

# On Quality of Monitoring for Multi-channel Wireless Infrastructure Networks

Arun Chhetri, **Huy Nguyen**, Gabriel Scalosub\*, and Rong Zheng

Department of Computer Science  
University of Houston, TX, USA

\*Department of Communication Systems Engineering  
Ben-Gurion University, Beer-Sheva, Israel



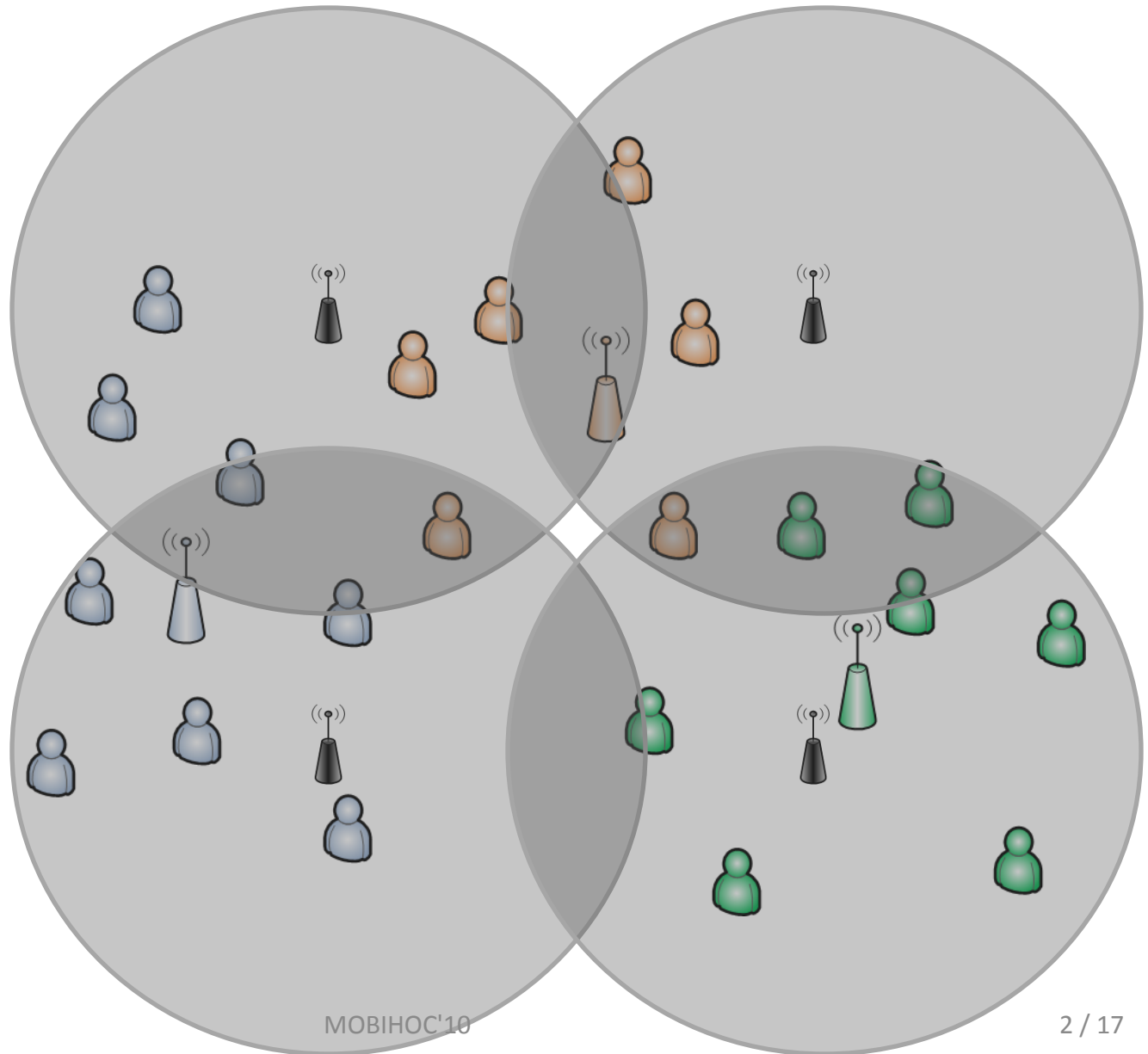
User



Access Point



Sniffer



# Outline

1. Introduction
2. Problem Formulation
3. QoM under User-centric Model
4. QoM under Sniffer-centric Model
5. Simulation and Experiments
6. Conclusion and Future Work

# Wireless Network Monitoring

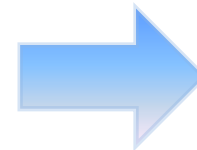
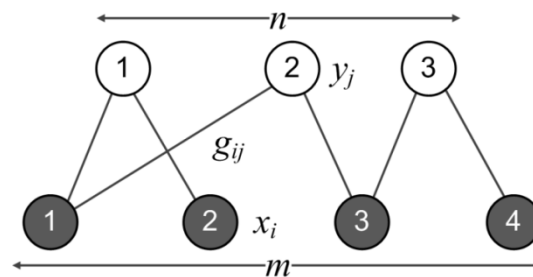
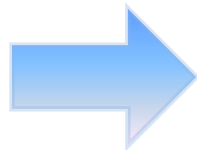
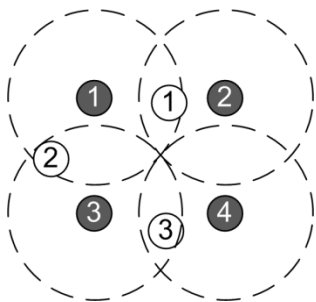
- Active vs. Passive monitoring
- Passive monitoring: using a dedicated set of **sniffers**
- Different sniffer capturing capability
  - **User-centric model**: frame-level information
  - **Sniffer-centric model**: binary channel activity
- Quality of Monitoring (**QoM**)
  - total expected number of active users captured

# Our Contributions

- Formulate the problem of **QoM** maximization
  - NP-hard
- Study **interaction** between two models: user-centric and sniffer-centric
- Propose a comprehensive scheme for **QoM** problem in both models
- Derive approximation algorithms to maximize **QoM** with proven lower bounds

# Problem Formulation

- Assumption: **user activities are independent**
- User-centric model
  - Consider a system of  $m$  sniffers,  $n$  users and  $k$  channels
  - A sniffer can only monitor a single channel at any time
  - Each user has a transmission probability  $p_u$
  - $\mathbf{G}$  : binary relationship matrix between sniffers and users
  - $\mathbf{G}$  and  $p_u$  are given



$$\mathbf{G} = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

# Problem Formulation

- Sniffer-centric model
  - User activity ( $\mathbf{y}$ ,  $p_u$ ) and mixing matrix  $\mathbf{G}$  are not available
  - Observation from  $m$  sniffers:  $\mathbf{x} = [x_1, x_2, \dots, x_m]$
  - Problem overview:  $x_i = \bigvee_{j=1}^n \underset{\text{(unknown)}}{g_{ij}} \wedge \underset{\text{(unknown)}}{y_j}$
- Versus User-centric model
  - Not as expressive as User-centric model
  - Aggregated statistics with less fluctuant tendency
  - Less hardware requirement and storage complexity

# MEC (Max Effort Coverage)

- With a set of sniffers, find the largest weighted set of users that can be monitored

$\begin{array}{ll} \text{max.} & \sum_{u \in U} p_u y_u \\ \text{s.t.} & \sum_{k=1}^K z_{s,k} \leq 1 \quad \forall s \in S \\ & y_u \leq \sum_{s \in N(u)} z_{s,c(u)} \quad \forall u \in U \\ & y_u \leq 1 \quad \forall u \in U \\ & y_u, z_{s,k} \in \{0, 1\} \quad \forall u, s, k. \end{array}$	$\left  \begin{array}{l} z_{s,k} = 1 \text{ if a sniffer is assigned to chan. } k \\ y_u = 1 \text{ if the user } u \text{ is monitored} \\ p_u \text{ is the weight associated with user } u \end{array} \right.$
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- Different from the problem in [Bagchi 09]
- Using a reduction from the problem of MON3SAT, we proved that
  - The unweighted MEC problem is NP-hard, even for  $k = 2$ .
  - The MEC problem is NP-hard to approximate to within a factor of



# Algorithms for MEC

- **Random**: sniffers are assigned randomly to a channel
- **Max**: sniffers are assigned to its busiest channel
- **Greedy**: sniffers are assigned to the channel that minimize the number of unmonitored users.

**Approximation factor:  $1/2$**

- **LP-Round**: solving the LP-relaxation of MEC, then round the fractional result into a integral solution (with e.g., probabilistic rounding technique in [Srinivasan 01]).

**Approximation factor:  $(1 - 1/e) \approx 0.632$**

# QoM under Sniffer-centric Model

- Convert Sniffer-centric to User-centric model
- Inferring  $G$ :
  - Classic Independent Component Analysis (ICA):  $\mathbf{x} = G \mathbf{y}$
  - ICA assumes continuous variables  $\rightarrow$  not directly applicable
  - Two steps process:
    1. Apply ICA to get an estimation of linear mixing matrix  $\hat{G}_L$ 
      - Non-negative fractional values in  $\hat{G}_L$
    2. Quantize  $\hat{G}_L$  to have the binary mixing matrix  $\hat{G}$ 
      - Normalizing and thresholding

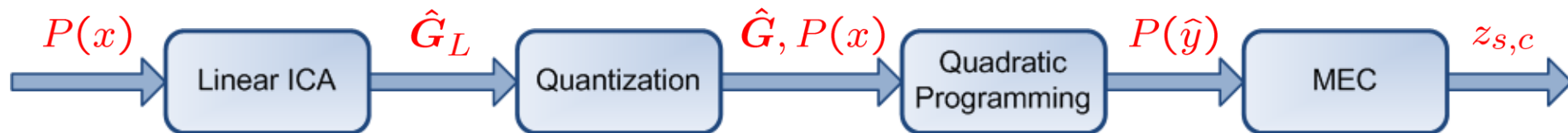
# QoM under Sniffer-centric Model

- Inferring  $p_u$  given  $G$ :
  - Assuming all users are independent

$$p(x_i = 0) = \prod_{\hat{g}_{ij}=1} p(y_j = 0)$$

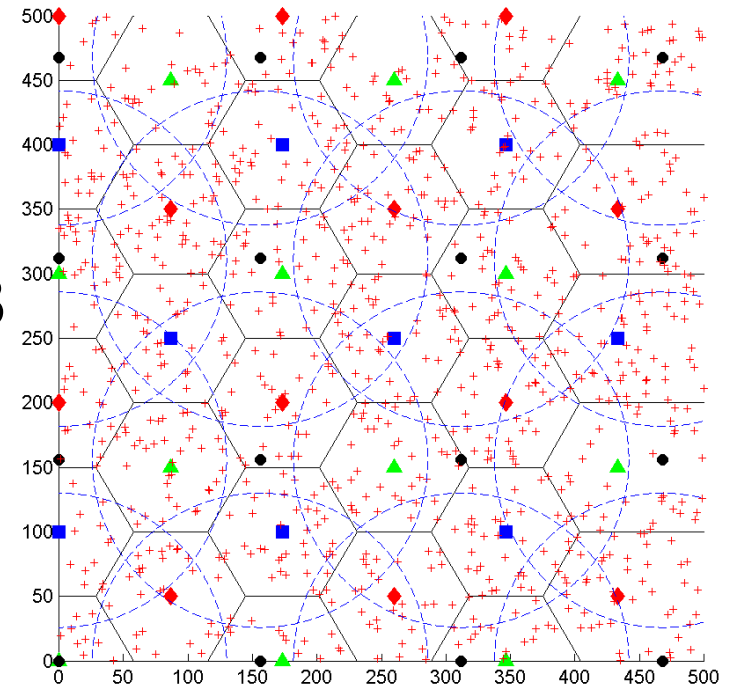
- Let  $\alpha_i = \log(p(x_i = 0))$ ,  $\beta_i = \log(p(y_j = 0))$
- Define  $\alpha = [\alpha_1, \alpha_2, \dots, \alpha_m]^T$ ,  $\beta = [\beta_1, \beta_2, \dots, \beta_n]^T$
- We have the quadratic programming problem

$$\begin{aligned} \min. \quad & \| \alpha - \hat{G}\beta \|^2 \\ \text{s.t.} \quad & \beta < 0, \end{aligned}$$

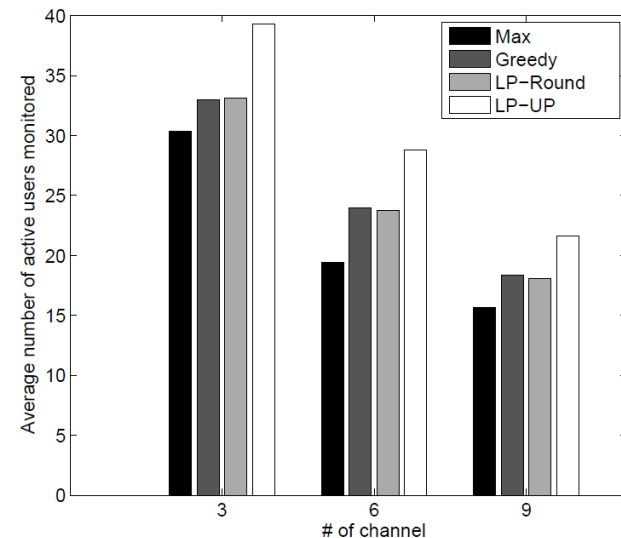
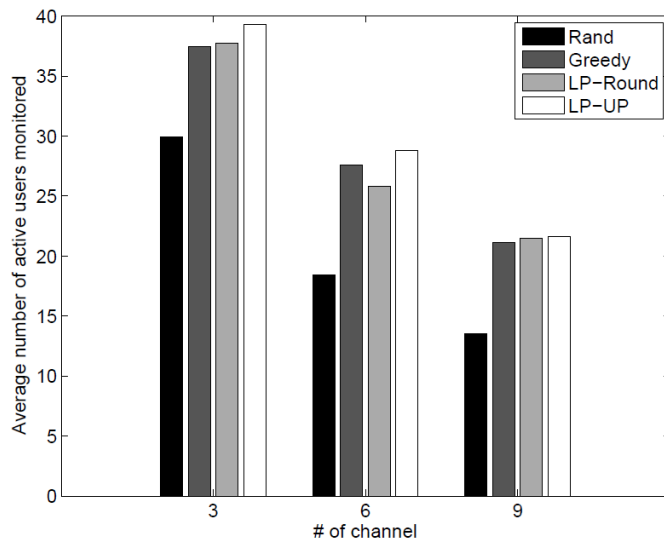


# Synthetic Traces

- 1000 wireless users are randomly placed on a 500x500 square meter area.
- 25 sniffers are deployed in a grid formation
- Distance = 100 meters
- Coverage radius = 120 meters
- $p_u \in (0, 0.06]$
- Average busy  $p$  on each cell  $\approx 0.3$
- Orthogonal channels =  $\{3, 6, 9\}$
- Observations  $T = 10,000$



# Synthetic Traces



## User-centric Model

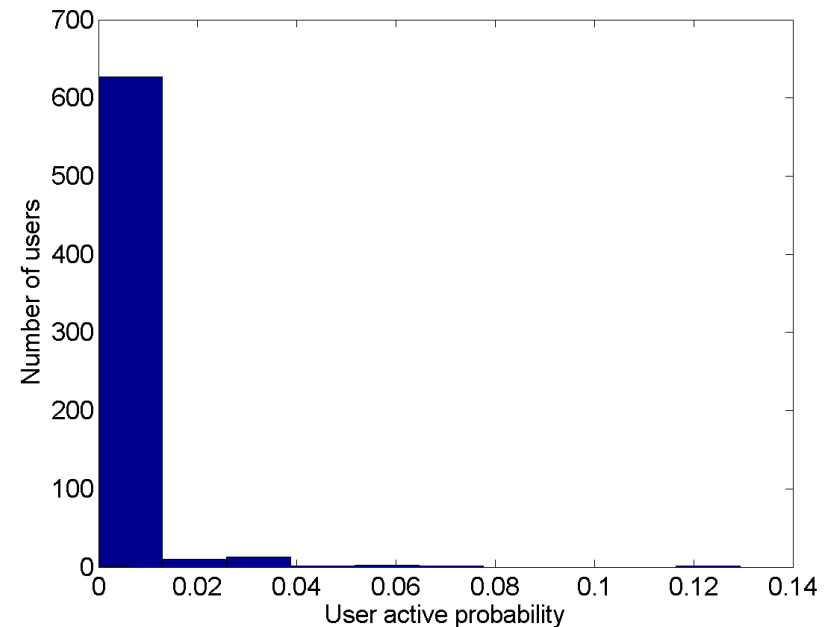
- Greedy and LP-Round are close to LP-UP
- Rand gets worst as no. of channels increases

## Sniffer-centric Model

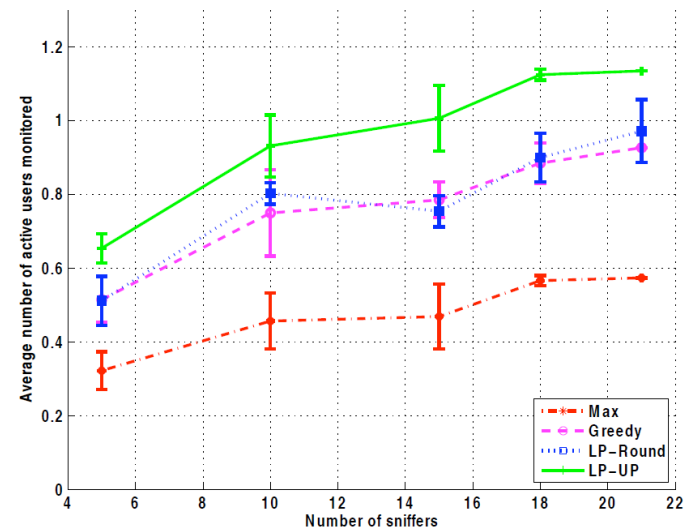
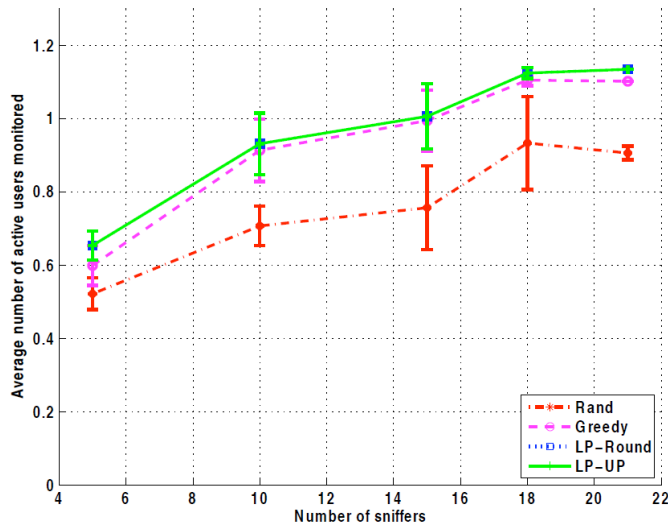
- Greedy and LP-Round outperform Max
- Degradations comparing to User-centric Model

# Real Traces

- 21 WiFi sniffers deployed at M.D. Anderson Library, U. of Houston
- Number of observations  $T = 300,000$
- 655 unique users identified on  
3 WiFi channels (1, 6, 11)
- Average  $p_u = 0.0014$
- Varies number of selected  
sniffers from 5 – 21



# Real Traces



## User-centric Model

- Greedy and LP-Round are close to LP-UP
- QoM and sniffer no. are monotonically increasing

## Sniffer-centric Model

- Greedy and LP-Round outperform Max
- Degradations comparing to User-centric Model

# Conclusion and Future Work

- Derive and address the problem of maximizing QoM (MEC) in multi-channel wireless networks
- Recover user-level info from sniffer observation
- Solve the MEC problem with proposed algorithms on synthetic and real systems
- Ongoing researches
  - Binary Independent Component Analysis with OR Mixtures  
<http://arxiv.org/abs/1007.0528>
  - Various network applications  
E.g.: PU separation in cognitive radio system, Link loss analysis in multicast network ...





# THANK YOU FOR YOUR ATTENTION



`nahuy@cs.uh.edu`  
`http://wireless.cs.uh.edu`