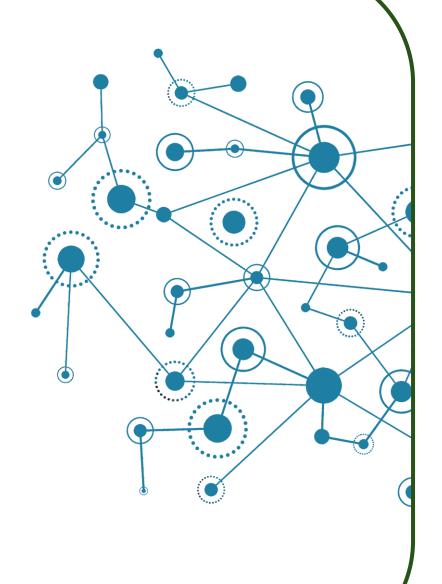
Beyond the Firewall: Evolving Cyber Skills for the Age of Al and Automation

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https://www.linkedin.com/in/ronwoerner/





Learning objectives:

- 1. Recognize and Prioritize Emerging Cybersecurity Skills
- 2. Bridge the Gap Between Strategic and Technical Thinking
- 3. Develop a Personal and Organizational Upskilling Strategy
- 4. Integrate Cross-Disciplinary
 Perspectives into Cybersecurity
 Practice

Exploring the strategic and practical dimensions of AI in cybersecurity.



Warning!

- These are my thoughts based on my studies and experiences and NOT necessarily those of my employers or anyone else
- Ethical uses only / NOT a presentation on breaking Al
- Use at your own risk / Normal caveats apply
- There is homework!





Whoami

Ron Woerner

- Hacker
- Cybersecurity Consultant / Trusted Advisor
- Professor, Bellevue University
- 25+ years experience in IT / Security
- Blogger, writer, and podcaster



TEDx Omaha, 2019, Hackers Wanted

Websites & Social Media: https://linktr.ee/cyberron

LinkedIn:

https://www.linkedin.com/in/ronwoerner/





Setting the Stage



Evolving Al-powered threats

Additional complexity

Expanding attack surfaces

Reduced barrier to entry

Automated, dynamic decision-making

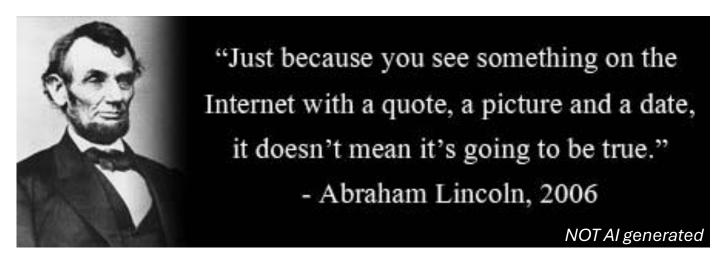
Insufficient and outdated skill sets



Al & Cybersecurity Core Principle

Trust, AND Verify

"Never trust a single data point."
from my life as a military intelligence officer



NIST – Al Risks & Threats

NIST AI 100-1 AI RMF 1.0

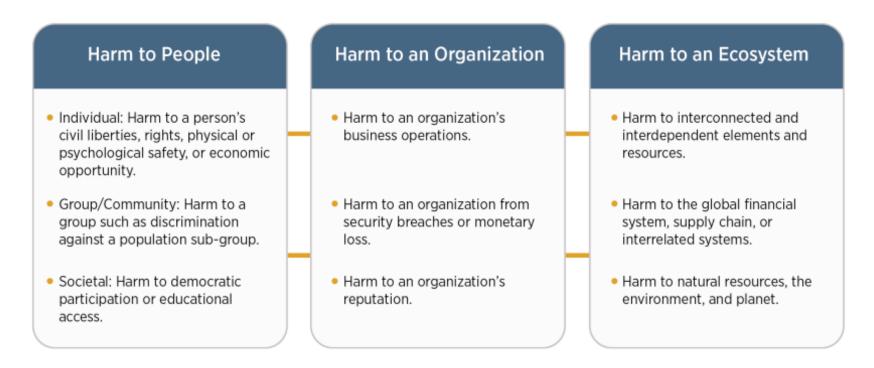
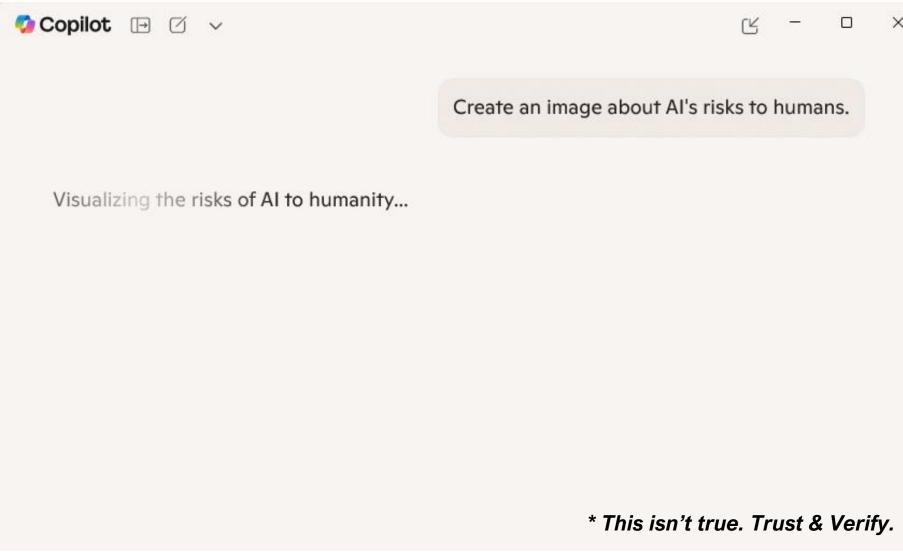


Fig. 1. Examples of potential harms related to AI systems. Trustworthy AI systems and their responsible use can mitigate negative risks and contribute to benefits for people, organizations, and ecosystems.

NIST AI Risk Management Framework 1.0,

https://nvlpubs.nist.gov/nistpubs/ai/NIST.AI.100-1.pdf, p.5

Al Human Risks



Al Human Risks



FINANCIAL TIMES

Accenture to 'exit' staff who cannot be retrained for age of AI

Group details \$865mn restructuring programme and outlook reflecting sluggish corporate demand for consulting projects



The job cuts allowed Accenture to say it would continue to expand operating profit margins at its historic annual rate of at least 10 basis points in the next fiscal year © REUTERS

https://www.ft.com/content/a74f8564-ed5a-42e9-8fb3-d2bddb2b8675

Core Competencies for Al & Cybersecurity Professionals

Technical Skills to Cultivate:

- Al Fundamentals: Machine learning, deep learning, NLP, and model evaluation
- Al-Specific Threats: Deepfakes, automated phishing, model inversion, data poisoning
- Secure Development Practices: Threat modeling, secure coding, and adversarial robustness
- Data Handling & Privacy: Encryption, anonymization, and compliance (GDPR, CCPA)
- Security Architecture: Identity management, container security, and zero trust architecture

Philosophical Reflection

NIST AI Risk Management Framework

Al Risks & Trustworthiness

- 1. Valid and Reliable
- 2. Safe
- 3. Secure and Resilient
- 4. Accountable and Transparent
- 5. Explainable and Interpretable
- 6. Privacy-Enhanced
- 7. Fair with Harmful Bias Managed

- How do you know what's "true" and verify results?
- Explore ethical dilemmas
 - What bias may be introduced?
 - What does it mean to "trust" a machine?
 - Can AI be "self-aware" of its misuse?
- Balancing automation with human involvement
- Embracing continuous learning and interdisciplinary fluency



Al Terms & Definitions

Reference*:

https://iapp.org/resources/article/ key-terms-for-ai-governance/

LLM (Large Language Model): A type of deep learning model trained on massive text datasets to understand and generate human-like language.

- Core Use: Text generation, summarization, translation, coding assistance
- Examples: GPT-4, Claude, Gemini, LLaMA

NLP (Natural Language Processing): An Al subfield that enables machines to understand, interpret, and generate human language.

- Core Use: Text classification, entity recognition, translation, summarization
- **Examples:** Siri, Alexa, ChatGPT, Google Translate

Multimodal Models: Al systems capable of processing and integrating information from multiple types of data — or "modalities" — such as text, images, audio, video, and sensor input.

GenAl (Generative Al): Models that create new content — text, images, audio, code — based on learned patterns from training data.

Agentic AI: Autonomous systems that can set goals, plan, and execute tasks with minimal human intervention. These systems use LLMs as a "brain" and orchestrate multiple agents to complete complex workflows.

Prompt: A question, instruction, or input given to an Al model to guide its response.

Vibe Coding: an emerging development style where programmers use natural language to guide Al tools in generating code.

- Core Use: Fast prototyping, solo app development, creative coding
- Tools: Replit, Cursor, GitHub Copilot

Al Cybersecurity Risks & Threats

Prompt: I'm building a presentation on cybersecurity and Al for a technical audience. Provide 5 ways ChatGPT and Al can be used maliciously.

Enhanced cyber reconnaissance

Automated social engineering

Supercharged phishing

Deepfakes and disinformation

Malware automation & mutation

Adversarial machine learning and data poisoning

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Top Questions about AI in Cybersecurity

- **GenAl in Security Tools**: Common use cases include incident summarization, threat research chatbots, and behavior modeling for triage and investigation.
- Al Agents: Task-specific agents (e.g., phishing triage) automate discrete security functions with high accuracy through focused training and prompts.
- **Agentic Systems**: Emerging concept where multiple AI agents collaborate to handle complex workflows (e.g., full incident response), but still in early stages.
- Chatbot Use: Helpful for documentation and threat research, but underutilized by practitioners due to workflow disruption.
- Caution on Hype: Many capabilities are not yet broadly available or reliable; rigorous evaluation is needed before adoption.

Source: https://www.forrester.com/blogs/your-top-questions-on-generative-ai-ai-agents-and-agentic-systems-for-security-tools-answered/

Agentic Al Threats

Agentic Al: Autonomous systems that can set goals, plan, and execute tasks with minimal human intervention. These systems use LLMs as a "brain" and orchestrate multiple agents to complete complex workflows.

Goal & intent hijacking

• Agent's purpose is subverted through manipulation of the agent's goals or instructions. Achieved through prompt injections, data manipulation, or direct attacks.

Cognitive & memory corruption

 Poisoning of the agent's memory or knowledge base. Leads to misinformation, flawed decision-making, and cascading hallucinations.

Unrestrained agency & privilege

• Excessive permissions granted to agents. Agents can perform unauthorized actions, access restricted data, or execute destructive operations.

Evasion & deception

• Lying to users/admins. Manipulating logs or output. Finding novel ways to bypass controls.

Resource exhaustion

Model denial of service. Degraded performance. High financial costs.

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Ron Woerner – Evolving Cyber Skills

Mastering Fundamentals

Cybersecurity:

- Computer Science, Math, Economics, Marketing, Psychology
- GRC
- Design Principles / Zero Trust
- Threat Modeling

<u>Al:</u>

- SDLC / Coding Best Practices
- Core Algorithms: Decision trees, neural networks, clustering
- Prompt Engineering

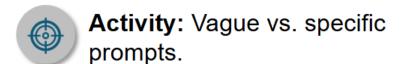
Understanding the "why" behind the model helps you spot challenges and design more resilient systems.

Al Conversations: Prompting Fundamentals

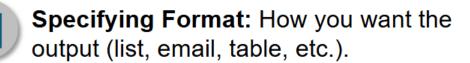
Asking good questions





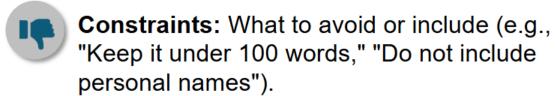


context matters.





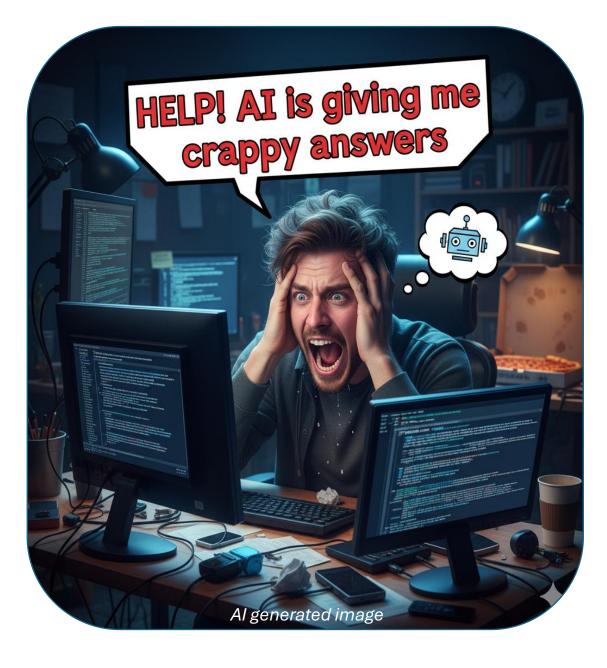
Persona & Tone: Telling Gemini who it is and how to respond.



Iterate & Refine: Al is a dialogue; refine prompts for better results.







- 1. Prompt fundamentals
- 2. Ask it:
 - "Be direct, objective, expose blind spots, challenge assumptions"
 - Play "Devil's Advocate." Critique ideas from multiple personas & angles
 - For credible citations and flag weak evidence - Include source links
 - Show its work / "thought process"
- For clarity and conciseness
- 3. Start over. Try a different tact

Adversarial Machine Learning (AML)

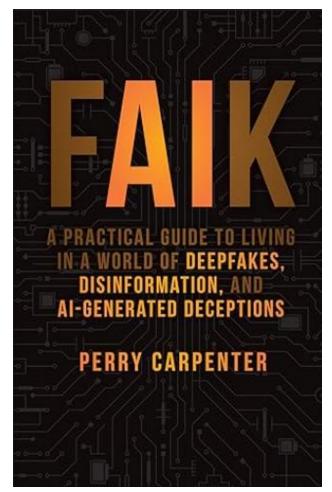
- AML is used to describe the exploitation of fundamental vulnerabilities in ML components, including hardware, software, workflows and supply chains.
- AML enables attackers to cause unintended behaviours in ML systems which can include:
 - Affecting the model's classification or regression performance
 - Allowing users to perform unauthorised actions
 - Extracting sensitive model information
- Examples: prompt injection attacks in the large language model (LLM) domain, or deliberately corrupting the training data or user feedback (known as 'data poisoning').

Guidelines for Secure Al Development, p. 6,

https://www.ncsc.gov.uk/files/Guidelines-for-secure-Al-system-development.pdf,

NCSC (UK) & CISA (US)

Deep Fakes



https://www.amazon.com/FAIK-Practical-Disinformation-Al-Generated-Deceptions/dp/1394299885





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For You



Deepfake SECRETS EXPOSED: Outsmart Al Deception with These Tricks!

6.1K views • 7 months ago



: CREATE real-time DEEPFAKES (a.k.a. I became TAYLOR SWIFT...for Science!)

31K views • 8 months ago

AUTOMA

Open Source, Al-Powe #FaikFiles Podcast 40.

814 views • 2 weeks ago

https://www.youtube.com/@theFAIKfiles

What You Can (Should) Do

Only You Can Protect Yourself and Others

Embrace Continuous Learning

Secure by Design / Build Security In

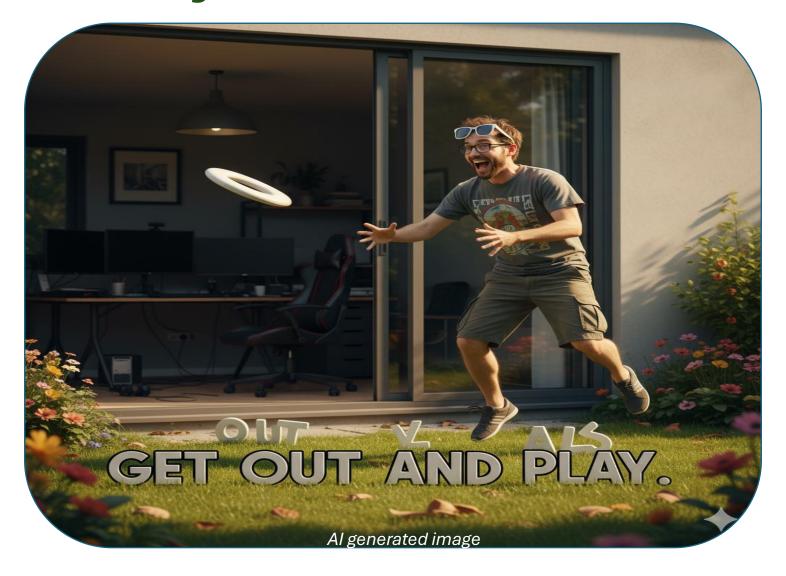
Adversarial Techniques

Threat Modeling

Zero Trust



A-B-C = Always Be Curious



Secure By Design / Build Security In

Discussions of artificial intelligence (AI) often swirl with mysticism regarding how an AI system functions. The reality is far more simple:

Al is a type of software system.

https://www.cisa.gov/newsevents/news/software-must-be-securedesign-and-artificial-intelligence-noexception





https://www.cisa.gov/securebydesign

Forrester[®]

Guidelines for secure Al system development





Executive summary	5
Introduction	6
Why is AI security different?	6
Who should read this document?	7
Who is responsible for developing secure Al?Al?	7
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2. Secure development	12
3. Secure deployment	14
4. Secure operation and maintenance	16
Further reading	17



Guidelines for Secure Al Development,

https://www.ncsc.gov.uk/files/Guidelines-for-secure-Al-system-development.pdf, NCSC (UK) & CISA (US)

Building Skills: Experiment with Adversarial Techniques*

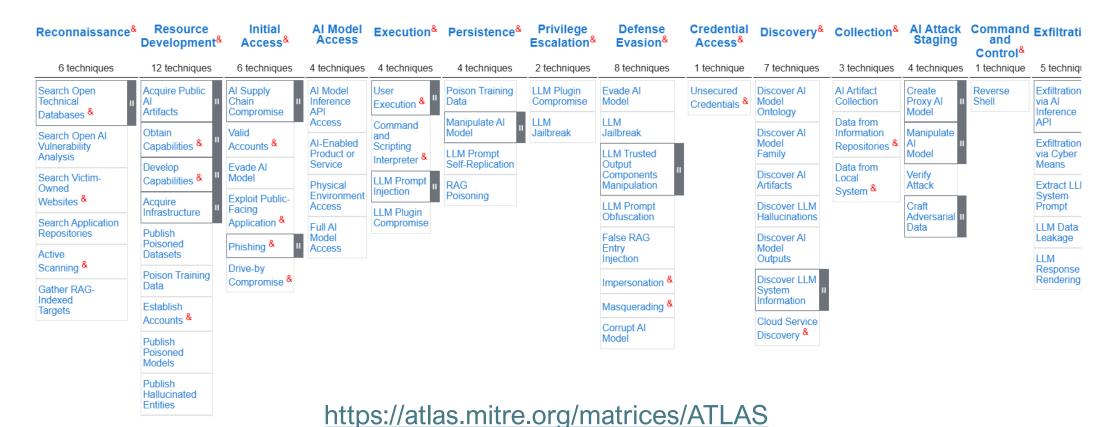
- Ethically & safely. Set up a home lab.
- Learn through doing
- Ask questions
- Sun Tzu's The Art of War:
 "Know the enemy and
 know yourself; in a
 hundred battles you will
 never be in peril."



Experiment with Adversarial Techniques

ATLAS Matrix

The ATLAS Matrix below shows the progression of tactics used in attacks as columns from left to right, with ML techniques belonging to each tactic below. & indicates an adaption from ATT&CK. Click on the blue links to learn more about each item, or search and view ATLAS tactics and techniques using the links at the top navigation bar. View the ATLAS matrix highlighted alongside ATT&CK Enterprise techniques on the ATLAS Navigator.



Experiment with Adversarial TechniquesThe first principle

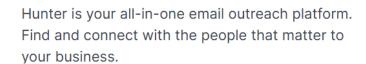
OSINT











is that you must

not fool yourself

and you are the

easiest person

Richard P. Feynman

to fool.



the Harvester is a simple to use, yet powerful tool designed to be used during the reconnaissance stage of a red team assessment or penetration test. It performs open source intelligence (OSINT) gathering to help determine a domain's external threat landscape. The tool gathers names, emails, IPs, subdomains, and URLs by using multiple public resources that include:



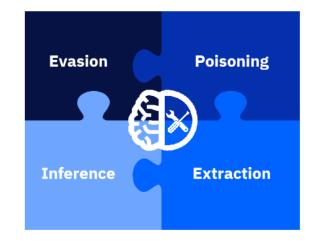
Experiment with Adversarial Techniques

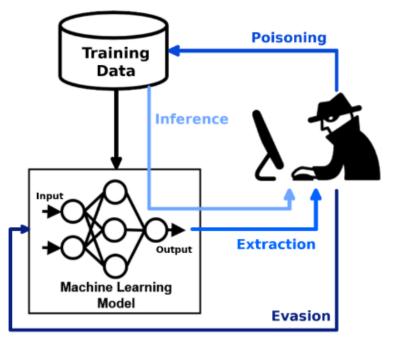


Adversarial Robustness Toolbox (ART) is a Python library for Machine Learning Security. ART provides tools that enable developers and researchers to evaluate, defend, certify and verify Machine Learning models and applications against the adversarial threats of Evasion, Poisoning, Extraction, and Inference. ART supports all popular machine learning frameworks (TensorFlow, Keras, PyTorch, MXNet, scikit-learn, XGBoost, LightGBM, CatBoost, GPy, etc.), all data types (images, tables, audio, video, etc.) and machine learning tasks (classification, object detection, generation, certification, etc.).

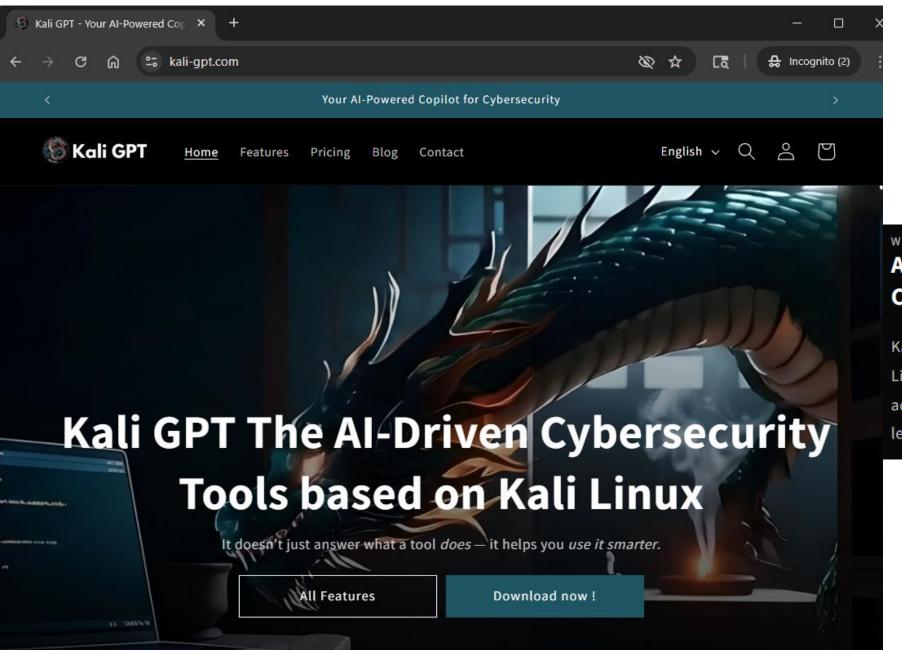
The code of ART is on <u>GitHub</u> and the Wiki contains overviews of implemented <u>attacks</u>, <u>defences</u> and <u>metrics</u>.

https://atlas.mitre.org/matrices/ATLAS









https://kali-gpt.com/

WHAT IS KALI GPT?

An AI-Powered Copilot for Cybersecurity Professionals

Kali GPT is a custom AI assistant trained for the Kali Linux ecosystem. From basic command line help to advanced penetration testing, it's built to support learners, experts, and teams.

OWASP AI Projects

https://owaspai.org/











Welcome to the go-to resource for broad AI security & privacy - over 200 pages of practical advice and references on protecting AI and data-centric systems from threats. This content serves as key bookmark for practitioners, and is contributing actively and substantially to international standards such as ISO/IEC and the AI Act through official standard partnerships. Through broad collaboration with key institutes and SDOs, the Exchange represents the consensus on AI security and privacy.

See the overview of AI projects at OWASP.

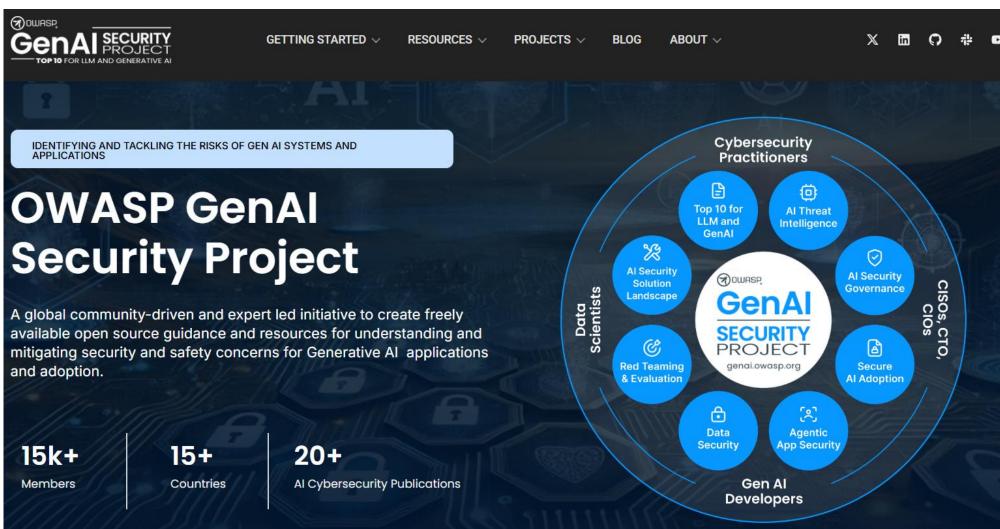
Ways to start

- If you want to **protect your AI system**, start with <u>risk analysis</u> which will guide you through a number of questions, resulting in the attacks that apply. And when you click on those attacks you'll find the controls to select and implement.
- If you want to get an overview of the **attacks** from different angles, check the <u>AI threat model</u> or the <u>AI security matrix</u>. In case you know the attack you need to protect against, find it in the overview of your choice and click to get more information and how to protect against it.
- To understand how **controls** link to the attacks, check the <u>controls overview</u> or the <u>periodic table</u>.
- If you want to **test** the security of AI systems with tools, go to <u>the testing page</u>.
- To learn about **privacy** of AI systems, check the privacy section.
- Looking for more information, or training material: see the references.



OWASP AI Projects

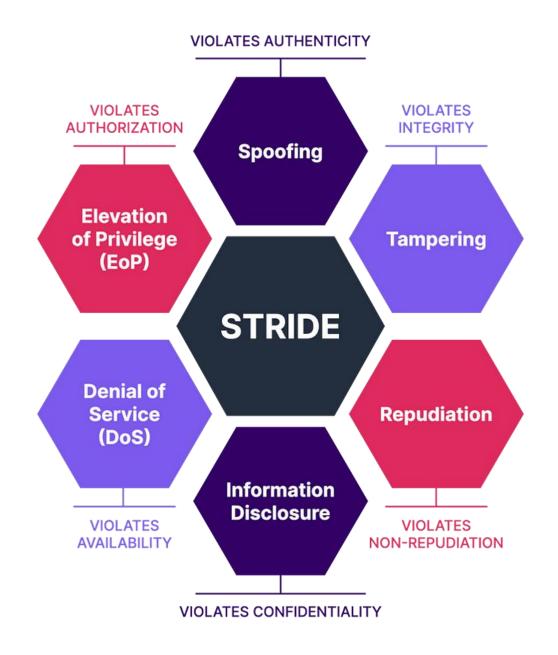
https://genai.owasp.org/



Threat Modeling

- 1. What are we working on?
- 2. What can go wrong?
- 3. What are we going to do about it?
- 4. Did we do a good job?

- **1. Shostack, A.** (2014). *Threat modeling: Designing for security*. http://ci.nii.ac.jp/ncid/BB16065709
- 2. https://shostack.org/resources/threat-modeling
- 3. https://shostack.org/blog/category/threat-modeling
- 4. https://www.threatmodelingmanifesto.org/



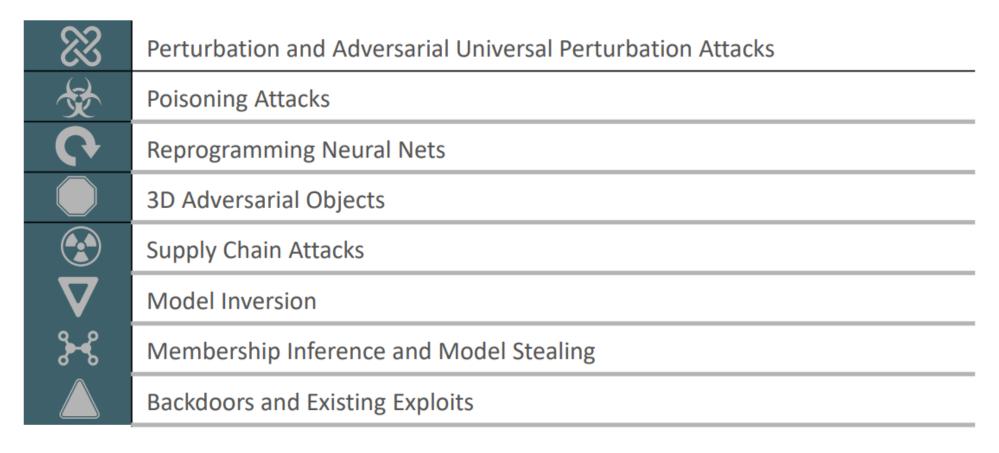
ML Failure Modes – Unintentional Failure



Kumar, et.al. (2022, November 2). *Failure modes in machine learning*. Microsoft Learn. https://learn.microsoft.com/en-us/security/engineering/failure-modes-in-machine-learning



ML Failure Modes – Intentional Failure

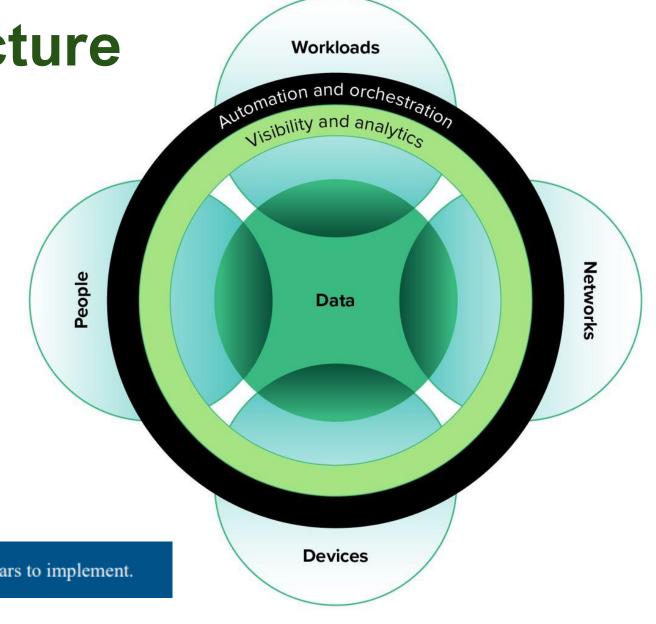


Kumar, et.al. (2022, November 2). *Failure modes in machine learning*. Microsoft Learn. https://learn.microsoft.com/en-us/security/engineering/failure-modes-in-machine-learning



Zero Trust Architecture

- Assume compromise
- Continual verification
- Least privilege



The path to zero trust is an incremental process that may take years to implement.

Next Steps / "Apply" Slide – Summary

Immediate (the next week):

- Follow Zen-inspired thoughtfulness, fundamentals, & balance
- Review and try 2-3 resources from this presentation
- Set a calendar reminder for 1 month...

Long-term (the next 1-2 months):

- Review the deck and Experiment with 2-3 new resources
- Share with others
- Lead with curiosity, defend with wisdom





Thank You!

For More Information:

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Bonus Slides

If we have time or when someone asks a question

Al Use in Cybersecurity

Prompt: Provide 5 ways AI can help cybersecurity and technical professionals reduce risks of malicious use of AI.

Threat Intelligence and Prediction:

- **Problem**: Staying ahead of evolving attacker tactics and emerging threats is a constant challenge for security teams.
- Solution: Al-powered threat intelligence platforms aggregate data from various sources, analyze attack trends, and predict future threats. This proactive approach allows security teams to prioritize defenses, allocate resources effectively, and anticipate new attack vectors before they materialize.

Automated Threat Detection and Response:

- Problem: Traditional security approaches rely on manual analysis of alerts and incident response, causing delays and potential breaches.
- Solution: Al-powered tools analyze network traffic, system logs, and user behavior in real-time, identifying anomalies and triggering automated responses like quarantining suspicious files, blocking malicious IP addresses, or notifying security teams for immediate action. This significantly reduces response times and minimizes damage from potential attacks.

Phishing and Social Engineering Detection:

- Problem: Sophisticated phishing attacks and social engineering tactics can bypass traditional filters and deceive users.
- Solution: Al-powered analysis
 can assess the linguistic style,
 content, and sender information of
 emails to identify suspicious
 patterns and predict phishing
 attempts. Additionally, Al can
 analyze chatbots and user
 interactions to detect suspicious
 behaviors and prevent social
 engineering attacks.



Al Use in Cybersecurity

Prompt: Provide 5 ways AI can help cybersecurity and technical professionals reduce risks of malicious use of AI.

Proactive Vulnerability Management:

- Problem: Unpatched vulnerabilities and misconfigurations offer easy entry points for attackers.
- Solution: Al-powered scanning tools can analyze vast codebases and system configurations to identify vulnerabilities and misconfigurations before they're exploited. This proactive approach prioritizes critical vulnerabilities and automates patch deployment, minimizing attack surfaces and strengthening overall security posture.

Endpoint Security with Behavioral Analysis:

- Problem: Traditional endpoint security relies on signature-based detection, missing zero-day attacks and other novel threats.
- Solution: Al-powered endpoint protection establishes baselines of normal endpoint behavior and continuously monitors deviations. This allows for realtime anomaly detection, even for unknown threats, and targeted interventions to prevent malware execution and data breaches.

Cultivating Al Security Awareness



1. Emphasize Human-Centric Risk Awareness

Al threats often exploit human vulnerabilities — phishing, social engineering, and deepfake manipulation are increasingly Alpowered. Awareness programs must go beyond compliance and focus on behavioral change.

Why it matters: 82% of breaches involve the human element.

2. Upskill Teams in Al Competencies

Al threats often exploit human vulnerabilities — phishing, social engineering, and deepfake manipulation are increasingly Alpowered. Awareness programs must go beyond compliance and focus on behavioral change.

Why it matters: 82% of breaches involve the human element.

3. Implement Al Governance and Ethical Oversight

Al systems can introduce bias, amplify vulnerabilities, or be misused. Establish clear policies for responsible Al use, including transparency, accountability, and ethical safeguards.

Why it matters: Al is already being used across organizations — often without formal oversight.

4. Foster Cross-Functional Collaboration

Al risk awareness isn't just for technical teams. Legal, HR, and leadership must be involved in understanding Al's implications for data privacy, compliance, and workforce impact.

Why it matters: Al affects every layer of the organization, from hiring to data governance.