

Al is only as good as its data. Is your organization protecting its models?

Al Data Poisoning

What: Al data poisoning is a type of cyberattack where adversaries intentionally introduce corrupt or deceptive data into an Al system's training or operational data pipeline.

Why: To *deliberately influence key decisions* when humans use the AI model to support decision-making (e.g., national defense, healthcare, finance)

This attack manipulates the model's learning process, so that the Al's outputs are aligned with the adversary's goals.



Preventing Al Data Poisoning

Secure Data Pipelines

- Vet and verify all training data sources *before* ingestion.
- Implement *strict access controls* to prevent unauthorized modifications.
- Use cryptographic techniques such as hashing and digital signatures to ensure data integrity.
- Log and audit data provenance to track its origin and modifications.



Preventing Al Data Poisoning

Anomaly Detection & Continuous Monitoring

- Monitor Al behavior for unexpected shifts or performance degradation.
- Use automated anomaly detection tools to *flag inconsistencies* in model outputs.
- Track *data drift*, as unexpected changes in real-world data can indicate poisoning attempts.
- Set up *real-time alerts* for unusual patterns in AI decision-making.



Preventing Al Data Poisoning

Adversarial Testing & Model Robustness

- Simulate data poisoning attacks during AI development to *identify vulnerabilities*.
- Conduct red-team testing on AI models to find weaknesses before attackers do.
- Use differential privacy techniques to make AI models less sensitive to small data manipulations.
- Retrain models periodically using verified datasets to correct potential poisoning effects.