

# Type-Based Unsourced Multiple Access over Fading Channels with Cell-Free Massive MIMO

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Kaan Okumus<sup>\*</sup>, Khac-Hoang Ngo<sup>†</sup>, Giuseppe Durisi<sup>\*</sup>, Erik G. Ström<sup>\*</sup>

<sup>\*</sup> Department of Electrical Engineering, Chalmers University of Technology, Gothenburg, Sweden

<sup>†</sup> Department of Electrical Engineering (ISY), Linköping University, Linköping, Sweden

okumus@chalmers.se

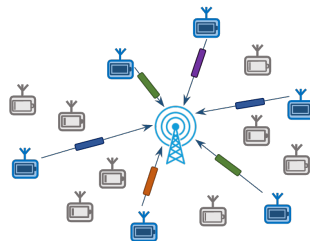
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# Massive uncoordinated access systems

Future 6G and IoT technologies will need to support massive machine-type communication (mMTC)



## Requirements for mMTC

- Data collection from low-cost sensor networks over a **shared medium**
- **Uncoordinated** and **sporadic** ( $\leq 1/\text{min}$ ) transmissions
- **Massive** number of devices ( $\sim 10^7/\text{Km}^2$ )

Any scheme that assigns dedicated resources to each user is not scalable

# Unsourced multiple access (UMA) [[Polyanskiy, 2017]]

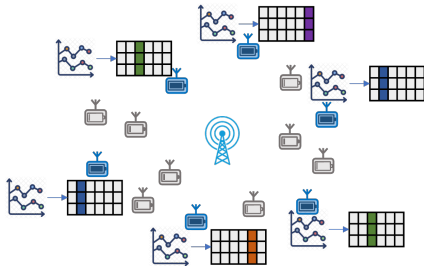
## UMA framework

- **Same encoder** for all users
- Receiver produces an **unordered** list of messages  $\Rightarrow$  “unsourced”

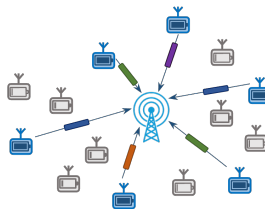
## Limitation of UMA

**Message collisions** treated as errors in UMA

# Do message collisions occur in practice?



Sampling and quantization



Data transmission

- Users may transmit the same quantized information
- Applications: multi-target tracking, federated learning
- We are interested in both the set of transmitted messages and their **multiplicities**

# Type-based UMA (TUMA)

An extension of UMA that handles message collision

**Goal:** Estimate type, i.e., empirical distributions of messages

## TUMA over Gaussian MAC [Ngo et al., 2024]

- Type estimation formulated as a compressed sensing problem
- Approximate message passing (AMP) algorithm employed for decoding
- Assumes perfect power control
- Multiplicities estimated from the received codeword power

Extension to fading is not straightforward

# How to handle fading channels?

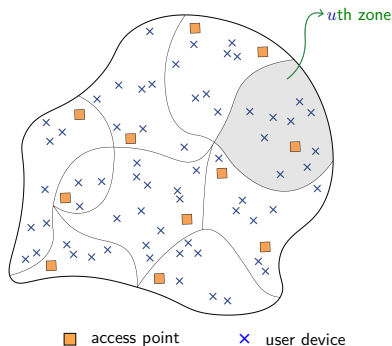
## The benefits of cell-free architectures in UMA

[[Cakmak et al., 2025]]: UMA within cell-free massive MIMO

- Rayleigh fading channel
- Perfect knowledge of large-scale fading model
- Location-based codebook partition
- Multisource AMP decoder

Same ingredients can be used for TUMA under fading

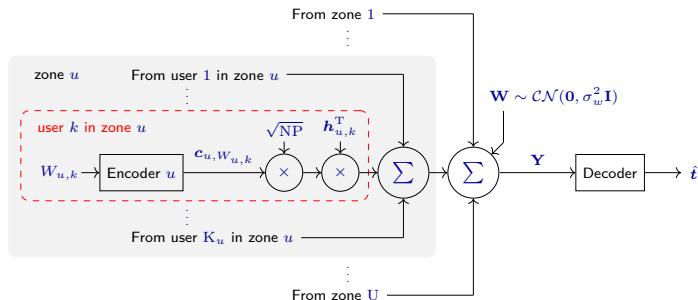
# TUMA over a cell-free massive MIMO architecture



## TUMA with cell-free setup:

- area partitioned into  $U$  zones
- users in a zone have similar large-scale fading coefficients
- codebook partitioned into  $U$  zones
- codebook for zone  $u$ :  $\mathbf{C}_u \in \mathbb{C}^{N \times M}$
- $N$ : blocklength
- $M$ : number of possible messages
- $B$  access points (APs) with  $A$  antennas each
- $F = B \times A$  antennas in total
- APs connected to a CPU via fronthaul links

# System overview of TUMA over CF Massive MIMO

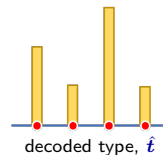
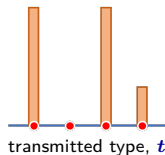


- User  $k$  in zone  $u$  has message  $W_{u,k}$  and transmits codeword  $c_{u,W_{u,k}}$ , scaled by  $\sqrt{NP}$
- Fading modeled via channel vector  $h_{u,k} \in \mathbb{C}^F$  (from user to all AP antennas)
- CPU receives superposition signal  $\mathbf{Y}$  and aims to estimate the type vector  $\mathbf{t}$  from it



# Type estimation and performance metric

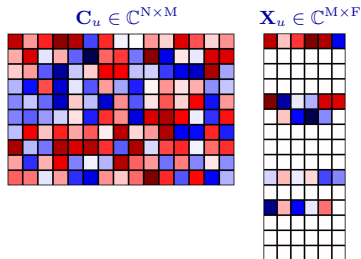
- Messages  $\{W_{u,k}\}$  form a global type vector  $\mathbf{t} = [t_1, \dots, t_M]$
- $t_m$  is the fraction of users transmitting message  $m$
- Decoder returns a type estimate  $\hat{\mathbf{t}}$  from  $\mathbf{Y}$ , using known codebooks  $\{\mathbf{C}_u\}$  and a statistical path loss model (no instantaneous CSI)



Performance metric: total variation distance

$$\mathbb{TV}(\mathbf{t}, \hat{\mathbf{t}}) = \frac{1}{2} \sum_{m=1}^M |t_m - \hat{t}_m|$$

# TUMA as a compressed sensing problem



- Users who transmit the same message contribute to the same row of  $\mathbf{X}_u$
- The  $m$ th row is:

$$x_{u,m} = \sum_{k: W_{u,k}=m} h_{u,k}$$

- Only a few messages are active  $\rightarrow \mathbf{X}_u$  is row-sparse

Global system equation:

$$\mathbf{Y} = \sqrt{NP} \sum_{u=1}^U \mathbf{C}_u \mathbf{X}_u + \mathbf{W}$$

This is a structured compressed sensing problem (multisource)

# Proposed centralized TUMA decoder

We use **multisource AMP** [Cakmak et al., 2025] to estimate  $\{\mathbf{X}_u\}$  and enable type estimation

For  $t = 1, 2, \dots$ , iterate:

- Effective observation:  $\mathbf{R}_u^{(t)} = \mathbf{C}_u^H \mathbf{Z}^{(t-1)} + \sqrt{\text{NP}} \mathbf{X}_u^{(t-1)}$
- Bayesian denoising:  $\mathbf{X}_u^{(t)} = \eta_{u,t}(\mathbf{R}_u^{(t)})$
- Onsager correction:  $\mathbf{Q}_u^{(t)}$  via Jacobian of  $\eta_{u,t}$
- Residual update:  $\mathbf{Z}^{(t)} = \mathbf{Y} - \sqrt{\text{NP}} \sum_{u=1}^U \left( \mathbf{C}_u \mathbf{X}_u^{(t)} - \frac{M}{N} \mathbf{Z}^{(t-1)} \mathbf{Q}_u^{(t)} \right)$

Asymptotically, the model  $\mathbf{Y} = \sqrt{\text{NP}} \sum_{u=1}^U \mathbf{C}_u \mathbf{X}_u + \mathbf{W}$  decouples into  $M$  independent submodels:

$$\mathbf{r}_{u,m}^{(t)} = \sqrt{\text{NP}} \mathbf{x}_{u,m}^{(t)} + \mathcal{CN}(\mathbf{0}; \mathbf{T}^{(t)}),$$

where  $\mathbf{T}^{(t)}$  is a diagonal covariance matrix computed via state evolution

The Bayesian denoiser  $\eta_{u,t}$  is designed for this decoupled model to estimate  $\mathbf{x}_{u,m}^{(t)}$

# Proposed centralized TUMA decoder

## Our contributions:

Bayesian denoiser  $\eta_{u,t}$  is tailored to the TUMA system using:

- Prior on multiplicities  $k_{u,m}$ : **truncated binomial distribution**
- Prior on user positions: **uniform distribution** over the zone

Challenges due to message collisions:

- High-dimensional integrals in denoiser and Onsager correction
- Solved using **Monte Carlo sampling** for scalability

Type estimation procedure:

- Bayesian estimation of type  $\hat{t}$  (using posteriors already computed to evaluate  $\eta_{u,t}$ )

# Proposed distributed TUMA decoder

To increase scalability, we design a **distributed decoder** inspired by **distributed AMP** [Bai et al., 2022]

**Key idea:** Move likelihood computations from CPU to the access points (APs)

**At each AP:**

- Process local received signal
- Compute local likelihoods using Monte Carlo sampling

**At the CPU:**

- Fuse them to form global likelihoods
- Perform Bayesian type estimation (same as centralized)

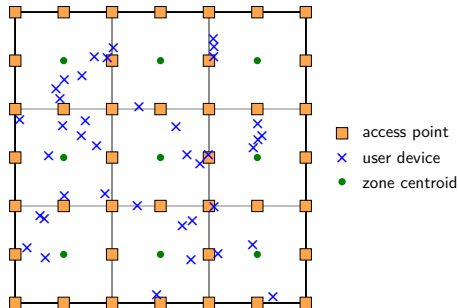
## Benefits compared to centralized decoder:

Reduced CPU load, reduced fronthaul bandwidth, scalable to large CF networks

## Drawbacks:

Slight performance loss compared to centralized decoder

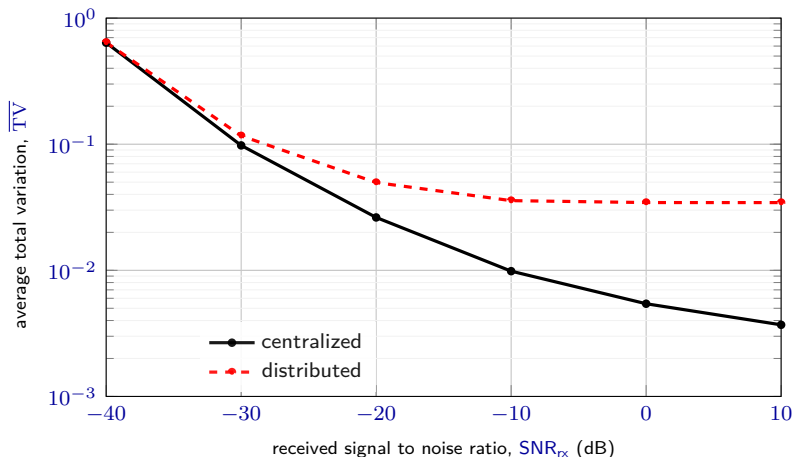
# Simulation settings



## Example of TUMA with cell-free setup:

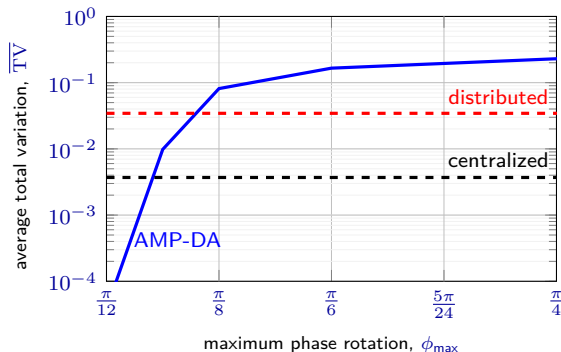
- $300\text{ m} \times 300\text{ m}$  coverage area
- $U = 9$  zones
- $B = 40$  APs with  $A = 4$  antennas each
- Blocklength  $N = 1024$
- $M = 2^{12}$  possible messages
- 180 active users (20 users per zone)
- 13 distinct messages per zone
- average multiplicity is 2
- maximum multiplicity is 7

# Simulation results—comparison between proposed decoders



# Simulation results—comparison with AMP-DA decoder

- AMP-DA [Qiao et al., 2024] is a type-based decoder tailored for federated learning
- AMP-DA requires perfect CSI at the UEs for channel pre-equalization
- In contrast, our TUMA decoders do not require instantaneous CSI
- Benchmark: performance of AMP-DA under imperfect CSI (random phase rotation)



Our decoders outperform AMP-DA even for small phase rotation



# Conclusion

- We extended the **TUMA** framework to **fading channels** using a **CF massive MIMO** architecture
- Proposed **centralized** and **distributed decoders** based on multisource AMP
- Our decoders handle **message collisions** and unknown user positions using **Bayesian estimation**
- **Monte Carlo sampling** enables scalable implementation of the denoiser
- Simulation results show improved **type estimation** and **robustness compared to AMP-DA**

Code and examples available at: [github.com/okumuskaan/tuma\\_fading\\_cf](https://github.com/okumuskaan/tuma_fading_cf)

