

Secret-Sharing-Based Secure User Recruitment Protocol for Mobile Crowdsensing

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Outline

- **Motivation**
- **Model & Problem**
- **Solution**
- **Extension**
- **Evaluation**
- **Conclusion**

Motivation

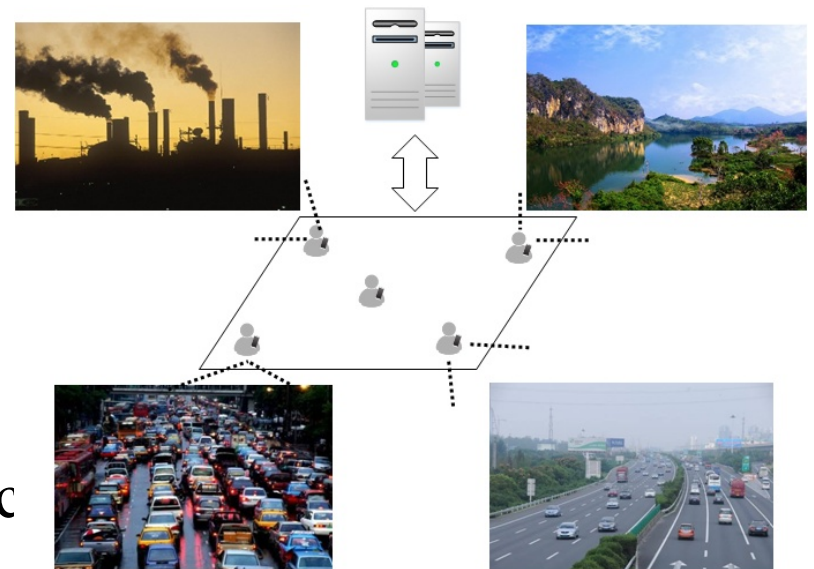
Mobile Crowdsensing

■ Flexible Sensing Paradigm

A requester can recruit a group of mobile users via a platform and coordinate them to perform some sensing tasks by using their smart phones

■ Applications

Urban WiFi characterization,
Traffic information mapping,
Noise pollution monitoring, etc



Motivation

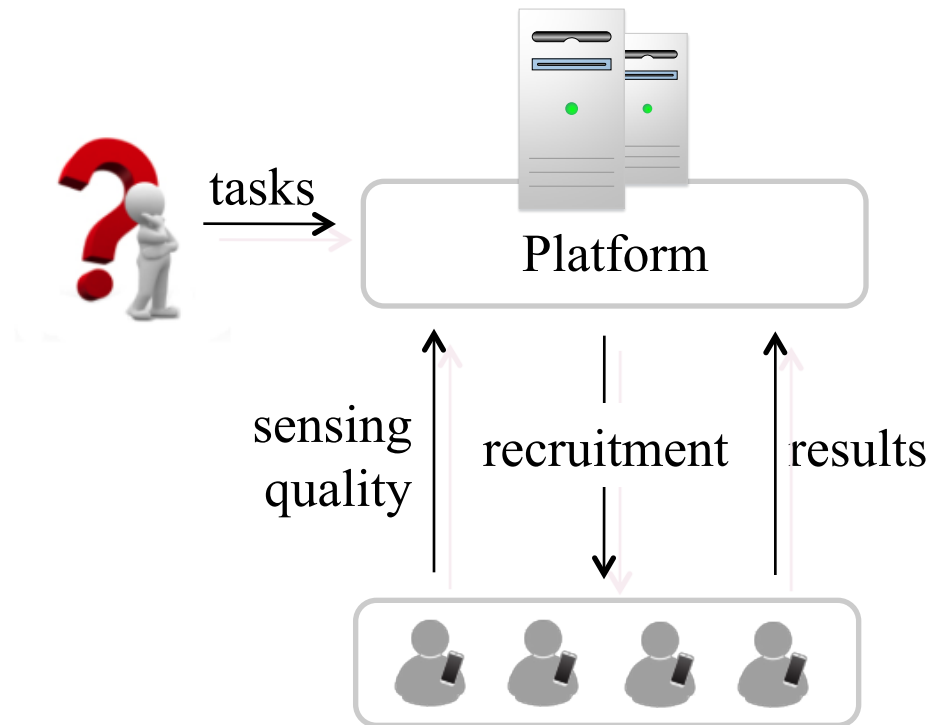
Mobile Crowdsensing

■ Typical system

a platform on the cloud
a group of mobile users
requesters

■ User recruitment

minimum users/cost
enough sensing quality



Motivation

Protecting users' privacy

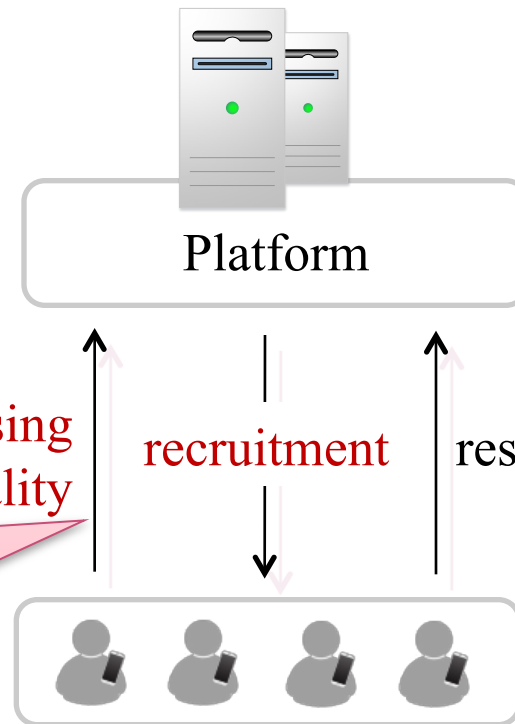
- Sensing results
- Privacy-preserving user recruitment

Imply which **locations** each user might visit, and the corresponding **frequency**, **distance**, **time**, etc.

sensing
quality

recruitment

results



[Rescuedp, infocom16]

Model

Quality-sensitive Crowdsensing Model

- m sensing **tasks**: $S = \{s_1, \dots, s_m\}$
- n mobile **users**: $U = \{u_1, \dots, u_n\}$
- each user might perform one or more tasks
- sensing **quality**: $q_{i,j} \in Z_p, 1 \leq i \leq n, 1 \leq j \leq m$
- $q_{i,j} = 0$ means that user u_i cannot deal with task s_j

Semi-honest Security Model

- the **dishonest** aspect: each user will try to derive the extra information from the received data
- the **honest** aspect: the user will also follow the whole user recruitment protocol, so as to benefit from the crowdsensing

Problem

Privacy-preserving User Recruitment Problem

- **Objective:** Securely recruit some users to perform all tasks so that we can **minimize the number of recruited users**, while ensuring that **the total sensing quality of each task** is no less than a given threshold θ .

- **Formalization**

$$\text{Minimize : } |\Phi|$$

$$\text{Subject to : } \Phi \subseteq U$$

$$Q_j \geq \theta, \quad 1 \leq j \leq m$$

Security under the semi-honest model

$$\text{Total sensing quality of task } s_j : Q_j = \sum_{u_i \in \Phi} q_{i,j}$$

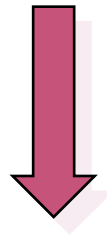
Solution

Problem Hardness Analysis

- **Theorem 1:** The user recruitment problem is NP-hard.

Basic idea

- **Basic User Recruitment (BUR)** protocol



Secret sharing scheme

Secure multi-party computation

- **Secure User Recruitment (SUR)** protocol

Basic User Recruitment Protocol

Utility Function

- **Utility function** $f(\Phi)$ indicates the total sensing qualities of all tasks in S contributed by the users in set Φ , until they reach the threshold θ :

$$f(\Phi) = \sum_{j=1}^m \min\{Q_j, \theta\} = \sum_{j=1}^m \min\left\{\sum_{u_i \in \Phi} q_{i,j}, \theta\right\}$$

- **Marginal utility**

$$\Delta_i f(\Phi) = f(\Phi \cup \{u_i\}) - f(\Phi)$$

Greedy User Recruitment Strategy

$$\Phi \leftarrow u_i : \arg \max_{u_i \in U \setminus \Phi} \Delta_i f(\Phi)$$

Basic User Recruitment Protocol

The Detailed BUR Protocol

Protocol 1 The BUR Protocol

Input: \mathcal{U} , \mathcal{S} , $\{q_{i,j} | u_i \in \mathcal{U}, s_j \in \mathcal{S}\}$, θ

Output: Φ , b_1, \dots, b_n

Phase 1: the requester publishes \mathcal{S} to \mathcal{U} via the platform;

Phase 2: users input their sensing quality values;

1: **for** $i = 1$ **to** n **do**

2: user u_i sends $\{q_{i,1}, \dots, q_{i,m}\}$ to the platform;

Phase 3: the platform makes the decision of user recruitment;

3: $\Phi = \emptyset$; $f(\Phi) = 0$;

4: **while** $f(\Phi) < m\theta$ **and** $|\Phi| < n$ **do**

5: Select a user $u_i \in \mathcal{U} \setminus \Phi$ to maximize $\Delta_i f(\Phi)$;

6: $\Phi = \Phi \cup \{u_i\}$;

Phase 4: the platform returns the results to users;

7: **for** $i = 1$ **to** n **do**

8: **if** $u_i \in \Phi$ **then**

9: the platform returns $b_i = 1$ to user u_i ;

10: **else**

11: the platform returns $b_i = 0$ to user u_i ;

Basic User Recruitment Protocol

Correctness and Approximation Ratio of BUR

- **Theorem 2:** $f(\Phi)$ is an **increasing** function with $f(\emptyset)=0$.
- **Theorem 4:** $f(\Phi)$ is a **submodular** function.
- **Theorem 5:** the BUR protocol is **correct**.
- **Theorem 6:** BUR can produce a **$(1+\ln\gamma)$ -approximation solution**, where $\gamma = \max_{u_i \in U} f(\{u_i\})$

Secure User Recruitment Protocol

Secret shares

the **shares of a secret** s among n users are denoted as

$$[s] \equiv (s[1], \dots, s[i], \dots, s[n])$$

$s[i]$ is the i -th user's share.

Shamir's secret sharing scheme

Let p be an odd prime and Z_p be a prime field. To share a secret s ($s \in Z_p$) among n users ($n < p$), Shamir's scheme determines **a random polynomial function**

$$g_s(x) = s + \alpha_1 x + \alpha_2 x^2 + \dots + \alpha_k x^k \pmod{p}$$

with randomly chosen for $\alpha_i \in Z_p$ for $1 \leq i \leq k$, $k \leq n$. Then, the share of the i -th user is **$s[i] = g_s(i)$** .

Secure User Recruitment Protocol

The Building Blocks

- Secure multi-party **addition/subtraction** operation

$$[z_1] \leftarrow \text{SecAdd}([x], [y]), \quad [z_2] \leftarrow \text{SecSub}([x], [y])$$

$$[x] \equiv (x[1], \dots, x[i], \dots, x[n])$$

$$[y] \equiv (y[1], \dots, y[i], \dots, y[n])$$



$$[z_1] \equiv (z_1[1], \dots, z_1[i], \dots, z_1[n])$$

$$[z_2] \equiv (z_2[1], \dots, z_2[i], \dots, z_2[n])$$

$$z_1[i] = x[i] + y[i] \pmod{p}$$

$$z_2[i] = x[i] - y[i] \pmod{p}$$

Secure User Recruitment Protocol

The Building Blocks

- Secure multi-party **multiplication/comparison** operation

$$[z_3] \leftarrow \text{SecMulti}([x], [y]), \quad [z_4] \leftarrow \text{SecCmp}([x], [y])$$

$$[x] \equiv (x[1], \dots, x[i], \dots, x[n])$$

$$[y] \equiv (y[1], \dots, y[i], \dots, y[n])$$



$$[z_3] \equiv (z_3[1], \dots, z_3[i], \dots, z_3[n])$$

$$z_3 = xy \pmod{p}$$



$$[z_4] \equiv (z_4[1], \dots, z_4[i], \dots, z_4[n])$$

$$z_4 = \begin{cases} 1 \pmod{p}, & x \leq y \\ 0 \pmod{p}, & x > y \end{cases}$$

Secure User Recruitment Protocol

The Building Blocks

■ Secure multi-party **max/min** operation

$$[z_5] \leftarrow \text{SecMax}([x], [y]), \quad [z_6] \leftarrow \text{SecMin}([x], [y])$$

$$[x] \equiv (x[1], \dots, x[i], \dots, x[n])$$

$$[y] \equiv (y[1], \dots, y[i], \dots, y[n])$$

$$[z_5] \equiv (z_5[1], \dots, z_5[i], \dots, z_5[n])$$

$$z_5 = \max\{x, y\} \bmod p$$

$$[z_6] \equiv (z_6[1], \dots, z_6[i], \dots, z_6[n])$$

$$z_6 = \min\{x, y\} \bmod p$$

$$\begin{aligned} \text{SecMax}([x], [y]) &\equiv \text{SecAdd}([x], \\ &\text{SecMulti}(\text{SecCmp}([x], [y], \\ &\text{SecSub}([x], [y]))) \end{aligned}$$

$$\begin{aligned} \text{SecMin}([x], [y]) &\equiv \text{SecAdd}([x], \\ &\text{SecMulti}(\text{SecSub}(1 - \text{SecCmp} \\ &([x], [y])), \text{SecSub}([x], [y]))) \end{aligned}$$

Secure User Recruitment Protocol

From BUR to SUR

■ Inputs

For each user's sensing quality: $q_{i,j} \rightarrow [q_{i,j}]$

■ Outputs

User recruitment result: $\Phi \rightarrow (b_1, \dots, b_n); u_i \in \Phi \rightarrow [b_i] = [1]$

■ Securely compute marginal utility

$$\begin{aligned}\Delta_i f(\Phi) &= f(\Phi \cup \{u_i\}) - f(\Phi) \\ &= \sum_{j=1}^m \min\{q_{i,j}, \theta - Q_j\}\end{aligned}$$



$$[\Delta_i f] \leftarrow \text{SecAdd}_{j=1}^m : \text{SecMin}([q_{i,j}], \text{SecSub}(\theta, [Q_j]))$$

Secure User Recruitment Protocol

From BUR to SUR

- Securely determine the recruited user

$$[\Delta_{\max} f] \leftarrow \text{SecMax}([\Delta_1 f], \dots, [\Delta_n f])$$

for $i = 1 \rightarrow n$ do

$$[z] \leftarrow \text{SecCmp}([\Delta_{\max} f], [\Delta_i f]);$$

$$[b_i] \leftarrow \text{SecAdd}([b_i], \text{SecMulti}(\text{SecSub}([1], [b_i]), [z]));$$

- Securely update the total sensing quality in each round

for $j = 1 \rightarrow m$ do

$$[\delta] \leftarrow \text{SecMin}([q_{i,j}], \text{SecSub}(\theta, [Q_j]));$$

$$[Q_j] \leftarrow \text{SecAdd}([Q_j], \text{SecMulti}([z], [\delta]));$$

Secure User Recruitment Protocol

The Detailed SUR Protocol

Protocol 2 The SUR Protocol

Input: \mathcal{U} , \mathcal{S} , $\{q_{i,j} | u_i \in \mathcal{U}, s_j \in \mathcal{S}\}$, θ

Output: b_1, \dots, b_n

Phase 1: the requester publishes \mathcal{S} to \mathcal{U} via the platform;

Phase 2: users input their sensing quality vectors;

- 1: **for** $i=1$ **to** n **do**
- 2: user u_i determines the sensing qualities $q_{i,1}, \dots, q_{i,m}$;
- 3: **for** $j=1$ **to** m **do**
- 4: user u_i generates the polynomial sharing $[q_{i,j}]$;
- 5: user u_i sends the share $q_{i,j}[i']$ to user $u_{i'}$;

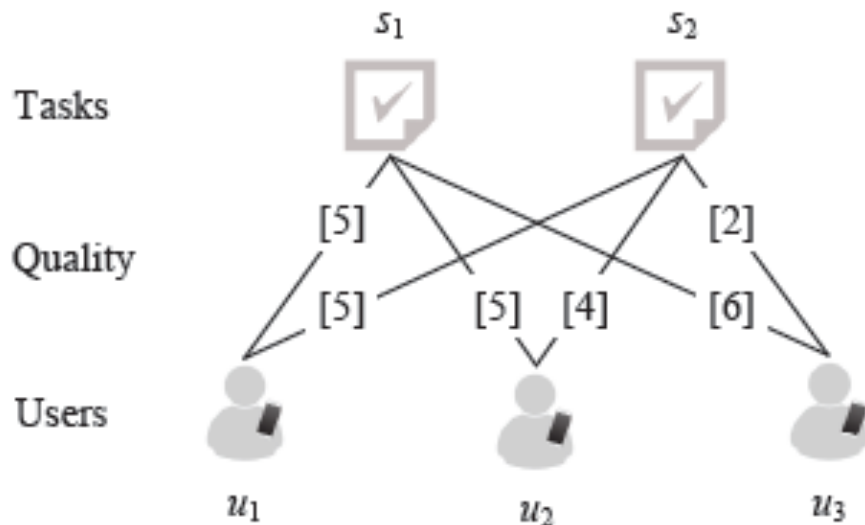
Phase 3: users jointly make the decision of user recruitment;

- 6: **for** $i=1$ **to** n **do**
- 7: $[b_i] \leftarrow [0]$;
- 8: **for** $j=1$ **to** m **do**
- 9: $[Q_j] \leftarrow [0]$;
- 10: **for** $round=1$ **to** n **do**

- 11: **for** $i=1$ **to** n **do**
- 12: $[\Delta_i f] \leftarrow [0]$;
- 13: **for** $j=1$ **to** m **do**
- 14: $[\delta] \leftarrow SecMin([q_{i,j}], SecSub(\theta, [Q_j]))$;
- 15: $[\Delta_i f] \leftarrow SecAdd([\Delta_i f], [\delta])$;
- 16: $[\Delta_i f] \leftarrow SecMulti([\Delta_i f], SecSub([1], [b_i]))$;
- 17: $[\Delta_{max} f] \leftarrow SecMax([\Delta_1 f], \dots, [\Delta_n f])$;
- 18: **for** $i=1$ **to** n **do**
- 19: $[z] \leftarrow SecCmp([\Delta_{max} f], [\Delta_i f])$;
- 20: $[b_i] \leftarrow SecAdd([b_i], SecMulti(SecSub([1], [b_i]), [z]))$;
- 21: **for** $j=1$ **to** m **do**
- 22: $[\delta] \leftarrow SecMin([q_{i,j}], SecSub(\theta, [Q_j]))$;
- 23: $[Q_j] \leftarrow SecAdd([Q_j], SecMulti([z], [\delta]))$;
- Phase 4:** the users reconstruct the results;
- 24: **for** $i=1$ **to** n **do**
- 25: user u_i collects all shares of $[b_i]$;
- 26: user u_i derives $b_i = \sum_{j=1}^m b_i[j]$;

Secure User Recruitment Protocol

Example



(a) Users, tasks and sensing qualities

round 1	round 2
<u>$[\Delta_1 f] = [10]$</u>	$[\Delta_1 f] = [0]$
$[\Delta_2 f] = [9]$	<u>$[\Delta_2 f] = [6]$</u>
$[\Delta_3 f] = [8]$	$[\Delta_3 f] = [5]$
<u>$[b_1] = [1]$</u>	$[b_1] = [1]$
$[b_2] = [0]$	<u>$[b_2] = [1]$</u>
$[b_3] = [0]$	$[b_3] = [0]$
$[Q_1] = [5]$	$[Q_1] = [8]$
$[Q_2] = [5]$	$[Q_2] = [8]$

(b) Intermediate results

The total sensing quality threshold $\theta=8$

Secure User Recruitment Protocol

Performance Analysis

$O(mn^2)$ invocations of secure multiplication operations

$O(mn^4l)$ bit-operations per user ($l = \lceil \log_2 p \rceil$)

$O(mn^2l)$ rounds of communication

Correctness and Approximation Ratio

- **Theorem 7:** SUR is correct, and it can also produce a $(1+\ln\gamma)$ -approximation solution, where $\gamma = \max_{u_i \in U} f(\{u_i\})$

Secure User Recruitment Protocol

Security of SUR

- **Theorem 8:** SUR can protect the sensing qualities of each user from being revealed to any κ semi-honest adversaries and the platform, even if they might collude, where κ (i.e., the degree of polynomial sharing) may be any integer less than n .

Extension

Extension: the total sensing quality function

- $Q(\bullet)$ becomes a general function about $q_{i,j}$

$$Q_j(\Phi) \equiv Q(q_{i,j} \mid u_i \in \Phi)$$

- **Example**

The sensing quality $q_{i,j}$ represents the probability of successful sensing

$Q(\bullet)$ may be defined as the joint successful probability

$$Q_j(\Phi) = 1 - \prod_{u_i \in \Phi} (1 - q_{i,j})$$

Extension

Extension

- **Theorem 9:** When $Q_j(\Phi)$ is a trivial function that can be securely computed by using the secure multi-party computation operations in SUR, **SUR will still be secure.**
- **Theorem 10:** When $Q_j(\Phi)$ is an increasing submodular function with $Q_j(\Phi=\emptyset)=0$, we have: 1) the utility function $f(\Phi)$ is still submodular; 2) **SUR can still produce a $(1+\ln\gamma)$ -approximation solution**, where $\gamma = \max_{u_i \in U} f(\{u_i\})$

Evaluation

Evaluate the User Recruitment Performance

■ Compared Protocols

MCUR: the user who can perform the most tasks is recruited first

MQUR: the user who performs tasks with the most sensing qualities is recruited first

■ Simulation Settings

Synthetic traces

Metric: the number of recruited users

Evaluation

Evaluate the User Recruitment Performance

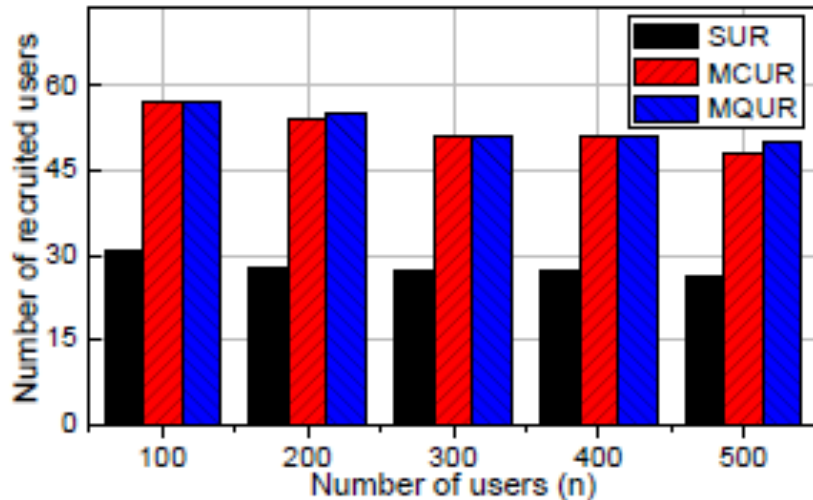
■ Simulation Settings

Parameter name	default	range
number of users n	200	100-500
number of tasks m	100	50-250
average sensing quality p	30	10-90
variance of sensing qualities σ	0.4	0.2-1.0
sensing quality threshold θ	100	20-250
largest number of tasks per user ρ	20	15-35

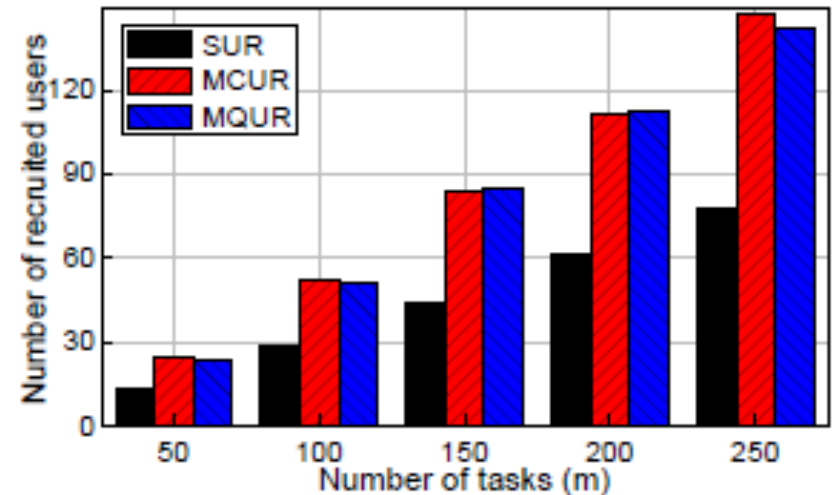
Evaluation

Evaluate the User Recruitment Performance

■ Evaluation Results



(a) $m = 100$



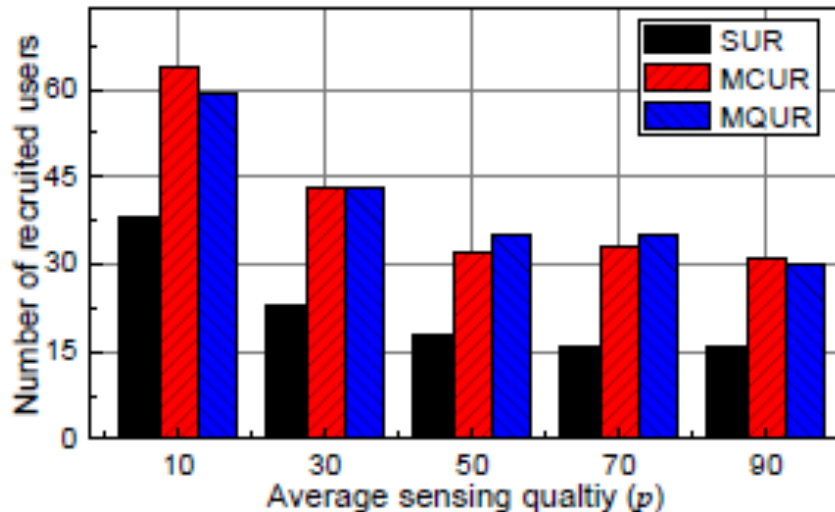
(b) $n = 200$

Number of recruited users vs. number of users and tasks

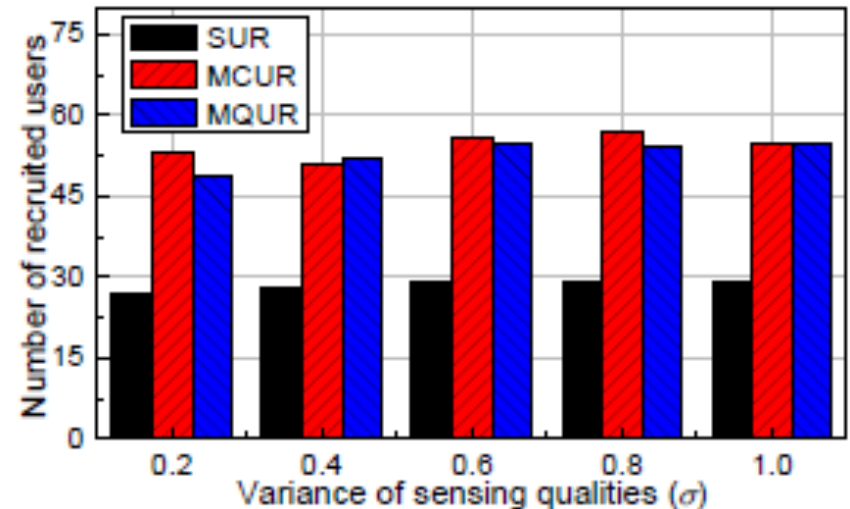
Evaluation

Evaluate the User Recruitment Performance

■ Evaluation Results



(a) $\sigma = 0.4$



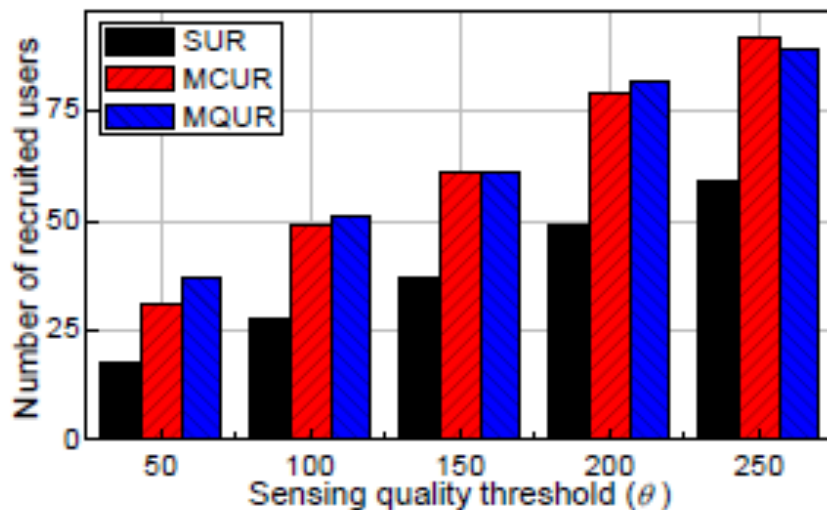
(b) $p = 30$

Number of recruited users vs. average sensing quality and variance

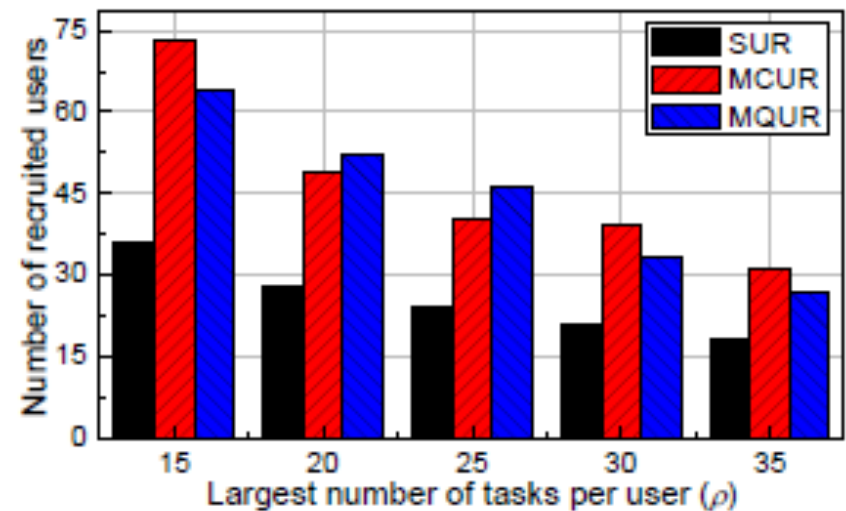
Evaluation

Evaluate the User Recruitment Performance

■ Evaluation Results



(a) $\rho = 20$



(b) $\theta = 100$

Number of recruited users vs. sensing quality threshold and largest number of tasks performed by each user

Evaluation

Evaluate the Time Efficiency

■ Compared Protocols

HEUR: Homomorphic-Encryption-based User Recruitment

GCUR: Garbled-Circuit-based User Recruitment

■ Experiment Settings

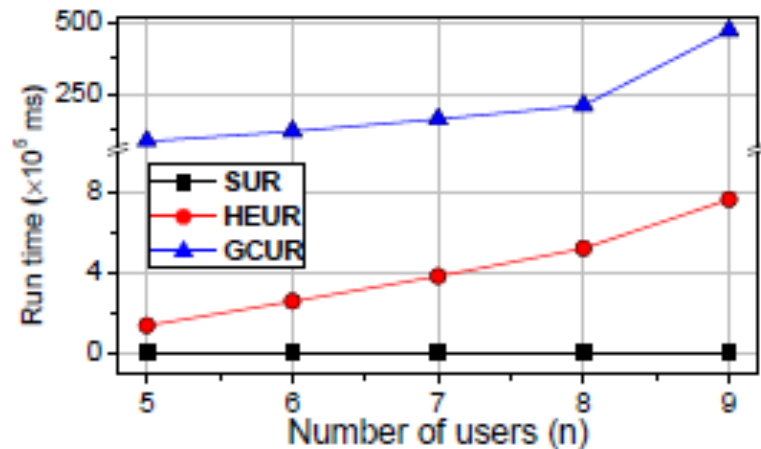
2.0GB memory

a processor of 4-core 2.2GHz plus 4-core 1.5GHz

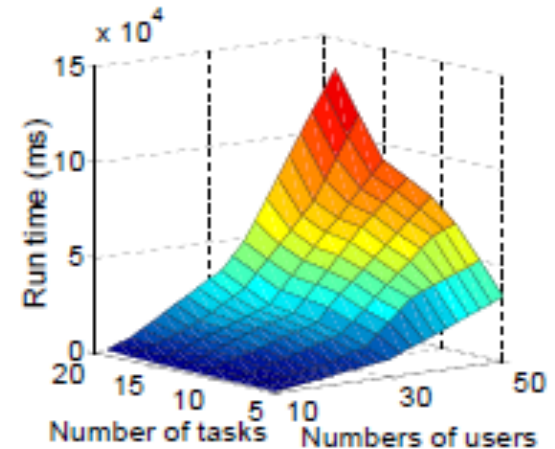
Evaluation

Evaluate the Time Efficiency

■ Experiment Results



(a) Run time of three protocols



(b) Run time of SUR

Evaluation: run time vs. the number of users and tasks

Conclusion

- **SUR can produce a solution with a logarithmic approximation ratio**
- **SUR can protect the inputs of each user from being revealed to the platform or to other users, even if they might collude**
- **Simulation results show that SUR can work well in real smartphones**

Q&A

Thank You!