

Experimental Study of Spectrum Estimation and Reconstruction based on Compressive Sampling for Cognitive Radios

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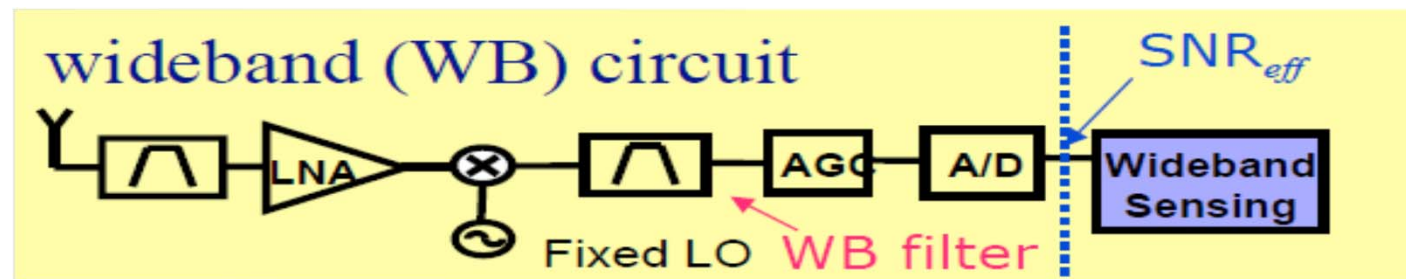
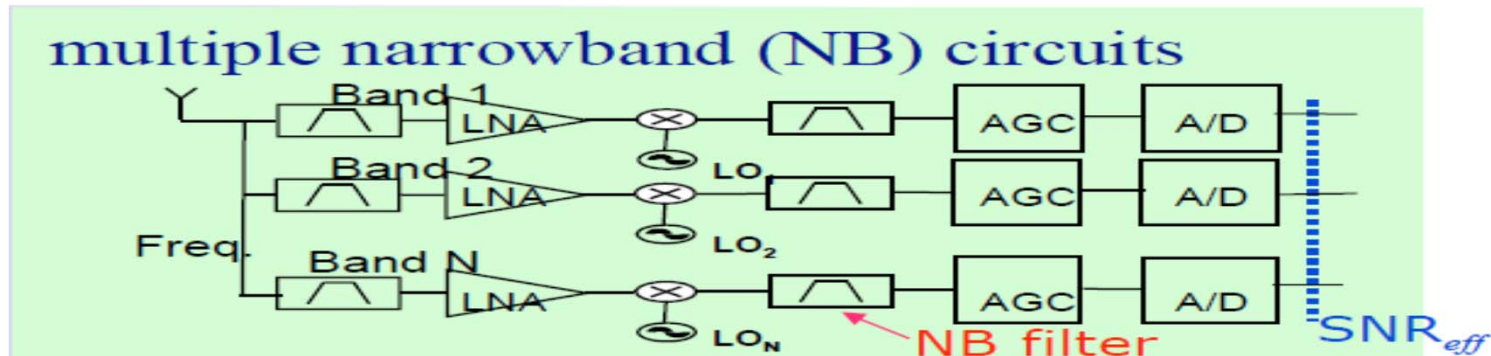
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Wideband Signal Acquisition

❑ Choices of RF Circuits: Multiple NB or single WB

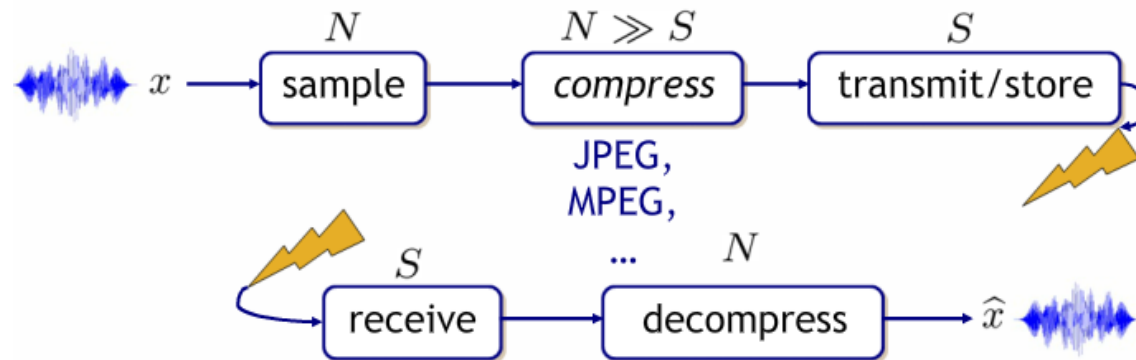


Compressed Sampling [Tian'07-08, Haupt'07]

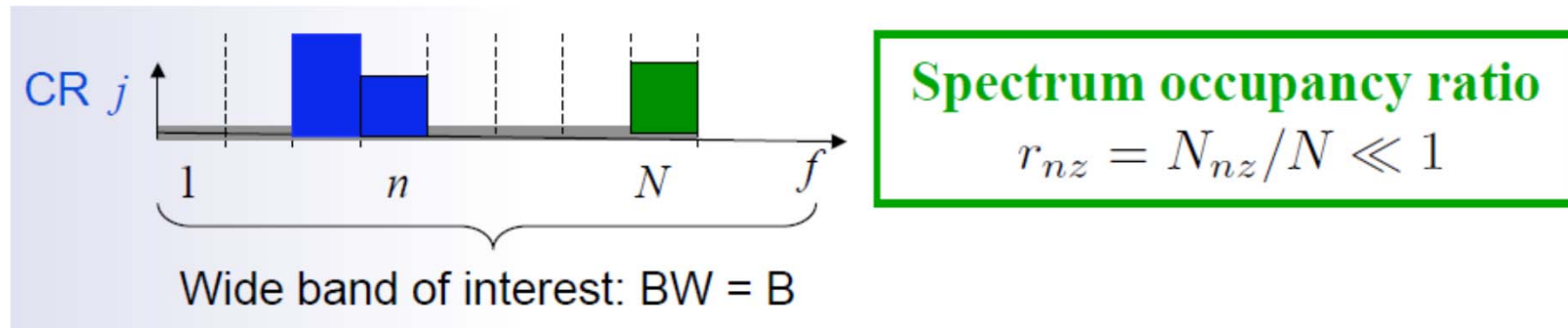
- ❑ Conventional spectral estimation methods require to operate at or above Nyquist rate.
- ❑ Requires high rate A/D or bank of low rate A/D for wideband signals.
- ❑ Compressed Sampling: sub-Nyquist rate sampling and reliable signal recovery via computationally feasible algos.
- ❑ Applicable to **sparse signals**.

Compressed Sampling

❑ CS is Essentially different from **Sample-then-Compress** (JPEG, MPEG etc)



Limits on Sampling Rates



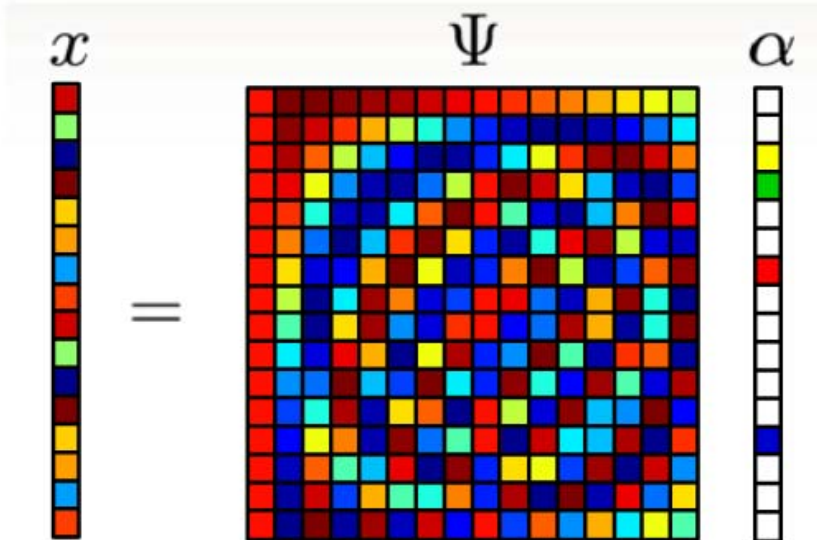
- ❑ Lower bounds on sampling rates f_s
 - ❑ Lowest f_s for reconstruction without aliasing
 - ❑ Nyquist Rate = $2B$
 - ❑ Lowest f_s for reconstruction of CR signals
 - ❑ Motivating factor for CR is low spectrum utilization
 - ❑ Landau rate = $2B_{\text{eff}} = 2r_{nz}B < \text{Nyquist Rate}$

CS Basics

$$\begin{array}{c} M \times 1 \\ \text{measurements} \end{array} \begin{array}{c} y \\ \text{[color bar]} \end{array} = \begin{array}{c} \Phi \\ \text{[M x N grid]} \\ M \times N \end{array} \begin{array}{c} x \\ \text{[color bar]} \\ N \times 1 \\ \text{sampled} \\ \text{signal} \end{array}$$

The diagram illustrates the basic equation of Compressed Sensing: $y = \Phi x$. On the left, a vertical column of 8 colored squares (light green, yellow, orange, red, light green, blue, cyan) represents the $M \times 1$ measurements vector y . In the center is a 10×10 grid of colored squares representing the measurement matrix Φ , with the dimensions $M \times N$ indicated below it. On the right, a vertical column of 10 colored squares (red, light green, dark blue, dark red, yellow, orange, blue, red, light green, orange) represents the $N \times 1$ sampled signal vector x . The word "sampled" is written in blue italics above "signal".

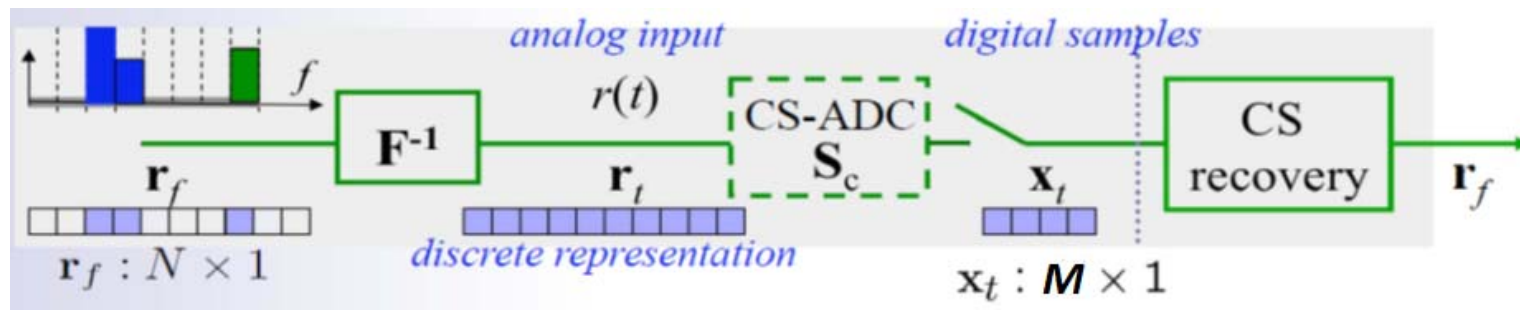
CS Basics

$$x = \sum_{j=1}^N \alpha_j \psi_j$$
$$= \Psi \alpha$$


□ Compressive sampling [Chen etal'98], [Candes etal'04-06]

□ Given y and Φ , unknown x can be found with high probability.

Sub-Nyquist rate Sampling



- ❑ Received signal: $r(t): t \in [0, NT_s]$
- ❑ Discrete representation: $\mathbf{r}_t \leftrightarrow \mathbf{r}_f = \mathbf{F}\mathbf{r}_t$

- ❑ Linear sampling: $\mathbf{x}_t = \mathbf{S}_c \mathbf{r}_t = \mathbf{S}_c \mathbf{F}^{-1} \mathbf{r}_f$
- ❑ Compression: $\mathbf{S}_c: M \times N$

- ❑ Various designs of random samplers [Kirolos etal'06, Hoyos etal'08]

Sub-Nyquist rate Sampling

❑ Reconstruction is achieved by solving the following convex optimization problem:

$$\arg \min_{\mathbf{r}_f} \|\mathbf{r}_f\|_1, \quad s. t. \quad \mathbf{x}_t = \mathbf{S}_c \mathbf{F}^{-1} \mathbf{r}_f$$

❑ Techniques to Solve the above problem:

❑ Linear Programming: Basis Pursuit [Chen et al'98]

❑ Iterative greedy algorithms: Matching Pursuit [Duarte et al'05] and Orthogonal Matching Pursuit [Tropp and Gilbert'07]

Formally the l_n -norm of \mathbf{x} is defined as:

$$\|\mathbf{x}\|_n = \sqrt[n]{\sum_i |x_i|^n}$$

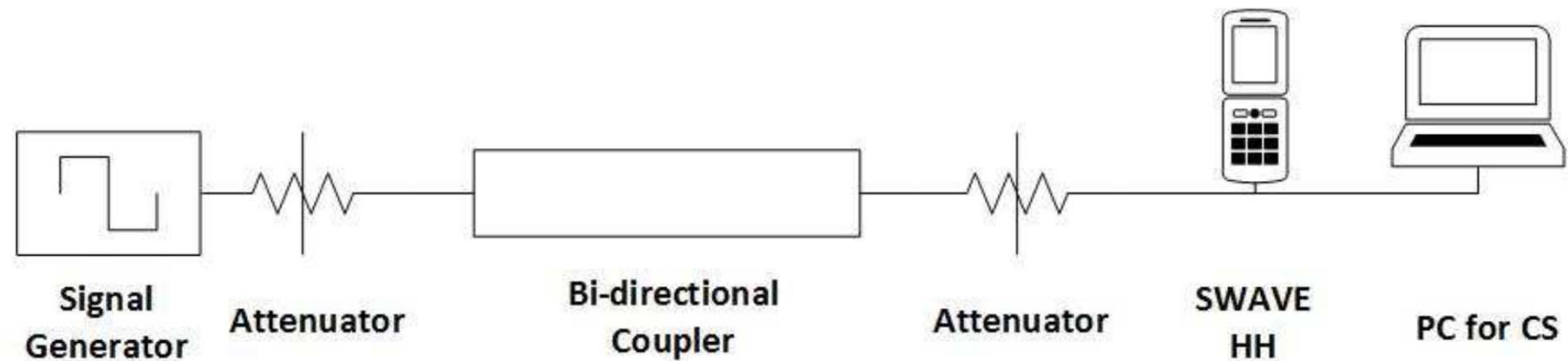
Test bed Architecture

- ❑ Physical architecture of the assembled test bed



Test bed Architecture

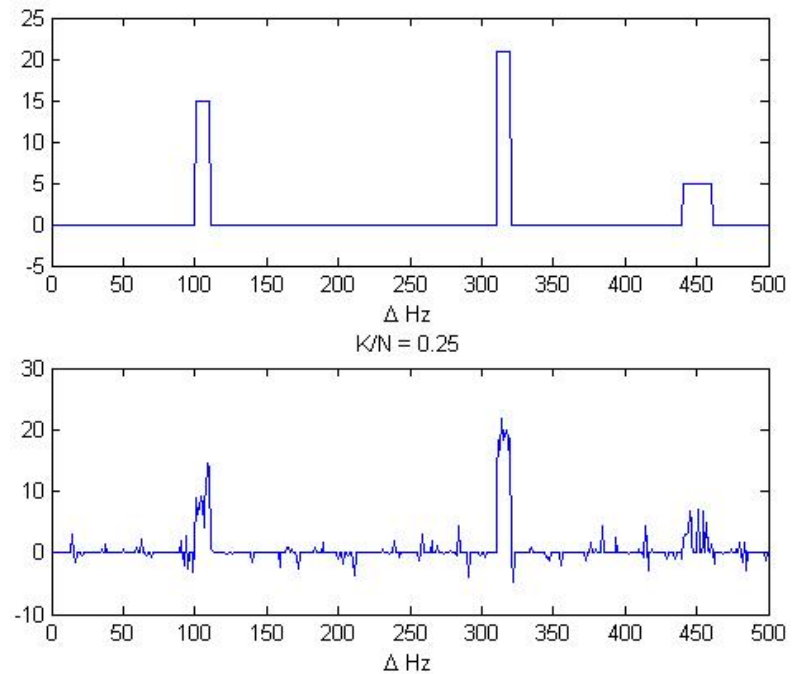
❑ Simplified block diagram of the assembled test bed



CS on simulated data

Simulation paras:

- 500 ΔHz wideband signal
- Non-zero elements = 40 ΔHz
- Occupancy = 40/500
- Basis Pursuit (BP) for compressed sensing.



CS on test bed data

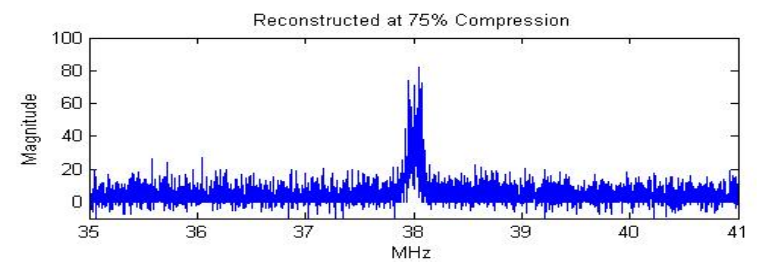
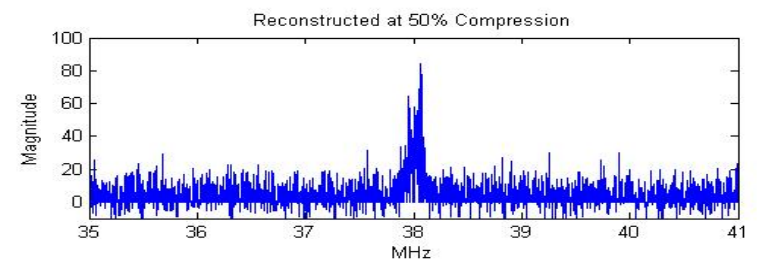
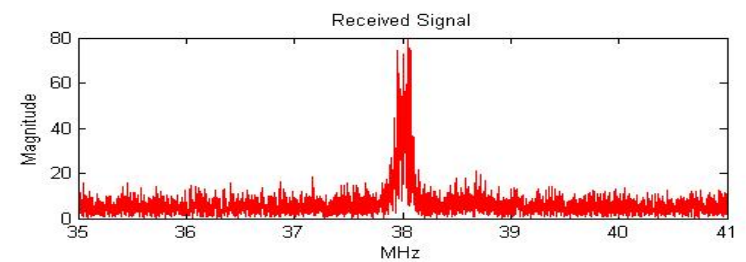
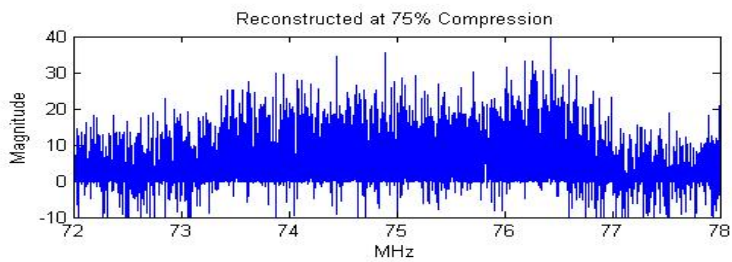
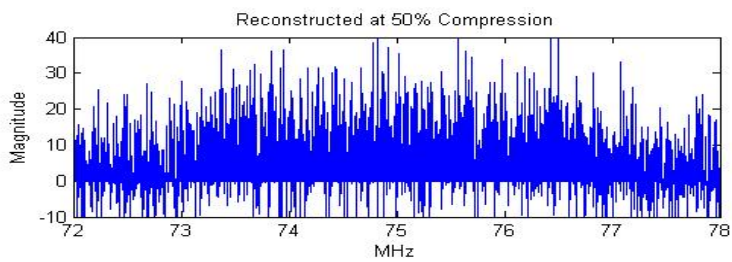
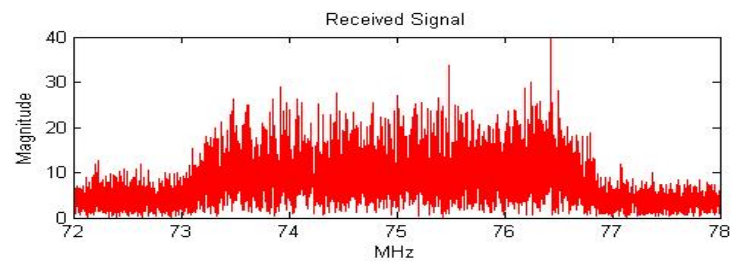
☐ Accumulated data:

- ☐ 120 MHz wideband spectrum
- ☐ Sampling rate of SWAVE HH = 250 Msamples/s
- ☐ FPGA output = 8192 samples / 1.3 s (not sufficient for meaningful waveform realization)
- ☐ 10 bursts of 8192 samples (8.192×10^4 samples)

☐ Analyzed waveforms:

- ☐ 3 MHz wideband Gaussian waveform
- ☐ 250 KHz narrowband GSM waveform

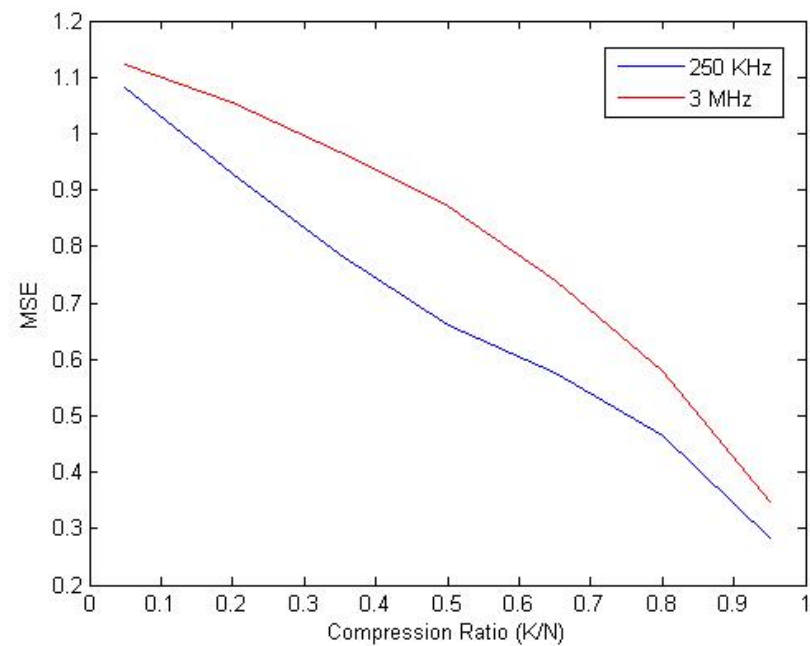
CS on test bed data



CS on test bed data

□ MSE Performance:

$$\text{MSE} = E \left\{ \frac{\|\hat{s} - s\|_2^2}{\|s\|_2^2} \right\}$$



Conclusions and Future Directions

- ❑ Successfully transmitted and received the real-world communication signals from the SDR test-bed.
- ❑ Successful implementation of CS algorithm on the test-bed data with acceptable performance.
- ❑ Developing an intelligent jammer detection algorithm for wideband radios using CS techniques.

Acknowledgements

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Thank you

Q & A