Anonymization of location data does not work: a large-scale measurement study

Sprir

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## CDRs are useful, but ...

- CDRs can be used for various purposes
  - >Marketing
  - >Business
  - >Security
  - >Location based applications and services
  - >Mobility modeling
- Privacy might be breached if such data is not anonymized and handled properly



## Outline

- CDR
- k-anonymity
- Dataset
- Factors impacting size of anonymity sets
  - > different location granularity levels
  - > Distance between locations
  - > geographical regions
  - >extra side knowledge
- Solutions
- >Time domain
- > Spatial domain



## CDR Example

| Attribute       | Value               |
|-----------------|---------------------|
| Mobile ID       | 123-456-7890        |
| Time of call    | 2010 02 02 12 33 02 |
| Call duration   | 300 seconds         |
| Start Cell ID   | 153                 |
| Start Sector ID | 2                   |
| End Cell ID     | 157                 |
| End Sector ID   | 1                   |
| Call direction  | incoming            |
| Caller ID       | 123-456-0987        |



| Attribute       | Value               |
|-----------------|---------------------|
| Mobile ID       | 0000001             |
| Time of call    | 2010 02 02 12 33 02 |
| Call duration   | 300 seconds         |
| Start Cell ID   | 153                 |
| Start Sector ID | 2                   |
| End Cell ID     | 157                 |
| End Sector ID   | 1                   |
| Call direction  | incoming            |
| Caller ID       | 0000002             |

# Simple anonymization for location record Sprint

| Attribute       | Value               | who is                    |
|-----------------|---------------------|---------------------------|
| Mobile ID       | 0000001 -           | 00000001                  |
| Time of call    | 2010 02 02 12 33 02 |                           |
| Call duration   | 300 seconds         |                           |
| Start Cell ID   | 153                 |                           |
| Start Sector ID | 2                   | Re-                       |
| End Cell ID     | 157                 | identification<br>Attacks |
| End Sector ID   | 1                   | Allacks                   |
|                 |                     |                           |
|                 |                     |                           |



Ouasi-

identifier

- Re-identification attacks
- Majority of US population can be uniquely identified by (gender, zipcode, birth-date)

 Anonymity set: individuals with the same (gender, zipcode, birth-date)

Re-identifiable if ||Anonymity Set|| = 1



## K-anonymity

- K-anonymity constraint
  - >At least k individuals have the same quasiidentifier
  - >||Anonymity set|| >= k
  - >E.g. using first 4 digits of zipcode, k = 2
- Our contribution: k-anonymity in location data from cellular networks



### Dataset

- Nation-wide CDR
- Feb April 2010
- 25 M subscribers (subset)
- 30 B records
- >100k locations



## **Quasi-Identifiers**

- Top N locations
- N = 1, 2, 3
- >User x's trace: 1-13, 1-23, 1-13, 1-23, 2-151
- >User x's top locations: 1-13:3, 1-23:2, 2-151:1
- >Anonymity sets:
  - Top 1: 1-13
  - Top 2: 1-13, 1-23
  - Top 3: 1-13, 1-23, 2-151
- Six granularity levels:
- >Sector, cell, zip-code, city, county, state
- For example,
  - >Top 1 location at sector level:1-13-1
  - >Top 3 locations at cell level: 1-13, 1-23, 2-151
  - >Top 2 locations at state level: CA, CA

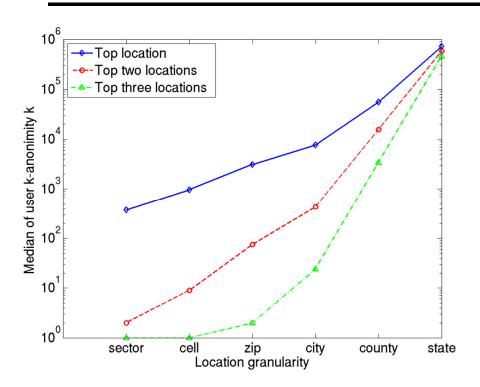


## Factors affecting anonymity

- N
- Location granularity
- Distance between top N locations
- Geographical regions
- Other information



## N & granularity



#### Median of users' k-anonymity at various granularity levels

### Top 1 location

| Location    | Size of anonymity set |                      |                       |         |
|-------------|-----------------------|----------------------|-----------------------|---------|
| granularity | 1 <sup>st</sup> %ile  | 5 <sup>th</sup> %ile | 10 <sup>th</sup> %ile | Median  |
| Sector      | 28                    | 71                   | 111                   | 372     |
| Cell        | 92                    | 220                  | 331                   | 967     |
| Zip code    | 184                   | 557                  | 909                   | 3125    |
| City        | 162                   | 487                  | 874                   | 7638    |
| County      | 802                   | 2972                 | 6272                  | 55649   |
| State       | 60139                 | 1.5e+05              | 2.6e+05               | 7.2e+05 |

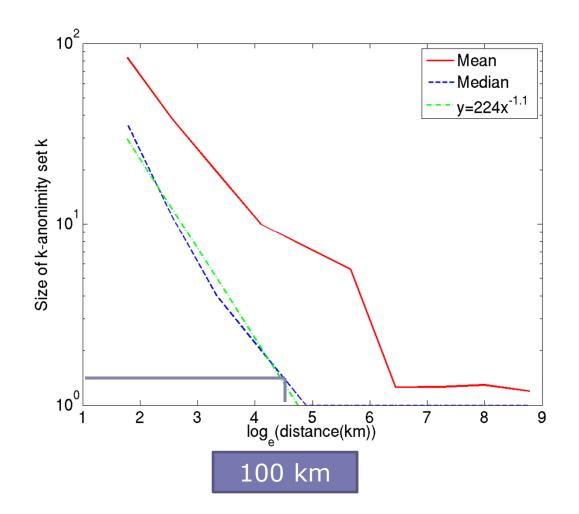
#### Top 2 locations

| Location    | Size of anonymity set |                      |                       |         |
|-------------|-----------------------|----------------------|-----------------------|---------|
| granularity | 1 <sup>st</sup> %ile  | 5 <sup>th</sup> %ile | 10 <sup>th</sup> %ile | Median  |
| Sector      | 1                     | 1                    | 1                     | 2       |
| Cell        | 1                     | 1                    | 1                     | 9       |
| Zip code    | 1                     | 1                    | 2                     | 75      |
| City        | 1                     | 2                    | 6                     | 437     |
| County      | 2                     | 23                   | 143                   | 15628   |
| State       | 530                   | 6912                 | 51291                 | 6.8e+05 |

#### **Top 3 locations**

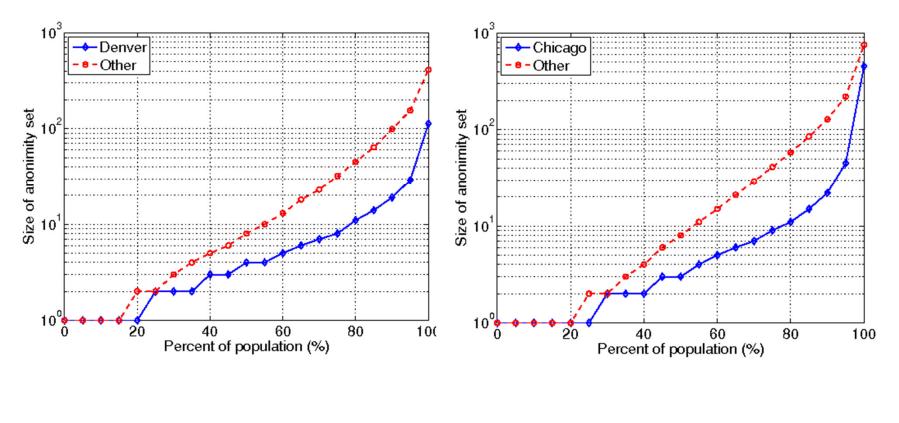
| Location    | Size of anonymity set |                       |                        |         |
|-------------|-----------------------|-----------------------|------------------------|---------|
| granularity | 1 <sup>st</sup> %tile | 5 <sup>th</sup> %tile | 10 <sup>th</sup> %tile | Median  |
| Sector      | 1                     | 1                     | 1                      | 1       |
| Cell        | 1                     | 1                     | 1                      | 1       |
| Zip code    | 1                     | 1                     | 1                      | 2       |
| City        | 1                     | 1                     | 1                      | 24      |
| County      | 1                     | 2                     | 7                      | 3407    |
| State       | 40                    | 1074                  | 5671                   | 4.6e+05 |







## Geographical Regions – urban vs. rural



Colorado

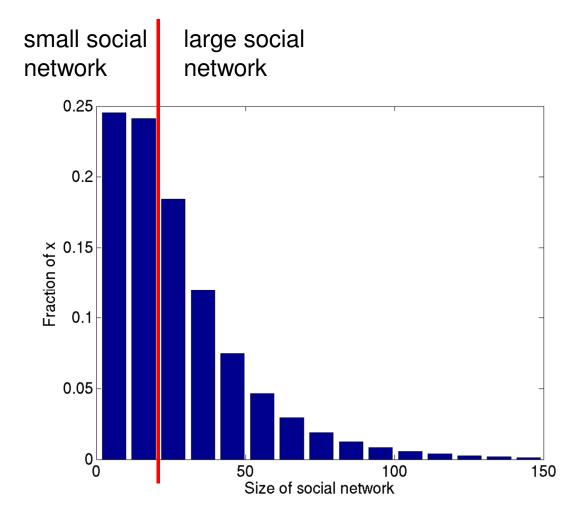
Illinois



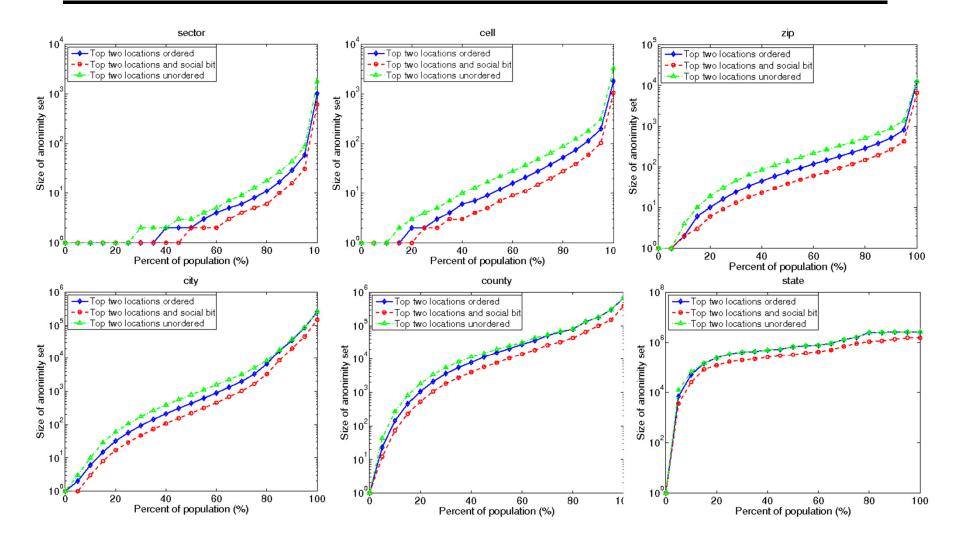
| Attribute       | Value               |
|-----------------|---------------------|
| Mobile ID       | 0000001             |
| Time of call    | 2010 02 02 12 33 02 |
| Call duration   | 300 seconds         |
| Start Cell ID   | 153                 |
| Start Sector ID | 2                   |
| End Cell ID     | 157                 |
| End Sector ID   | 1                   |
| Call direction  | incoming            |
| Caller ID       | 0000002             |



## Extra side information



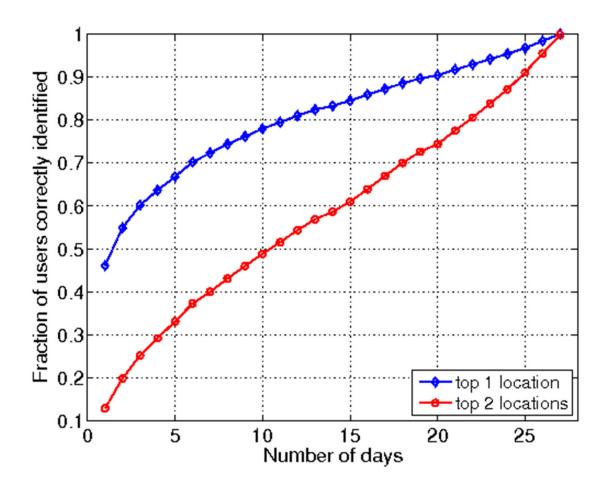
## Extra side information



Red curves: size of anonymity sets reduces by half

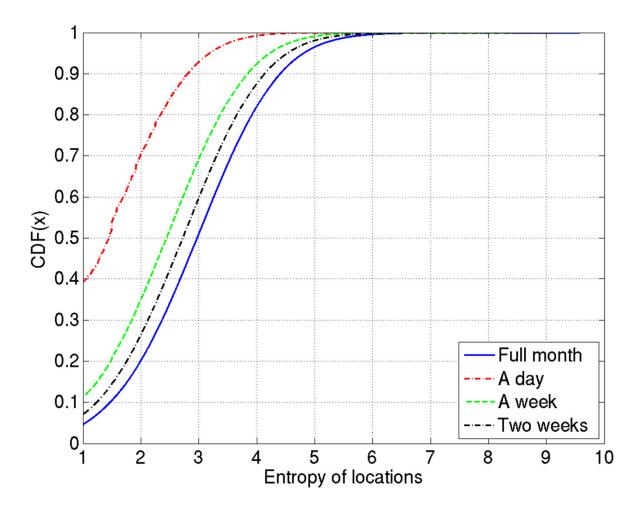
- Spatial and time domain solutions:
  Publish traces at zip-code granularity or above
  Publish short traces, such as a day
- Reduction of utility of published traces
  Mobility modeling
  - >Identifying preferred locations

## Fraction of users whose top locations are<sub>sprint</sub> <u>correctly extracted</u>



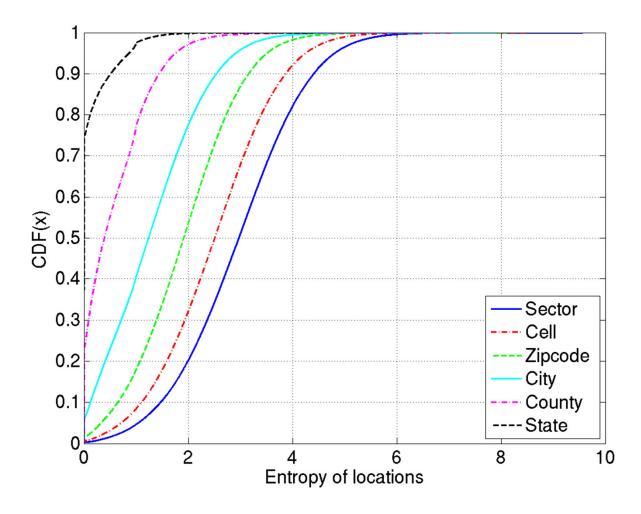


## Entropy of traces of different durations





## Entropy of traces of different location granularity





## Conclusions

- Availability of large scale cell phone data has enabled and will continue to enable a wide range of new services and applications
- Cell phone data are economically valuable
- Subscribers' privacy is at risk if such data is not anonymized and handled properly
- >Anonymity depends on N, granularity, geographical regions, etc.
- >Time domain and spatial domain approaches are proposed to improve anonymization



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