

# Data Protection & Privacy

## First Homework: k-Anonymity

October 12, 2018

### Goal

The goal of this homework is to implement the k-anonymity algorithm depicted in fig. 1 which is based on the approach presented in [1]. Feel free to implement the algorithm in your favourite programming language.

<p><b>Input:</b> Private Table <b>PT</b>; quasi-identifier <math>QI = (A_1, \dots, A_n)</math>, <math>k</math> constraint; hierarchies <math>DGH_{A_i}</math>, where <math>i=1, \dots, n</math>.</p> <p><b>Output:</b> <b>MGT</b>, a generalization of <math>PT[QI]</math> with respect to <math>k</math></p> <p><b>Assumes:</b> <math> PT  \geq k</math></p> <p><b>Method:</b></p> <ol style="list-style-type: none"><li>1. <math>freq \leftarrow</math> a frequency list contains distinct sequences of values of <math>PT[QI]</math>, along with the number of occurrences of each sequence.</li><li>2. <b>while there exists</b> sequences in <math>freq</math> occurring less than <math>k</math> times that account for more than <math>k</math> tuples <b>do</b><ol style="list-style-type: none"><li>2.1. <b>let</b> <math>A_j</math> be attribute in <math>freq</math> having the most number of distinct values</li><li>2.2. <math>freq \leftarrow</math> generalize the values of <math>A_j</math> in <math>freq</math></li></ol></li><li>3. <math>freq \leftarrow</math> suppress sequences in <math>freq</math> occurring less than <math>k</math> times.</li><li>4. <math>freq \leftarrow</math> enforce <math>k</math> requirement on suppressed tuples in <math>freq</math>.</li><li>5. <b>Return</b> <b>MGT</b> <math>\leftarrow</math> construct table from <math>freq</math></li></ol>
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Figure 1: Datafly algorithm

Where:

- PT : Table to anonymize.
- DGH : Domain generalization hierarchies.
- MGT: Generalization of PT that satisfies k-anonymity.

### Dataset

In the homework folder there are 3 databases: `db_10000.csv`, `db_50000.csv` and `db_100000.csv` which contain 10K, 50K and 100K records, respectively. The attributes of each database are:

- 1) `id` that is the EI.

2) `age`, `city of birth` and `zip code` that are QI.

3) `disease` that is the SD.

All the generalizations of the QI are also provided in the folder `:age_generalization.csv`, `city_generalization.csv` and `zip_code_generalization.csv`.

Such files provide a domain generalization hierarchy in the following format:

```
level_0_generalization, level_1_generalization, level_2_generalization
....
```

For example, for the `zip code` the following domain generalization is provided:  
`67002`, `6700*`, `670**`, `67***`, `6****`.

## Output

You are required to:

1. implement the algorithm depicted in fig. 1;
2. test your implementation by executing the algorithm for several values of  $k$  on all three databases.
3. for each test, count the number of equivalence classes/clusters which have homogeneity.
4. evaluate the relationship between the value of  $k$  and the number of homogeneous equivalence classes.

## Contacts

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## References

- [1] L. SWEENEY, “ACHIEVING k-ANONYMITY PRIVACY PROTECTION USING GENERALIZATION AND SUPPRESSION,” *International Journal of Uncertainty, Fuzziness and Knowledge-Based Systems*, vol. 10, no. 05, pp. 571–588, 2002. [Online]. Available: <http://www.worldscientific.com/doi/abs/10.1142/S021848850200165X>