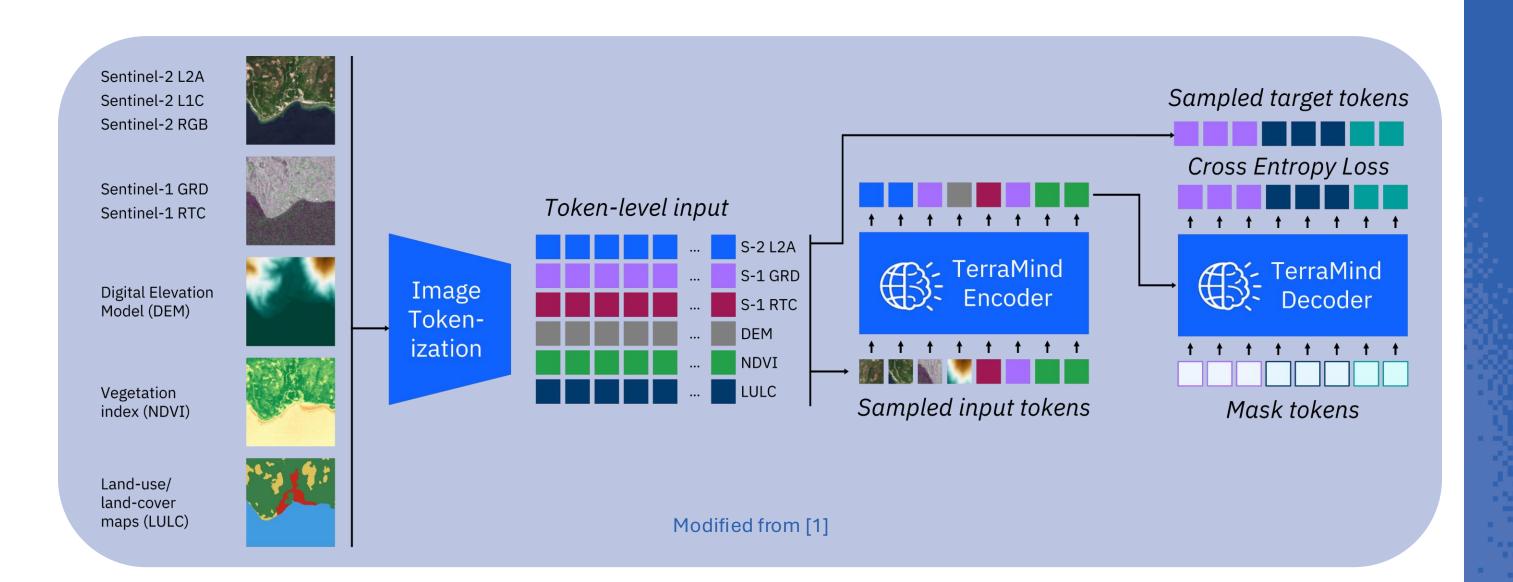


REVOLUTIONISING GEOSPATIAL DATA WITH AI COMPRESSION

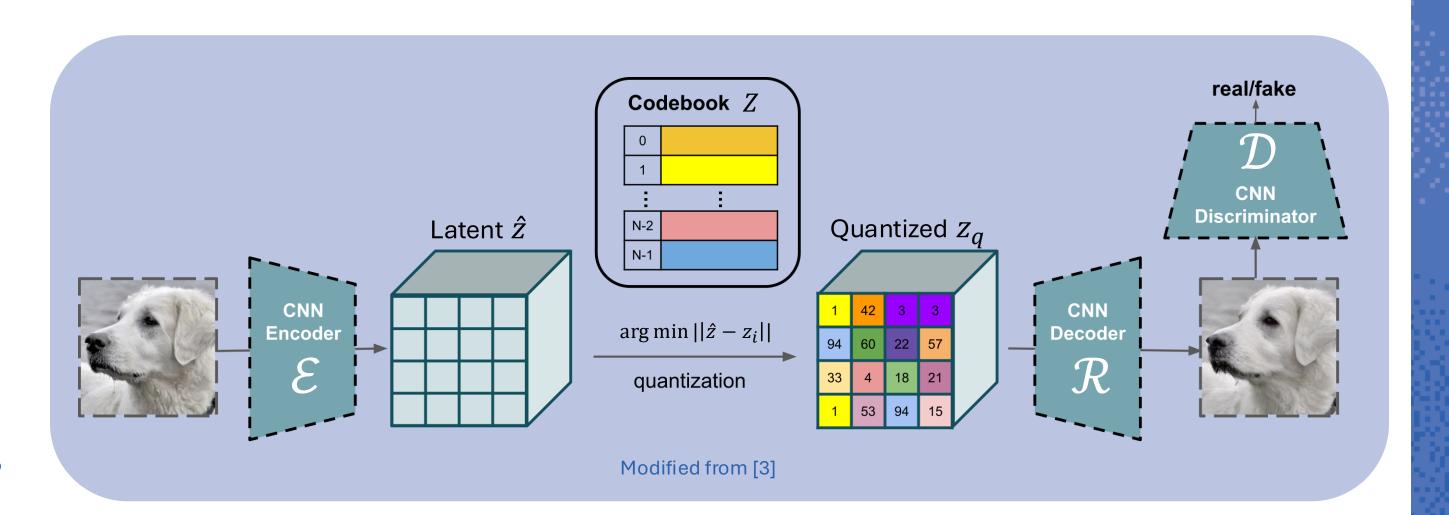
Motivation

- Continuous EO acquisitions generate petabyte-scale archives.
- Larger deep learning models demand more training data and I/O
- Vector-quantization (VQ) compression stores only codebook indices, reducing size by > 450x
- Transformer backbones like TerraMind [1] already rely on a fixed VQ codebook
- Experments use MajorTOM-Core dataset [2]

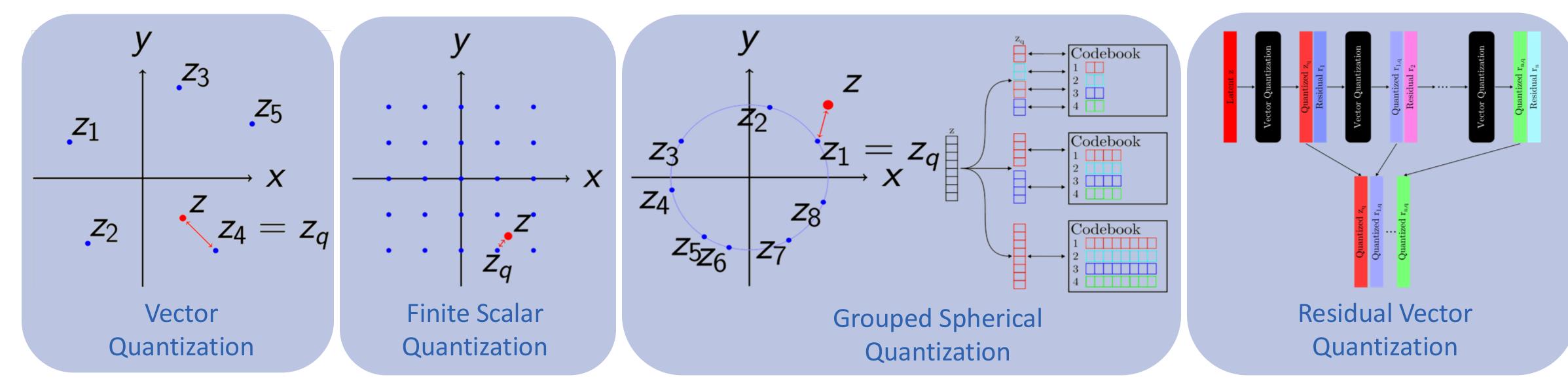


Vector Quantization

- Encoder (VQ-GAN [3]) produces latent representation
- Quantizer assigns each latent vector to its nearest codebook entry
- Decoder reconstructs the image from the quantized embeddings
- An optional discriminator influences training dynamics



Quantization Methods



Bibliography

[1] Jakubik, J., Yang, F., Blumenstiel, B., Scheurer, E., Sedona, R., Maurogiovanni, S., Bosmans, J., Dionelis, N., Marsocci, V., Kopp, N., Ramachandran, R., Fraccaro, P., Brunschwiler, T., Cavallaro, G., Bernabé-Moreno, J., & Long'ep'e, N. (2025). TerraMind: Large-Scale Generative Multimodality for Earth Observation.

[2] Blumenstiel, B., Fraccaro, P., Marsocci, V., Jakubik, J., Maurogiovanni, S., Czerkawski, M., Sedona, R., Cavallaro, G., Brunschwiler, T., ErraMesh: A Planetary Mosaic of Multimodal Earth Observation Data.

[3] Esser, P., Rombach, R., & Ommer, B. (2021). Taming transformers for high-resolution image synthesis. Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern

[4] Wang, J., Qin, Z., Zhang, Y., Hu, V. T., Ommer, B., Briq, R., & Kesselheim, S. (2024). Scaling Image Tokenizers with Grouped Spherical Quantization. arXiv preprint arXiv:2412.02632.

[5] Bosmans, J., Dionelis, N., Maurogiovanni, S., Czerkawski, M., Sedona, R., Cavallaro, G., Brunschwiler, T., Bernabé-Moreno, J., & Long'ep'e, N. (2025).

[6] Blumenstiel, B., Fraccaro, P., Marsocci, V., Jakubik, J., Maurogiovanni, S., Czerkawski, M., Sedona, R., Cavallaro, G., Brunschwiler, T., Bernabé-Moreno, J., & Long'ep'e, N. (2025).

[7] Blumenstiel, B., Fraccaro, P., Marsocci, V., Jakubik, J., Maurogiovanni, S., Czerkawski, M., Sedona, R., Cavallaro, G., Brunschwiler, T., Bernabé-Moreno, J., & Long'ep'e, N. (2025).

[8] Blumenstiel, B., Fraccaro, P., Marsocci, V., Jakubik, J., Maurogiovanni, S., Czerkawski, M., Sedona, R., Cavallaro, G., Brunschwiler, T., Bernabé-Moreno, J., & Long'ep'e, N. (2025).

[9] Blumenstiel, B., Fraccaro, P., Marsocci, V., Jakubik, J., Maurogiovanni, S., Czerkawski, M., Sedona, R., Cavallaro, G., Brunschwiler, T., Bernabé-Moreno, J., & Long'ep'e, N. (2025).

[9] Blumenstiel, B., Fraccaro, P., Marsocci, V., Jakubik, J., Maurogiovanni, S., Czerkawski, M., Sedona, R., Cavallaro, G., Brunschwiler, T., Bernabé-Moreno, J., & Long'ep'e, N. (2025).

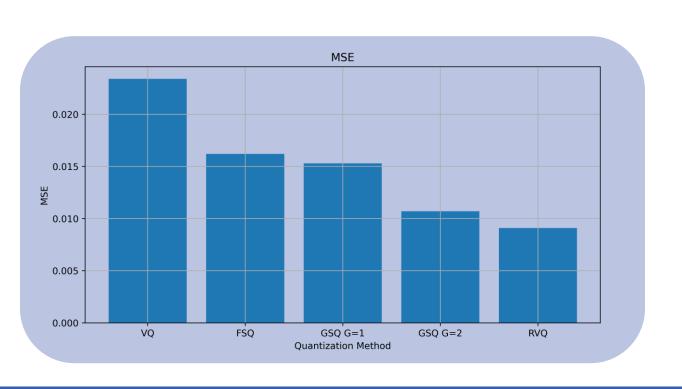
[9] Blumenstiel, B., Fraccaro, P., Marsocci, V., Jakubik, J.,

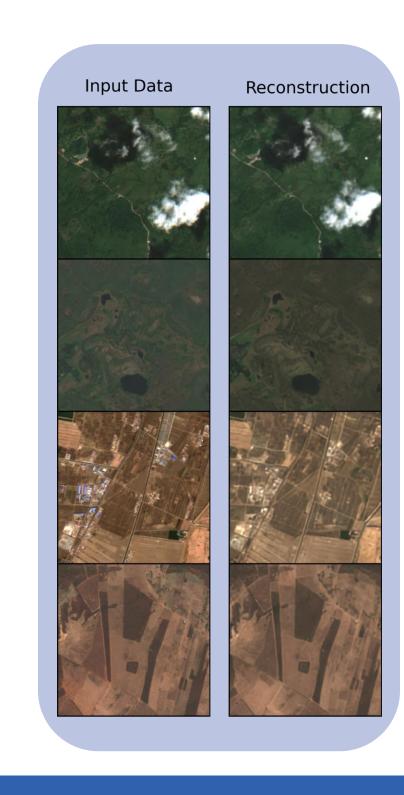
Scalable Efficient Compression in Large-Scale Earth Observation

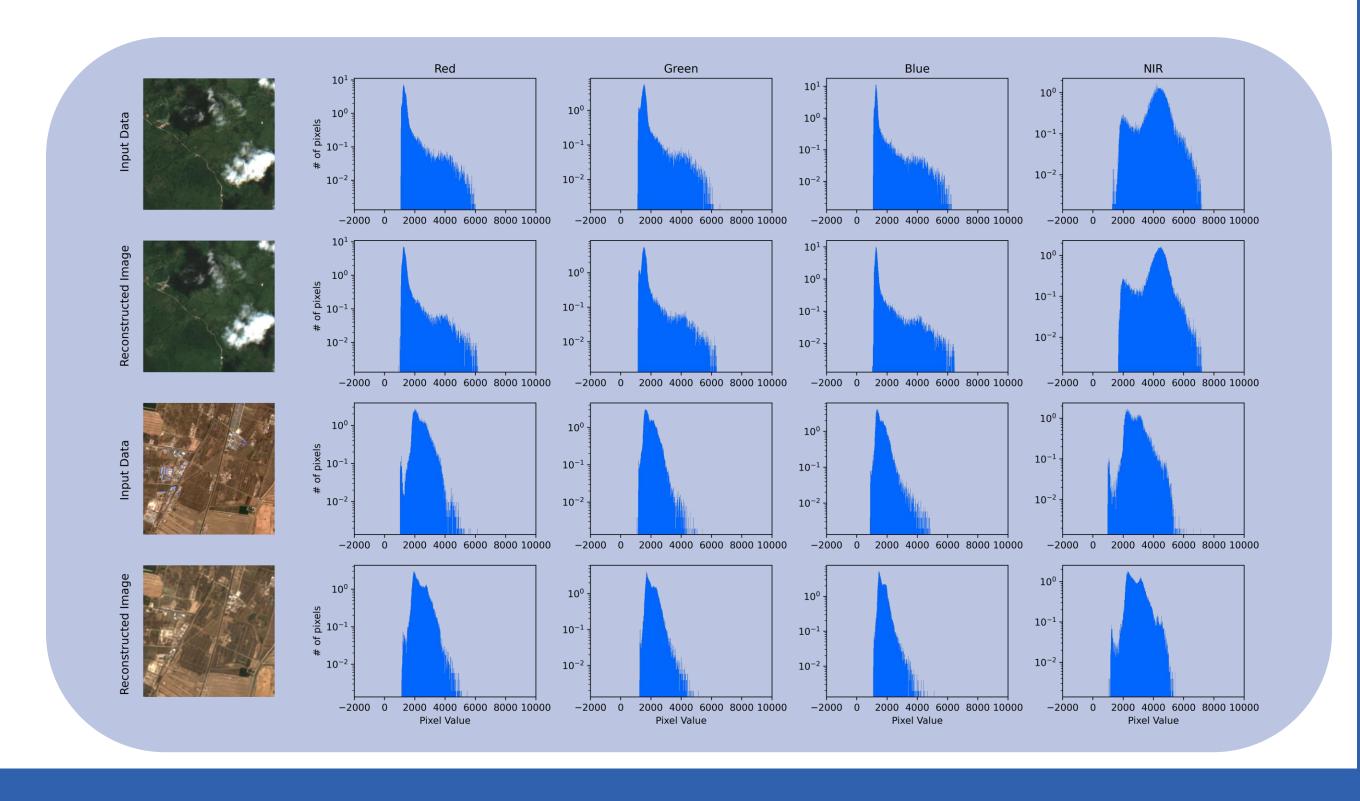
Erik Scheurer, Jiangtao Wang, Rocco Sedona, Stefano Maurogiovanni, Benedikt Blumenstiel, Johannes Jakubik, Paolo Fraccaro, Thomas Brunschwiler, Stefan Kesselheim, Gabriele Cavallaro

Reconstruction Accuracy

- Reconstructions look visually accurate
- Details missing due to downsampling
- Spectral distribution matches
- Well suited for image-level tasks

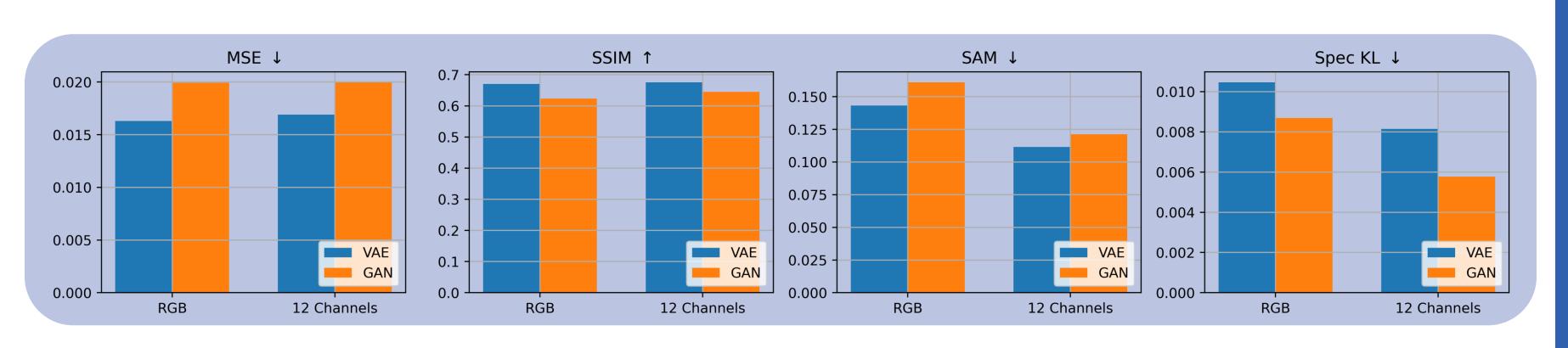






Using a Discriminator

- For non-EO data: Discriminator found to improve visual fidelity of images [4]
- For EO data: beyond visual aspect, spectra must match, and each pixel should be as close to original as possible



VQ-VAE for Compression

- Additional correlations in multispectral data aids compression
- Larger codebook sizes improve reconstruction fidelity
- Larger embedding dimension underutilized and harder to optimize codebook
- Decomposed VQ helps utilize large latent space [4]

