Improvement of Estimate Distribution with Local Differential Privacy

Hikaru Horigome and Hiroaki Kikuchi Meiji University

The 19th International Conference on Modeling Decisions for Artificial Intelligence

Background : Location based service and location privacy



Local Differential Privacy (LDP)



Problems of Most Likelihood Estimation (MLE)



Our idea

- Our Approach
 - Expectaion Mazimization (EM) algorithm estimation
- Proposal
 - We apply the EM algorithm to the randomization used in RAPPOR

	Randomization	Estimation
Google RAPPOR[1]	Bloom Filter	MLE
Ours	Randomized Response (RR)	EM

Why EM algorithm works better?

• EM algorithm [Dempster, et al., 1997]

- Iterative process for which posterior probabilities are updated based on Bayes' theorem.



6

How to apply the EM algorithm to RAPPOR

marginal probability $\widehat{\theta}^{(t)} = (\Pr[1, 0, 0], \Pr[0, 0, 1], \Pr[0, 0, 1])$



Research Questions

- RQ1.
 - Dose the proposed method estimate the frequency more accurately than MLE used in RAPPOR?
- RQ2.
 - Is there any correlation between privacy badget ε and the estimation accuracy?



Result(1): Estimation Distribution

Privacy budget $\varepsilon = 0.5$



Result (2) : Mean Absolute Error (MAE)

RQ2. Is there any effect of safety parameter ε on the estimation accuracy?



Regardless of the value of ε , the proposed method estimated more accurately.

Conclusions

- We studied the privacy preservation of time-series location traces using LDP algorithm RAPPOR and proposed the EM for estimating distributions.
- Our experiment using 6,528 individuals' location traces in Tokyo demonstrates that the proposed algorithm performs better than the MLE used in RAPPOR for any privacy budgets ε.