

RF Interference Mitigation for UWB SAR using Image Sparsity

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RFI Suppression and Sparse Image Formation for UWB SAR: Objectives/Motivations

- Propose an image formation algorithm for ultra wideband (UWB) synthetic aperture radar (SAR) (VHF/UHF spectrum).
- Suppress radio frequency interference (RFI) and reconstruct a SAR image.
- RFI suppression shouldn't introduce artefacts.
- Achieved by leveraging image sparsity model.

Outline

- 1 Background
- 2 UWB SAR Model
- 3 RFI Suppression and Image Formation
- 4 Numerical Experiments

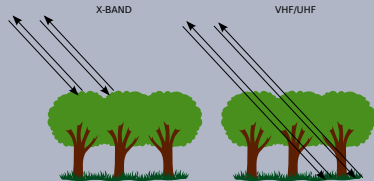
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Background

Why use VHF/UHF Spectrum?

- Foliage Penetration (FoPEN) Radar
- Ground Penetration Radar (GPR)
- Scattering is dependent on wavelength.



Background

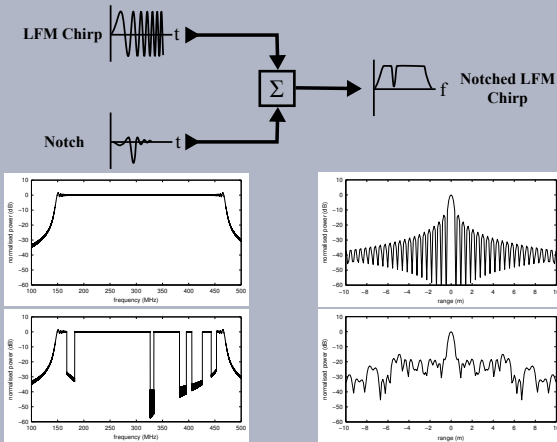
Issues which effect the VHF/UHF spectrum

- Interference between SAR systems and radio, television and communications systems.
- Radio frequency interference (RFI)
- Interference Types:
 - 1 SAR systems can interfere with other spectrum users.
 - 2 Other users in the spectrum can interfere with SAR system.



Background

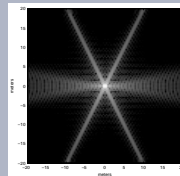
Notched LFM on Transmit



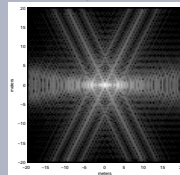
Background

RFI suppression

- Strong interference from AM/FM transmitters.
- RFI pre-processing suppression methods: estimate-and-subtract or linear filter.
- Estimate-and-subtract: estimate the frequencies and phases of the RFI and then abstract.
Can be computationally expensive and approximation dependent.
- Linear filter: minimise RFI using linear filter, e.g. LMS filter and Wiener filter.
Can produce large side lobes.



example PSF



filtered PSF

Outline

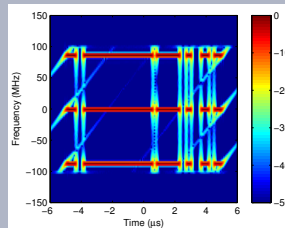
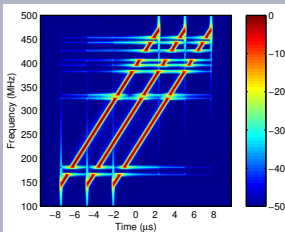
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UWB SAR Model

Dechirping of Notched LFM

- Received signal is weighted by the spectrum of the transmitted notched LFM chirp.

$$\tilde{y}(t) \approx |S(\omega_0 + 2\alpha t)| y(t)$$



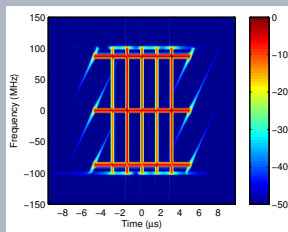
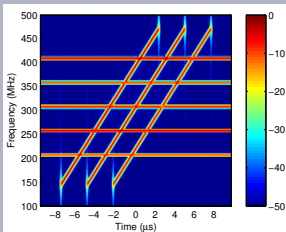
UWB SAR Model

Dechirping of RFI

- The RFI signal after dechirping is approximately given by:

$$\tilde{s}_{\text{rfi}}(t) \approx \left[\frac{1}{2\pi} S_{\text{rfi}}(\omega_0 + 2\alpha t) \exp(j\alpha t^2) \right] * \left[\sqrt{\frac{\pi}{j\alpha}} \frac{\omega_{\text{if}}}{2\pi} \text{sinc}\left(\frac{\omega_{\text{if}} t}{2}\right) \right]$$

- If $s_{\text{rfi}}(t)$ is a random signal then the $\mathbb{E}\left(|\tilde{s}_{\text{rfi}}(t)|^2\right)$ will be the PSD of $s_{\text{rfi}}(t)$ convolved with a squared sinc function.



UWB SAR Model

Notation

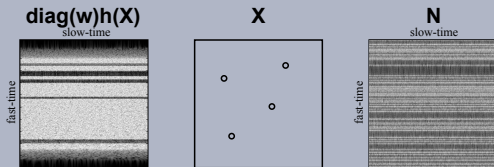
Spotlight-mode Dechirped SAR with RFI System Model:

$$\mathbf{Y} = \text{diag}(\mathbf{w})h(\mathbf{X}) + \mathbf{N}$$

$$\mathbf{Y} \in \mathbb{C}^{M' \times N'}, \mathbf{N} \in \mathbb{C}^{M \times N}, \mathbf{w} \in \mathbb{R}^M$$

Linear Reconstruction:

$$\hat{\mathbf{X}} = g(\mathbf{Y}), \text{ where, } g(h(\mathbf{X})) = \mathbf{X}$$



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RFI Suppression

Filter-based RFI suppression

Linear RFI Filtered Reconstruction:

$$\hat{\mathbf{X}} = g(\mathbf{H} \text{vec}(\mathbf{Y}))$$

$$\mathbf{H} = \text{diag}([\mathbf{H}_1, \dots, \mathbf{H}_{N'}])$$

Wiener Filter:

$$\mathbf{H}_{n'} = \mathbf{I} - \mathbf{Q}_{n'} (\mathbf{Q}_{\tilde{\mathbf{y}}_{n'}} + \mathbf{Q}_{n'})^{-1} \text{ for } \mathbf{Q}_{\mathbf{x}} = \mathbb{E} [\mathbf{x}\mathbf{x}^H]$$

Reconstruction Error:

$$\|\mathbf{X} - \hat{\mathbf{X}}\|_F \leq \|g\| \|(\mathbf{I} - \mathbf{H}) \text{vec}(h(\mathbf{X}))\|_F + \|g\| \|\mathbf{H} \text{vec}(\mathbf{N})\|_F$$

which can be compared with

$$\|\mathbf{X} - \hat{\mathbf{X}}\|_F \leq \|g\| \|\mathbf{N}\|_F$$

RFI-aware Sparse Image Formation

RFI-aware Sparse Image Formation

$$\hat{\mathbf{X}} = \underset{\mathbf{X}}{\text{minimise}} \|\mathbf{X}\|_1$$

subject to $\|\mathbf{Y} - h(\mathbf{X})\|_{\mathbf{Q}_N^{-1}} \leq \epsilon,$

where, $\|\mathbf{A}\|_{\mathbf{Q}} = \text{vec}(\mathbf{A})^H \mathbf{Q} \text{vec}(\mathbf{A})$

Motivation Theory: Compressive Sensing

Modified Restricted Isometry Property:

$$C(1 - \delta_{2S}) \|\mathbf{X}_{2S}\|_F^2 \leq \|h(\mathbf{X}_{2S})\|_{\mathbf{Q}_N^{-1}}^2 \leq C(1 + \delta_{2S}) \|\mathbf{X}_{2S}\|_F^2$$

Reconstruction Error Bound:

$$\|\mathbf{X} - \hat{\mathbf{X}}\|_F \leq C_{1,S} \|\mathbf{N}\|_{\mathbf{Q}_N^{-1}} + C_{2,S} S^{-1/2} \|\mathbf{X} - \mathbf{X}_S\|_1,$$

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Numerical Experiments

VHF/UHF SAR simulation Parameters

parameter	value
carrier frequency (ω_0)	$2\pi \times 308$ Mrad/s
altitude	7000 m
number of targets	20
chirp bandwidth ($2\alpha T$)	$2\pi \times 324$ Mrad/s
stand-off distance	7000 m
number of interferes	80
IF bandwidth	$2\pi \times 60$ Mrad/s
aperture length	7000 m
signal to noise ratio (SNR)	60 dB
scene radius (L)	75 m
number of aperture samples	300
signal to interference ratio (SIR)	-30 dB
transmit notch centre frequencies	175, 330, 389, 416 and 448 $\times 2\pi$ Mrad/s
transmit notch bandwidths	15, 7, 13, 20 and 10 $\times 2\pi$ Mrad/s

Numerical Experiments

RFI-aware Sparse Image Formation Implementation

Estimate Noise Covariance:

Estimate \mathbf{Q}_N using ten “dead-time” measurements.

Assume elements of \mathbf{N} are independent.

Unconstrained Optimisation:

$$\hat{\mathbf{X}} = \underset{\mathbf{X}}{\text{minimise}} \|\mathbf{X}\|_1 + \lambda(\|\mathbf{Y} - \text{diag}(\mathbf{w})h(\mathbf{X})\|_{\mathbf{Q}_N^{-1}} - \epsilon)$$

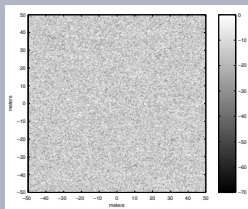
Approximately solved using thirty iterations of a fast iterative shrinkage thresholding algorithm.

Project onto Domain of $h(\cdot)$:

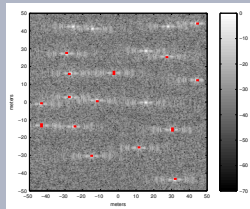
$$\hat{\mathbf{X}} \leftarrow g(h(\hat{\mathbf{X}}))$$

Numerical Experiments

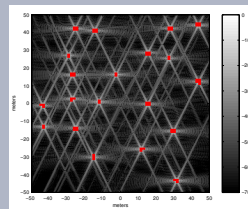
CFAR Results



Filtered back-projection



Wiener filtered followed by filtered back-projection



RFI-aware sparse image formation

Conclusions and Future Work

Conclusions

- Theoretically near optimal performance for ideally sparse SAR images.
- Improved performance for simulated UWB SAR.

Future Work

- Balance trade off between image quality and computational complexity (parallel processing?).
- Propose a modified scheme for systems with interference that is not stationary.
- Investigate algorithm on real UWB SAR data.

Thank you for your attention!

Questions?

