Differential privacy

DATA PRIVACY AND ANONYMIZATION IN R



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Why differential privacy

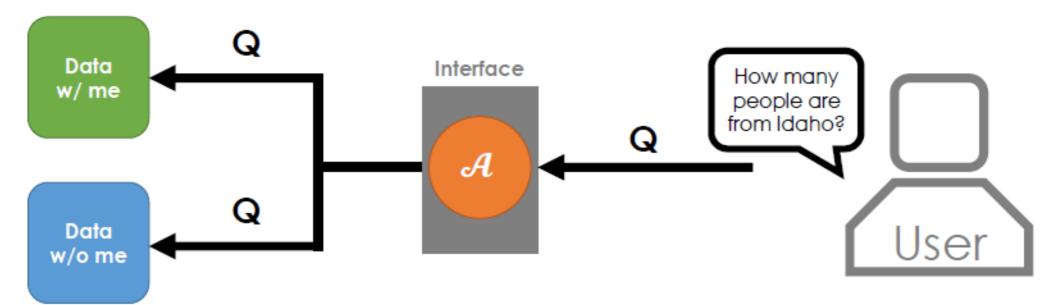
- Quantifies privacy loss via a privacy budget
- Assumes worst-case scenario; no assumptions about the data intruder



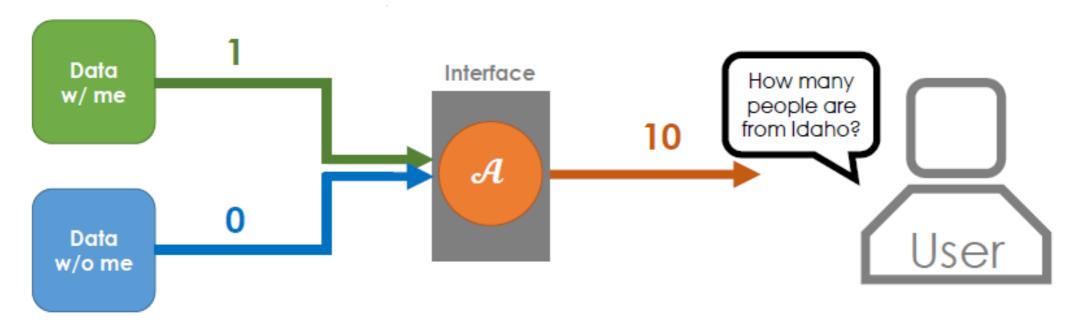
Epsilon, the privacy budget



Differential privacy: general concept

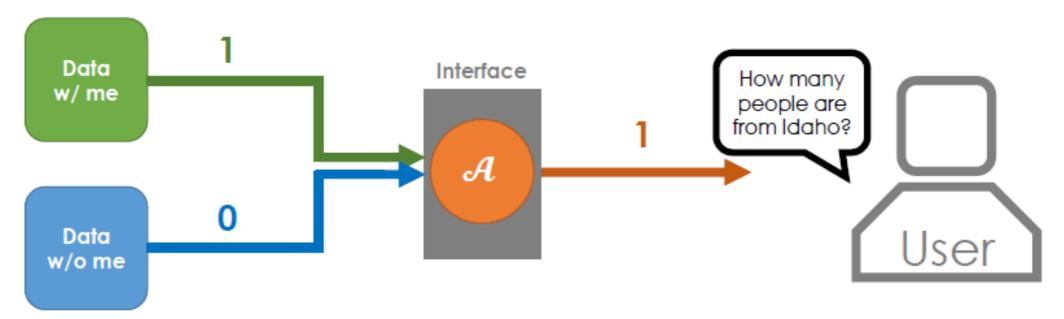


Differential privacy: small privacy budget



 Smaller privacy budget means less information or a noiser answer.

Differential privacy: large privacy budget



 Larger privacy budget means more information or a more accurate answer.

Let's practice!

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Global sensitivity

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Global sensitivity of counting queries



Global sensitivity of other queries

- n is total number of observations
- ullet a is the lower bound of the data
- ullet b is the upper bound of the data

- Counting: 1
- Proportion: 1/n
- Mean: (b-a)/n
- Variance: $(b-a)^2/n$

Global sensitivity and noise

- Small global sensitivity results in less noise
- Large global sensitivity results in more noise

Let's practice!

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Laplace mechanism

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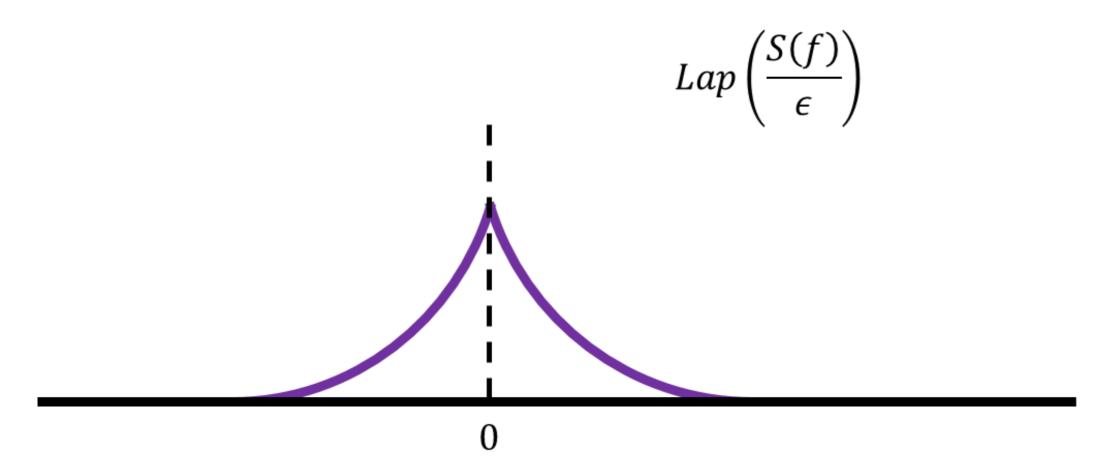


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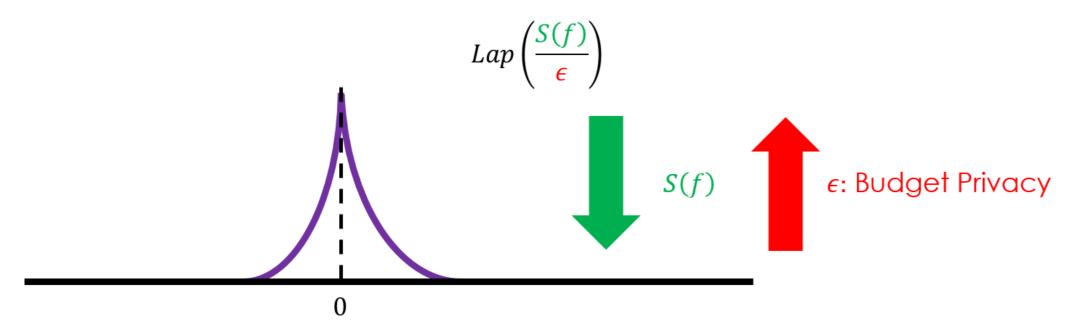
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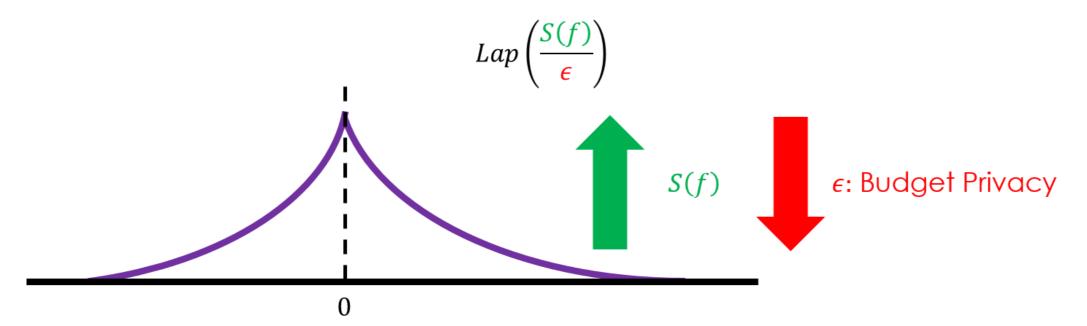
Laplace mechanism Part I



Laplace mechanism Part II



Laplace mechanism Part III



Coding the Laplace mechanism

```
library(dplyr)
fertility %>%
    summarize_at(vars(Child_Disease), sum)
```

```
library(smoothmest)
# rdoublex(draws, mean, shaping)
set.seed(42)
rdoublex(1, 87, 1 / 10)
```

```
set.seed(42)
rdoublex(1, 87, 1 / 0.1)
```

87.01983

88.98337

Let's practice!

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