



**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  $\sigma^2$  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  $\sigma^2$
- Cyclostationary model reduces
- dependence

# Mathematical Model 1: Signal Observation Model

## Binary Hypothesis Testing Model

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

## Random Variables Defined

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

## Statistical Assumptions

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

## 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
- $\sigma^2$  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:

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

Cyclic autocorrelation at frequency  $\alpha$ :

→  $R_y^\alpha(\tau)$

## 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

- $R_x^a(\tau) \rightarrow$  estimated cyclic autocorrelation
- $T(mc) \rightarrow$  test statistic
- $\lambda \rightarrow$  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  $T(mc) \geq \lambda \rightarrow \text{Decide } H_1$
- ❖ Else  $\rightarrow \text{Decide } H_0$

## Flow of Randomness

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

## Core Challenge

- Probability of Detection ( $P_d$ )
- Probability of False Alarm ( $P_f$ )

# 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  $P_f$
- 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!