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Emerging Hardware Architectures for AI- Enabled Spectrum Systems

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Initiatives

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NSI Vision and Approach

Vision

Accelerate the delivery of advanced capability needs of the Department of War and the Intelligence Community through leading-edge, high impact research and workforce development

Approach

- **Applied Research**
Coordinate world-class capabilities of NSI and CU campus to mature concepts from basic ideas to products
- **Workforce Development**
Train students to meet unique national security workforce requirements
- **Collaboration**
Foster close relations with government and industry to develop effective partnerships for transition of capabilities

Core Capabilities

Multiple-research teams aligned with campus and regional strengths

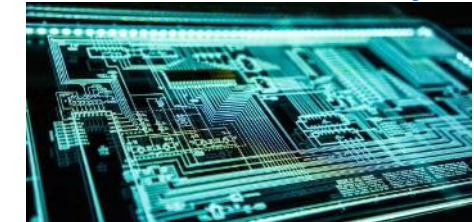
Hypersonics



RF Engineering



Micro-Electronics & Cyber



Space Domain Awareness



Remote Sensing





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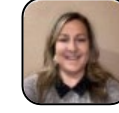
Heather Shocker
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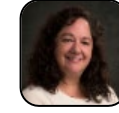
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Key Campus Partners

Office of Contracts and Grants
Office of Research Integrity
(including Research Security and Export Control)
Office of Advancement
(including Industry and Foundation Relations)
Office of Information Technology
(including Research Computing)

Key Campus Faculty Collaborators

40+ faculty across campus:
College of Engineering and Applied Sciences
College of Arts and Sciences



College of Engineering
& Applied Science



1. Motivation – Next Generation Systems Powered by AI/ML
2. Digital-at-the-Element Front End Architectures
 - Simultaneous Transmit and Receive
 - Wideband Arrays
3. Metasurface Front End Architectures
4. AI/ML Enabled Spectrum Systems
 - Gradient-based design and optimization
 - Fully differentiable digital twins
 - Physics informed AI/ML co-design
5. Conclusions

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Significant changes driving hardware redesign for 6G/FutureG:

**Full-Duplex
Communications
(FD)**

**Non-Terrestrial
Networks
(NTN)**

**Holographic
Communications
(HOLO)**

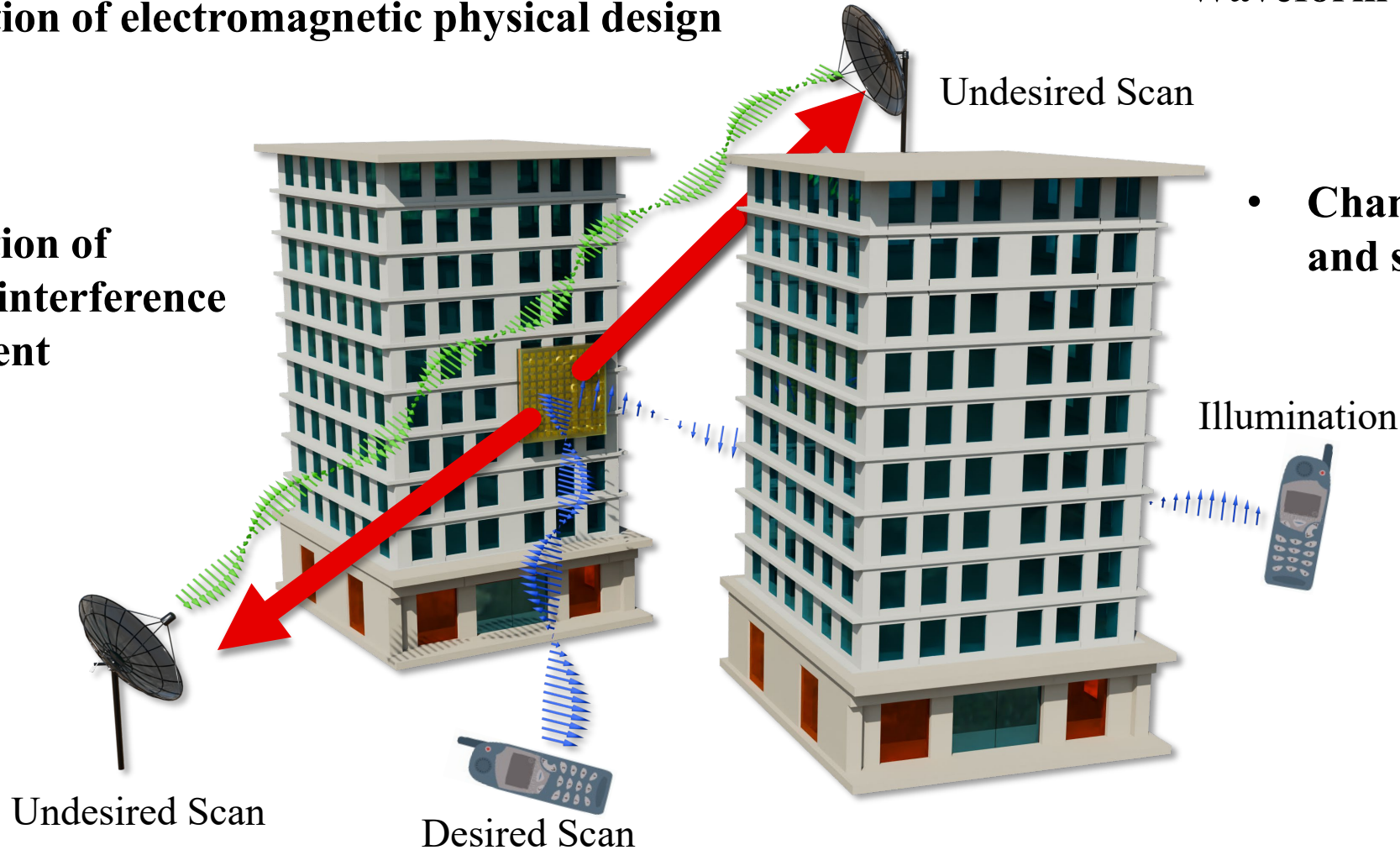
**Integrated
Sensing and
Communications
(ISAC)**

- Optimization of adaptive array state
- Optimization of electromagnetic physical design

- Waveform design and control

- Coordination of resources/interference management

- Channel estimation and state-awareness



If AI/ML is imagined to mitigate these complexities – how can we enable it?

Three steps are crucial to the effective enablement of AI/ML spectrum systems:

1. Flexible, scalable, and capable RF front ends

- Without adequate front-end architectures, AI/ML enabled systems are limited

2. Electromagnetic-aware models and digital twins

- Models and twins must be high fidelity
- Capture full-wave effects such as mutual coupling, and large-scale effects such as fading
- Can be used to pre-train and design systems/networks

3. Data capture and calibration

- Data capture is highly challenging for NG systems
- Therefore, pre-training on simulated results is necessary
- Data capture alone does **NOT** solve co-design problem of next-generation systems

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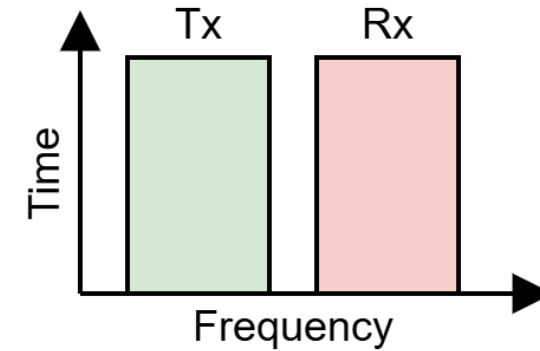
AKA – In Band Full Duplex (IBFD)
AKA – Same-Frequency STAR (SF-STAR)
Uses Self-Interference Cancellation (SIC)

Nominal required isolation:

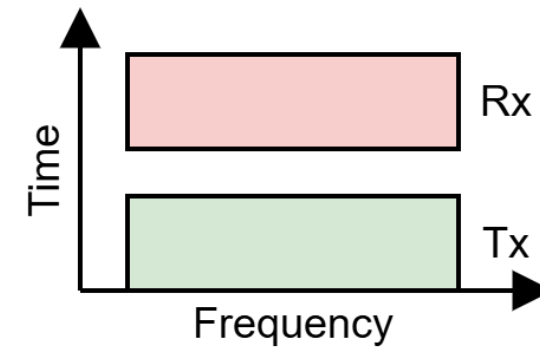
- ~100dB for 1mW TX power
- ~150dB for 100W TX power

In Band Full Duplex (IBFD) allows for the simultaneous transmission and intelligible reception of radio signals at the same frequency, time, and polarization.

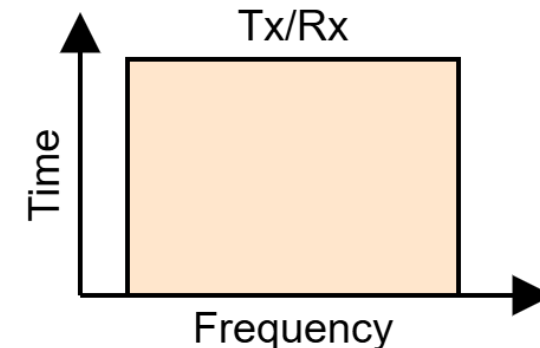
Frequency multiplexing



Time multiplexing



In-Band Full-Duplex



- This is achieved through self interference cancellation (SIC), which is achieved through 3 domains:

- Propagation
- Analog
- Digital

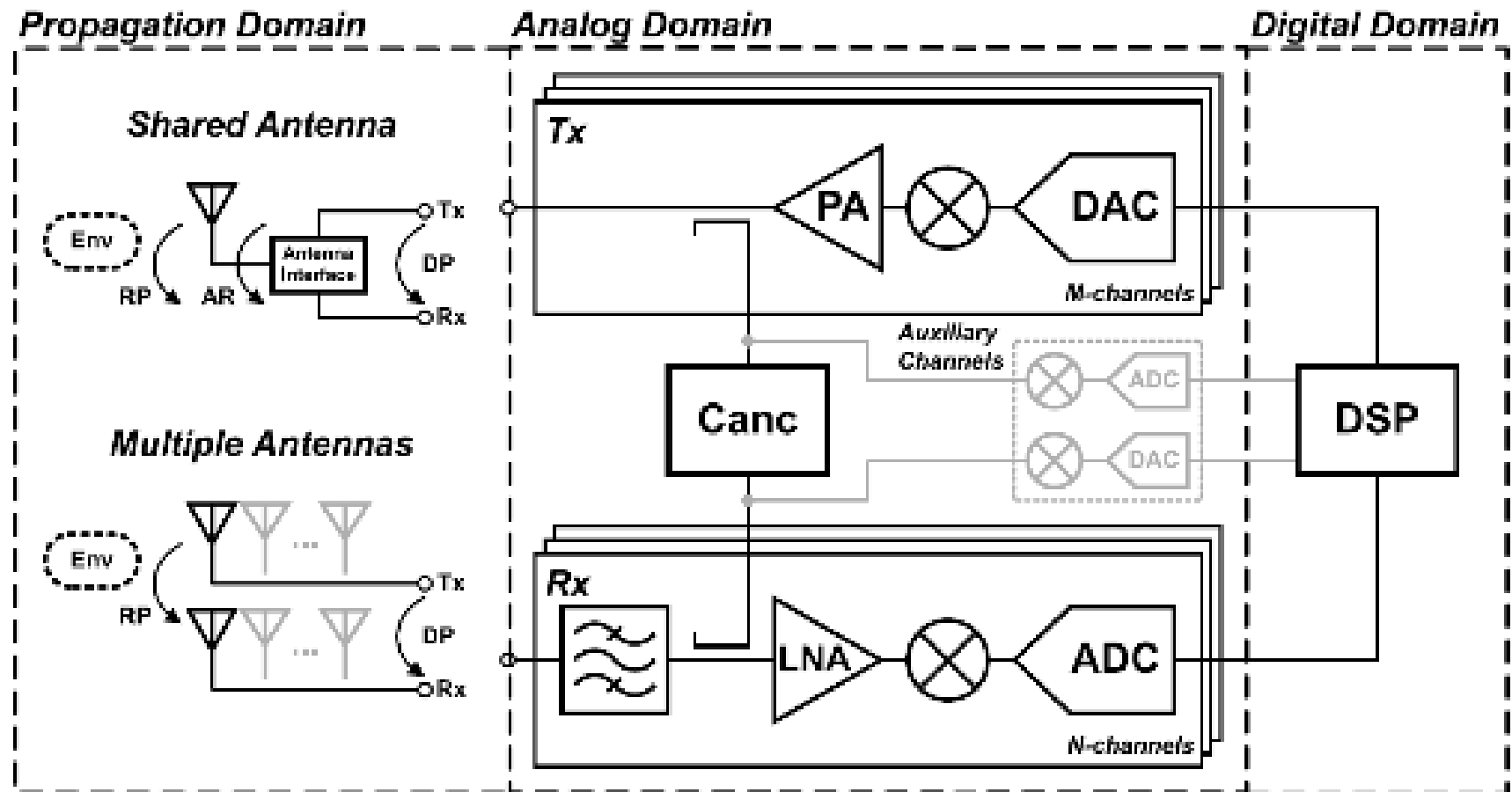
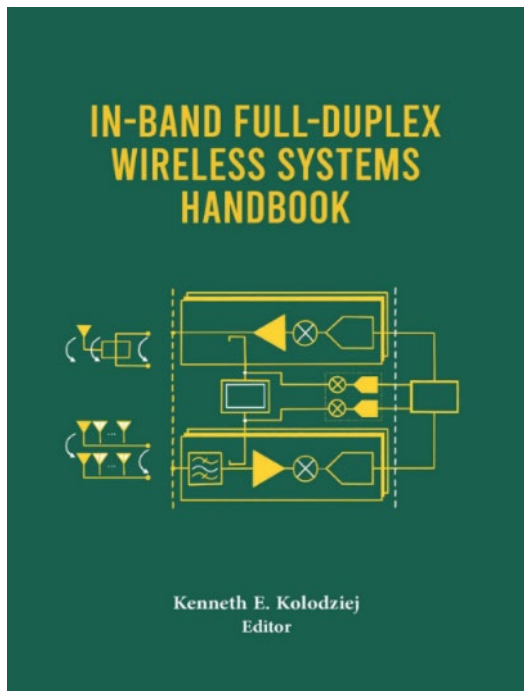
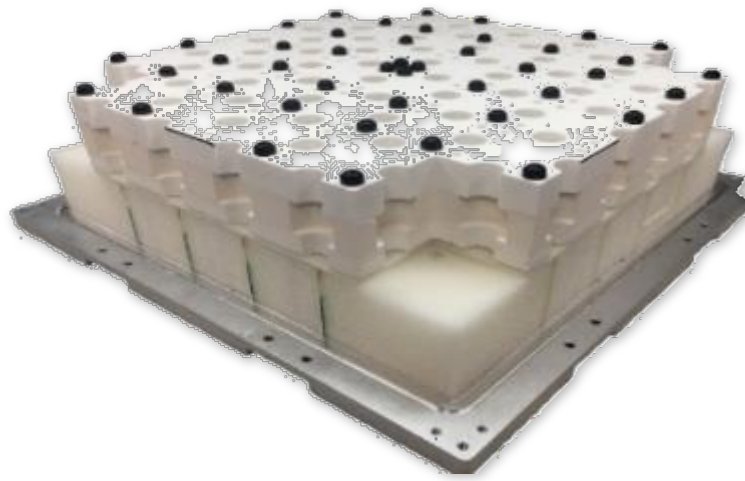


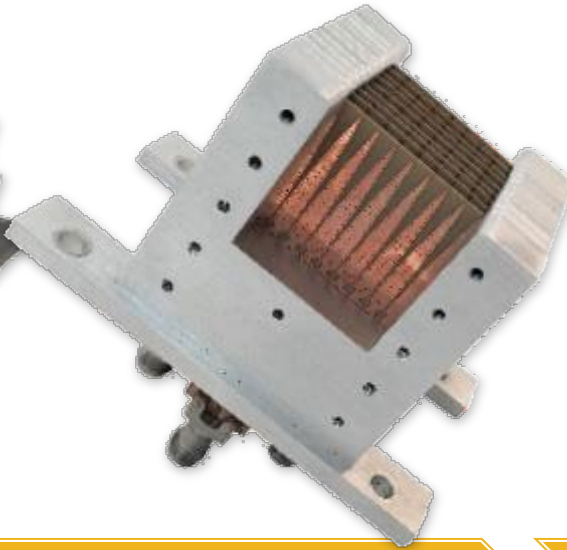
Figure 1.1 Generic IBFD transceiver, highlighting the propagation, analog, and digital domains.

~0.5-3.5 GHz



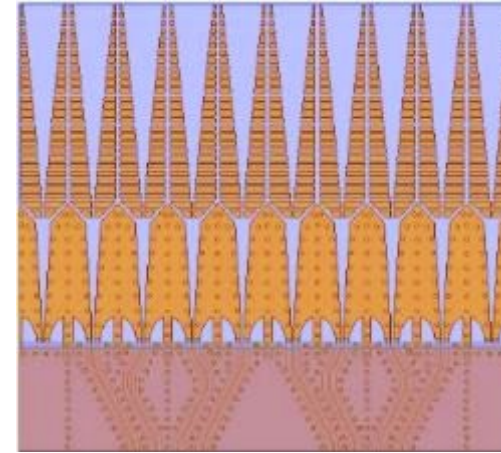
Tightly Coupled
Dipole Arrays

~5-60 GHz



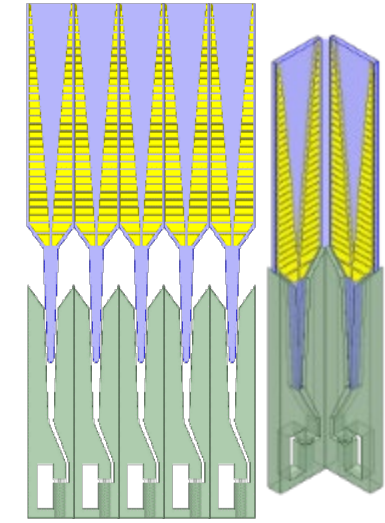
Tapered Slot
Antenna Arrays

~3.5-60 GHz



Sliced Notch
Antenna Arrays

~3.5-60 GHz



Integration

Digital at-the-element architectures enable:
Multi-Beam Steering, T/R Isolation, MIMO, Multi-Mission, Multi-Function

- Digital-at-the-element arrays have poor SWAP-C
 - Most flexible architecture
 - Power and cost limit scalability

Hybrid architectures are most economical and/or practical!

TABLE 2. Beamforming networks - performance vs. trade-offs.

BFN	Pros	Cons
Analog	Cost & complexity, Moderate power	Limited flexibility, poor isolation
Digital	Calibration, reconfiguration	Device count, high power
Hybrid	Optimize between analog and digital	Device count, Complexity
Lens	Low power, low cost	Large footprint, unused ports, fix beam angles
Circuit	Simple & scalable, low power & cost	Lossy, fix beam angles

Goal: multi-mission, multi-function modular hybrid front-ends

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Programmable, passive array formed by many elements

Metasurface:

- Tightly coupled
- Typically very subwavelength (e.g., $\lambda/10$)

Frequency selective surface (FSS):

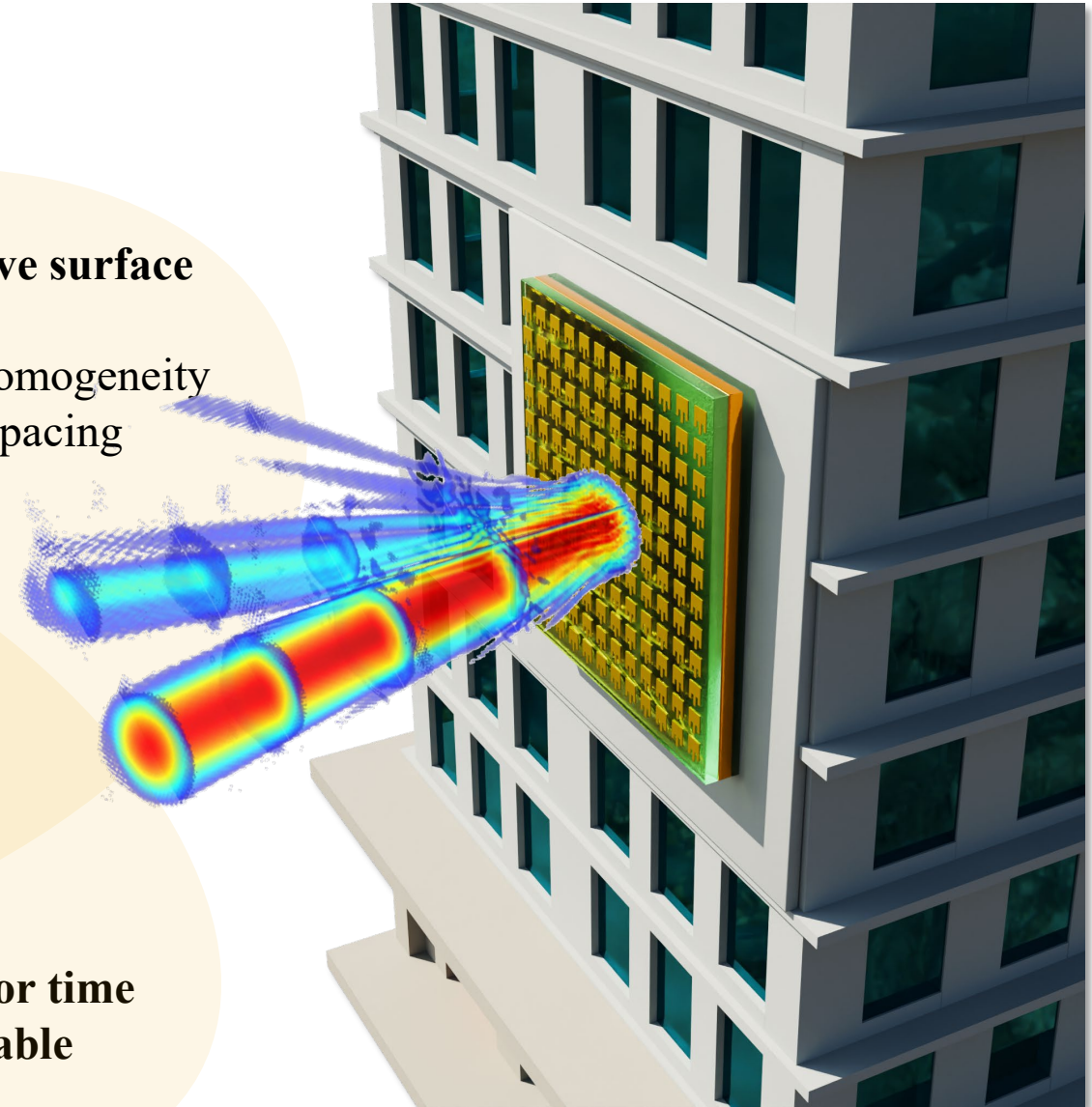
- Large scale inhomogeneity
- $\sim \lambda/2$ to $\lambda/10$ spacing

Reconfigurable Intelligent Surface (RIS)

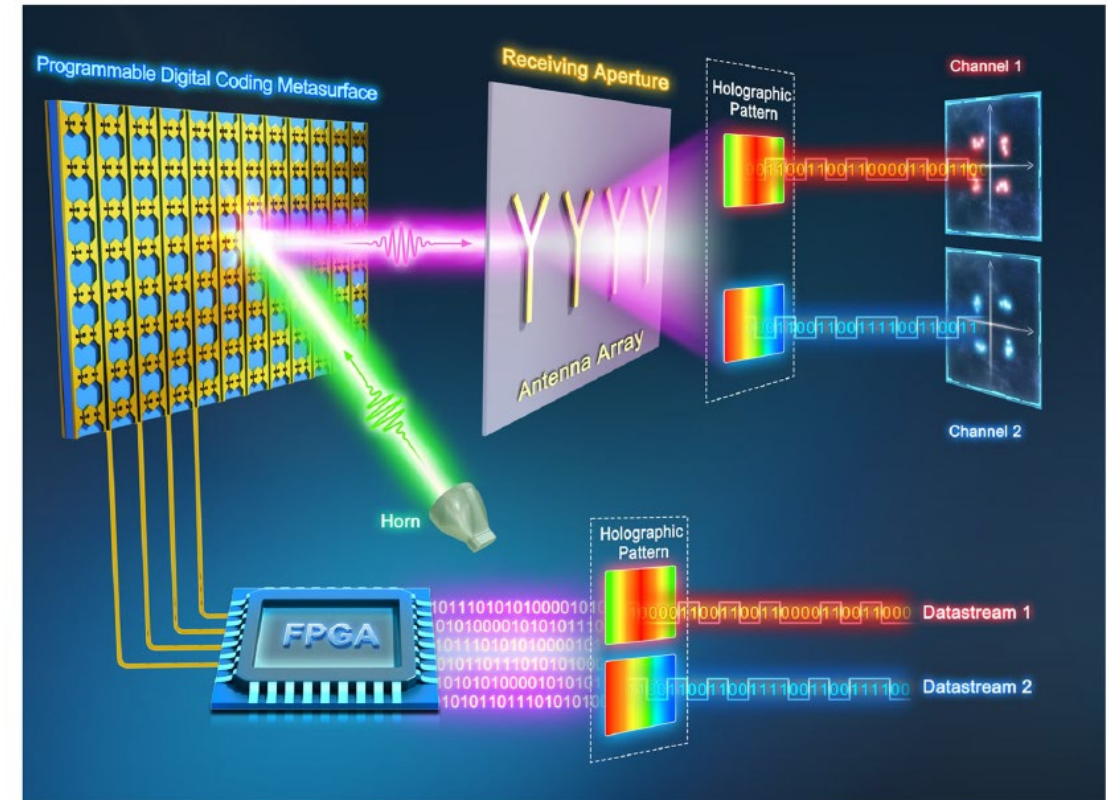
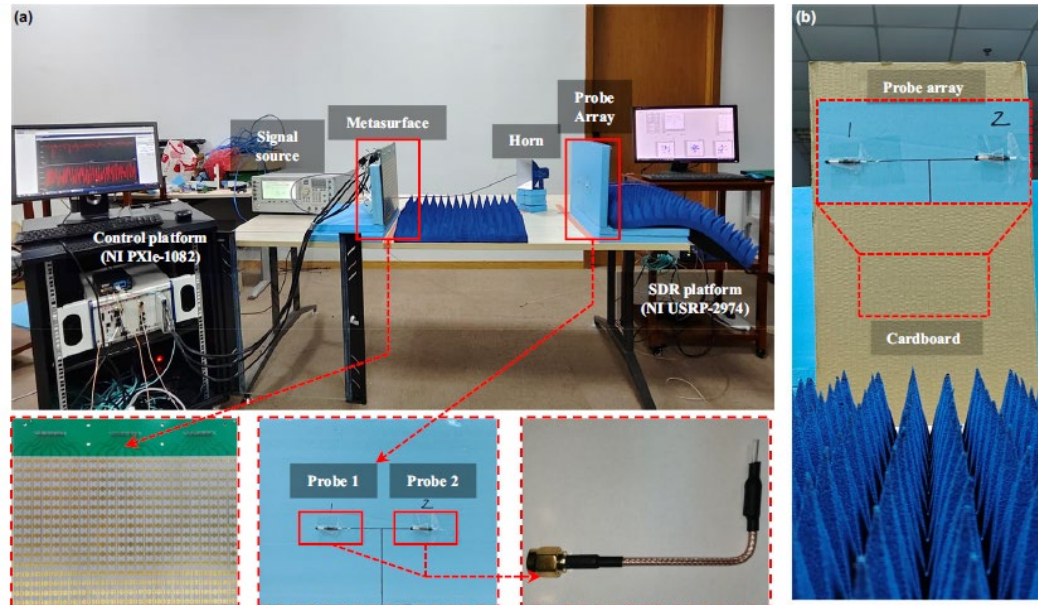
Electrical control

- Varactor diodes
- PIN diodes

Space and/or time programmable



- Integrated **active** and **passive** elements
- Massive apertures enabling near-field holography
 - Example: Aperture level duplexing



Hybrid metasurface-phased array designs achieve modularity and scalability!

There are (3) complexity challenges with metasurface-hybrid front ends

Theoretical Modelling

- Interactions between active and passive elements
- Potentially significant mutual coupling
- Edge-effects/finite array modelling

Hardware Design

- Complex, multi-leveled design
- Performance depends on deployment/wireless environment

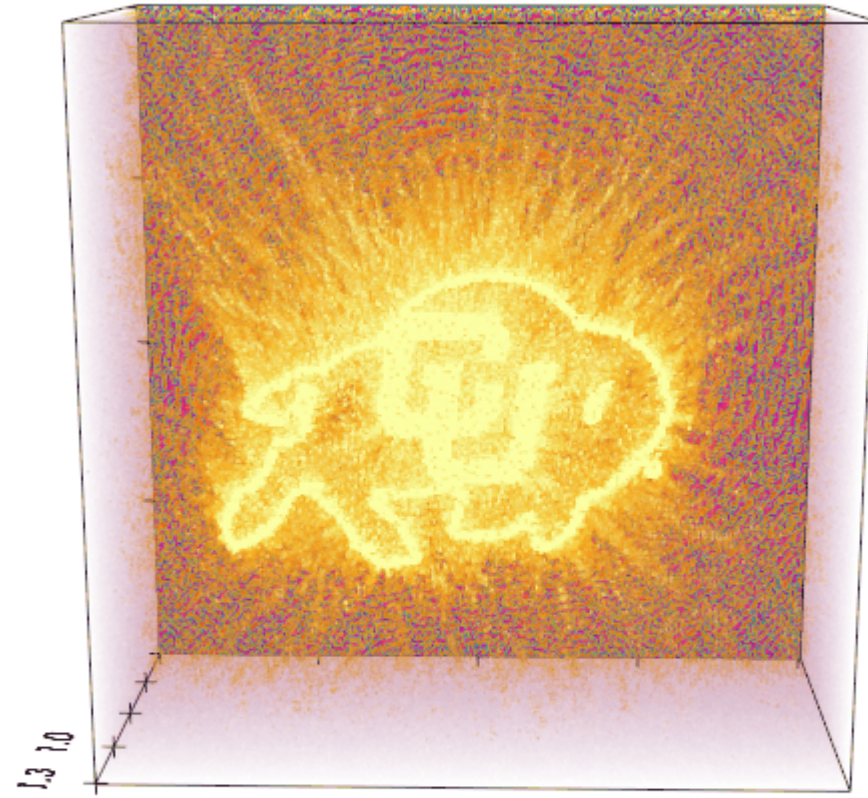
Control/Operation

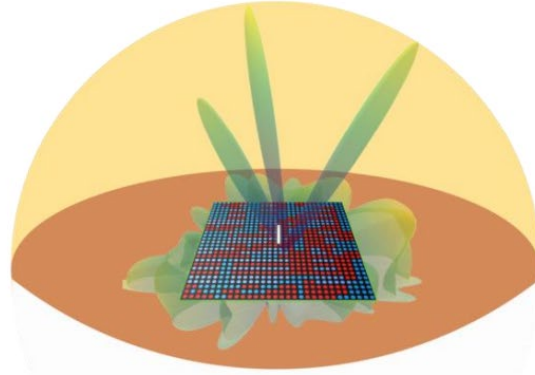
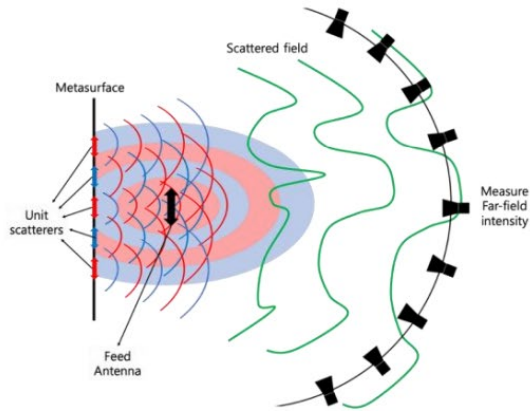
- Large, reconfigurable front-ends have many degrees-of-freedom
- Optimal state is highly dependent on in-situ wireless conditions

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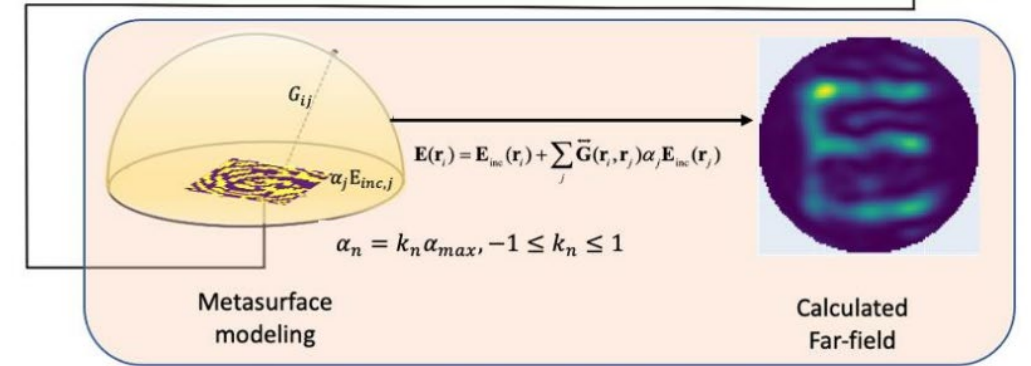
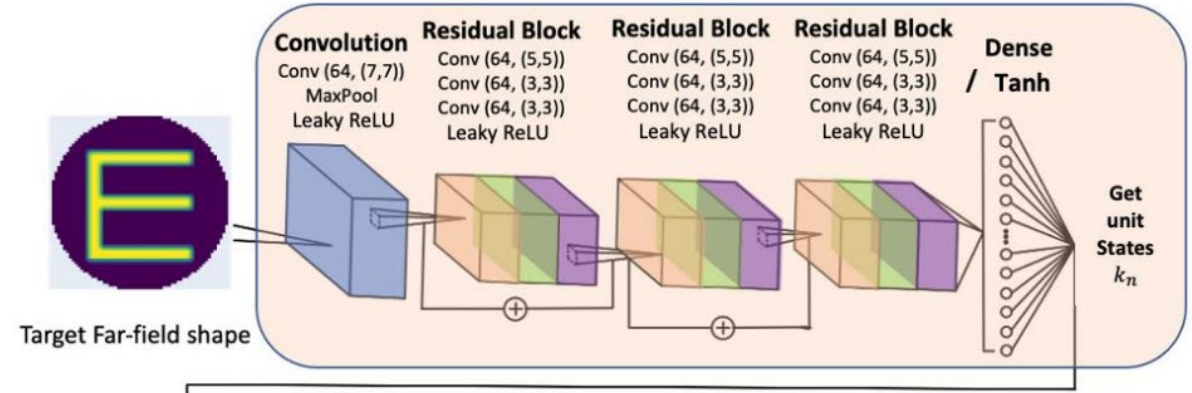
- Mutual coupling aware synthesis methods **are differentiable** [1]
- Benefits from GPU/TPU accelerated libraries (Tensorflow, Jax, Pytorch)

**Spacetime modulated near field scattering:
12.5 GHz carrier, 100 MHz framerate, 131k
tunable elements**





Encoder (Neural Network)



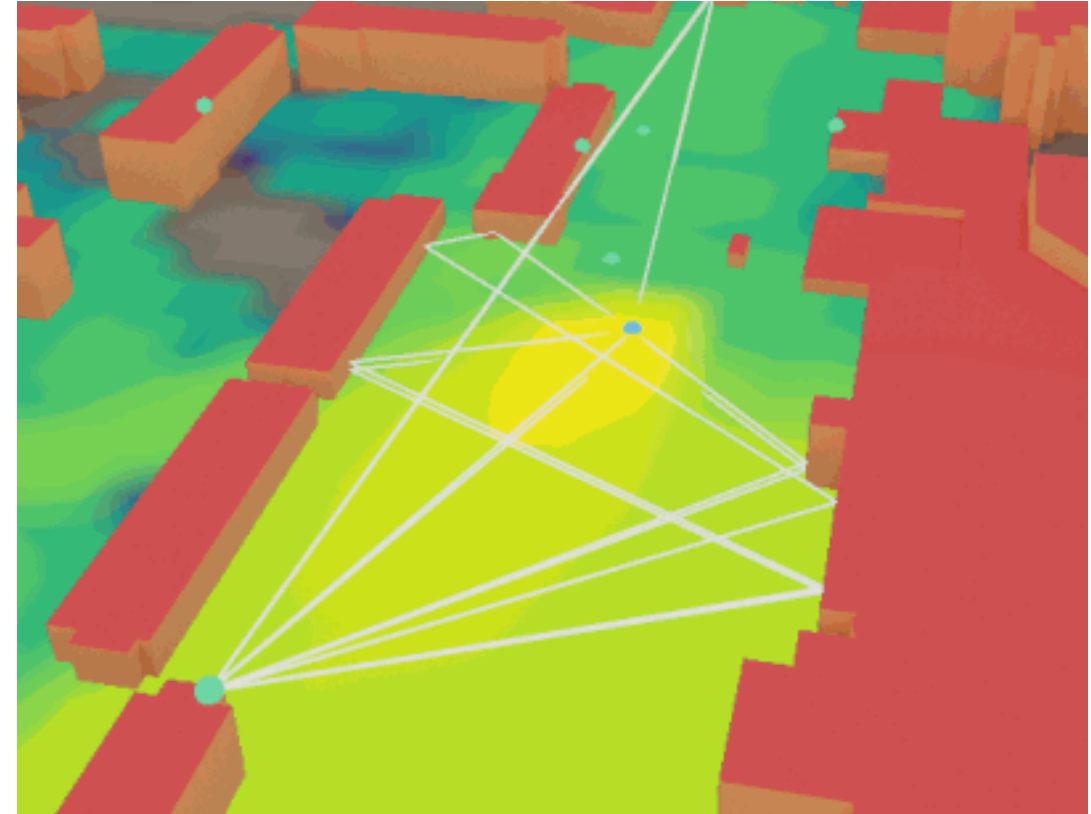
Decoder (Scattering Equation)

Methods	Calculation time	
	Training time	Calculation time
AutoEncoder & first-order Born	1854 sec	194 μ s
AutoEncoder & second-order Born	1921 sec	192 μ s
AutoEncoder & third-order Born	2077 sec	185 μ s
Genetic Algorithm	232 sec	
Gerchberg-Saxton	1.25 sec	

Table 1. Calculation time according to the proposed methods, calculation time except Genetic Algorithm is averaged over the MNIST dataset. Genetic algorithm calculation time is averaged over the cases of Fig. 4.

Ma, H., et al. "Deep-learning-assisted reconfigurable metasurface antenna for real-time holographic beam steering," in *Nanophotonics*, vol. 12, no. 13, pp. 2415–2423, 2023.

- Geometric ray tracers can simulate channel conditions based on digital twin
- Differentiable ray tracers [1] track gradients through 5G waveform (link-layer) and environment scattering (physical layer)



Evolving channel conditions, simulated and visualized using NVIDIA Sionna [1]

[1] Hoydis, J., Cammerer, S., Aıt, F., Aoudia, A., Vem, A., Binder, N., Marcus, G., & Keller, A. (2022). *Sionna: An Open-Source Library for Next-Generation Physical Layer Research*. <https://arxiv.org/pdf/2203.11854>

Ansys

HFSS

- Full-wave simulation (FEM)
 - **Inputs:** geometry, materials, sources
 - **Outputs:** S-parameters



python™

+



TensorFlow

Diode Ports

v_1^+
 v_2^+
 v_3^+
 v_4^+
 v_5^+

$$\begin{bmatrix} S_{VV} & S_{V\tilde{E}} \\ S_{\tilde{E}V} & S_{\tilde{E}\tilde{E}} \end{bmatrix}$$

$$\Gamma_r = e^{j\gamma_r}$$

Plane Wave Ports

$\tilde{E}_{\tilde{k}_1}^-$
 $\tilde{E}_{\tilde{k}_2}^-$
 $\tilde{E}_{\tilde{k}_3}^-$
 $\tilde{E}_{\tilde{k}_4}^-$
 $\tilde{E}_{\tilde{k}_5}^-$

Forward-Scattering Models

- Network-parameter based
 - **Inputs:** incident plane waves, lumped element impedances
 - **Outputs:** outgoing plane waves

$\tilde{E}_k^-$

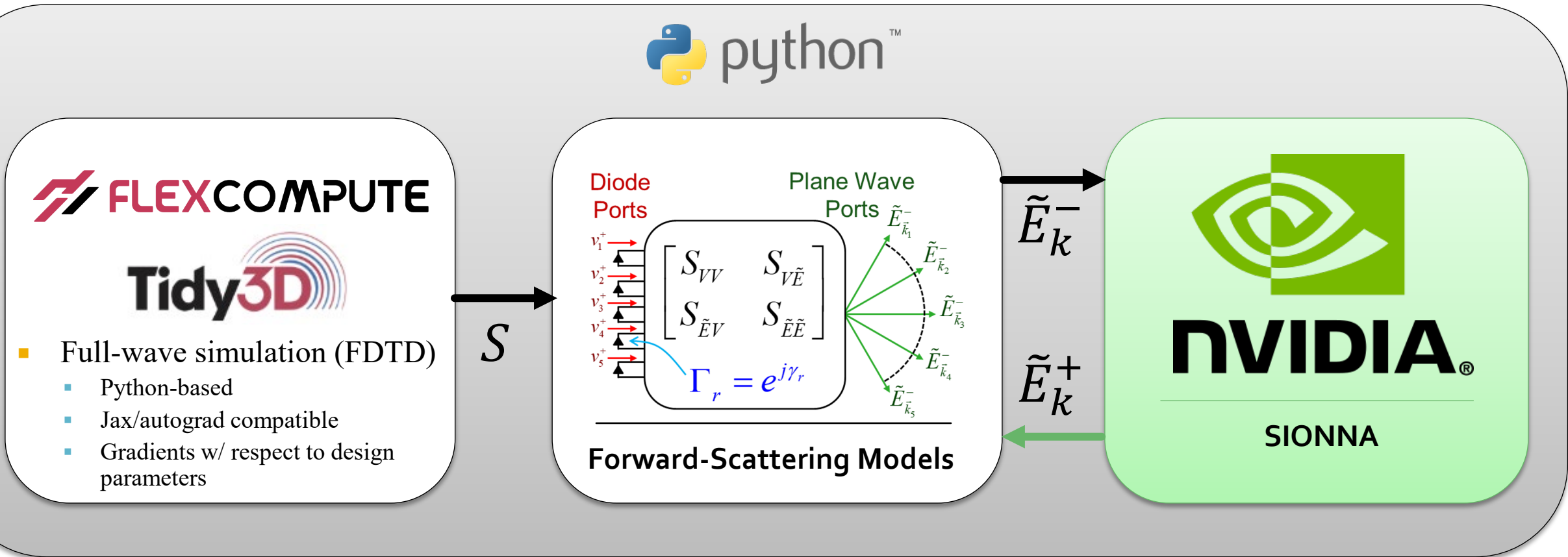
$\tilde{E}_k^+$



NVIDIA®

SIONNA

- Geometric ray tracing
 - **Inputs:** 3D scene/geometry, large movement, material properties
 - **Outputs:** channels, incident/outgoing plane waves, telecom metrics (BER)



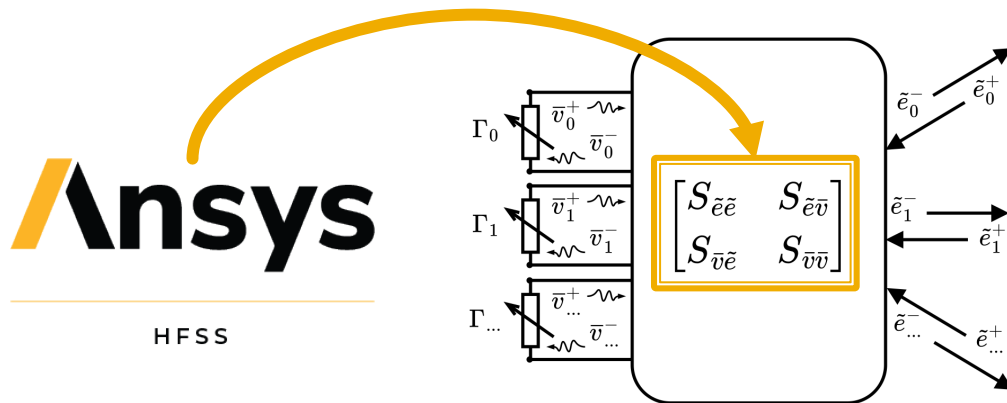
Python + automatic differentiation

(1) Inverse design, (2) Physics-informed machine learning (3) Fully-scriptable

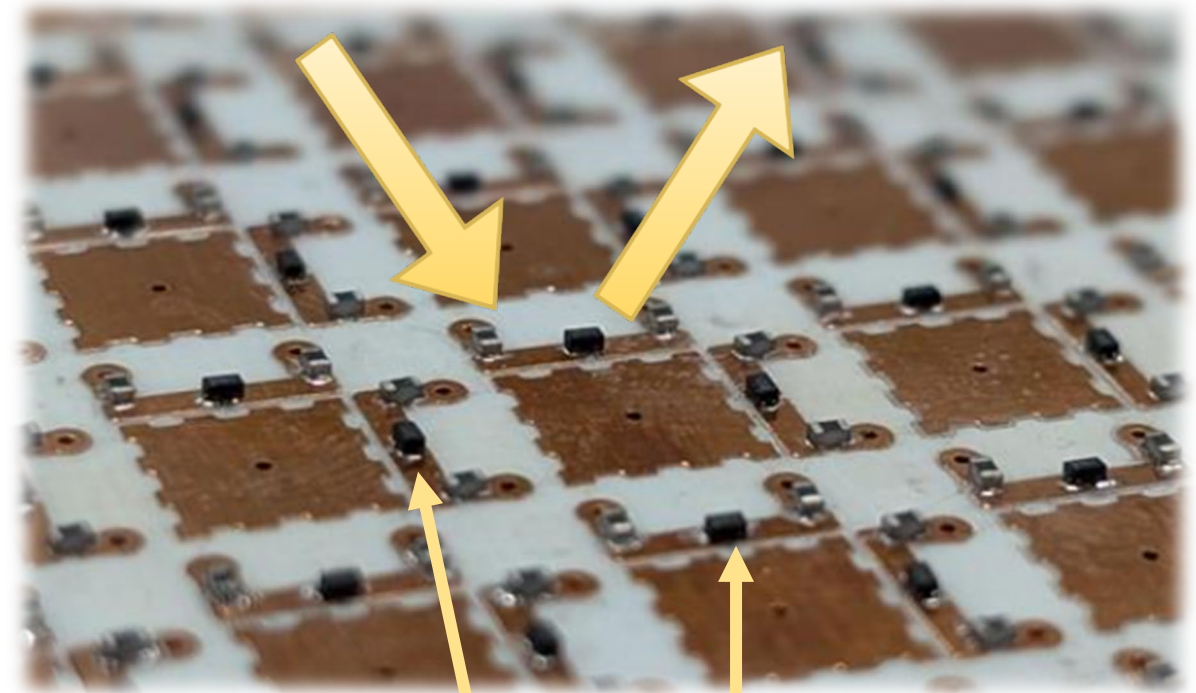
- Next-generation networks are complex, and interdependent between hardware, software, and situational context
- Adaptive, capable systems require adaptive and capable RF front ends
- Design and optimization of AI/ML spectrum systems requires:
 1. Electromagnetically aware digital twins
 2. Ideally, differentiable digital twins
 3. In-situ data capture to refine systems and models (future work)

1. Relate the **spatial** port voltages of a periodic RIS to incoming/outgoing plane wave spectrum
2. Determine the **outgoing** plane wave spectrum given biased element states $Z \rightarrow \Gamma$

Full-wave simulation performed **only once**, then scattering can be solved efficiently during deployment!

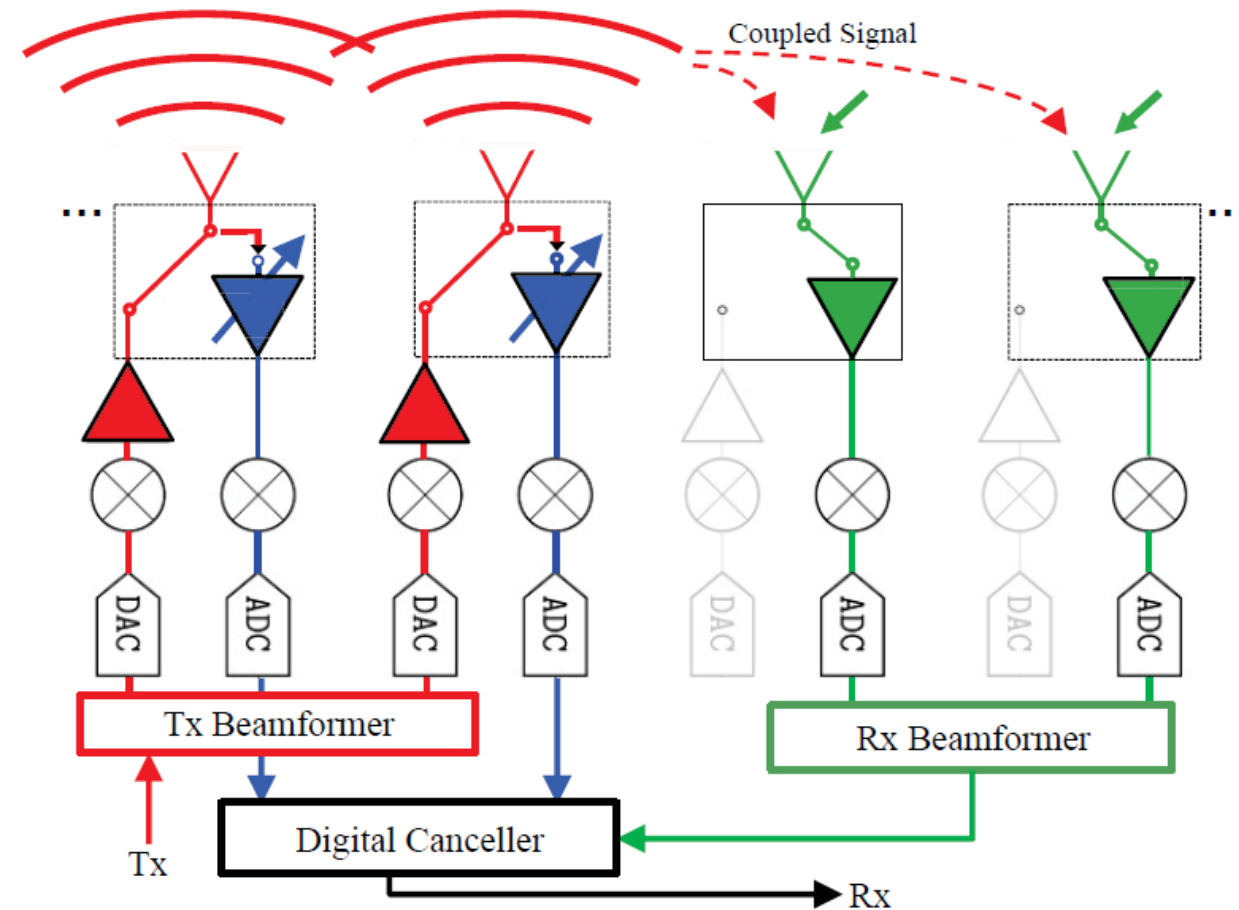


Incoming/outgoing plane wave spectrum



Tunable reflection coefficients

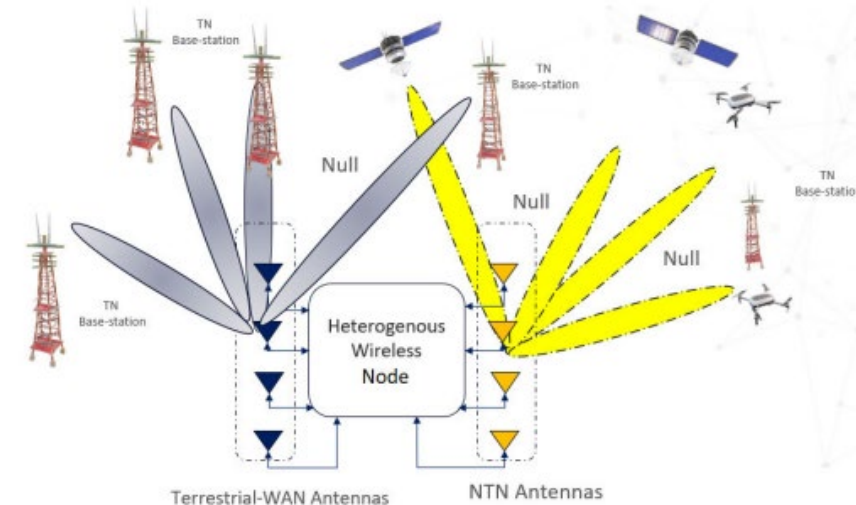
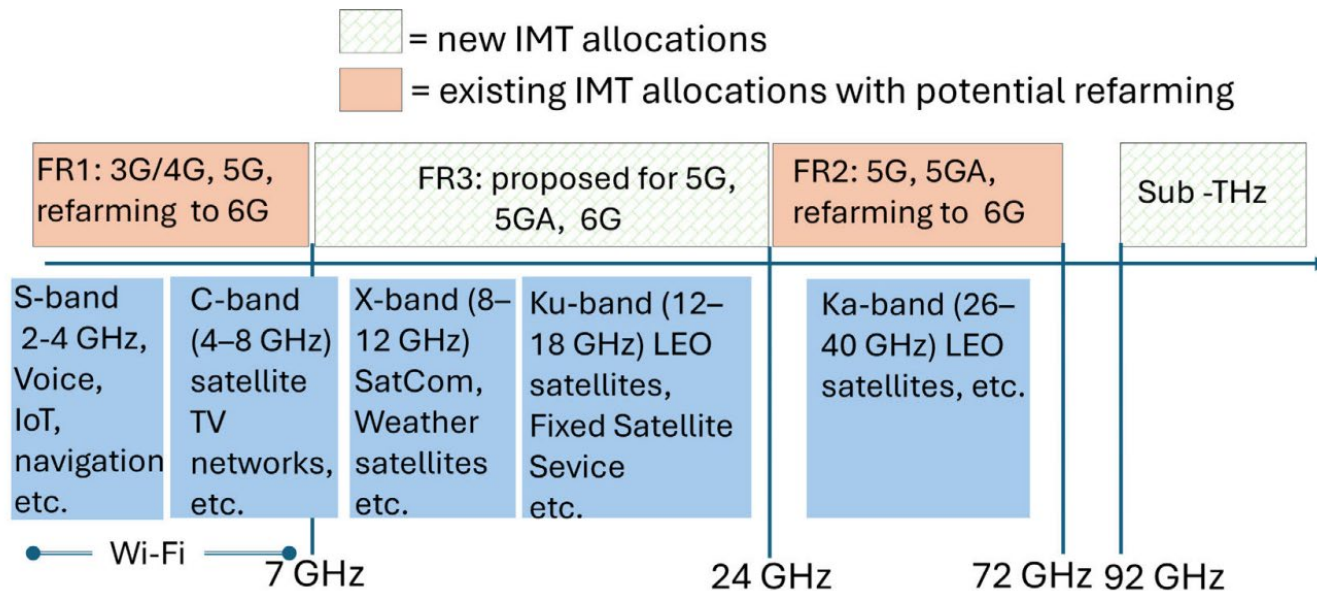
- Switch isolation for sampling of the transmit-path signal
 - Assist digital self-interference cancellation (SIC)
- A fraction of the elements transmit while the remaining elements receive
- Switched half-duplex T/R module in each element
 - Reconfigured in terms of number and location of each T and R element



Adaptive beamforming on TX & RX
Aperture-level simultaneous transmit and receive (ALSTAR)

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- Significant spectrum overlap in 5G, 6G
- High fidelity spatiotemporal control needed for interference mitigation



[1] Overlapping spectrum bands of cellular networks, satellites, and Wi-Fi

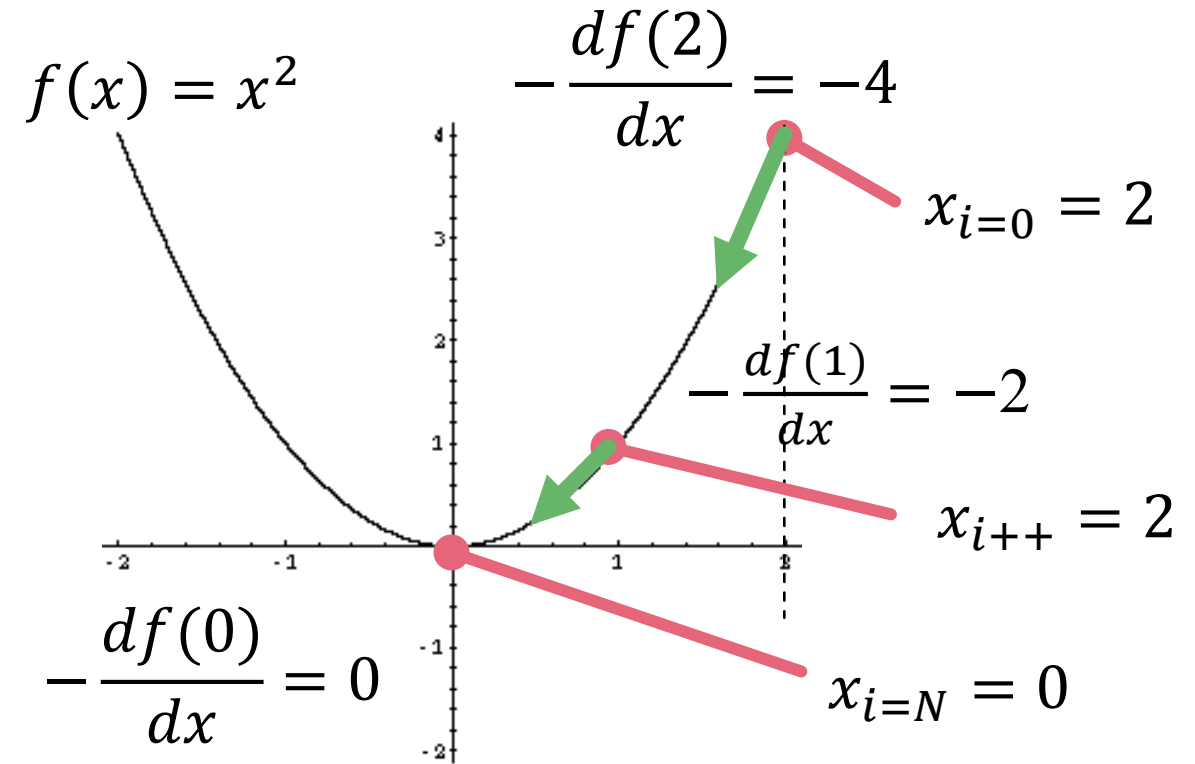
[1] Interference mitigation through antenna beam optimization

Flinck, H., Kumar, T., Ramabadran, P., Katranaras, E., Garcia, L. G. U., Hmedoush, I., Hassan, S. S., Wilhelm, M., Meyer, S., Drewes, C., & Mähönen, P. (2025). An Overview of Flexible Spectrum Management Approaches for 6G. *IEEE Access*, 13, 207396–207411.
<https://doi.org/10.1109/ACCESS.2025.3638140>

The **gradient** of a function is a vector of partial derivatives

- E.g., given some error function, it is the vector of derivatives with respect to each parameter

The gradient can be used to optimize (minimize or maximize) some objective function f



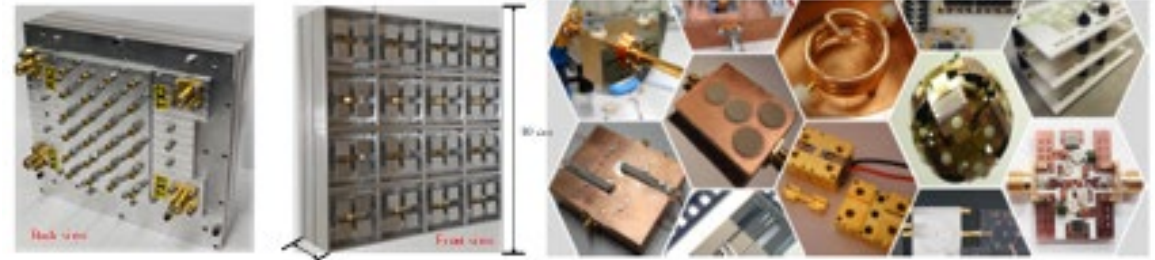
e.g., 1-D convex optimization (trivial)

Modern AI/ML relies on automatic differentiation and the adjoint variable method (backpropagation) to efficiently compute gradients for highly parameterized neural networks

Overview

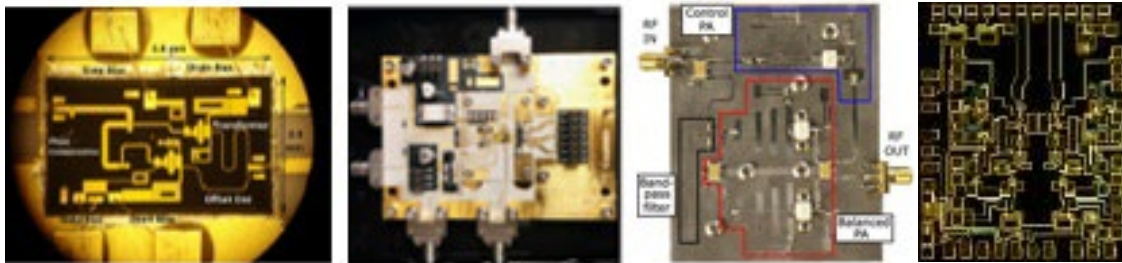
- **Missions:** RF Communications, Spectrum Operations, Electronic Warfare, Remote Sensing
- **Challenges:** Bandwidth, power, size, integration, resilience, LPI/D, reconfigurability, security
- **CU Strengths:** 5G/XG, antennas, power amps, filters, RF AI/ML, MMIC design, photonics
- **Federal Agencies:** DARPA, Air Force, Army, Navy, NSF, NASA, NIST, Space Force, OUSD...
- **Industry Partners:** Lockheed-Martin, Northrop Grumman, Boeing, Qorvo, L3-Harris, BAE, HRL, CACI, SRI, Azure Summit...

Apertures and Passive Circuits



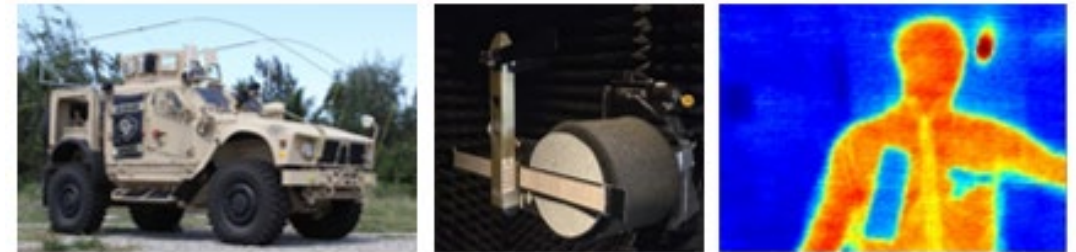
- Wideband, small, efficient, high-power antennas
- STAR antennas, arrays, and analog front-ends
- RF co-designed electrically-tunable front-ends
- High power, tunable, small RF filters

Active and Integrated Circuits



- High-efficiency power amplifiers for telemetry, radar, communications, and medical applications
- Low phase noise oscillators, time and frequency stability
- Low noise amplifiers, MMIC design
- Package design for high active device efficiency

Quantum Sensing & Metrology



- RF & RF-photonic front-ends for EW, communications, radar, telemetry, full-duplex radios, small satellites
- Quantum sensors (quantum magnetometers, Rydberg sensors)
- Wireless power transfer, MRI, non-invasive sensing, and other medical applications