



Cyber
Security
for Europe

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Research on Data Sharing in Open Banking & Medical Data Exchange

Personal Data Sharing - Emerging Technologies

Brussels, 7 October 2022

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Joint work with David Goodman, Juan Carlos Perez Baun, ...

CyberSec4Europe



More than **€63.5 million** invested in **4 projects**

CONCORDIA
Cyber security eCompetence Centres for Research and Innovation

Partners: **46**

EU Member States involved: **14**

Key words

SME & startup ecosystem
Ecosystem for education
Socio-economic aspects of security
Virtual labs and services
Threat Intelligence for Europe
DDoS Clearing House for Europe
AI for cybersecurity
Post-Quantum cryptography

Cyber Security for Europe

Partners: **43**

EU Member States involved: **20**

Key words

Cybersecurity for citizens
Application cases
Research Governance
Cyber Range
Cybersecurity certification
Training in security

ECH

Partners: **30**

EU Member States involved: **15**

Key words

Network of Cybersecurity centres
Cyber Range
Cybersecurity demonstration cases
Cyber-skills Framework
Cybersecurity certification
Cybersecurity early warning

SPARTA

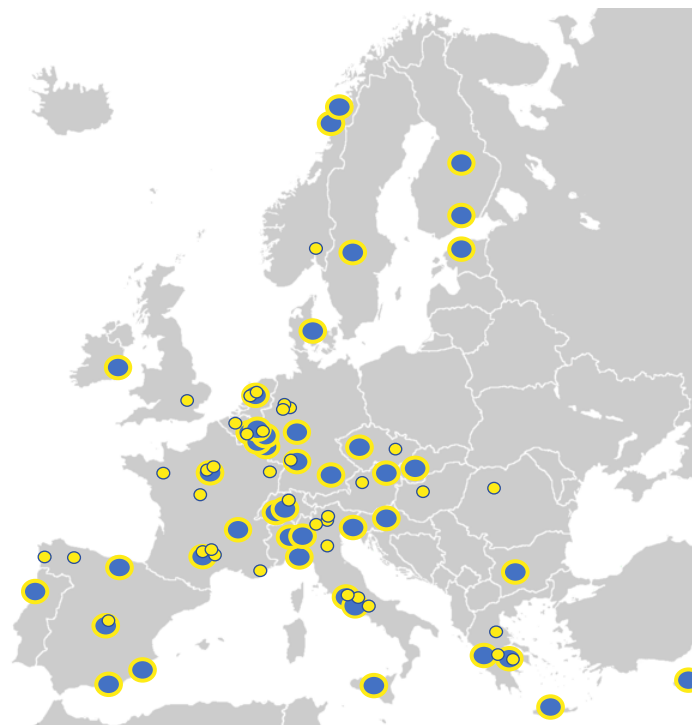
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Key words

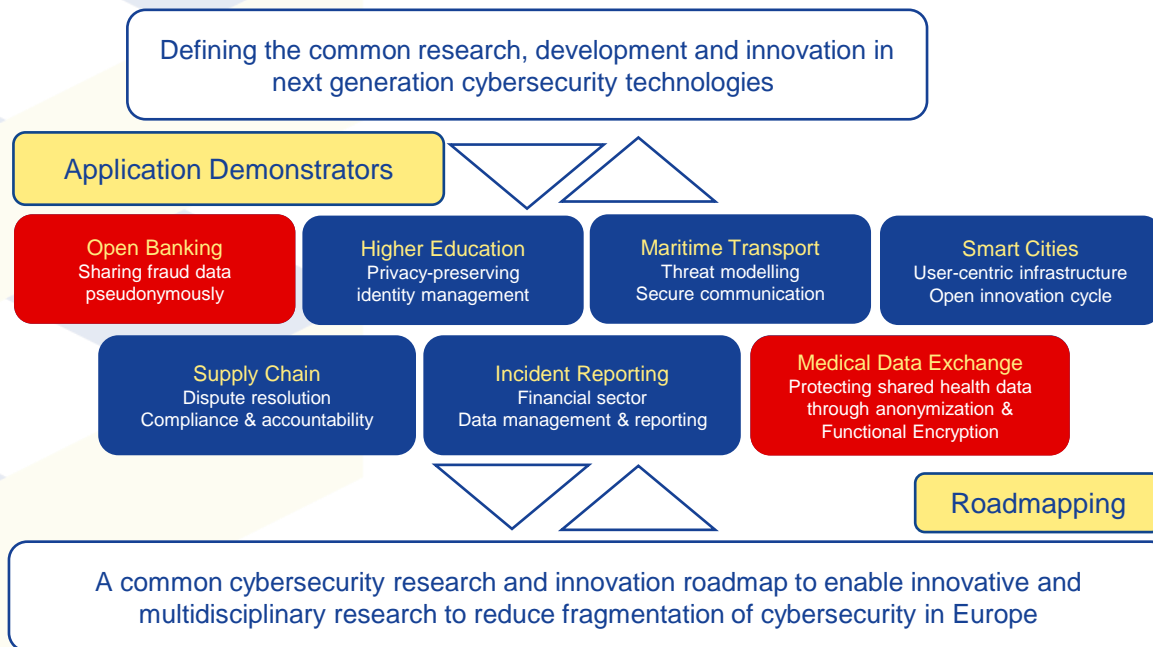
Research Governance
Cybersecurity skills
Cybersecurity certification
Community engagement
International cooperation
Strategic Autonomy

Logo updated 8 March 2019



● Partner ● Associate

From Research to Innovation to Industry





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Open Banking

OBSIDIAN

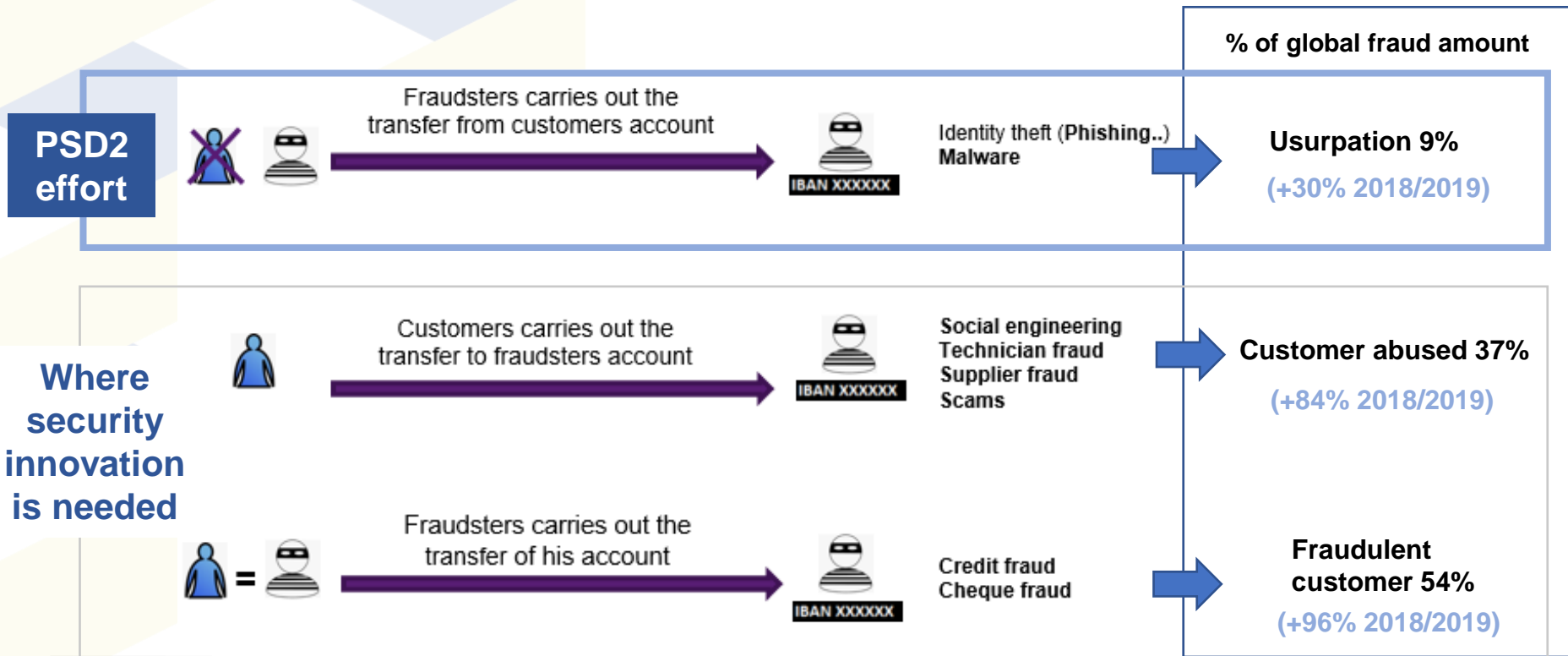
Open Banking Sensitive Data Sharing Network for Europe



- High incidence of fraud regularly cost financial institutions – and their customers – very large sums of money.
- OBSIDIAN is a pilot **data sharing network** which supports the fight against fraud by **sharing false IBANs between banks** and is an effective approach to **detecting money laundering or terrorist financing**
- Participating banks experiencing potential fraud attempts will request confirmation of the right decision to take
 - The requests provide the core transaction data (IBAN et al) in a format which guarantees privacy and security network requirements.

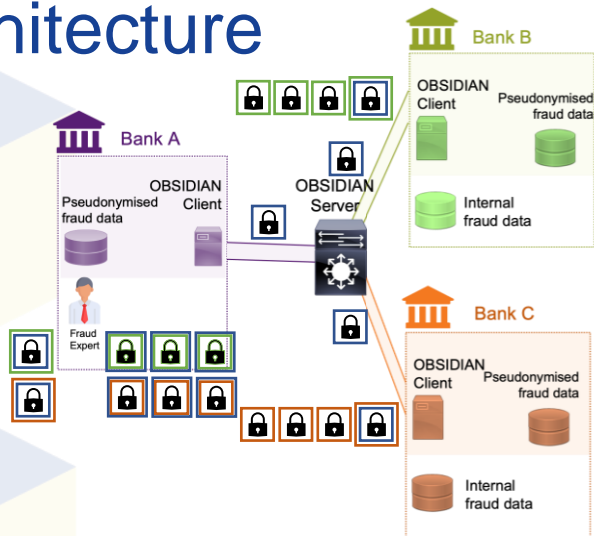
Fraud Evolution

Why sharing fraud information is a good idea



OBSIDIAN – Architecture

- ❖ A fraud expert (or system) detects a suspicious transaction and uses the OBSIDIAN network to check the beneficiary's IBAN
- ❖ The OBSIDIAN client applies a pseudonymisation and sends the request to the server
- ❖ OBSIDIAN server broadcasts Bank A requests to the other network participants (Banks B and C)
- ❖ Each bank receiving the request re-randomizes the request



- ❖ Banks that received the request send back the re-randomized request and pseudonyms on fraudulent accounts
- ❖ The server relays the responses back to Bank A
- ❖ Bank A checks for matches

Reasons why banking secrecy is protected:

- ✓ Banks B and C don't know the request came from Bank A
- ✓ Bank A cannot identify the origin of the responses
- ✓ Banks B and C do not know the result of the request



In the exchange, no one knows who exchanged information with whom

OBSIDIAN – Innovation

Regulatory compliance



- ✓ GDPR Compliant
- ✓ Banking secrecy protected

International applicability



- ✓ Applicable to the countries with the most restrictive secrecy laws
- ✓ Can easily be expanded into a European network

Data never handed over



- ✓ The OBSIDIAN server **does not store fraud data**
- ✓ IBANs are always pseudonymised when exchanged
- ✓ Banks can take back their data whenever necessary (GDPR right to erasure)

Easy integration



- ✓ No complicated mathematics : technology easily understood by IT experts
- ✓ Simple integration : one client connected to one server only

OBSIDIAN – Collaborating Banks

- An OBSIDIAN network has been piloted with 8 major French banks under the umbrella of the French Banking Federation
- The crucial non-technical issues that OBSIDIAN addresses are:
 - GDPR compliance
 - Banking secrecy

One of the issues is that most banks are not motivated to share their (data) assets and are innately conservative

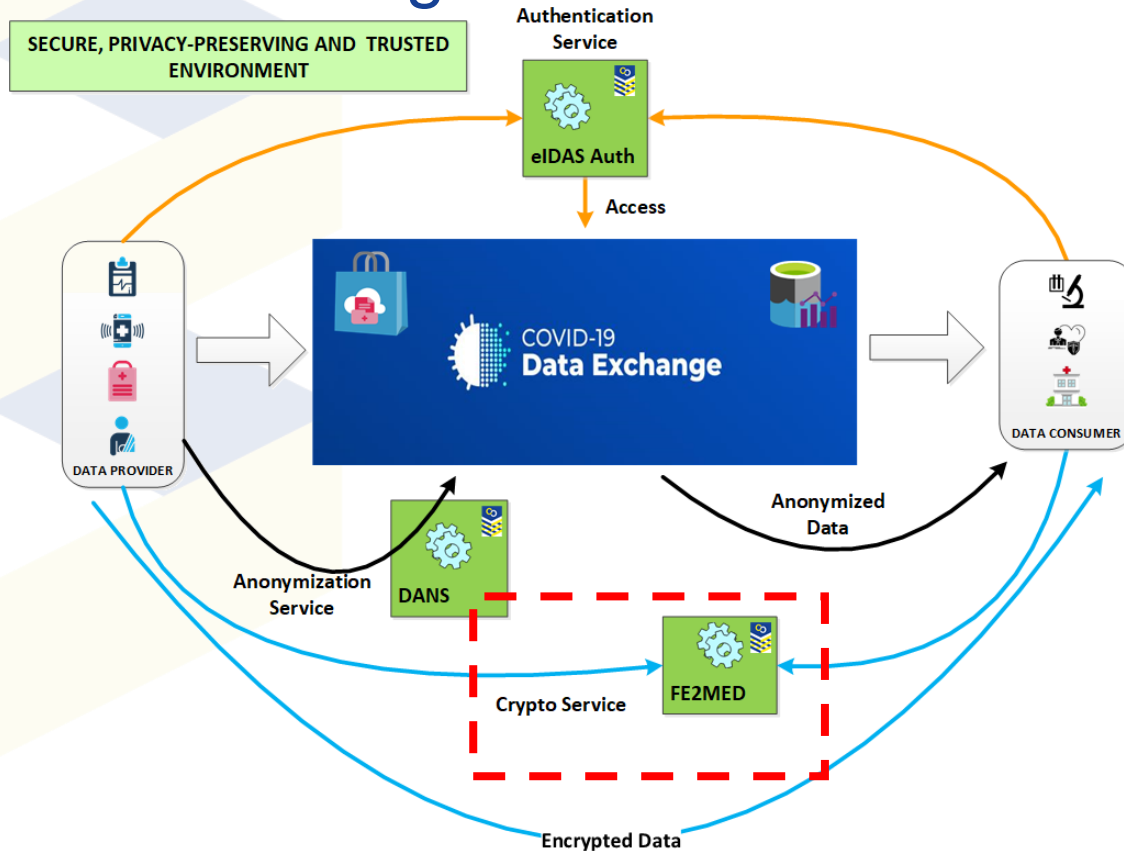
However, the underlying premise – bank fraud – is getting worse, year-on-year and it is recognised that compliant data sharing is the way forward



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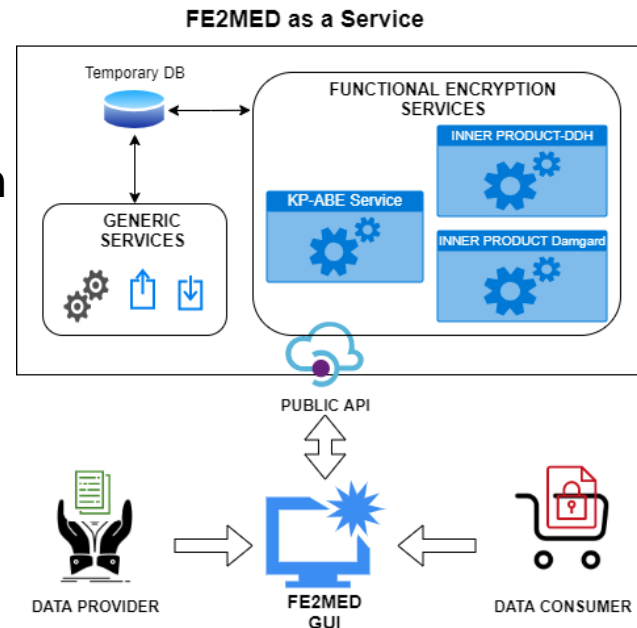
Medical Data Sharing

Medical Data Exchange



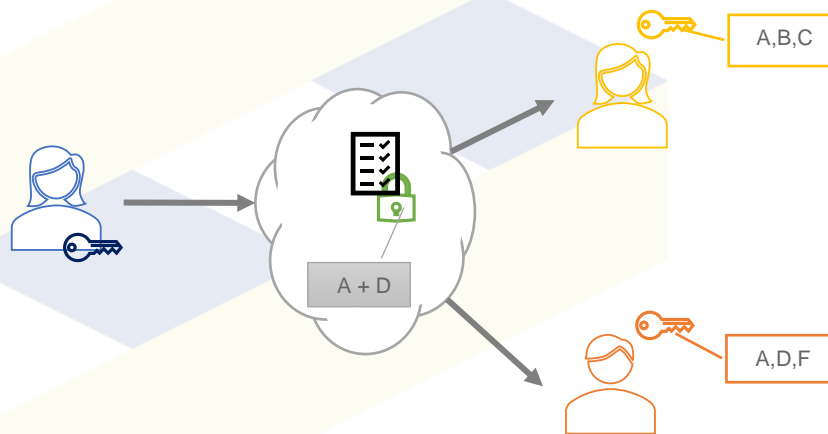
FE2MED (Functional Encryption To Medical Data)

- Secure personal and sensitive data
- Ensure confidentiality and data integrity
- The encrypted data are only accessible to certain allowed users
- Leverage two advanced primitives:
 - Attribute-based encryption
 - Functional encryption

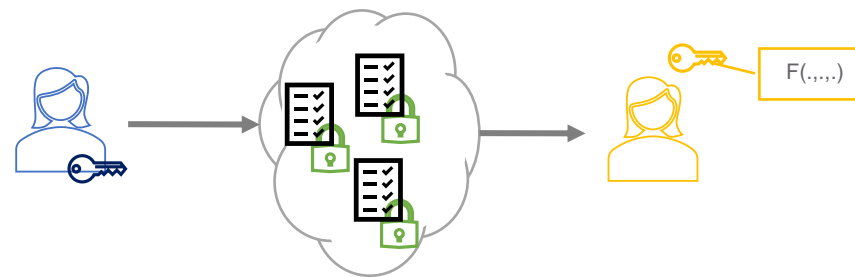


Advanced Encryption Schemes

Attribute-Based Encryption

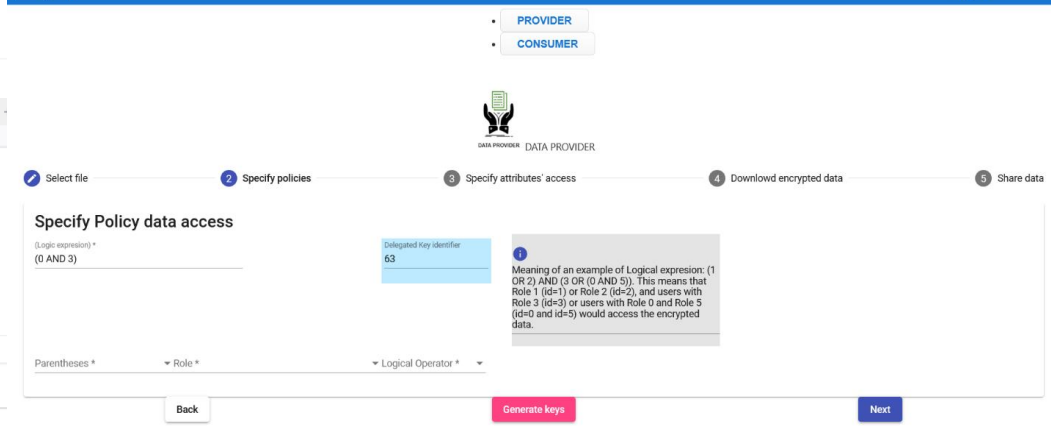
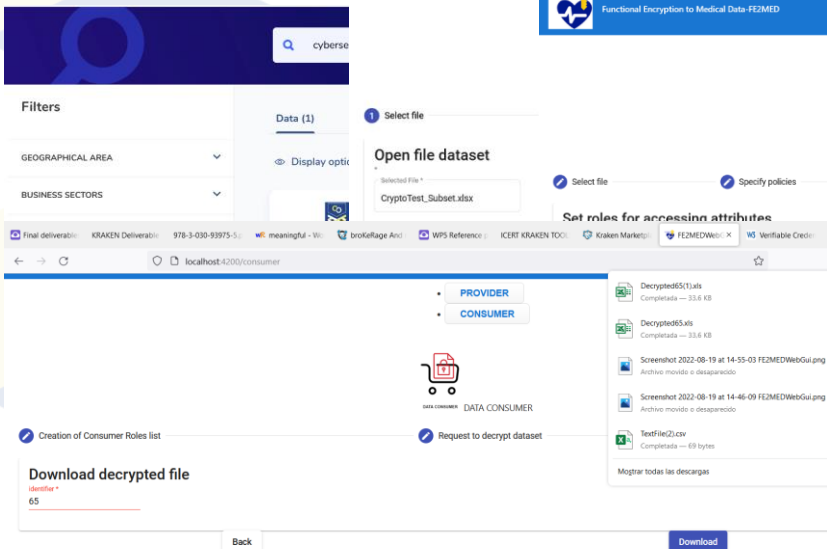


Functional Encryption



FE2MED GUI

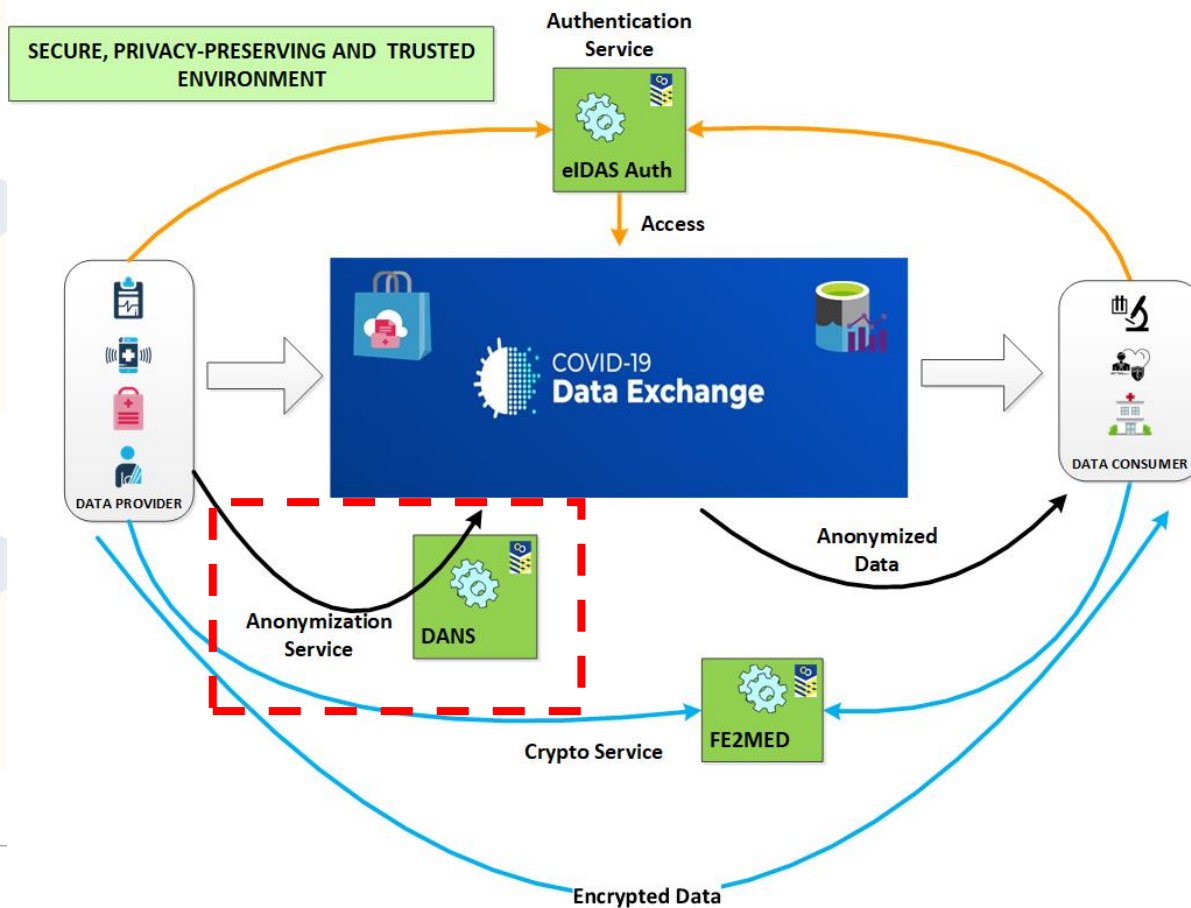
Screenshots



Please, check your Downloads folder for getting the decrypted Decrypted65.xls file



Data Anonymization



Data Anonymization

- ***k*-anonymity**

- Each quasi-identifier tuple has at least *k* records in the anonymized dataset
- Concepts: suppression, generalization

- ***l*-diversity**

- For each set of rows with identical quasi-identifiers, there are at least *l* distinct values for each sensitive attributes

Name	Age	Gender	Disease
Alice	31	F	Cancer
Bob	28	M	No illness
Charlie	34	M	Heart-related
Dan	38	M	Cancer



Name	Age	Gender	Disease
*	31-40	F	Cancer
*	21-30	M	No illness
*	31-40	M	Heart-related
*	31-40	M	Cancer

DANS GUI

Screenshots



Data ANonymisation Service-DANS



1 Select... 2 Select at... 3 Attribute ... 4 Quasi-Identifying... 5 Attributes ... 6 Privacy ... 7 Anonymisation ... 8 Conne...

Open file dataset

Selected File
MYsample_03forDAN!

Data ANonymisation Service-DANS



1 Select file 2 Select attribute 3 Attribute format 4 Quasi-Identifying attribute 5 Attributes Resume 6 Privacy Models 7 Anonymisation Resume 8 Connect DEP

Quasi-Identifying attribute: age_label

Data ANonymisation Service-DANS



1 Select file 2 Select attribute 3 Attribute format 4 Quasi-Identifying attribute 5 Attributes Resume 6 Privacy Models 7 Anonymisation Resume 8 Connect DEP

k-anonymity

k parameter *
5
Set k-anonymity value

l-diversity

l parameter *
3
Set l-diversity value

Back

Next

Name	Data Type	Format Type	Attribute Type	Transformation Method	Hierarchy file	k Anonymity	l Diversity
age	integer	string	quasi-identifying	Generalization	adult_hierarchy_age.csv		
disease	string	string	sensitive			5	3

This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No. 830929.

Cyber Security for Europe project. The content of this website reflects only the consortium view.



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Thank you!

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Save the date:

MOMENTUM!

1 – 2 December, 2022

Brussels

<https://cybersec4europe.eu/>