

Type-based unsourced multiple access

Giuseppe Durisi

Chalmers, Sweden

March, 2025



Wireless-communication-enabled massive connectivity



source: IoTpool

Challenge

- 🔧 Collect **in an energy efficient way** data from a **massive number of low-cost sensors**

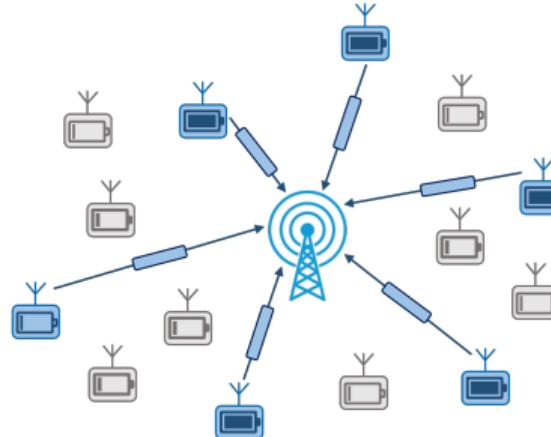
Massive wireless connectivity

Massive machine-type communication (mMTC)

- Mostly uplink
- Small information payload (**100** bits)
- High user density (10^7 devices per Km²)
- Sporadic transmission (less than once per minute)

Challenging problem

Around **120** complex degrees of freedom per user
for a **20 MHz** system

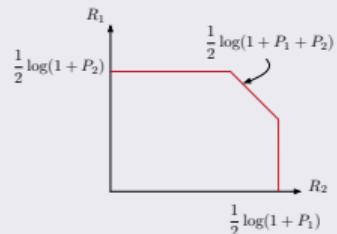


Key design question

How to transmit around **100** bits in around **100** d.o.f. per user over a MAC, under stringent energy-efficiency requirements

Traditional multiple access models and their limitations [Gallager '85]

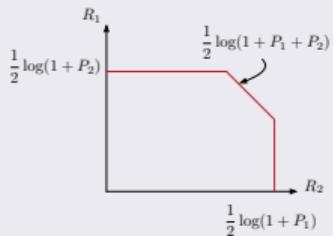
Multiaccess IT [Cover '75, Wyner '74]



- ✖ All users **active** (no sporadicity)
 - Each user is given a **different codebook**
- ✖ Not feasible for mMTC (**overhead** too large)

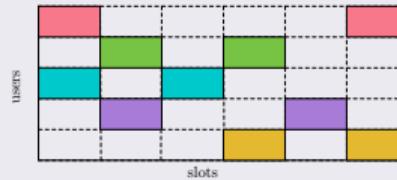
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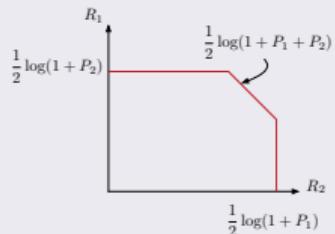
Collision resolution [Abramson '70, Roberts '72, Liva '11]



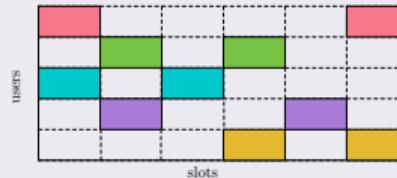
- ✓ Infinitely many, **sporadically** active users
- ✗ Crude modeling of communication aspects
 - De-facto **standard** for mMTC

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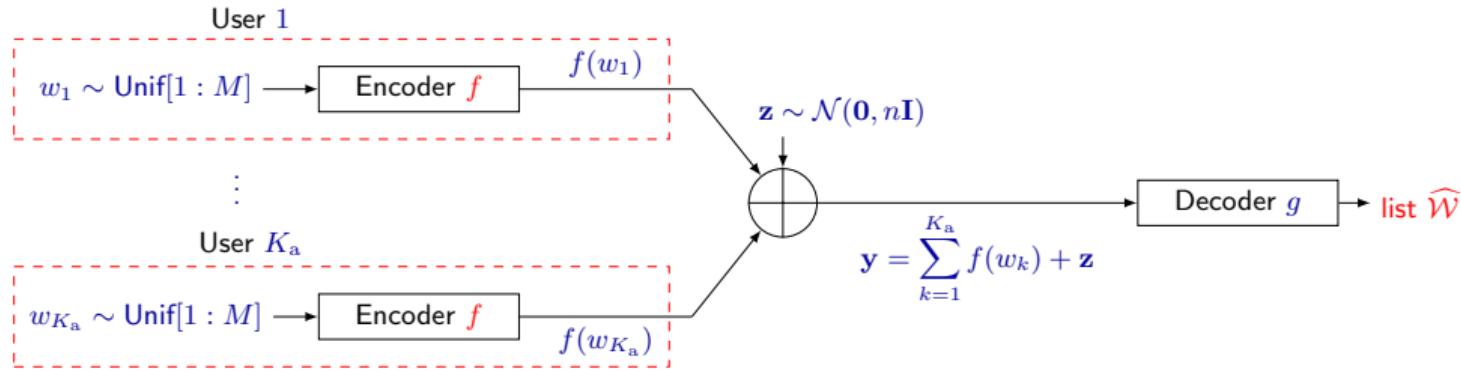
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Addressing these limitations

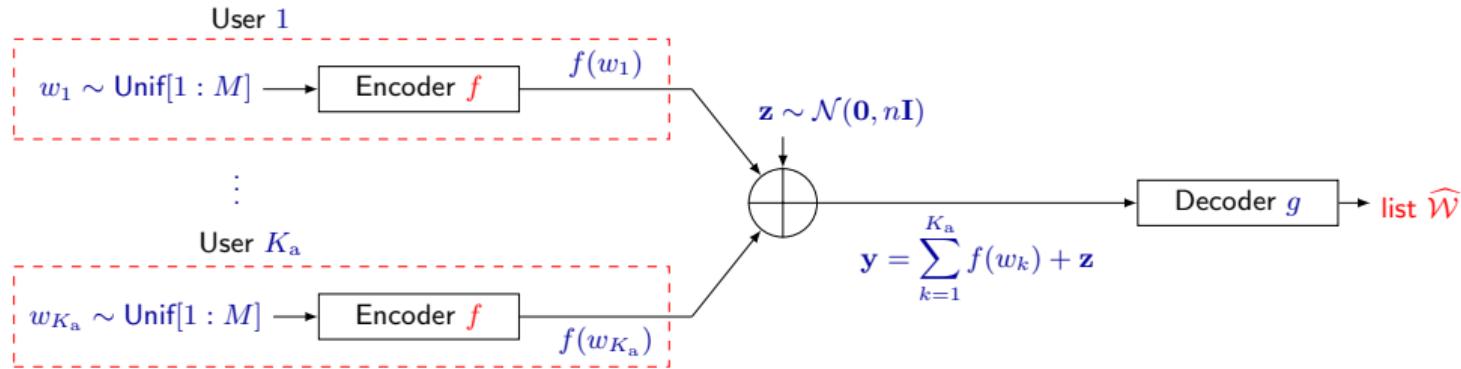
- Noiseless adder channel (e.g., [Bar-David et al., '97])
- More general information-theoretic perspective [Polyanskiy '17]

Unsourced Gaussian MAC model [Polyanskiy '17]



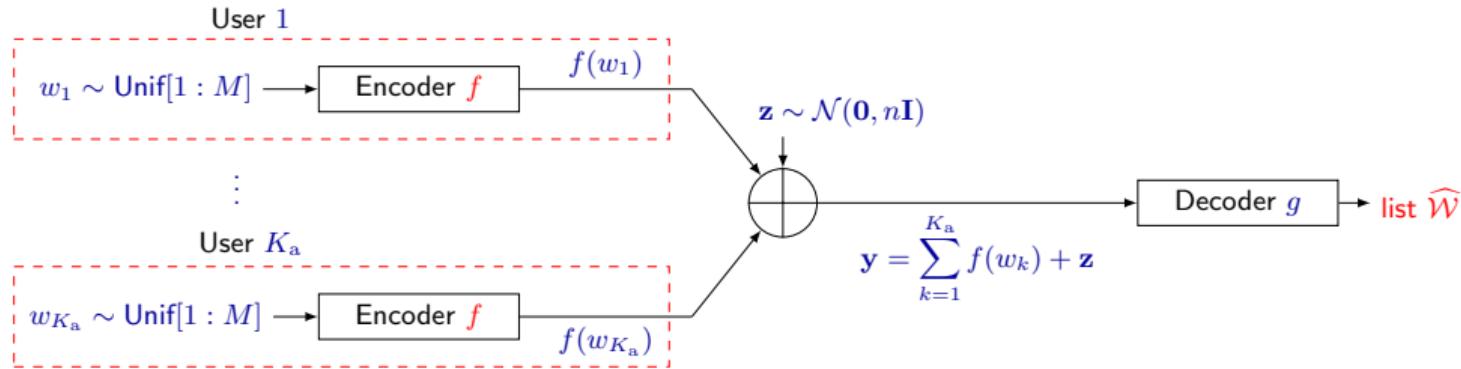
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Unsourced Gaussian MAC model [Polyanskiy '17]



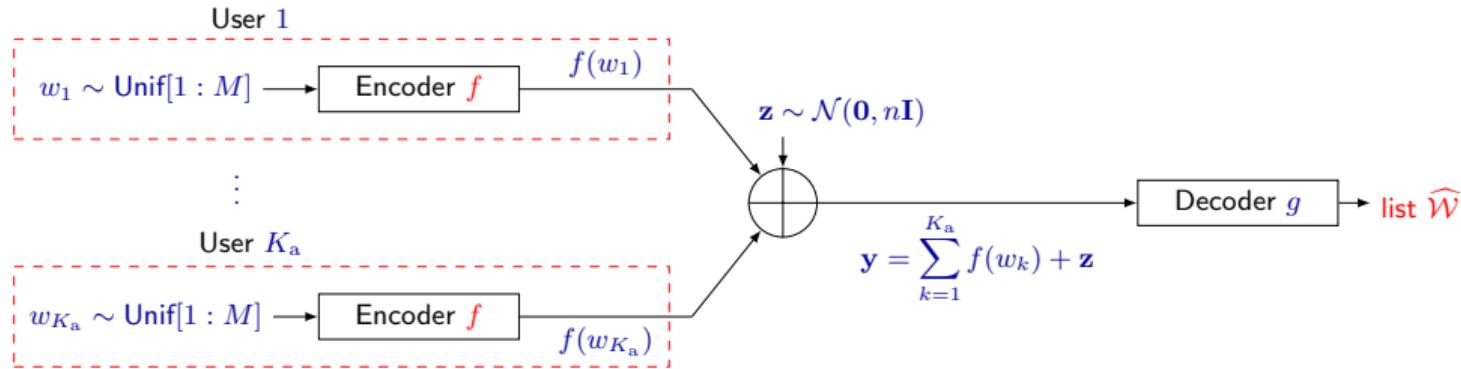
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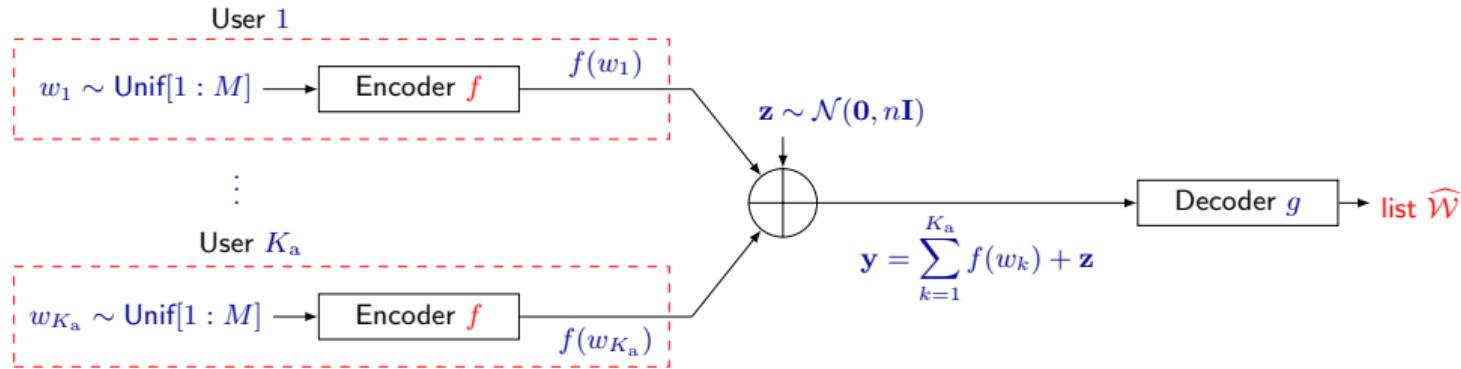
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[Polyanskiy '17]: achievability bound on $E_b/N_0 = \frac{nP}{\log_2 M}$ required to achieve $P_e \leq \epsilon$

Random coding achievability bound for UMA

Random-coding achievability bound (K_a known) [Polyanskiy '17]

For every $P' < P$, there exists an (M, n, ϵ) code for the K_a -user unsourced GMAC with power constraint P satisfying

$$\epsilon \leq \sum_{k=1}^{K_a} \frac{k}{K_a} \min\{p_k, q_k\} + p_0, \quad \text{where}$$

$$p_0 = \frac{\binom{K_a}{2}}{M} + K_a \mathbb{P}\left[\frac{1}{n} \sum_{j=1}^n z_j^2 > \frac{P}{P'}\right]$$

$$p_k = e^{-E(t)}$$

$$E(t) = \max_{0 \leq \rho_1, \rho_2 \leq 1} -\rho_1 \rho_2 k R_1 - \rho_2 R_2 + E_0(\rho_1, \rho_2)$$

$E_0(\rho_1, \rho_2)$: complicated expression in ρ_1, ρ_2, k, P'

$$q_k = \inf_{\gamma} \mathbb{P}[I_k \leq \gamma] + e^{n(kR_1 + R_2) - \gamma}$$

I_k : related to inf. dens.

$$R_1 = \frac{1}{n} \log M - \frac{1}{nk} \log k!$$

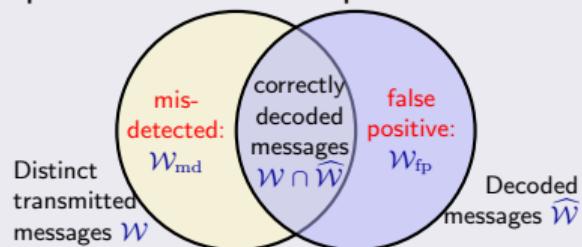
$$R_2 = \frac{1}{n} \log \binom{K_a}{k}$$

Key ideas in the proof

- Gaussian codebook
- Message collisions or power-constraint violations treated as **error**
- **Decoder**: unordered list $\widehat{\mathcal{W}}$ of decoded messages obtained by solving

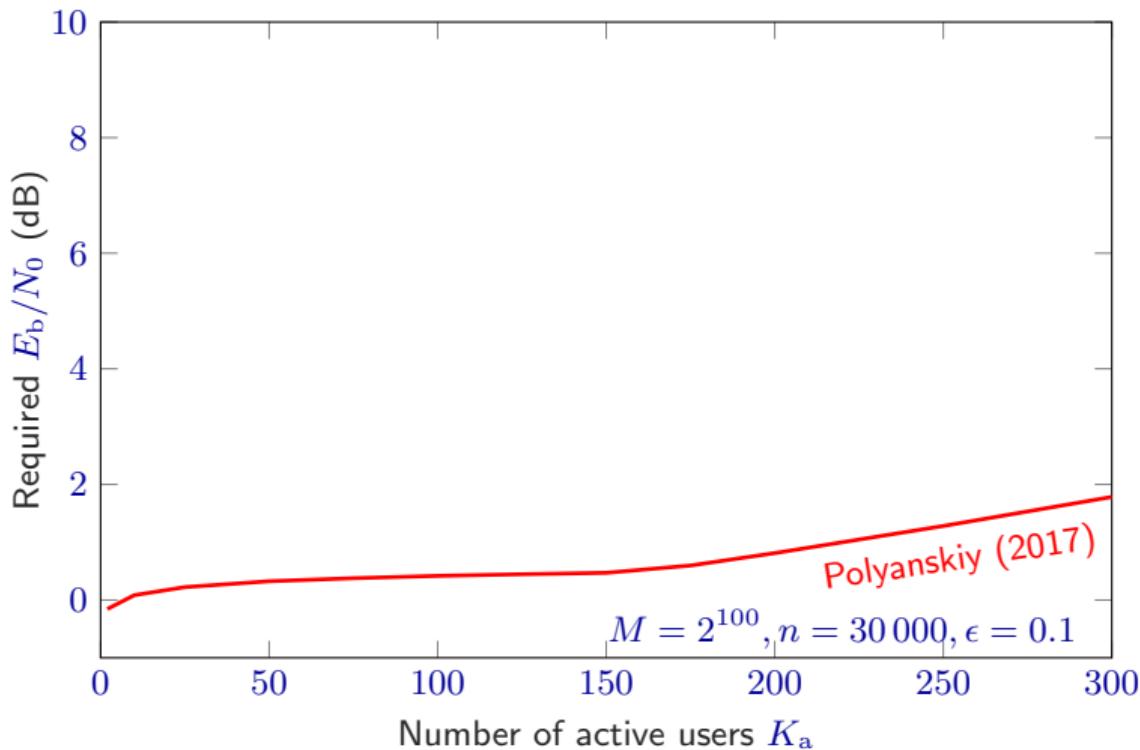
$$\widehat{\mathcal{W}} = \arg \min_{\mathcal{W}' \subset [1:M], |\mathcal{W}'| = K_a} \|\mathbf{y} - \mathbf{c}(\mathcal{W}')\|, \quad \text{with} \quad \mathbf{c}(\mathcal{W}') = \sum_{w \in \mathcal{W}'} \mathbf{c}_w$$

- Space of error events parameterized by number of **misdetected/false-positive** messages

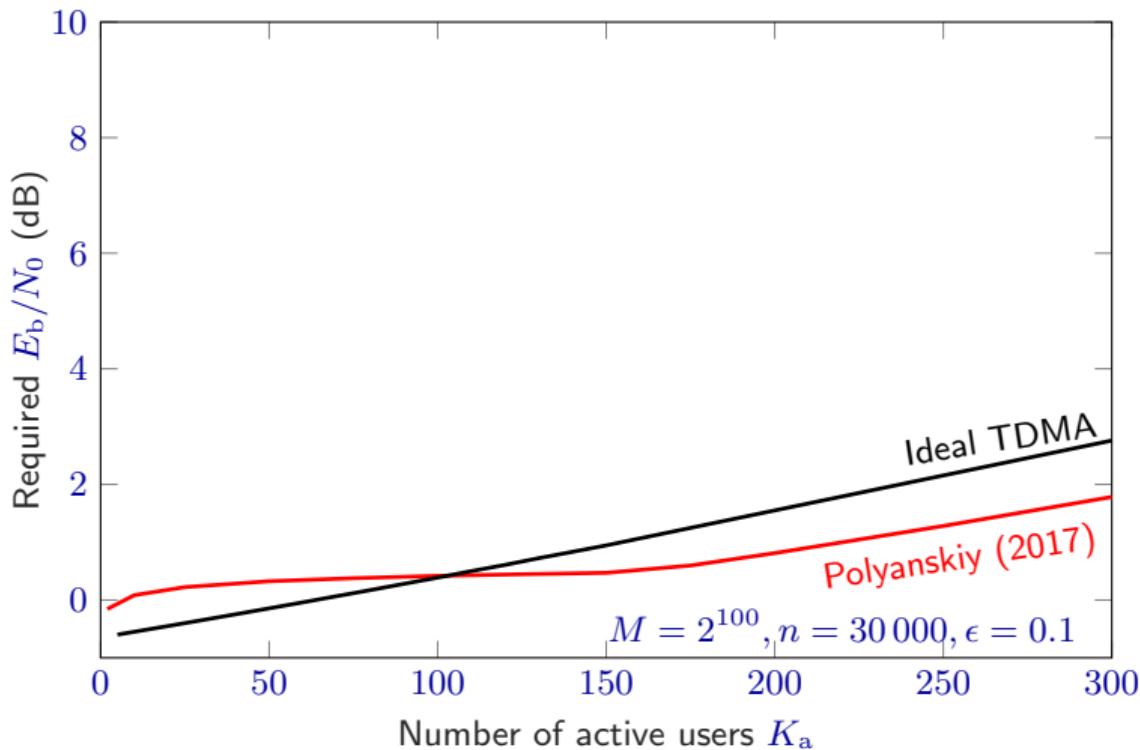


- **Error exponent** analysis

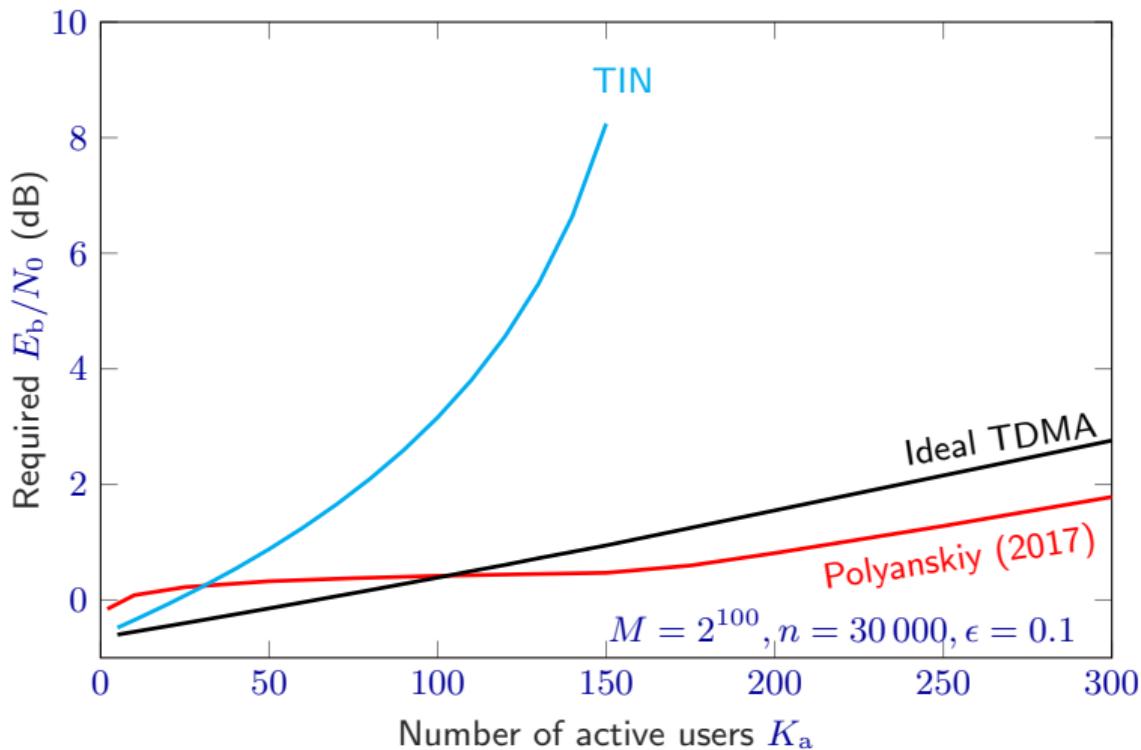
Numerical evaluation of the bound



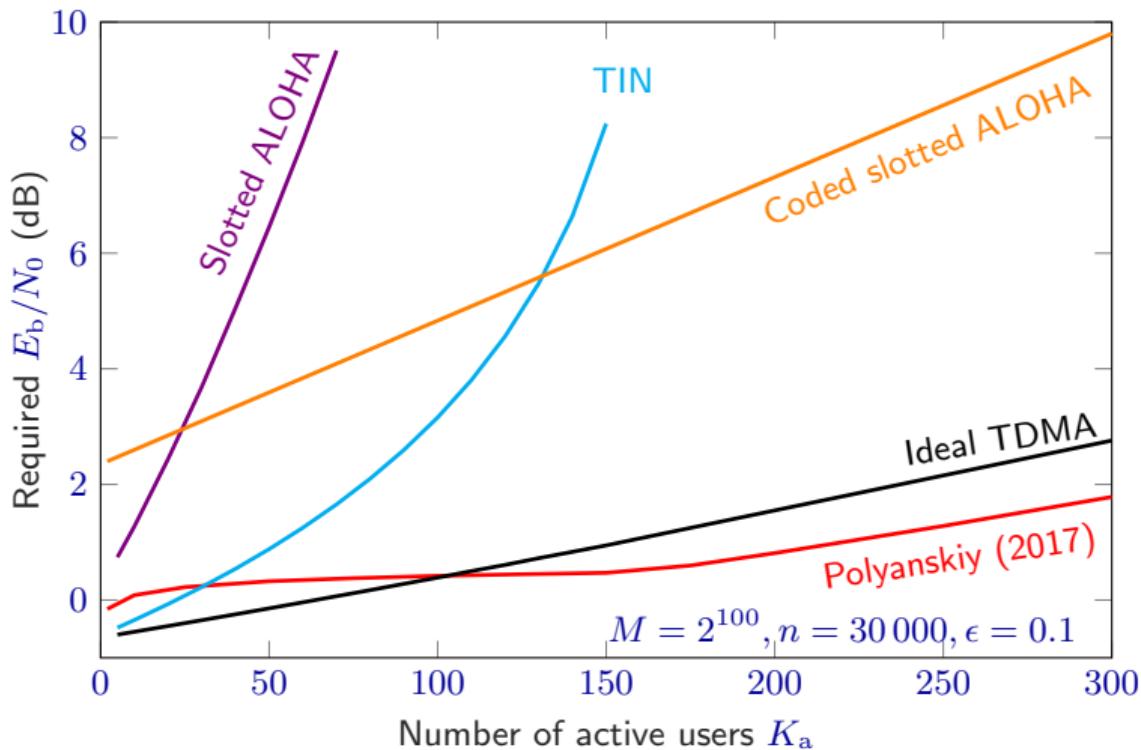
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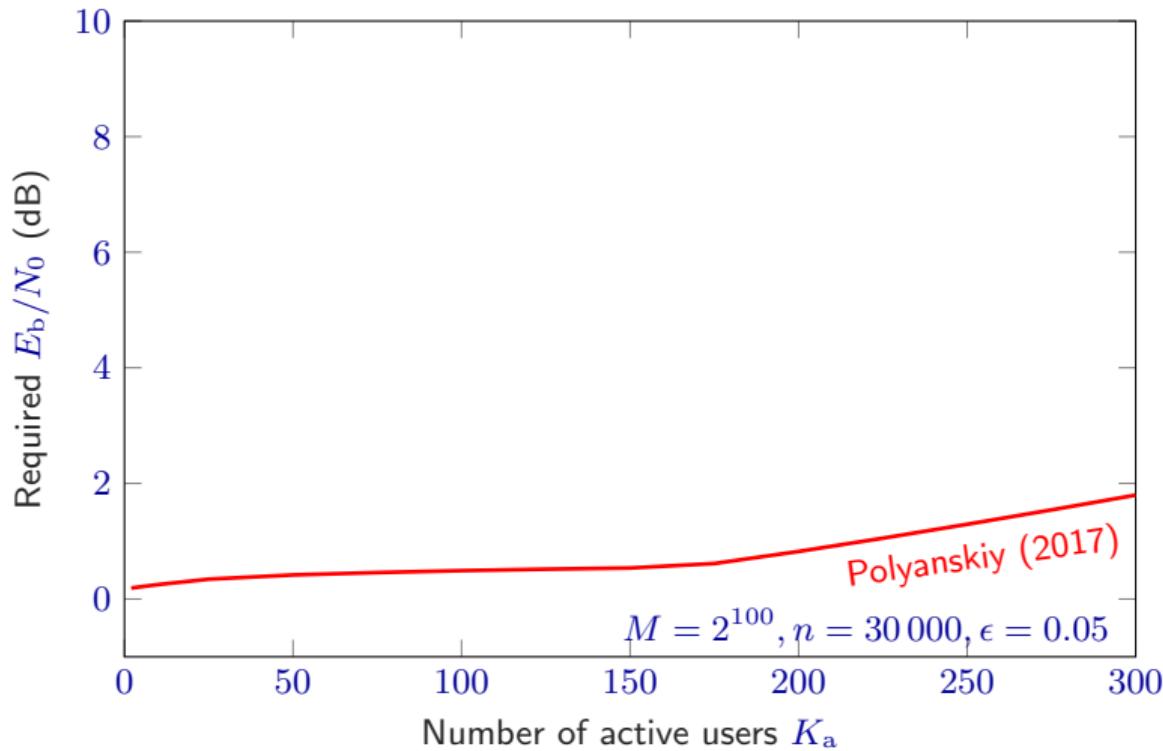
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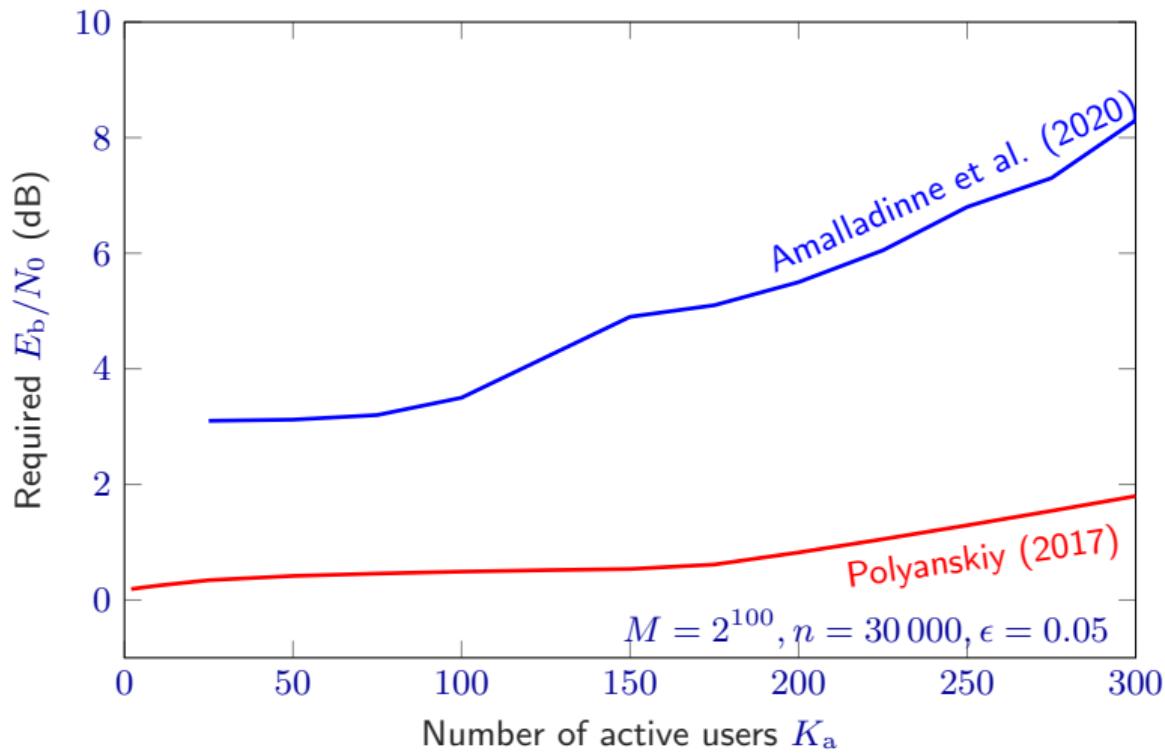
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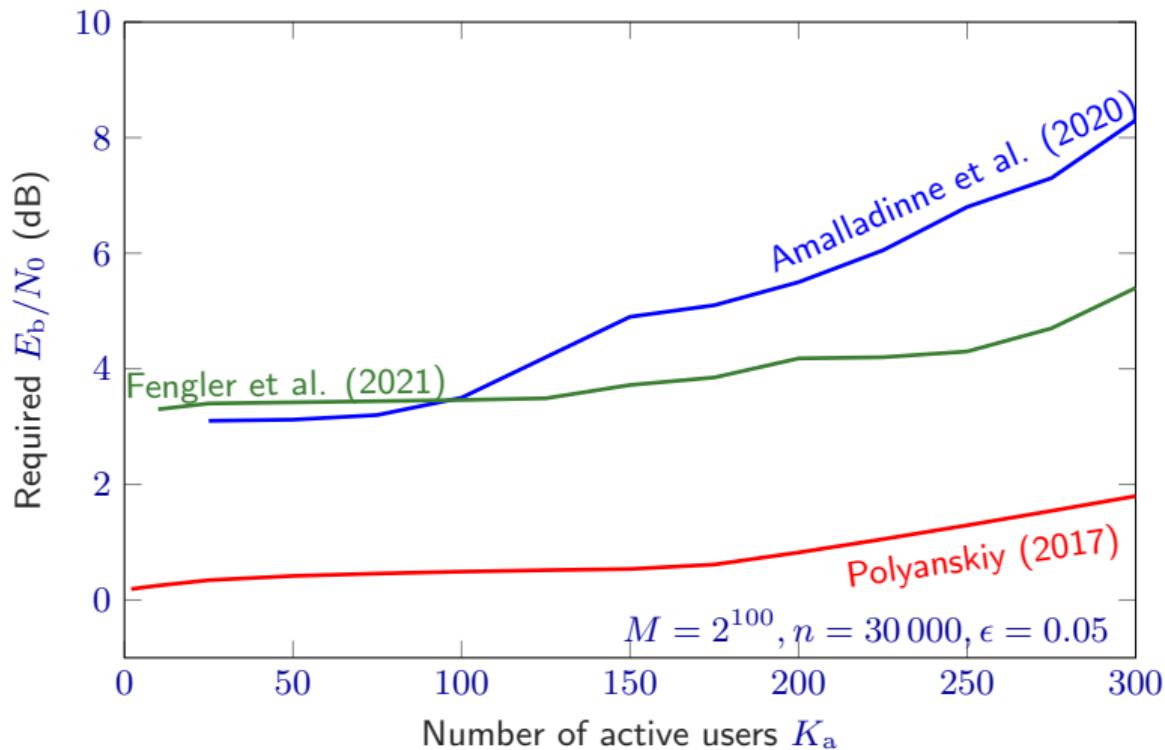
Novel coding schemes (2017–2022)



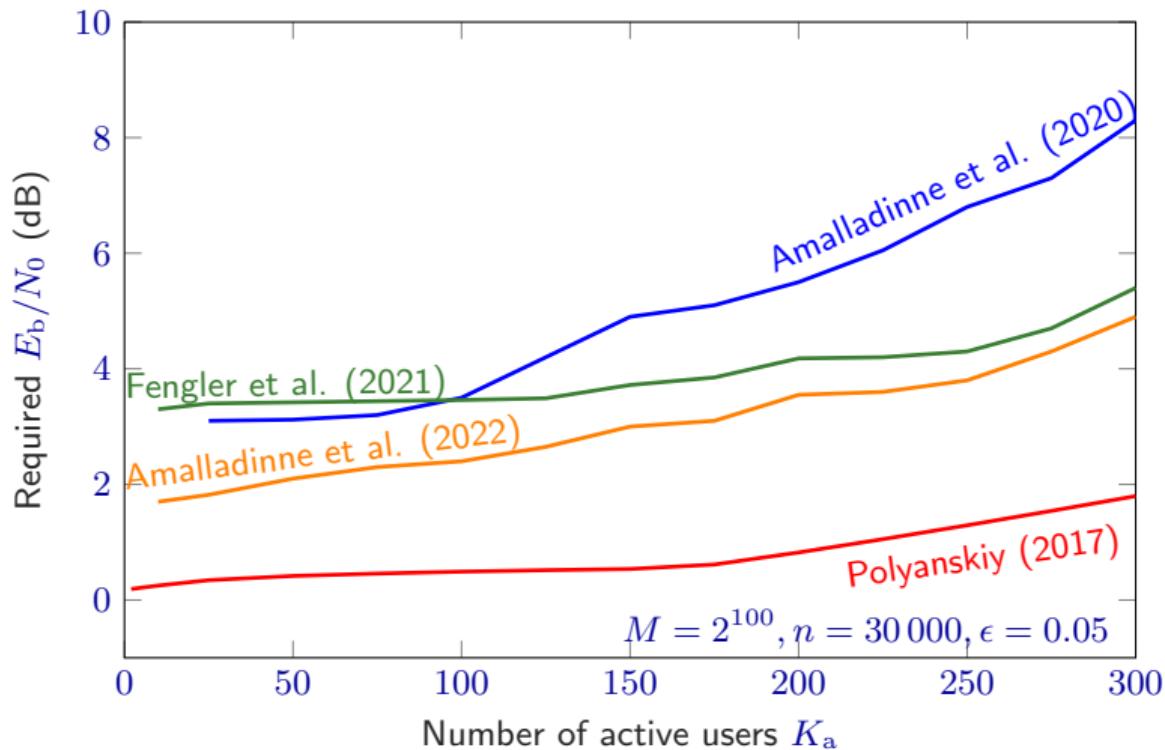
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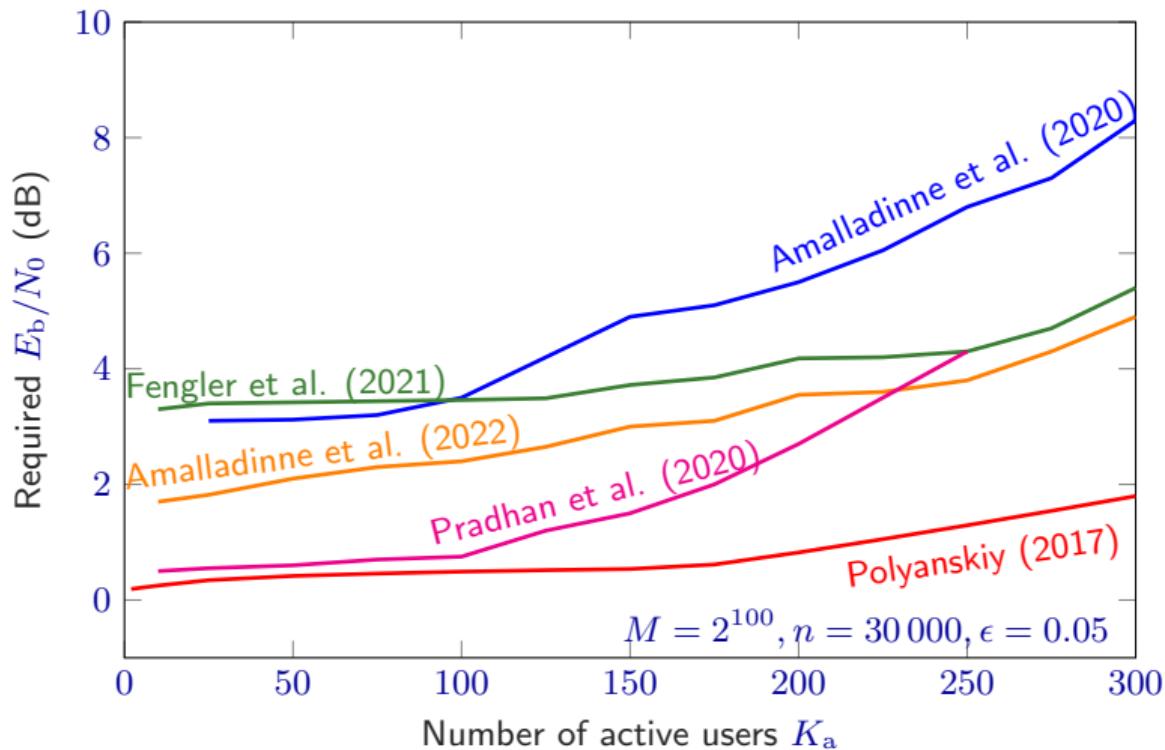
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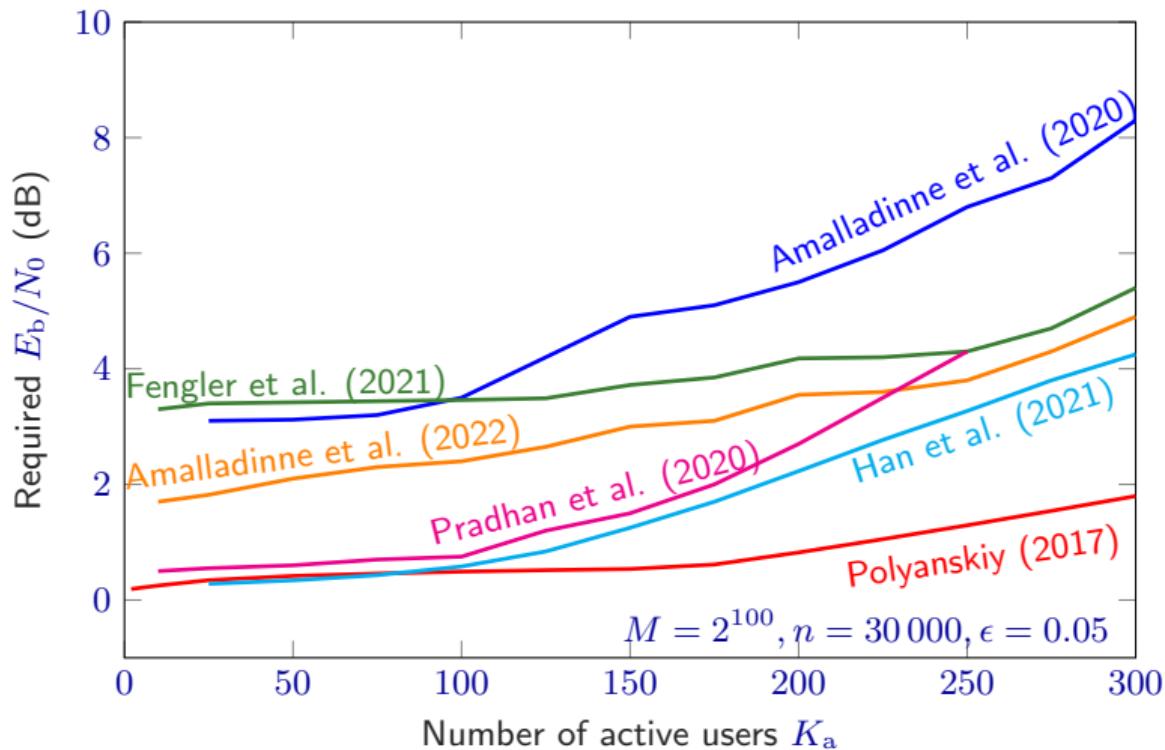
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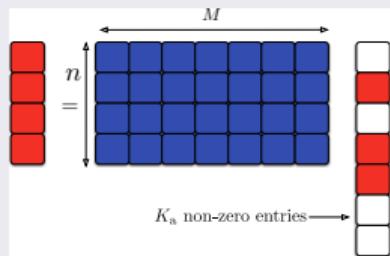


Novel coding schemes (2017–2022)



The coded compressive sensing approach

Message detection is a compressive sensing problem

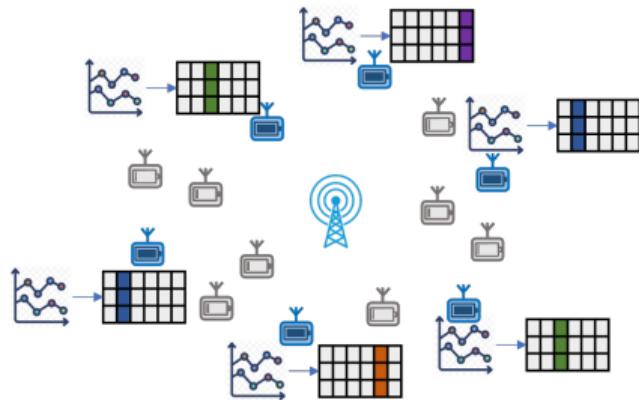


- ✓ We could use compressive sensing solvers such as AMP...
- ✗ ... but the width of the matrix is huge ($M = 2^{100}$)!

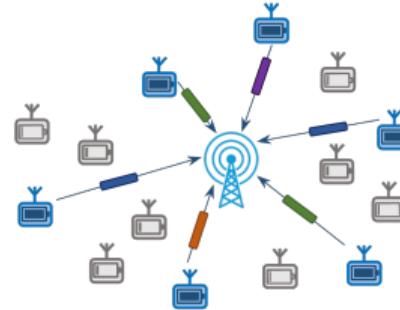
Solution

- 💡 **Fragment** message into smaller sub-blocks (e.g., 16 bits each), transmit one sub-block at a time, and use AMP to decode
- 💡 Add **outer tree code** to stitch together sub-blocks

From UMA to TUMA



Sampling and quantization

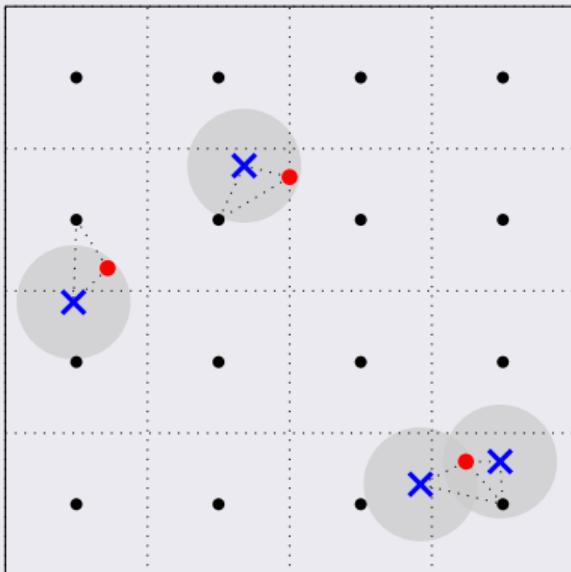


Data transmission

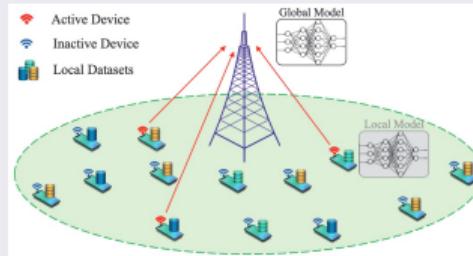
- Users may transmit the same quantized information
- Interested in both the **set** if transmitted messages and their **multiplicities**
- **Goal:** estimate **empirical message distribution (type)**

Two examples

Multi-target position tracking

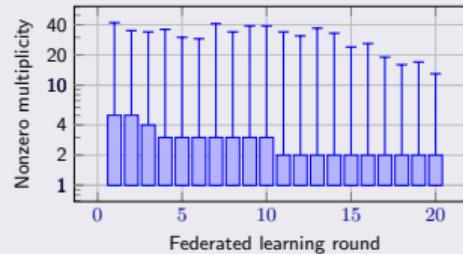


Over-the-air aggregation in federated learning



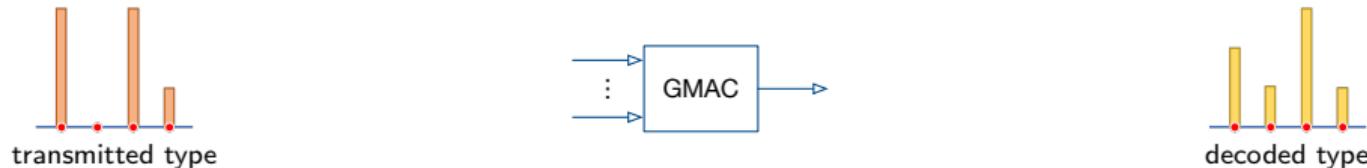
source: Qiao, Gao, Li, and Gündüz (2023)

Message multiplicity: 10th-90th percentile box with full range whiskers



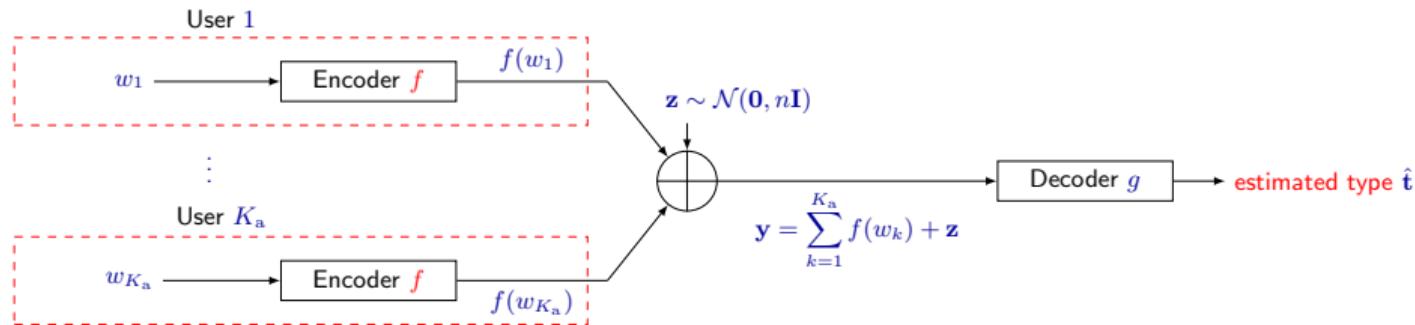
50 clients with 80% random activation

Type-based estimation over multiple access channel



- Considered by [Mergen & Tong, 2006], however...
 - Assumed that each message can be associated to an orthogonal codeword
 - Multiuser interference over a Gaussian MAC can be eliminated via matched filtering
- Our scenario—**type-based unsourced multiple access (TUMA)**
 - Number of possible messages (2^{100}) \gg frame length (30 000): orthogonalization not possible
 - Only few messages are active (still a compressive sensing problem)

TUMA over a Gaussian MAC



Key differences with respect to UMA

- Messages w_1, \dots, w_{K_a} forming type $\mathbf{t} = [t_1, \dots, t_M]$ where $t_m = \frac{1}{K_a} \sum_{k=1}^{K_a} \mathbf{1}\{w_k = m\}$
- Decoder returns a type estimate $\hat{\mathbf{t}}$
- (Communication) performance metric: total variation distance

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

Random coding achievability bound for TUMA [Krishnan et al. 2025]

Random-coding achievability bound (K_a and M_a known)

For every $P' < P$, there exists an (M, n, ϵ) TUMA code satisfying

$$\epsilon \leq \sum_{k=1}^{K_a} \frac{k}{K_a} p_k + p_0 + p_1, \quad \text{where}$$

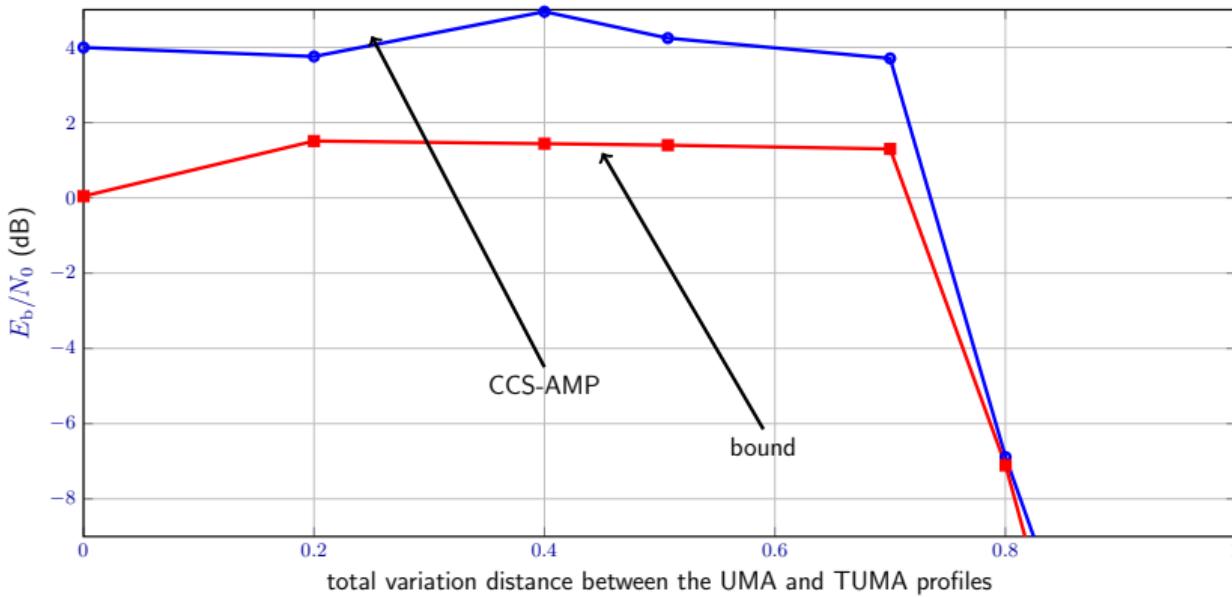
$$p_0 = M_a \mathbb{P} \left[\frac{1}{n} \sum_{j=1}^n z_j^2 > \frac{P}{P'} \right]$$

$\delta = \text{complicated expression}$

$$p_1 = e^{-N\delta^2/8}$$

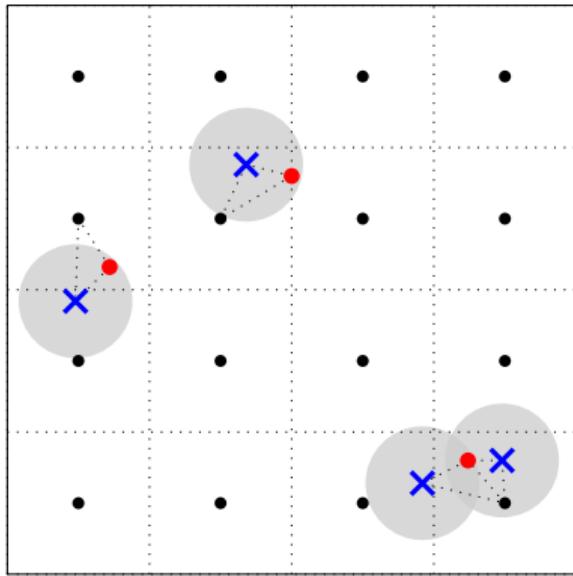
$p_k = \text{even more complicated expression...}$

Example: UMA vs TUMA and CCS performance

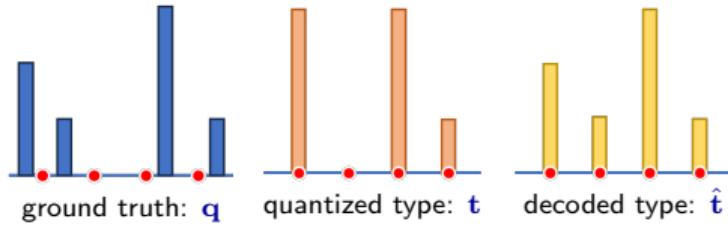


$$n = 38\,400, \epsilon = 0.05, K_a = 100, M_a \in [2, 100]$$

Application example: multi-target position tracking over AWGN channel

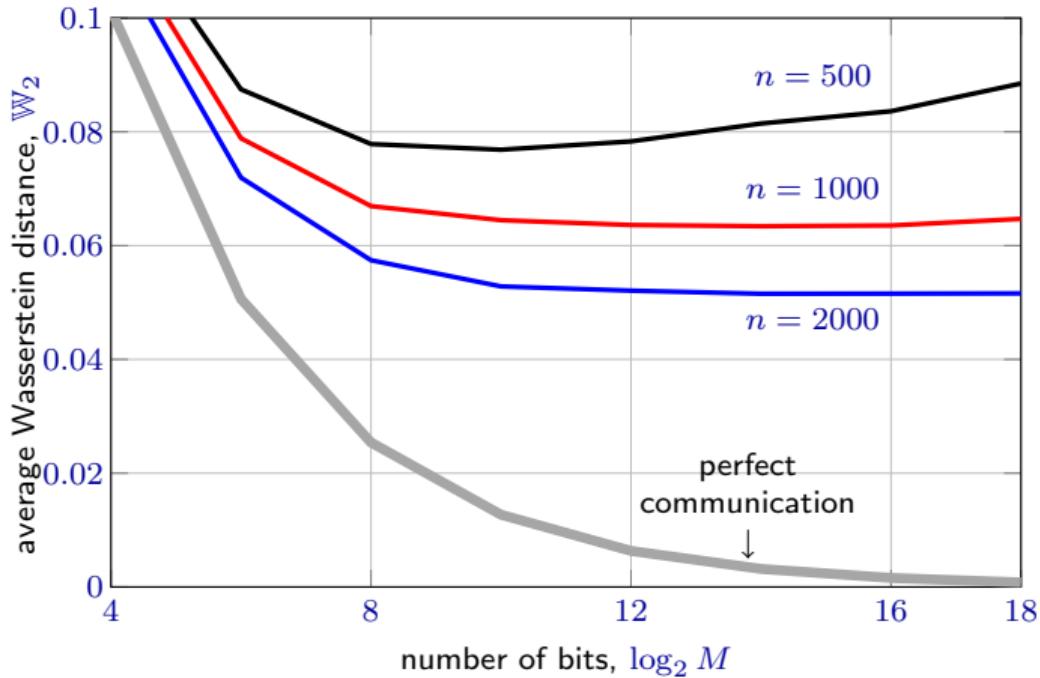


- sensor
- target
- quantiz. position



- $\text{TV}(\mathbf{t}, \hat{\mathbf{t}})$ captures only communication performance
- Overall performance: Wasserstein distance $\mathbb{W}_2(\mathbf{q}, \hat{\mathbf{t}})$

Tradeoff between communication and quantization [Ngo et al., 2024]



$K_a = 100$ sensors, $M_a = 10$ targets; AMP detection algorithm

Beyond Gaussian MAC: TUMA over fading channels

So far Gaussian MAC...

- 💡 Perfect power control: $\mathbf{y} = \sum_{k=1}^{K_a} f(w_k) + \mathbf{z}$
- 💡 Multiplicity can be estimated from power of received codeword
- ✖ Over a fading channel, this would require channel inversion [Qiao et al., 2024]

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The benefits of cell-free architectures in UMA

- [Gkiouzepi et al., 2024]: UMA within cell-free massive MIMO architectures
 - 🏆 Message recovery + channel estimation + estimation of position
 - ✓ Perfect knowledge of large-scale fading coefficients
 - ✓ Location-based codeword partition
 - ✓ Multi-source AMP

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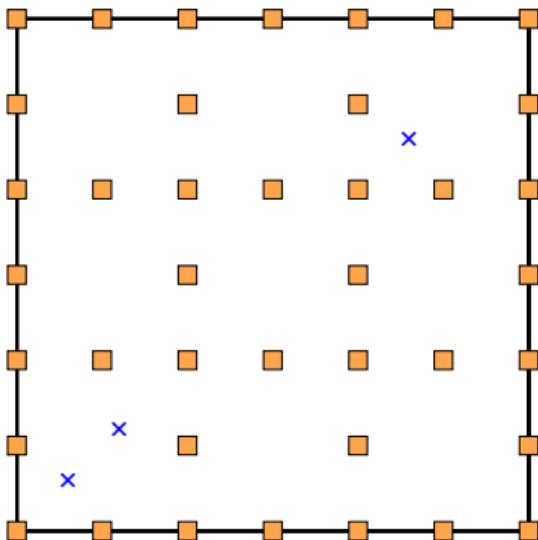
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The same ingredients can be used to perform TUMA over a cell-free massive

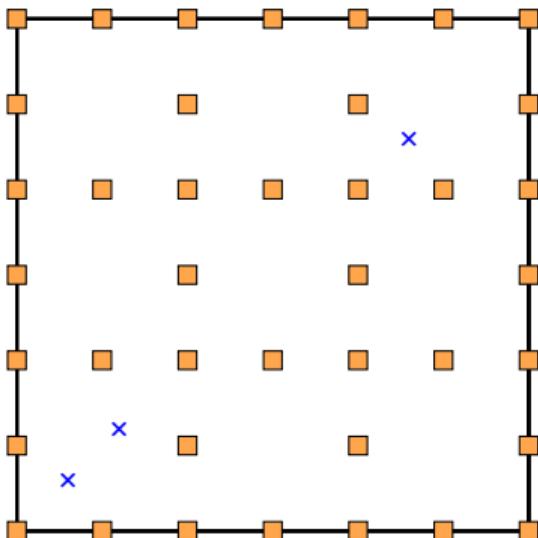
TUMA over a cell-free massive MIMO architecture



Activity detection

- ✓ Perfect knowledge of large scale fading coefficients (LSFC) of each user
- ✓ Each user has its own signature
- ✓ Association between users and LSFC

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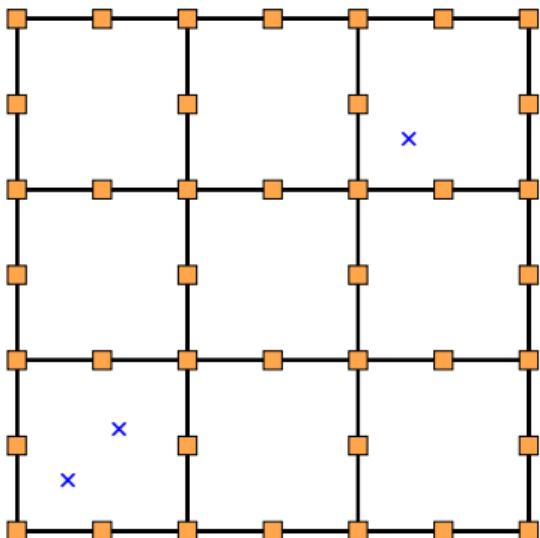
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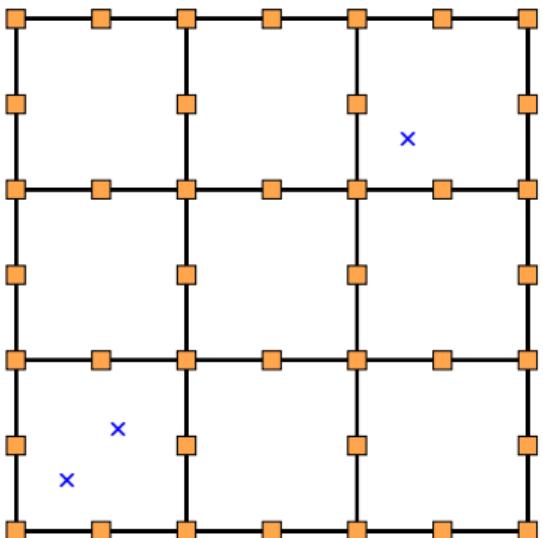
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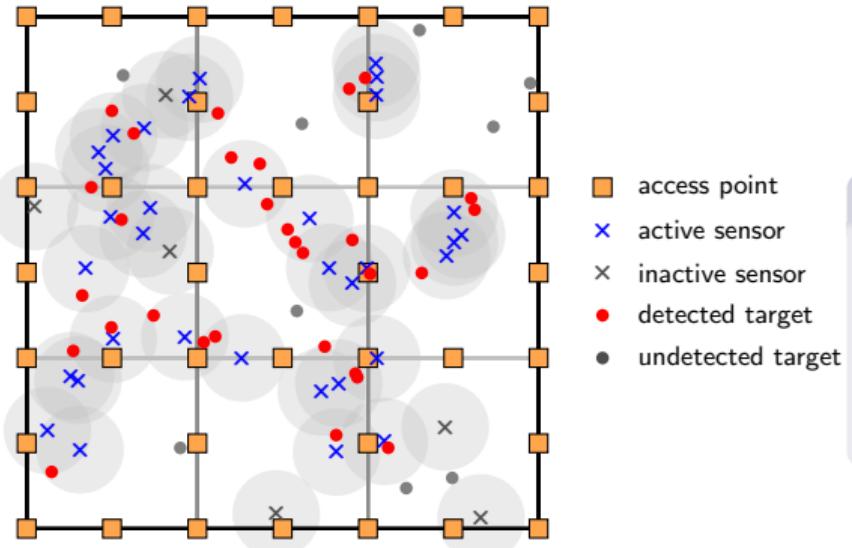
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TUMA

- ✓ Possible to estimate multiplicities

TUMA over a cell-free massive MIMO architecture [Okumus et al., 2025]

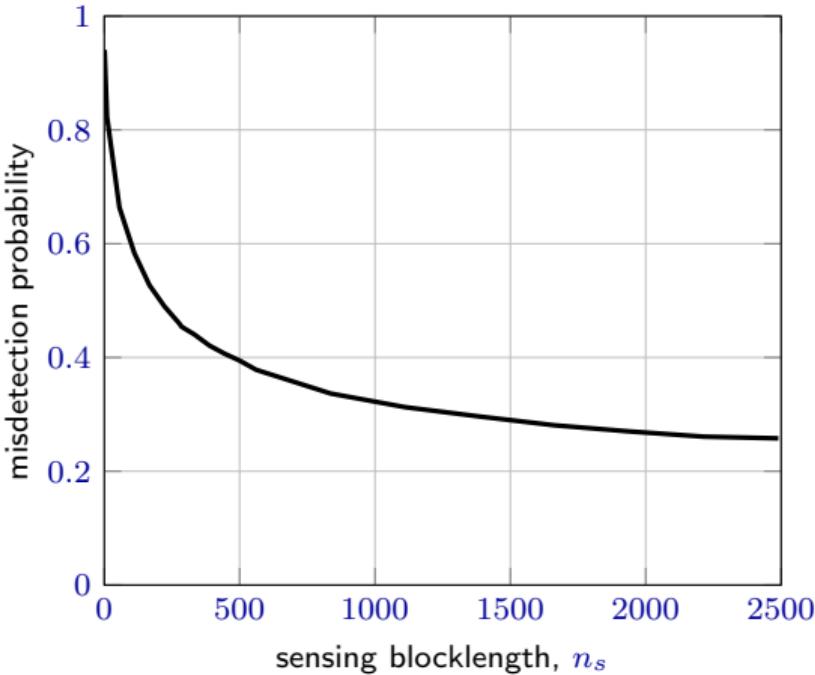
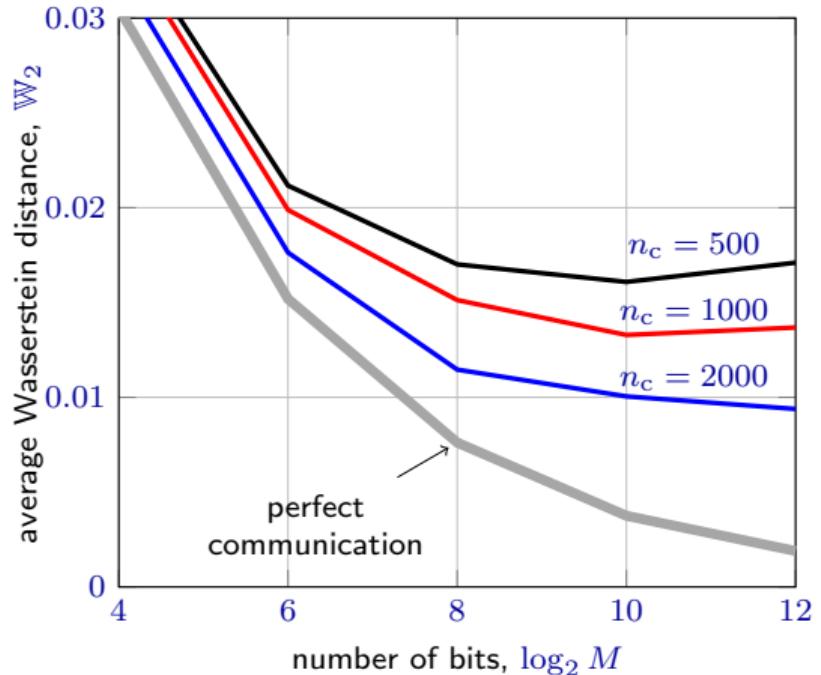


Multi-target tracking example

- 40 access points with 4 antennas each
- 100 sensors, 50 targets
- Sensing phase followed by communication phase: $n = n_s + n_c = 2500$

TUMA over a cell-free massive MIMO architecture [Okumus et al., 2025]

100 sensors, 50 targets, $n = n_s + n_c = 2500$, multi-source AMP detection algorithm



Conclusion

Type-based unsourced multiple access (TUMA)

A framework to collect in an efficient way data from a massive population of sporadically active sensors

Theoretical bounds and practical algorithms

