

WiFi Threat Detector

Deep Learning + RF Fingerprinting Framework

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Introduction

- **Wireless Threats:** MAC spoofing, Rogue Access Points, Beacon/Probe Flooding (DDoS), Passive Sniffers.
- **Why Existing IDS Fail:**
 - Signature-based → only detects known attacks.
 - MAC/IP anomaly detection → spoofable.
 - Lack RF-layer visibility.

Methodology

- 01 **RF Data Acquisition** – Collect I/Q samples, CSI, and RSSI using SDRs (HackRF, RTL-SDR) and Intel 5300 CSI Tool.
- 02 **Preprocessing** – Apply STFT for spectrograms, extract CSI phase shifts & RSSI variance, normalize and augment data.
- 03 **Feature Representation** – Generate spectrogram images for CNN and sequential CSI/RSSI inputs for LSTM.
- 04 **Deep Learning Model** – Train a hybrid CNN-LSTM network to capture both spectral fingerprints and temporal patterns.
- 05 **Evaluation & Validation** – Test on 10,000+ labeled signals across attack scenarios, achieving 98.9% accuracy with <300 ms latency.

Results

The hybrid CNN-LSTM model achieved 98.9% detection accuracy, outperforming standalone CNN (98.2%) and LSTM (97.5%) models. It successfully identified spoofing, rogue APs, DDoS, and passive sniffers with low inference latency (<300 ms) and strong generalization to unseen devices.



Use Cases

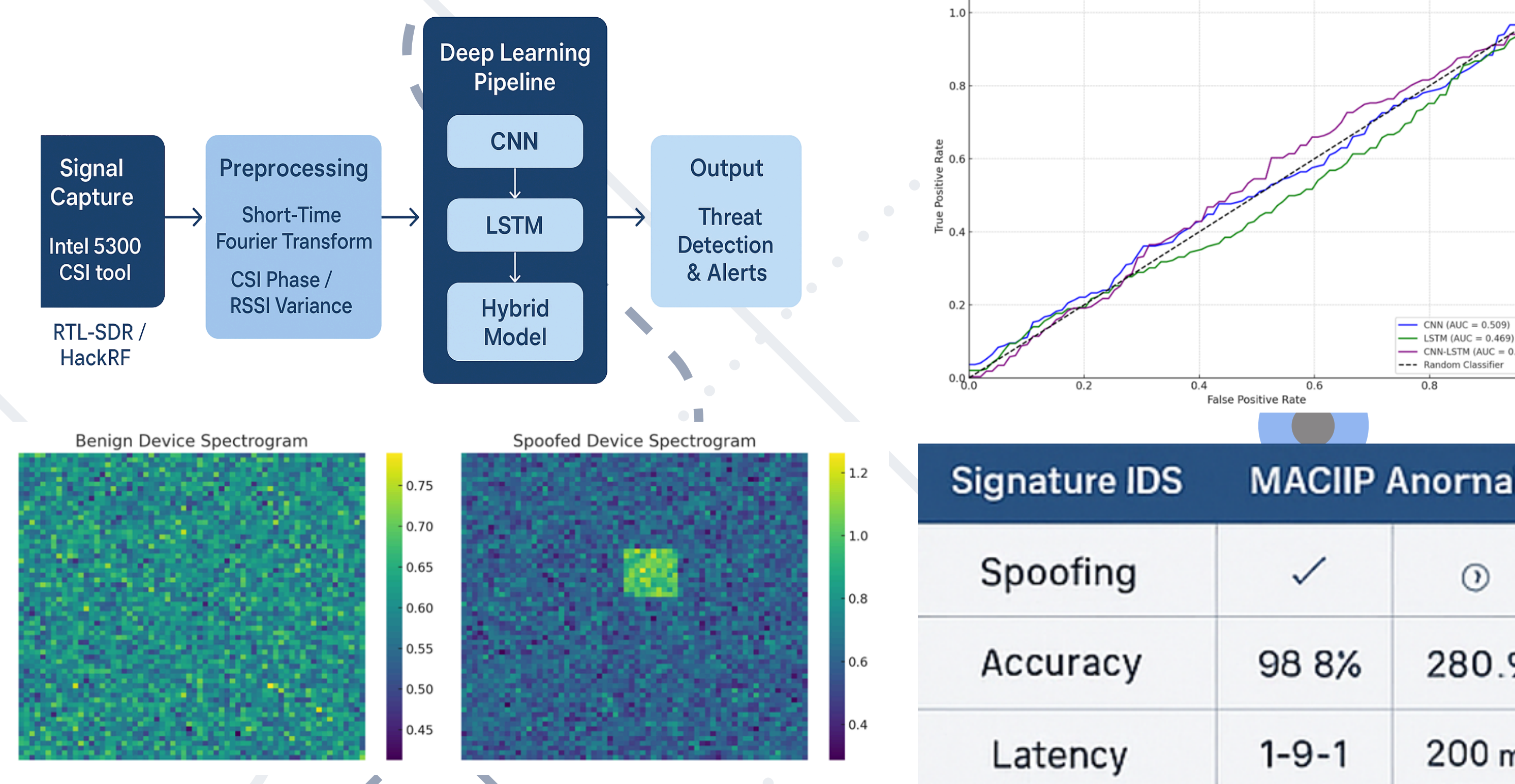
- **Smart Homes:** Detect spoofed IoT sensors
- **Smart Cities:** Protect EV chargers, drones, traffic sensors
- **Enterprises:** Defend against rogue APs & DDoS attacks

Conclusion

The WiFi Threat Detector demonstrates that combining RF fingerprinting with deep learning provides a powerful defense against stealthy wireless threats, achieving high accuracy and low latency. By securing the physical layer, it offers a scalable and explainable solution that strengthens smart homes, enterprises, and smart city networks.

Objective

This work aims to build a real-time Wi-Fi threat detection system that leverages RF fingerprinting and deep learning to identify stealthy attacks like spoofing, rogue APs, and DDoS at the physical layer. The goal is to provide a scalable, low-cost, and explainable security framework for smart homes, enterprises, and smart cities.



Key Sources