

Ezgi Ozyilkan

New York University
Tandon School of Engineering



NYU

TANDON SCHOOL
OF ENGINEERING



The iREDEFINE workshop is supported by NSF grant #2408861.

Education

- NYU Tandon School of Engineering, Ph.D. (2021 - May 2026 [expected]). Area: Neural (distributed) lossy data compression.
Advisor: Prof. Elza Erkip
- Imperial College London, M.Eng., First Class Honors (2017 - 2021). Area: Electrical Electronics Engineering. Thesis advisor: Prof. Deniz Gündüz.

Service & Teaching

- Co-organizer of the IEEE ISIT 2025 "Learn to Compress & Compress to Learn" Workshop.
- Co-organizer of the NeurIPS 2024 Machine Learning and Compression Workshop.
- Lead organizer of the IEEE ISIT 2024 "Learn to Compress" Workshop.
- Member of the IEEE IT Society Student and Outreach Subcommittee, 2024 - present.
- Probability and Stochastic Processes, Fall 2024 & 2022, Graduate Teaching Assistant
- Deep Learning, Spring 2022, (Head) Graduate Teaching Assistant.

Selected Awards

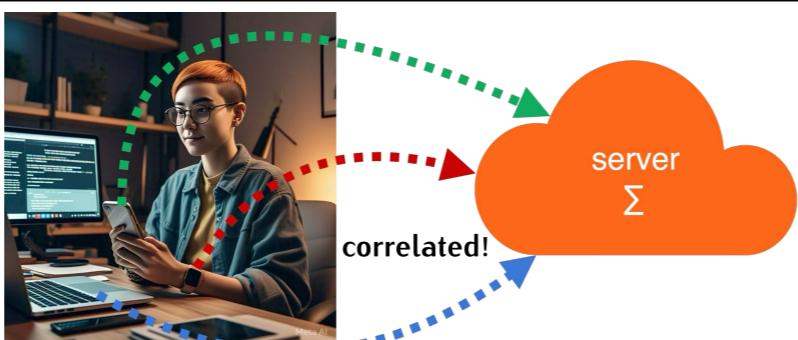
- IEEE Signal Processing Society Scholarship (2024–2026).
- Best Reviewer Award, Neural Compression Workshop @ ICML 2023.
- 2021 Ivor Tupper Prize in Signal Processing, Imperial College London.

Industry Experience

- (Incoming) ML/CV Intern at Apple
- Research Intern at InterDigital Video Lab
- Research Intern at InterDigital AI Lab

General Research Area

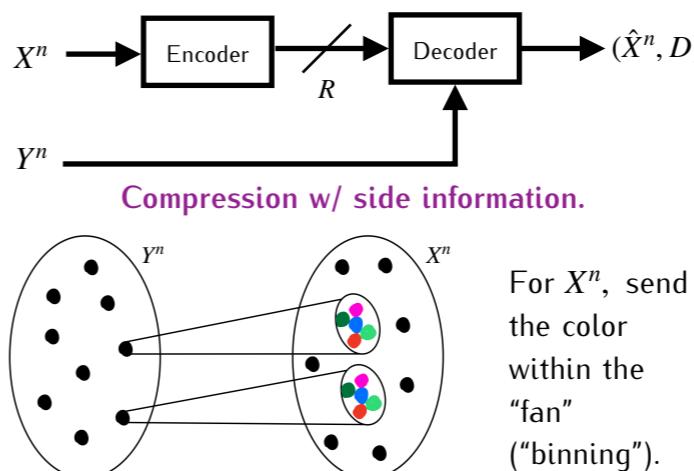
- **Research interests** Neural (data) compression, quantization, information theory, machine learning, signal processing, explainable AI.
- **PhD topic** Bridging Theory and Practice: Advancing Distributed Data Compression and Communication via Machine Learning
- Traditional compression removes redundancy within one source, while "distributed" compression leverages correlations **across multiple data sources** encoded independently but collaboratively.



Example of a distributed compression and inference scenario, e.g., federated learning across a single user's personal devices.

- **Challenge** Exploiting correlation across users/devices in distributed networks (IoT, cameras, sensors...) for practical distributed compression & inference.
- **Our solution** Translating insights from information theory into real-world applications via machine learning.
- **Impact** Edge computing/inference, federated learning, multi-view image/video compression. Foundations for fully distributed case.
- **Ongoing** Diverse correlation structures and multi-modal data within users/devices; fully distributed, scalable and robust algorithms.

Research Contributions



Contribution I: Leveraging Side Information Efficiently

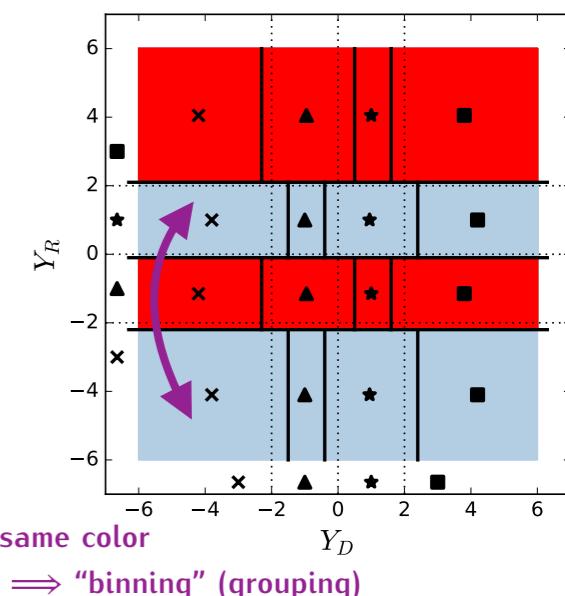
Approach Neural networks learn theoretically optimal "binning" (grouping), outperforming state-of-the-art neural compressors.

Result Near-optimal and interpretable performance that aligns with theory.

Supported by Google Research Collabs Program.

Contribution II: Cooperative Communication & Relaying

Approach First proof-of-concept for learning-based "compress-and-forward" relaying, building onto our prior work on distributed compression.



Selected Publications

- E. Ozyilkan*, F. Carpi*, S. Garg, and E. Erkip, "Learning-based compress-and-forward schemes for the relay channel," *IEEE Journal on Selected Areas in Communications*, 2025.
- E. Ozyilkan, J. Ballé, and E. Erkip, "Neural distributed compressor discovers binning," *IEEE Journal on Selected Areas in Information Theory*, 2024.
- E. Ozyilkan, J. Ballé, S. Bhadane, A. B. Wagner, and E. Erkip, "Breaking smoothness: The struggles of neural compressors with discontinuous mappings," *Wksp. on Machine Learning and Compression @ NeurIPS 2024*.
- E. Ozyilkan, J. Ballé, and E. Erkip, "Learned Wyner-Ziv compressors recover binning," *IEEE Int. Symp. on Information Theory*, 2023.

