

# Binary Blind Identification of Wireless Transmission Technologies for Wide-band Spectrum Monitoring

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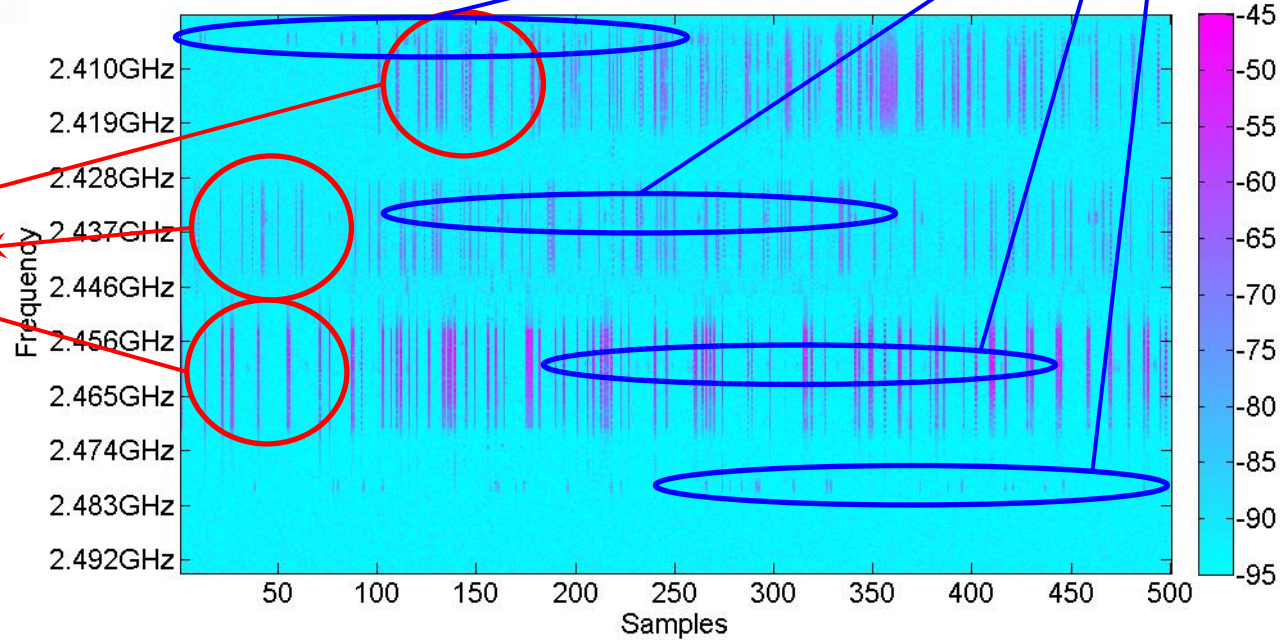
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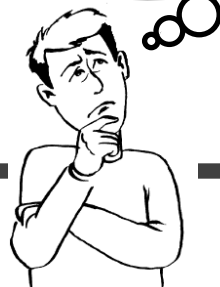


WiFi 1,6,11?



Some Zigbee devices?

How many different technologies are there?  
And what are they?



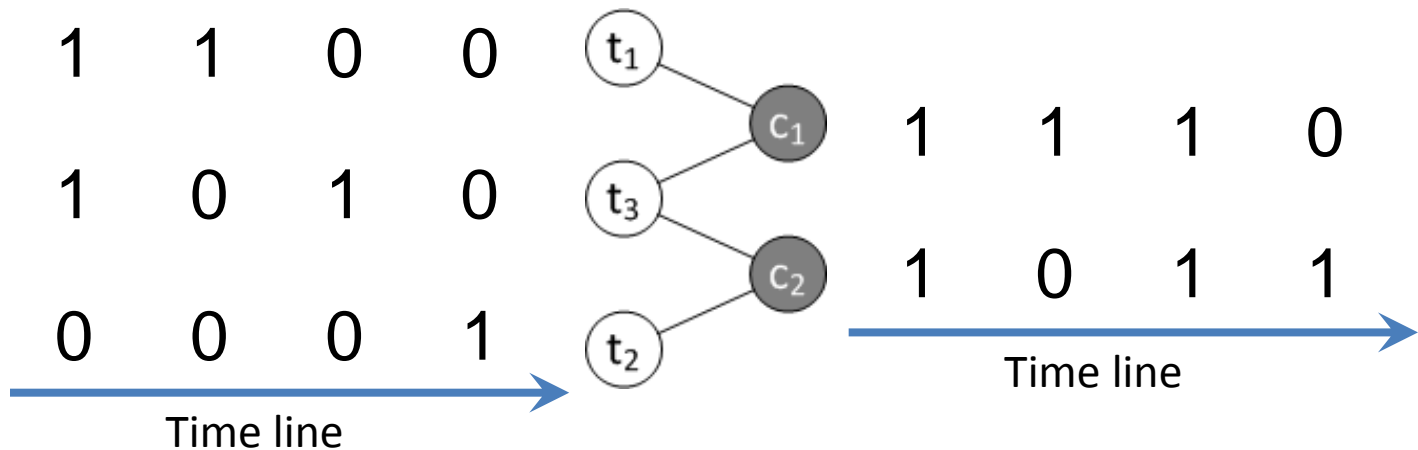
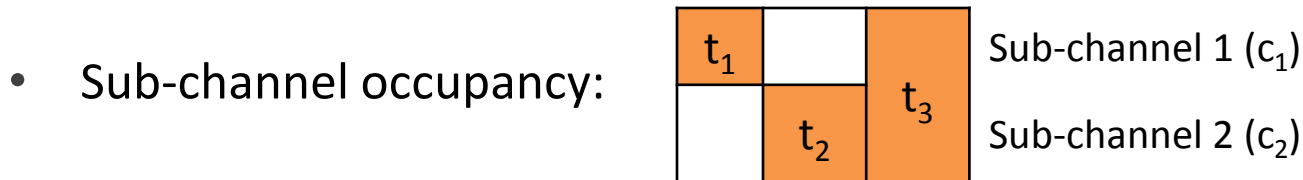
# Observations and Contributions

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- Technologies occupy different ranges of spectrum
- When a wireless device occupies a sub-channel, **adjacent sub-channels** are likely to be activated (correlated)
- **Blind identification**: no prior knowledge about the high level features wireless technologies, purely based on spectrum occupation
- Contributions:
  - Formulate the **blind technology identification** problem
  - Propose a **binary framework** to solve the problem

# Problem Binary Formulation

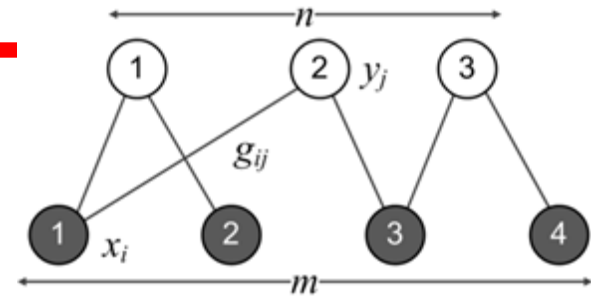
- Assuming the activities of devices using different technologies are **independent**
- A RF transmission will cause a **power surge** in its associated sub-channels
- Consider an example: 3 technologies operate on 2 sub-channels



$$Y \otimes G = X$$

(unknown)                      (unknown)

# Binary Independent Component Analysis (bICA)



- **Network model**: a bipartite graph  $\mathbf{G} = g_{ij}$   
 $n$  independent binary sources  $\mathbf{y} = [y_1, y_2, \dots, y_n]$   
 $m$  observable binary variables  $\mathbf{x} = [x_1, x_2, \dots, x_m]$
- Observations  $\mathbf{X}$  are **disjunctive mixtures** of latent sources  $\mathbf{Y}$   
$$x_i = \bigvee_{j=1}^n (g_{ij} \wedge y_j), i = 1, 2, \dots, m \quad \text{or} \quad \mathbf{X} = \mathbf{G} \otimes \mathbf{Y}$$
- **Problem**: given  $\mathbf{X}$ , infer the mixing matrix  $\mathbf{G}$  and the source  $\mathbf{Y}$
- Original ICA assumes continuous variables  $\rightarrow$  not applicable

# Proposed Procedure

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Spectrum  
monitoring



Spectrum  
occupancy  
inference



Sub-channel  
clustering



Inference and  
post-processing



Technology  
identification



Wireless  
technologies

1. **Spectrum occupancy inference**: remove noise and identify useful signals
2. **Sub-channel clustering**: cluster similar sub-channels to reduce inference complexity
3. **Inference and post processing**: use bICA to infer the channel occupancy matrix
4. **Technology identification**: from the center frequencies and bandwidths

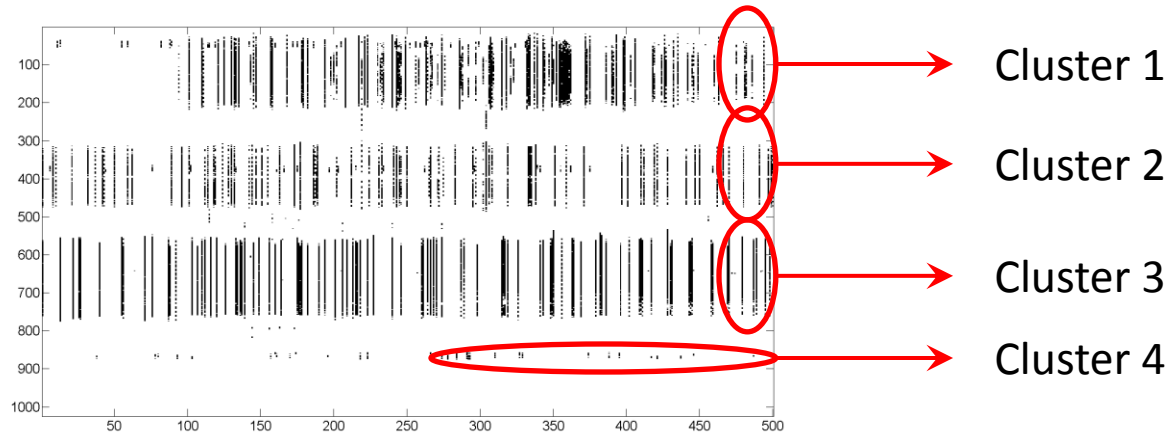
# Spectrum Occupancy Inference

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- Separate useful signals from noise
- Mean Shift (MS) clustering method
- Noise and useful signals form different clusters
- Two-step procedure:
  - Cluster the spectrum power by applying FFT to the measurement data, determine the **noise floor**
  - Apply MS on 2-d data to determine all **cluster means**
- Clusters with means  $>$  noise floor are useful clusters

# Sub-Channel Clustering

- Reducing computation complexity of bICA
- **Observation:** wireless technologies tend to occupy contiguous sub-channels



→ Cluster similar, contiguous sub-channels using Girvan-Newman community detection algorithm



# Wireless Technology Inference

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- Use bICA to infer the independent technologies occupying sub-channels
- Un-cluster the inferred result to obtain the original channel occupancy matrix ( $\mathbf{G}$ )
- From  $\mathbf{G}$ , determine the center frequency and bandwidth of each group
- Identify the associated wireless transmission technologies

# Evaluation Setup

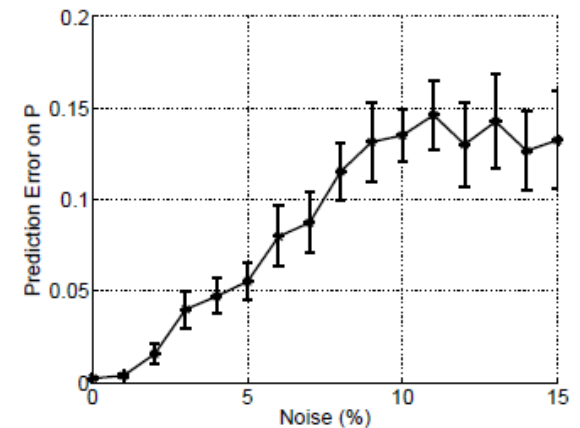
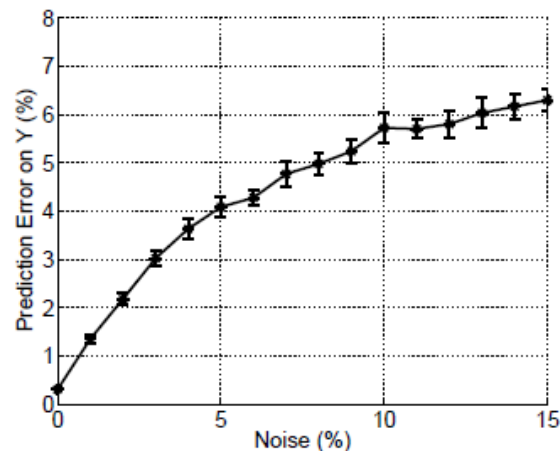
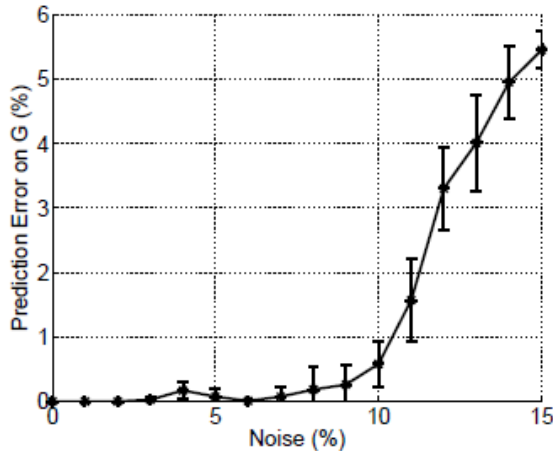
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- Synthetic trace
  - 3 WiFi devices on channels 1, 6, 11 and 8 ZigBee devices on channels 11 – 18
  - Device transmission prob. in  $[0.05, 0.1]$
  - Data noise ratio in  $[0, 0.1]$
- Real trace
  - 3 WLAN devices on channels 1, 6, 11 and 4 TmoteSky ZigBee devices on channels 11, 17, 22, 26
  - 1024 points sampling for each measurement
  - Measure each 10 secs, for 500 times

# Synthetic Trace Result

Matlab implementation

Channel noise = 0% – 15%

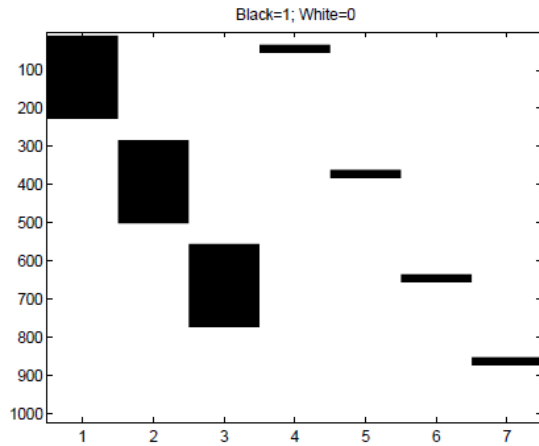


**Structure error ratio:**  
% inference error on  $G$

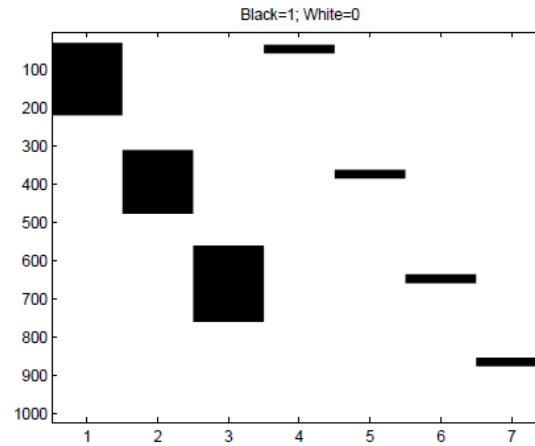
**Activity error ratio:**  
% inference error on  $Y$

**Transmission prob. error:**  
inference error on the active  
probability of technologies

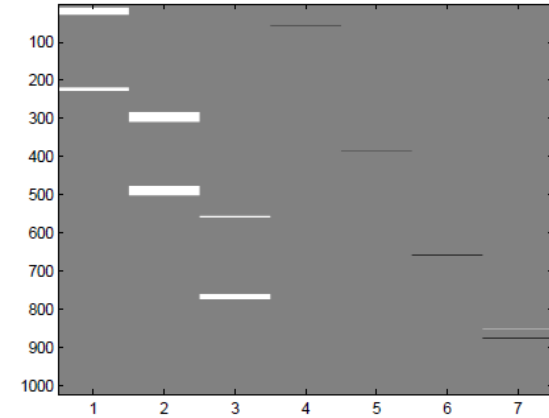
# Real Trace Result



(a) Original  $G$



(b) Inferred  $\hat{G}$  after *structure matching*



(c) Difference matrix

Component	$f_c$ (GHz) Inferred	$f_c$ (GHz) Ground truth	$b$ (MHz) Inferred	$b$ (MHz) Ground truth	Technology
1	2.4124	2.412	17.766	22	WiFi ch. 1
2	2.4371	2.437	15.604	22	WiFi ch. 6
3	2.4615	2.462	18.706	22	WiFi ch. 11
4	2.4052	2.405	2.068	2	ZigBee ch. 11
5	2.4352	2.435	2.068	2	ZigBee ch. 17
6	2.4603	2.460	2.162	2	ZigBee ch. 22
7	2.4803	2.480	2.068	2	ZigBee ch. 26

# Conclusion

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- Identifying transmission technologies **without prior knowledge** with only binary sensing is feasible
  - Frequency domain only
- What to do next?
  - Validation using large-scale spectrum data
  - Improve accuracy and computation efficiency of the proposed algorithm
  - Incorporate cyclostationary spectrum density



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