



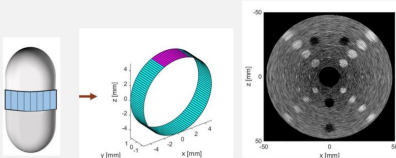
GANs for Ultrasound Compressed Sensing

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Introduction

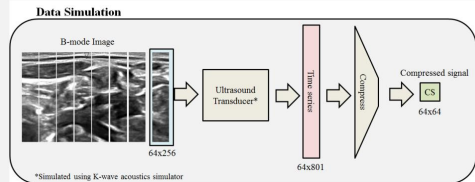
- Motivation:** Ultrasound (US) implants have a small array size with limited power, space, and bandwidth
- Idea:** Proposing an end-to-end compression and reconstruction solution based on generative adversarial networks (GANs)
- Aim:** Reconstruct high quality B-mode images from highly compressed US measurements
- Why GANs?**
 - They perform well in image restoration in similar biomedical applications (e.g. MRI reconstruction [1])
 - They have not yet been investigated for **ultrasound** CS reconstruction



Operation of cylindrical ultrasound array
Simulated B-mode image
(Pill-sized ultrasound device, credit: Arbabian Lab)

Data

- Ultrasound dataset:** publicly available as part of Kaggle nerve segmentation contest [2]
 - 5000 B-mode images of neck, 47 patients
 - Evaluation set:** Images from 7 randomly selected patients
- Each B-mode image is cropped to 9 slices of size 64x256
- K-wave acoustics simulator** is used to extract time series data corresponding to each target B-mode slice
- Timeseries data is then compressed to 8% of its original size using a binary random sampling matrix



*Simulated using K-wave acoustics simulator

Results

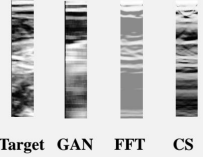
Image quality comparison

	Target	FFT
CS	7.5 dB	22 dB
GAN	15 dB	7.3 dB
FFT	7.6 dB	X

Algorithmic metrics

	Time	CSP*
CS	0.02s	8%
GAN	0.004s	8%
FFT	0.02s	100%

- PSNR is used for comparing quality of image reconstruction to target
- GAN generates closer image to target than CS & FFT



* Compression sensing percentage

Conclusion and Future Work

- GANs show promise in ultrasound image reconstruction with benefits over traditional CS algorithms
- Benefits:**
 - Faster at image reconstruction (5 fold)
 - Produce better quality images than CS when under-sampling ratio is high
- Next steps:**
 - Professional assessment for generated images
 - Training on larger and more diverse dataset
 - Experimenting with different loss functions

Challenges

In compressed sensing:

- Aliasing artifacts due to undersampling
- Low SNR with more undersampling

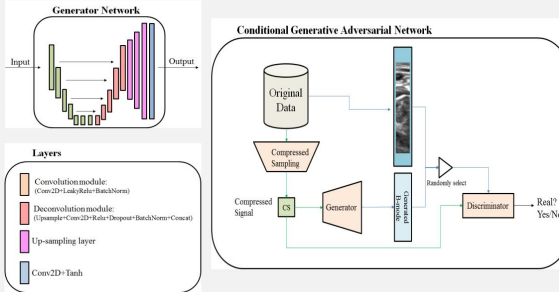
In small size US arrays:

- Low resolution in lateral direction

In our approach:

- RF data & ultrasound B-mode images are scarce
 - Find ways to create training data
- GANs can get unstable during training
- Loss function for compressed sensing is not well defined

Model



References

- Mardani, Morteza, et al. "Deep generative adversarial networks for compressed sensing automates MRI." arXiv preprint arXiv:1706.00051 (2017).
- <https://www.kaggle.com/c/ultrasound-nerve-segmentation>