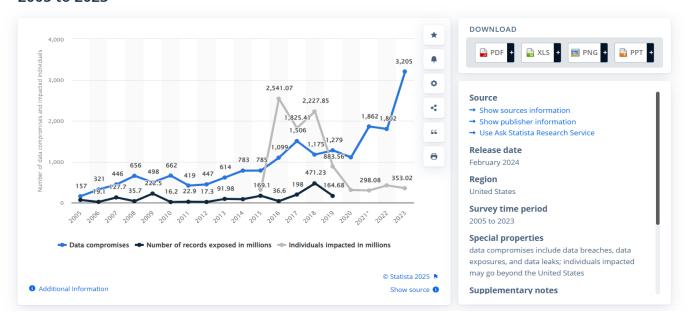
Annual number of data compromises and individuals impacted in the United States from 2005 to 2023



Keyways AI can contribute to data breaches in finance:

Increased Attack Surface:

Integrating AI into financial operations expands the potential entry points for attackers, as they can target vulnerabilities within the AI models themselves, data pipelines, or the infrastructure supporting them.

Data Leakage through Training Data:

The vast amount of sensitive financial data used to train AI models can inadvertently leak information if not properly anonymized or secured, exposing customer details to potential attackers.

Model Manipulation:

Cybercriminals can manipulate AI models through "adversarial attacks" by feeding them carefully crafted data to produce incorrect outputs or reveal sensitive information.

• Lack of Explainability ("Black Box" Problem):

The complex nature of many Al algorithms makes it difficult to understand how they reach decisions, hindering the identification and mitigation of security vulnerabilities.

Automated Attacks:

Al can be used to automate malicious activities like phishing campaigns, credential stuffing, or even complex financial fraud schemes, making attacks more efficient and harder to detect.

Insider Threats:

Malicious insiders with access to AI systems could exploit their knowledge to manipulate data or launch targeted attacks.

Examples of Al-related data breaches in finance:

Al-powered Fraud Detection Bypass:

Cybercriminals can use AI to modify transaction data in a way that bypasses traditional fraud detection systems powered by AI.

Al Chatbot Exploits:

Hackers can exploit vulnerabilities in Al-powered customer service chatbots to gain access to sensitive customer information.

Data Breach through Model Reverse Engineering:

Advanced attackers may attempt to reverse engineer an AI model to extract sensitive information embedded within its training data.

Mitigating the risks:

Robust Data Security Practices:

Implementing strong data encryption, anonymization techniques, and access controls for training data.

Model Monitoring and Validation:

Continuously monitoring AI models for anomalies and potential vulnerabilities, including regular testing and validation processes.

Explainable AI (XAI):

Utilizing techniques that provide insights into how AI models make decisions to identify potential biases and security risks.

Cybersecurity Expertise:

Building a team with specialized knowledge in AI security to proactively identify and address emerging threats.