EN4720: Security in Cyber-Physical Systems Exercise — Privacy

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This is an individual exercise!

Due Date: 24 June 2023 by 11.59 PM

Section 1

Provide answers to questions 1 through 7 in the given space referring to the Tables 1 and 2 given below.

Table 1.	Example	dataset	on the	weekly	usage of	taxis by	y certain individuals	2
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Name	Age	Gender	Occupation	Average No. of taxi
				trips per week
943145	21	Female	Legal Counsel	15
416765	38	Male	Data Privacy Officer	2
356891	44	Female	Database Administrator	3
723145	25	Female	Administrative Assistant	1
239976	31	Male	Data Privacy Officer	5
562396	42	Female	Programmer	3
964825	22	Female	Administrative Assistant	4
873892	30	Female	Legal Counsel	2

1. What type of data anonymization technique is used for the dataset given in Table 1?

De-identification: actual names of the individuals have been replaced by a numerical identifier. Because, the attribute 'Name' of a data record is a direct identifier which can be used to identify a given individual uniquely.

2. Is this technique sufficient to protect the privacy of the associated individuals? If not, why?

No. Removing only the name can not guarantee the preservation of privacy, because it can be **susceptible to linkage attacks** which leads to re-identification of the individuals by linking the information with other external information (zip, sex, birth date).

- 3. Briefly describe three other data anonymization techniques that can be used to anonymize data.
 - Generalization: Rather than including the specific information about an attribute (eg: age = 21) in a record, it can be replaced with more generalized value (eg: range of ages = (20 to 30)).

- Suppression: Some fields of the data base can be entirely removed, if that field contains highly unique identifiers such as national ID card numbers and telephone numbers of the individuals.
- Data swapping: some of the values of a given record can be swapped with some other record's value while keeping the statistical properties of the database intact.

Table 2: Modified example dataset on the weekly usage of taxis by certain individuals

Age	Gender	Occupation	Average No. of taxi
			trips per week
21 to 30	Female	Legal Counsel	15
31 to 40	Male	Data Privacy Officer	2
41 to 50	Female	IT	3
21 to 30	Female	Administrative Assistant	1
31 to 40	Male	Data Privacy Officer	5
41 to 50	Female	IT	3
21 to 30	Female	Administrative Assistant	4
21 to 30	Female	Legal Counsel	2

- 4. The example dataset given in Table 1 was modified to improve anonymity. The new dataset is provided in Table 2. Mention all the data anonymization techniques that were used to achieve this.
 - Generalization: ages have been replaced with ranges of ages, some value of the occupation filed hs been replaced with its parent set's value IT = {Database Administrator, Programmer, ...}
 - Suppression: Name field has been removed entirely
- 5. Can k-anonymity be observed in the data given in Table 2? If so, what is the value of k?

Yes.
$$K = 2$$

6. Is privacy guaranteed for the data given in Table 2? Justify your answer.

Although the above methods reduce the risk of re-identification, it is not possible to grantee the preservation of privacy of data. The database can be still susceptible to attribute disclosure attacks, if a given equivalence class (the group of k records with the same quasi-identifiers) has the same value for the given sensitive attribute. Let's assume the sensitive attribute in the Table 2 is the Average No. of taxi trips per week. Then if we consider the 3rd and 6th records in that table they have the same value for the sensitive attribute.

7. Calculate the risk of re-identification for data given in Table 2 (mention as a percentage). Justify your answer.

Consider below tables with rearranged records.

Table 3: Original dataset

Name	Age	Gender	Occupation	Average No. of taxi
				trips per week
943145	21	Female	Legal Counsel	15
873892	30	Female	Legal Counsel	2
416765	38	Male	Data Privacy Officer	2
239976	31	Male	Data Privacy Officer	5
356891	44	Female	Database Administrator	3
562396	42	Female	Programmer	3
723145	25	Female	Administrative Assistant	1
964825	22	Female	Administrative Assistant	4

Table 4: Rearranged example dataset

Age	Gender	Occupation	Average No. of taxi trips per
			week
21 to 30	Female	Legal Counsel	15
21 to 30	Female	Legal Counsel	2
31 to 40	Male	Data Privacy Officer	2
31 to 40	Male	Data Privacy Officer	5
41 to 50	Female	IT	3
41 to 50	Female	IT	3
21 to 30	Female	Administrative Assistant	1
21 to 30	Female	Administrative Assistant	4

8. Suggest ways to enhance the privacy of this dataset considering l-diversity.

L-diversity ensures that withing each equivalence group (quasi-identifier group) there are at least l distinct values for each attribute. This make it harder to learn the sensitive attribute through attribute disclosure attacks. In order to achieve l-diversity, generalization and suppression techniques has to be used on the original data set recursively until we get l distinct values inside every quasi-identifier group.

Section 2

1. What is differential privacy?

Your answer here

2. Briefly describe what is **protected** and **not protected** by applying differential privacy.

Your answer here

3. The following mechanism is used to satisfy differential privacy where Z denotes the noise added.

$$F(x) = f(x) + Z$$

(a) Provide an equation to show the relationship between ϵ (privacy budget) and Z? Are they directly or inversely related?

Your answer here

(b) What does it mean to have $\epsilon = 0$?

Your answer here

(c) What would be the result of adding a larger amount of noise?

Your answer here