



Ahmedabad
University

Spectrum Sensing in Cognitive Radio Using Cyclostationary Features

CSE400 - Section 1
Group 11

Project Overview

Problem Statement

- Detect presence of primary OFDM user
- Work under unknown noise variance
- Ensure controlled false alarm probability

Core Idea of Paper

- Use cyclostationary structure of OFDM
- Combine multiple cyclic frequencies
- Derive analytical false alarm expression

Milestone-2 Focus

- Identify and explain 4 core mathematical models
 - Analyze random variables and distributions
- Understand F-distribution derivation

Problem Framing: Spectrum Sensing Under Uncertainty

Primary Goal

- Detect presence of OFDM primary user

Mathematical Formulation

- Binary hypothesis testing
- $H_0 : x[n] = w[n]$
- $H_1 : x[n] = s[n] + w[n]$

Core Challenge

- Noise is random
 - Noise variance is unknown
 - Finite samples only
-
- ❖ $w[n]$ noise & $s[n]$ OFDM signal

Why a Probabilistic Detection Framework is Required?

Observation Under Uncertainty

- $x[n]$ is random
- $w[n]$ is random
- Noise variance σ^2 is unknown

Decision is Not Deterministic

- Cannot be 100% certain
- Risk of false alarm and missed detection

Therefore: Probabilistic Modeling Required

- Model distributions of test statistic
- Derive probability of false alarm
- Select threshold analytically

Motivation for Structural Detection

- Energy detector depends on σ^2
- Cyclostationary model reduces dependence

Mathematical Model 1: Signal Observation Model

Binary Hypothesis Testing Model

- $H_0 : x[n] = w[n]$
- $H_1 : x[n] = s[n] + w[n]$

Statistical Assumptions

- $w[n] \sim \text{Gaussian}(0, \sigma^2)$
- Samples independent
- σ^2 unknown

Random Variables Defined

- $w[n] \rightarrow$ Additive noise
- $s[n] \rightarrow$ OFDM signal
- $x[n] \rightarrow$ Observed sample

Purpose of This Model

- Formalizes detection problem
- Defines randomness source
- Foundation for all later models

Mathematical Model 2: Noise Statistical Model

Noise Statistical Characterization

- $w[n] \sim \mathcal{N}(0, \sigma^2)$

Random Variables Defined

- $w[n] \rightarrow$ additive noise sample
- $\sigma^2 \rightarrow$ noise variance (unknown parameter)

Statistical Assumptions

- Zero mean
- Independent across time
- Gaussian distribution
- σ^2 not perfectly known

Purpose of This Model

- Quantify randomness in received samples
- Identify main source of uncertainty
- Explain why energy-based detection becomes unreliable

Mathematical Model 3: Cyclostationary Signal Model

Cyclostationary Characterization of OFDM Signal

Time-varying autocorrelation:

$$\rightarrow R_y(t, \tau) = E[y(t) y(t + \tau)]$$

Cyclic autocorrelation at frequency α :

$$\rightarrow R_{y^\alpha}(\tau)$$

Random Variables Defined

$y(t) \rightarrow$ received signal process

$R_y(t, \tau) \rightarrow$ time-varying autocorrelation

$R_{y^\alpha}(\tau) \rightarrow$ cyclic autocorrelation component

Statistical Assumptions

- OFDM contains cyclic prefix
- Signal exhibits periodic statistical behavior
- Noise has zero cyclic autocorrelation for $\alpha \neq 0$

Purpose of This Model

- Capture structural property of OFDM
- Differentiate signal from noise using periodicity
- Provide feature independent of energy magnitude

Mathematical Model 4: Multi-Cycle Detection & Distribution Model

Multi-Cycle Test Statistic

- $T(mc) = (\text{Sum of cyclic feature energies}) / (\text{Reference energy})$

Random Variables Defined

- $Rx^a(\tau)$ → estimated cyclic autocorrelation
- $T(mc)$ → test statistic
- λ → detection threshold

Statistical Assumptions

- Multiple cyclic frequencies combined
- Numerator and denominator independent under H_0
- Large sample approximation
- Under H_0 : $T(mc)$ follows F-distribution
- $P_f = 1 / (\lambda + 1)$

Purpose of This Model

- Construct robust detection statistic
- Cancel unknown noise variance
- Derive analytical threshold
- Control probability of false alarm

Decision Logic and Randomness Flow

Final Decision Rule

- ❖ If $T(mc) \geq \lambda \rightarrow$ Decide H_1
- ❖ Else \rightarrow Decide H_0

Flow of Randomness

- ❖ $w[n] \rightarrow x[n] \rightarrow R_x^{\wedge} a(\tau) \rightarrow T_{mc} \rightarrow$ Decision

Core Challenge

- Probability of Detection (Pd)
- Probability of False Alarm (Pf)

Model-Implementation Alignment

How Theory Connects to Practice

- ❖ Observation model → Sample collection
- ❖ Cyclostationary model → Feature extraction
- ❖ Detection model → Statistic computation
- ❖ Distribution model → Threshold selection

Simulation Validation

- Matches analytical Pf
- Works across OFDM modulations
- Robust under noise uncertainty

Limitations and Milestone-2 Contribution

Current Assumptions

- Independent Gaussian noise
- Large sample approximation
- Ideal synchronization

Future Extensions

- Channel fading models
- Small-sample distribution analysis
- Hardware imperfections

Milestone-2 Contribution

- Clear identification of 4 core probabilistic models
- Structured random variable analysis
- Analytical understanding of F-distribution result
- Stronger theoretical clarity than Milestone 1

Thank you!