

Towards Deep Learning Augmented Robust Hem Regmi **D-Band Millimeter-Wave Picocell Deployment** Sanjib Sur

D-Band Deployment & Challenges

NextG wireless networks are difficult to design and deploy

- can increase the **capacity** of existing networks.
- Large number of **picocells** are needed to support high throughput demand of clients.
- Deploying **picocells** that are **operating** at **D-band** frequency has three major challenges.

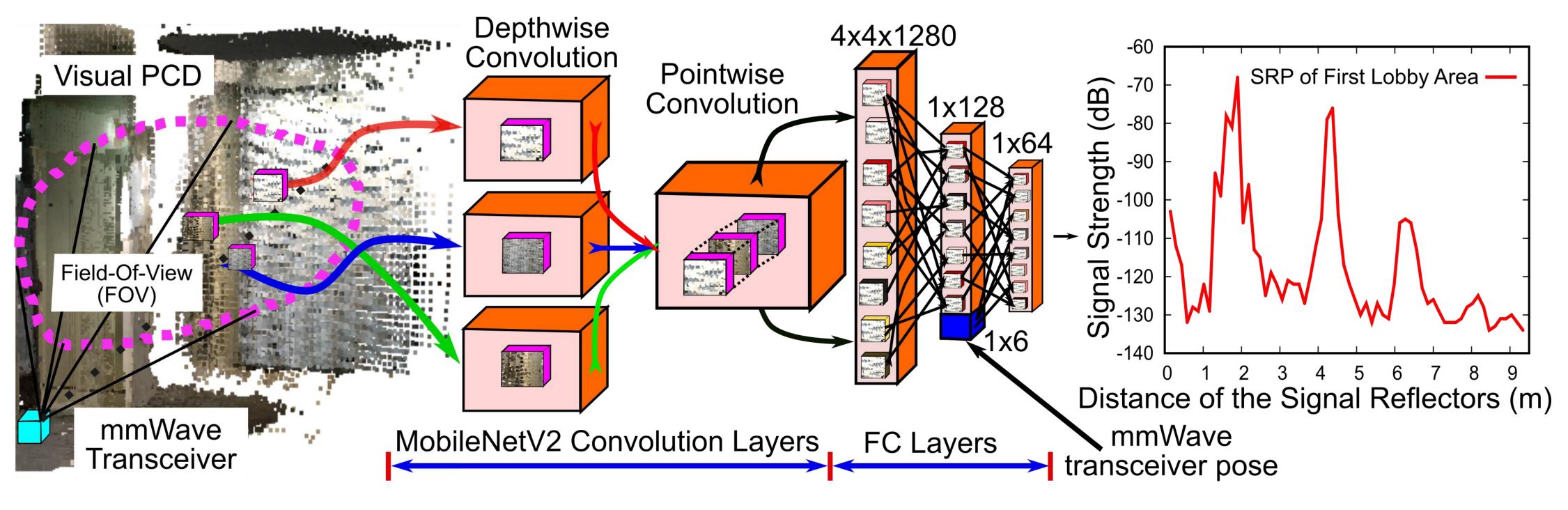
Challenges with picocell deployment at D-band

- of Non-Line-of-Sight (NLOS) path models at high frequency.
- Accurate and thorough manual site surveys might help for picocell deployment, but they are time-consuming and costly.
- Even **minor** changes in the **environment** highly impact the **effectiveness** of deployment and may require to update deployment locations.

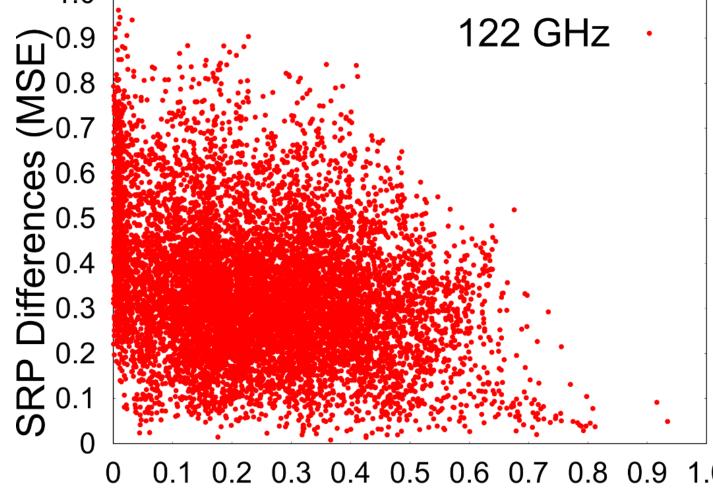
System Design

Deep Convolutional Neural Network (DCNN) accurately predict SRPs

- Preprocess visual **Point Cloud Data** (PCD) to obtain **Inverse Depth Image** for MobileNetV2 network.
- **Depthwise and Pointwise convolution** layers of **MobileNetV2** extracts highlevel abstract features from Inverse Depth Image.
- Fully connected layers use abstract features and mmWave transceiver pose to accurately **predict SRPs** at the output layer.



Millimeter-wave at D-Band (110 GHz to 170 GHz) offers wider bandwidth and



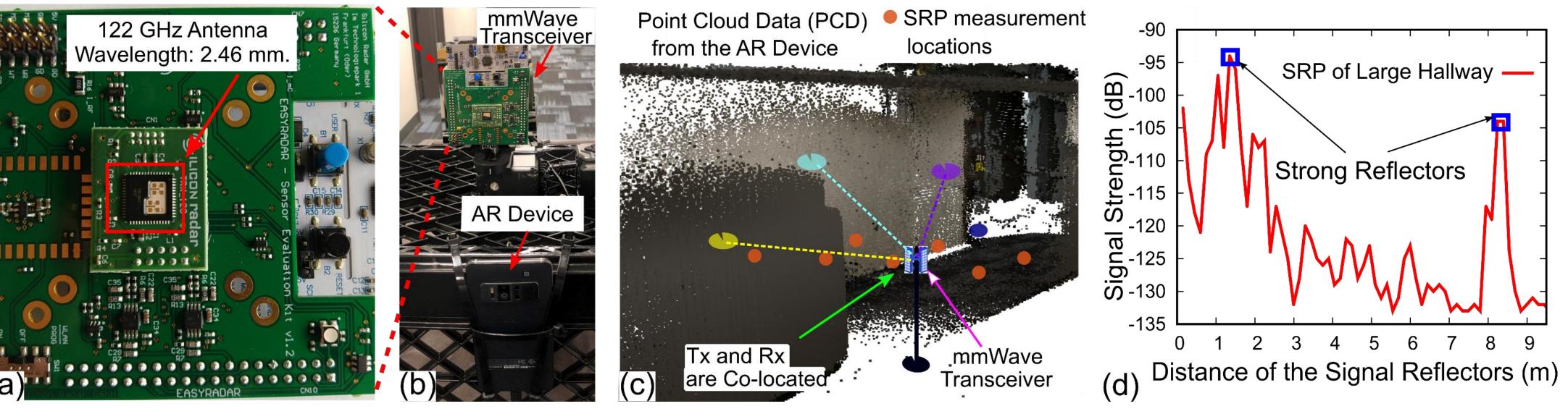
Visual Similarity (SSIM)

Ray-tracing simulation based **picocell deployment** are **inaccurate** due to lack

Our Approach

Can we deploy picocells at D-band more effectively?

- Empirical analysis across diverse environments show that visually similar objects produce similar millimeter-wave Signal Reflection Profiles (SRPs).
- Similarity between visual images and SRPs show the non-linear relationship, making it **difficult to model** their relationship (scatter plot of Block 1).
- Deep learning-based models can accurately learn the mapping between nonlinear data with limited number of environmental observations.
- A co-located mmWave and AR devices can quickly collect visual images and SRPs at random locations to train deep learning model and predict more SRPs.



Preliminary Results & Future Goals

DCNN model accurately predicts SRPs across diverse environments

- **Deep learning** model **doesn't require** extensive **manual site surveys** by accurately predicting SRPs at **unobserved** locations.
- Model predicts SRP at **D-band** with \widehat{m} an average **3.5 dB** median error.

Ongoing Work

- Include **semantic** awareness in the model to **improve** SRP prediction.
- Find **best picocell locations** at **122** GHz for diverse environments.

Future Goals

- Perform "what-if" analysis of picocell deployment.
- Explore 122 GHz and 24 GHz joint picocell deployment and simulate reconfigurable intelligent surfaces to maximize coverage.

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