

Simplifying Differential Privacy for Non-Experts

The ENCRYPT Project Approach



Enhancing Data Security and Utility Across Federated Data Spaces

Overview of Data Privacy Challenges

- **Data Explosion:** Massive amounts of data driving research and innovation
- **Privacy Risks:** Cybersecurity threats and data misuse
- **Regulatory Compliance:** Importance of adhering to GDPR and other regulations
- **Public Concerns:** Growing demand for stronger privacy measures
- **ENCRYPT Project:** Aims to enhance data security and privacy across federated data spaces.



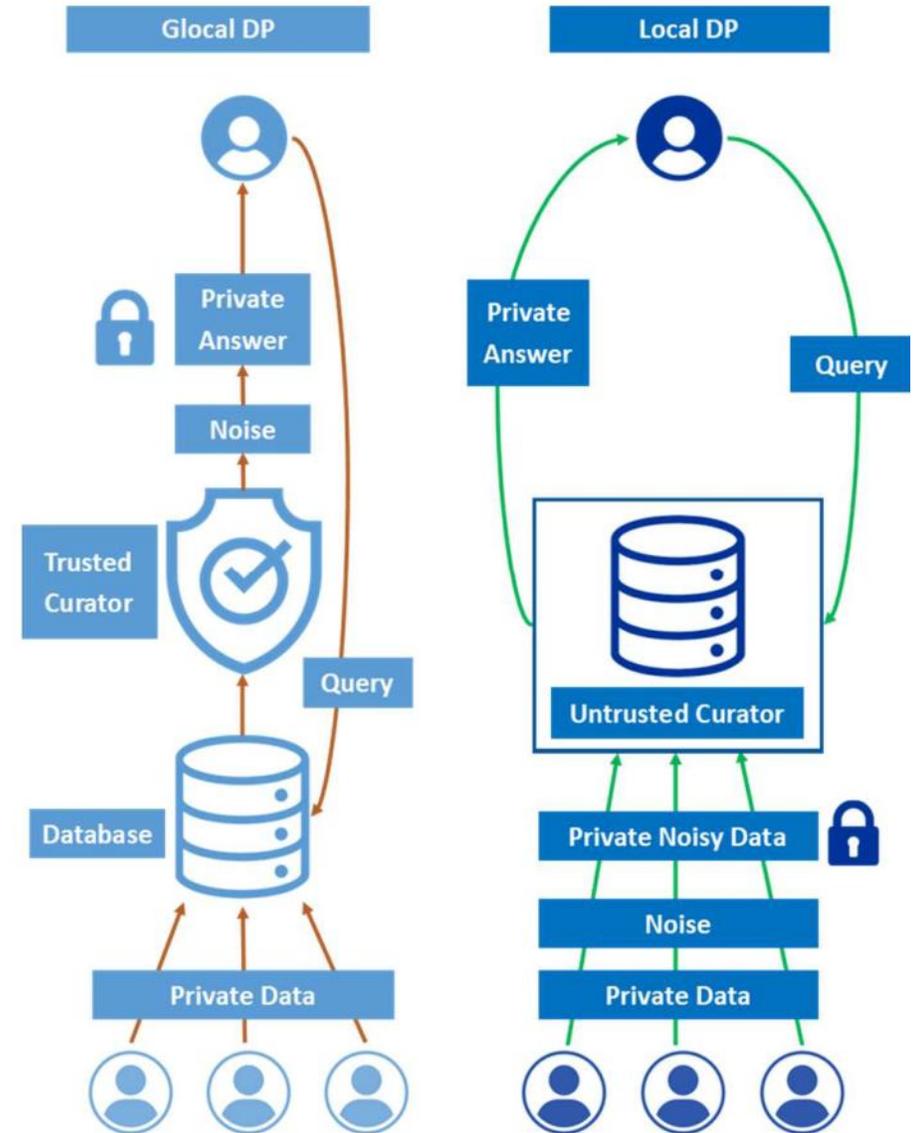
What is Differential Privacy?

- **Core Concept:** Protects individual privacy by adding noise to data
- **Mathematical Foundation:** Balances privacy with data utility using the privacy loss parameter (ϵ)
- **Key Feature:** Ensures statistical analysis results are consistent whether or not a specific individual's data is included
- **Real-World Applications:** Used by companies like Apple, Google, and in the U.S. Census



Local vs. Global Differential Privacy

- Local Differential Privacy (LDP):
 - ✓ **Privacy at the Source:** Noise added before data leaves the user's device
 - ✓ **Use Case:** Ideal when users don't fully trust the data collector (e.g., Apple iOS)
- Global Differential Privacy (GDP):
 - ✓ **Privacy at the Server:** Noise added at a centralized database
 - ✓ **Use Case:** Suitable for large-scale analysis with a trusted central entity (e.g., U.S. Census)
- ENCRYPT's Choice:
 - ✓ Focuses on LDP to empower user control over data privacy

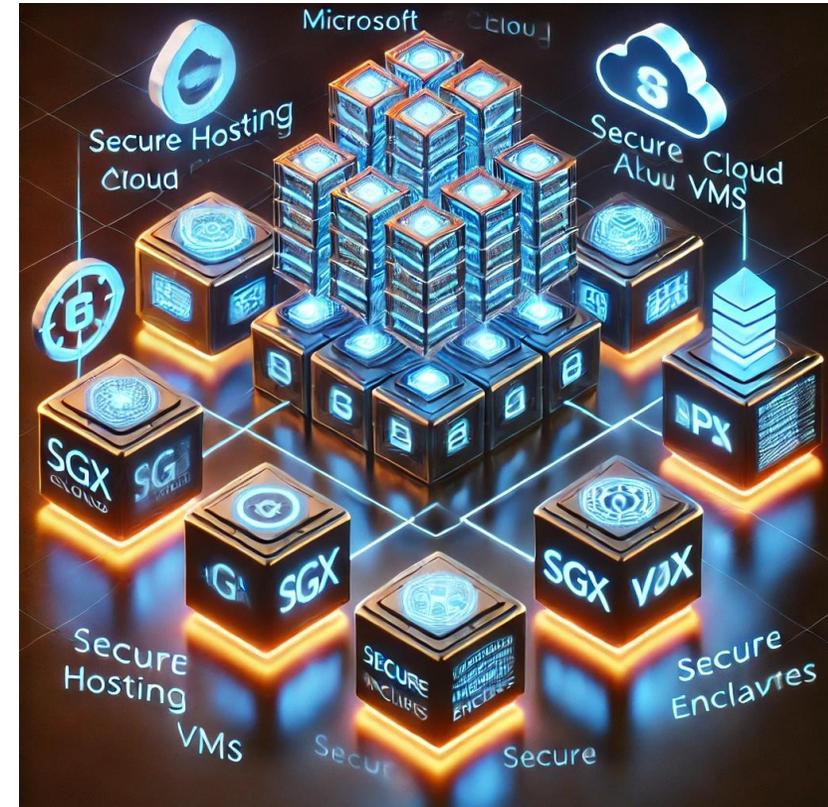


The ENCRYPT Platform Architecture

- **Modular Design:** Flexible and scalable structure for integrating various technologies
- **Key Components:**
 - ✓ **User Interface:** Accessible design for non-experts
 - ✓ **Secure Hosting:** Microsoft Azure Cloud with SGX-enabled VMs for secure enclaves
 - ✓ **APIs and Communication:** Ensures reliable and consistent interaction between components
- **Purpose:** Facilitates secure, efficient, and GDPR-compliant data processing

Making DP Accessible with the Recommendation Engine

- **AI-Driven Recommendations:** Suggests appropriate privacy-preserving technologies
- **Customizable:** Tailored suggestions based on user's data and privacy needs
- **Continuous Updates:** Adapts to new research, technologies, and regulations
- **User-Centric Design:** Provides explanations to build trust and understanding



User Interaction and Workflow in ENCRYPT

- **User-Friendly Interface:** Simplifies data pre-processing and privacy parameter selection
- **Local Noise Addition:** Privacy-enhancing noise is added locally on the user's device
- **Secure Data Upload:** Noised data is securely uploaded to the ENCRYPT platform
- **Automated Model Training:** ENCRYPT trains and optimizes AI models using the noised data
- **Consistent Analysis:** Users can apply the same privacy settings to new datasets for consistent results



Artificial Neural Networks



Validating DP in Real-World Scenarios

- **Fintech Use Case:** Applied DP in financial data analysis
- **Experiment Results:** High accuracy (~88%) even with significant noise added
- **Key Insight:** DP maintains data utility while ensuring strong privacy protections
- **Practical Impact:** Demonstrates the feasibility of DP for real-world applications in sensitive domains

Decision Tree Classifier



Gaussian Naive Bayes



Future Challenges and Research Directions

- **Privacy vs. Utility Trade-off:** Balancing strong privacy with high data accuracy
- **Scalability Issues:** Addressing computational challenges with large datasets
- **User-Friendly Tools:** Need for more intuitive interfaces for non-experts
- **Interoperability:** Ensuring seamless integration with various systems and platforms

Summary

- **ENCRYPT's Mission:** Simplifies privacy-preserving technologies for non-experts
- **DP's Role:** Balances data utility and privacy effectively
- **Real-World Impact:** Proven applicability in sectors like finance
- **Looking Forward:** Continued innovation needed to address challenges and expand use cases



Q&A



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