

# SNR Walls for Cognitive Radio Spectrum Sensing

Anant Sahai

presenting joint work with students:

Danijela Cabric    Rahul Tandra

BWRC and Wireless Foundations Center  
U.C. Berkeley

**Major support from the National Science Foundation**

*I<sup>2</sup>R* Seminar: 8th January, 2008

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- Error correcting codes: already approaching Shannon limits
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- Increase spectrum utilization
  - ▶ Colonize new spectrum: very promising work on 60GHz radios
  - ▶ **Make better use of the spectrum we have by exploiting frequency agile devices.**

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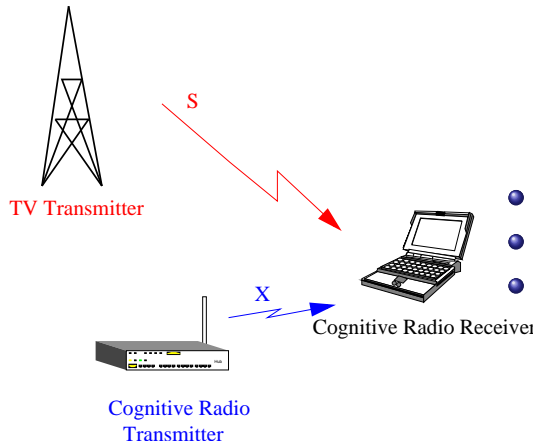
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- **Can't have everything!**

# Outline

- Introduction
- **Why is robust sensing important?**
- Noncoherent
- Coherent
- Cyclostationary
- Capacity/Robustness tradeoff

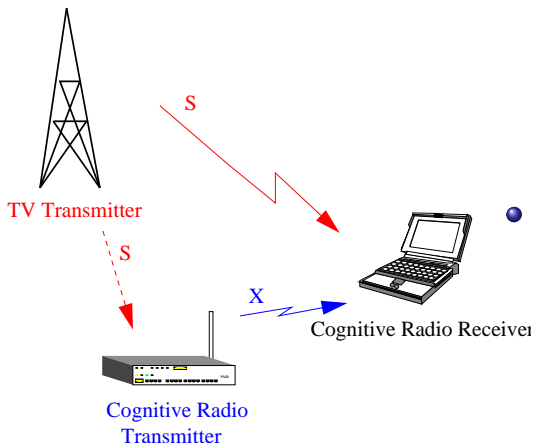
# Interference due to the primary



$$Y = X + S + Z$$

- Signal  $X$  has power  $P$
- Interference  $S$  has power  $Q$
- Noise  $Z$  has power  $N$

# Assume known at transmitter



$$Y = X + S + Z$$

- We get advance warning of the interference.

# “Canceling” known interference

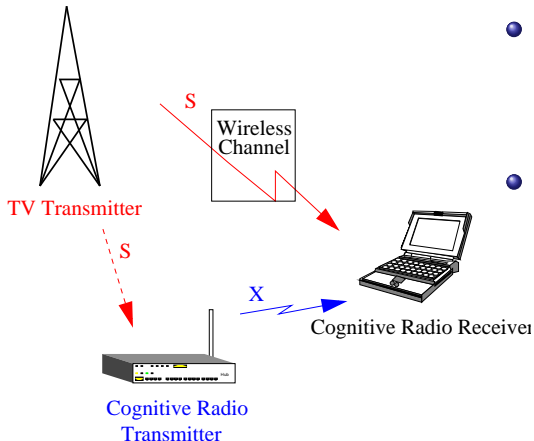
## Dirty-paper coding: Costa'83

Capacity of (complex) AWGN channel with interference known *a priori* is same as that with **no** interference

$$C = \log \left( 1 + \frac{P}{N} \right)$$

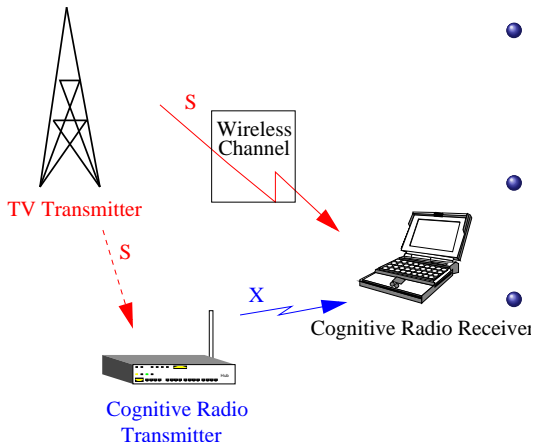
- Explored by *Tarokh et al* and *Viswanath et al* :  
If primary interference is known *a priori*, achievable secondary rates are **higher**

# Is the interference *really* known?



- Primary's transmissions are different from the **interference** caused by primary's transmissions
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# Is the interference *really* known?



- Primary's transmissions are different from the **interference** caused by primary's transmissions
- **Wireless fading channel** connects the primary transmitter to the secondary receiver.
- Would you bet your rate on knowing the phase of the channel?

# Simplify: just phase uncertainty

$$Y = X + Se^{j\theta} + Z$$

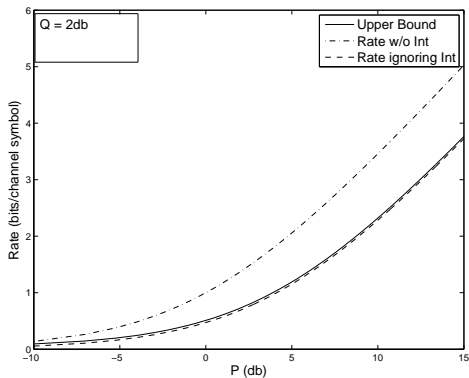
- $\theta$  is same for the whole block
- $\theta$  is unknown at the transmitter

## Maximum Achievable Rate

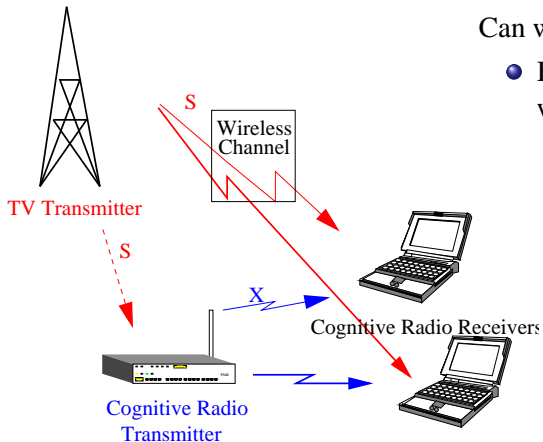
What is the maximum rate for reliable communication for unknown  $\theta$  but known  $S$  at the transmitter?

# Our upper bound for phase uncertainty

$$R \leq \frac{1}{2} \log \left[ \frac{(P + Q + N)^2}{4QN} \right]$$



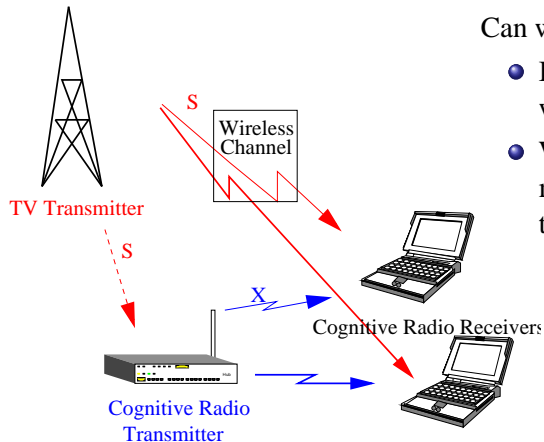
# Incidental architectural implications



Can we do “network-coding?”

- Based on broadcast nature of wireless.

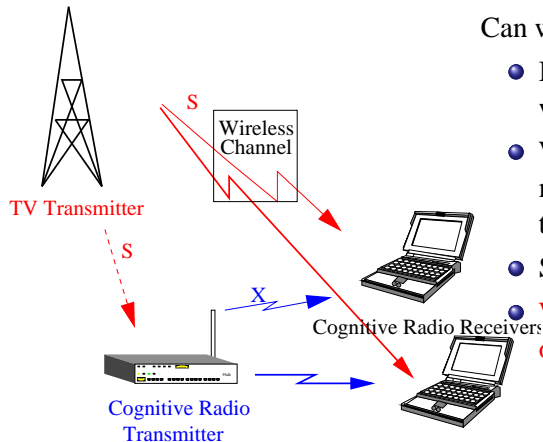
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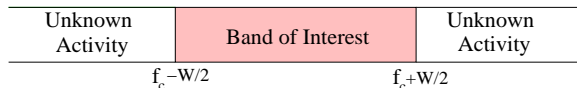
- Based on broadcast nature of wireless.
- With feedback, the two phases may be estimated at the transmitter.
- Still two phases = rate-loss!
- **Writing on interference is not omnidirectional!**

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# Sensing the primary's presence

Spectrum picture



- Look for the primary in the ‘band of interest’
- Within band model:
  - ▶ Unknown signal:  $X(t)$
  - ▶ Independent white noise:  $W(t)$
- Idealization: sample the band of interest at Nyquist rate

# Simplest detector: energy detector

- Received energy used for detection
- Test statistic and decision rule:

$$\sum_{n=1}^N Y^2[n] \underset{\mathcal{H}_0}{\overset{\mathcal{H}_1}{\geq}} \gamma(N)$$

- Sample complexity:

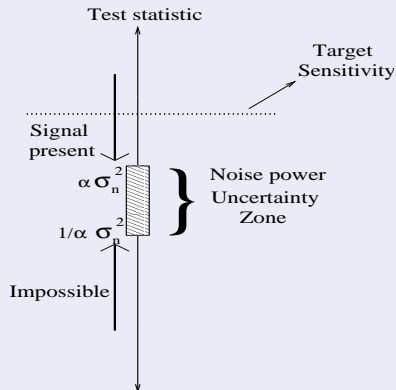
$$N \approx 2 \left[ Q^{-1}(P_{FA}) - Q^{-1}(P_D) \right]^2 SNR^{-2}$$

# Impact of uncertainty: energy detector

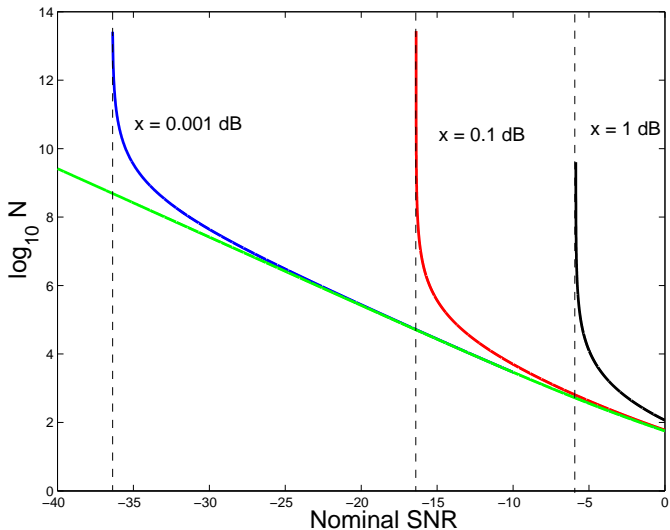
- Actual noise power,  $\sigma_a^2 \in [\frac{1}{\alpha}\sigma_n^2, \alpha\sigma_n^2]$
- If

$$P + \sigma_a^2 \leq \alpha\sigma_n^2$$
$$\Rightarrow P \leq \frac{\alpha^2 - 1}{\alpha}\sigma_n^2$$

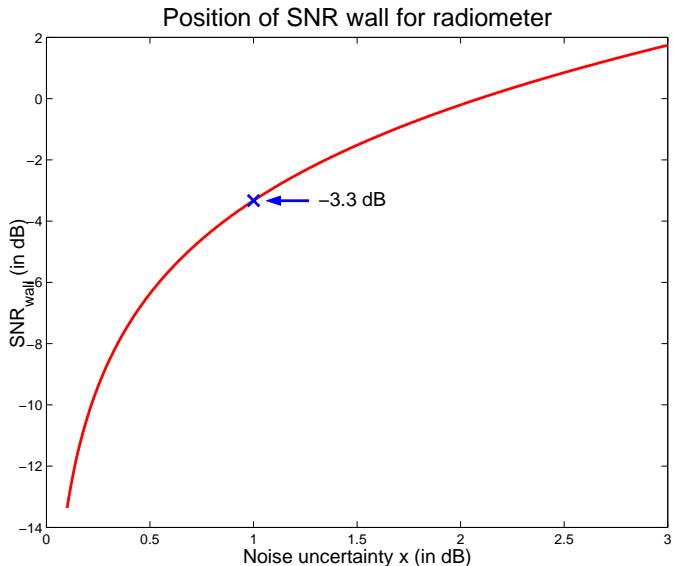
Energy detector fails to detect the signal



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- Since  $SNR \ll 1$ , use Taylor series approximation
- First order terms of the observation  $\mathbf{y}$  vanish for a zero-mean constellation
- Only quadratic terms of  $\mathbf{y}$  remain

# “Noise” is uncertain in distribution

- Noise is usually assumed to be Gaussian
- Sources of uncertainty:

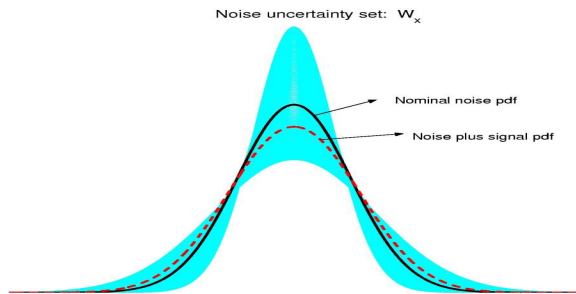
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- Sources of uncertainty:
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  - ▶ **Interference from other transmissions**
  - ▶ Nonlinear mixing from other bands

# Realistic model for noise uncertainty

- Receiver knows the noise distribution only up to an uncertainty set  $\mathcal{W}_x$ .

## Realistic model

- Noise cloud includes a range of Gaussians with variance  $\in [\frac{1}{\alpha}\sigma_n^2, \alpha\sigma_n^2]$  as well as other similar distributions.
- $W_a \in \widetilde{\mathcal{W}}_x$  iff

$$\mathbb{E}W_a^{2k} \in \left[ \frac{1}{\alpha^k} \mathbb{E}W_n^{2k}, \alpha^k \mathbb{E}W_n^{2k} \right], \quad \alpha = 10^{x/10}$$

- Implication: Energy detector like wall for all detectors

## Theorem

Consider detecting a weak BPSK signal with the noise distribution lying in  $\widetilde{\mathcal{W}}_x$ . Under this model, there exists an absolute SNR wall ( $snr_{wall}^*$ ) for any possible robust detector.

$$snr_{wall}^* = \min_{k>0} snr_{wall}^{(2k)} = \alpha - 1$$

# Energy detector summary

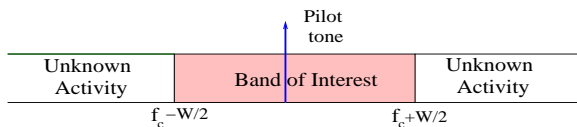
- Noise uncertainty is always present
  - ▶ At least about 1-2 dB of device level uncertainty
    - ★ Cannot see 3 dB below the noise
  - ▶ Easily 10-30 dB of interference level uncertainty
    - ★ Cannot see below it at all!
- E.g.: In a 6MHz TV band, Digital TV receiver sensitivity = -85dBm.
  - ▶ We must have -116dBm sensitivity to deal with rare fading.
  - ▶ Thermal Noise alone at -106dBm !
  - ▶ With interference could be -91dBm !
- What can we do to mitigate this?

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# Coherent detection: maximally detailed modeling

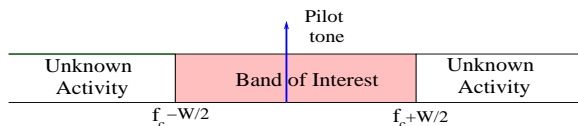
## Spectrum picture



- Look for the primary pilot in the ‘band of interest’

# Coherent detection: maximally detailed modeling

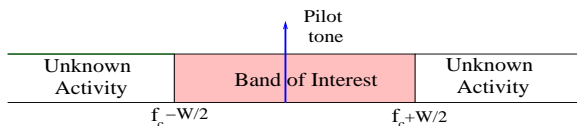
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  - ▶ Denial pilot (very critical)
  - ▶ Permissive pilot (non-critical)
- Within band model:
  - ▶ Unknown component,  $X(t)$
  - ▶ Pilot signal,  $X_p(t)$
  - ▶ Independent white noise,  $W(t)$

# Hypothesis testing model: coherent detection

- Distinguish between the following hypotheses:

$$\mathcal{H}_0 : Y[n] = W[n]$$

$$\mathcal{H}_1 : Y[n] = W[n] + \sqrt{(1-\theta)}X[n] + \sqrt{\theta}X_p[n]$$

- Basic assumptions:

- ▶ Signal samples  $X[n]$ 's are white or orthogonal to the pilot
- ▶ Noise samples  $W[n]$ 's are white
- ▶  $X_p[n]$  is a known pilot tone
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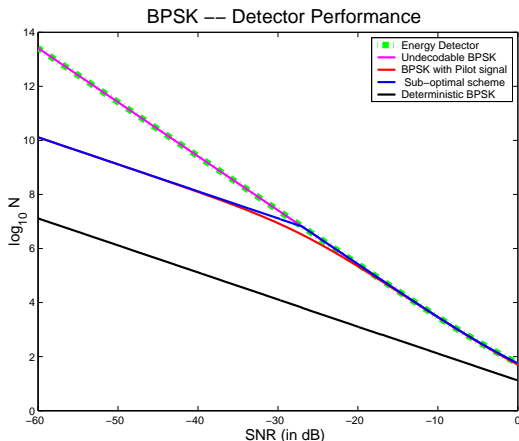
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- Main idea: Coherent processing gain can overcome noise level uncertainty

$$T(\mathbf{y}) = \frac{1}{N} \sum_{n=1}^N Y[n] \hat{X}_p[n] \underset{\mathcal{H}_0}{\overset{\mathcal{H}_1}{\gtrless}} \gamma(\sigma^2)$$

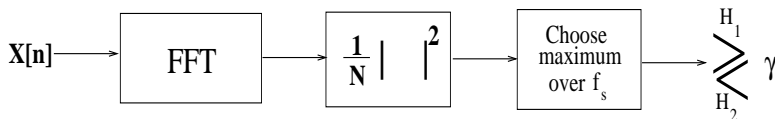
# Matched filter performance

- $X[n]$  are BPSK modulated
  - ▶  $X[n] \sim \text{Bernoulli}(\frac{1}{2})$ , taking values in  $\{\sqrt{P}, -\sqrt{P}\}$
- $\theta = 0.01$ , *ie.* 1% energy in the pilot

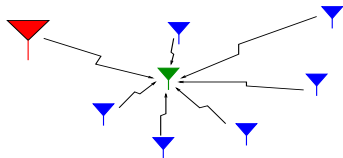


# Easy uncertainties

- Unknown but steady frequency and time offsets
- Effect on pilot:
  - ▶  $X_p[n] = A \cos(2\pi f_s n + \phi)$
  - ▶  $f_s$  and  $\phi$  unknown
  - ▶ Phase offset is easy to deal with
- Approach: Search in many bins
- Implementation and computational complexity
  - ▶ Need to compute an  $N$ -point FFT
  - ▶ Search for the maximum over  $f_s$
  - ▶ Computationally more involved than the energy detector



## Noise uncertainty: closer look

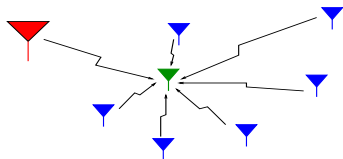


- Noise = receiver/background noise + secondary interference,  
 $\sigma^2 = \sigma_0^2 + \sigma_i^2$

### Effect of unknown interference

- Interference is uncertain
- Dominating term in noise uncertainty

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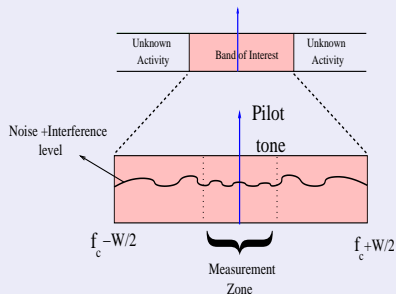
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## Proposed remedy

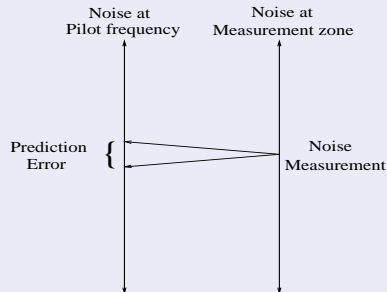
- Robustly estimate  $\sigma_i^2$
- Significant reduction in noise uncertainty

# Setting threshold by learning

## In-band measurement



## Noise calibration error



### • Sources of calibration error:

- ▶ Long-term frequency selectivity in secondary signals
- ▶ “Coherence bandwidth/time” for secondary signals

# Fundamentally challenging uncertainty: fast fading

- Assume *worst-case channel coherence time*:  $T_c$ 
  - ▶ Fading is assumed constant during this coherence time
  - ▶ Matched filter can be applied in each coherent block
  - ▶ Test statistic:

$$T(\mathbf{y}) = \frac{1}{N} \sum_{n=0}^{N-1} \left[ \frac{1}{\sqrt{N_c}} \sum_{k=1}^{N_c} Y[n] \hat{X}_p[nN_c + k] \right]^2$$

- ▶  $N_c$ : Length of single coherence block
- Reduced to the energy detector case,
  - ▶ Processing gain due to the coherence time  $T_c$
  - ▶ Less interference uncertainty.

# Summary: finite gains from coherent processing

- Primary coherence time:  $[100 \mu\text{s}, 10 \text{ ms}]$   
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- Effective *SNR* wall:  $[10, 50] \text{ dB}$  lower  
Dwell times:  $[0, 30] \text{ dB}$  better
- System could be ‘Wall limited’ or ‘Dwell limited’

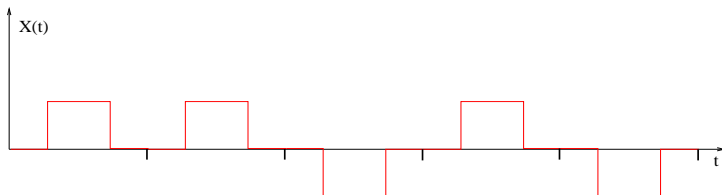
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# Refined model: Cyclostationary signals

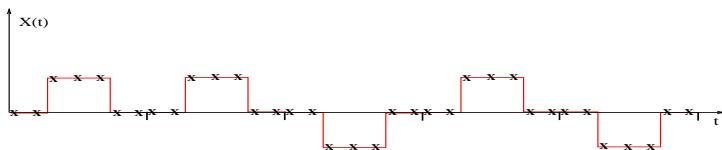
William Gardner

- Pulse amplitude modulated signal:
  - ▶ BPSK modulated over a rectangular pulse shape



- Nyquist sampling — Stationary signal
- Oversampling — Cyclostationary signal
  - ▶ Periodicity corresponding to data rate
  - ▶ Feature detectors
    - ★ Sample complexity same as energy detector —  $O(SNR^{-2})$

# Spectral Correlation Function



- Spectral correlation function can be estimated by

$$\tilde{S}_X^\alpha = \lim_{t \rightarrow \infty} \lim_{T \rightarrow \infty} \frac{1}{t} \frac{1}{T} \int_{-\frac{t}{2}}^{\frac{t}{2}} X_T \left( t, f + \frac{\alpha}{2} \right) X_T^* \left( t, f - \frac{\alpha}{2} \right) dt$$

where

$$X_T(t, f) = \int_{t-\frac{T}{2}}^{t+\frac{T}{2}} x(t) e^{-j2\pi fu} du$$

is the spectral component of  $x(t)$  at frequency  $f$  with bandwidth  $\frac{1}{T}$

- $S_x^0$  is the power spectral density of  $x(t)$
- $S_x^\alpha = 0$  for a WSS process for  $\alpha \neq 0$

# Single-cycle feature detector

- Estimate Spectral Correlation Function:  $\tilde{S}_Y^\alpha(f)$
- Compute the following test-statistic

$$T(Y) = \int \tilde{S}_Y^\alpha(f) S_X^\alpha(f)^* df$$

## Under Hypothesis $\mathcal{H}_0$

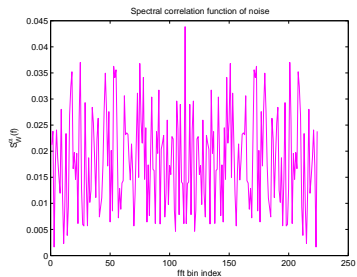
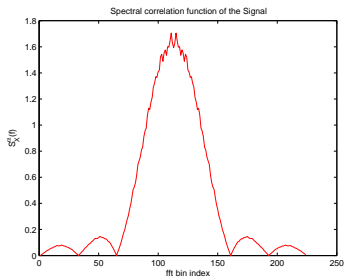
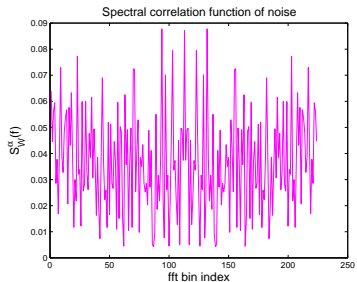
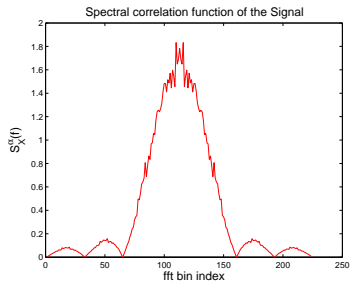
$$\begin{aligned} T(Y|\mathcal{H}_0) &= \int \tilde{S}_W^\alpha(f) S_X^\alpha(f)^* df \\ &\rightarrow \int S_W^\alpha(f) S_X^\alpha(f)^* df \\ &= \int 0 \cdot S_X^\alpha(f)^* df = 0 \end{aligned}$$

## Under Hypothesis $\mathcal{H}_1$

$$\begin{aligned} T(Y|\mathcal{H}_1) &= \int [\tilde{S}_X^\alpha(f) + \tilde{S}_W^\alpha(f)] S_X^\alpha(f)^* df \\ &\rightarrow \int [S_X^\alpha(f) + S_W^\alpha(f)] S_X^\alpha(f)^* df \\ &= \int |S_X^\alpha(f)|^2 df \end{aligned}$$

- Feature detection works at every SNR, given sufficiently large  $N$ !

# BPSK example



# Robust to flat-fading

- Flat fading does not change the pulse-shape, just the amplitude.

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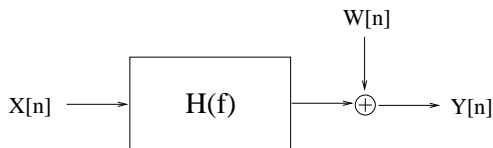
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- Amplitude is changing due to data modulation anyway.
- **More robust** than even coherent detection in this regard!

# Impact of frequency-selective fading

- Need to incorporate frequency-selective fading into channel model

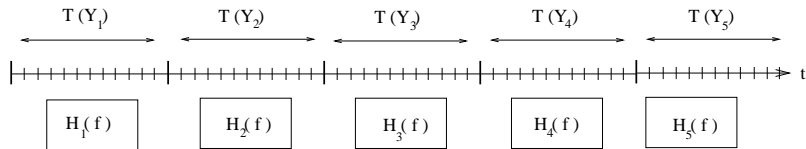


- Unknown channel filter  $H(f)$
- Spectral Correlation Function

$$S_Y^\alpha(f) = H\left(f - \frac{\alpha}{2}\right) H^*\left(f + \frac{\alpha}{2}\right) S_X^\alpha(f) + S_W^\alpha(f)$$

- Fading smears the features
- **Can kill the feature completely in the worst case!!!**

# Fundamental limitation: “delay” coherence time

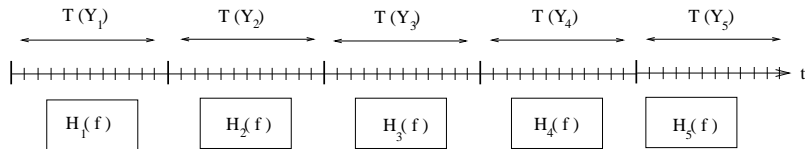


- In reality, fading response varies with time.
- Scale of variation – channel ‘delay-coherence’ time
- Single cycle feature detector:

$$\begin{aligned} T(Y) &= \int \tilde{S}_Y^\alpha(f) S_X^\alpha(f)^* df \\ &= \frac{1}{M} \sum_{i=1}^M T(Y_i) \\ &\rightarrow 0 \end{aligned}$$

- Averaging over delay coherence times kills features!

# Fundamental limitation: “delay” coherence time



- Modified test-statistic:

$$T(Y) = \frac{1}{M} \sum_{i=1}^M |T(Y_i)|^2$$

- Key observation:

- ▶  $\mathbb{E}[T(Y)|\mathcal{H}_0] \neq 0$

- Consequence:

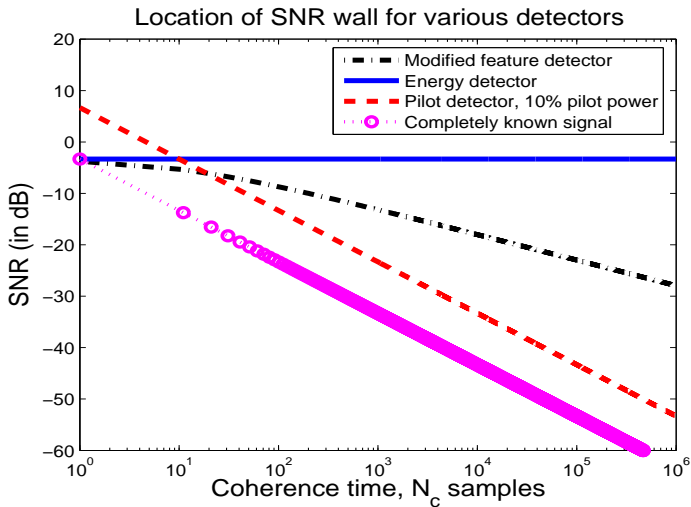
- ▶  $\mathbb{E}[T(Y)|\mathcal{H}_1] - \mathbb{E}[T(Y)|\mathcal{H}_0] \leq \Delta_{\mathbb{E}[T(Y)|\mathcal{H}_0]}$

- ▶  $\Rightarrow$  SNR wall

# What about noise calibration?

- Can learn the overall noise/interference level.
- Like the coherent case, the true wall comes from unknown color.

# Detector robustness with coherence time



# Outline

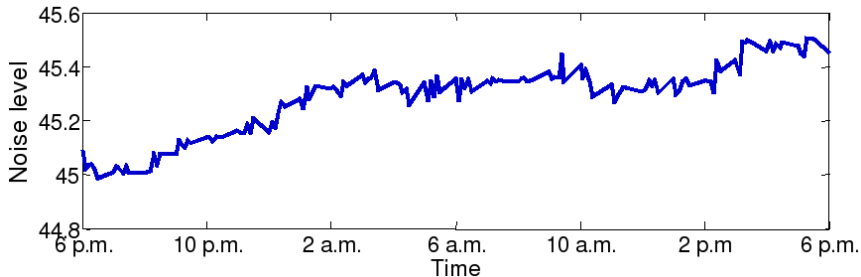
- Introduction
- Why is robust sensing important?
- Noncoherent
- Coherent
- Cyclostationary
- **Capacity/Robustness tradeoff**

# Are we just being paranoid?

- Experimental validation
- Used BEE2 and 2.4 GHz radio front-ends

# Are we just being paranoid?

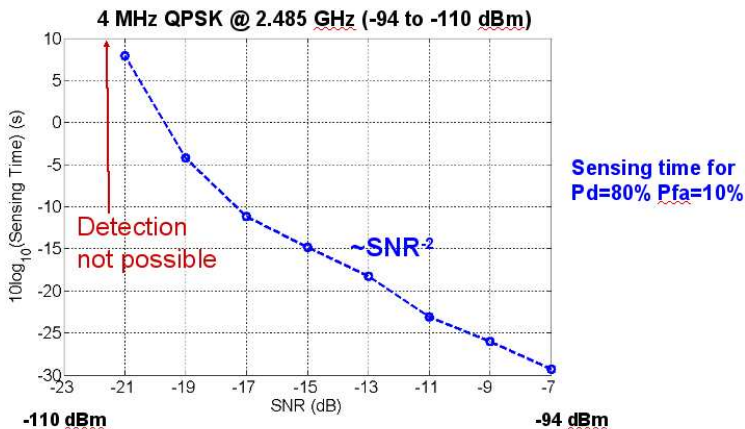
- Experimental validation
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Noise levels move around over a day.

# Are we just being paranoid?

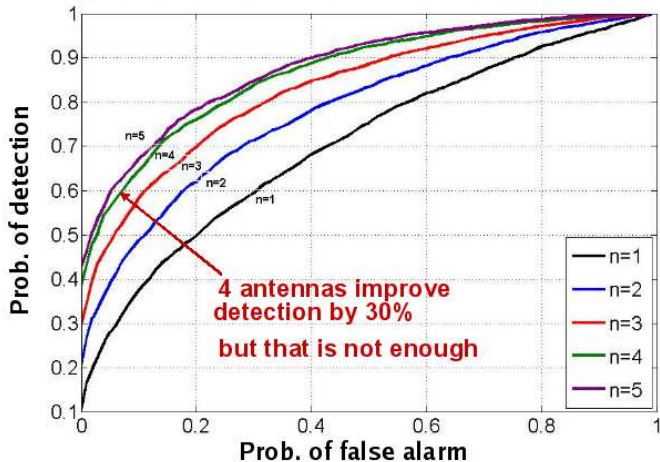
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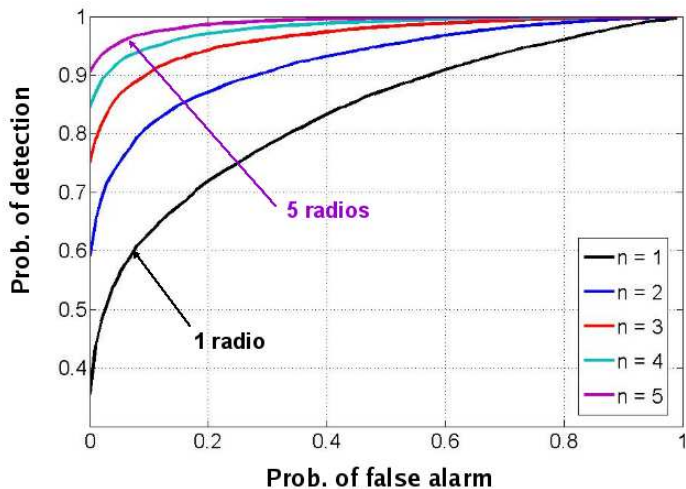
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## Single radio with 1- 5 antennas in fading environment



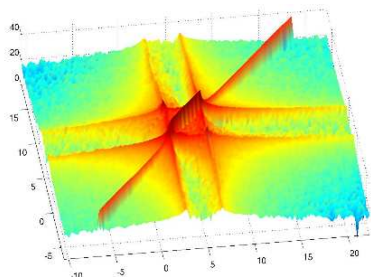
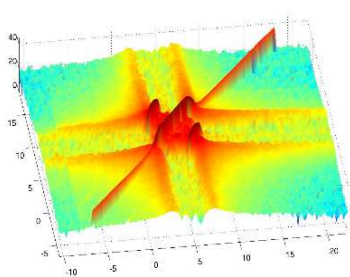
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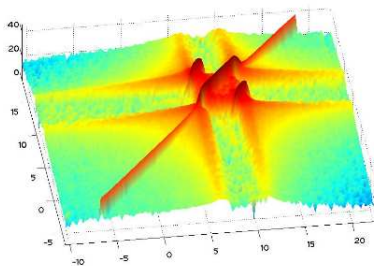


The spectral correlation function shows spectral redundancy in a transformed domain.

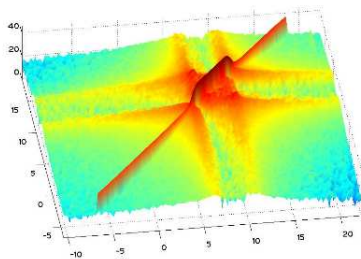
# Are we just being paranoid?

- Experimental validation
- Used BEE2 and 2.4 GHz radio front-ends

16000 spectral averages  
and perfect sampling



16000 spectral averages  
and 100 Hz sampling offset (25 ppm)



But this redundancy is blurred away by fast fading.

# The core tension

- Flexibility of architecture: primary **must lose** to allow setting sensitivity for cognitive users.

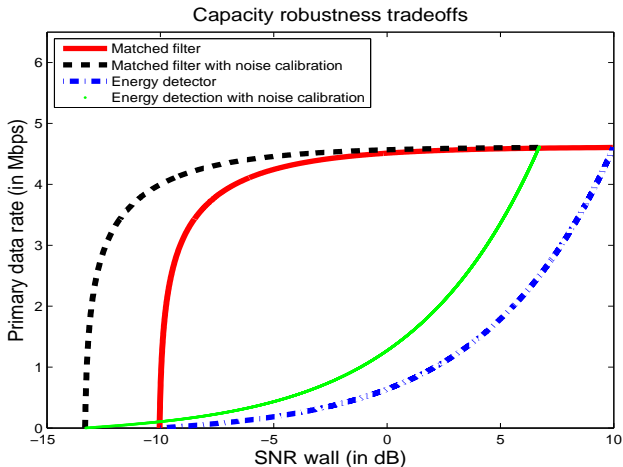
# The core tension

- Flexibility of architecture: primary **must lose** to allow setting sensitivity for cognitive users.
- Flexibility of deployment: the goal of sensing

# The core tension

- Flexibility of architecture: primary **must lose** to allow setting sensitivity for cognitive users.
- Flexibility of deployment: the goal of sensing
- Flexibility of use
  - ▶ How much can we assume about the primary modulation?
  - ▶ Do these “features” come at a price to the primary?

# Tradeoffs illustrated



# Tradeoffs illustrated

