### EXTRACTION AND DENOISING OF HUMAN SIGNATURE ON RADIO FREQUENCY SPECTRA VIA CONDITIONAL GENERATIVE ADVERSARIAL NETWORK

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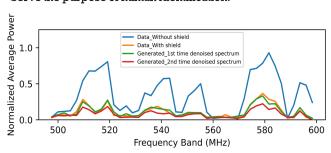
**Problem Statement** 

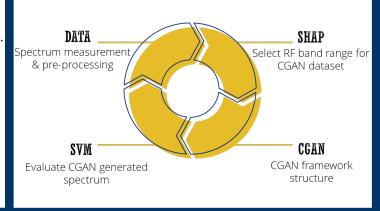
Framework

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#### Our goal in this paper:

- Capture the PRF spectrum contains human signature.
- Extract and denoise the human signature from the RF spectrum.
- Serve the purpose of human identification.





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# Proposed Method

### 01 HARDWARE

- Two cognitive radio devices (SDR)
- Physical shielded environment







#### Three phases of datasets are constructed:

DATASETS

- Phase 1 & 2: two rounds of SHAP analysis
- Phase 3: CGAN training and SVM

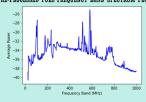
PHASE 1 & 2 DATASET PREPARED FOR SHAP:

	Phase 1	Phase 2	
Frequency Band	24-1000MHz	200-800MH2	
Sampling Step Size	2.4MHz	1.2MHz	
Sampling Rate	2.4MHz	2.4MHz	
Background Condition	With Shield Only	With Shield Only	
Total Sample Quantity	600	400	
Subjects	Human Subject 1-6, & Unoccupied		
Data Prepared For	SHAP	SHAP	
Dataset Notation	$D_{SHAP-1}$	$D_{SHAP-2}$	

PHASE	3	DATASET	PREPARED	FOR	CGAN&SVM

	Phase 3			
Dataset Notation	$D_{\mathrm{w}}$	$D_{wo}$		
Frequency Band	500-600MHz			
Sampling Step Size	2.4MHz			
Sampling Rate	2.4MHz			
Background Condition	With Shield	Without Shield		
Total Sample Quantity	800	800		
Subjects	Human Subject 1 & 2, & Unoccupied			
Data Prepared For	CGAN & SVM	CGAN		

PRE-PROCESSED FULL FREQUENCY BAND SPECTRUM PLOT.



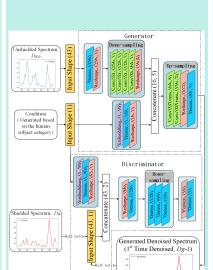
- Down-sampling for feature extraction
- Up-sampling for handling data imbalance

CGAN

- All the Dense, Convolutional, and Convolutional Transpose layers are followed with a Leaky ReLU layer.
- Hyper parameters are tested to get better

#### Discriminator:

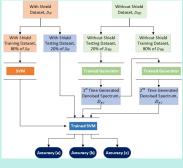
- Binary classifier.
- Dense layers outperform Convolutional layers due to data type and size.





SVM is used in this research to classify the human subjects in 3 datasets:

- The testing samples from D\_W,
- The generated denoised spectrum
- The generated additionally denoised spectrum D (q-2).



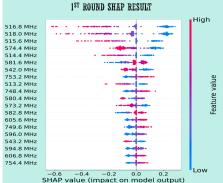
# **Experiment & Results**

#### SHAP: Used on frequency band selection

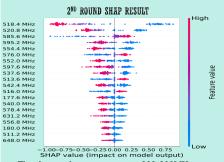
Preliminary frequency band range selection:

- Spectrum initially captured in full frequency
- SHAP result shows top 20 frequencies fall within 100-850MHz.
- Phase 1 dataset frequency is set to 24-1000MHz Phase 1 & 2: Two rounds of SHAP are performed

respectively to narrow down the RF band range.



- The phase 1 dataset frequency range: 24-1000MHz.
- The top ten SHAP results all fall in 513-753MHz.



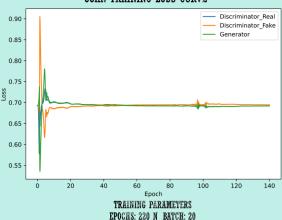
- The dataset frequency range: 200-800MHz.
- The top ten frequencies fall within 513.6-597.6MHz

The phase 3 dataset (CGAN training dataset) frequency range set to: 500-600MHz

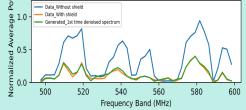
#### CGAN: Training & synthesized spectrum

- CGAN model training dataset consist of 608 unshielded samples and 608 shielded samples.
- Discriminator input: shielded samples as target
- Generator training input: unshielded samples.

#### CGAN TRAINING LOSS CURVE

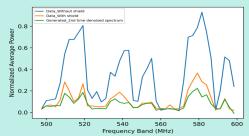


# CGAN SYNTHESIZED SPECTRUM - 1ST TIME DENOISED



- The synthesized VS shielded spectrum
- Small deviations

#### CGAN SYNTHESIZED SPECTRUM - 2ND TIME DENOISED

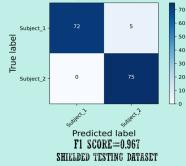


- Generator input: the synthesized 1st denoised spectrum.
- The same trained CGAN model is used.

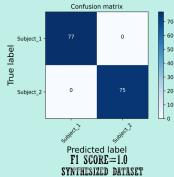
## Evaluation of the synthesized spectrum

- SVM Classify the synthesized spectra dataset.
- SVM is trained on the shielded training dataset.
- CGAN training target: obtain same F1 score as the shielded testing dataset.

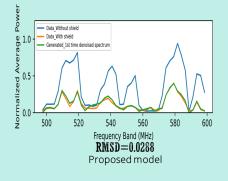
#### SVM CLASSIFICATION RESULTS



20% of shielded spectrum dataset



Both 1st and 2nd time denoised synthesized dataset



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# Conclusion & Future works

#### CONCLUSION

- Synthesized spectrum exceeds the expectation, outperform the target spectrum.
- CGAN up-sampling achieved additional signature extraction function by altering data weight.
- The CGAN training strategy:
- The CGAN structure down-sampling and up-sampling tunning. 1.
- 2. The CGAN generator hyper parameters setting.
- The CGAN discriminator Dense layer VS Convolutional layer.

#### **FUTURE WORKS**

Experiment setup:

- Increase human subject quantity and variety.
- Change of baseline condition.
- Upgrade shielding material.

Improvement on algorithm:

- Increase labelling dimensions.
- Adding baseline information into condition data.
- Investigate ResNet into CGAN.