

Beamsense: Rethinking Wireless Sensing with MU-MIMO Wi-Fi Beamforming Feedback

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AR/VR: Shaping the Future of Interactions

Global market value is predicted to reach 1066 Billion by 2034 [1]



We are very far from truly seamless and immersive
AR/VR experiences

What is the Current Bottleneck?

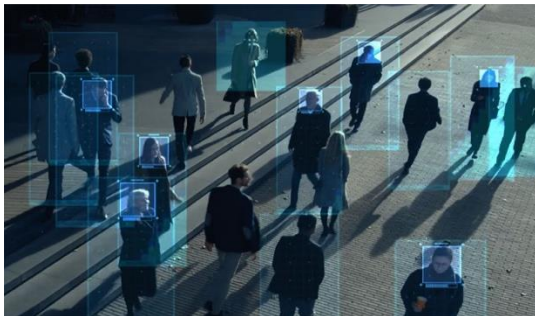
Semantic segmentation



Compute-intensive
AI/ML tasks



Object recognition



120 Hz frame
rate and 8K
resolution are
needed! [1]

State of Current Wireless AR/VR Headsets



Meta Quest 3
1.5-2 hours

Weight
515 grams

Refresh rate
Up to 90 Hz



Controllers



Apple Vision Pro
Up to 2 hours

Weight
650 grams plus
353 grams battery

Refresh rate
Up to 90 Hz



External
battery

Wireless Edge Computing and Sensing

Offload intensive compute task to edge server

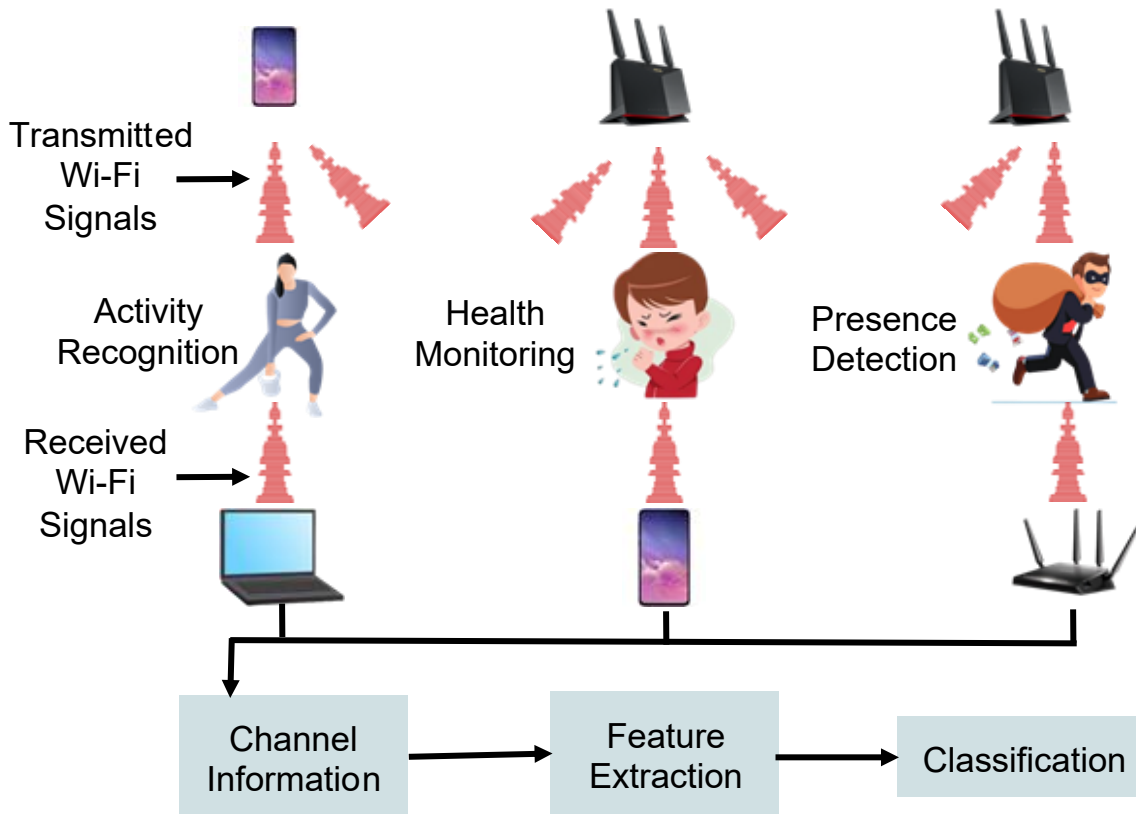


Wireless sensing for surrounding awareness

Light-weight & device free
AR/VR experience

Introduction on Wireless Sensing

- Perceived by the changes in the received signal
- Main approaches [1] are based on:
 - Received Signal Strength
 - Passive Wi-Fi Radar
 - **Channel Frequency Response (CFR)**



Prior Work and Existing Challenges

Need Firmware
Modifications

Need Dedicated CSI Tools

Limited Hardware

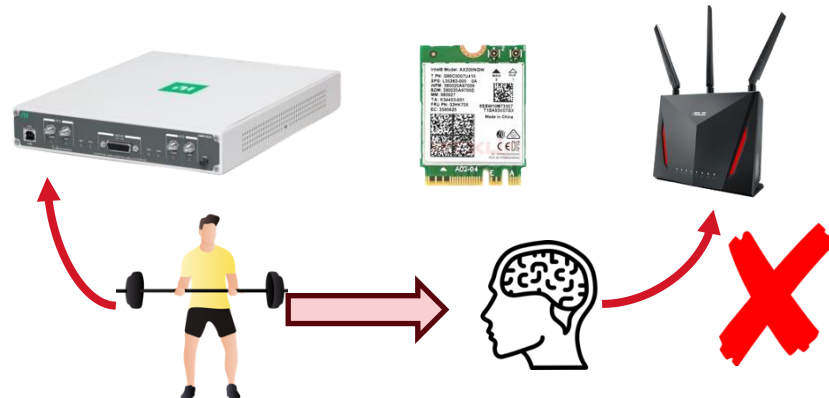
Can't Generalize Domains

nexmon

802.11n
CSI Tool

PicoScenes

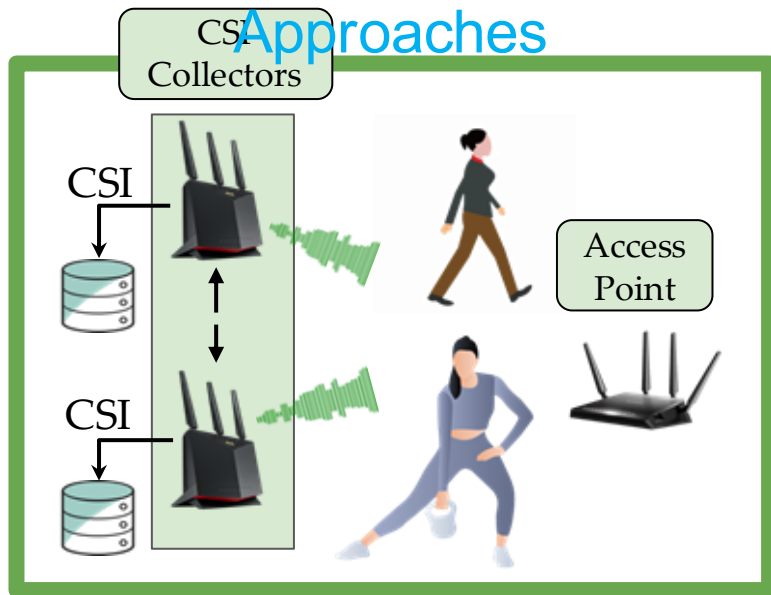
AX-CSI



- Bahadori, N., Ashdown, J. and Restuccia, F., 2022, June. ReWiS: Reliable Wi-Fi sensing through few-shot multi-antenna multi-receiver CSI learning. In 2022 IEEE WoWMoM (pp. 50-59).
- Meneghello, D. Garlisi, N. Dal Fabbro, I. Tinnirello, and M. Rossi, "SHARP: Environment and Person Independent Activity Recognition with Commodity IEEE 802.11 Access Points," IEEE TMC, 2022.
- R. Xiao, J. Liu, J. Han, and K. Ren, "OneFi: One-Shot Recognition for Unseen Gesture via COTS WiFi," in Proceedings of the ACM Conference on Embedded Networked Sensor Systems (SenSys), pp. 206–219, 2021
- Yin, G., Zhang, J., Shen, G. and Chen, Y., 2022. FewSense, towards a scalable and cross-domain Wi-Fi sensing system using few-shot learning. IEEE TMC, 23(1), pp.453-468.
- Na, Y., Zhou, G., Wang, S., Zhao, H. and Jung, W., 2018. SignFi: Sign language recognition using WiFi. Proceedings of the ACM IMWUT, 2(1), pp.1-21.

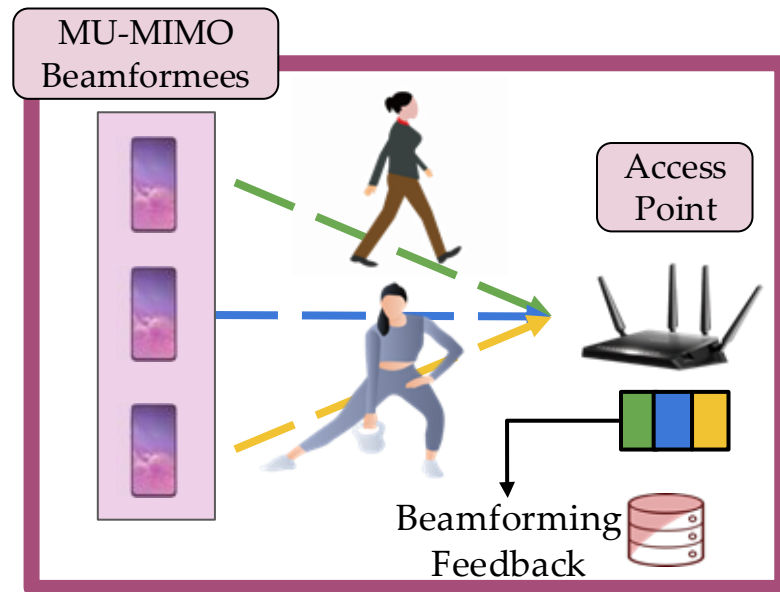
BeamSense: Overview and Novelties

Existing CFR based Approaches



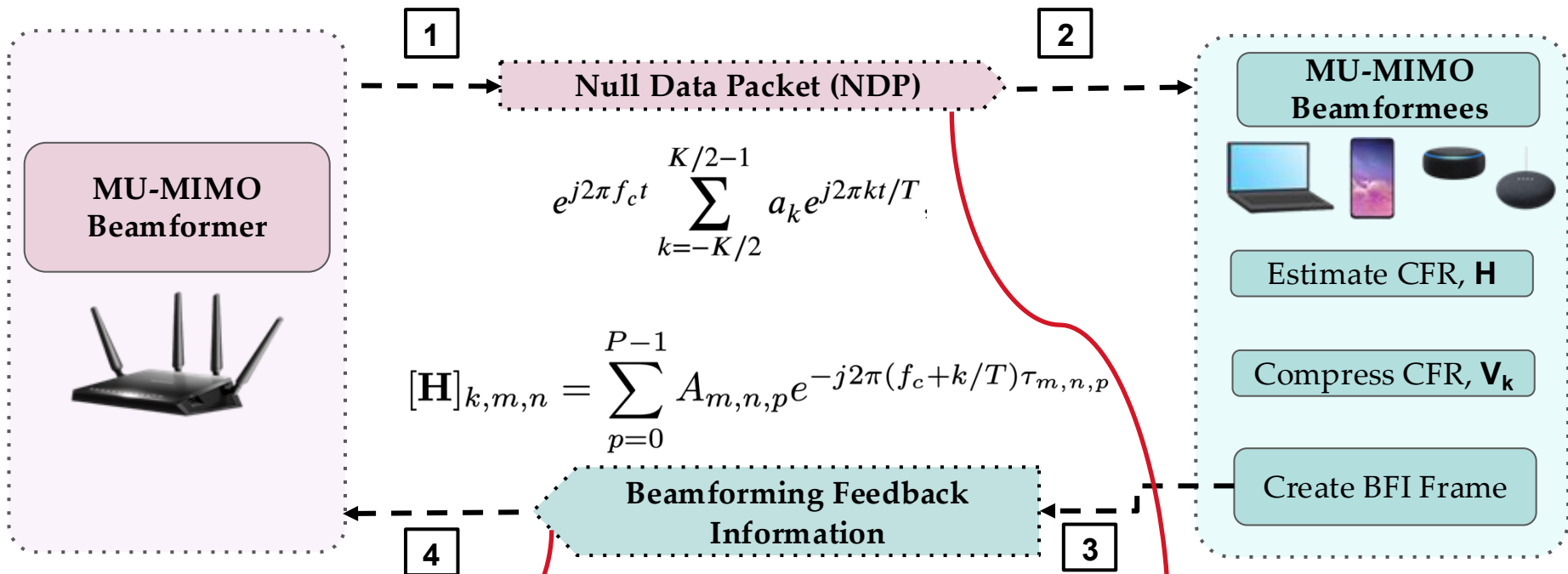
Requires specialized equipment
Requires Direct Access to Devices
Introduces additional overhead

BeamSense- based on BFI



No specialized equipment
No Direct Access Required
No additional overhead

From CFR to BFI



52 0.032155 BeijingX_76:65:3e ... 1a:e0:65:4e:ea:ca ... 802.11 61 VHT/HE NDP Announcement, Sounding Dialog Token=114, Flags=.....C

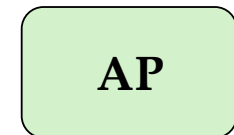
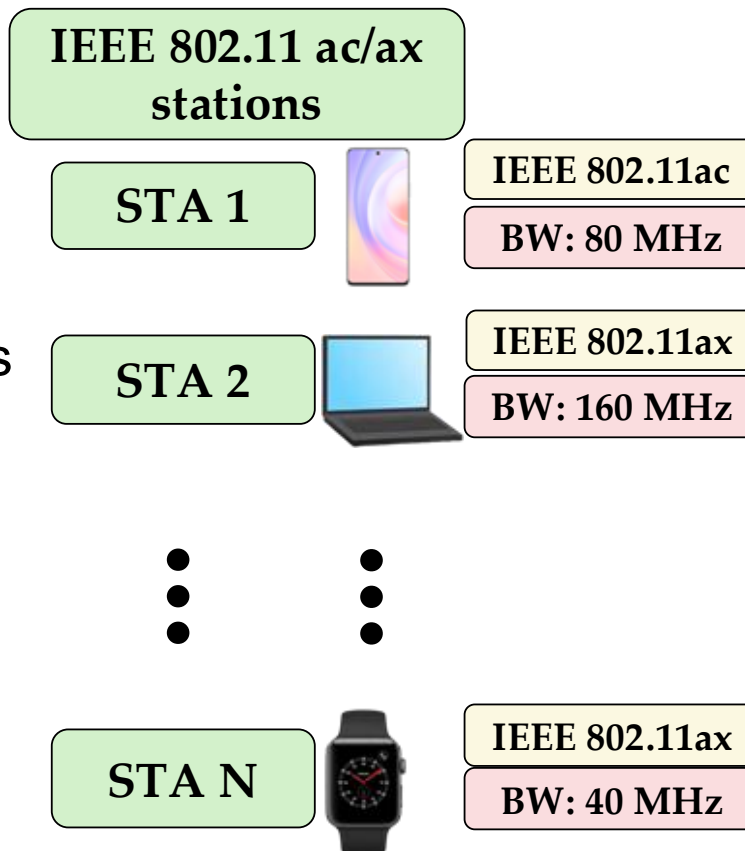
53 0.033717 1a:e0:65:4e:ea:ca BeijingX_76:65:3e 802.11 1011 Action No Ack, SN=28, FN=0, Flags=.....C

Transmitted unencrypted

Packet Bytes

Operation of BeamSense

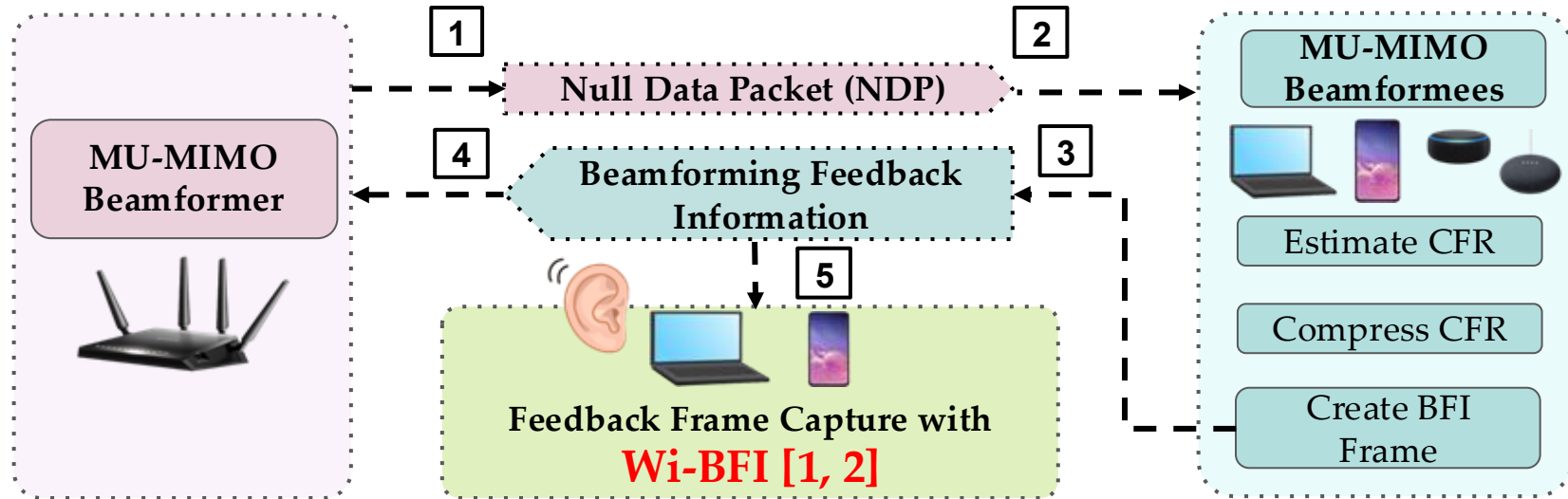
- Works with any device supporting MIMO
- Support across multiple standard
- Works with multiple bands simultaneously
- No direct access to the network needed
- Can be collected with a single capture



Edge Server

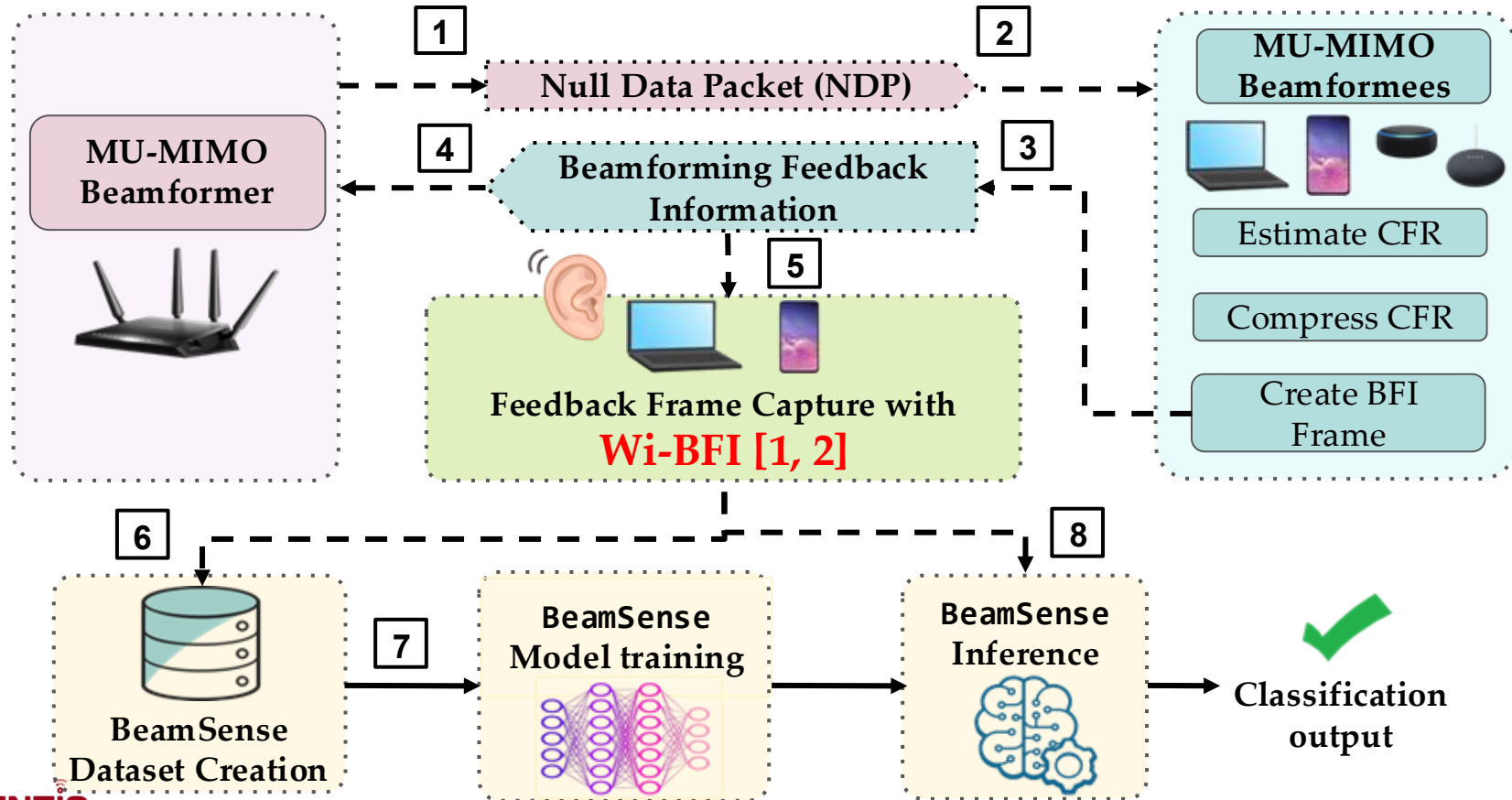


Capture BFI with Wi-BFI Tool

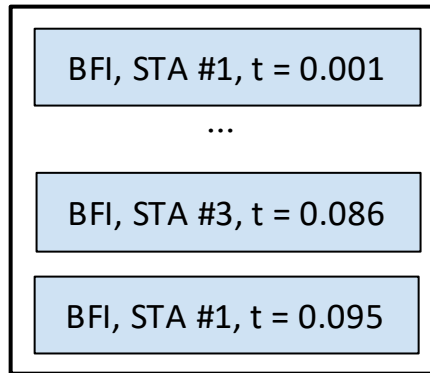


1. **Wi-BFI** captures the beamforming feedback packet
2. Decode the **Wi-Fi BFAs** based on the network configuration
3. Supports both **online and offline** extraction for **IEEE 802.11ac and 802.11ax**

BeamSense Walkthrough



BFI packets in a window

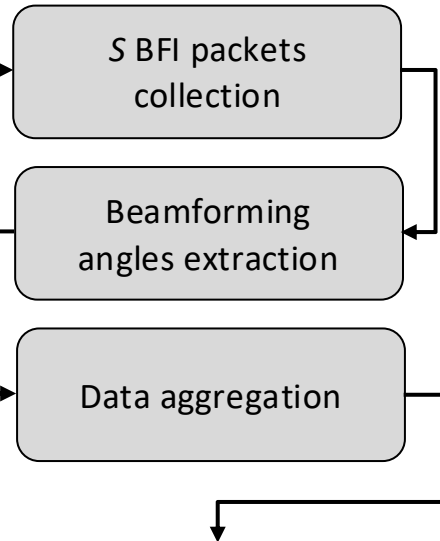


S: no. samples

K: no. OFDM sub-channels

A: no. BFI angles in a packet

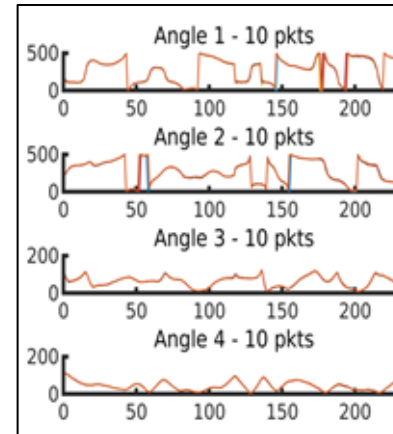
U: no. MU-MIMO users



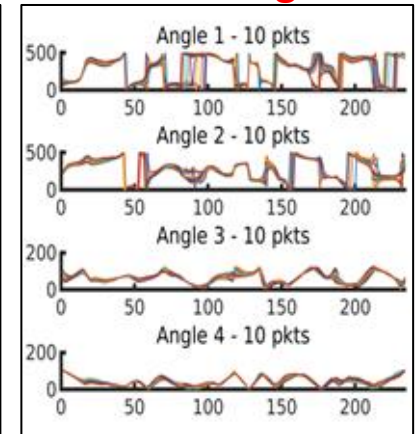
$(S \times K \times A \times U)$
Input Tensor

Each sample captures the variations of **Angle 1** over **subcarriers (234)** and **time (0.1s)**

Browsing phone



Walking

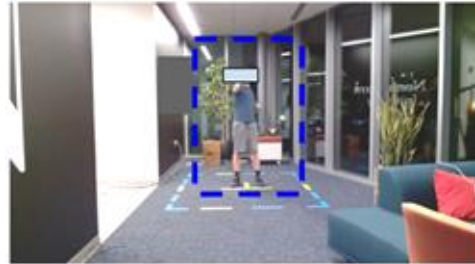


Sub-channels

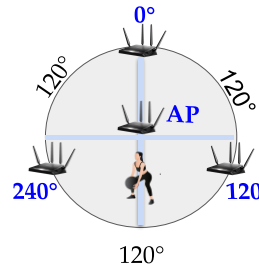
Sub-channels

Data Collection Campaign

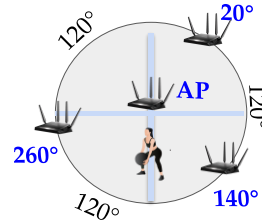
- Data collection campaign is done with 3 different persons, with 20 activities each in 3 different environment
- Data has been collected with 3 different orientations at different times of day
- Video Streams has been recorded as ground truth



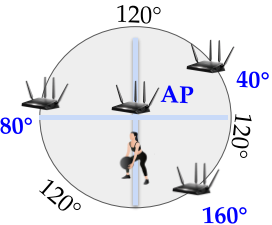
-  Kitchen $\Rightarrow$ Subject 1 $\Rightarrow$ side bend
-  Classroom $\Rightarrow$ Subject 2 $\Rightarrow$ reading book
-  Living room $\Rightarrow$ Subject 3 $\Rightarrow$ boxing



Orientation 1

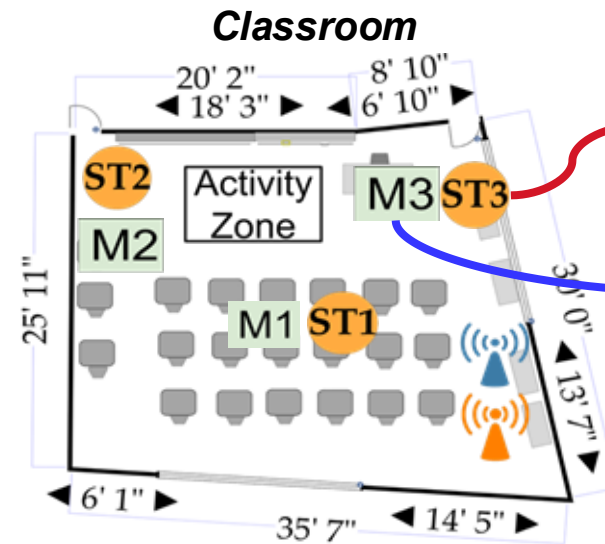
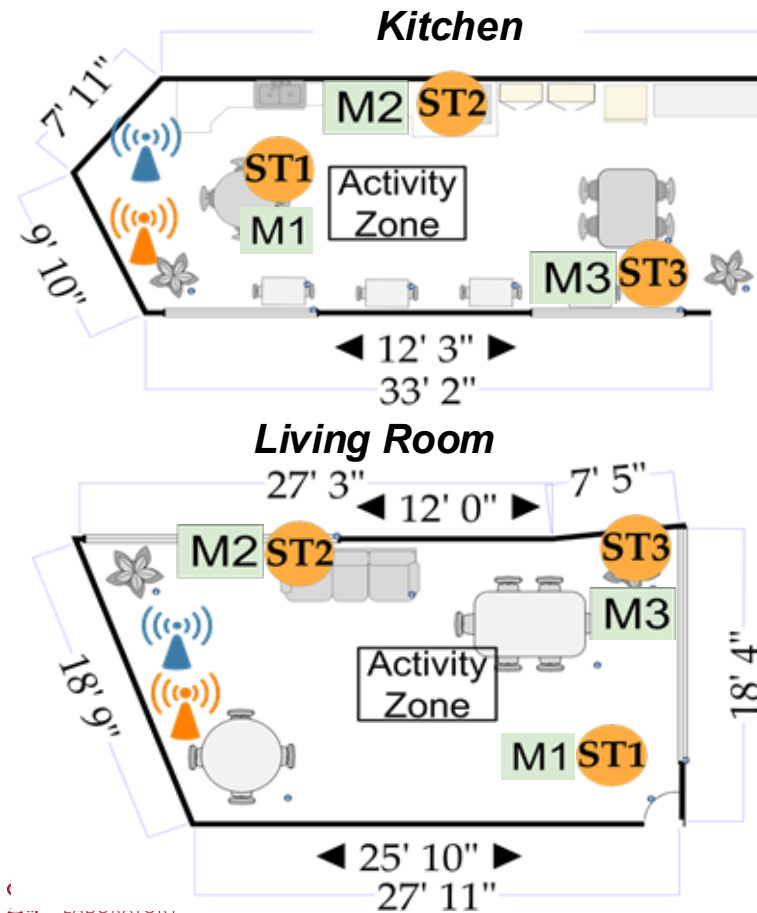


Orientation 2



Orientation 3

Experimental Testbeds



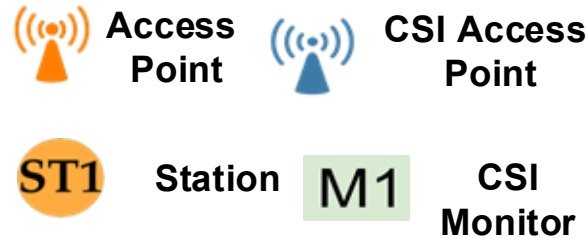
BeamSense STA:
Netgear X4S AC2600

CSI Monitor:
ASUS AC86U

Standard: IEEE 802.11ac

MU-MIMO Config: 3 X 3

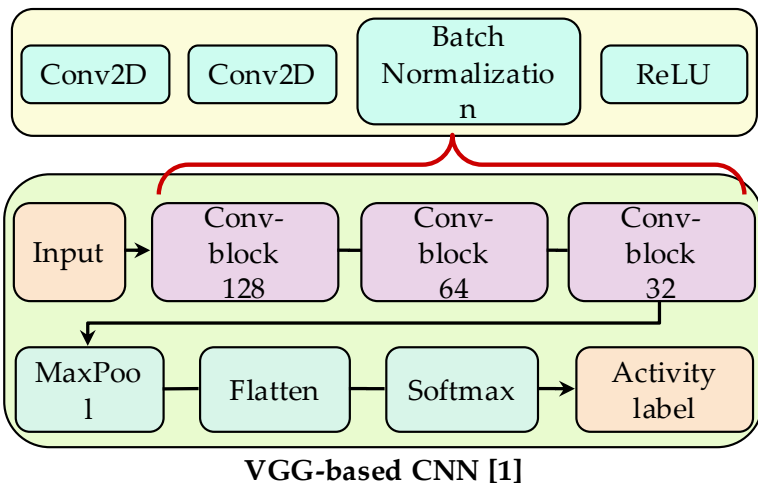
Channel BW: 80 MHz



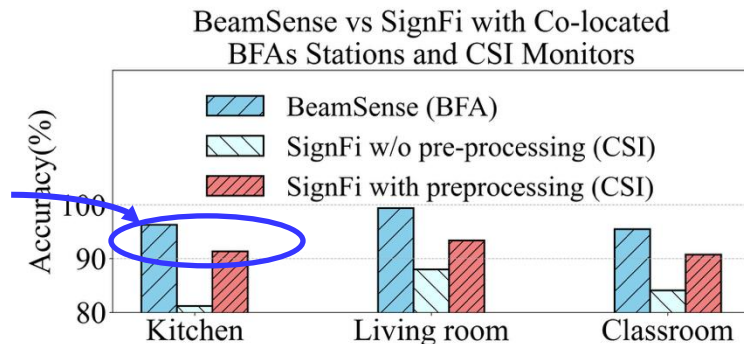
Performance with CNN Baseline

The baseline CNN consists of **three conv-blocks with 128, 64 and 32 filters**

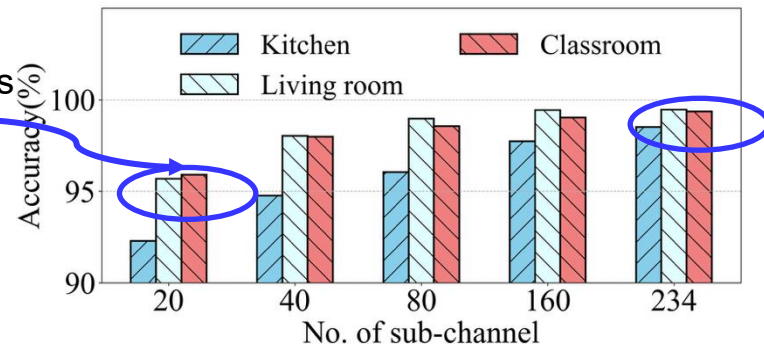
Conv-block



BeamSense improves the performance by **11%**, compared to **SOTA**



The accuracy decreases by only **4.52%** switching from **234 to 20 sub-channels**

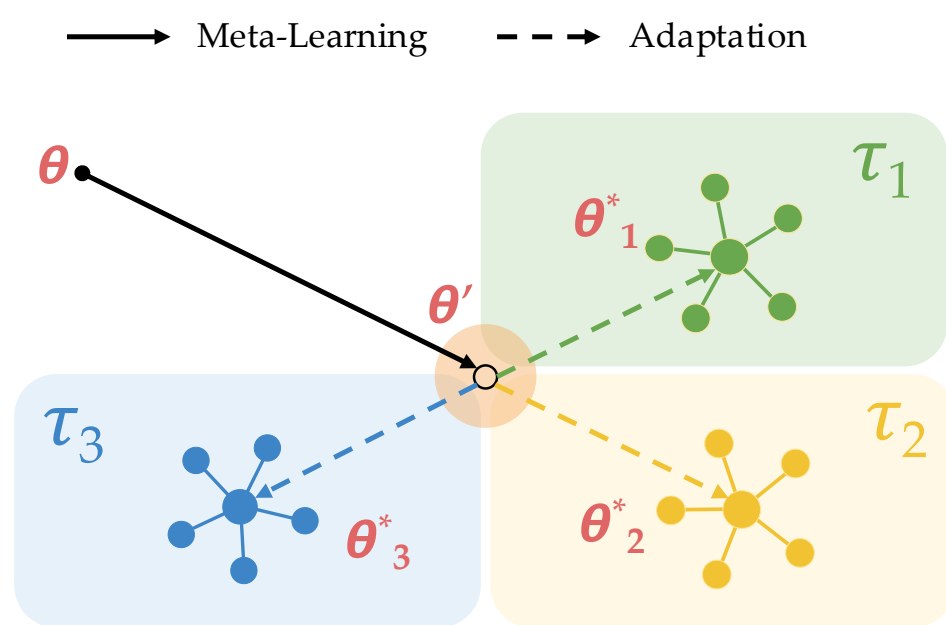


CNN cannot generalize to new domains

Meta-Learning through Few-shots

- Find common features among different tasks
- Meta-learning is the warm-up phase before learning
- Fine-tuning on a specific task is faster with meta-learned weights

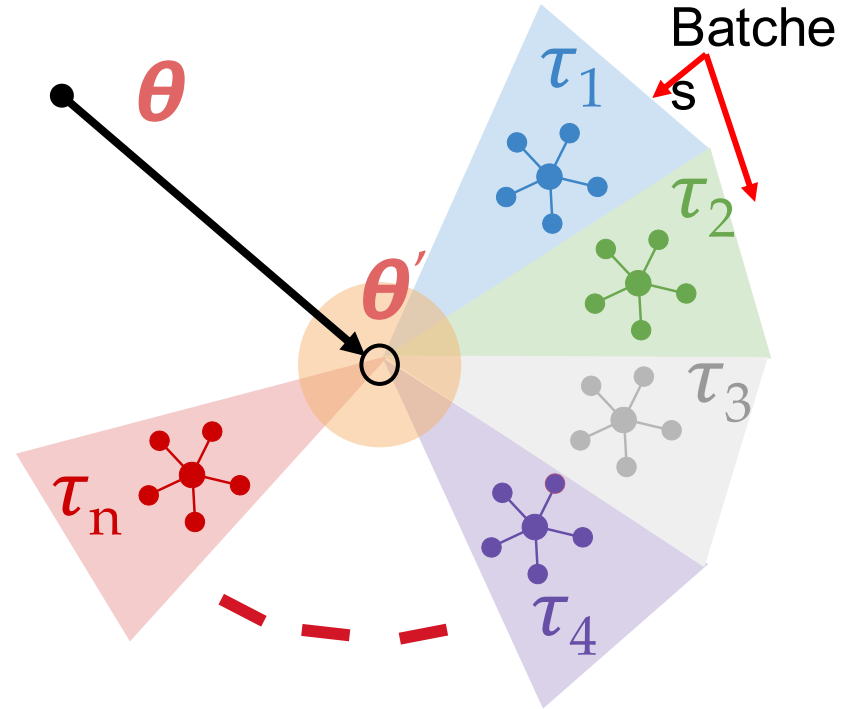
$$\begin{aligned}\theta' &= \theta - \beta \frac{1}{n} \sum_{\tau=1}^n \left(\frac{1}{m} \sum_{i=1}^m \nabla_{\theta} L_{\tau} [f(x_i, y_i | \theta)] \right) \\ &= \theta - \beta \frac{1}{n} \sum_{\tau=1}^n (\theta - \tilde{\theta}),\end{aligned}$$



- **Due to infinite number of tasks** Reptile can't be directly adopted
- We consider each batch of training data as a new task in meta-learning

Algorithm 2: The FAMReS Algorithm

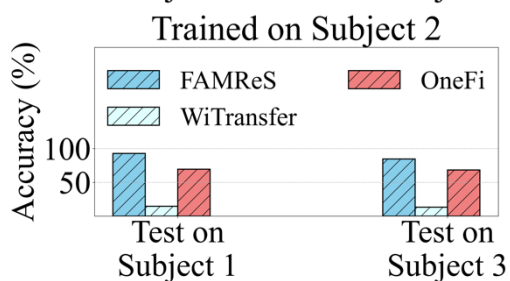
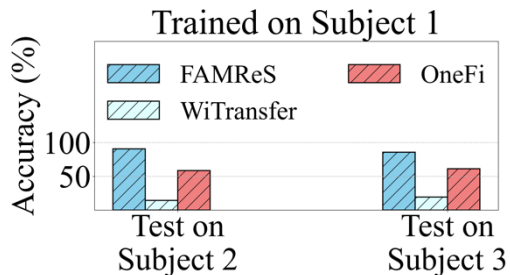
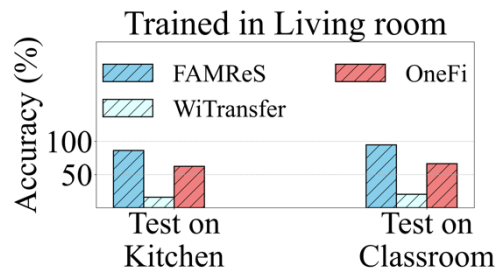
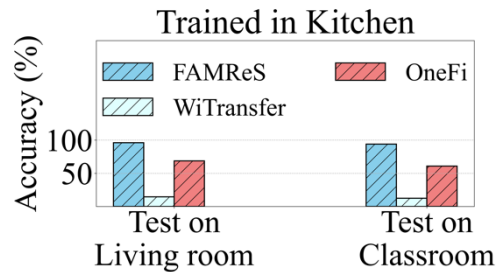
```
Require: step size  $\beta$ , micro dataset  $\mathbb{D}$ ;  
Initialize: a set of parameters  $\theta$ ;  
for  $iteration = 1, 2, \dots$  do  
    sample  $k$  points of data from  $\mathbb{D}$ ;           /*stage i*/  
    compute  $\tilde{\theta}$  using the SGD formulation;  
    update the parameters:  $\theta \leftarrow \theta + \beta (\tilde{\theta} - \theta)$ ;  
for  $epoch = 1, 2, \dots$  do  
    update  $\theta$  running SGD on  $\mathbb{D}$ ;           /*stage ii*/
```



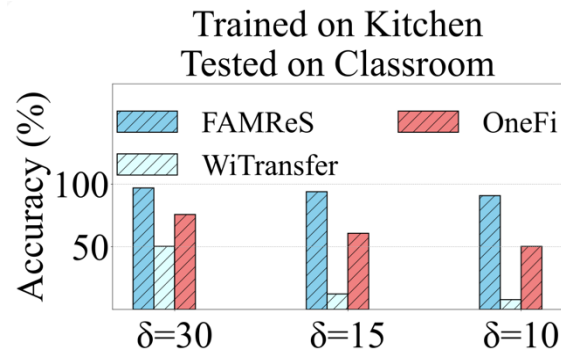
we only use a small portion of data in meta-learning and micro-learning for domain adaptation

Performance of Domain Adaptation

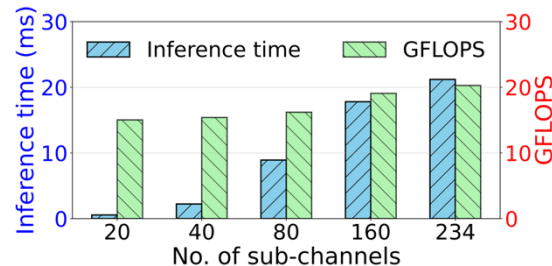
FAMReS improves cross-environment
and cross-subject accuracy by up to
35%



From 30s to 10s of data from new En
, FAMReS accuracy drop by only 5.3



20 GFLOPs of computation with
20 ms of inference time



Beamsense: Rethinking Wireless Sensing with MU-MIMO Wi-Fi Beamforming Feedback

- ✓ No hardware specific firmware, or firmware modification needed
- ✓ Data Extraction can be performed with any Wi-Fi compliant device
- ✓ No dedicated extraction tool needed with tight synchronization
- ✓ Can generalize to new environments, subjects or orientations



- Sensing without **no additional data at all** from new environment
- Sensing with **larger band to have more sensing granularity**

Thank You