

Privacy-preserving data analytics via neural network models



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Problem statement

Data analytics can provide valuable knowledge to companies that leverage collected data to derive relevant information. However, processed data are often **highly sensitive** and thus their disclosure may harm **individuals' privacy**. Nowadays, the current European **General Data Protection Regulation** (GDPR) represents a major challenge for companies, as they are required to follow a privacy-by-design approach to protect data and yet allow their processing.

Proposed solution

We developed dedicated **privacy-preserving data analytics modules** able to extract analytics on **protected** data, via the application of **artificial intelligence**. This approach ensures data subjects' privacy while still being **cost-effective** and **accurate**.

PAPAYA and the General Data Protection Regulation

Advances in technology and the capability of big data analytics and artificial intelligence have made it easier to create **profiles** and make **automated decisions**. Article 29 Data Protection Working Party, on its "Guidelines on Automated Individual decision-making and Profiling", reports that "profiling and automated decision-making can pose significant risks for individuals' rights and freedoms which require **appropriate safeguards**".

How PAPAYA addresses the challenge



Explicit consent

Data subjects have the ability to give or withdraw consent



Security measures

PETs are employed to extract analytics from subjects' data



Transparency

Data subjects can visualize the disclosed data and their rights



Auditability

Data controllers can visualize audit logs and handle DPIA

Applications in the healthcare field

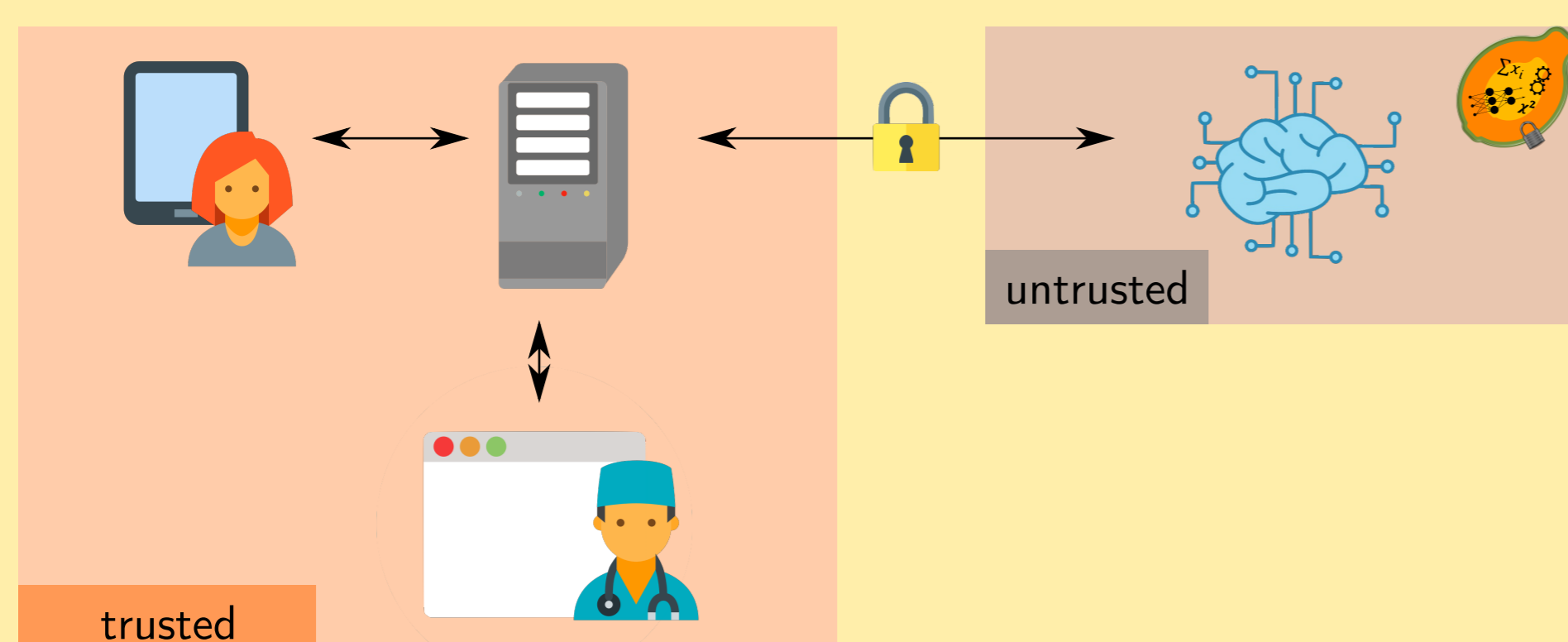
Single data owner

The data controller needs to extract data analytics from his sensitive data in a privacy-preserving manner, thus he applies adequate **protection** (e.g., encryption) beforehand, and then outsources the task to a **third-party data processor** (e.g., the PAPAYA platform).

Advantages:

- data subjects' privacy is preserved
- the computational burden is on the data processors

Use case: ECG analysis



Multiple data owners

Multiple data controllers would like to perform analytical tasks over their **combined datasets**, disclosing the extracted analytics and keeping protected their original datasets. Thus, each controller **protects** his dataset (e.g., via encryption) before outsourcing it for analytics extraction.

Advantages:

- data subjects' privacy is preserved
- better models are built on larger datasets

Use case: Stress management

