

Securing Contrastive mmWave-based Human Activity Recognition against Adversarial Label Flipping

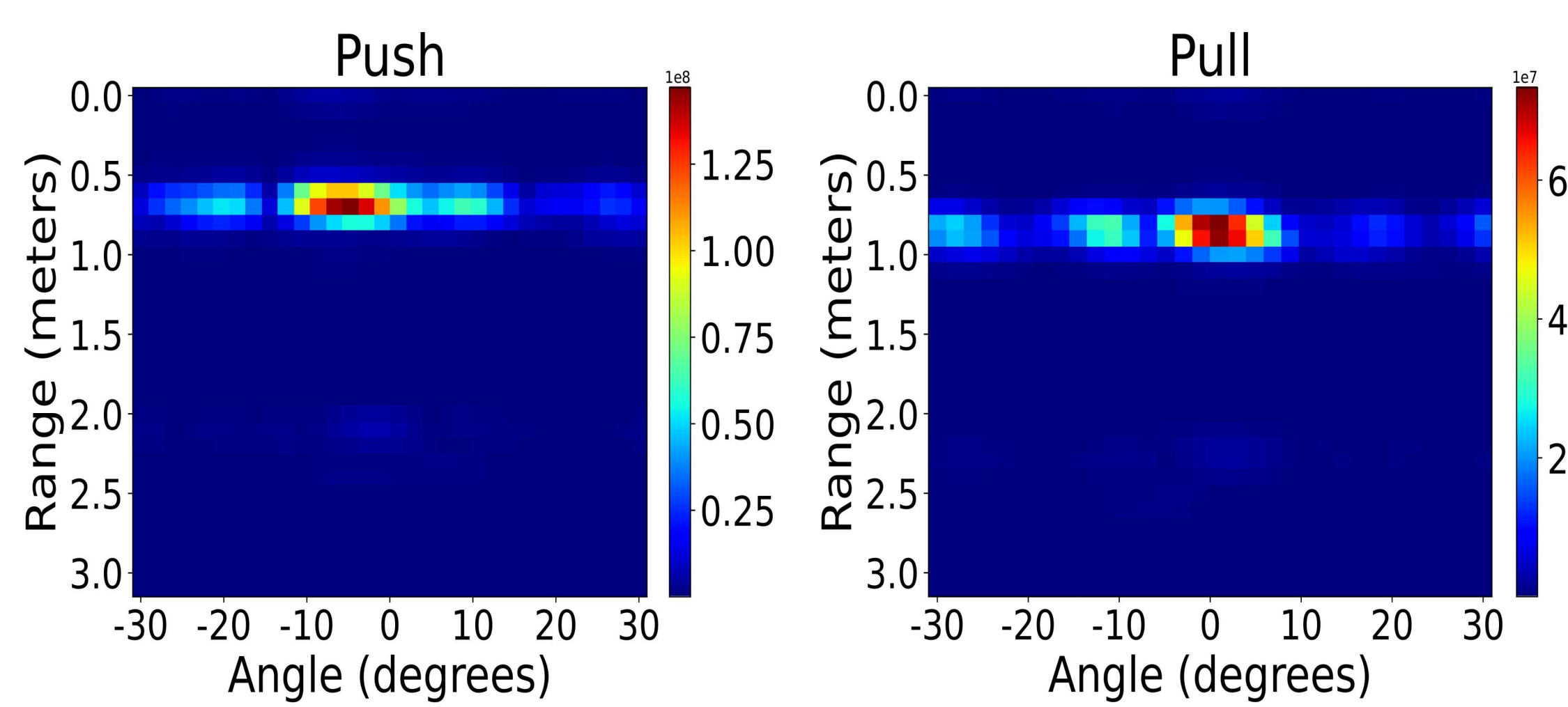
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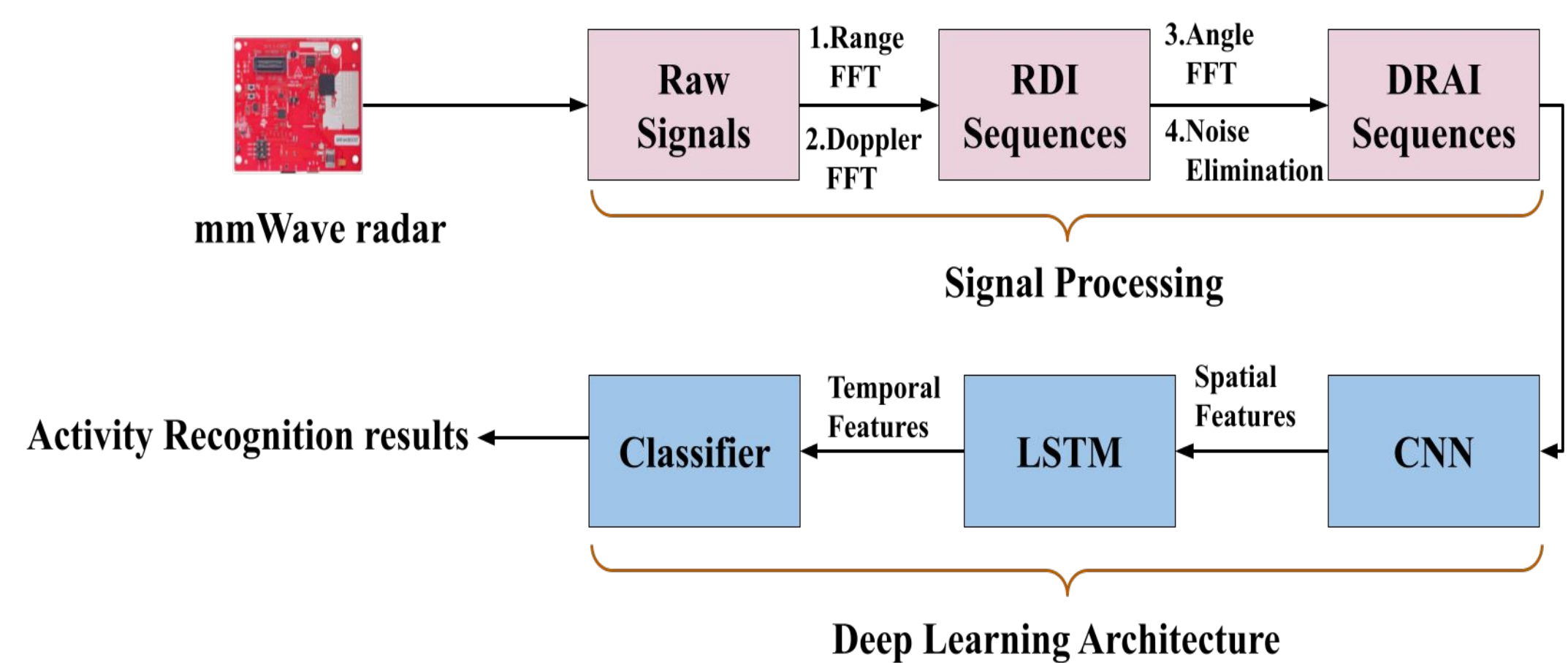
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The objective of this research is to identify vulnerabilities within mmWave-based Human Activity Recognition (HAR) systems to adversarial label poisoning attacks under supervised contrastive learning (SCL) frameworks. We identify three types of label poisoning attacks on contrastive mmWave-based HAR systems and propose a corresponding defense termed selective supervised contrastive learning (Sel-CL).

Heatmaps of two activities



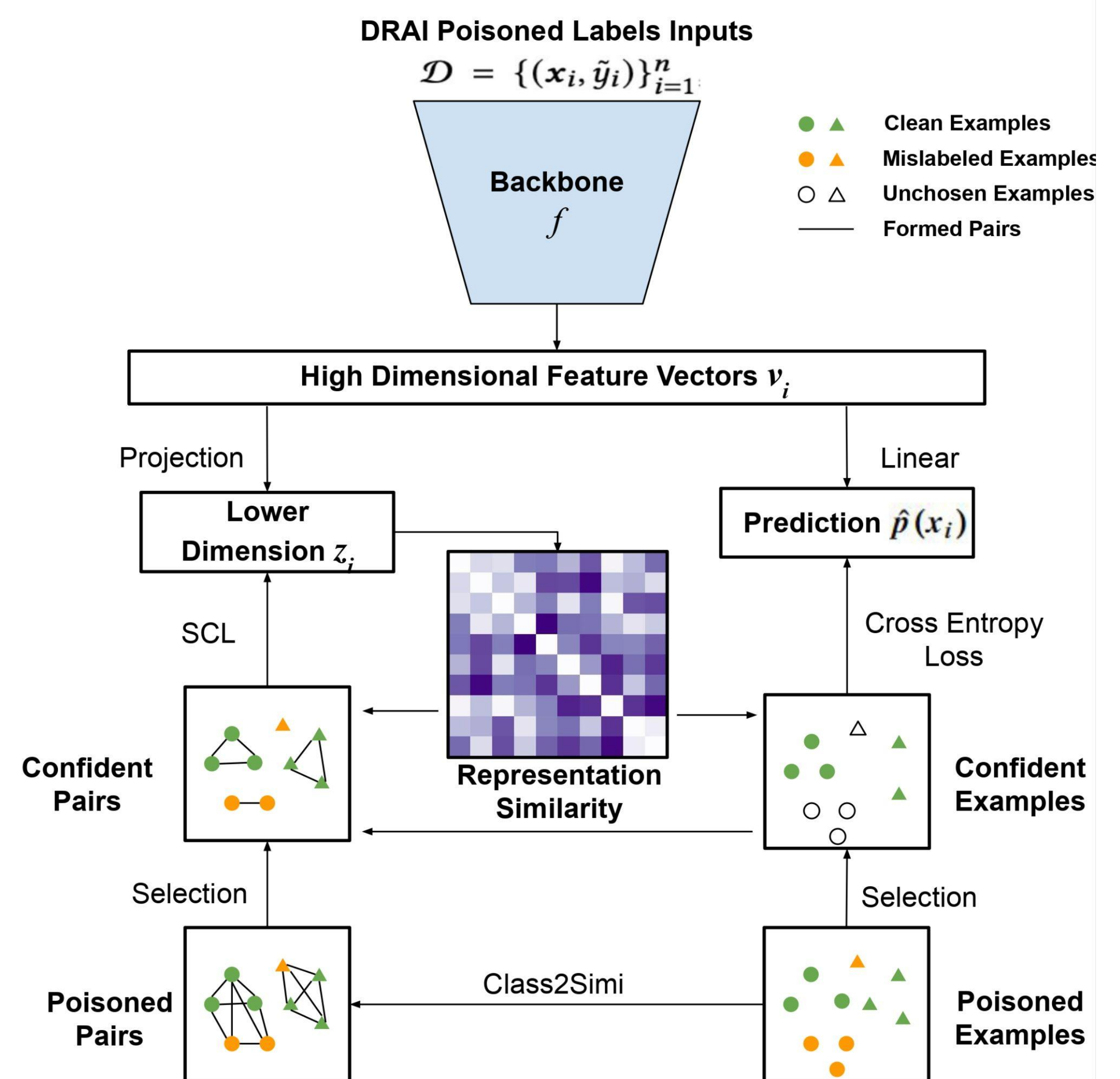
A basic mmWave-based HAR system



Attacks

- **Adversary Model:** Attackers aim to degrade HAR system performance by manipulating activity labels in the training dataset. This can be done through intentional data mislabeling, acquiring mislabeled data, or outsourcing training to a malicious third party.
- **Random Attacks:** Randomly alter labels in the training dataset to other arbitrary labels.
- **Across Trajectory Attacks:** Modify labels of activities to those of other activities with different trajectories.
- **Inner Trajectory Attacks:** Involve altering the labels of activities to those of other activities with similar trajectories, aiming to hide malicious tampering.

Defenses



Results

Sel-CL is highly effective against label flipping attacks, outperforming traditional SL and SCL.

