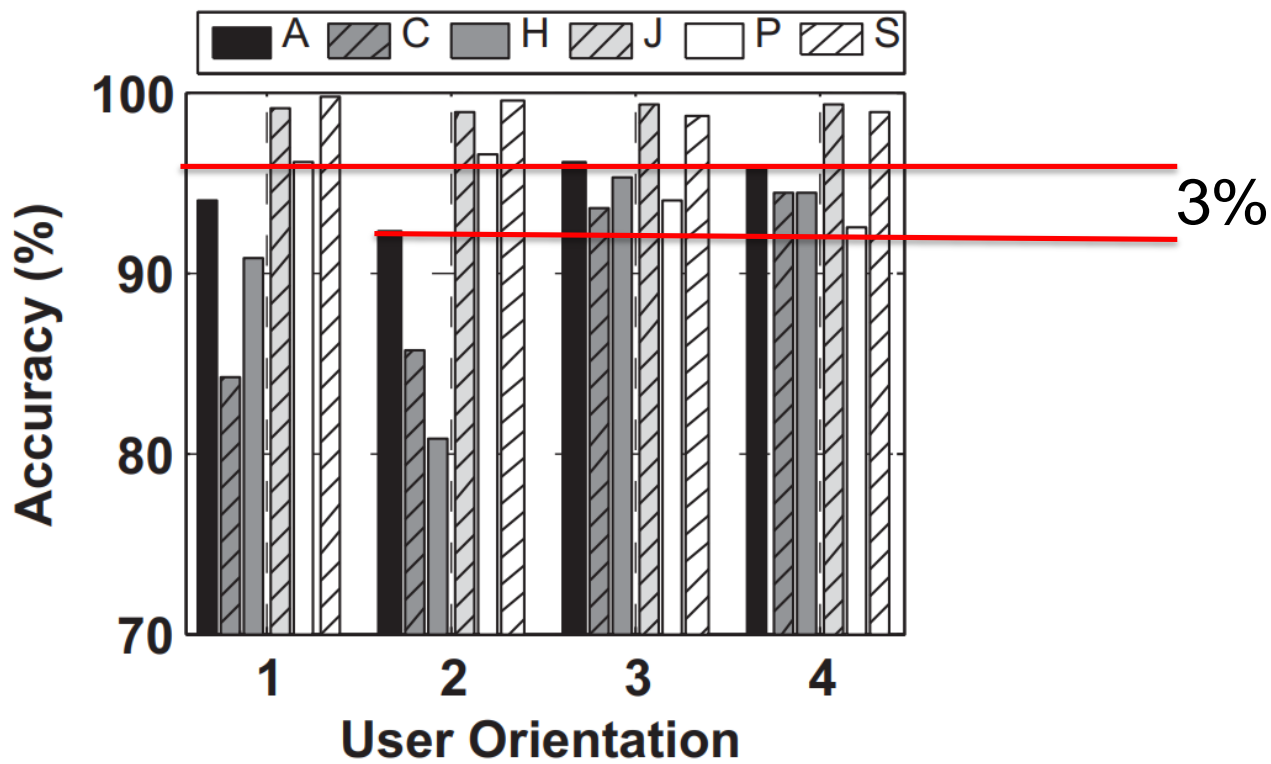
1. Recognition Accuracy : Unseen User Positions



Average : 95.2%

Evaluation

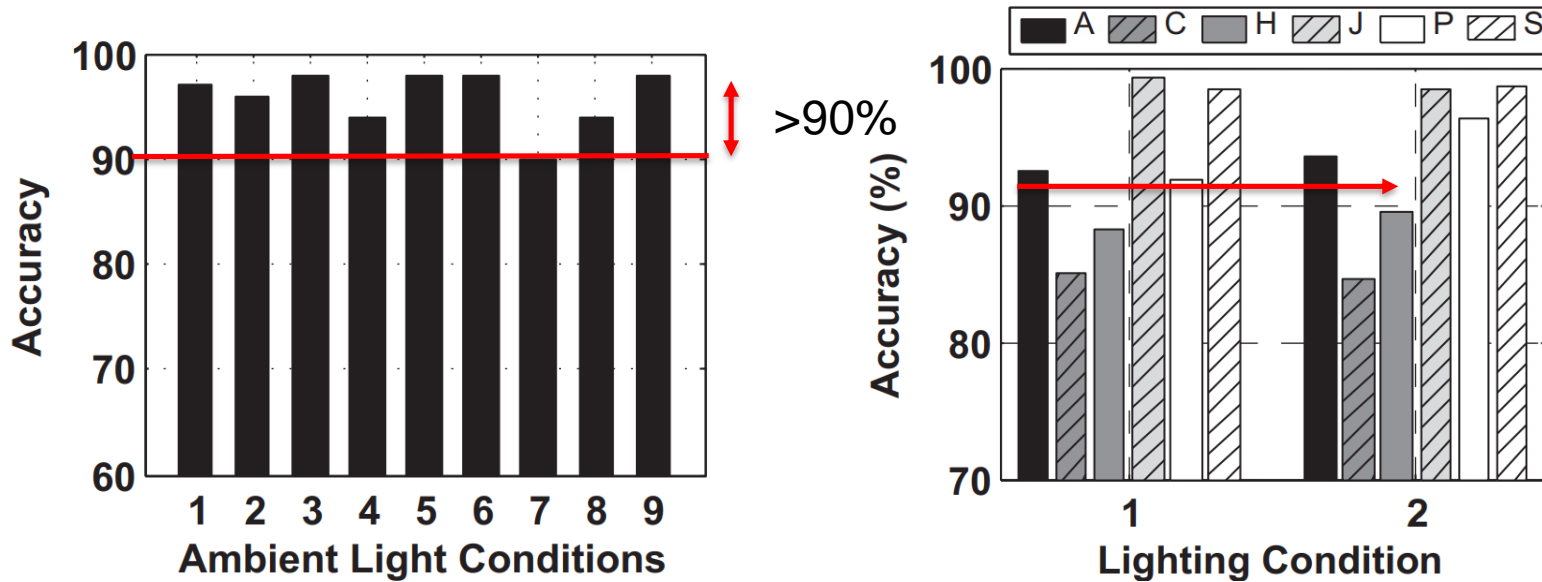
2. Recognition Accuracy : Unseen User Orientations



Average : 94.5%

Evaluation

3. Recognition Accuracy : Unseen lighting conditions



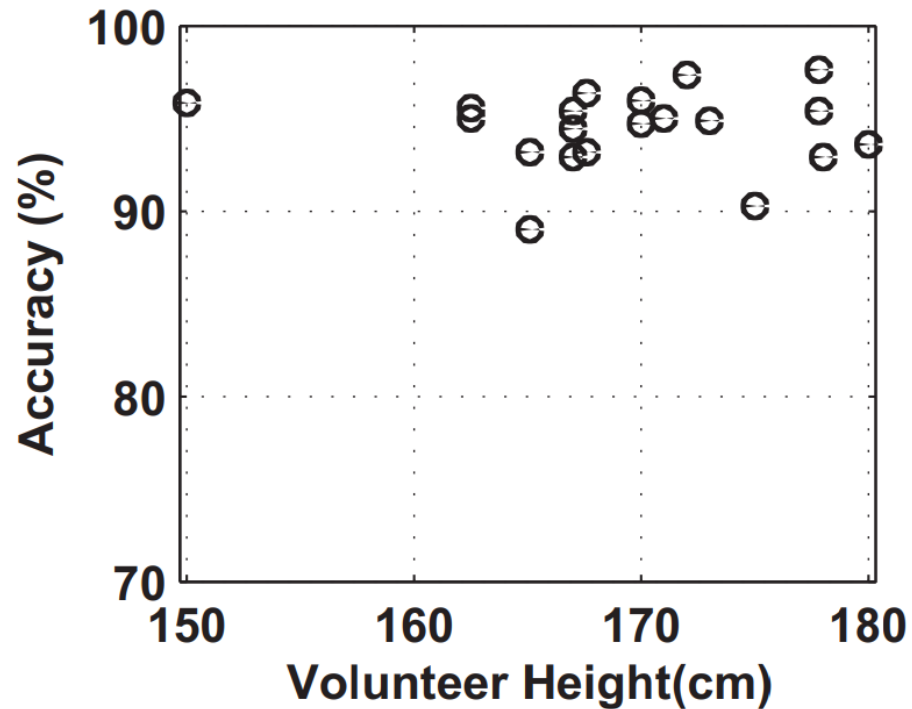
Living Room (Sunlight)
[200,1700 lux]

Lab (Fluorescent)
[280,320 lux]

Average : 96.1% , 93%

Evaluation

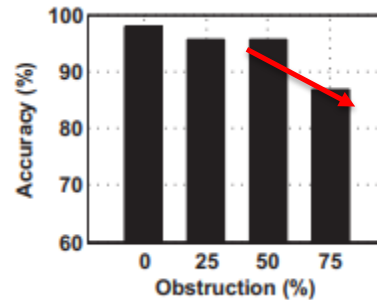
4. Recognition Accuracy : Unseen Users



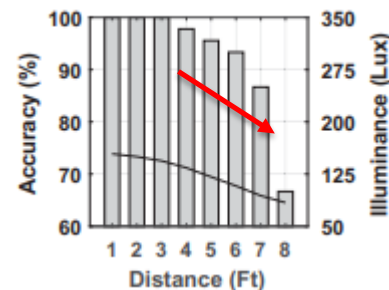
Average : 94.64%

Limitations

1. Obstructions



2. Sensitivity to low-illumination levels (<300 Lux)



3. Users cannot walk during gesture

Key Takeaways

1. Demonstrated a **gesture recognition system using only ambient light**.
2. Developed **feature extraction methods** agnostic to changing lighting conditions, user positions, user orientations, users.
3. Extensively evaluated a prototype using **low-cost commercially available sensors**.
4. Demonstrated **average accuracy (96%)** comparable to existing RF-based gesture recognition systems

Thank You

Questions?