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# Capricorn: Towards Real-time Rich Scene Analysis Using RF-Vision Sensor Fusion

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**Ziqi Wang, Ankur Sarker, Jason Wu, Derek Hua, Gaofeng Dong, Akash Deep Singh, and Mani B. Srivastava**

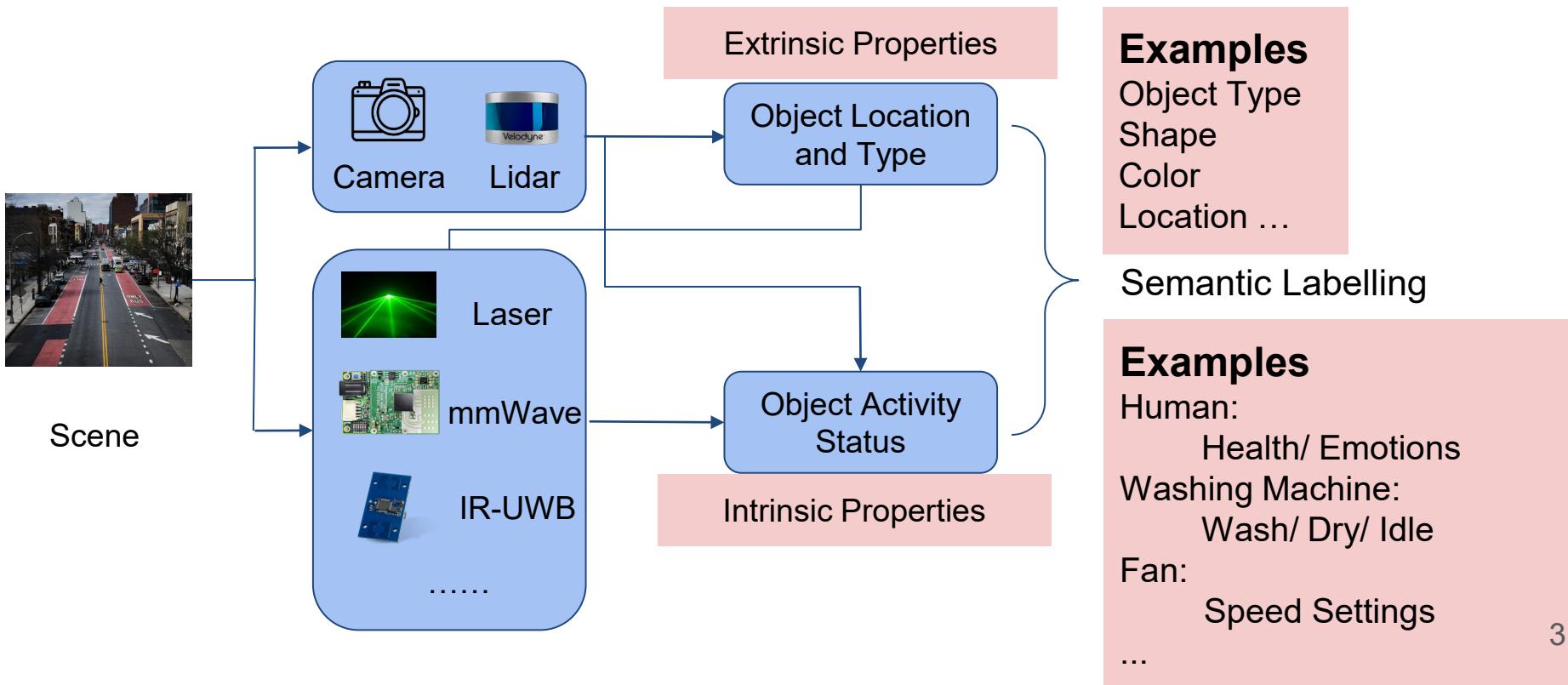
University of California, Los Angeles, Electrical and Computer Engineering Department

# Real-time Rich Scene Analysis

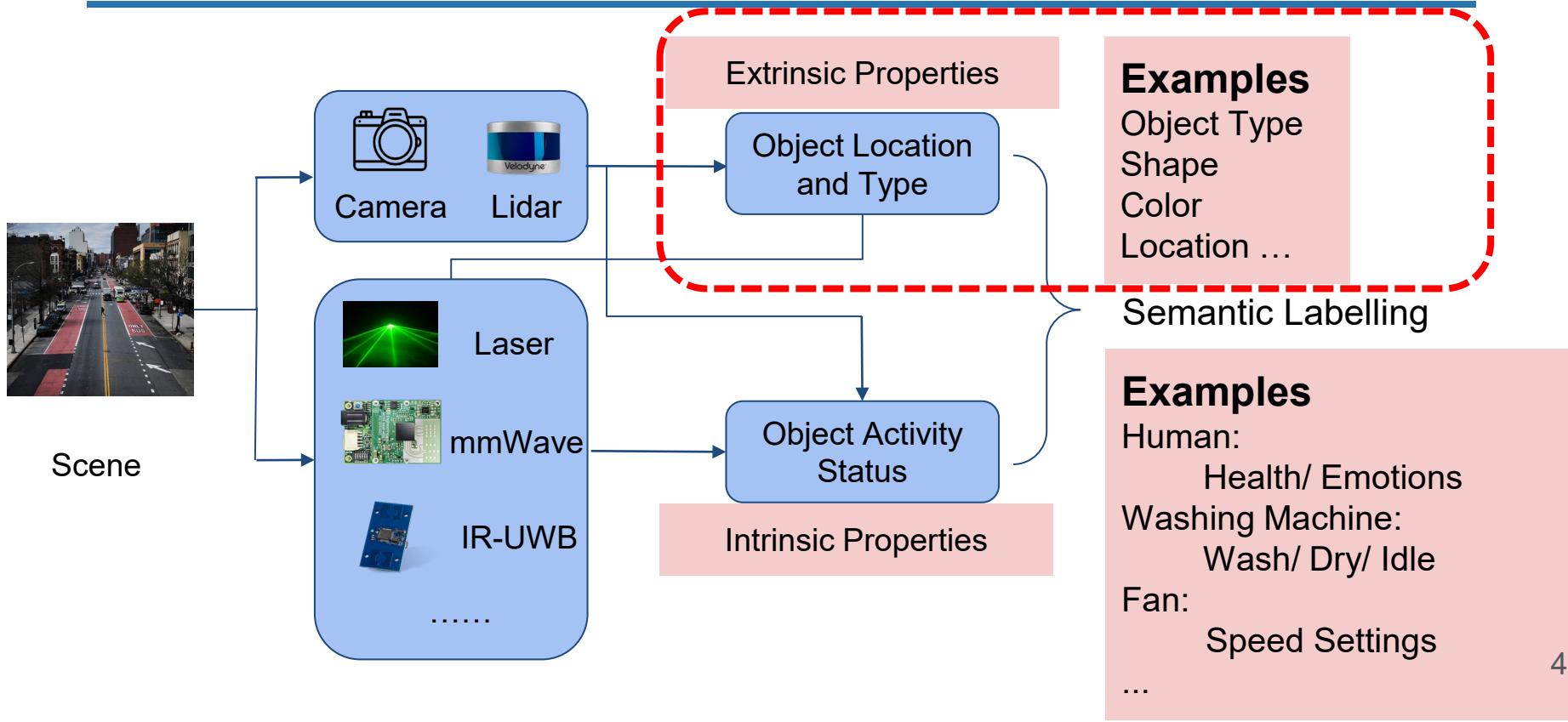
- Situation awareness (SA) is crucial to smart systems
- SA requires the collection and analysis of sensory data to estimate the states of an environment



# Real-time Rich Scene Analysis

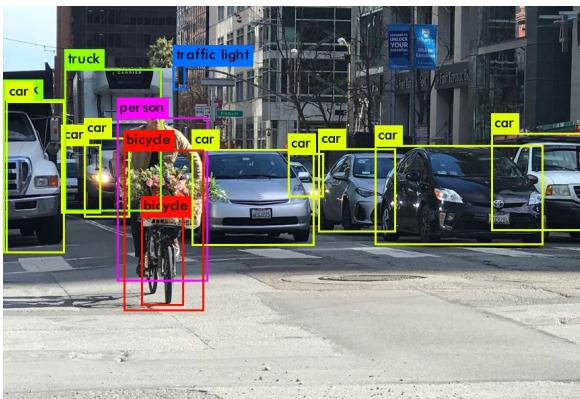


# Real-time Rich Scene Analysis

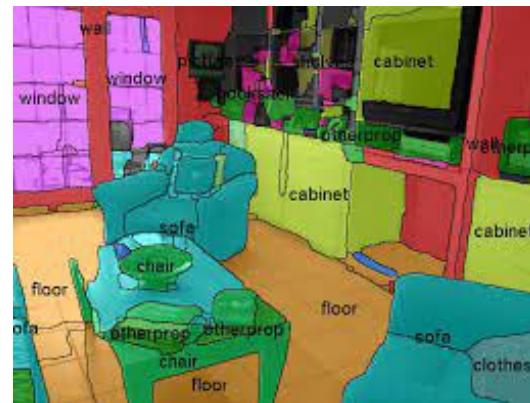


# Computer Vision for Extrinsic States

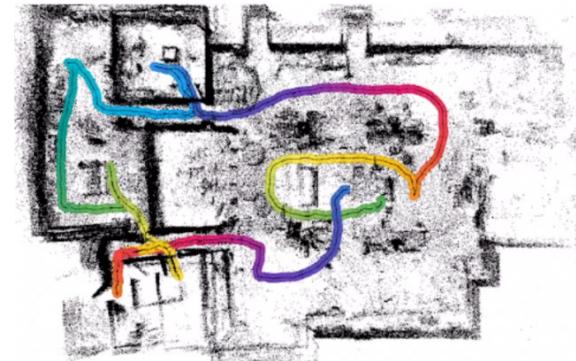
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Object Detection and Recognition

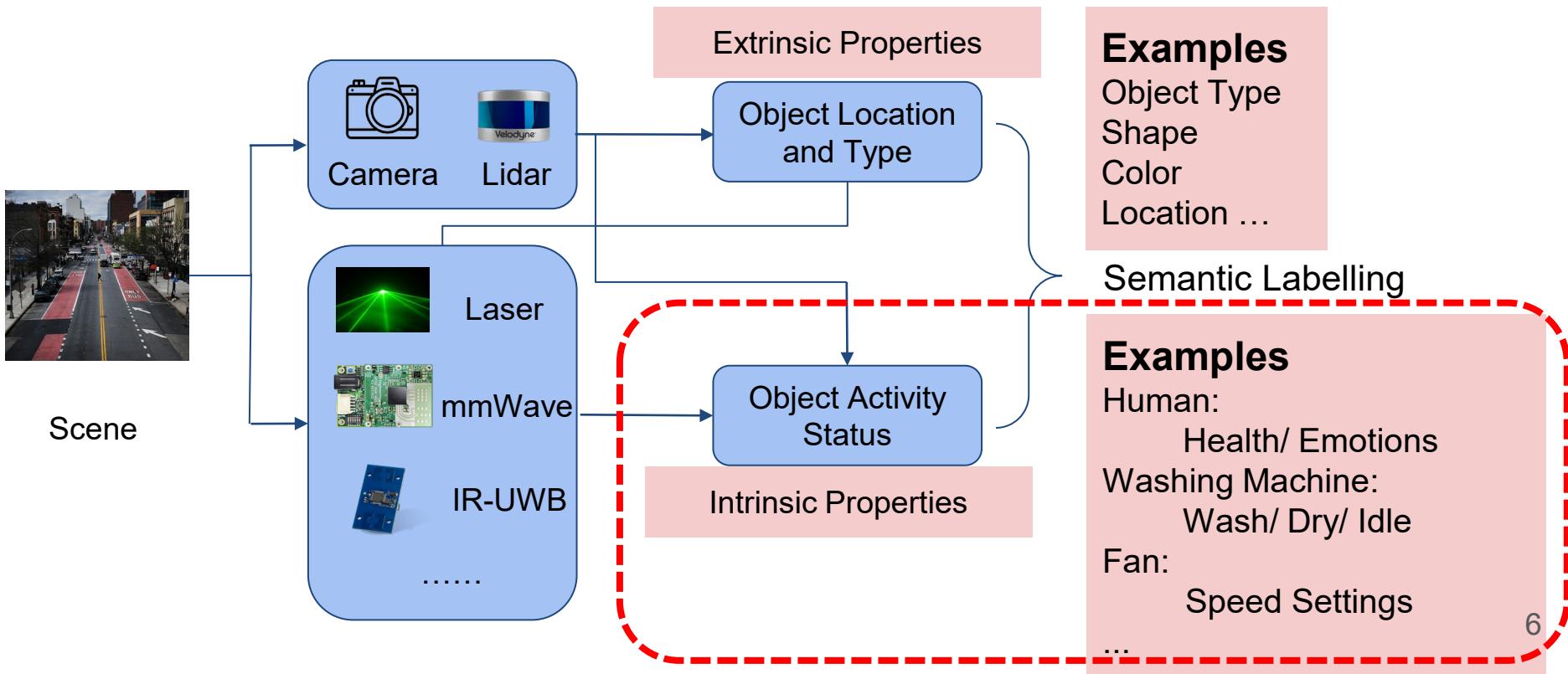


Semantic Segmentation



Localization and Mapping

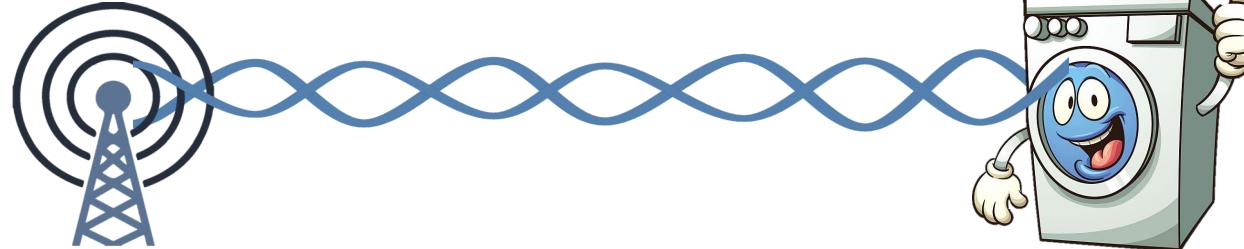
# Real-time Rich Scene Analysis



# Wireless Vibrometry for Intrinsic States

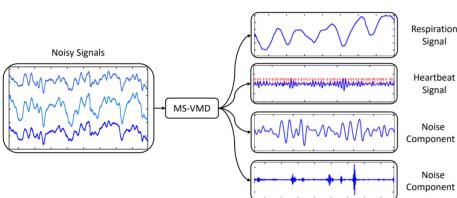
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Emit a probing RF signal

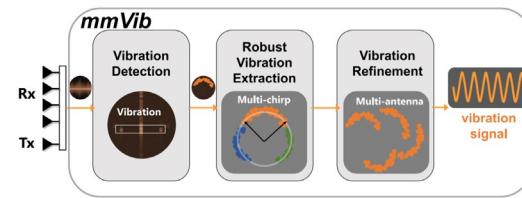


Washing machine  
State: washing

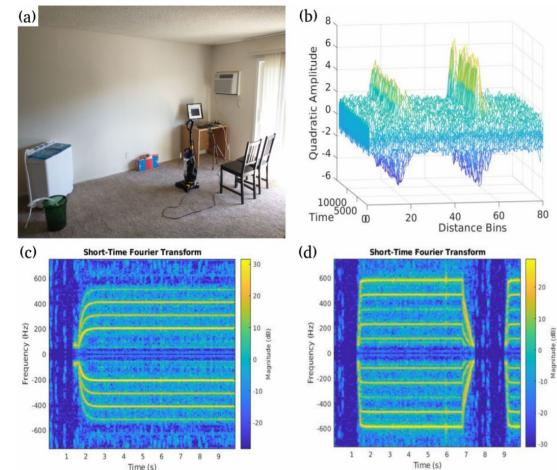
# Wireless Vibrometry for Intrinsic States



Vital Signals / Activities from Human Operating States of Industrial Machinery  
(Vi2Fi, IMWUT'20)



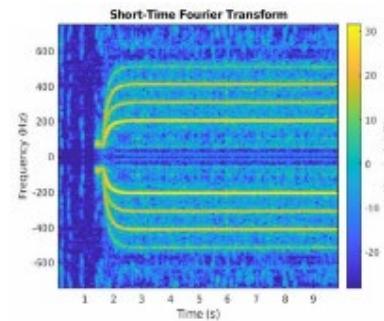
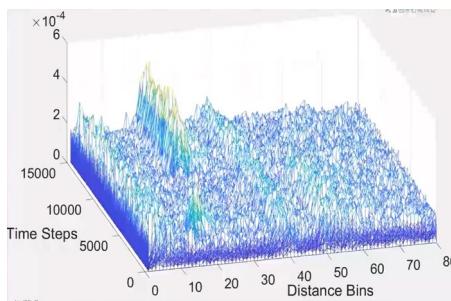
Operating States of Industrial Machinery  
(mmVib, MobiCom'20)



Usage of Home Appliances  
(UWHear, SenSys'20)

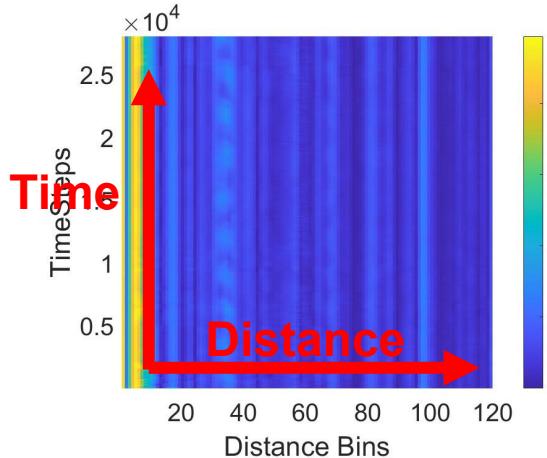
# RF Signals Require Context to Make Sense

- Presumed object type: existing wireless vibrometry sensing systems presuppose the existence of a particular type of object (e.g., a person) in the scene to process the signal accordingly.



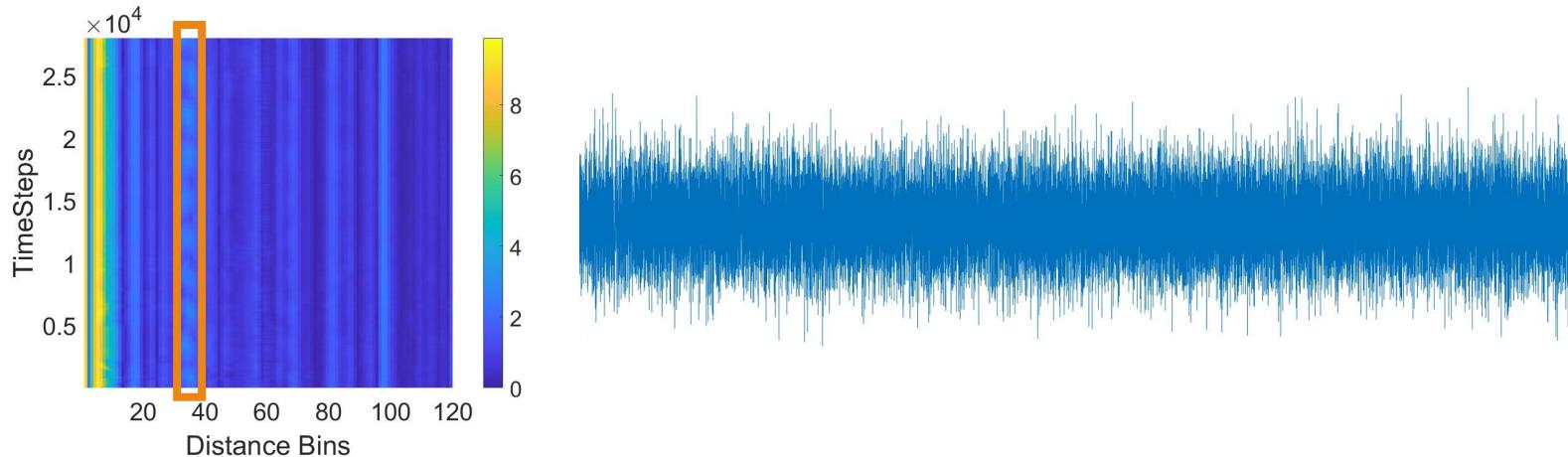
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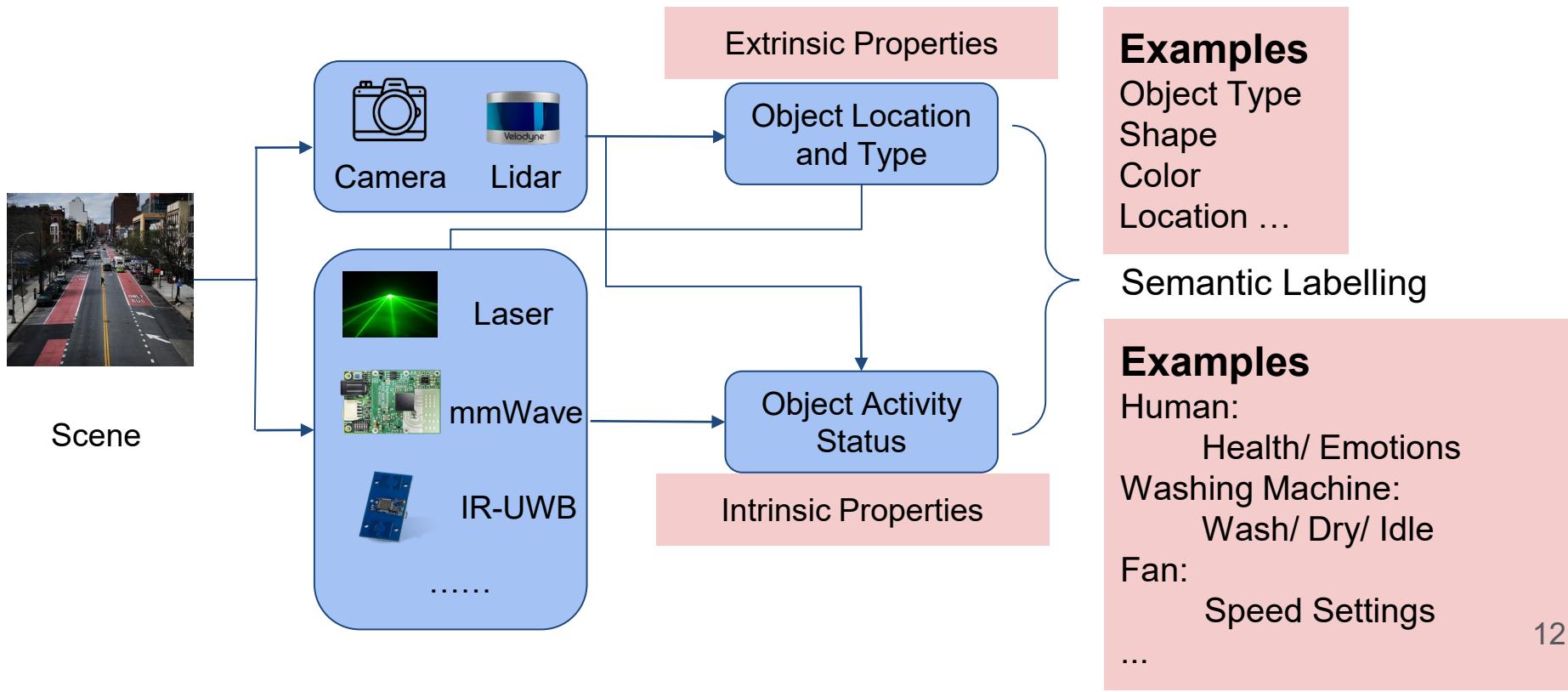


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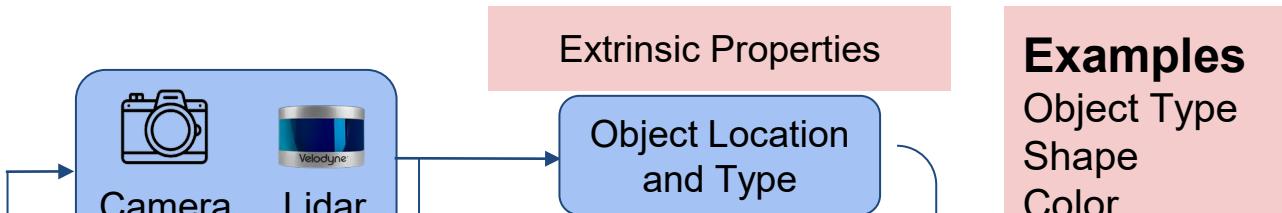
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# Our Vision: Multimodal Sensor Fusion

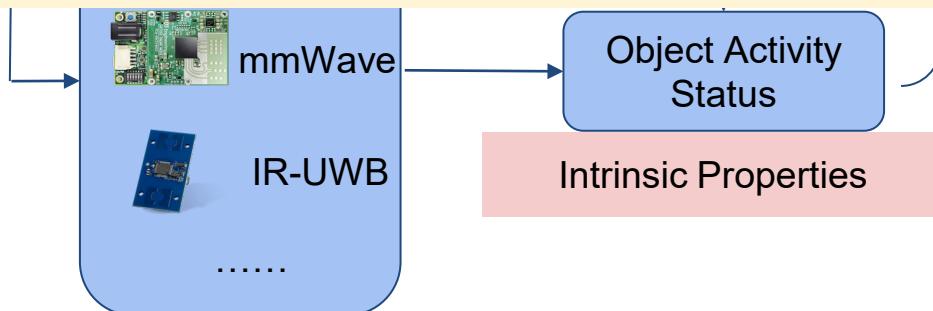


# Our Vision: Multimodal Sensor Fusion



Multiple sensing modalities can offer complimentary capabilities

Scene



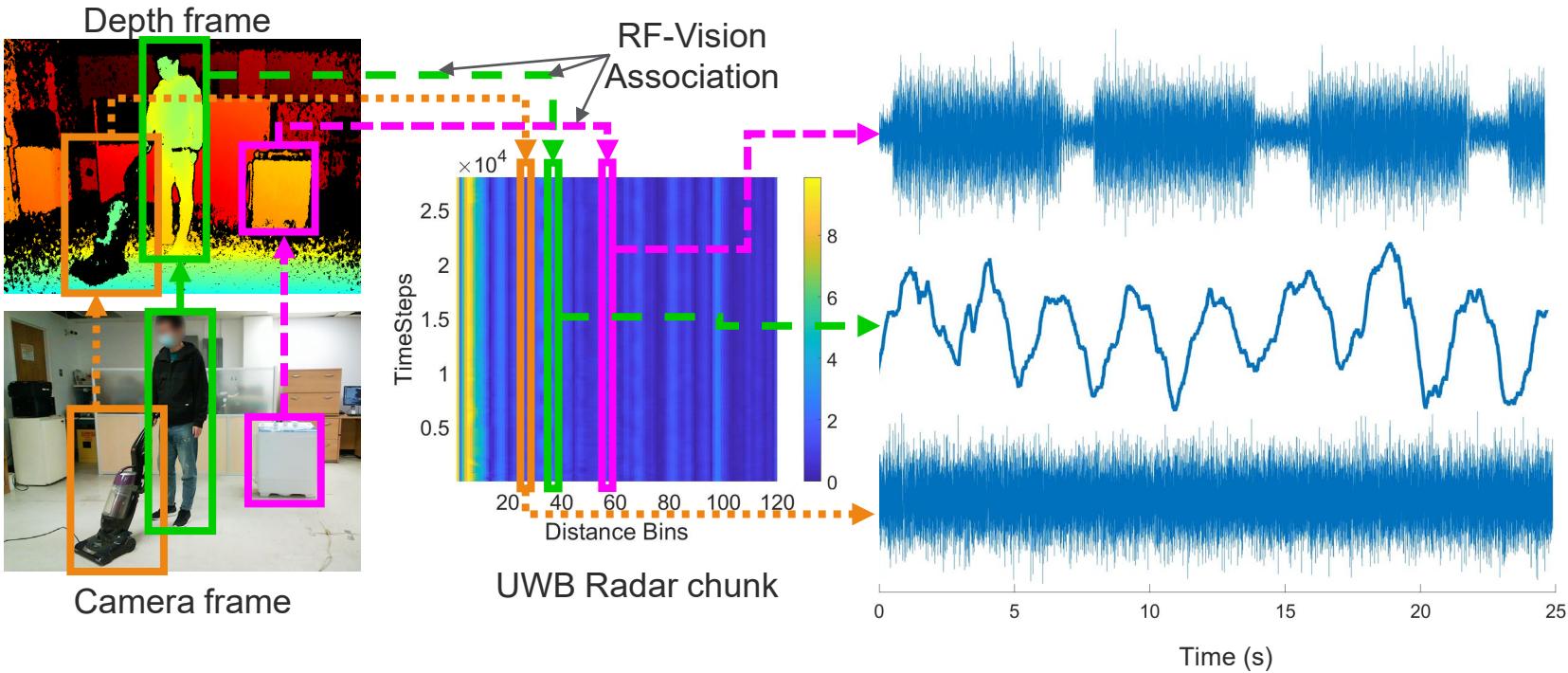
Examples

Human:  
Health/ Emotions  
Washing Machine:  
Wash/ Dry/ Idle  
Fan:  
Speed Settings  
...

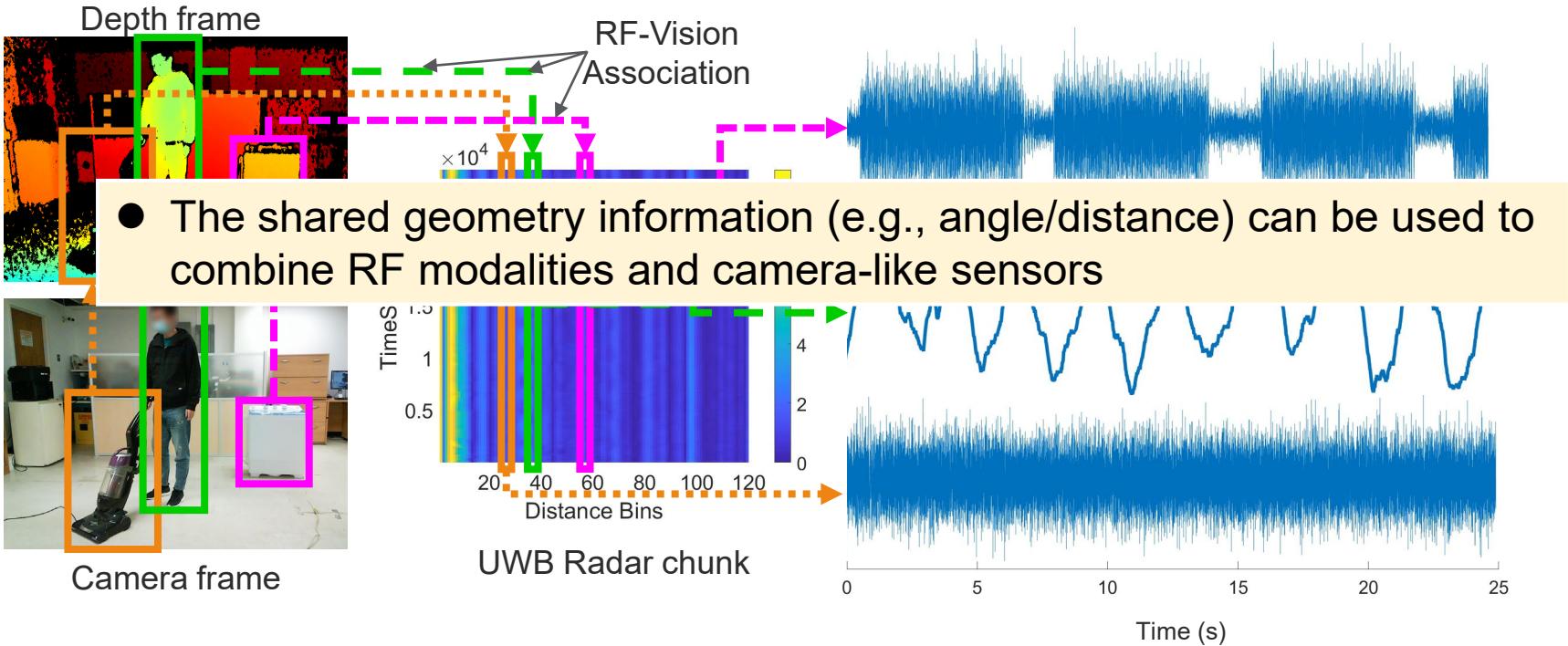
# Our Insights: Multimodal Sensor Fusion

- Presumed object type: existing wireless vibrometry sensing systems presuppose the existence of a particular type of object (e.g., a person) in the scene to process the signal accordingly.
- Presumed number of objects: existing systems either target at a fixed number of subjects or rely on threshold-based search algorithms
- Multiple sensing modalities can offer complimentary capabilities
- The shared geometry information (e.g., angle/distance) can be used to combine RF modalities and camera-like sensors
- One sensor's inference can become the prior information for another and expedite the signal processing algorithms

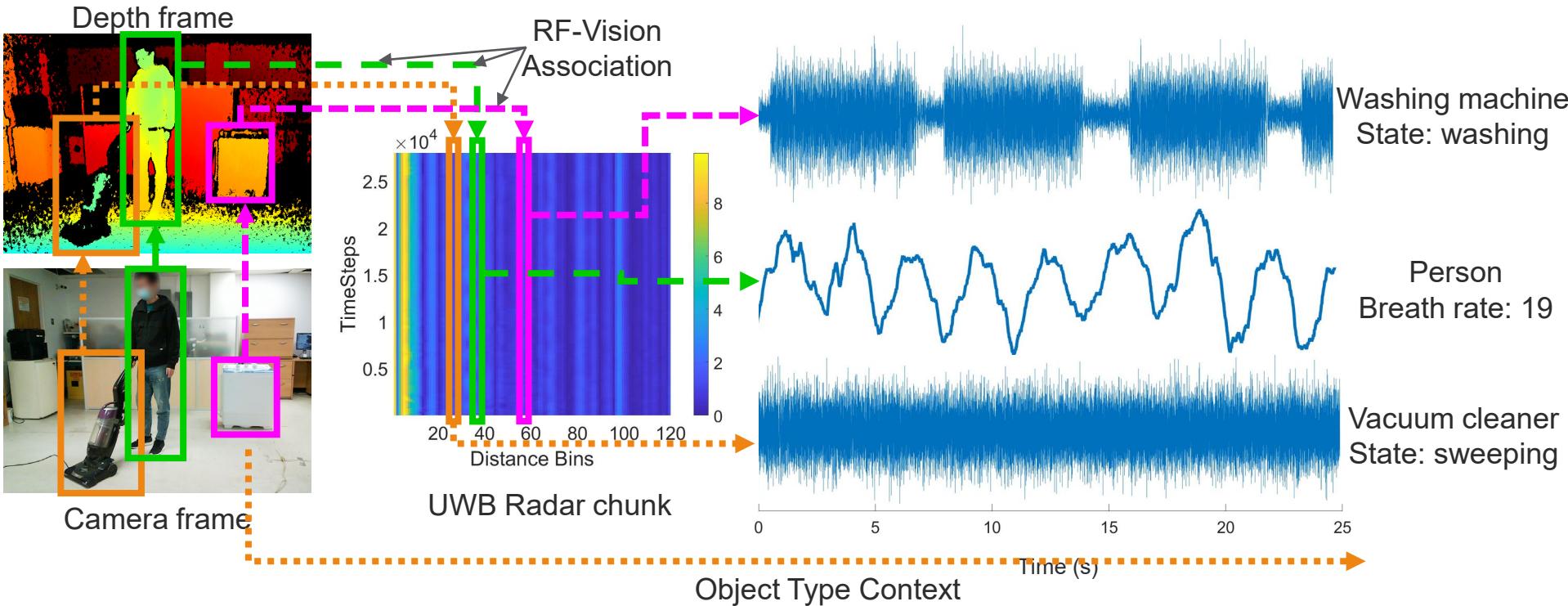
# Capricorn Design



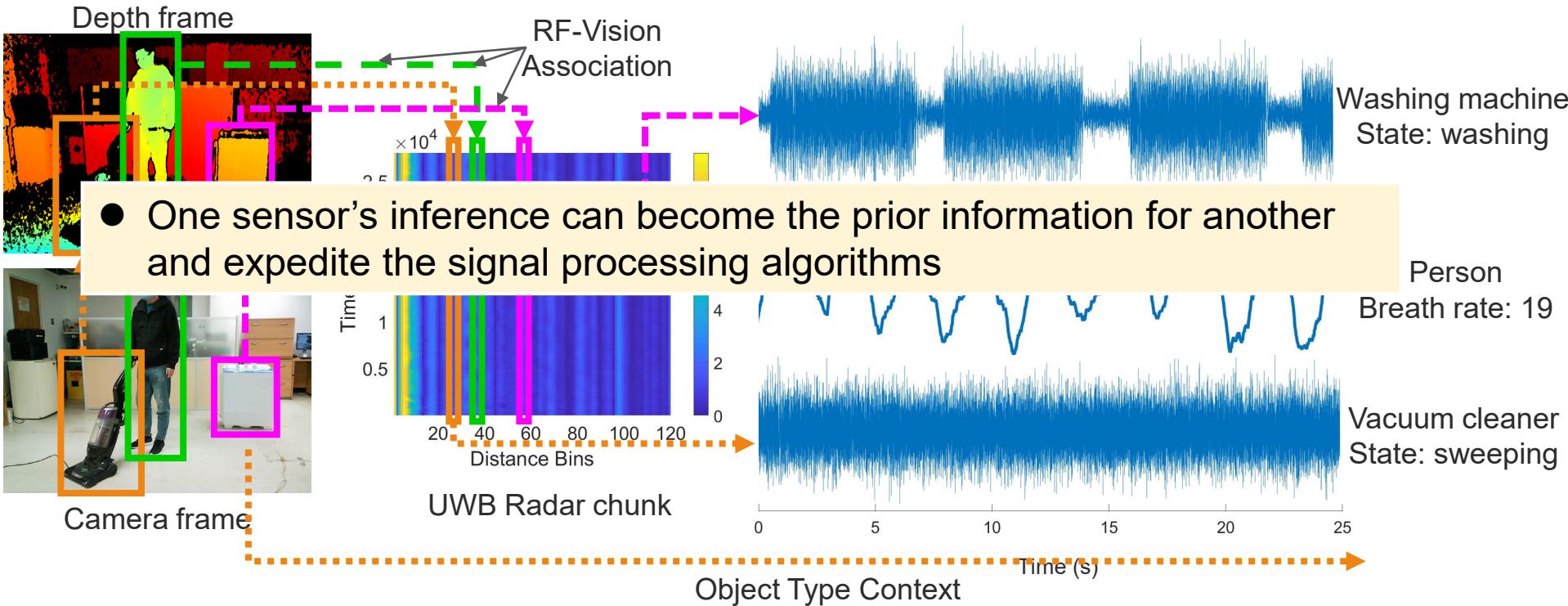
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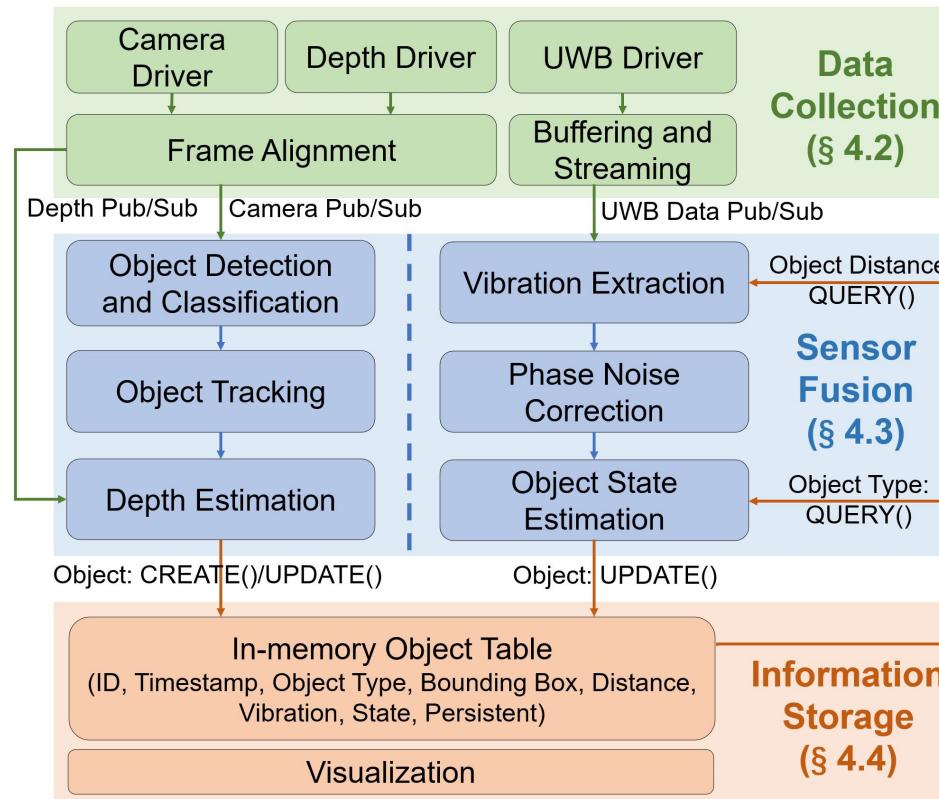
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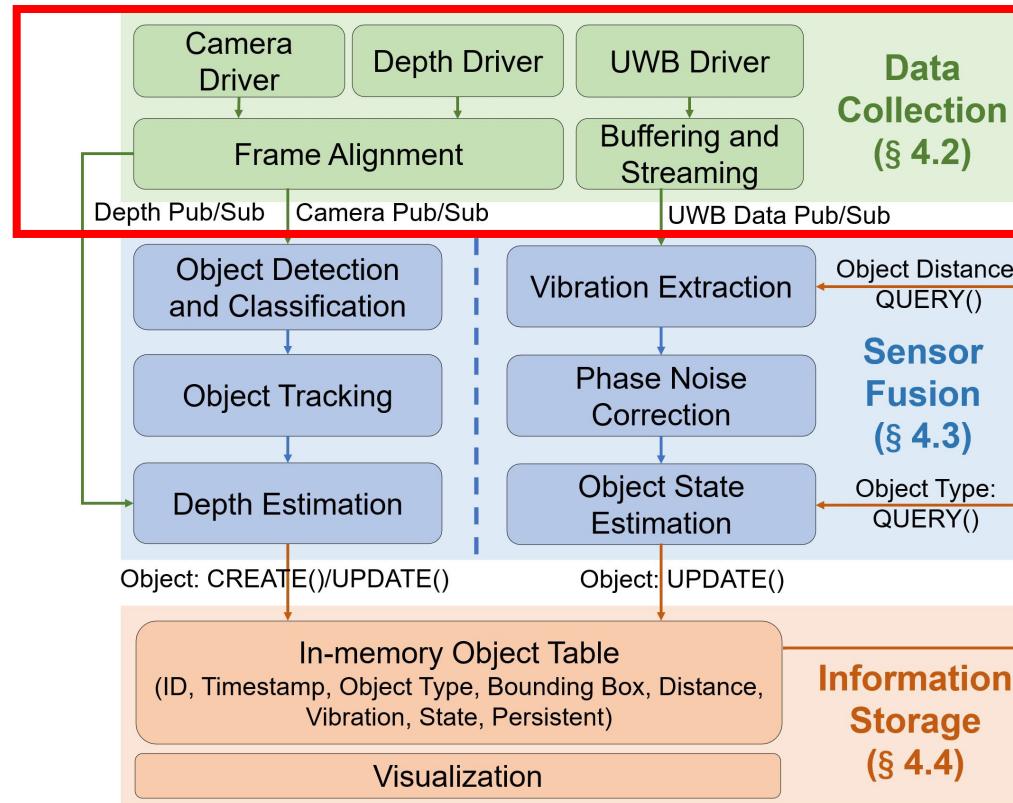
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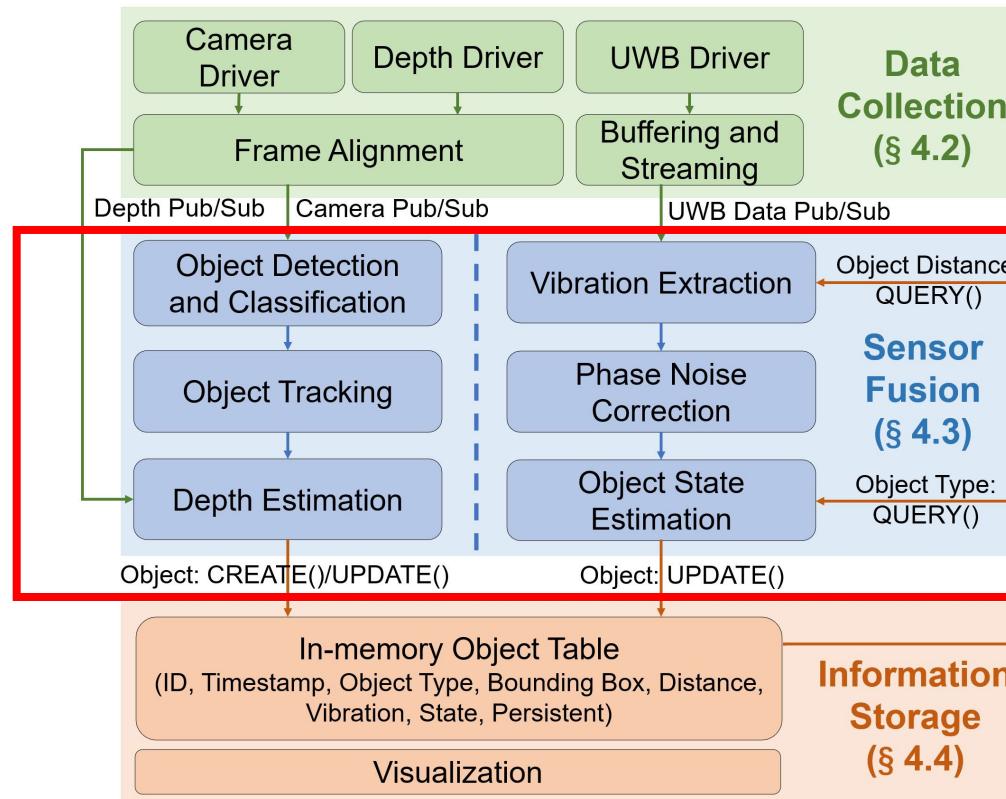
# Capricorn Architecture



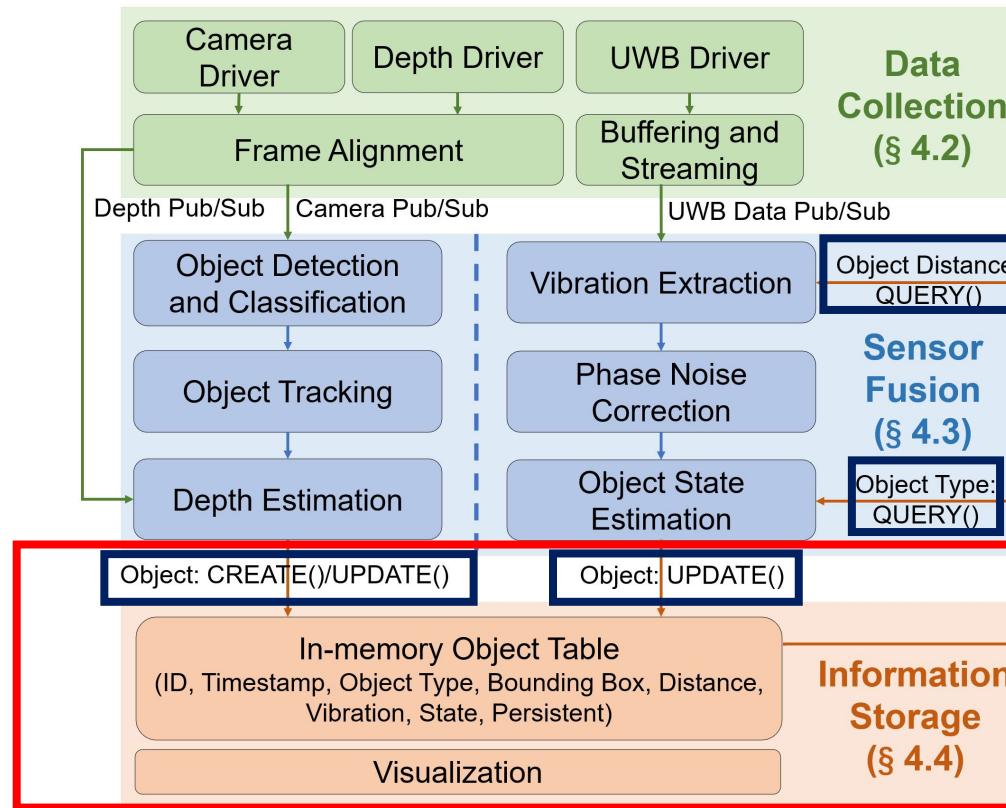
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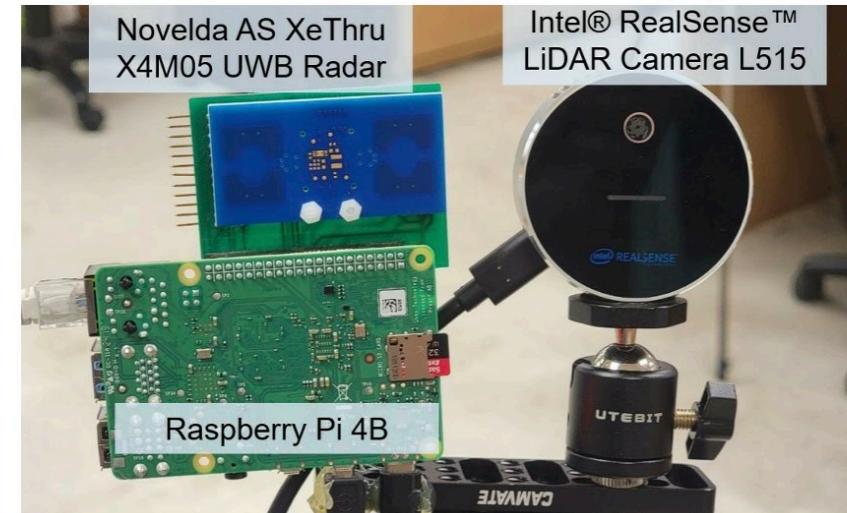
# Capricorn Architecture



# Implementations

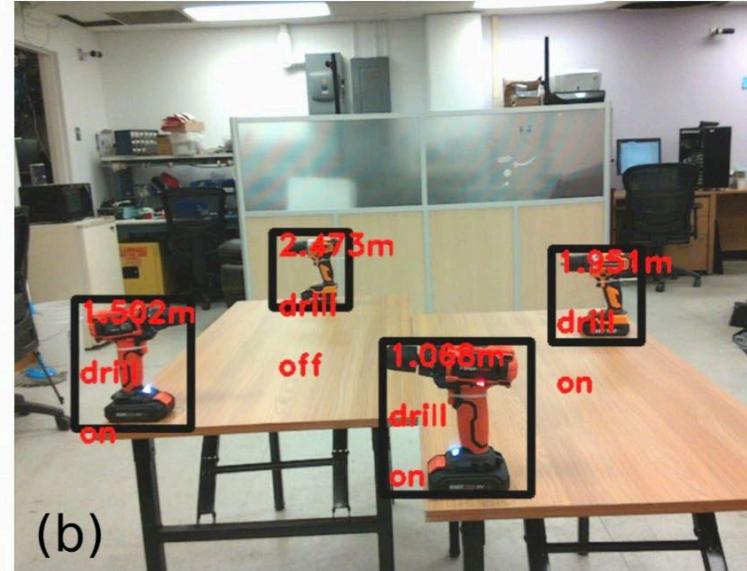
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- UWB Radar: Xethru X4M05
- LiDAR-Camera: Intel RealSense L515
- UWB Radar Host: Raspberry Pi 4B
  - Cortex-A72 Processor
  - 8 GB RAM
- Main Computation: Intel NUC
  - Intel i7-6770HQ CPU
  - 16 GB RAM
  - No GPU or any hardware accelerator



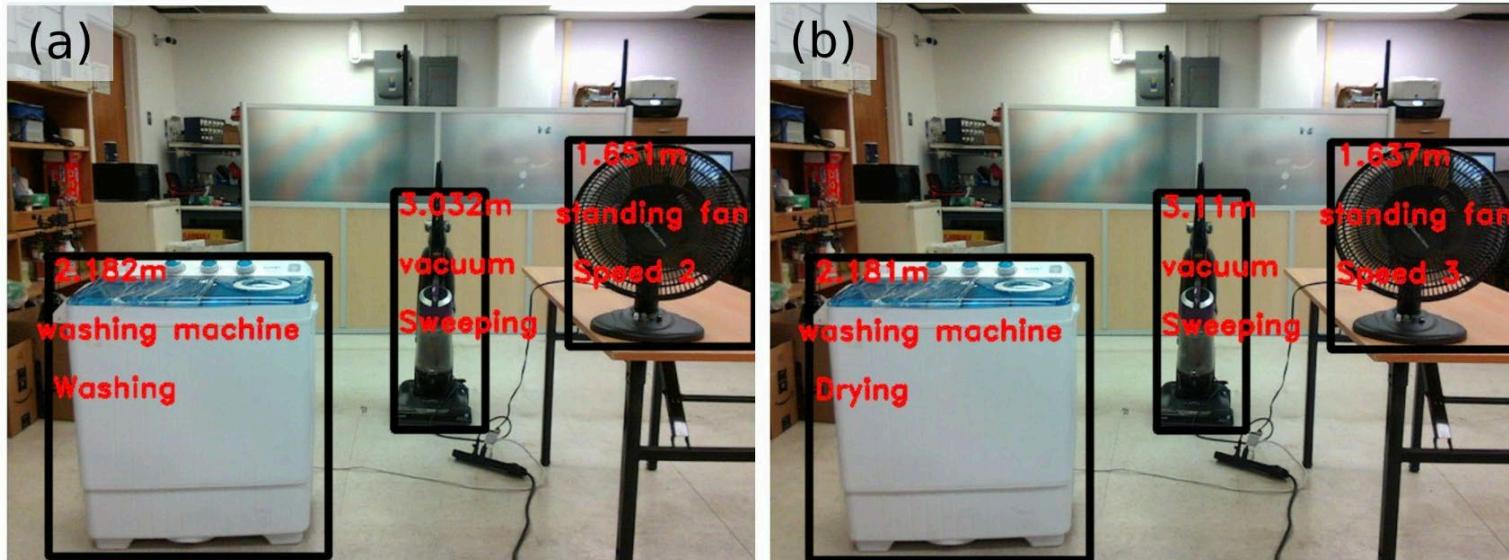
Hardware platform

# Sample Applications: It's all real-time!

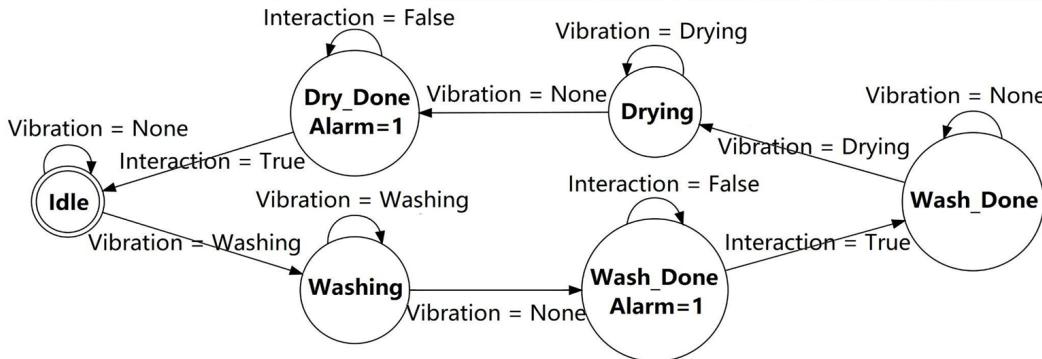


Workshop scene: drill state detection

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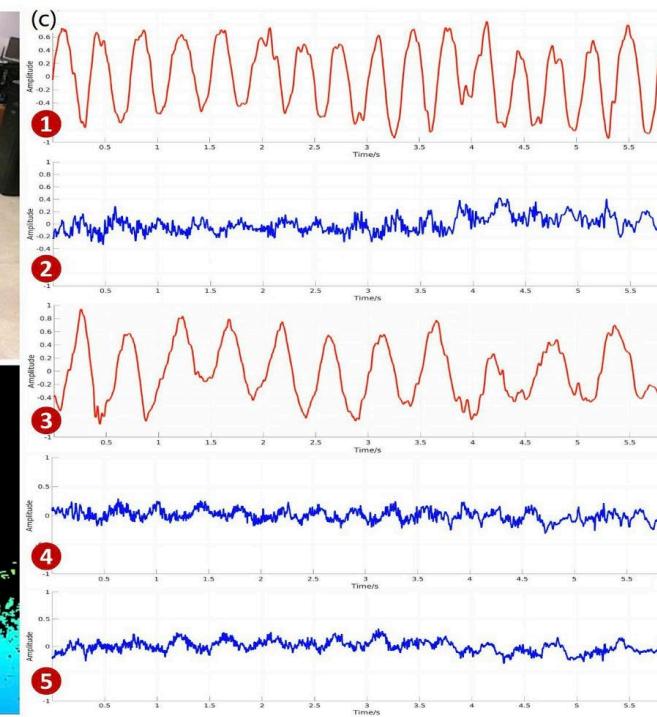
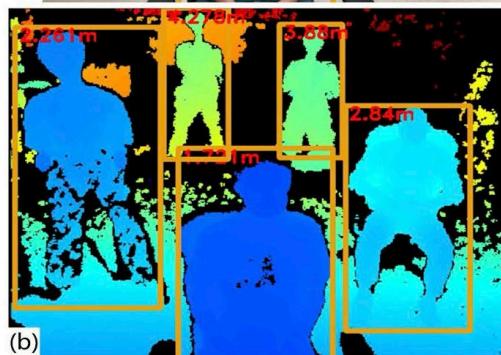


SmartHome scene: appliance usage tracking



Provide a **richer** set of  
**atomic events** for  
complex event detection

# Sample Applications: It's all real-time!

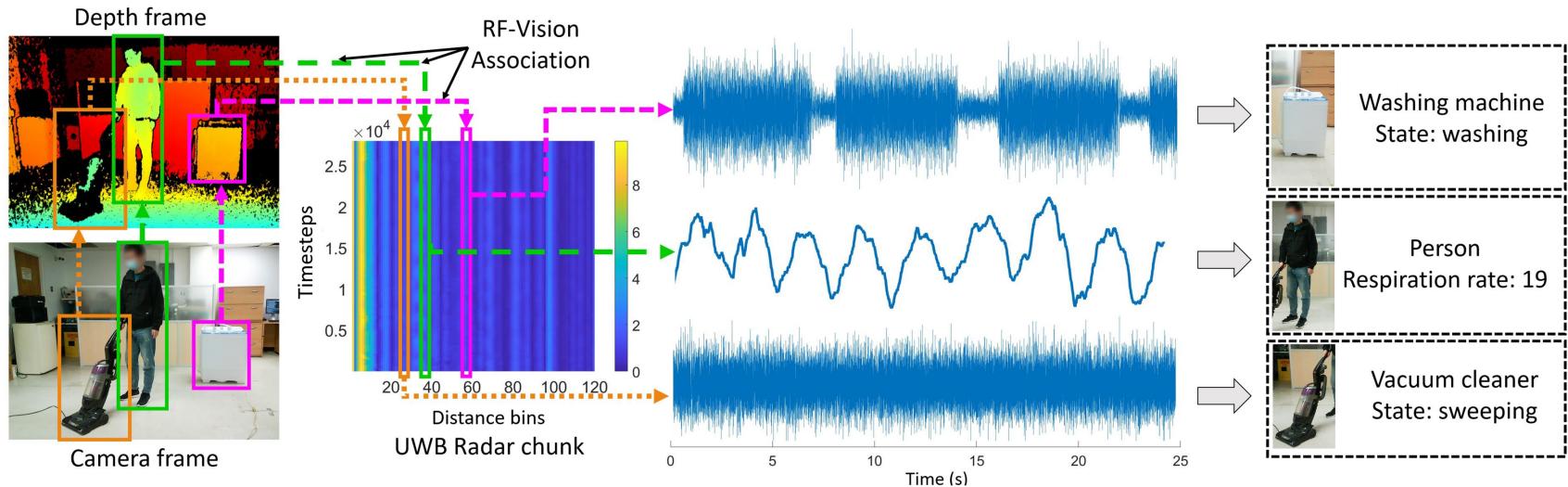


Multi-person respiration rate estimation

# Limitations and Future Work

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- **Integrate More Sensing Modalities:**
  - Currently we integrate a LiDAR camera and a UWB Radar
  - More sensing modalities will bring in richer information about the scene
  - e.g., Thermal camera, mmWave Radar, .....
- **Enable Mobility:**
  - Current system is mounted statically on a tripod
  - Mobility will give us more perspectives and a better scene understanding
  - SLAM on robots will enable the system to explore a totally new environment
- **Applications:**
  - Integrate sensor fusion information into AR/VR platform
  - Create an enhanced “digital twin” between the physical and virtual world
  - Neural symbolic-based complex event detection



**Thank you!**

**UCLA NESL**  
Networked & Embedded Systems Laboratory

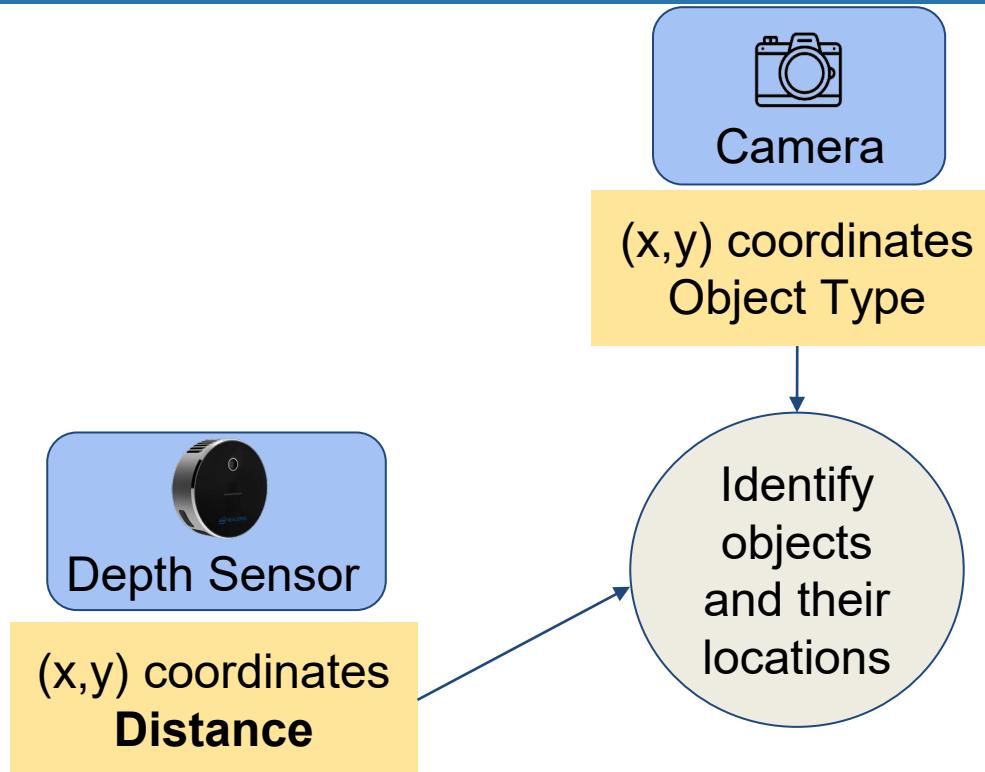


# Backup Slides

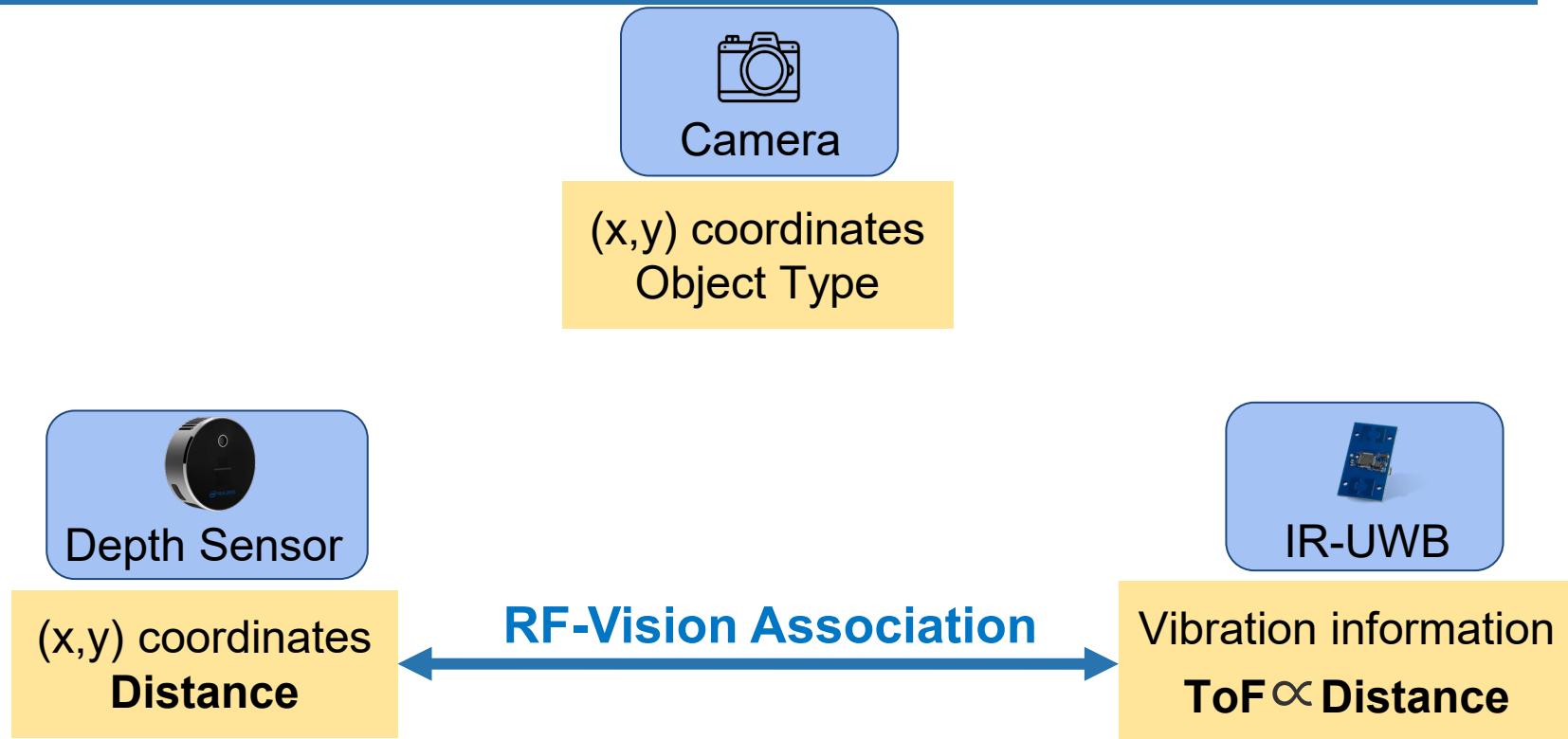
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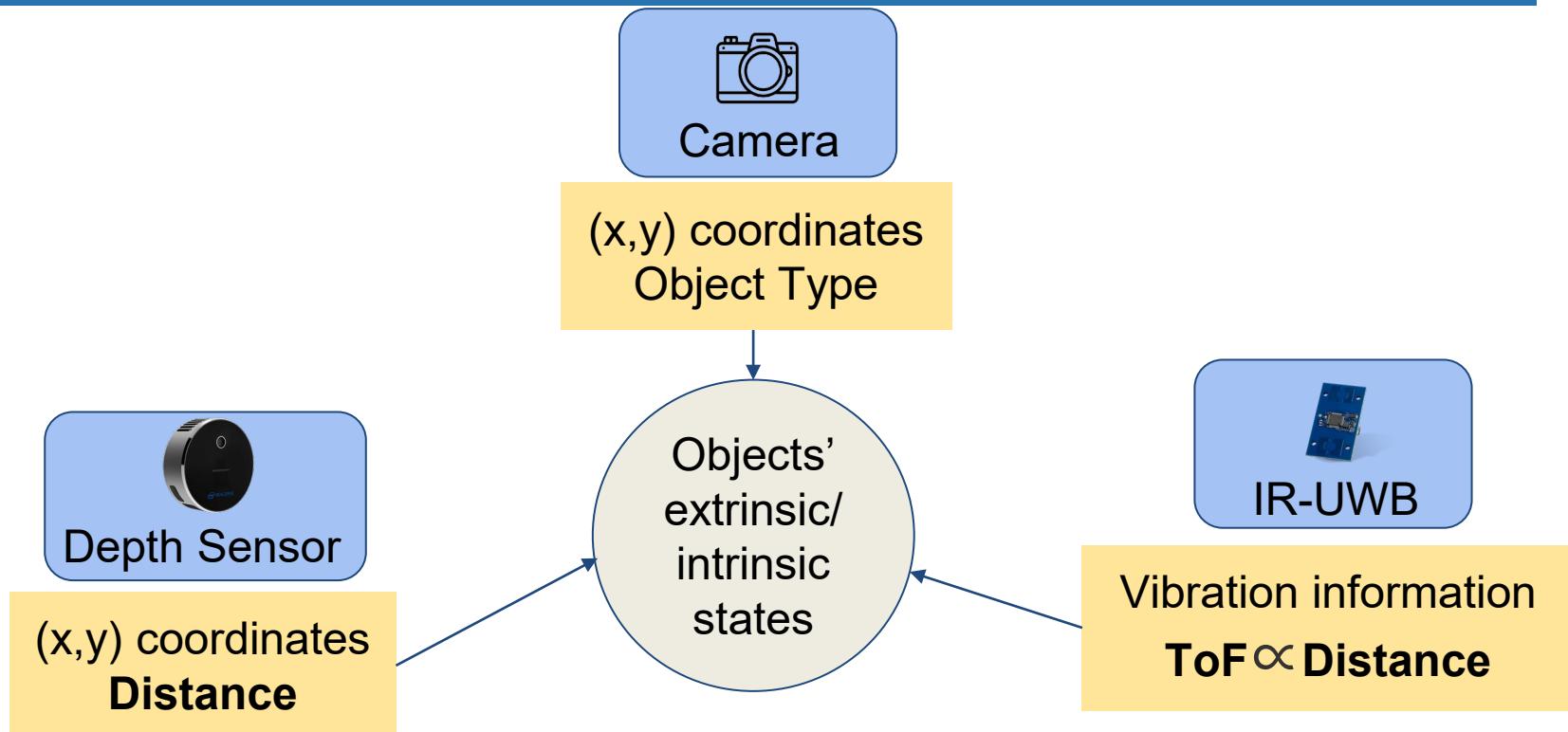
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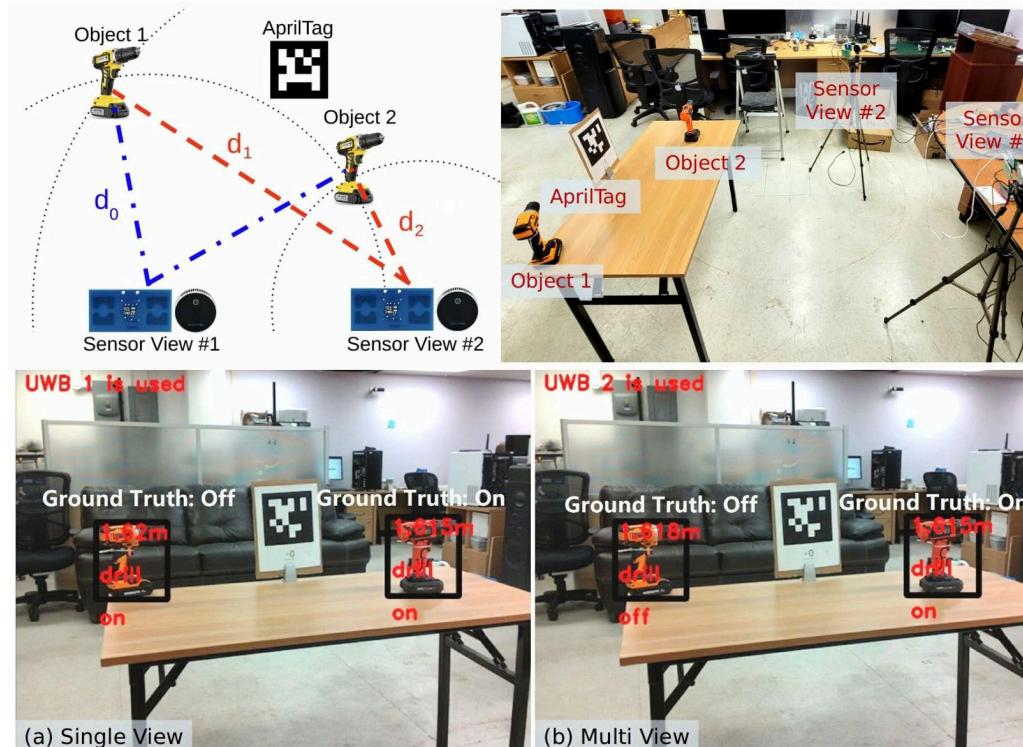
# Our Vision: Multimodal Sensor Fusion



# Our Vision: Multimodal Sensor Fusion



# Multi-view Capricorn



Multi-view version to distinguish objects at the same distance

# Back Up Slides: Latency

Capricorn Component	Mean(ms)	Std(ms)
Camera/Depth Pub-Sub Delay	1.08	0.13
YOLOv5	38.35	5.25
YOLOv5 (GPU)	6.28	1.29
Whole Extrinsic Sensing Pipeline	42.81	6.3
UWB Chunk Pub-Sub Delay	171.61	21.87

**Table 2: Latency analysis of Capricorn in the appliance usage classification scene.**

