

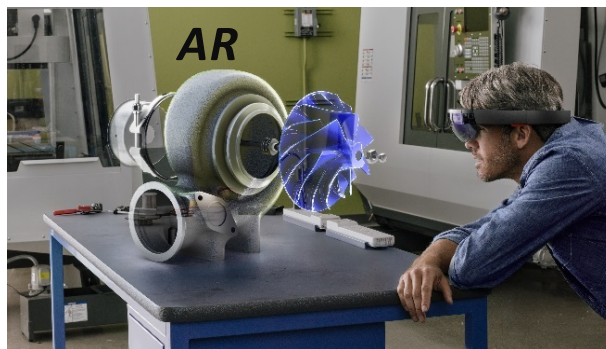
# Extracting and Quantifying Visual Information from Mobile Devices

Minh N. Do



# Mobile Devices with Visual Perception

- Automobile
- Service
- Consumer
- Medical
- Entertainment
- Education
- Domestic
- Manufacturing
- Military
- Augmented Reality



# Required Perception of Mobile Devices

## Mapping

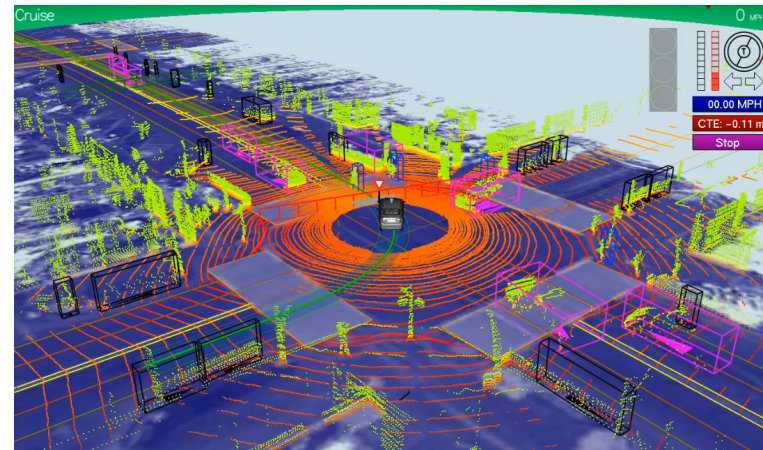
- Location
- Geometry
- Semantics
- Updates

## Object

- Distance
- Dimension
- Category
- Instance

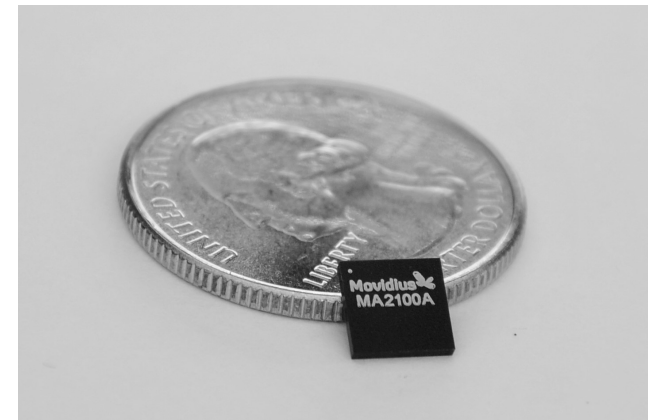
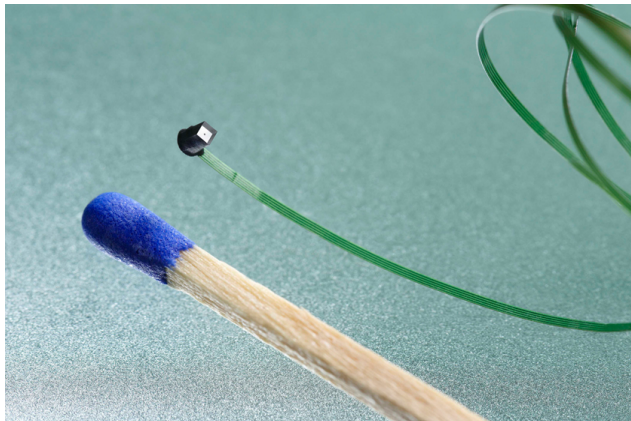
## Dynamics

- Motion
- Behavior
- Interaction



# Vision as Sensing Input


- High resolution provides details about complex scenes
  - Typical camera has ~1.3 mega-pixel, running at 36FPS (~50Mbyte per second)
  - More than 70% of human body's sensors are in the eyes
- Visual information: Shape vs. Appearance
  - Most complex situations are defined by appearance (texture) more than shape:
    - e.g. road markings, traffic signs, person identity, object instance, etc.
- Cheap and versatile in size and configuration



# Computer Vision

past

present

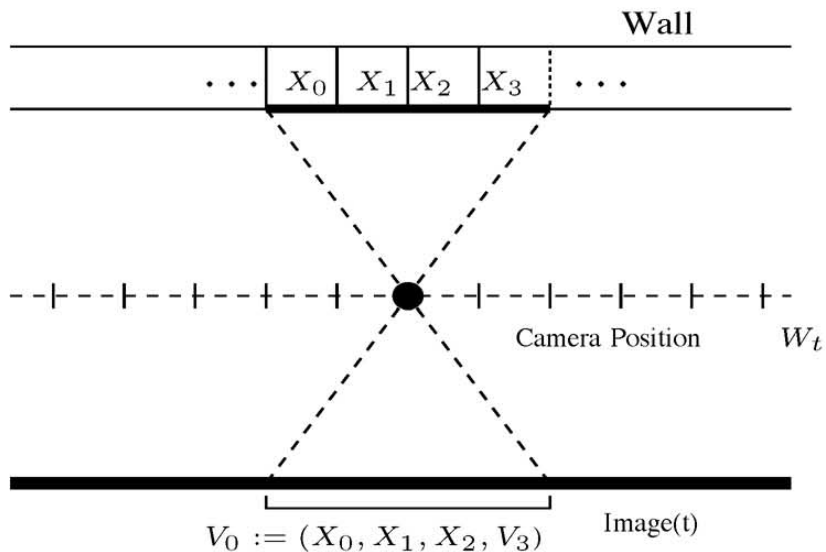
- 
- Single image
  - Static scene
  - RGB only
  - Limited data
  - Limited computation power
  - Slow algorithms
- Video
  - Dynamic scene
  - Depth, IMU, GPS
  - Large amount of data
  - Visual computing chips
  - Real-time algorithms
- Key problems in mobile vision:
    - Localization and mapping
    - Object and place recognition
    - Motion and dynamics

## Vision from a Mobile Device

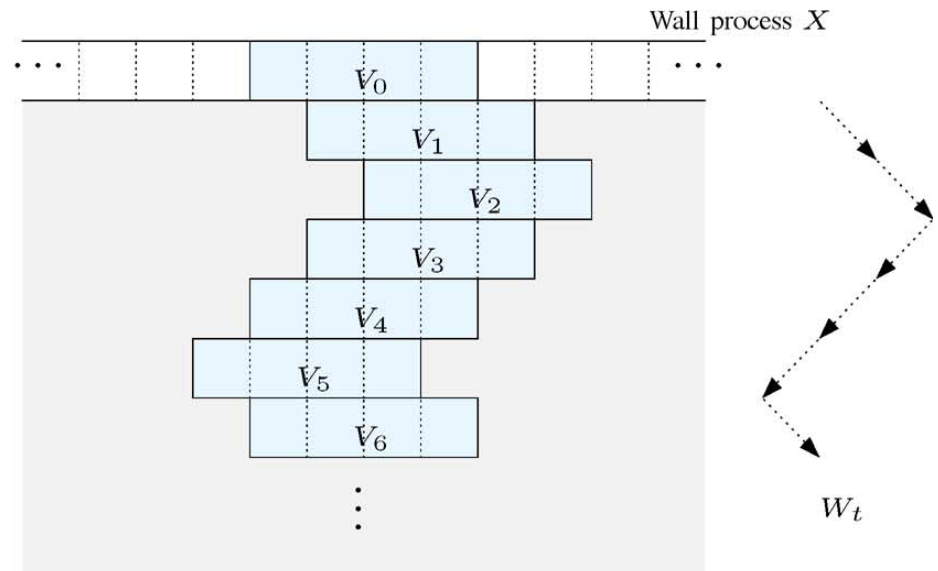


Source: Seattle Police Department

# On the Information Rates of Mobile Vision



(a)



(b)

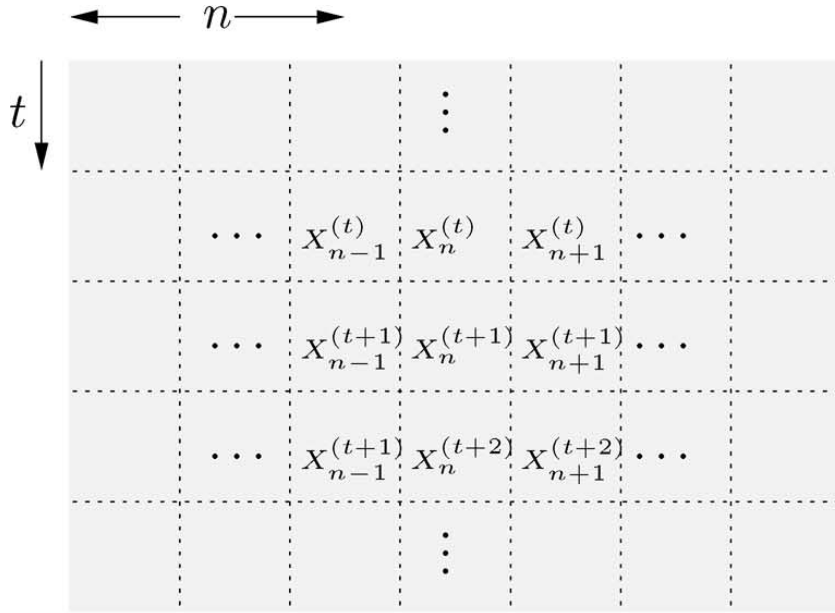
*Theorem:* The entropy rate and rate-distortion function of  $V$  satisfies

$$(1 - 2p_W)H(X) + H(p_W) - H(P_e) \leq H(V) \leq (1 - 2p_W)H(X) + H(p_W)$$

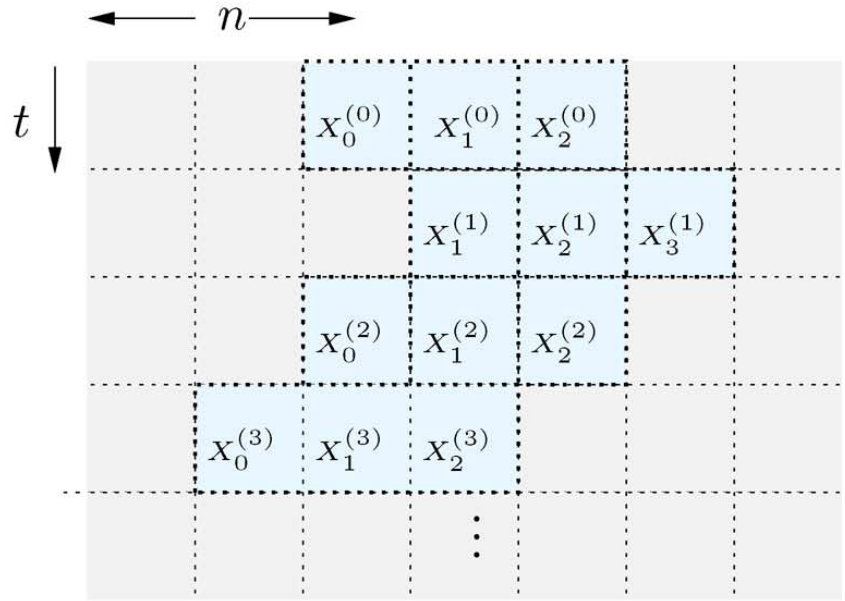
$$R_V(D) \leq (1 - 2p_W)R_D(X) + H(p_W)$$

Cunha, Do, Vetterli (2010)

# Mobile Vision of Dynamic Environments



(a)



(b)

*Theorem:*

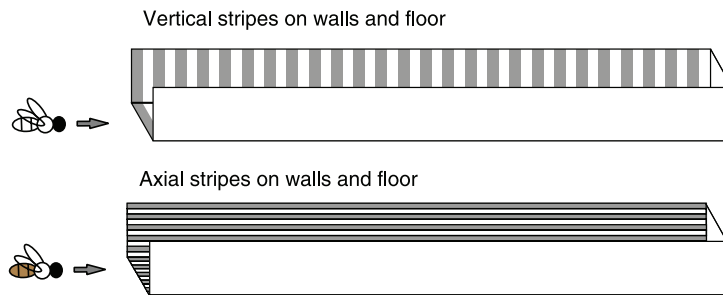
$$H(V|W) + H(p_W) - H(P_e) \leq H(V) \leq H(V|W) + H(p_W)$$

where  $H(V|W) = (L - 1)H(X_0^{(1)}|X_0^{(0)}) + (1 - 2p_W)H(X_0^{(\infty)}) + \sum_{i=0}^{\infty} H(X_0^{(i)}|X_0^{(0)})\Pr\{\mathcal{T}^i\}$

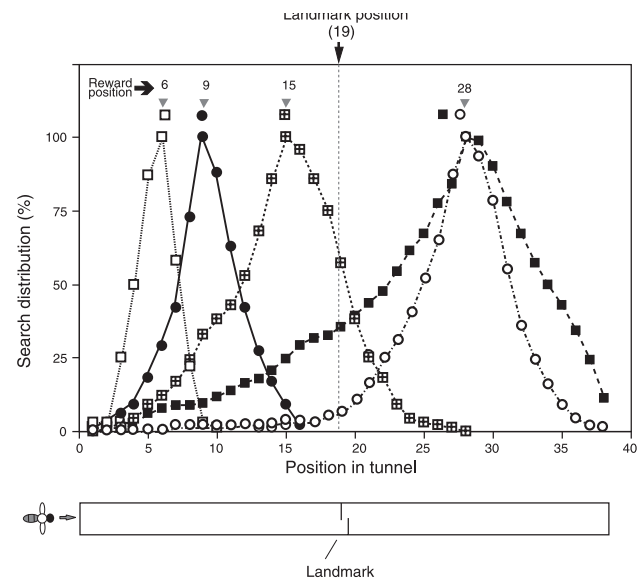
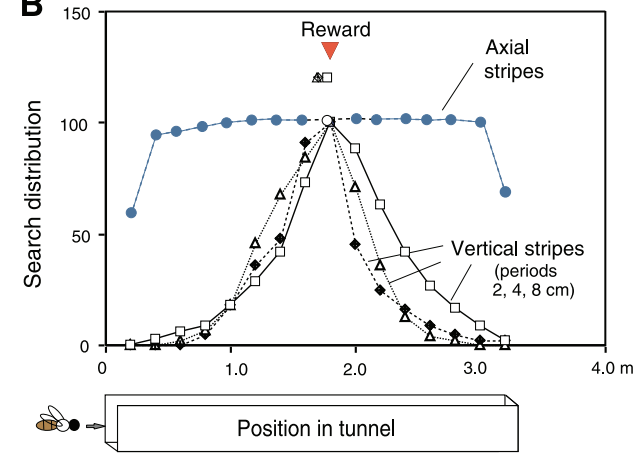


# Lessons from Bees

**A**



**B**



Srinivasan (1997, 2011)

# Holistic Mobile Vision

Geometric Reconstruction meets  
Semantic Recognition for 3D holistic vision:  
**Real-time, robust, geometry-centric vision**

Where: geometry, location...

What: semantics, action...

## Reconstruction

- Real-time **camera pose** localization
- 3D **environment** mapping
- **Depth** and **motion** estimation
- Large-scale **urban** reconstruction

## Recognition

- Visual **place/scene** recognition
- **Object** recognition, localization
- **Human** re-identification
- **Action** recognition, tracking

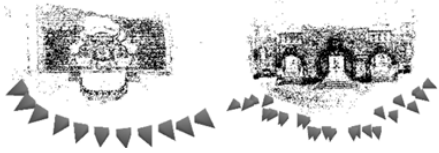
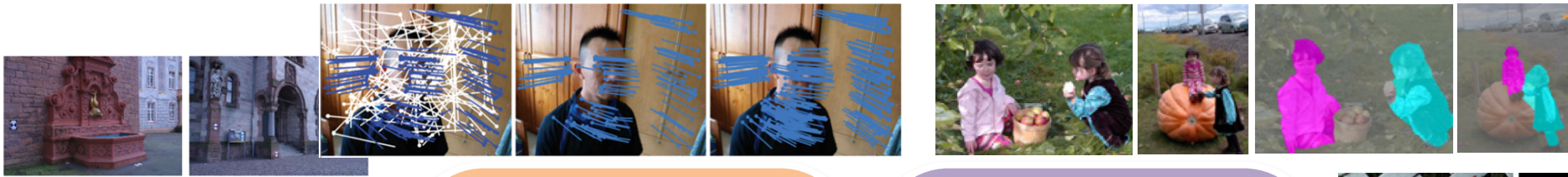
How: model, solve, compute

## Key techniques

- Multiple sensors
- Modern vehicles
- Moving robots
- Opportunistic scan

- Geometry-aware filtering
- Fast randomized algorithms
- Efficient inference models
- Deep learning innovation

- Mobile cameras
- Big visual data
- Rich annotations
- Powerful machine



### Scene structure

- Depth enhance.
- Slanted stereo
- Optical flow
- Live FG segment
- 3D city recons.

### 3D vision & percept.

CODE [ECCV'14, '16]  
Feature matching

DSE [CVPR'15]  
Camera pose rec.

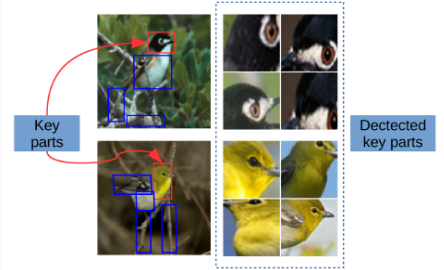
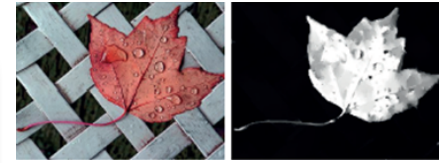
SLAM-O [TCSVT'16]  
3D object prop.

### Recogn. & segment.

Fine-grained  
recog. [TIP'16]

PISA [CVPR'13b]  
Pixel-acc. saliency

MHIC [TMM'16]  
Recog & Co-seg.



### Edge-aware filtering

CLMF [CVPR'12]  
Local EAF

FGS [TIP'14]  
Global EAF

FGI [ECCV'16]  
Sparse to dense

### Efficient inference

PMF [CVPR'13]  
Local optim.

DFG [CVPR'14]  
Generalized

SPM-BP [ICCV'15]  
Global optim.

### Comp. imaging

- Photo refocus
- Rain removal
- Image stitching
- Multi-scale dec.
- Structure extra.
- Image warping
- Colorization

### Current topics

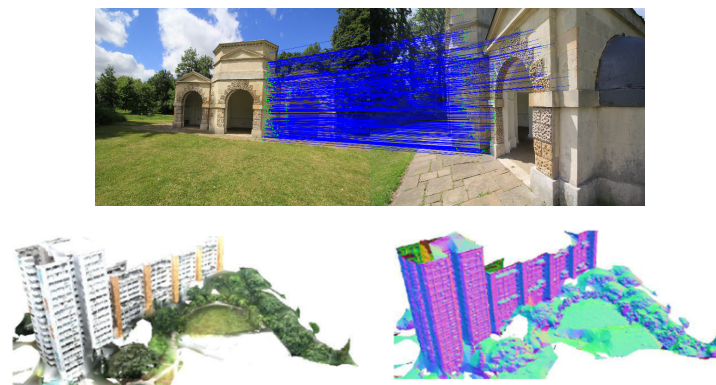
- *Scene flow*
- *Robust SLAM*
- *Motion seg.*
- *Text detection*
- *Place recogn.*
- *Scene labeling*
- *Action recog.*
- *Deep learning*

# Holistic Mobile Vision

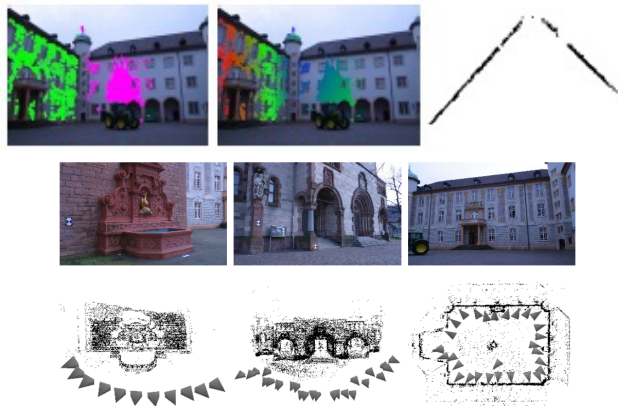
## Dense depth & motion



## Feature matching and mapping



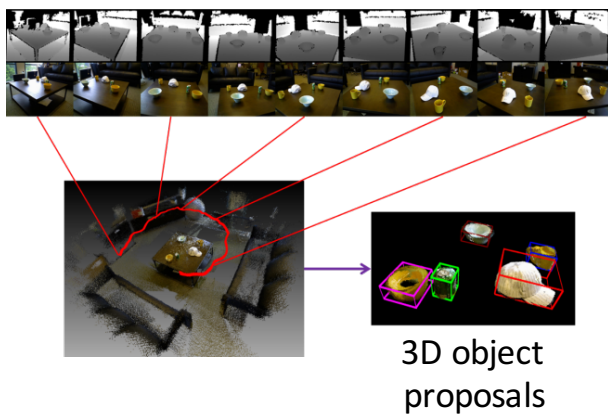
## Camera pose localization



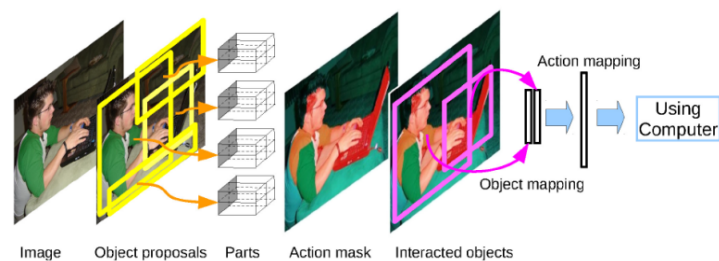
## Stitching & visualization



## Localizing 3D object proposals



## Action recog. w/ min. labelling



# Summary

- Visual perception is crucial for mobile devices and autonomous systems
  - Small
  - Cheap
  - **Fast**
- Key problems:
  - Localization and mapping
  - Object and place recognition
  - Motion and dynamics
- Holistic mobile vision:
  - Combine geometric reconstruction with semantic recognition
  - Supported by information rates
  - Challenges: real-time and robust methods