

# *Disrupting Next G*

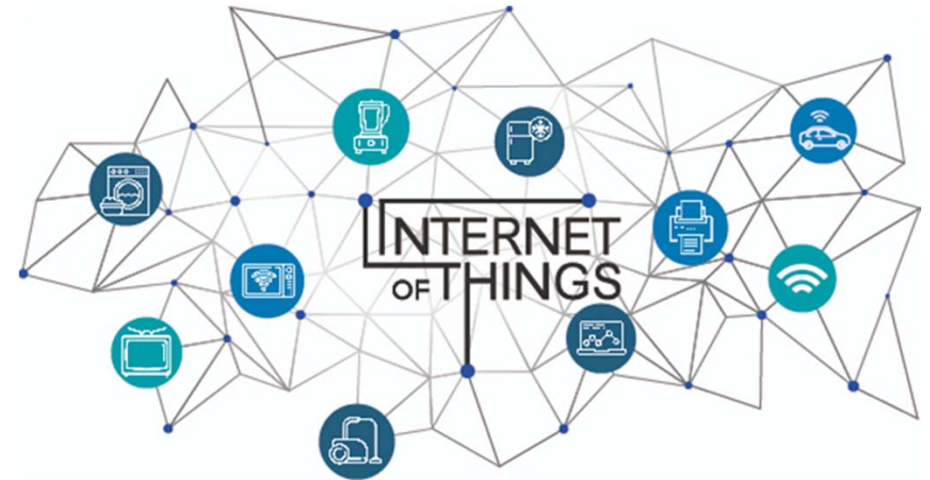
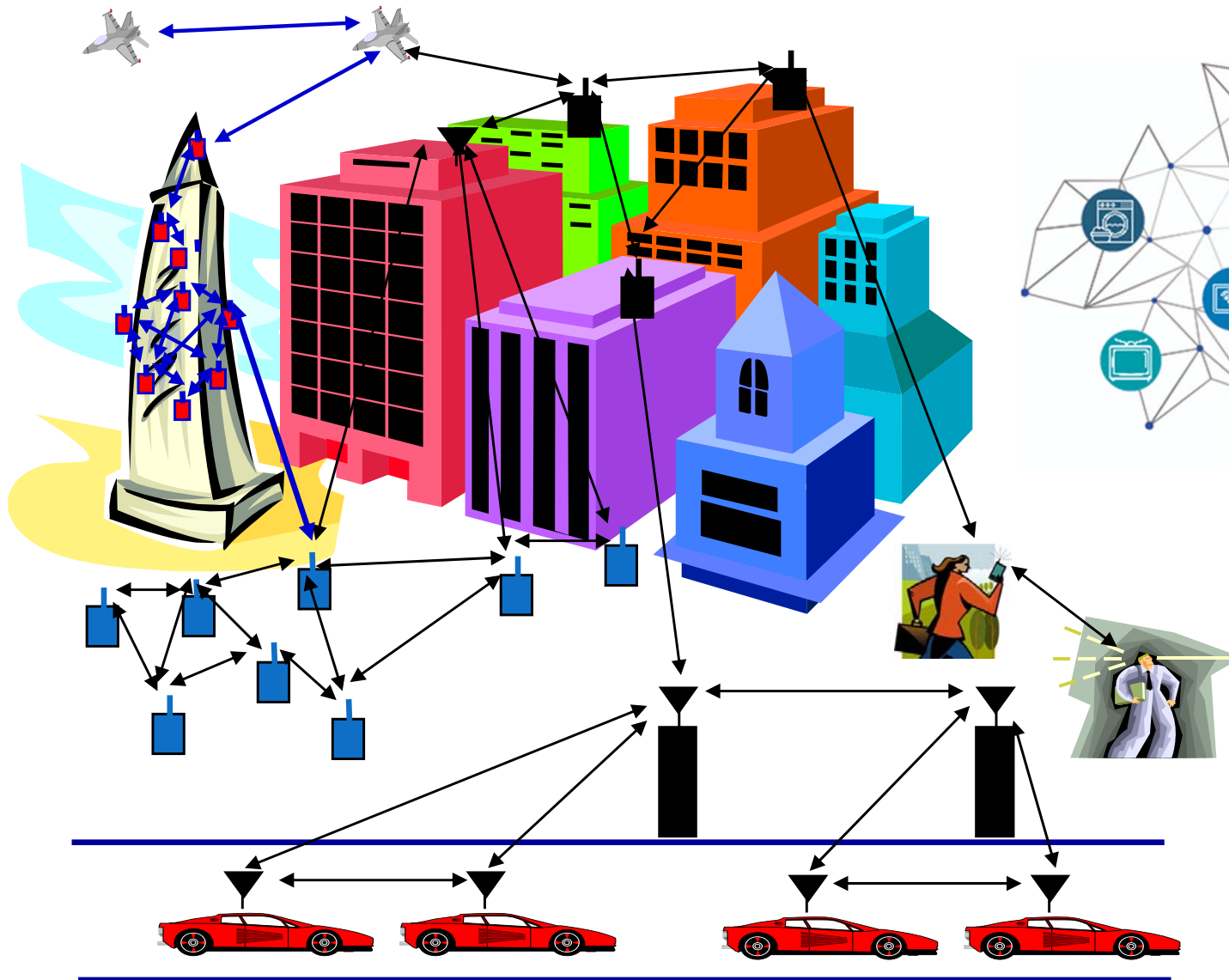


ICASSP Plenary Lecture  
June 6, 2023  
Rhodes, Greece



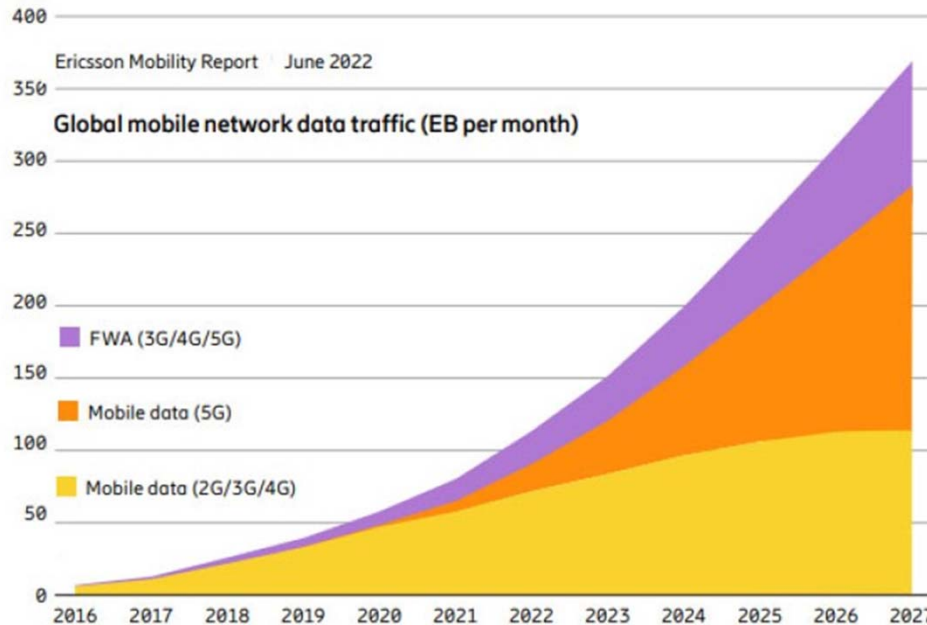
# Future Wireless Networks

*Ubiquitous Communication Among People and Devices*



Next-Gen Cellular/WiFi  
Smart Spaces/Cities  
Autonomous Vehicles  
“Enhanced” Reality  
Body-Area Networks  
Massive Sensing  
All this and more ...

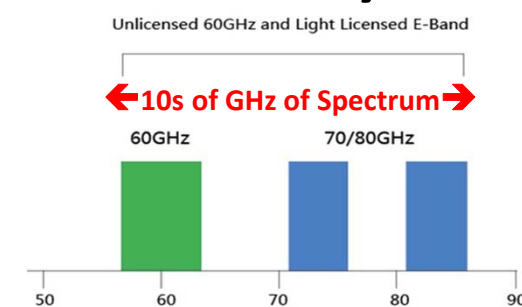
# The Licensed Airwaves are "Full"



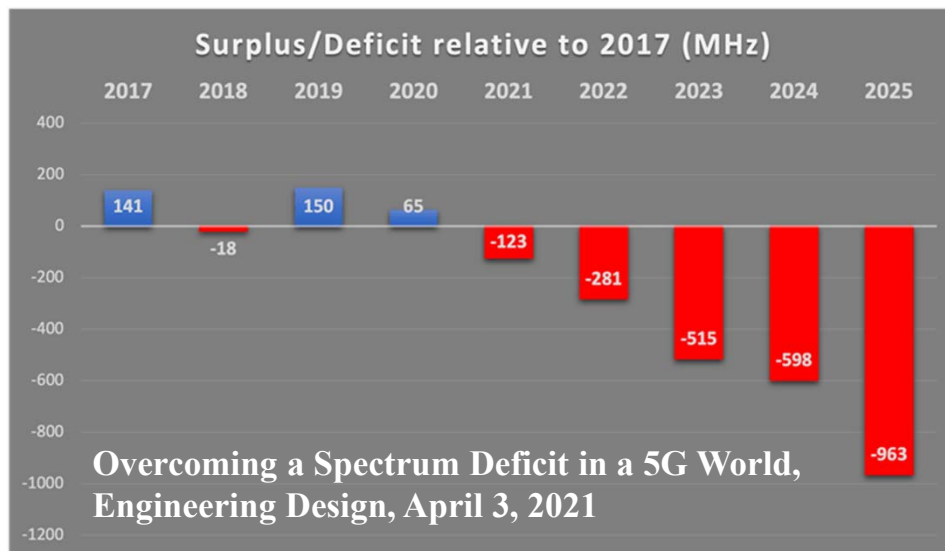
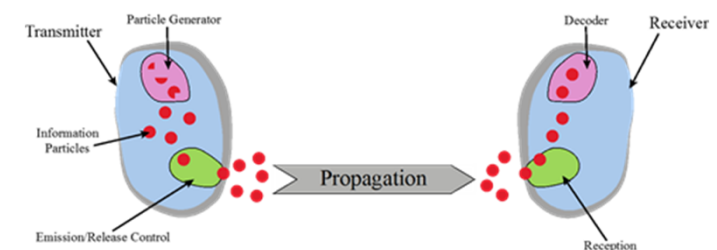
*Also have Wifi*



*And mmW/THz*



*And molecular*



# On the Horizon, the Internet of Things

U.S. IoT analytics market, by vertical, 2014-2025 (USD Billion)

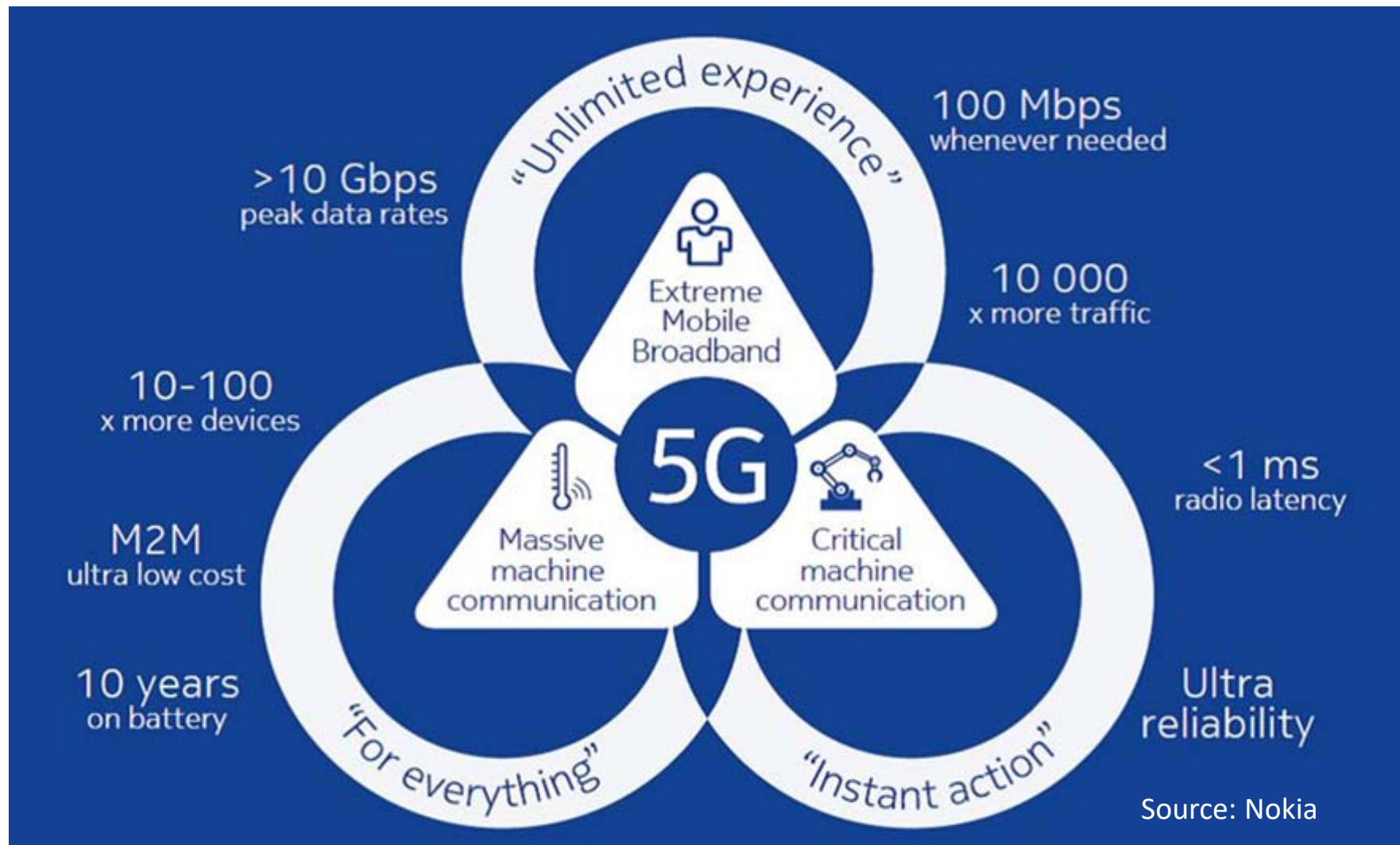


Source: Grand View Research  
Report ID: GVR-2-68038-142-9

- Different requirements than smartphones
  - Low rates, latency, energy consumption
  - Also security, privacy, and resilience

# Promise of 5G

Challenges: high data rates, low energy, low latency





Promises and pie-crust are  
made to be broken.

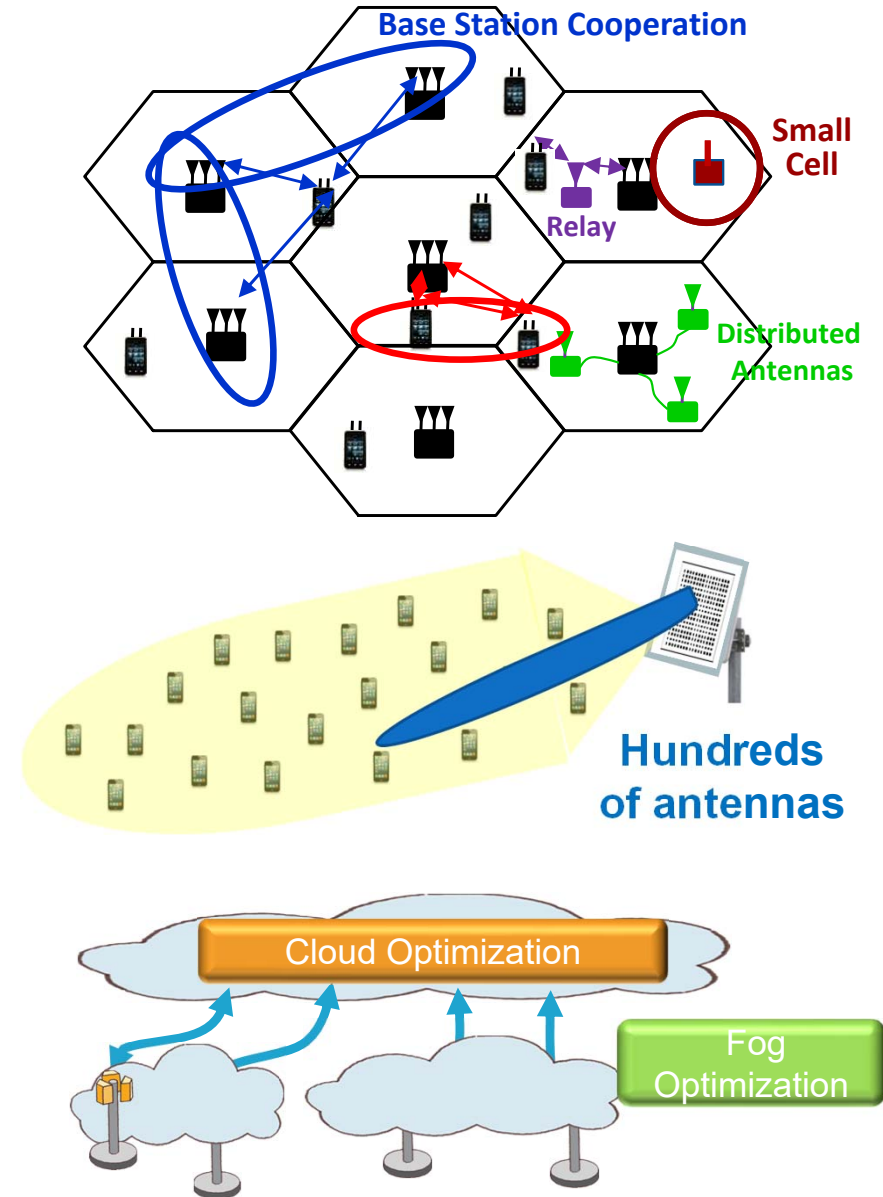
~ Jonathan Swift

**Especially in wireless standards, as it drive the next G**

# Enabling Technologies for NextG networks

- Rethinking cellular system design
- Rethinking backbone network design
- Utilizing more spectrum (mmw/THz)
- Very low power radios
- Massive MIMO
- New PHY and MAC techniques
- Multihop routing
- Edge computing and caching
- Cloud and fog optimization
- Security, privacy and resilience
- Machine learning

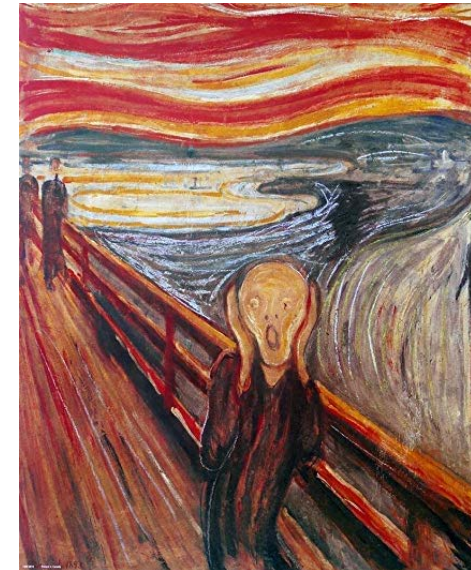
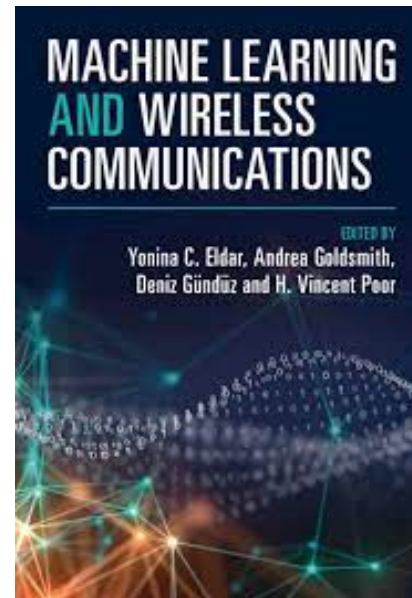
Can't ML enable everything?



# ML Today is a Bandwagon



Should we jump on?



Or run screaming?

- ML **will** play an important role in NextG design/operation
  - ML **can** “beat” theory in PHY-layer design (equalizers, encoding/decoding)
  - ML may provide approximate solutions to intractable network resource allocation problems
  - ML may provide breakthroughs in security, privacy and resilience
  - ML may provide the key to breakthroughs in cross-layer design

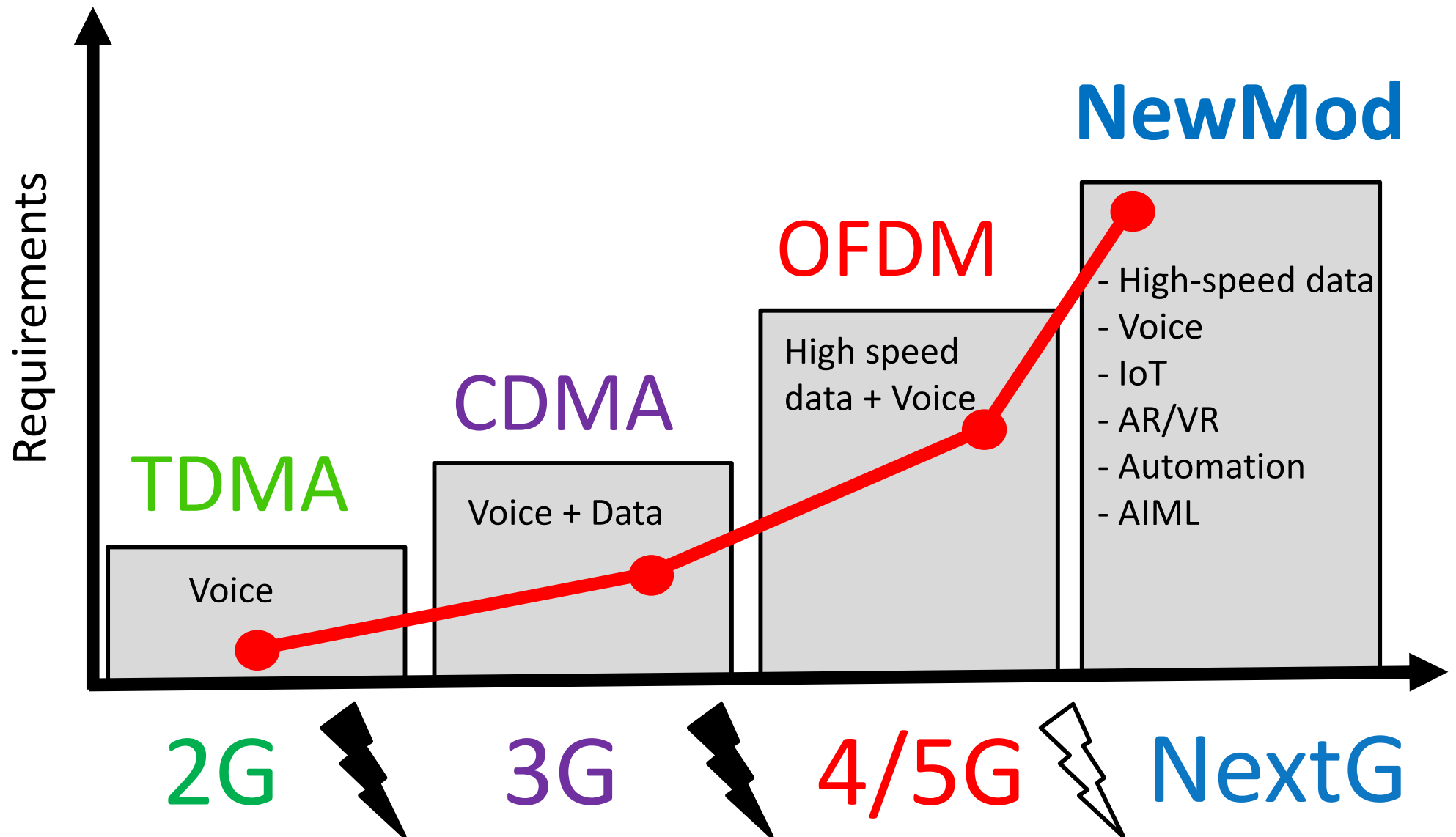


# New PHY and MAC Techniques

- New Waveforms
  - More bandwidth/energy efficient/robust to changing channels
  - More flexible and efficient subcarrier allocation
- New Coding Techniques
  - New channels and new waveforms
  - New requirements (low latency and low complexity)
- New Detection Techniques
  - Lower complexity, robust to impairments, blind
- New Multiple Antenna Techniques
  - New space-time modulation, coding, and detection methods
  - Massive MIMO
- New Access Techniques
  - Efficient (non-orthogonal) access
- Machine learning in PHY/MAC design and operation

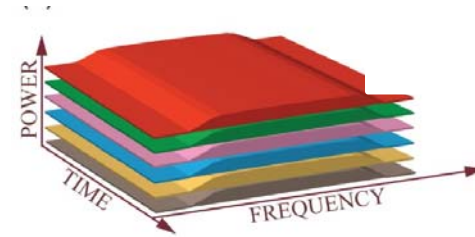


# Waveform Disruption

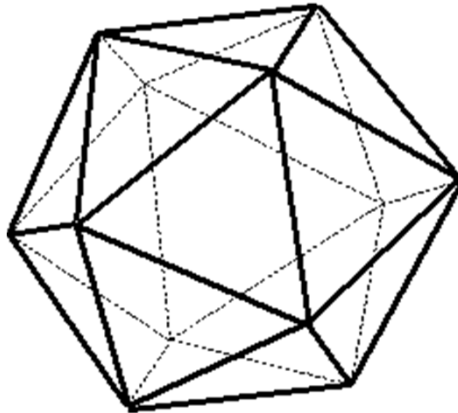


# Data can be modulated onto different domains: *time, frequency, doppler, delay*

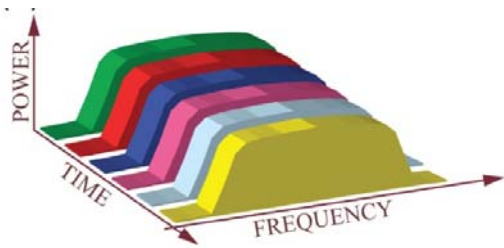
CDMA



NewMod



TDMA



OFDM

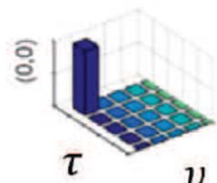
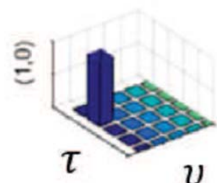
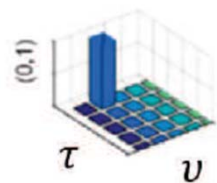
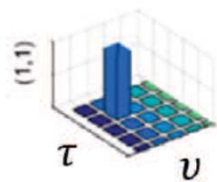


# One Strong Candidate: Orthogonal Time-Frequency-Space Modulation (OTFS)\*

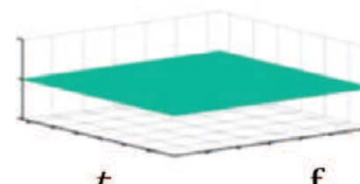
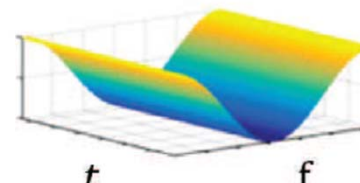
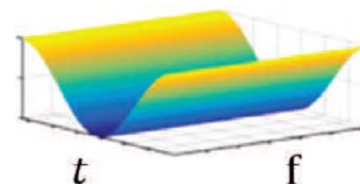
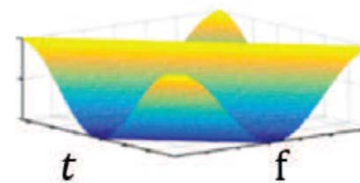
*Data adaptively modulated in delay-Doppler domain*

**Delay-Doppler  
Domain**

**Zak Domain**



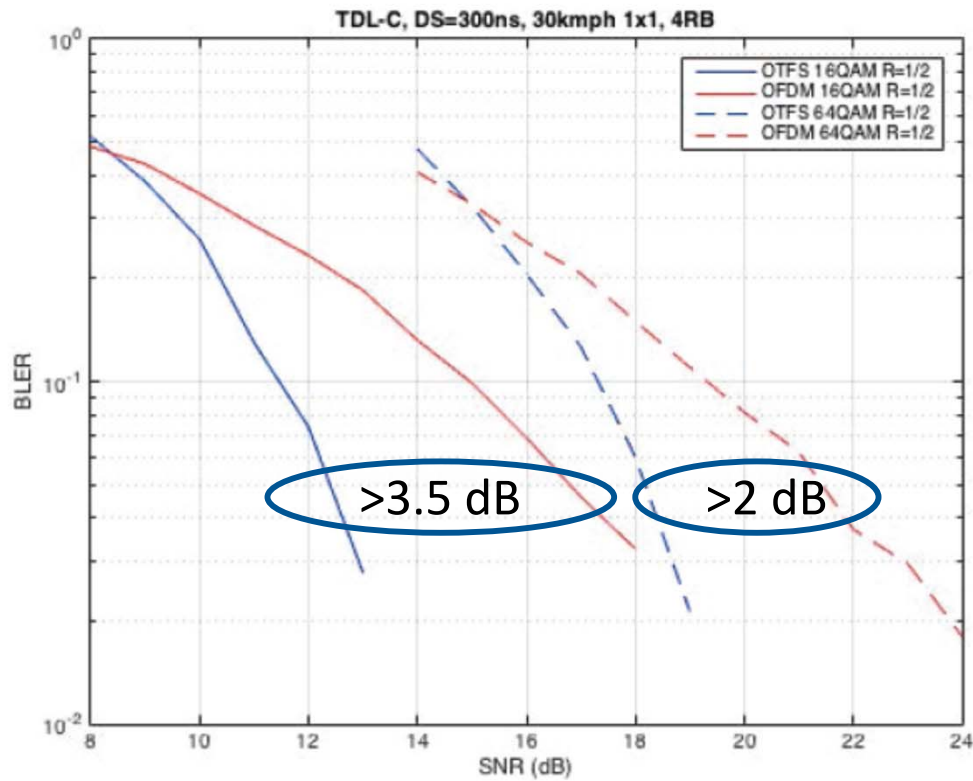
*Zak  
Transform*



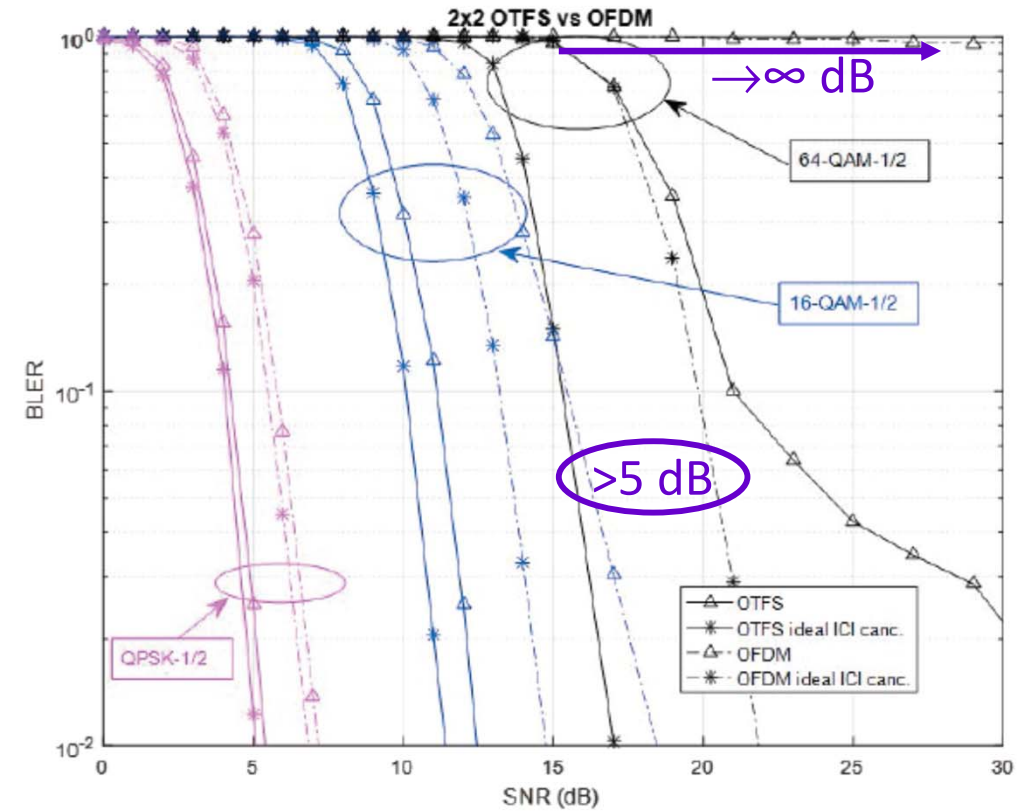
**Time-Frequency  
Domain**

**All symbols spread  
over entire time-  
frequency channel;  
obtain full diversity**

# OFDM vs OTFS Performance

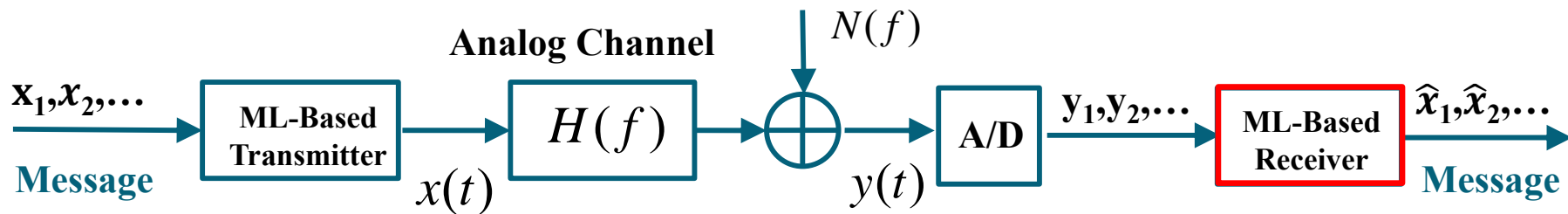


Short packets, 1x1, 16/64 QAM, 30 km/h



Long packets, 2x2, 4/16/64 QAM, 500 km/h

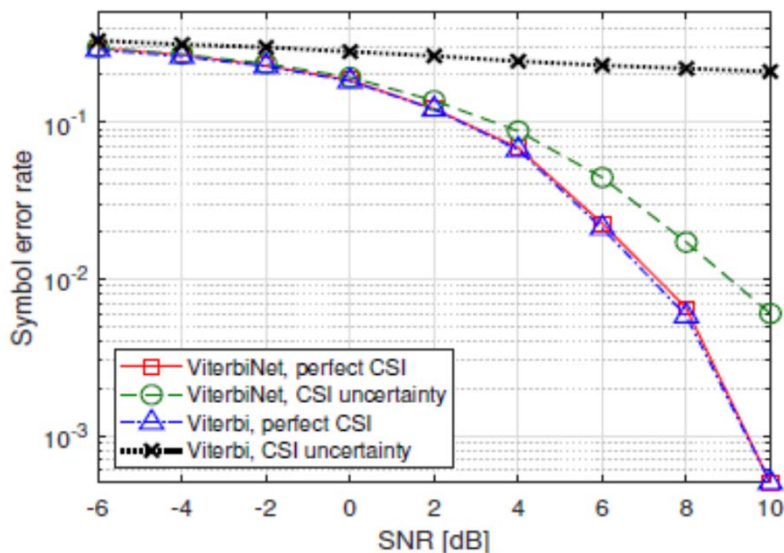
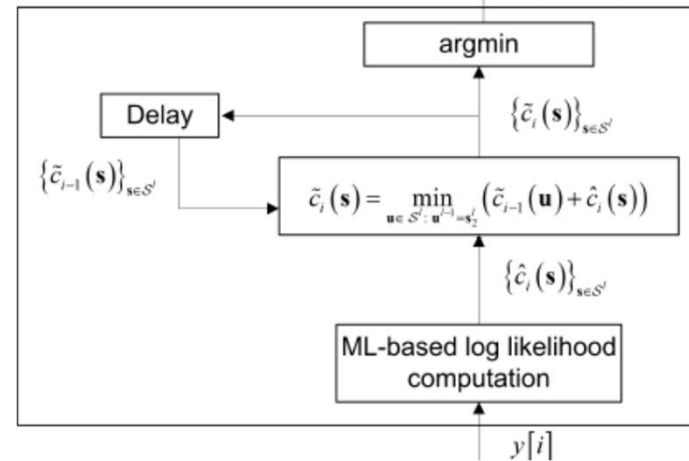
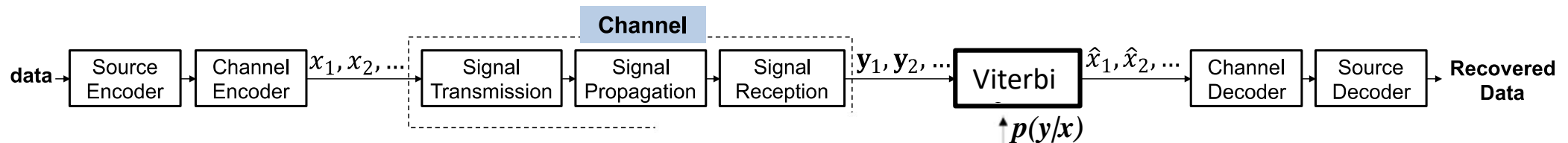
# ML-Based Receiver Design



- PHY transmitter and receiver design typically based on a mathematical channel model
  - Accurate channel models may not be known
  - Models may not enable computationally efficient PHY algorithms (decoding, detection, message recovery)
- How does ML-Based RX design solve this?
  - No need for channel model or its parameters
  - Learn the RX design directly from data
  - Solution is robust to estimation error
  - Requires large amount of training (many  $x_i$ s)

# ViterbiNet:

## Standard Viterbi with ML to learn $p(y|x)$

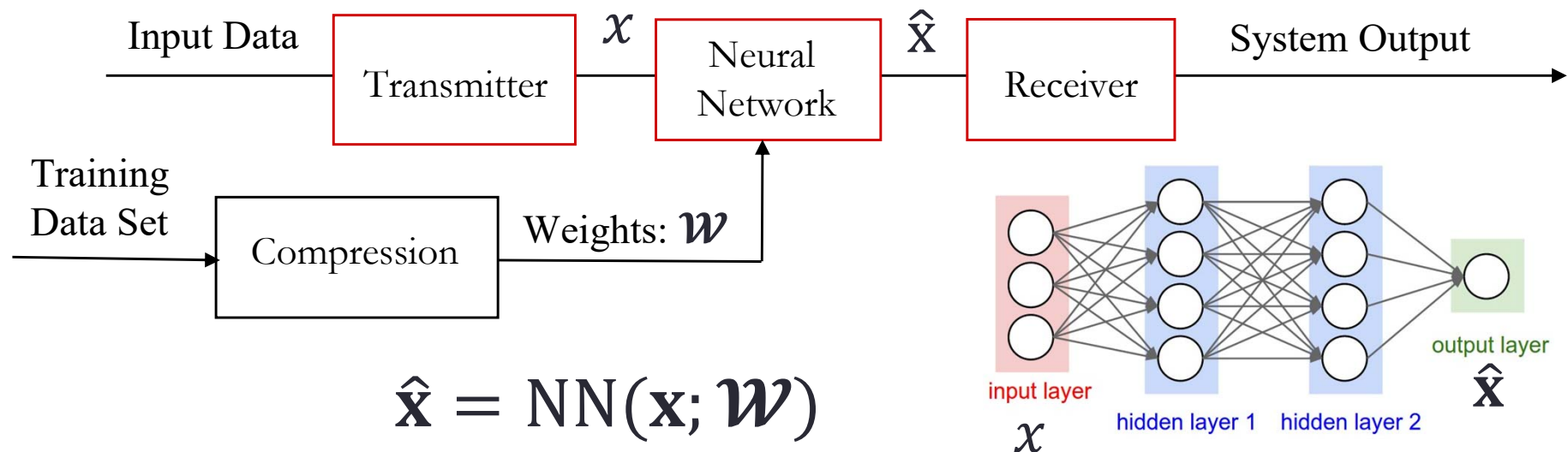
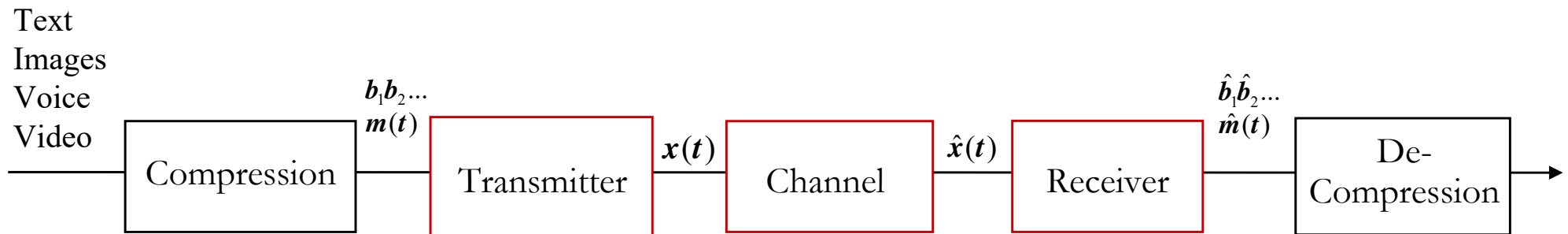


Much faster training and lower complexity than end-to-end ML-based receiver design

Joint work with Eldar, Farsad, Shlezinger

# Neural Nets as a Communication System

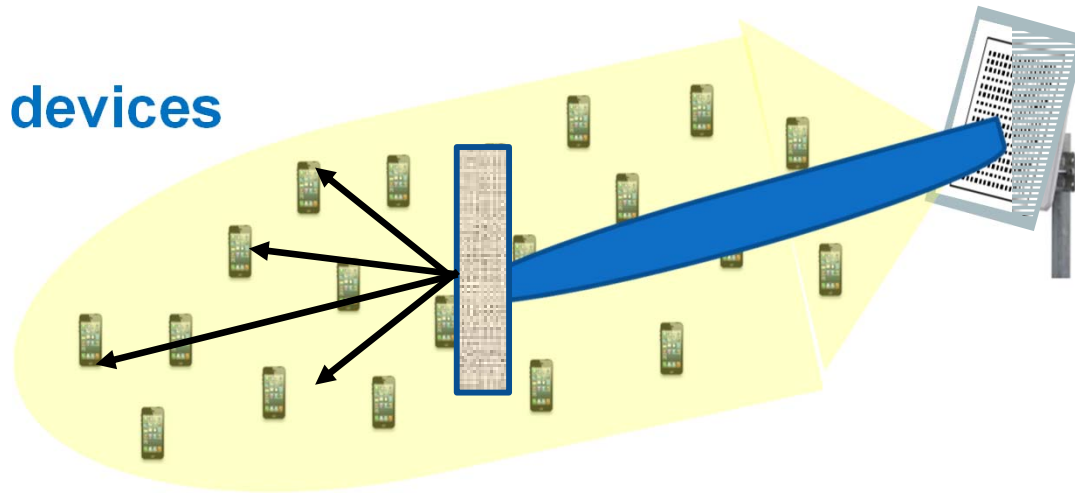
- Classic Communication System





# Massive MIMO

Dozens of devices

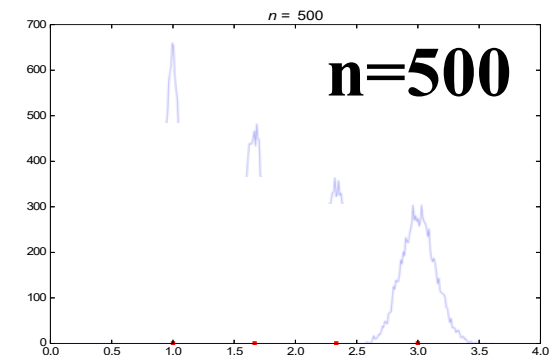
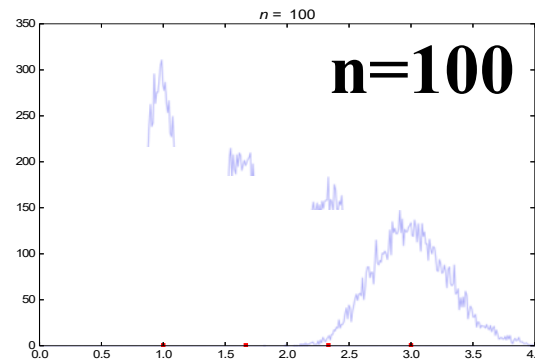
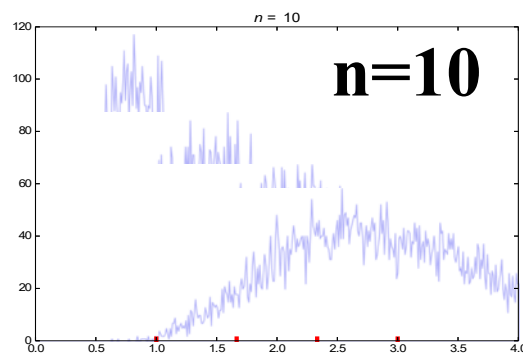


Hundreds of antennas

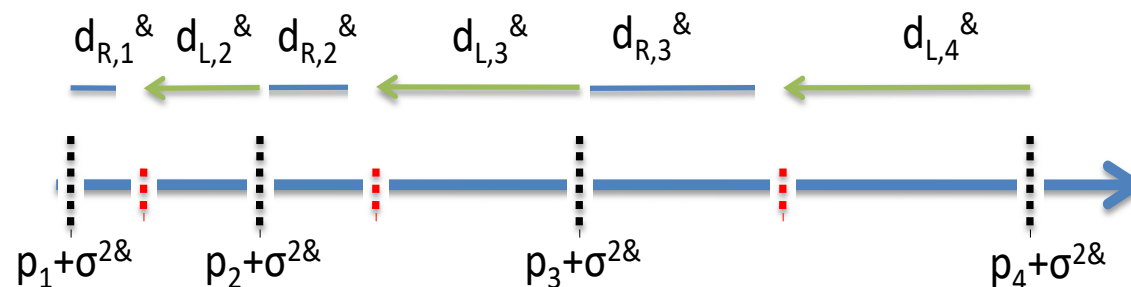
- Massive MIMO removes attenuation, fading, interference
- Bottlenecks: channel estimation, complexity
- Ideal beamforming disappears with shadowing
- Need multihop/mesh networks

# Non-coherent massive MIMO

- Propose simple energy-based modulation
- No capacity loss for large arrays:  $\lim_{n \rightarrow \infty} C_{nocsi} = \lim_{n \rightarrow \infty} C_{csi}$ 
  - Holds for single/multiple users (1 TX antenna, n RX antennas)



- Constellation optimization: unequal spacing



# Blind MIMO Decoding via Vertex Hopping

Given samples in the form:

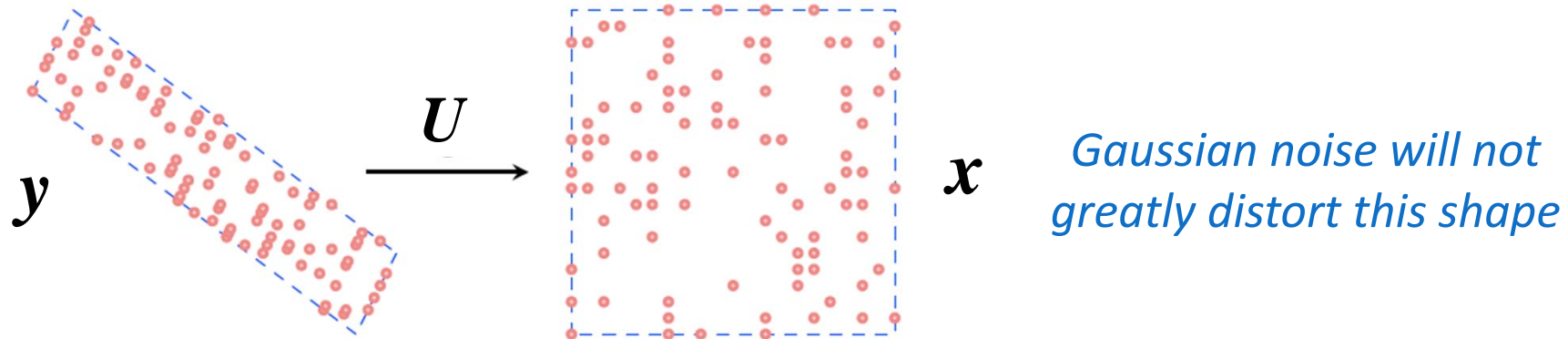
$$\mathbf{y} = \mathbf{A}\mathbf{x} + \mathbf{e}$$

- $\mathbf{A} \sim \mathcal{N}(0, 1)^{n \times n}$ ,  $\mathbf{e} \sim \mathcal{N}(0, \sigma^2)^{n \times n}$ .
- $\mathbf{x}$  is drawn from an MPAM or BPSK constellation (source must be hypercubic).
- Rich scattering, small MIMO ( $2 \leq n \leq 12$ ).

## Blind MIMO Decoding

In a block-fading environment, estimate  $\mathbf{A}$  and recover  $\mathbf{x}$  given only  $k$  samples of  $\mathbf{y}$

# Fitting a Parallelepiped --- Algorithms



$$\underset{U}{\text{maximize}} \quad \log |\det U| \quad (*)$$

$$\text{subject to} \quad |U y_i|_{\infty} \leq 1 + c, i = 1, \dots, k$$

- (\*) is a non-convex optimization problem
- Constrained gradient descent works but is slow.
- The Vertex Hopping algorithm uses concepts from solving mixed-integer linear programming to solve (\*).

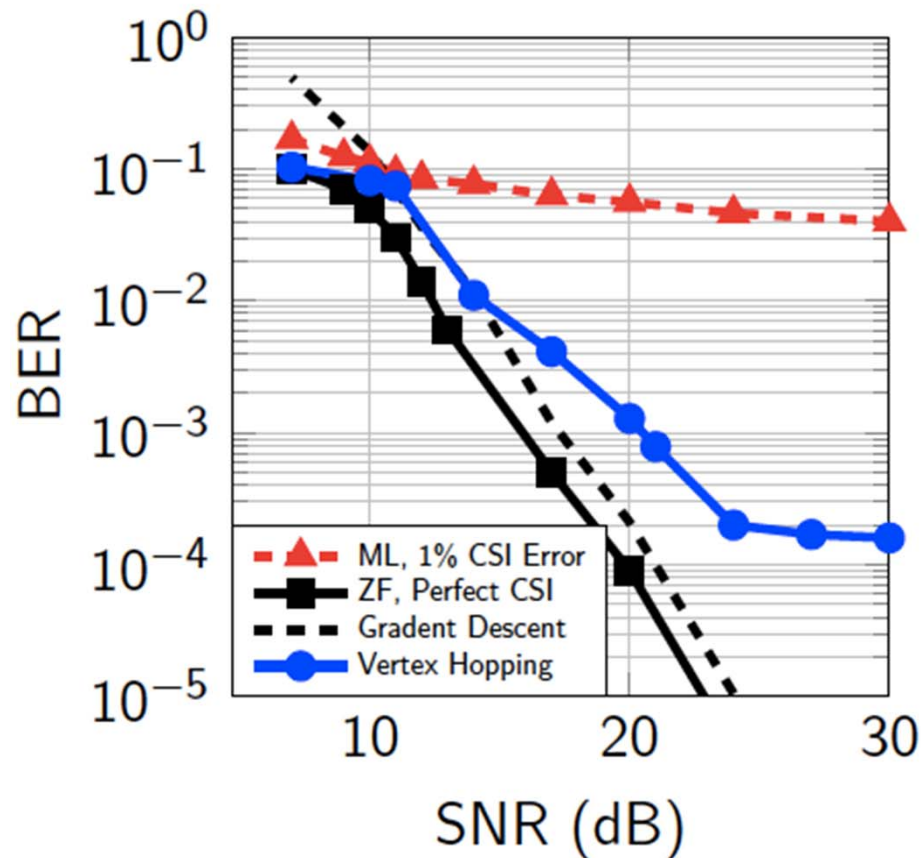
# Runtime Performance

		Vertex Hopping*		Gradient Descent†	
n	k	Pr[success]	Time (s)	Pr[success]	Time (s)
2	8	1.0	$1.83 \times 10^{-5}$	0.99	$3.01 \times 10^{-2}$
3	13	1.0	$6.46 \times 10^{-5}$	0.99	$6.33 \times 10^{-2}$
4	18	1.0	$1.74 \times 10^{-4}$	0.99	0.13
5	18	1.0	$2.96 \times 10^{-4}$	0.97	0.30
6	22	1.0	$8.52 \times 10^{-4}$	0.93	0.59
8	30	0.99	$4.99 \times 10^{-3}$	0.80	3.5
10	45	0.99	$5.36 \times 10^{-2}$	0	-
12	60	0.99	$3.70 \times 10^{-1}$	0	-

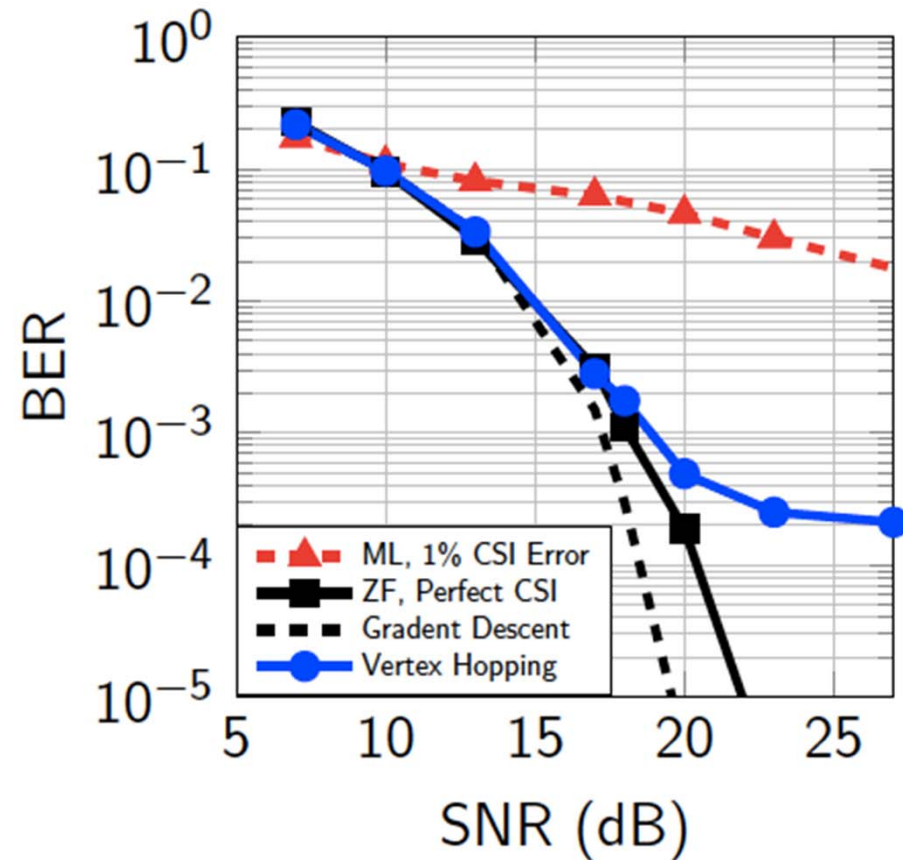
\***Implemented in Rust**  
†**MATLAB's fmincon**

# AWGN and Fading Performance

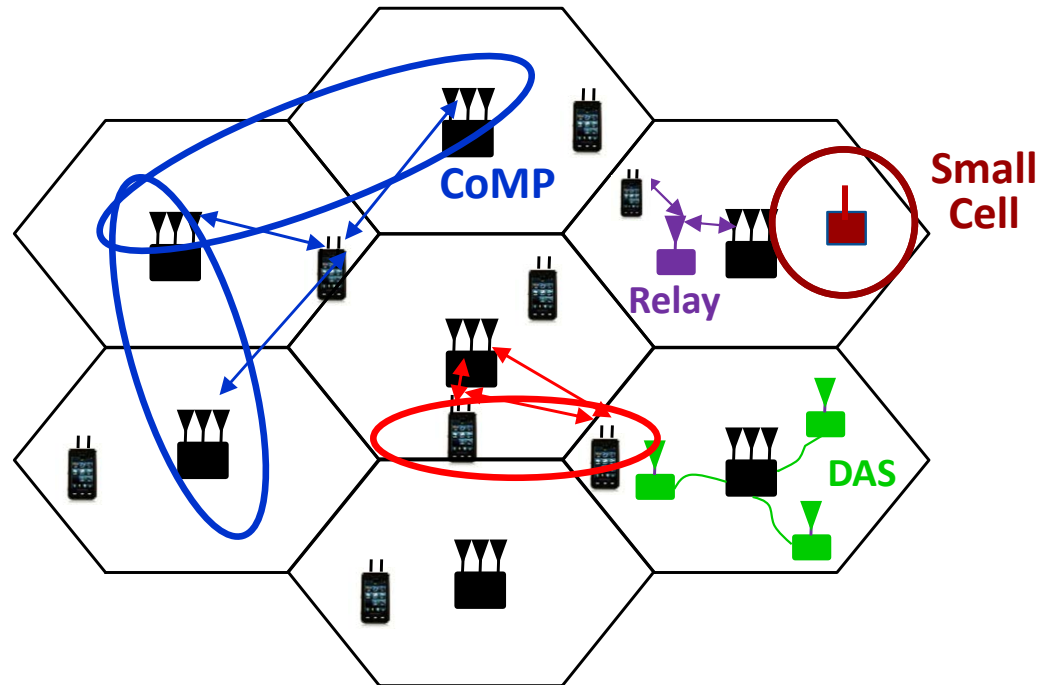
$$a_{i,j} \sim \mathcal{N}(0, 1)$$



Rayleigh Fading



# Rethinking Cellular System Design



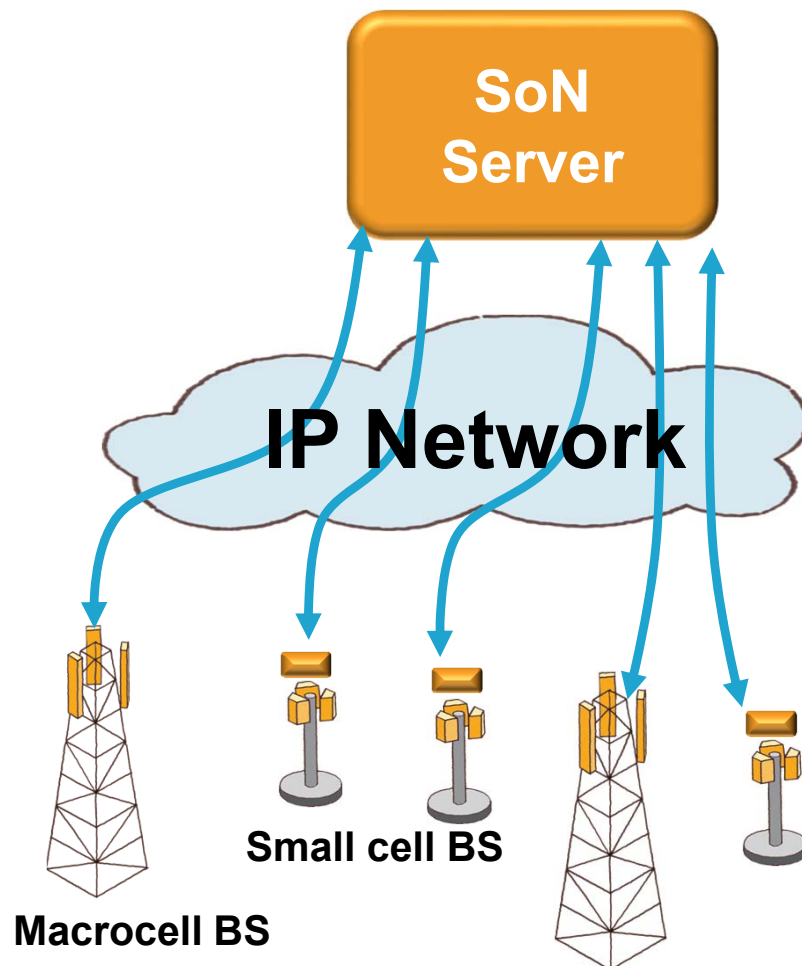
*How should cellular systems be designed?*

*Will gains be big or incremental; in capacity, coverage or energy?*

- Cellular systems reuse channels/timeslots in different cells
  - Traditional design assumes system is “interference-limited”
  - Capacity unknown; upper bound based on BC/MAC with pooled antennas
- No longer the case with recent technology advances:
  - MIMO, multiuser detection, cooperating BSs (CoMP) and relays
  - Raises interesting questions such as “what is a cell?”
- Dynamic self-organization (SoN) needed for deployment and optimization

# Small cells are the solution to increasing cellular system capacity

In theory, provide exponential capacity gain



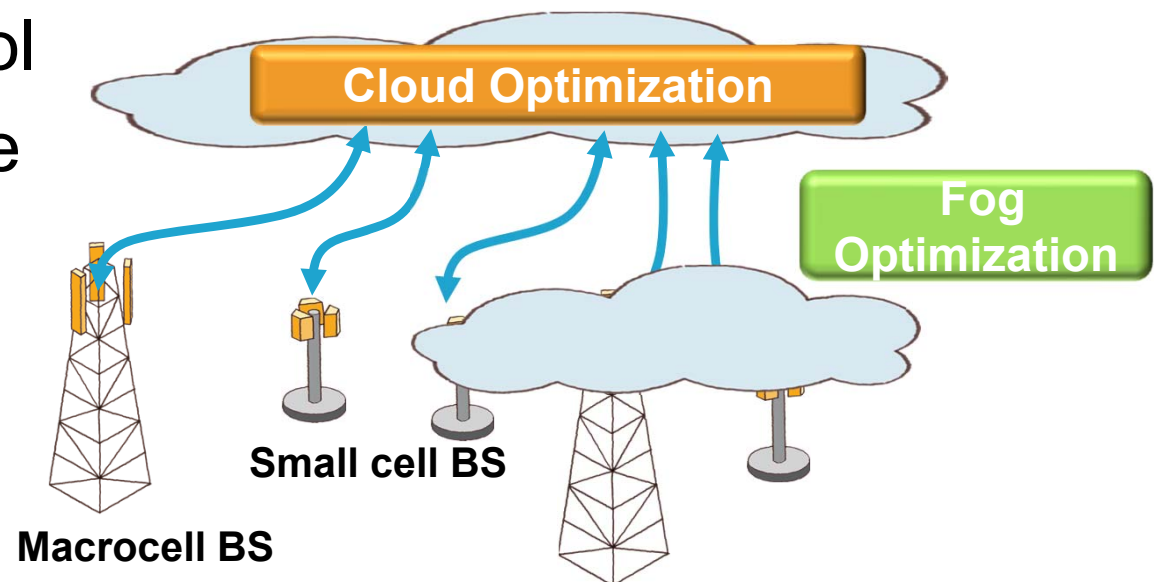
- Future cellular networks will be hierarchical
- Large cells for coverage
- Small cells for capacity and power efficiency
- Small cells require self-optimization in the **cloud**



# HetNet Optimization Challenge

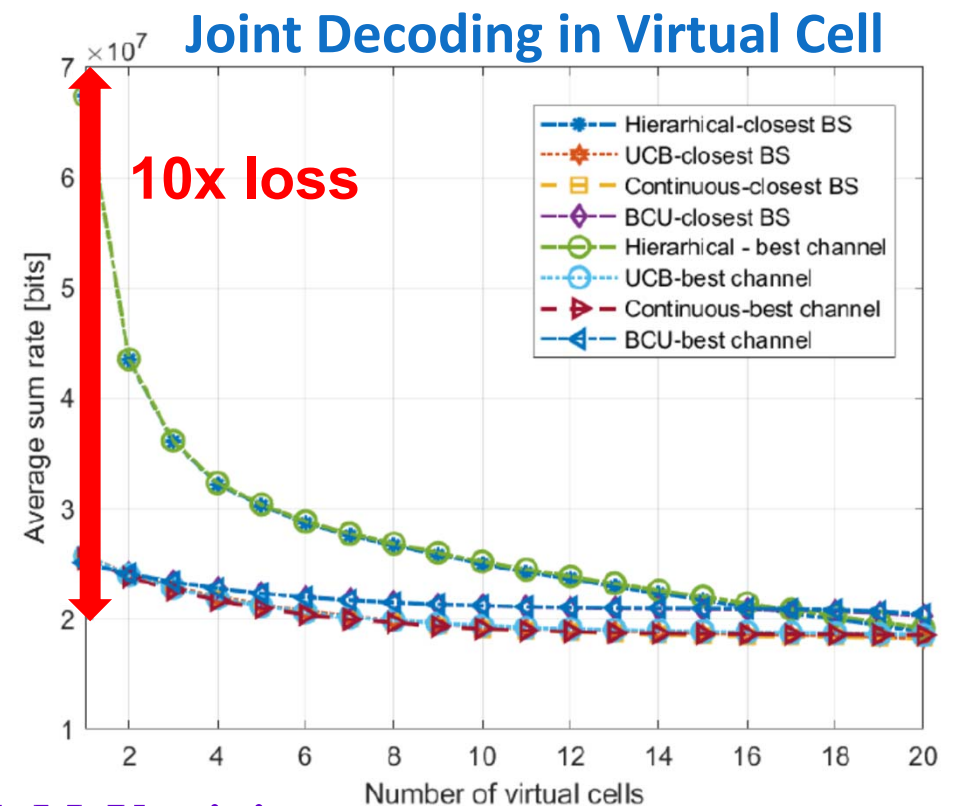
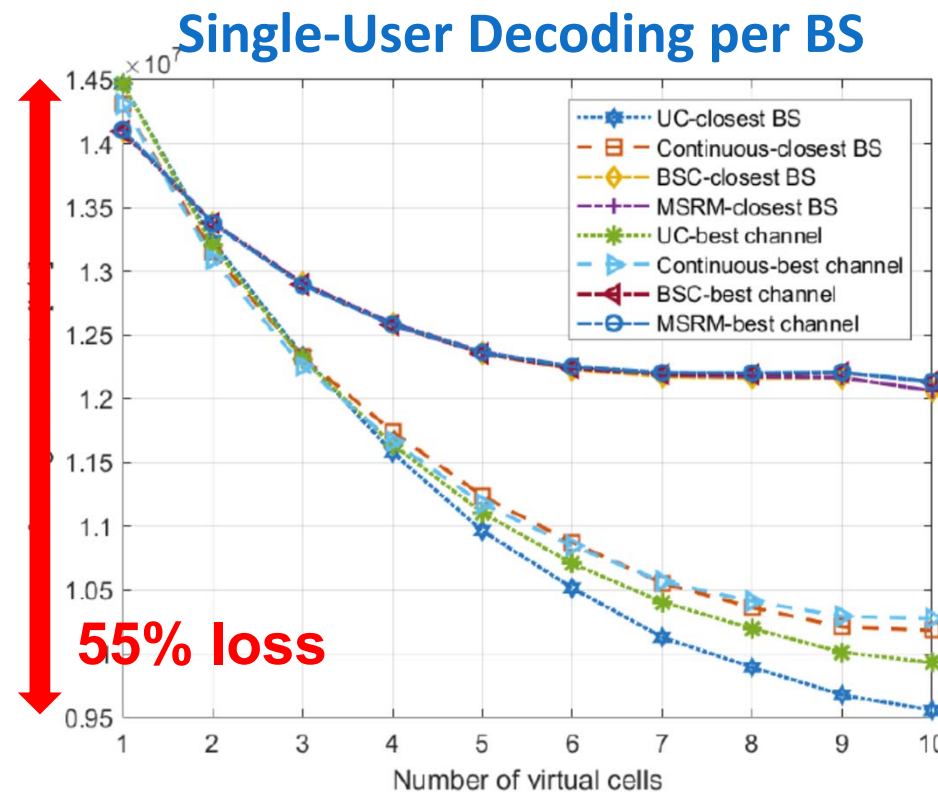
- Algorithmic complexity
  - Frequency allocation alone is NP hard
  - Also have MIMO, power control, CST, hierarchical networks: *NP-really-hard*
  - Advanced optimization tools needed, including a **combination** of centralized (cloud) distributed, and locally centralized (fog) control
  - **ML** can also play a role

Next challenge:  
optimizing caching  
and edge computing



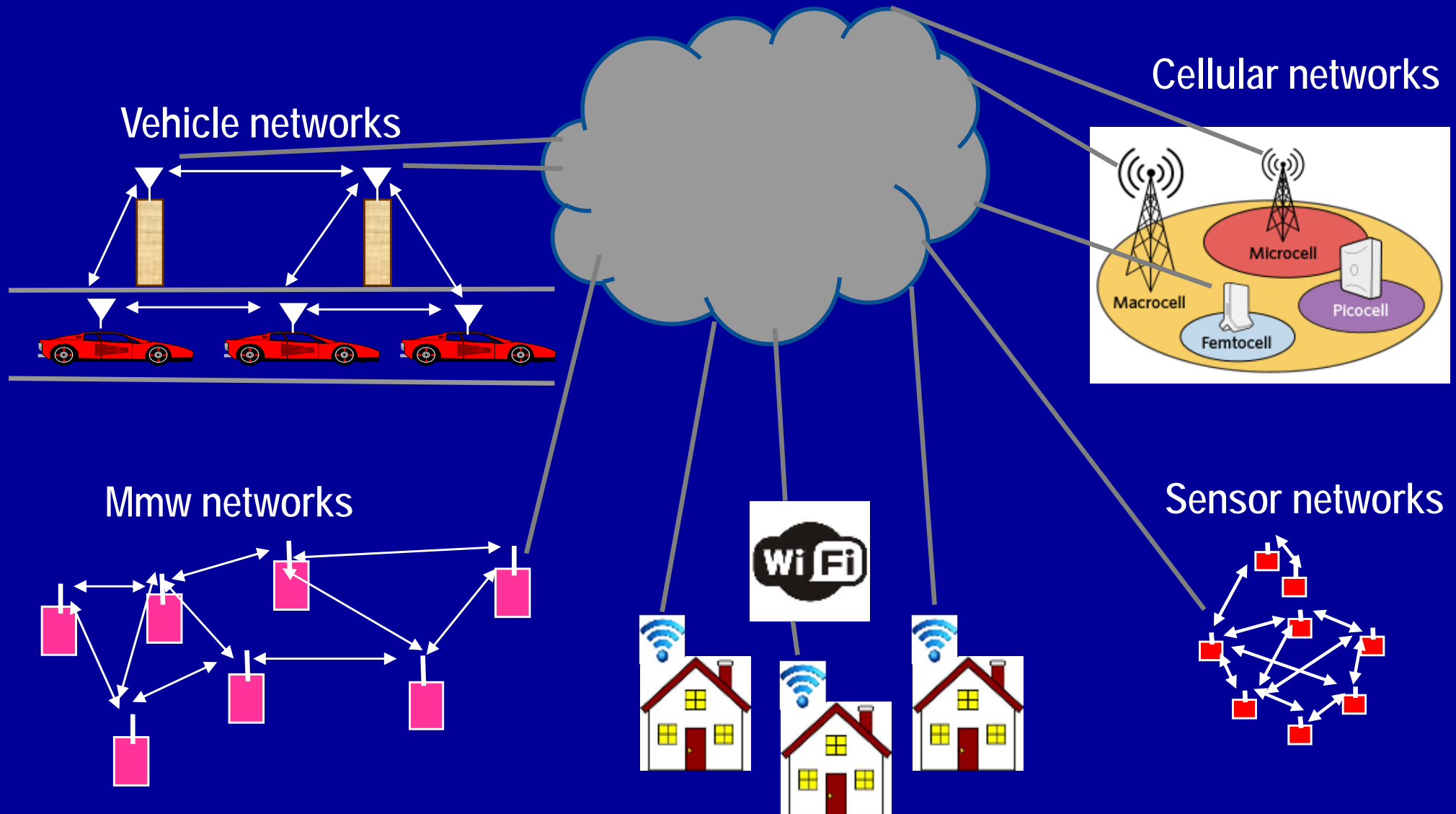
# Fog-Optimization vs. Centralized

- Use clustering technique to cluster BSs, then optimize power allocation to maximize uplink sum rate
  - Consider multiple clustering techniques (not much difference)
  - Nonconvex approximation for optimization

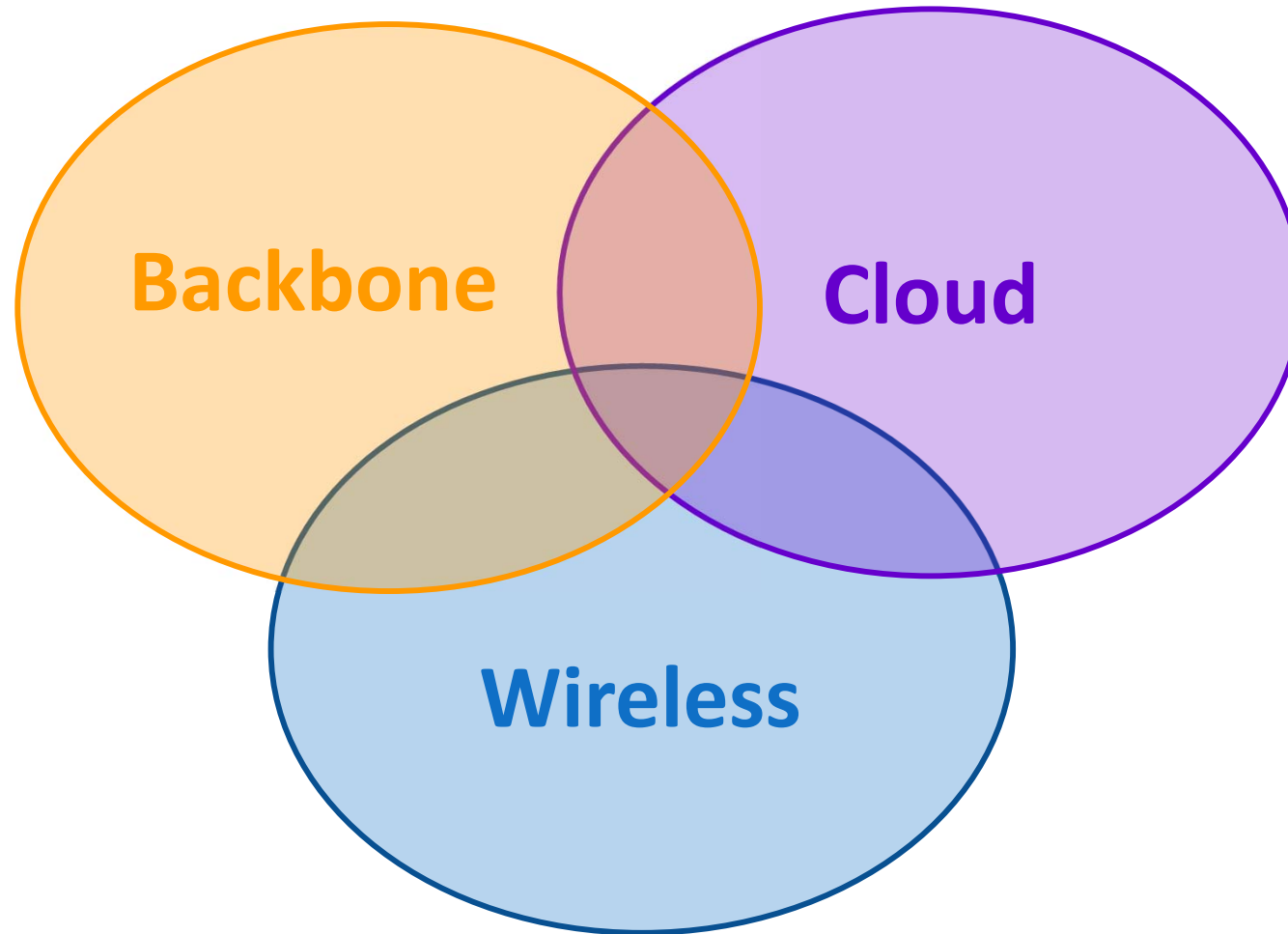


Joint work with M. Yemini

# Cloud-Enabled Wireless Networks (includes “fog”)

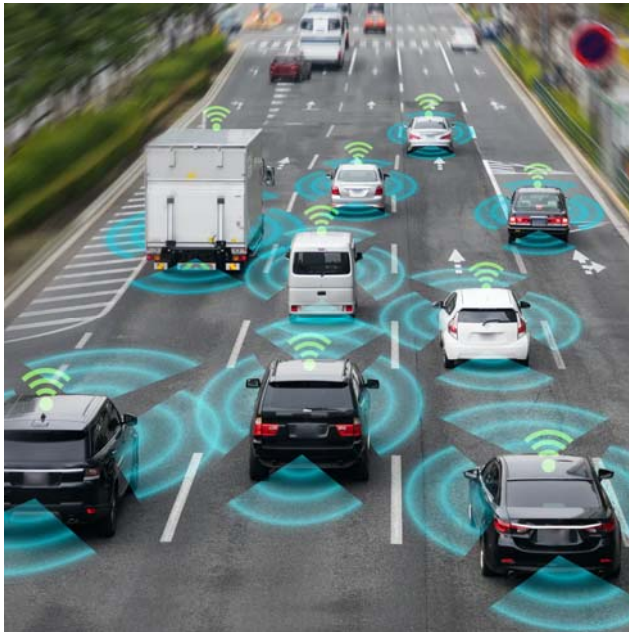


# NextG Network Convergence

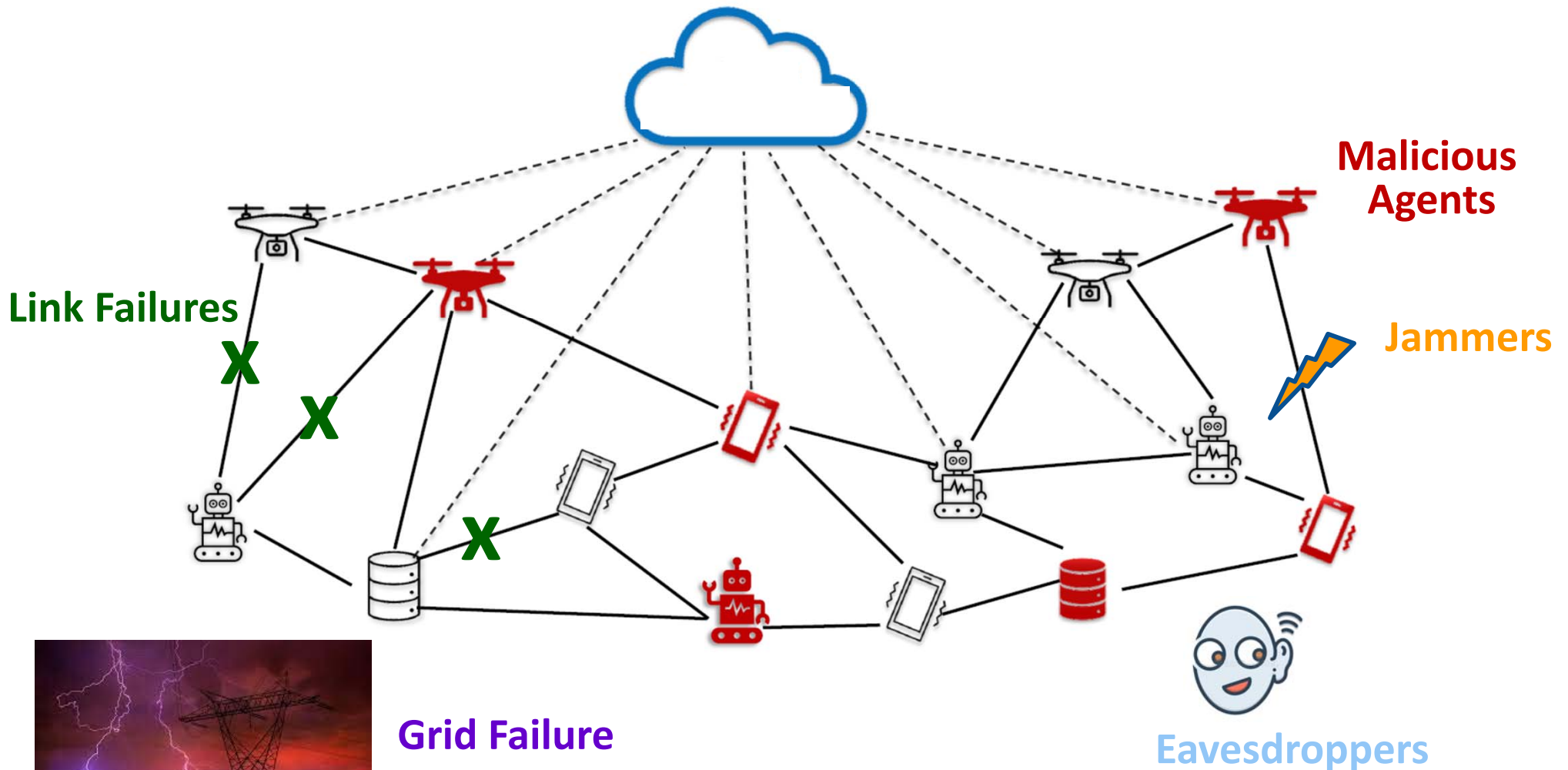


**What design principles should drive this convergence?**

# This design is critical for NextG applications

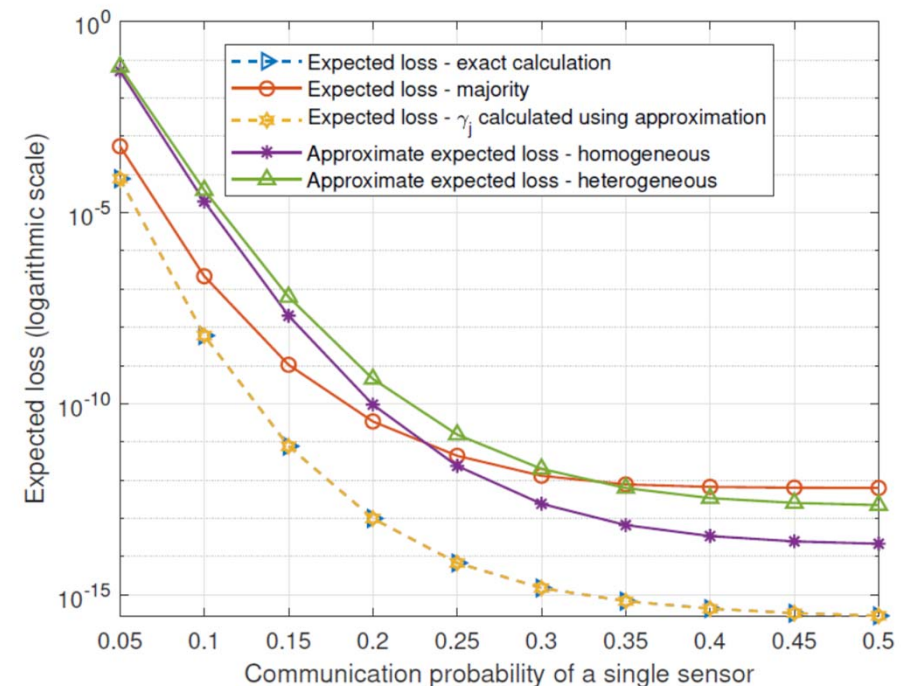
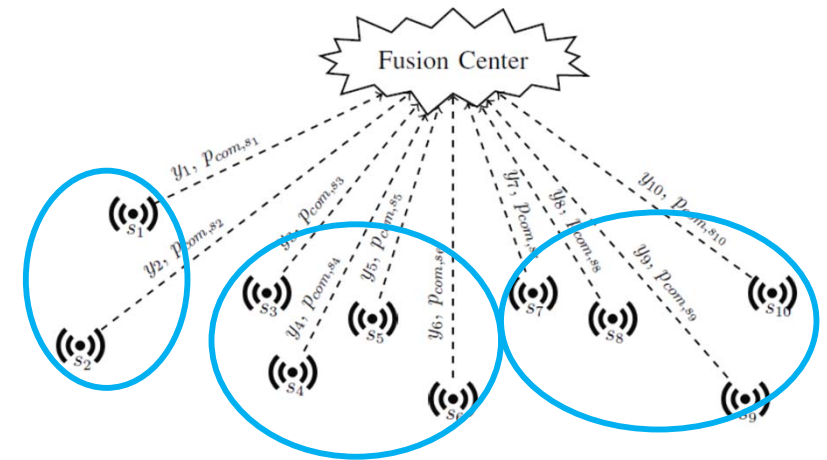


# Collaboration over Wireless Networks: Security, Privacy, and Resilience



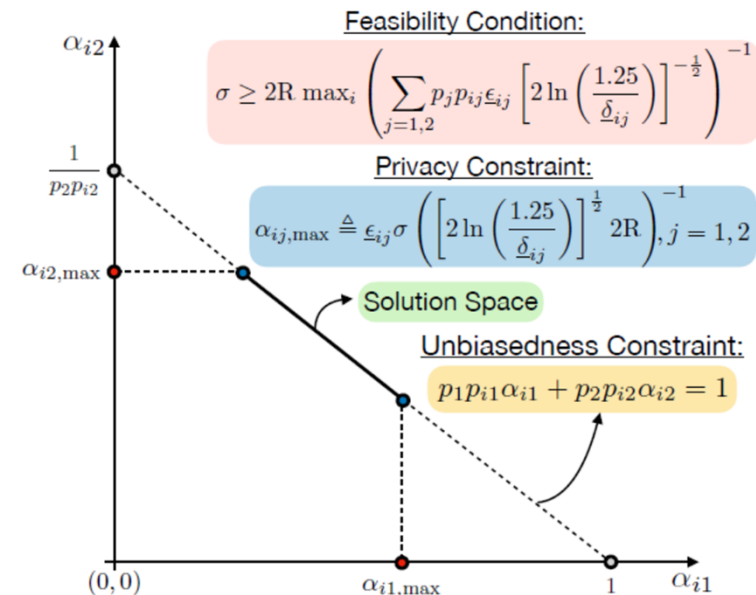
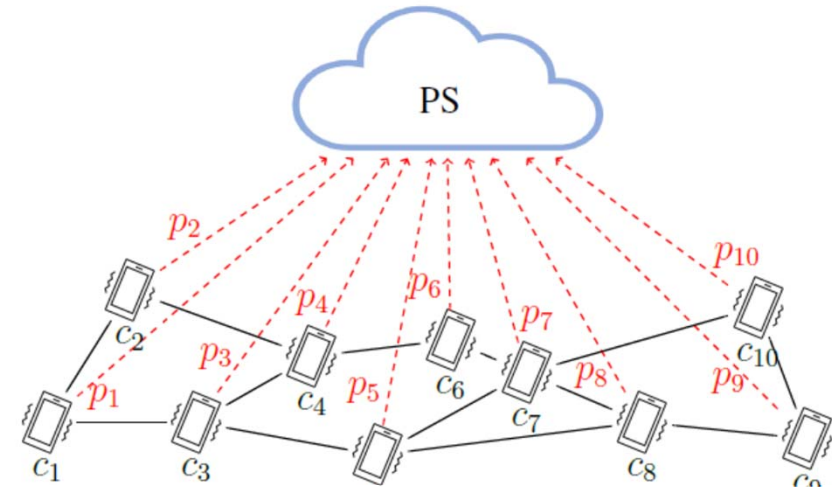
# Centralized Detection with Flaky Links

- Sensors have intermittent connectivity to fusion center (FC)
- Sensors collaborate within predefined sensor clusters
  - Fuse their noisy sensor data to reach a common local estimate
  - Common estimate sent to FC
- FC fuses the received estimates for event detection
- Determine optimal decision rule with tractable complexity
- Excellent detection performance



# Federated Learning with Flaky Links

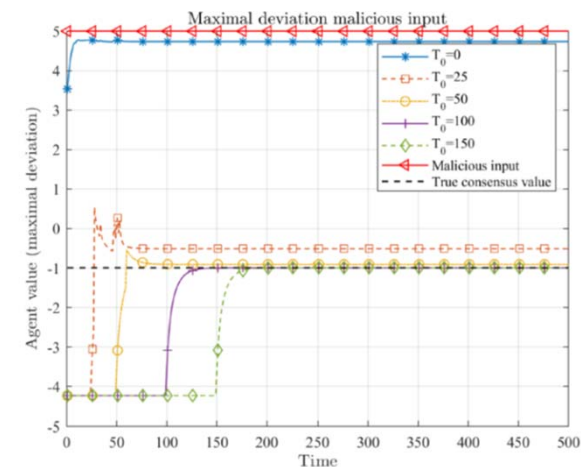
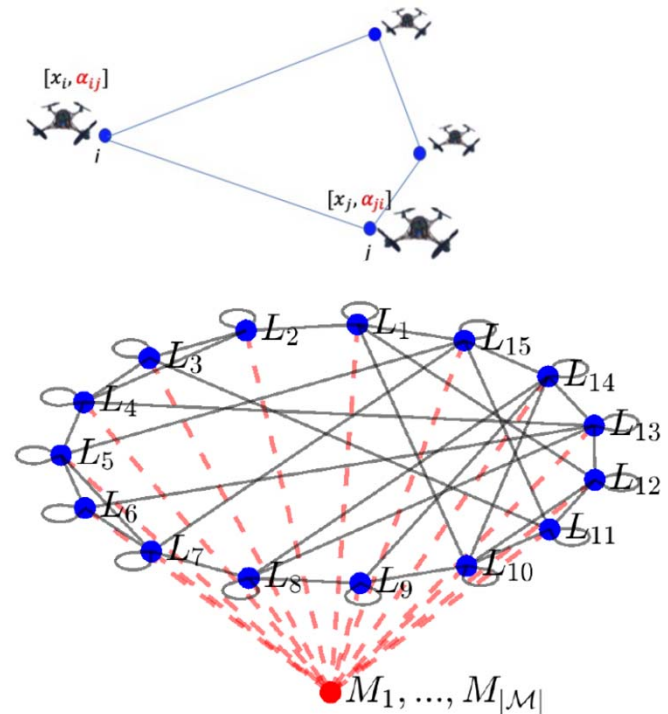
- Neighbors collaborate to send data to central parameter server
- Each client computes consensus estimate from neighbors
  - Weighted by connectivity
- Algorithm exhibits improved convergence and accuracy
- Can also introduce privacy constraints





# Trust and Resilience in Distributed Consensus

- Consensus algorithms assume trusted agents
- We ask: how to achieve consensus when there are malicious agents
- Propose a consensus algorithm with untrusted agents, with almost sure convergence
  - Even when malicious agents constitute more than half of the network connectivity
- The true consensus value bounded with  $p \rightarrow 1$ .
- Correct classification of malicious and legitimate agents can be attained in finite time
- Expected convergence rate decays exponentially with trust



# Challenges to NextG technology disruption



- Complexity of current networks
- Lack of significant research investment for a decade
- Standards process can stifle innovation
- NextG hardware and software proprietary and closed
- Research silos across wireless, backbone networks, cloud, electronics, and applications

# New era for US technology innovation

2022  
CHIPS and Science Act  
2022



# Killer App for 6G?



***Connecting the next billion***

# BRIDGING THE DIGITAL DIVIDE



- Cellular architectures for long-range coverage
- Satellite systems
- Low-cost hardware and devices
- We have the technology today
  - A question of cost, commitment, policy, and politics

# Summary

- The next wave in wireless technology is upon us
  - Will enable new applications that will change people's lives
- Future wireless networks must support high rates, extreme energy efficiency, and low latency
  - With robust security, privacy, and resilience
- Many challenges to creating the technical innovations and disruptions needed to achieve this vision
- Connecting the next billion is the killer app for 6G