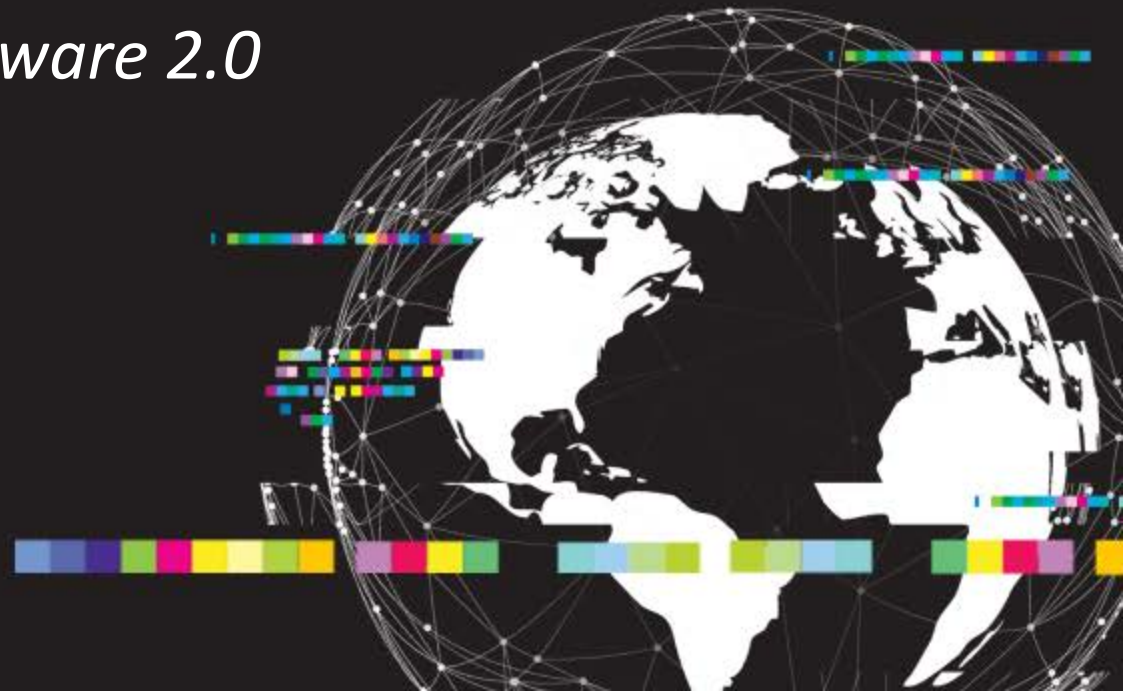




# Introducing MLSecOps

*DevSecOps for Software 2.0*





# Introduction



# About



## Eugene Neelou

- Built various security products since 2011



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- **Pioneering Security of AI since 2019**
  - Author of the Industry Report on Adversarial ML
  - Organizer of the MLSEC AI Hacking Competition track
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## Adversa AI

- World-first AI vulnerability research company
- World-first commercial AI red team
- World-first AI protection patent
- and many more



Mission



# Adversa AI Increases Trust in AI Systems

by Improving the Security of ML Algorithms

TOP MEDIA

Forbes

FORTUNE

VICE

THE WALL STREET JOURNAL  
WSJ



TECH AND SECURITY MEDIA

DARKReading

HELPNETSECURITY

SECURITYWEEK

The Daily Swig  
*Cybersecurity news and views*

AITHORITY  
AI TECHNOLOGY INSIGHTS

BIOMETRIC  
UPDATE.COM



ADVERSA





# Agenda



- AI is Software 2.0
- Adversarial Machine Learning
- MLSecOps Pipeline Stages
- Takeaways





# AI is Software 2.0





# AI for Old Products

- AI is becoming a core technology for **existing products**
  - Most tech companies will transform into AI companies

## **Nvidia CEO: Software Is Eating the World, but AI Is Going to Eat Software**

Jensen Huang predicts that health care and autos are going to be transformed by artificial intelligence.





# AI for New Products

- AI is becoming a go-to technology for **new products**
  - Most disruptive products will emerge from AI-first ideas

## **Nvidia CEO: Software Is Eating the World, but AI Is Going to Eat Software**

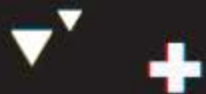
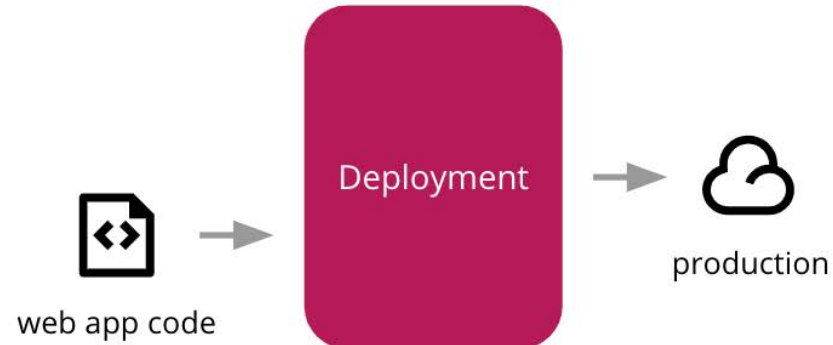
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# AI is Software 2.0



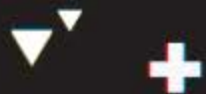
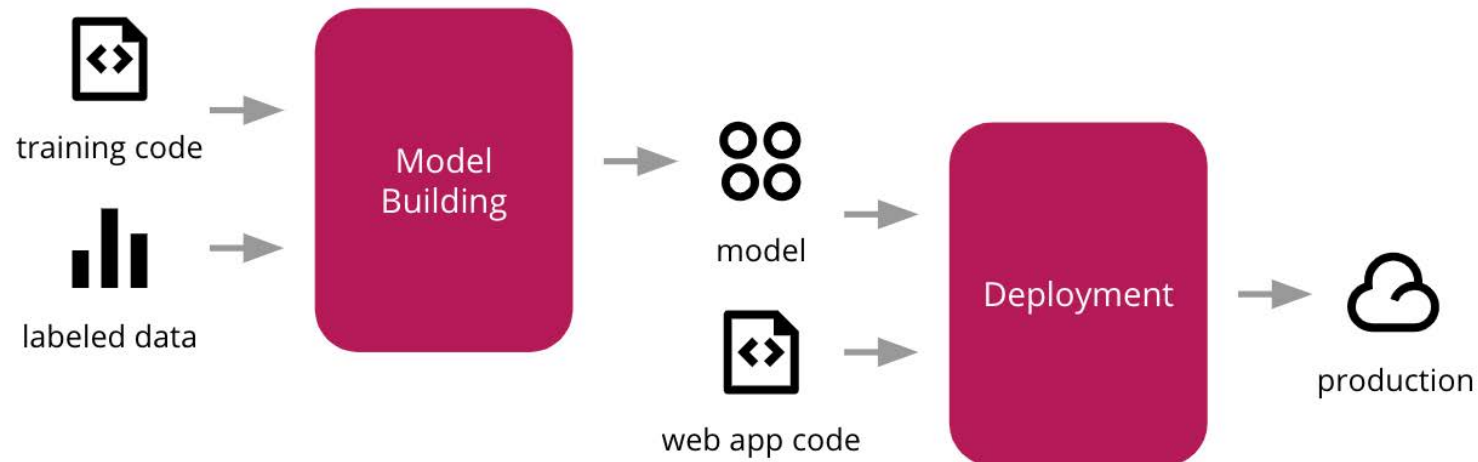
- AI is the **new paradigm** of software development



# AI Development Process



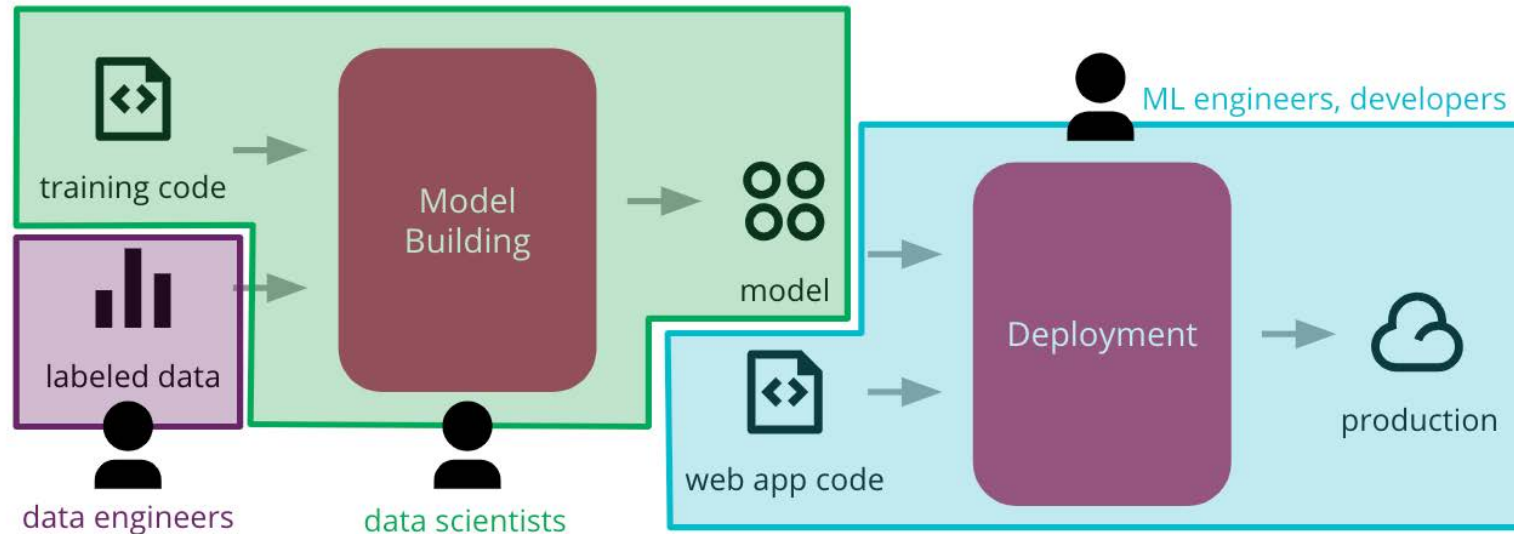
- AI development brings **model** and **dataset** assets



# AI Development Process



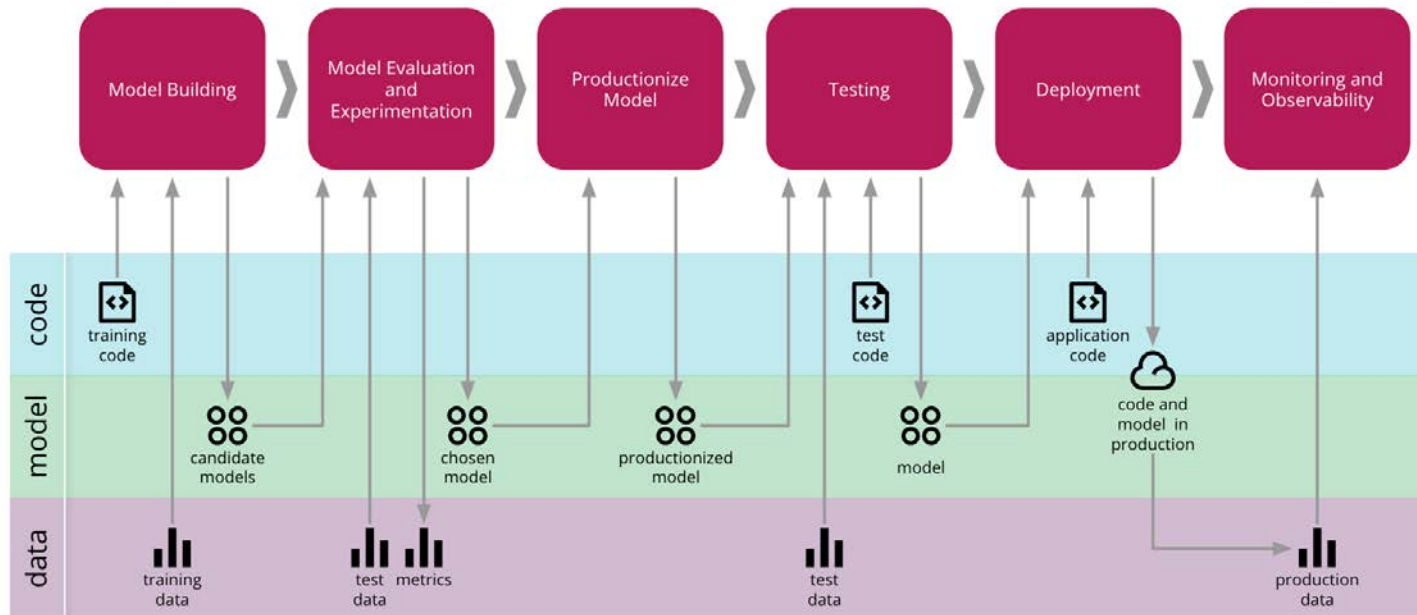
- AI development requires **data engineers** and **data scientists**



# AI Development Process



- AI development enabled by **MLOps pipelines**



*Continuous Delivery for Machine Learning (CD4ML)*







# Adversarial Machine Learning



# AI Threats Rising



**HELPNETSECURITY**

## AI industry alarmingly unprepared for real-world attacks

**VICE**

## Hackers Fool Facial Recognition Into Thinking I'm Mark Zuckerberg

**The Daily Swig**  
*Cybersecurity news and views*

Machine learning security vulnerabilities are a growing threat to the web, report highlights

**Ti Tech Informed**  
*The frontier of tech news*

## Bad Robot: the rise of adversarial AI

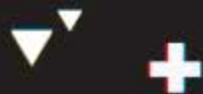
As the popularity of AI applications grows, so too do the risks associated with it

**WSJ**

## Faces Are the Next Target for Fraudsters

**DARKReading**

## Expect an Increase in Attacks on AI Systems



# AI Hacking Challenge



## Fooling facial recognition

- Preserving image quality
- Impersonating target identity

## Evading phishing detection

- Preserving code rendering
- Bypassing malware detector



<https://adversa.ai/conf42-mlsec>



# History of Adversarial ML



## Stats for 10 years of AI attacks

- AI is the new attack vector
- Real incidents In AI systems
- AI applications under attacks

## AI Security Case Study

- How to attack AI systems
- How to defend AI systems
- Secure lifecycle for AI systems



<https://adversa.ai/conf42-hitb>



# Common AI Attacks



## Infection

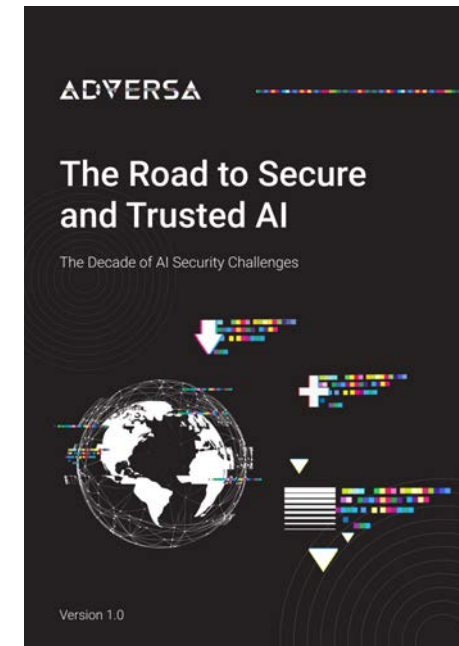
- Poisoning training data
- Supplying trojan models

## Manipulation

- Fooling model decisions
- Sabotaging model performance

## Exfiltration

- Stealing model algorithm
- Extracting private data



<https://adversa.ai/conf42-report>



# AI Attack Surface



	Traditional Software	AI = Software 2.0
<b>Nature</b>	Fixed program logic	Dynamic model training
<b>Workflow</b>	Commands based on algorithm	Decisions based on learning
<b>Interaction</b>	Graphical UI using menus and buttons	Cognitive UI using images, voice, and text
<b>Security issues</b>	Improper validation, access control issues, incorrect system configurations	Manipulation, exfiltration, infection of models & datasets





# MLSecOps Principles



# Building Security In



- ❑ Security as enabler, not blocker

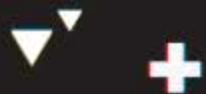




# Building Security In



- ❑ Security as enabler, not blocker
- ✓ Automated and repeated validation



# Building Security In



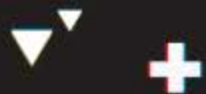
- Security as enabler, not blocker
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- Add security as early as possible



# Building Security In



- Security as enabler, not blocker
- ✓ Automated and repeated validation
  
- Add security as early as possible
- ✓ Accumulated controls across lifecycle





# MLSecOps Pipeline Stages



# MLSecOps Intro



Tasks



# MLSecOps Intro



## Problems



# MLSecOps Intro



**Solutions**



# MLSecOps: Planning



## Tasks

- **Requirements**
- Stakeholders
- Compliance



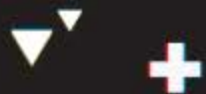


# MLSecOps: Planning



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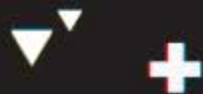


# MLSecOps: Planning



## Tasks

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# MLSecOps: Planning



## Problems

- No risk assessment
- No threat modeling
- No security governance



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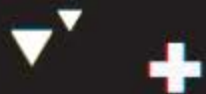


# MLSecOps: Planning



## Solutions

- ✓ **Risk register**
- ✓ Attack surface
- ✓ Security baseline

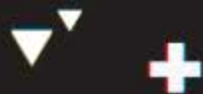


# MLSecOps: Planning



## Solutions

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# MLSecOps: Planning



## Solutions

- ✓ Risk register
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# MLSecOps: Development



## Tasks

- **Data understanding**
- Data preparation
- Data packaging

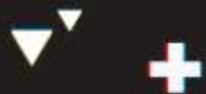


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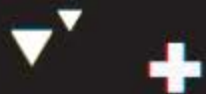


# MLSecOps: Development



## Tasks

- Data understanding
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# MLSecOps: Development



## Problems

- Privacy breach
- Unreliable sources
- Data poisoning



# MLSecOps: Development



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## Problems

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# MLSecOps: Development



## Solutions

- ✓ **Data privacy**
- ✓ Data integrity
- ✓ Data sanitization



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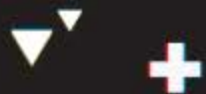


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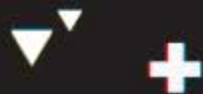


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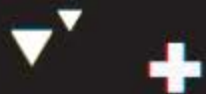


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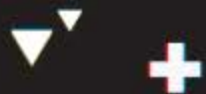


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- Supply chain risks
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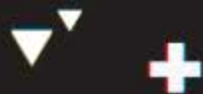


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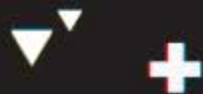


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# MLSecOps: Validation



## Tasks

- **Model evaluation**
- Robustness testing
- Compliance checks



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# MLSecOps: Validation



## Problems

- Bad testing coverage**
- Weak testing scenarios
- Vulnerable infrastructure

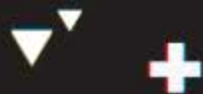


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## Solutions

- ✓ **Security governance**
- ✓ Validation playbooks
- ✓ Secure environment

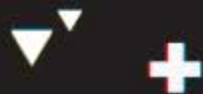


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# MLSecOps: Validation



## Solutions

- ✓ Security governance
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# MLSecOps: Deployment



## Tasks

- **Model deployment**
- Model inference
- Model serving



# MLSecOps: Deployment



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- Model deployment
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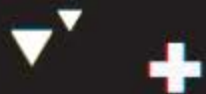


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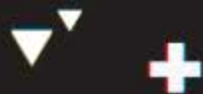


# MLSecOps: Deployment



## Problems

- Model hijacking**
- Adversarial attacks
- Unrestricted access



# MLSecOps: Deployment



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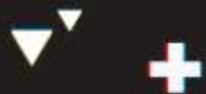


# MLSecOps: Deployment



## Solutions

- ✓ **Model authenticity**
- ✓ Adversarial robustness
- ✓ Secure communications



# MLSecOps: Deployment



## Solutions

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# MLSecOps: Deployment



## Solutions

- ✓ Model authenticity
- ✓ Adversarial robustness
- ✓ **Secure communications**



# MLSecOps: Monitoring



## Tasks

- **Model performance**
- Anomaly detection
- Feedback loops



# MLSecOps: Monitoring



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- Model performance
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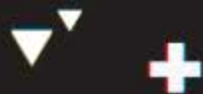


# MLSecOps: Monitoring



## Tasks

- Model performance
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# MLSecOps: Monitoring



## Problems

- Lack of monitoring
- Lack of analytics
- Lack of resilience





# MLSecOps: Monitoring



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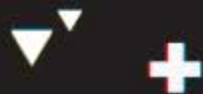


# MLSecOps: Monitoring



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# MLSecOps: Monitoring



## Solutions

- ✓ **Activity monitoring**
- ✓ Analytics playbooks
- ✓ Detection & response

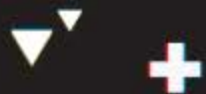


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# MLSecOps: Monitoring



## Solutions

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- ✓ Analytics playbooks
- ✓ **Detection & response**



# MLSecOps: Full Pipeline



# MLSecOps: Full Pipeline



\* Security of Workspaces



# MLSecOps: Full Pipeline



\* Security of Workspaces

\* Security of Pipelines





# Takeaways



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# Takeaways



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- Traditional cybersecurity solutions don't help



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# Thanks!

Let's talk about AI Security & Safety

[linkedin.com/in/eneelou](https://www.linkedin.com/in/eneelou) | [twitter.com/eneelou](https://twitter.com/eneelou)

