

Subspace-based 1-bit Wideband Spectrum Sensing

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This work focuses on

- Power-efficient wideband spectrum sensing for cognitive radio sensor networks

We consider

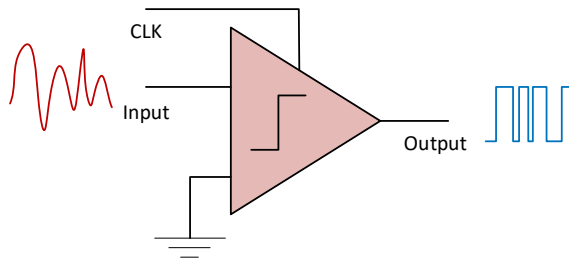
- Spectrum sensing in a wideband cognitive radio system where 1-bit ADCs are adopted at the RF sensors

The objective is

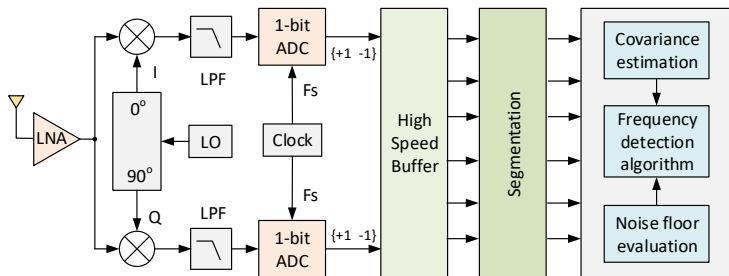
- To detect the occupation states of individual sub-bands simultaneously in a wide frequency range

1-bit ADCs for wide-band spectrum sensing?

- Can be implemented using a single comparator
- Ultra-low driving power and circuit complexity
- Incurs only a small performance loss compared to high-resolution ADCs in low-SNR regime
- Have been considered for massive MIMO, low-cost radar

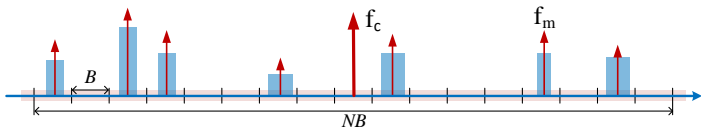


System architecture for 1-bit wideband spectrum sensing



- Homodyne RF architecture
- No automatic gain control (AGC) required
- Size of buffer can be greatly reduced
- Low signal processing complexity

1-bit Wideband Quantized Signal Model



Continuous analog signal:

$$y(t) = \sum_{m=1}^M \alpha_m(t) e^{j2\pi f'_m(t-\tau_m)} + w(t), \quad (\text{P1})$$

Discrete received signal:

$$y[n] = \sum_{m=1}^M \alpha_m[n] e^{j2\pi f'_m \left(\frac{n}{F_s} - \tau_m \right)} + w[n] \quad (\text{P2})$$

1-bit quantized signal:

$$q[n] = \frac{1}{\sqrt{2}} (\text{sign}(\Re\{y[n]\}) + \mathbf{j} \text{sign}(\Im\{y[n]\})) \quad (\text{P3})$$

Problem Formulation

- The M signals with frequencies $\{f_m\}_{m=1}^M$ are assumed to lie in exactly M sub-bands
- The objective of the RF sensor is to provide an N -bit digital word representing the states of the spectrum sub-bands
- We define $2N$ binary hypotheses $\{\mathcal{H}_{0,n}\}_{n=1}^N$ and $\{\mathcal{H}_{1,n}\}_{n=1}^N$, in which $\mathcal{H}_{0,n}$ denotes the idle state of the n -th sub-band and $\mathcal{H}_{1,n}$ represents the active state
- For each sub-band, a test statistics χ_n is formulated based on the 1-bit sampled data, and a test decision is given as follows:

$$\begin{cases} \text{Choose } \mathcal{H}_{0,n}, & \text{if } \chi_n < \theta_n, \\ \text{Choose } \mathcal{H}_{1,n}, & \text{if } \chi_n > \theta_n, \end{cases} \text{ for } n \in \{1, 2, \dots, N\}, \quad (\text{P4})$$

Subspace-based Technique for Wideband Spectrum Sensing

- Based on signal covariance, typical methods are MUSIC and ESPRIT

Received signals in vector form:

$$\mathbf{y} = \mathbf{s} + \mathbf{w} = [y[0], y[1], \dots, y[N-1]]^T, \quad (\text{P5})$$

Covariance Matrix for \mathbf{y} :

$$\mathbf{R}_{yy} = \mathbb{E} \left\{ (\mathbf{s} + \mathbf{w})(\mathbf{s} + \mathbf{w})^H \right\} = \mathbf{A}\mathbf{\Delta}\mathbf{A}^H + \sigma_w^2 \mathbf{I} \quad (\text{P6})$$

- We have eigen-decomposition $\mathbf{R}_{yy} = \mathbf{U}(\mathbf{\Lambda} + \sigma_w^2 \mathbf{I})\mathbf{U}^H$
- The signal and noise spaces are orthogonal for \mathbf{R}_{yy} , we have

$$\mathbf{U} = [\mathbf{U}_s \ \mathbf{U}_n]$$

- \mathbf{U}_n of size $N \times (N - M)$ defines the noise subspace

Subspace-based Technique for Wideband Spectrum Sensing

The core idea is to estimate frequencies using the pseudo-spectrum

$$P_{\text{pseu}}(f) = \frac{1}{\mathbf{v}^H(f) \mathbf{U}_n \mathbf{U}_n^H \mathbf{v}(f)} = \frac{1}{\|\mathbf{U}_n^H \mathbf{v}(f)\|_2^2}. \quad (\text{P7})$$

where $\mathbf{v}(f) = \left[1, e^{\frac{j2\pi}{F_s} f}, e^{\frac{j4\pi}{F_s} f}, \dots, e^{\frac{j2(N-1)\pi}{F_s} f} \right]^T$ is the frequency-domain steering vector.

- If f equals one of the carrier frequencies of the spectrum components, the denominator is small, and there will be M largest peaks.

How to estimate the covariance based 1-bit quantized data?

- With 1-bit ADC, we only have $\mathbf{R}_{\mathbf{q}\mathbf{q}} = \mathbb{E}\{\mathbf{q}\mathbf{q}^H\}$
- According to **Bussgang theorem** and **Vleck's arcsine law**, we have

$$\mathbf{R}_{\mathbf{q}\mathbf{q}} = \frac{2}{\pi} \left[\arcsin \left(\boldsymbol{\Sigma}_{\mathbf{y}}^{-\frac{1}{2}} \mathbf{R}_{\mathbf{y}\mathbf{y}} \boldsymbol{\Sigma}_{\mathbf{y}}^{-\frac{1}{2}} \right) \right], \quad (\text{P8})$$

where $\boldsymbol{\Sigma}_{\mathbf{y}} = \text{diag}(\mathbf{R}_{\mathbf{y}\mathbf{y}})$ and $\arcsin(\cdot)$ is element-wise.

- The normalized covariance for unquantized \mathbf{y} can be approximated as

$$\bar{\mathbf{R}}_{\mathbf{y}\mathbf{y}} \doteq \frac{\pi}{2} \mathbf{R}_{\mathbf{q}\mathbf{q}} + \left(1 - \frac{\pi}{2}\right) \mathbf{I} \quad (\text{P9})$$

- For the an eigenvector \mathbf{v} of $\mathbf{R}_{\mathbf{y}\mathbf{y}}$ with $\mathbf{R}_{\mathbf{y}\mathbf{y}} \mathbf{v} = \lambda \mathbf{v}$, we have

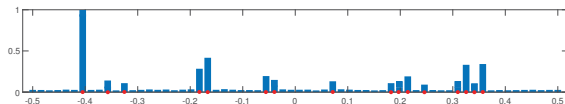
$$\frac{\pi}{2} \mathbf{R}_{\mathbf{q}\mathbf{q}} \mathbf{v} \doteq \left(\frac{\lambda}{p} - 1 + \frac{\pi}{2} \right) \mathbf{v}, \quad (\text{P10})$$

which implies that $\mathbf{R}_{\mathbf{q}\mathbf{q}}$ and $\mathbf{R}_{\mathbf{y}\mathbf{y}}$ have identical signal and noise spaces

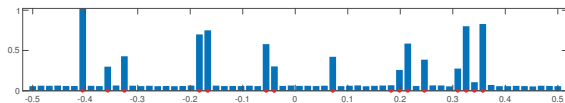
Subspace-based 1-bit wideband spectrum sensing algorithm

1. Acquire L snapshots of 1-bit quantized data $\{\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_L\}$
2. $\hat{\mathbf{R}}_{\text{qq}} \leftarrow \frac{1}{L} \sum_{l=1}^L \mathbf{q}_l \mathbf{q}_l^H$, $\hat{\mathbf{R}}_{\text{yy}} \leftarrow \frac{\pi}{2} \hat{\mathbf{R}}_{\text{qq}} + (1 - \frac{\pi}{2}) \mathbf{I}$
3. $\hat{\mathbf{R}}_{\text{yy}} = \hat{\mathbf{U}} \hat{\mathbf{\Lambda}} \hat{\mathbf{U}}^H$, where $\hat{\mathbf{U}} = [\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_N]$, and $\hat{\mathbf{\Lambda}} = \text{diag}\{\lambda_1, \lambda_2, \dots, \lambda_N\}$ with $\lambda_i \geq \lambda_j$ for $i < j$
4. Estimate the number of spectrum components using a Minimum Description Length(MDL) estimator
5. Partition $\hat{\mathbf{U}}$ into $[\mathbf{U}_s \ \mathbf{U}_n]$
6. Compute pseudo-spectrum $\frac{1}{\|\mathbf{U}_n^H \mathbf{v}(f)\|_2^2}$ for $f \in \{f_1, f_2, \dots, f_N\}$
7. Find the $N - M$ smallest elements in pseudo-spectrum, estimate the noise floor P_{noise} as the mean of the $N - M$ smallest elements
8. If $p_s(n) > 10^{\frac{\gamma}{10}} P_{\text{noise}}$ ($\gamma = 3$ dB), mark the n -th sub-band as occupied

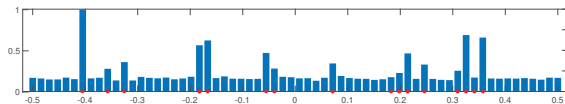
Performance Evaluation



(a) Subspace-based pseudo-spectrum



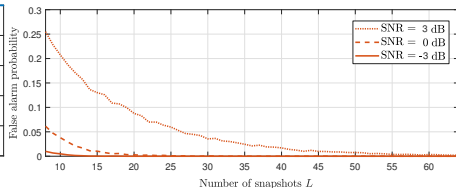
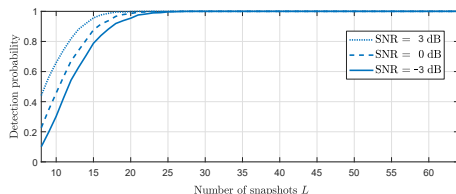
(b) DFT-based spectral power



(c) Correlation-based spectral power

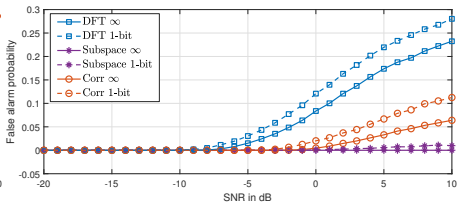
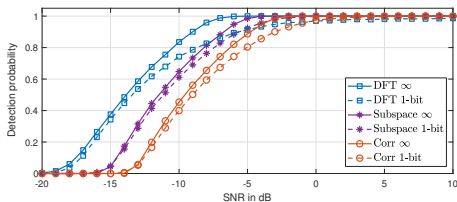
- Subspace-based method has a more distinguishable floor compared to FFT-based and correlation-based method

Time Resolution vs Detection Performances



- When SNR is 0, the proposed method has perfect performances with 32 snapshots, corresponds to a time-resolution of $3.2 \mu s$
- When SNR is high, more snapshots of data are needed to attain a zero false alarm rate
- In high SNR regime, more samples are needed to average out the 1-bit quantization distortion in estimating the empirical covariance matrix
- 1-bit wideband spectrum sensing has a preferred operational SNR range

Performance Comparisons under Different SNR Conditions



- Performances with 1-bit ADCs are comparable to those with infinite-resolution ADCs
- The detection probability of the proposed method is lower than that of DFT-based and higher than correlation-based
- The proposed method achieves almost zero false alarm and is superior compared to the other two

Concluding Remarks

- We have proposed a subspace-based 1-bit wideband spectrum sensing method, it exhibits ultra-low power consumption, low memory and computation demands, and is suitable for larger-scale RF sensor network deployments.
- Our results suggest that the superiority of the subspace technique in parameter estimation translates into efficacy in 1-bit wideband spectrum sensing.
- We show by simulations that the proposed method exhibits near-zero false alarm while achieves similar detection probability as compared to other typical sensing methods.