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# Collaborative Location Privacy with Rational Users

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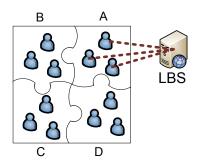
#### **Presentation Outline**

- **1** Introduction
- 2 Two-Agent Game
- Multiple-Agent Game
- Conclusion

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#### **Privacy in Location Based Systems**

- Users have GPS devices
- Communicate wirelessly
- Access location based service (LBS)
- Find shops, restaurants, hotels, etc...



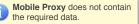
#### **Privacy threat**

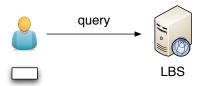
What will the LBS do with the collected data?

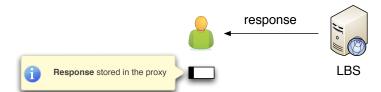


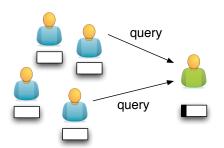














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#### **Sharing also has Costs**

#### **Advantages**

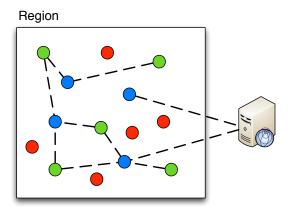
- Fewer queries to LBS
- Users gain privacy

#### **Problems**

- Some users still need to query the LBS
- Sharing data has costs

Are users willing to cooperate by sharing regional data to avoid the privacy threat of connecting to the LBS?

# **System Model**



# Legend:



## System Model

• Cumulative cooperation effort:

# of times agent i receives help – # of times helps others:

$$\phi_i(t) = \operatorname{rc}_i(t) - \operatorname{tr}_i(t)$$

State of an agent:

current role  $\times$  cumulative cooperation effort  $\phi = \phi_i(t)$ :

$$\mathcal{K}_{\phi}, \mathcal{I}_{\phi}, \mathcal{R}_{\phi}$$

Threshold strategy:

*i* cooperates if it is informed and  $\phi_i(t)$  above threshold  $\alpha$ :

$$s_i(t) = \left\{ egin{array}{ll} {\sf cooperate} & {\sf if} \ \phi_i(t) > lpha \wedge i \in \mathcal{I}_{\phi_i(t)} \ {\sf otherwise} \end{array} 
ight.$$

#### **Two-Agent Game**

- 2-agents confined to one region
- Information expires after one game stage
- One agent randomly picked to query the LBS server
- Will it cooperate by sharing this data with its peer?

$$0 \le c_{\text{com}} < c_{\text{srv}} < b_{\text{inf}} \le 1$$



#### **Utility Function**

$$u_i(t) = \left\{egin{array}{ll} b_{ ext{inf}} & s_j(t) = \mathsf{C} \ b_{ ext{inf}} - c_{ ext{srv}} & s_j(t) = \mathsf{D} \ -c_{ ext{com}} & s_i(t) = \mathsf{C} \ 0 & s_i(t) = \mathsf{D} \end{array}
ight.$$

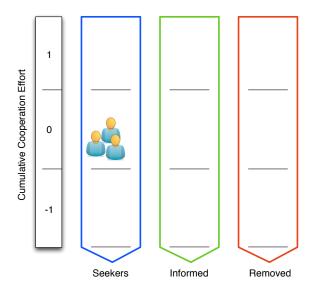
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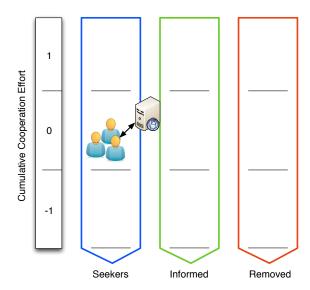
#### **Two-Agent Game Analysis**

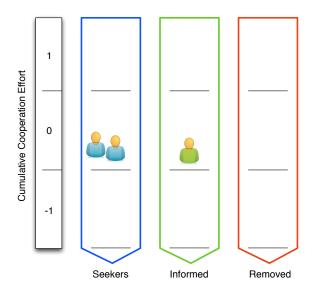
#### **Expected Aggregate Discounted Reward**

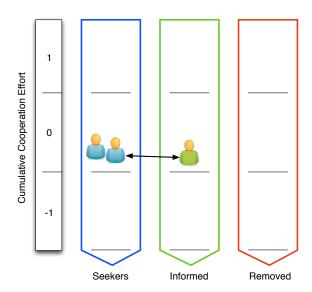
$$U_i^*(lpha) = \left\{ egin{array}{l} rac{1}{1-\delta} \left( b_{\mathsf{inf}} - c_{\mathsf{com}} 
ight) & \mathsf{if} \ lpha \leq -1 \ rac{1}{1-\delta} \left( b_{\mathsf{inf}} - c_{\mathsf{srv}} 
ight) & \mathsf{otherwise} \end{array} 
ight.$$

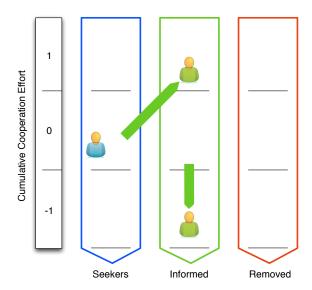
- Nash equilibrium for  $\alpha > 0$ : both defect
- Nash equilibrium for  $\alpha = -1$ : mutual cooperation maximal payoff, minimizes sharing
- Choice of  $\alpha = -1$  is **Pareto** optimal

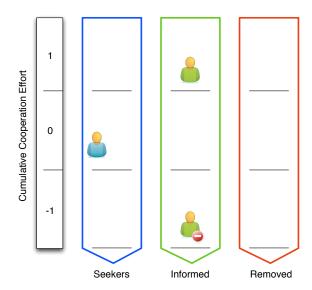


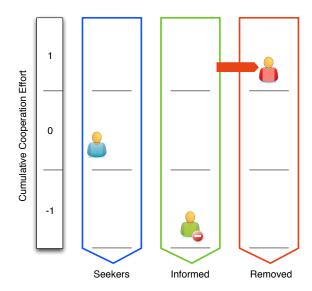


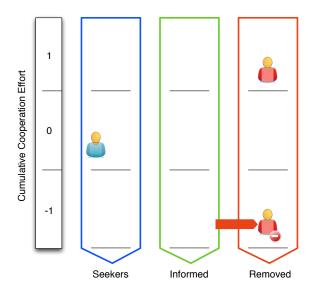


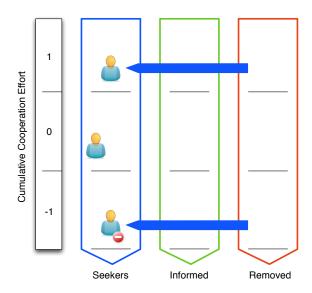


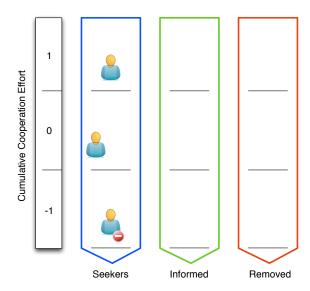








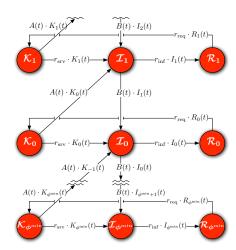




#### **SIR Model for the Multiple-Agent Game**

#### Key

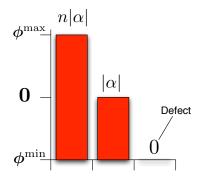
- $\mathcal{X}_{\phi}$ : agent state
- $X_{\phi}(t)$ : fraction agents
- r<sub>srv</sub>: LBS contact rate
- 1/ $r_{inf}$ : info lifetime
- $r_{req}$ : rate  $\mathcal{R} \to \mathcal{K}$
- A(t):  $\propto$  informed agents
- B(t):  $\propto$  seeker agents



#### **Cumulative Cooperation Effort Bounds**

#### **Maximum & Minimum**

- $\phi^{\text{max}} = (1 n)\alpha$
- $\bullet$   $\phi^{\min} = \alpha$

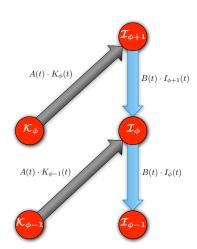


- Amount of cooperation is limited
- Proportional to number of agents n and threshold α
- *i* cooperates, at most,  $n|\alpha|$  times when  $\phi_i(t) = \phi^{\text{max}}$
- At  $\phi^{\min}$  agent defects
- Total agent-states:

$$|\mathcal{S}| = 3(1 - n\alpha)$$

#### **Conservation of Information**

- Total information received
  - Total information shared
- No loss of data
- Lost pkts retransmitted



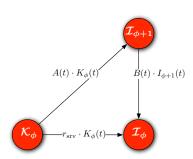
# **Transition Payoffs**

- $\mathcal{K}_{\phi} \to \mathcal{I}_{\phi+1}$ : seekers contact informed peers **Payoff**:  $b_{\text{inf}}$
- $\mathcal{K}_{\phi} \to \mathcal{I}_{\phi}$ : seekers connect to LBS

Payoff:  $b_{inf} - c_{srv}$ 

•  $\mathcal{I}_{\phi+1} \to \mathcal{I}_{\phi}$ : informed agents share data

Payoff:  $-c_{com}$ 



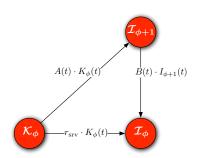
$$0 \le c_{com} < c_{srv} < b_{inf} \le 1$$

### **Total Discounted Payoff**

Benefit b(α, t):

$$R_{\{\mathcal{K}_{\phi} \to \mathcal{I}_{\phi+1}\}}(\alpha, t) \cdot b_{\mathsf{inf}}$$

- +  $R_{\{\mathcal{K}_{\phi} o \mathcal{I}_{\phi}\}}(lpha,t) \cdot (b_{\mathsf{inf}} c_{\mathsf{srv}})$
- Cost  $c(\alpha, t)$ :  $R_{\{\mathcal{I}_{\alpha+1} \to \mathcal{I}_{\alpha}\}}(\alpha, t) \cdot c_{\mathsf{com}}$

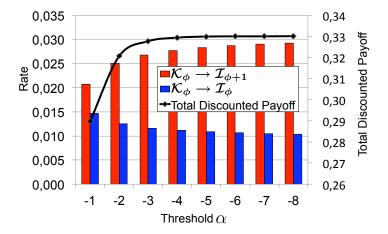


# Total Discounted Payoff

$$U_{\text{total}}^*(\alpha) = \int_0^\infty \delta^t \left( b(\alpha, t) - c(\alpha, t) \right) dt$$

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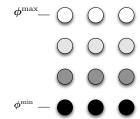
#### **Example Total Discounted Payoff**

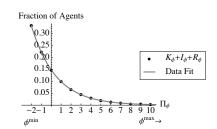


Payoff:  $b_{inf} - c_{srv}$ 

 $\mathcal{K}_{\phi} 
ightarrow \mathcal{I}_{\phi+1}$ Payoff:  $b_{\mathsf{inf}}$ 

#### **Estimating Distribution of Cumulative Cooperation Effort**





# Probability an agent *i* has $\phi_i(t) = \phi$

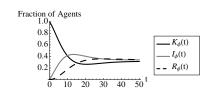
$$\Pr\{\exists_i:\phi_i(t)=\phi\}\approx \frac{1}{1-\alpha}\cdot \frac{1}{(1-1/\alpha)^{\phi-\alpha}}$$

# **Optimal Threshold Maximizes Agent Payoff**

• Agent *i* in state  $\mathcal{X}_{\phi}$ :

$$(1/3) \cdot \Pr\{\exists_i : \phi_i(t) = \phi\}$$

- Estimate rates and total discounted payoff:  $\widehat{U}^*_{\text{total}}(\alpha)$
- An agent's expected total pay:  $\widehat{U}^*(\alpha) = \frac{1}{n} \cdot \widehat{U}^*_{total}(\alpha)$



Homogeneous K, I, R

#### **Optimal Threshold**

$$\alpha^{\text{opt}} = \max\{\alpha \leq -1 : \widehat{U}^*(\alpha - 1) - \widehat{U}^*(\alpha) < \epsilon\}$$

#### Conclusion

#### Goal

Are users willing to share data to gain privacy?

2-agents: Pareto optimal equilibrium favoring cooperation

$$\alpha^{\text{opt}} = -1$$

• *n*-agents: Found  $\alpha$  that maximizes an agent's payoff

$$\alpha^{\mathsf{opt}} < -1$$

To maximize their payoffs, agents benefit by sharing

How will humans behave?

#### **Future Steps**

- Consider threshold strategies with an independent (per agent) threshold
- Experiment with different types of reactive strategies
- Introduce statistical processes describing the rate at which agents enter and leave a region

#### References

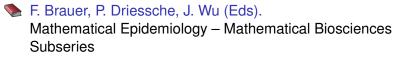


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