

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

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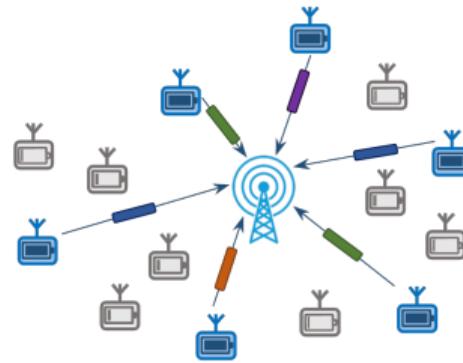
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Massive uncoordinated access systems

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]]

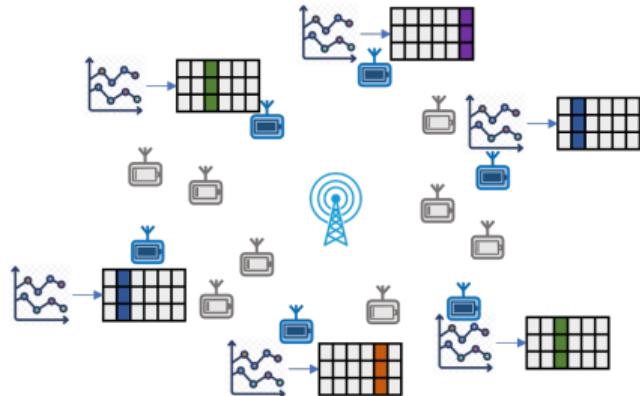
UMA framework

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

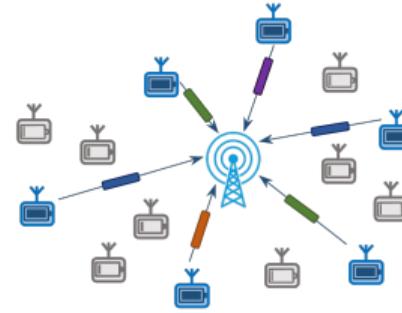
Limitation of UMA

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?

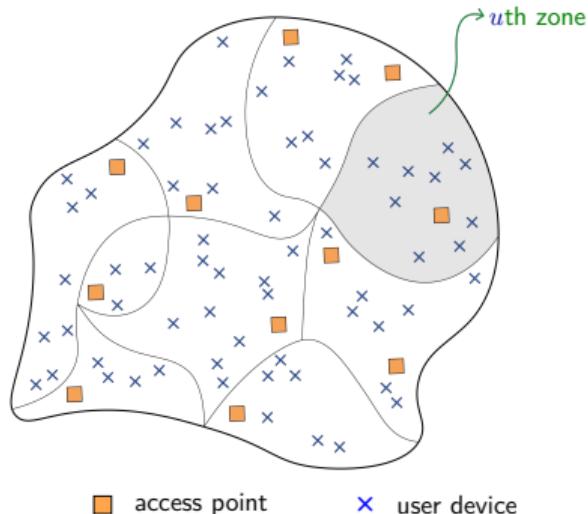
The benefits of cell-free architectures in UMA

[[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

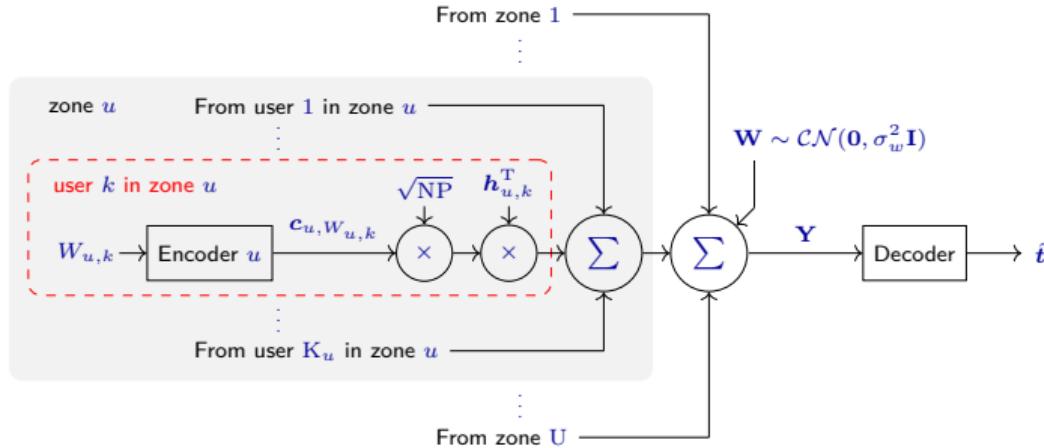
TUMA over a cell-free massive MIMO architecture



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

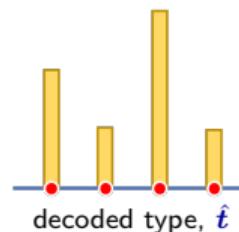
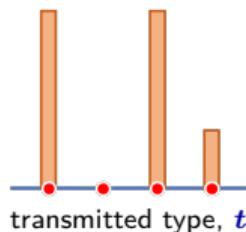
System overview of TUMA over CF Massive MIMO



- 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)

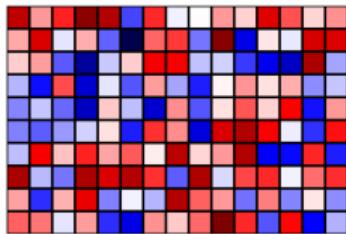


Performance metric: total variation distance

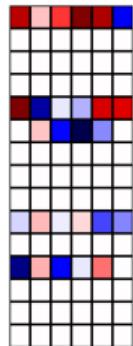
$$\mathbb{TV}(\mathbf{t}, \hat{\mathbf{t}}) = \frac{1}{2} \sum_{m=1}^M |t_m - \hat{t}_m|$$

TUMA as a compressed sensing problem

$$\mathbf{C}_u \in \mathbb{C}^{N \times M}$$



$$\mathbf{X}_u \in \mathbb{C}^{M \times F}$$



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

$$\mathbf{x}_{u,m} = \sum_{k: W_{u,k}=m} h_{u,k}$$

- 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: $\mathbf{R}_u^{(t)} = \mathbf{C}_u^H \mathbf{Z}^{(t-1)} + \sqrt{NP} \mathbf{X}_u^{(t-1)}$
- 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: $\mathbf{Z}^{(t)} = \mathbf{Y} - \sqrt{NP} \sum_{u=1}^U \left(\mathbf{C}_u \mathbf{X}_u^{(t)} - \frac{M}{N} \mathbf{Z}^{(t-1)} \mathbf{Q}_u^{(t)} \right)$

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

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

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)

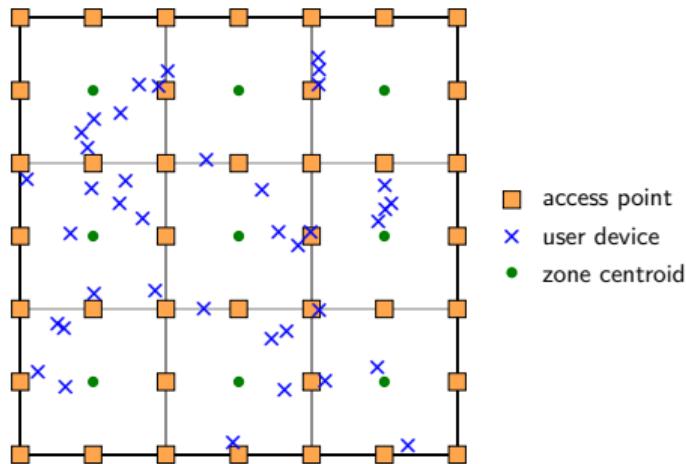
Benefits compared to centralized decoder:

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

Drawbacks:

Slight performance loss compared to centralized
decoder

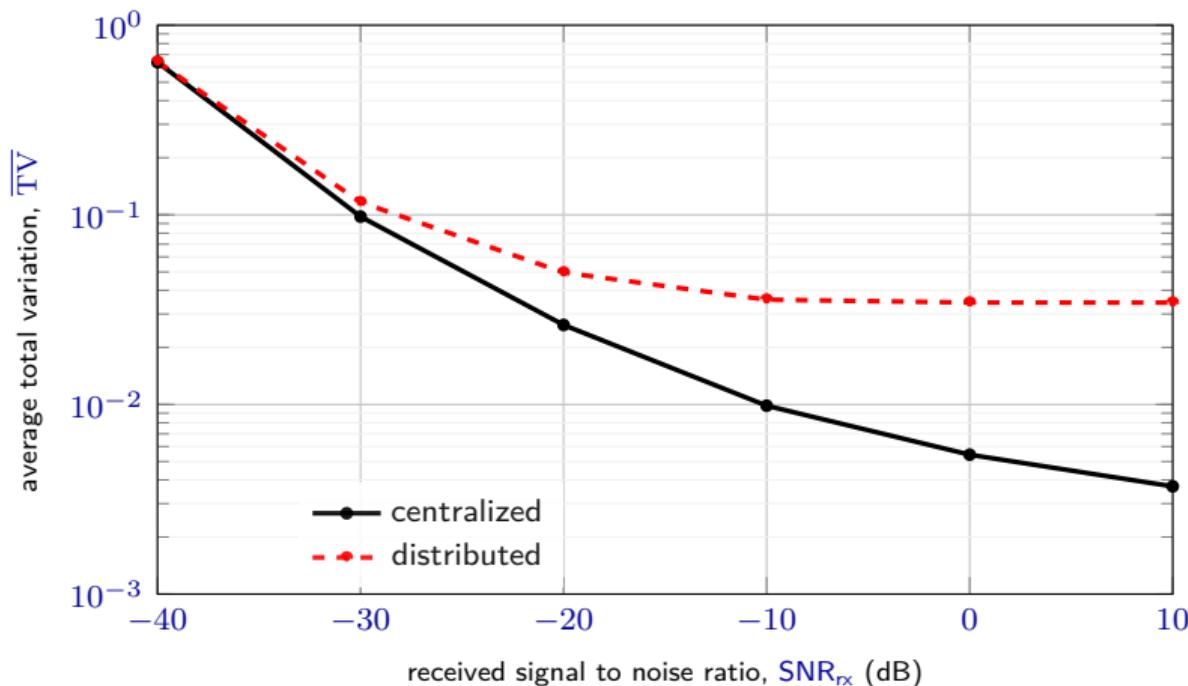
Simulation settings



Example of TUMA with cell-free setup:

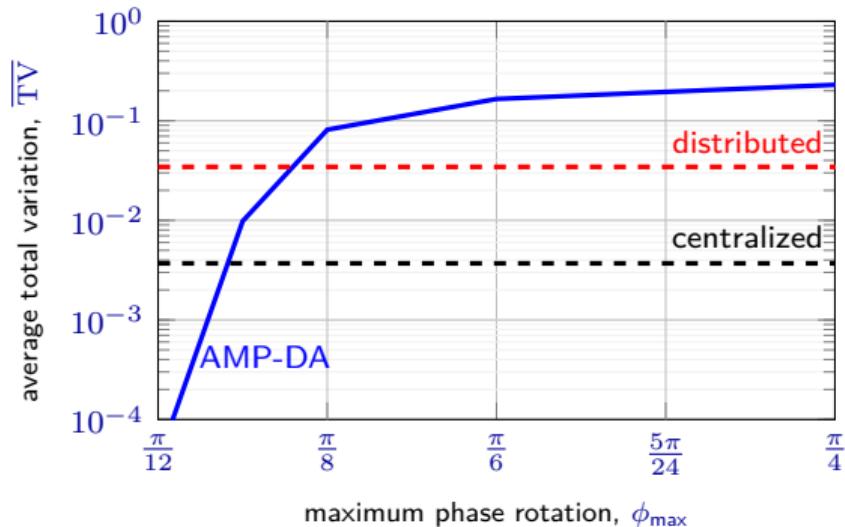
- $300 \text{ m} \times 300 \text{ m}$ coverage area
- $U = 9$ zones
- $B = 40$ APs with $A = 4$ antennas each
- Blocklength $N = 1024$
- $M = 2^{12}$ possible messages
- 180 active users (20 users per zone)
- 13 distinct messages per zone
- average multiplicity is 2
- maximum multiplicity is 7

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

