

SubNyquist Electronic Surveillance

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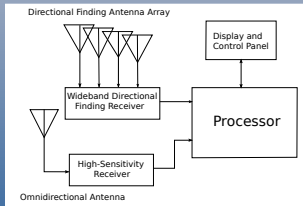


Source Separation and Sparsity Theme Meeting,
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Electronic Support Measures (ESM)

- The task is to detect all RF emitters, for identifying presence of threats.
- It has a passive monitoring system.
- While ESM signals are very dense, e.g. can be of millions of pulses per second, they have very sparse TF representations.
- ESM systems can be noise limited, rather than sparsity limited.



Techniques for ESM receivers

- Instantaneous Frequency Measurements: limited spectral sensitivity.
- Rapid Frequency Swapping A/D's: limited temporal sensitivity.
- Wideband Analog to Digital Convertors: multi GHz A/D's.
- **Proposal:** Sub-Nyquist Analog to Information Convertor.

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SubNyquist Sampling

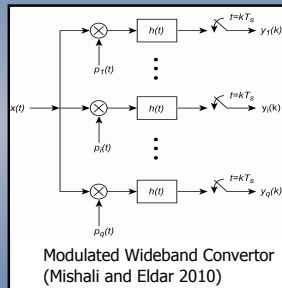
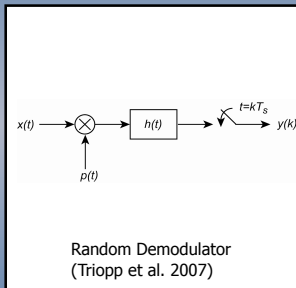
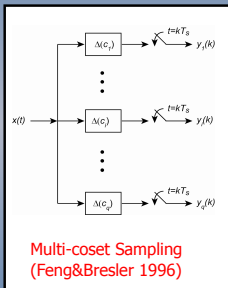
- Why?
 - 1 Sampling at the rate of Nyquist is difficult or costly for some applications, *e.g.* Wideband A/D's and Wideband Digital Receivers.
 - 2 It is waste of resources, if we sample at a rate, much higher than the information rate.
 - 3 Allows us to have an application specific sampling strategy, *i.e.* exploring signal structures.
- How?
 - 1 Using underlying signal structures, *e.g.* sparsity.
 - 2 Incorporating non-uniform sampling (random?) in the sensing framework.
 - 3 Non-linear reconstruction of signals.

SubNyquist Sampling

- What would be the challenges?
 - ① *Analog Hardware*: How efficiently can we design the analog part?
 - ② *Computational Complexity*: How efficient can we implement the non-linear recovery algorithm?
 - ③ *Noise Sensitivity*: Sensitivity to the input noise?
 - ④ *Robustness*: How much the sub-Nyquist algorithm is sensitive to the **signal model mismatch** and **circuit design tolerances**.

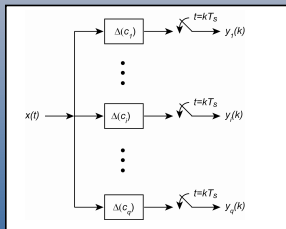
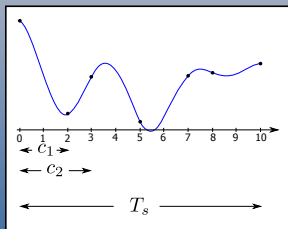
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SubNyquist Sampling Techniques



Multi-coset Sampling Framework

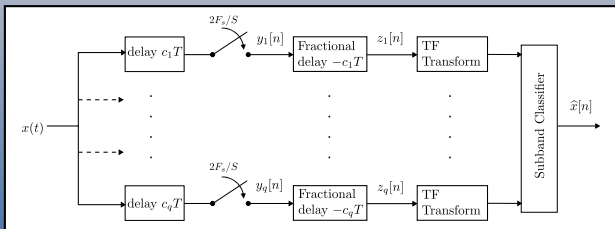
- Non-uniform sampling technique [Feng and Bresler, 1996].
- Sparse multiband signal model.
- A subspace method for reconstruction by Feng et al.
- A convex optimisation problem for reconstruction by [Mishali and Eldar 2009].



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Proposed SubNyquist Sampling Framework

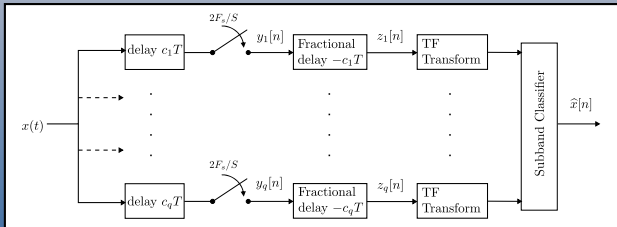
- A Multicoset sampling strategy.
- Avoiding any complicated operations e.g. SVD, ℓ_1 minimisation.
- The signal model have to fit into the ESM.



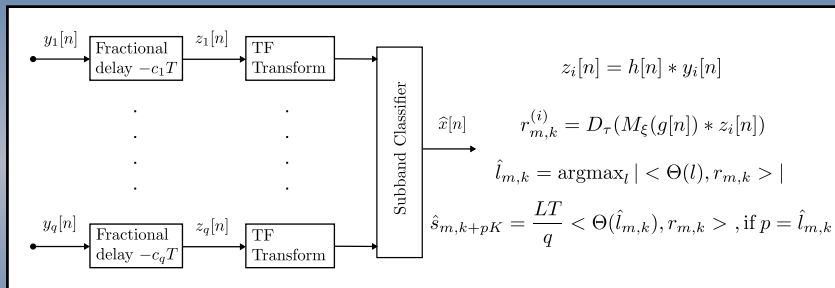
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Components of Proposed Framework

- *A bank of multicoset channels:* it has distinguished delays.
- *Digital Fractional Delay (DFD) filters.*
- *Time-Frequency transform:* STFT currently has been used.
- *Subband Classifier:* Composed of a linear operator (Harmonic Frame), followed by a simple maximum-absolute value operator.



Low-Complexity MC Reconstruction Algorithm

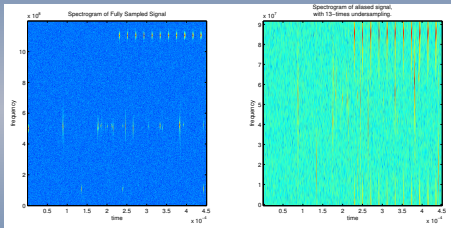


- Θ : Harmonic Frame
- $\hat{x}[n]$ can be reconstructed using inverse TF transform.

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Assumptions and Properties of Proposed Framework

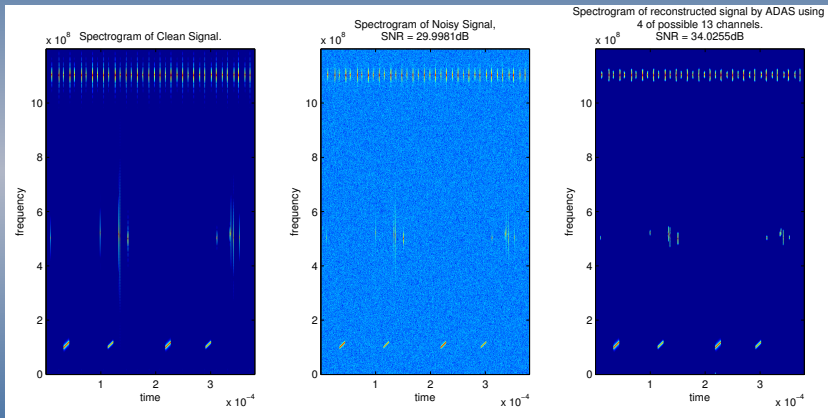
- *Approximate Disjoint Aliased Support (ADAS)*: it is different to the sparsity.



- *Does not require random sampling*: optimal delay parameters can be yielded using Harmonic Grassmannian Frames (HGF).
- *Practical Issues with DFD filters*: finite length filters introduce distortion → Combining TF and DFD filters.

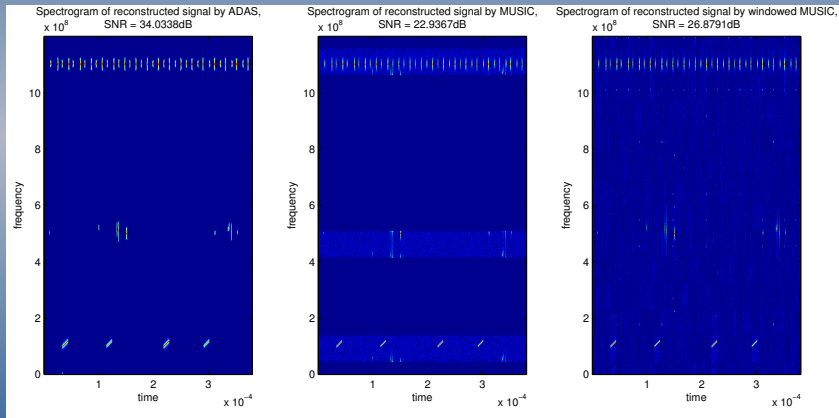
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Evaluation with Radar ESM signals



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Comparison with MUSIC type recoveries [Feng and Bresler, 1996]



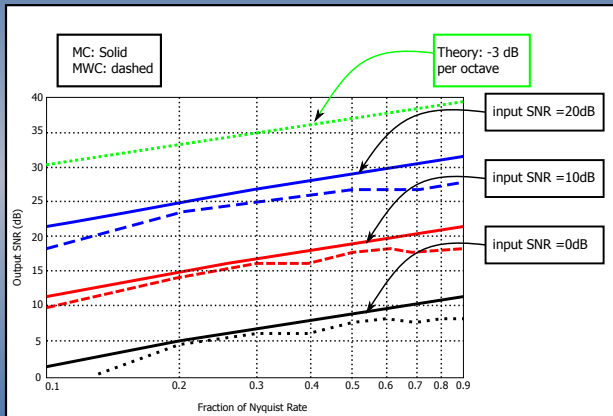
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Advantages of the Proposed Method:

- **Non-iterative**: it may be pipelined.
- Can use only **a few** Multi-coset channels, e.g. as few as $q = 2$.
- Uses a different signal model, i.e. **ADAS**, which matches well to some classes of signals, e.g. ESM.
- **Simple analog hardware (digitiser)**: periodic non-uniform sampling pattern: easier to implement than a random sampling pattern.
- **Large Dynamic Range**, e.g. 70 dB, which makes it suitable for the low probability of intercept signals.
- **Continuously monitoring** wideband RF signals, in a contrast with the rapid frequency swapping technique.

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Noise Folding in Sub-Nyquist Sampling



Conclusions

- A low SWAP subNyquist algorithm was presented for ESM application.
- The proposed technique uses parsimonious signal structures.
- When ESM signals are structurally sparse in some TF domains, we can assure signal recovery, by selecting a moderate undersampling factor.
- The proposed algorithm out performs the canonical MUSIC based recovery algorithms for the given ESM signals.

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Future Work

- An optimal TF selection to maximise coherent processing gain.
- Sensitivity and robustness analysis.
- More simulations around about comparisons with canonical methods.
- Further optimisation of implementation of DFD filters.



Thanks for your attention.