

Secondary User Monitoring in Unslotted Cognitive Radio Networks with Unknown Models

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Outline

□ 1. **Background**

- ◆ 1.1 Problem
- ◆ 1.1 Challenges
- ◆ 1.2 Related works

□ 2. Motivation

□ 3. Problem Formulation

□ 4. Simulation

□ 5. Conclusion & Future Work

1. Background



- ❑ Why & How to perform secondary user data capture in a Cognitive Radio Networks with sniffers?
- ❑ What is our goal and measurement?

1.1 Problem

- Monitoring the detailed characteristics of an operational cognitive radio network is critical
 - ◆ spectrum policy enforcement, wireless advisory
 - ◆ fault diagnosis, anomaly detection, attack detection
 - ◆ forensics, resource management
 - ◆ critical path analysis for network upgrades



1.1 Problem

- Very limited work has been done
- Efficient capture with limited number of monitors
- Customable capture

1.2 Challenges

- ❑ Secondary users' activities are unknown in priori
 - ◆ highly dynamic wireless network
- ❑ Monitor resource is limited
 - ◆ CRNs have much wider spectrum than traditional wireless networks
 - ◆ limitation of hardware technology
- ❑ Sniffer channel assignment
 - ◆ tradeoff : Exploration or Exploitation
- ❑ Customable capture
 - ◆ capture data of interested secondary users coexisting with PUs and unconcerned SUs

1.3 Related Work

□ Multichannel Wireless Network Monitoring

- ◆ Chhetri et al. MobiHoc'10 [7] - QoM
- ◆ Arora et al. INFOCOM'11 [8] - MAB

□ Cognitive Radio Network Monitoring

- ◆ Chen et al. ICNP'11 [5] – SVR

□ Opportunistic Spectrum Access

- ◆ Tehrani et al. CAMSAP'11 [14] – Channel assignment

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2. Motivation

- Efficient capture with limited number of monitors
 - ◆ sharing information among sniffers
 - ◆ Tradeoff: exploration and exploitation
 - ◆ Dynamic assigning the sniffers
- Customable capture
 - ◆ Data from different users have priorities
 - ◆ Maximize the expected captured data of interested secondary users

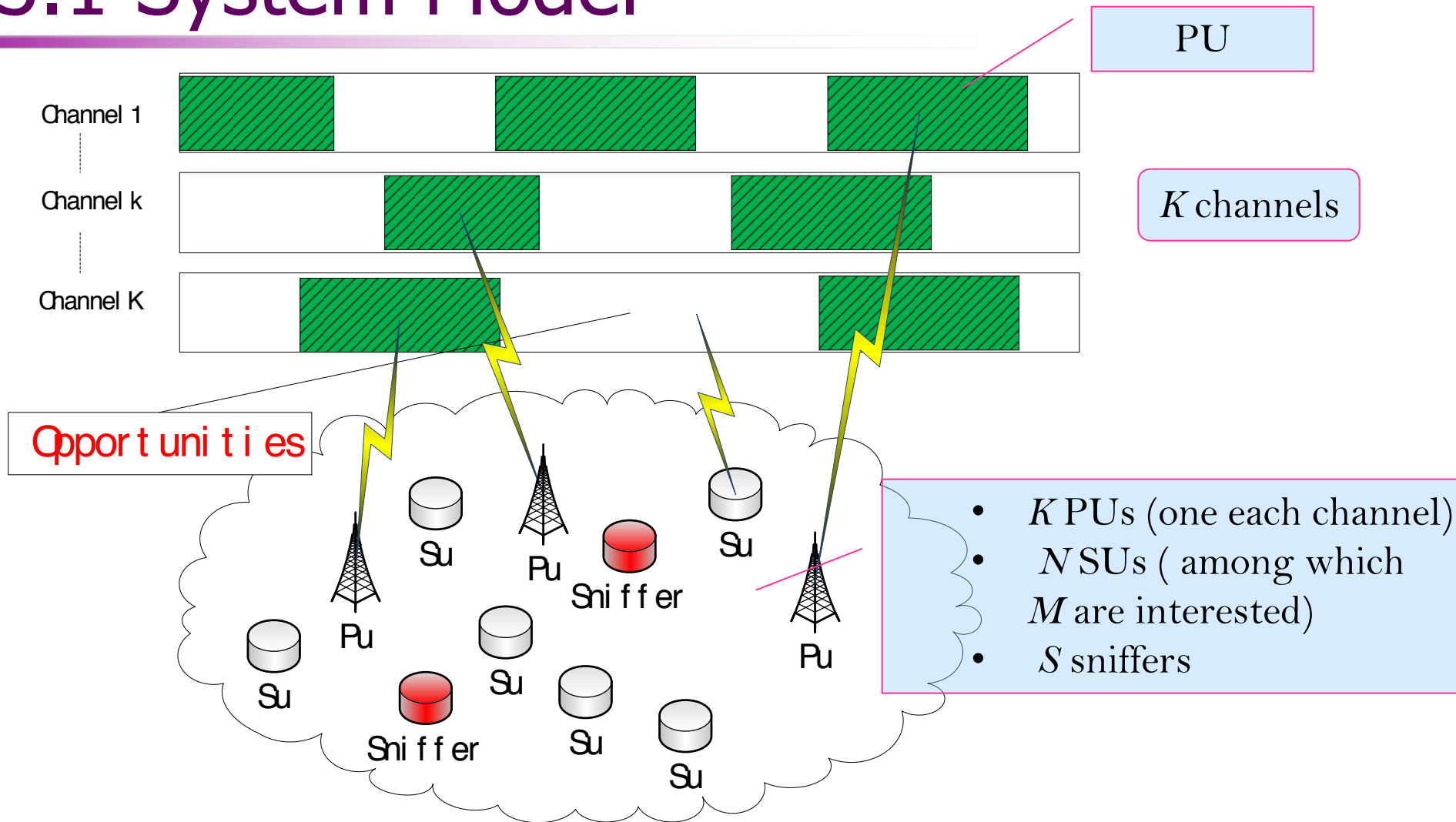
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3. Problem Formulation

- 3.1 System Model
- 3.2 Traffic Model
- 3.3 Objective
- 3.4 Multi-Armed Bandit Problem
 - ◆ 3.4.1 Reward
 - ◆ 3.4.2 Regret
 - ◆ 3.4.3 Data Capture Policy

3.1 System Model



3.1 System Model

□ Behavior of sniffer

- ◆ can only observe one channel at a time
- ◆ can switch to other channel at anytime
- ◆ Information is shared

□ Behavior of SU

- ◆ keep silent when the primary user shows up

3.2 Traffic Model

- The traffic of SU&PU is modeled as on-off renewal process.
 - ◆ The duration of state follows exponential distribution.
- Unslotted system
 - ◆ No need of time synchronization, no extra overhead
 - ◆ more common and adaptable
 - ◆ more complex due to the arbitrary starting and ending times of transmission.

3.3 Objective

- Maximize the expected captured data of interested secondary users
 - ◆ unknown prior information
 - exploration or exploitation
 - ◆ limited monitoring resource
 - sniffer assignment

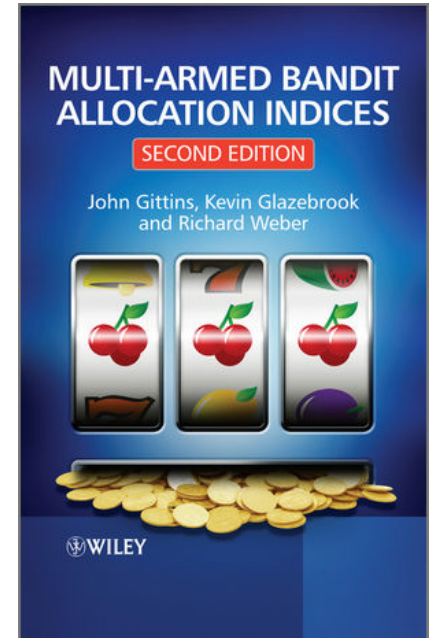
3.4 Multi-Armed Bandit Problem

□ N independent arms with
Known states $x_1(t), \dots, x_N(t)$

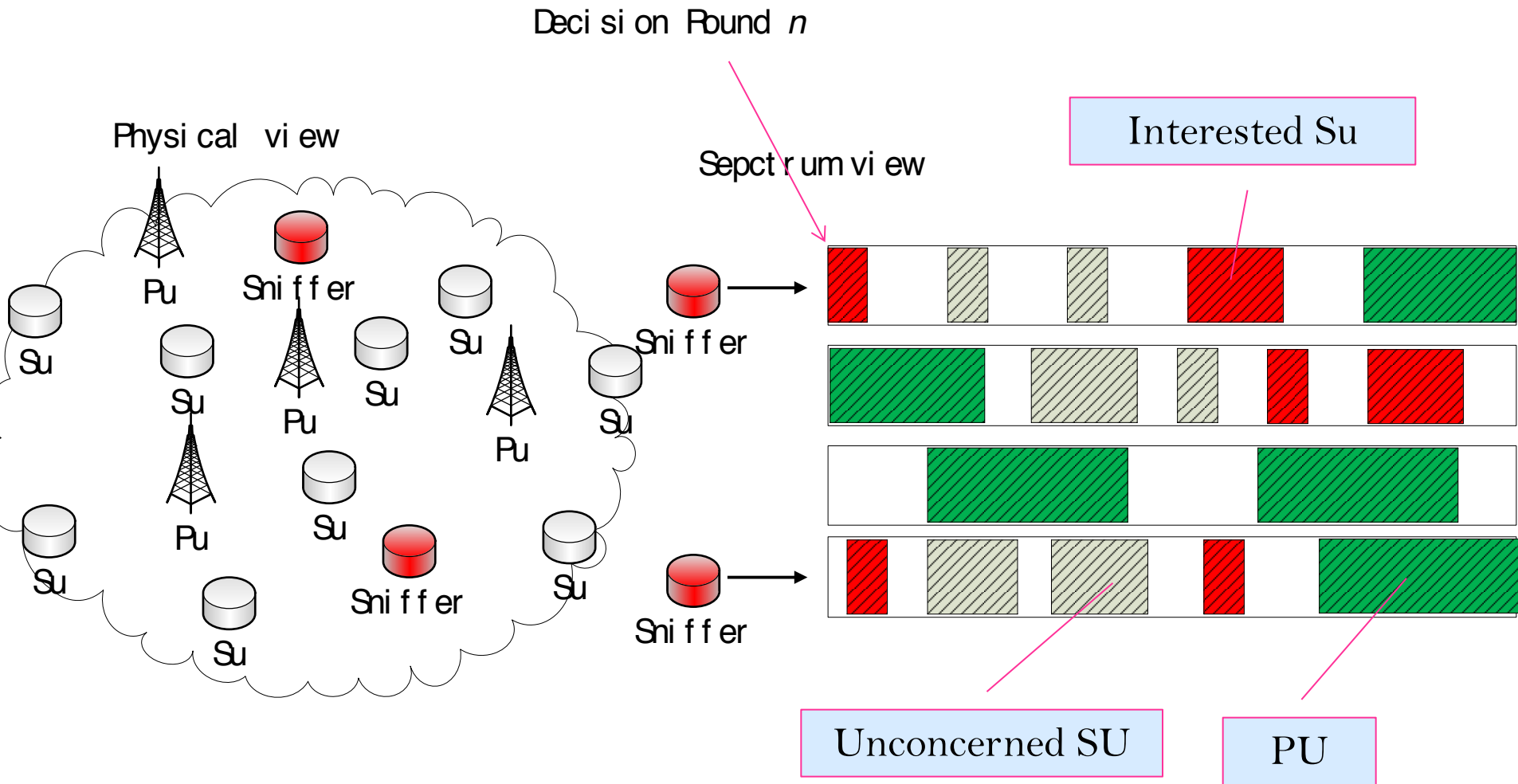
□ At each time t

- ◆ One arm is to be activated
- ◆ Others are not

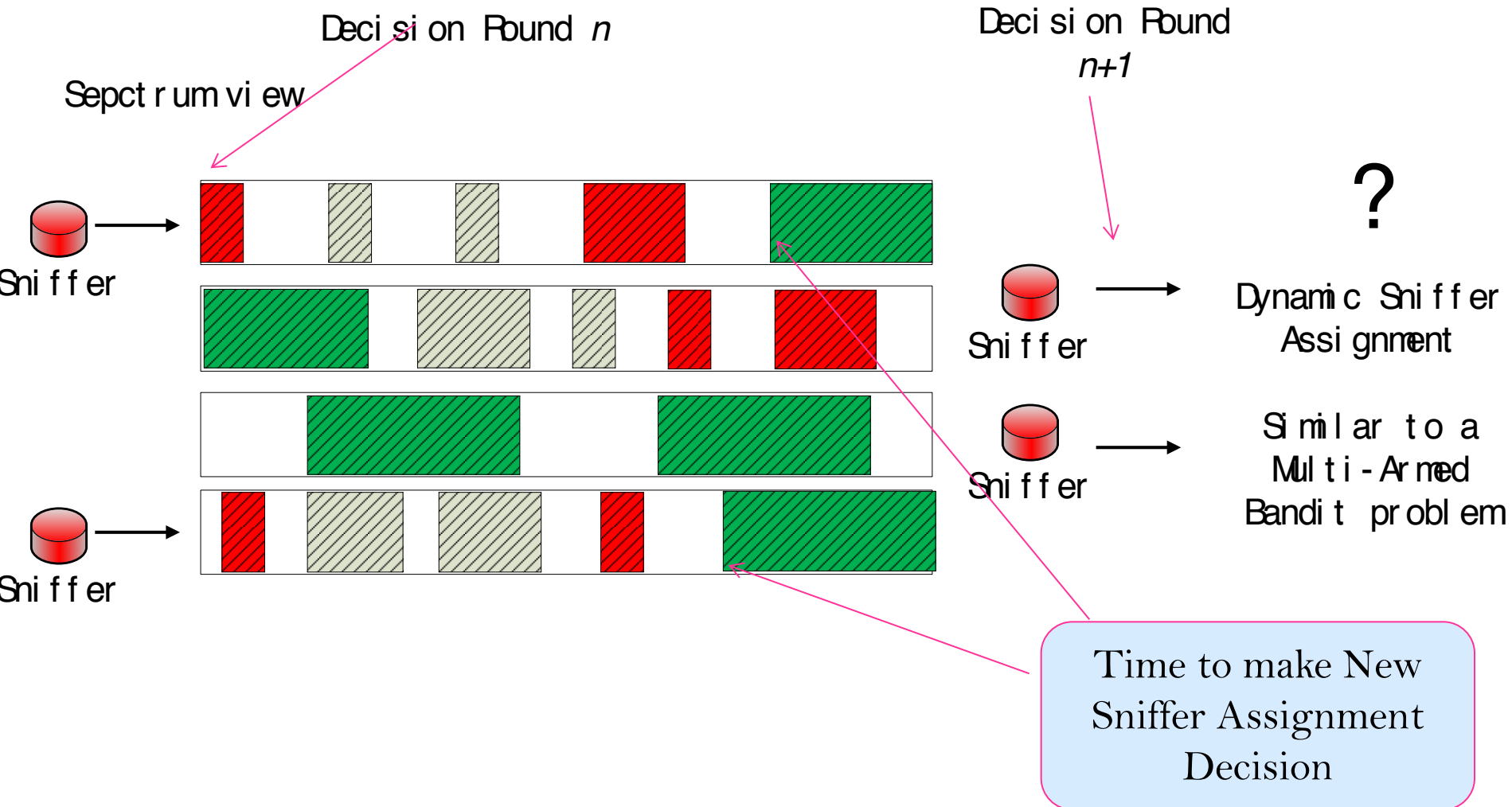
□ Objective: maximize the expected
reward



3.4 Multi-Armed Bandit Problem



3.4 Multi-Armed Bandit Problem



3.4 Multi-Armed Bandit Problem

- Player(Observer)

- ◆ Sniffer

- Arm

- ◆ Channel

- Reward

- ◆ Valid captured data of SUs

- Data Capture Policy

- ◆ channel assignment rule for sniffers

3.4.1 Reward

□ The total captured data of sniffers

$$V_o(t - t_0) = \sum_{n(t_0)}^{n(t)} \sum_k \sum_m y_{k,m}^o(n)$$

Summed up by length of every captured data of all the interested SUs

□ Virtual reward of channel k

$$V_k = \sum_m \frac{\hat{y}_{m,k}^o}{\hat{y}_{m,k}^o + \hat{x}_{m,k}^o} p_{m,k}^o$$

Summed up by average On/(Off+On) length ratio of all the interested SUs with weight value of the appearance probabilities of SUs

3.4.2 Regret

□ Genie in MAB

- ◆ Genie is ideal and omnipotent
- ◆ Genie knows prior information
- ◆ Genie always makes the right decision

□ Regret - Performance Measure

$$\mathbb{E}V^*(T) - \mathbb{E}V_o(T)$$

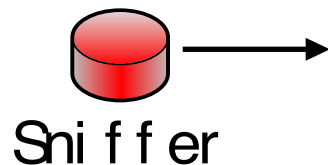
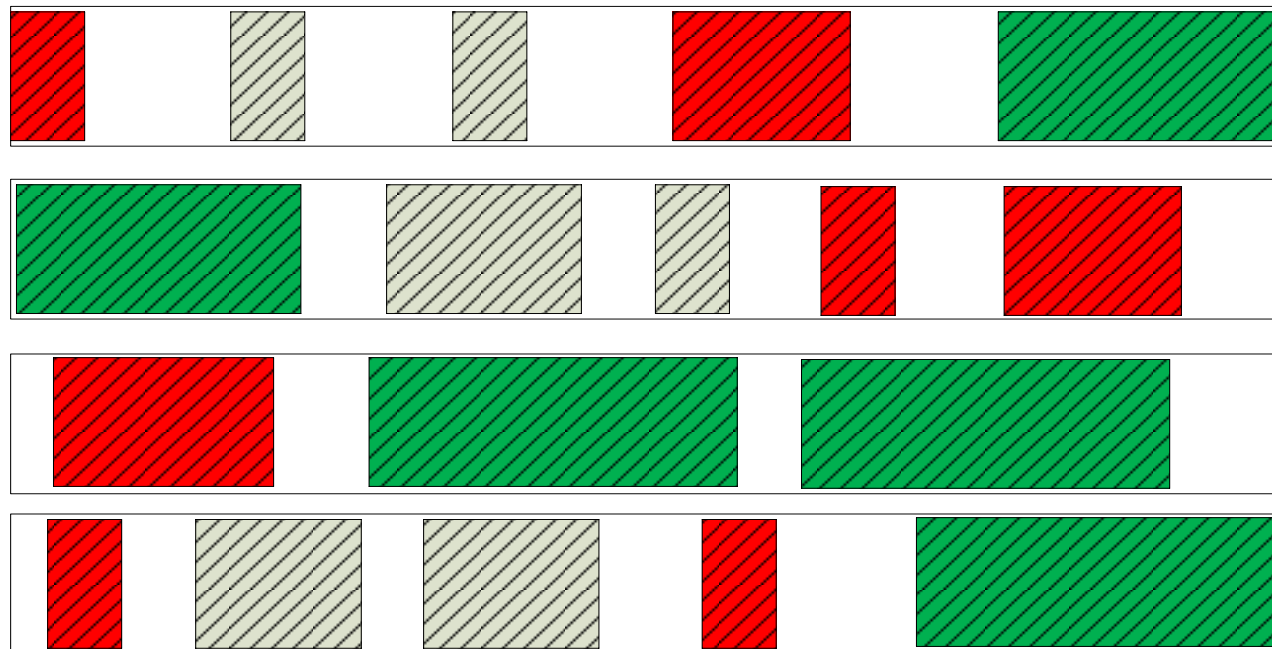
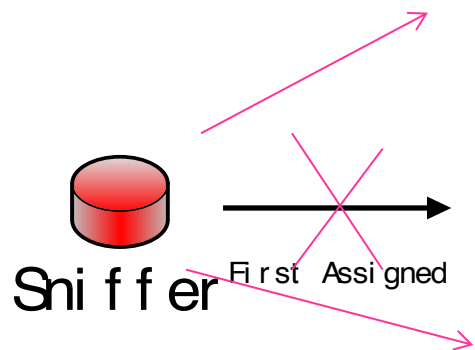
difference of expected value of gained reward between the “genie” and the proposed method

3.4.3 Data Capture Policy

□ Monitor Rule

- ◆ Random switching when a sniffer senses that the currently assigned channel is occupied by a primary user

Sequential view



3.4.3 Data Capture Policy

□ UCB1 (Upper Confident Bound 1)

Deterministic policy: UCB1.

Initialization: Play each machine once.

Loop:

- Play machine j that maximizes $\bar{x}_j + \sqrt{\frac{2 \ln n}{n_j}}$, where \bar{x}_j is the average reward obtained from machine j , n_j is the number of times machine j has been played so far, and n is the overall number of plays done so far.

Auer et al.
Finite-time
Analysis of
the Multi-
Armed
Bandit
Problem

3.4.3 Data Capture Policy

□ Modified UCB1

◆ Decision Index

◆ Update rule

$$I_k(t) = \overline{V}_k(t) + \sqrt{\frac{2 \log(t)}{\sum_m n_{k,m}(t)}}$$

Average virtual
reward from
channel k

$$\overline{V}_k(t + \Delta t) = \begin{cases} \frac{\overline{V}_k(t) + V_k(t, t + \Delta t)}{n_{k,m}(t) + 1} & \text{if SU } m \text{ is captured at channel } k; \\ \overline{V}_k(t) & \text{else} \end{cases}$$

$$n_{k,m}(t + \Delta t) = \begin{cases} n_{k,m}(t) + 1 & \text{if SU } m \text{ is captured at channel } k; \\ n_{k,m}(t) & \text{else} \end{cases}$$

◆ Index chosen rule

$$k = \arg \max_k I_k(t)$$

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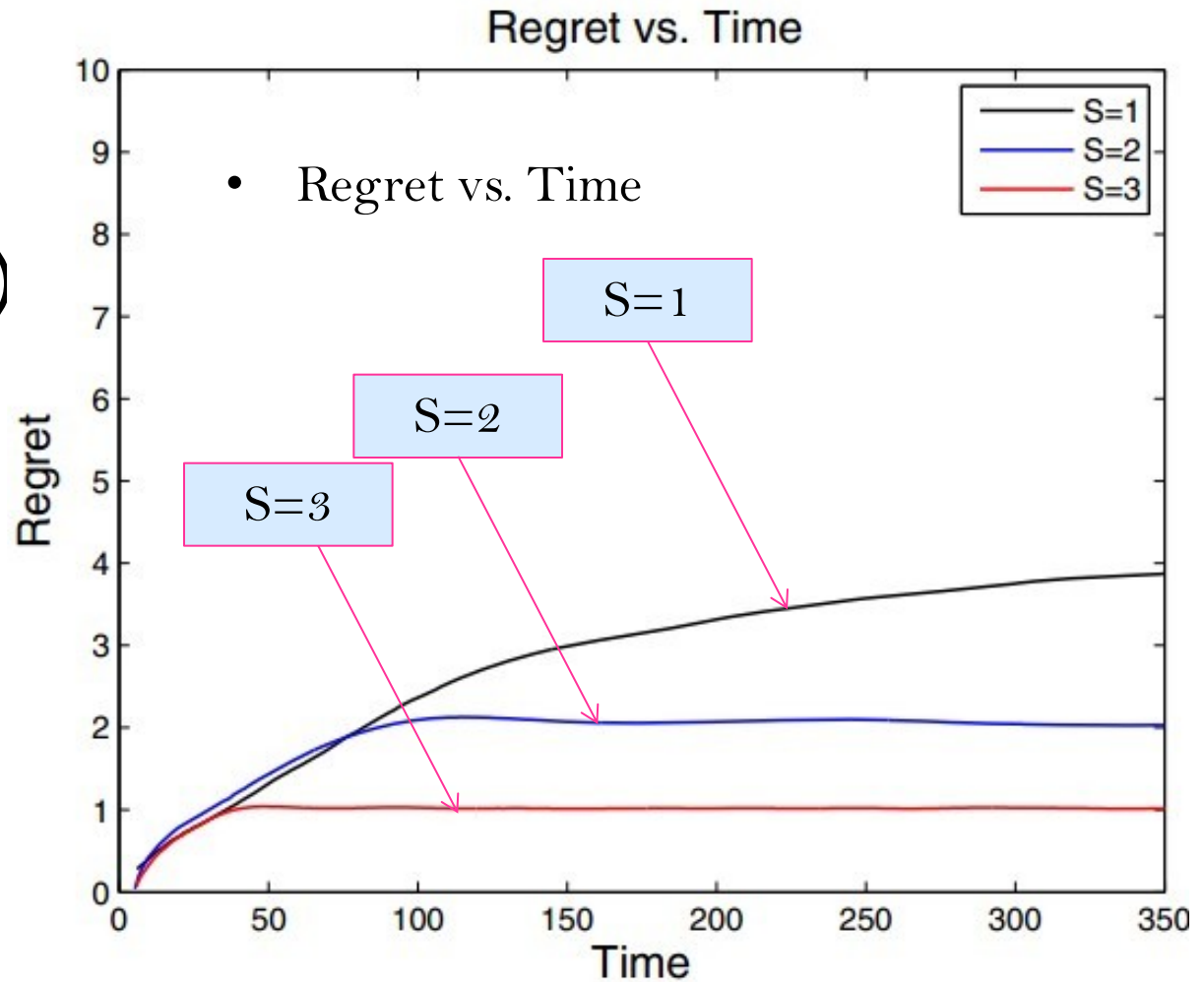
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4.1 Simulation Setting

- Number of channels $K = 4$
- Number of interested SUs/SUs $M/N=[4, 8, 12]/12$
- Number of sniffers $S=[1, 2, 3]$
- busy/idle periods of users follow exponential distributions
- The appearance probabilities of SUs in channels are randomly generated

4.2 Simulation Result

- ❑ Sniffer = [1,2,3]
- ❑ 4 Channels
- ❑ 8 Target SUs(12)



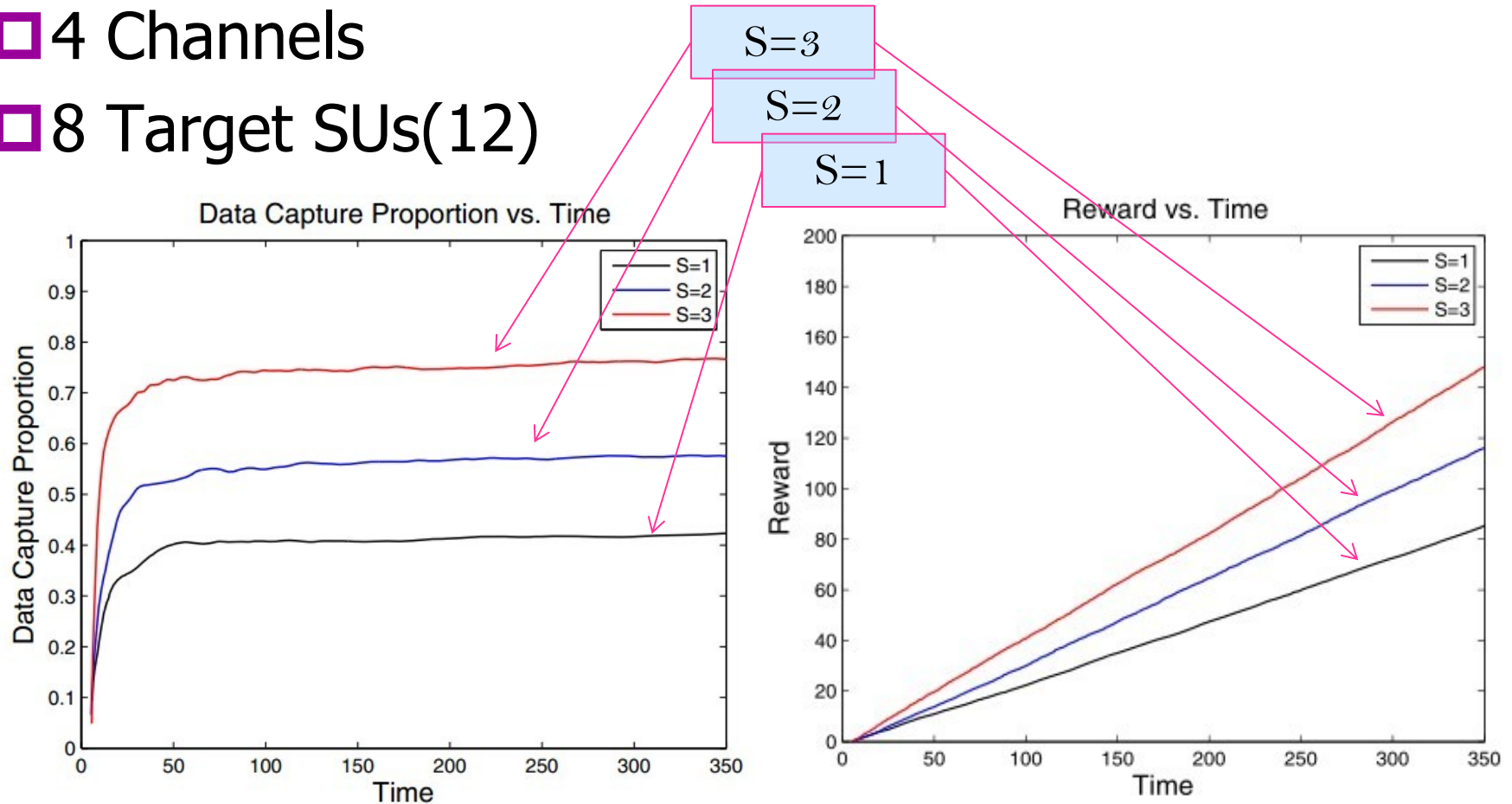
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- Data Capture Proportion vs. Time
- Reward vs. Time



4.2 Simulation Result

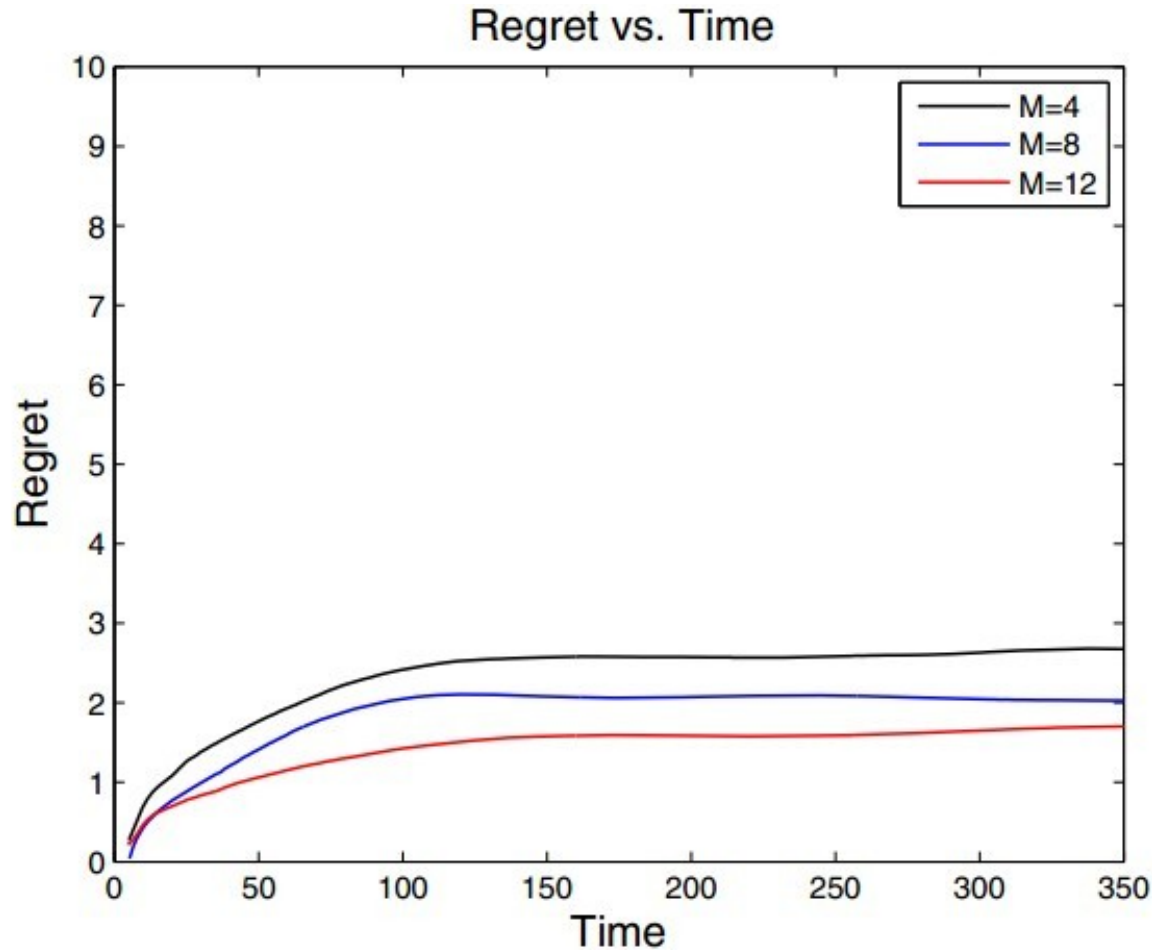
□ Interested

SUs/SUs=[4,8,12]/12

□ 2 sniffers

□ 4 channels

- Regret vs. Time



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5.1 Conclusion

- Propose a secondary user monitoring problem in unslotted cognitive radio networks with unknown user traffic statistics
- Formulate the problem as a MAB problem with weighted virtual reward
- Apply UCB1 method to solve the proposed problem
- Simulation shows that the proposed policy can achieve a logarithmic regret with relative scalability

5.2 Future Work

- ❑ study the secondary user monitoring problem in more complicated scenarios
- ❑ considering secondary user channel switching and its influence on our policy design

Thanks!

Any Questions?