## Sub-Nyquist Sampling of Wideband Signals

#### **Deborah Cohen**

Technion – Israel Institute of Technology

Sub-Nyquist Sampling (Xampling) – Smart Sampling Seminar

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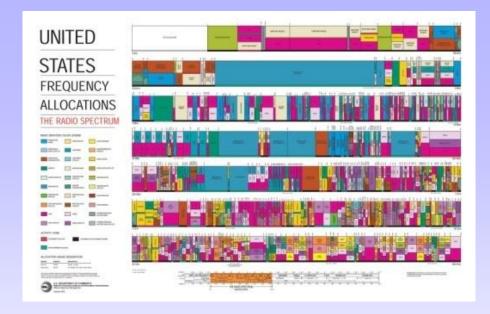
# Outline

Motivation
Algorithms

Sampling: MWC and Multicoset
Recovery

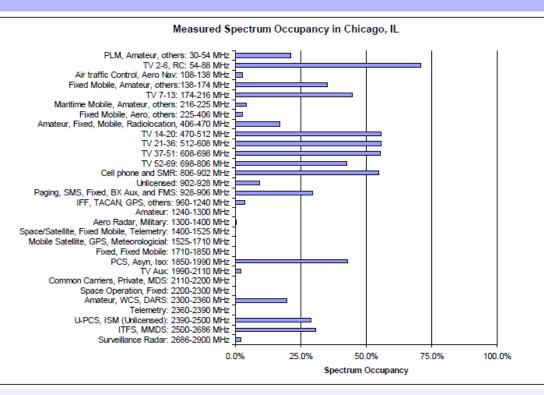
Challenges and Trade-Offs
Treatment of Noise

# **Spectrum Saturation**



- Licensed frequency bands to Primary Users (PUs): TV, radio stations, mobile carriers, air traffic control...)
- Spectrum is too crowded
- Cannot allocate frequency bands to new users!

# **Spectrum Sparsity**



Spectrum is underutilized

 In a given place, at a given time, only a small number of PUs transmit concurrently

Shared Spectrum Company (SSC) - 16-18 Nov 2005

Can we exploit temporarily available spectrum holes for opportunistic transmissions?

# **Cognitive Radios**

## Principle:

- Perform spectrum sensing to search for available spectrum holes
- Monitor spectrum during transmission to detect any change in PUs' activity

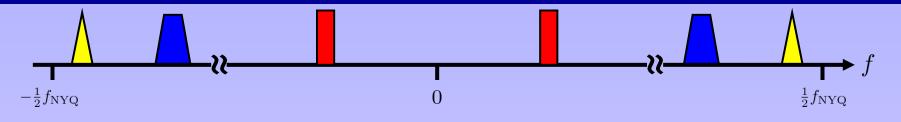
## Requirements:

- Wideband spectrum sensing
- Real-time
- Reliability
- Minimal hardware and software resources (mobile)

Nyquist sampling is not an option!

How do we efficiently perform detection on a wideband signal?

## Model

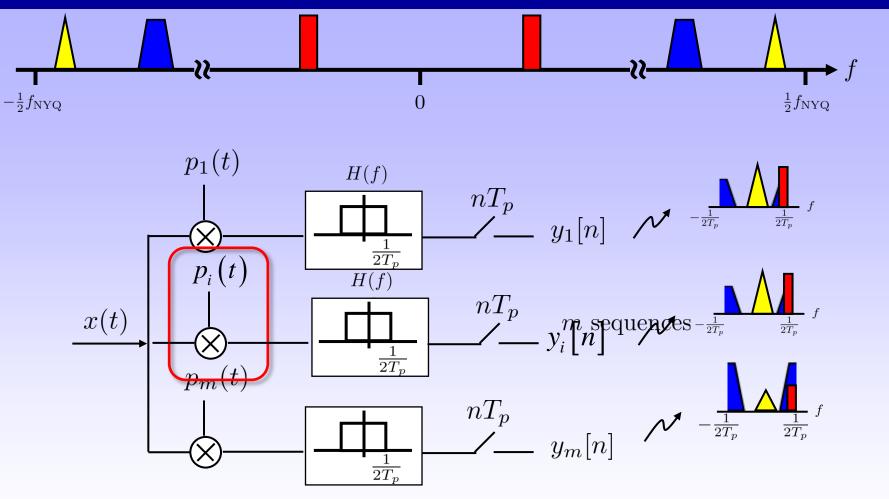


Multiband model:
 N – max number of transmissions
 B – max bandwidth of each transmission

Goal: blind detection

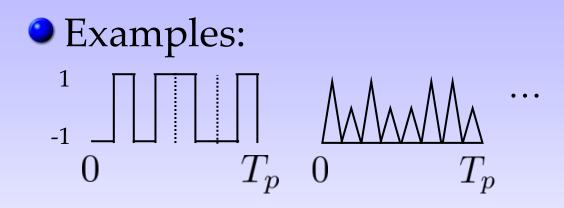
Minimal achievable rate: 2NB << f<sub>NYQ</sub>

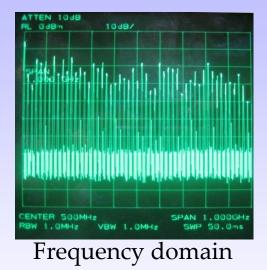
# The Modulated Wideband Converter (MWC) Mishali & Eldar '10



# MWC – Mixing & Aliasing

## • Mixing function $p_i(t)$ periodic with period $T_p$



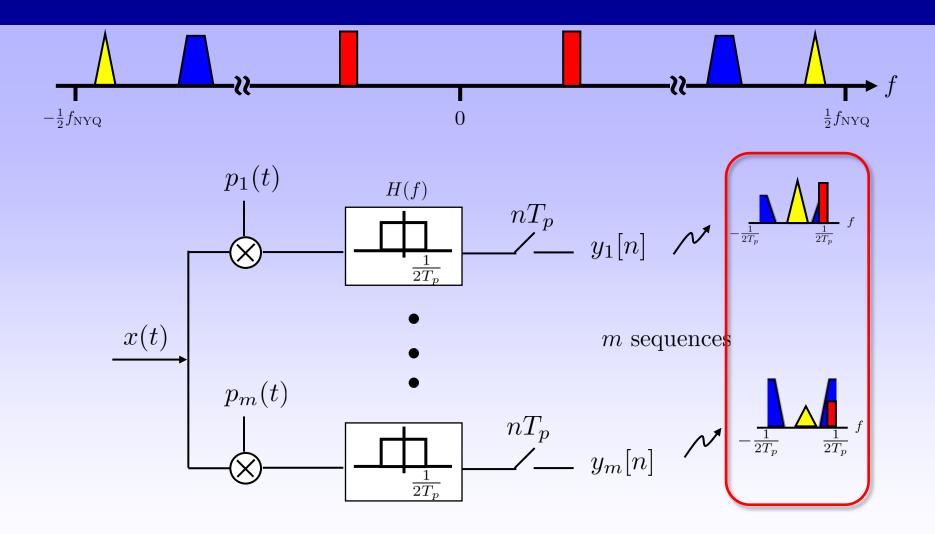


## • Practical considerations:

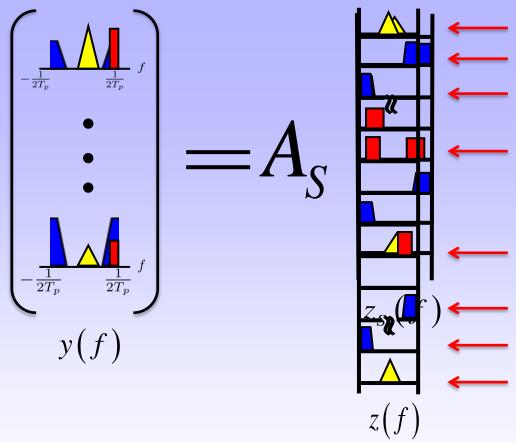
Can't design nice sign patterns at high frequency
 Only marie disity and frequency are eather as a method.

Only periodicity and frequency smoothness matter

## MWC – Aliasing



## **MWC – Recovery**



- Support S recovery
- Signal reconstruction:  $z_s(f) = A_s^{\dagger} y(f)$

# MWC – Support Recovery (CTF)

**Problem: infinite number of linear systems (f is continuous)** 

- Solve in the time domain for each n:  $\mathbf{y}(f) = \mathbf{Az}(f) \iff \mathbf{y}[n] = \mathbf{Az}[n]$
- Time consuming
  Not robust to noise
  CTF (Continuous To Finite):

$$\begin{array}{c} \mathbf{y}[n] \\ \bullet \\ \mathbf{Q} = \sum \mathbf{y}[n] \mathbf{y}^{H}[n] \\ \bullet \\ \mathbf{Q} = \mathbf{V} \mathbf{V}^{H} \\ \bullet \\ \mathbf{V} = \mathbf{A} \mathbf{U} \\ \bullet \\ \mathbf{S} = \mathrm{supp}(\bar{\mathbf{U}}) \\ \bullet \\ \bullet \\ \mathbf{frame \ construction} \\ \end{array}$$

Infinite problem (IMV)  $\rightarrow$  One finite-dimensional problem

# MWC – Single Channel

m channels at rate 
$$f_s \rightarrow 1$$
 channel at rate  $mf_s$ 

# • A system with $f_s = qf_p$ provides q equations for each physical channel

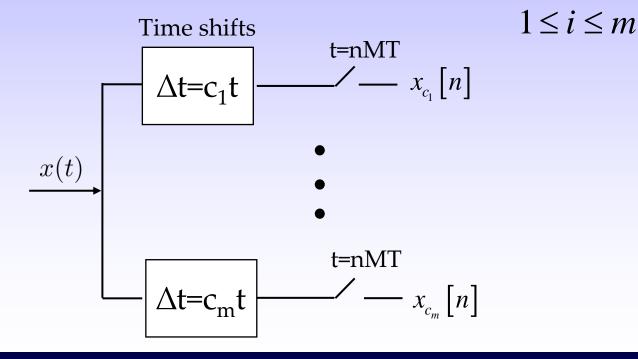
## Trade-off:

- Fewer channels: big hardware savings
- Increased rate in each channel

### Alternative: Multicoset Sampling Mishali & Eldar '09

Selection of certain samples from the Nyquist grid at rate  $f_s = \frac{1}{MT}$ :

$$x_{c_i}[n] = x(nMT + c_iT), \qquad 0 \le c_i \le M - 1$$



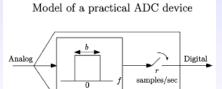
# Multicoset vs. MWC

## Same...

- Minimal sampling rate
- Relation between samples and original signal
- Reconstruction scheme

## ... But Different

- **×** Difficult to maintain accurate time shifts
- × Practical ADCs distort the samples

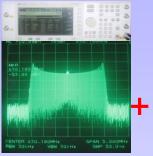


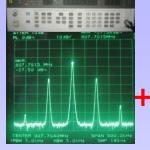
Easier to implement – less hardware
Solve digital bottleneck in case of low bandwidth

# **Sub-Nyquist Demonstration**

Mishali & Eldar, '10

#### Carrier frequencies are chosen to create overlayed aliasing at baseband





FM @ 631.2 MHz

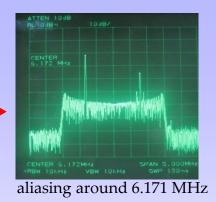
AM @ 807.8 MHz

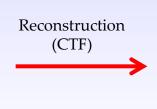


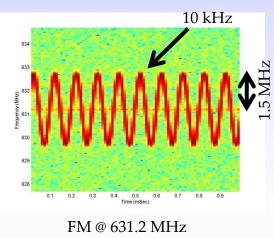
Sine @ 981.9 MHz

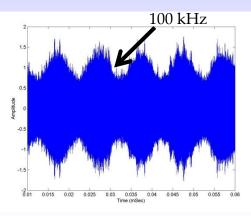


MWC prototype









AM @ 807.8 MHz

## But...

Joint work with Cores, UCLA

### **Problem: High sensitivity to noise**

Energy detection fails in low SNR regimes

**Solution:** New detection scheme

• Using a property of communication signals that is not exhibited by noise S(f)

# Cyclostationarity

## Definition:

Process whose statistical characteristics vary periodically with time

## Example:

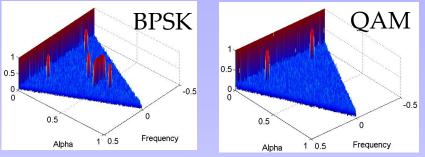
Communication signals

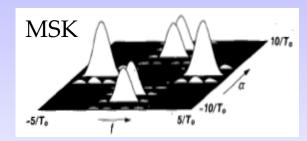
## Characterization:

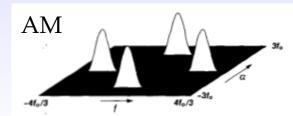
- Spectral correlation function (SCF)
- Exhibits spectral peaks at certain frequency locations called cycle frequencies

# **SCF – Examples**

(Gardner)



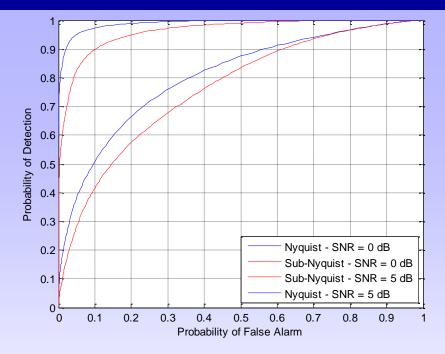




Modulation	Peaks at $(\alpha, f)$
BPSK	$\left(rac{1}{T},f_{c} ight),\left(2f_{c},0 ight),\left(2f_{c}\pmrac{1}{T},0 ight)$
MSK	$\left(rac{1}{T},f_{c} ight),\left(2f_{c}\pmrac{1}{2T},0 ight)$
QAM	$\left(rac{1}{T},f_{c} ight)$
AM	$(2f_c,0)$

## Results

#### Cohen, Rebeiz et. Al, '11



	Nyquist	Sub-Nyquist
Sampling rate	$f_{nyq} = 10GHz$	$m \cdot f_s = 30 \cdot 12MHz = 360MHz$

We can perform recovery from MWC samples in low SNR regimes using cyclostationary detection

# Conclusions

- Cognitive radios: solve the spectrum congestion issue
- Crucial task: wideband analog spectrum sensing
- Sensing mechanism: low-rate, quick, efficient and reliable
- Robustness to noise: exploit communication signals cyclostationarity

## References

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