

Raw Data Compression for Synthetic Aperture Radar using Deep Learning

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Overview



- Synthetic Aperture Radar (SAR)
- Lossy data compression
- Proposed methodology
- Experiments
- Conclusions







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Synthetic Aperture Radar (SAR)



- Transmitter/receiver emits pulses (chirps) and records echoes.
- Each received pulse is digitised to an azimuth line, consisting of range samples.
- Travel-times taken by electromagnetic waves to propagate are utilized in focusing the raw data.
- Range-Doppler Algorithm (RDA) is widely used.



Synthetic Aperture Radar (SAR) – Image focusing





Contains modified Copernicus Sentinel data [2022]

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- Usually, components are optimized independently.
- If one component is upgraded, potentially, re-optimization needed.
- Quality of higher level products not always considered.



Lossy data compression - SAR



- Block Adaptive Quantization (BAQ)-based algorithms perform quantization on blocks of samples.
- Quantizers and consequently bit-rate varies depending on block statistics.
- Entropy coding following quantization.
- No data transformation is performed.

Lossy data compression - SAR



- SAR data compression should attain:
 - High compression ratio on raw data,
 - Retain high quality in the SLC image.



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- Lossy SAR raw data compression using deep learning.
- Vector-Quantized Variational Auto-Encoder (VQVAE) [1].
- Learned analysis/synthesis transforms, vector quantization.
- Two strategies followed: one compresses raw data to raw data directly. Second compresses raw data to Single Look Complex (SLC) images.

[1] A. Oord, K. Kavukcuoglu, and O. Vinyals. Neural discrete representation learning, Advances on Neural Information Processing Systems (NIPS), 2017.



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Decoder:

• Same as encoder in reverse order



- 16 channels each (except last layer with 128)
- Codebook size of 512 codewords.
- Normalization and ReLU

VQVAE-RDA (raw data to SLC image)





Encoder:

- 5 convolutional layers,
- 8 channels each (except last layer with 128)
- Codebook size of 512 codewords.
- Normalization and ReLU

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- Mean squared error between original SLC images and estimated SLC images.
- Intermediate data before RDA layer might not be directly usable.
- Differentiable RDA layer implemented to enable gradients to pass through during backwards pass.





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- Number of layers,
- Number of channels,
- Number of striding,
- Codebook size,
- Normalization and activation functions.

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- Two-way travel time between platform and targets.
- 2. Phase shift on the chirp depending on distance travelled.
- Antenna gain depending on measurement angle
- 4. Circular complex Gaussian noise.

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- **Centre frequency:** 5.405 GHz
- Chirp bandwidth: 50 MHz
- Chirp pulse duration: 3 µs
- Range sampling frequency: 150 MHz
- Pulse Repetition Frequency: 1871
- Antenna length: 12 metres
- Platform height: 690 km

Note: These do not correspond to any actual Sentinel-1 acquisition mode, chosen for simplicity and ease of experiments.

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2500





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2500

Training data from simulations



- 300 raw data volumes (1871, 2500, 2), 300 SLC images,
- 250 training data, 50 test data.
- Raw data normalized between -1 and 1, by dividing with maximum absolute value.





- Bits/sample = (first dimension of indices*second dimension of indices*9)/(azimuth lines*range samples*2),
- where 9 bits per index are used (codebook size = 512), 2 is
 for real and imaginary parts of raw data.
- Indices dimensions (935, 1249)
- Bit-rate = 1.12 bits/sample.

Examples







SLC image formed from VQVAE-recovered data



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Examples



Original SLC image



SLC image formed from VQVAE-RDA



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- Image bias issues.
- Data normalization during pre-processing.
- Normalization within each network layer.
- We continue with Impulse Response Function (IRF) analysis.



Impulse Response Function (IRF) metrics



- Spatial resolution: -3dB IRF widths in both azimuth and range.
- Peak side-lobe ratio (PSLR): ratio of
 intensity of the largest peak outside the
 main-lobe to main-lobe [in dB].



Impulse Response Function (IRF) metrics





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Impulse Response Function (IRF) in azimuth





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Range Impulse Response Function (IRF)





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Azimuth Impulse Response Function (IRF)





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Rate, Resolution & Inference Time (averaged 50 images) Cesa

Input size (azimuth lines = 1871, range samples = 2500, 2)

	Bit-Rate [bits/samples]	Mean Resolution (Rg\Az) [metres]	Mean PSLR (Rg\Az) [dB]	Mean Compression time [seconds]	Mean De-Compression time [seconds]
Original	-	2.62 \ 5.91	-13.32 \ -19.30	-	-
VQVAE	1.12	2.67 \ 6.28	-13.39 \ -22.48	0.29	0.28
VQVAE-RDA	1.12	2.67 \ 5.83	-13.60 \ -18.99	0.21	1.71

Inference time was calculated on an NVIDIA Tesla T4.

We continue the experiments with VQVAE



Sentinel-1 Instrument simulator



- Stripmap mode, emulates the Sentinel-1 SAR instrument.
- Random location of point targets across 2 datasets (1 for training, 1 for testing).
- (Azimuth line: 16659, Range samples: 25742, 2).
- Entire azimuth line during testing VQVAE, less image biases
- FDBAQ implemented nominally for Sentinel-1.
- Custom RDA implementation for image focusing.

Sentinel-1 Instrument simulator – Range IRF



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Approximately 60% data reduction using VQVAE

Sentinel-1 Instrument simulator – Azimuth IRF



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Approximately 60% data reduction using VQVAE

- Data acquired by Sentinel-1 available from Copernicus
 Open Access Hub (Copernicus Service Information, 2022).
- Data already FDBAQ-processed (quantized, de-quantized).
- Goal to test VQVAE on more realistic scenes, albeit already processed.
- Stripmap-6 Mode, Sao Paulo, Brazil (sea, ships, land, mountains, city).



- Four large data volumes (three for training, one for testing).
- (Azimuth lines = 29934, Range samples = 19950, 2).
- Experimented with and without splitting into range blocks.
- Results with input size (2000, 2500, 2).
- Choice of range block size is not trivial, effect on image quality and resources needs to be studied further.

Focused from FDBAQ-processed data



Contains modified Copernicus Sentinel Data [2022].

Focused from VQVAE-recovered data (1.12 bits/sample)



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Focused from FDBAQ-processed data



Focused from VQVAE-recovered data (1.12 bits/sample)



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Conclusions



- Lossy SAR data compression methodology proposed.
- VQVAE and VQVAE-RDA performed similarly on IRF metrics.
- Image biases were observed in image formed from recovered data. Further studies on data normalization are essential.
- Range block size investigation is required.
- Entropy coding not used. This could further decrease bit-rate.





- Architecture choices and their effect on bit-rate, image quality and inference time need to be studied.
- Quality of phase component in image to be investigated.
- Greater generalizability has not been tested. Experiments with data from different regions of the world are required.
- Preliminary results illustrate the potential of deep learning for lossy SAR raw data compression.