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# Adversarial Image Attacks Against Automotive Systems

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# Autoliv-Veoneer-Arriver-Qualcomm

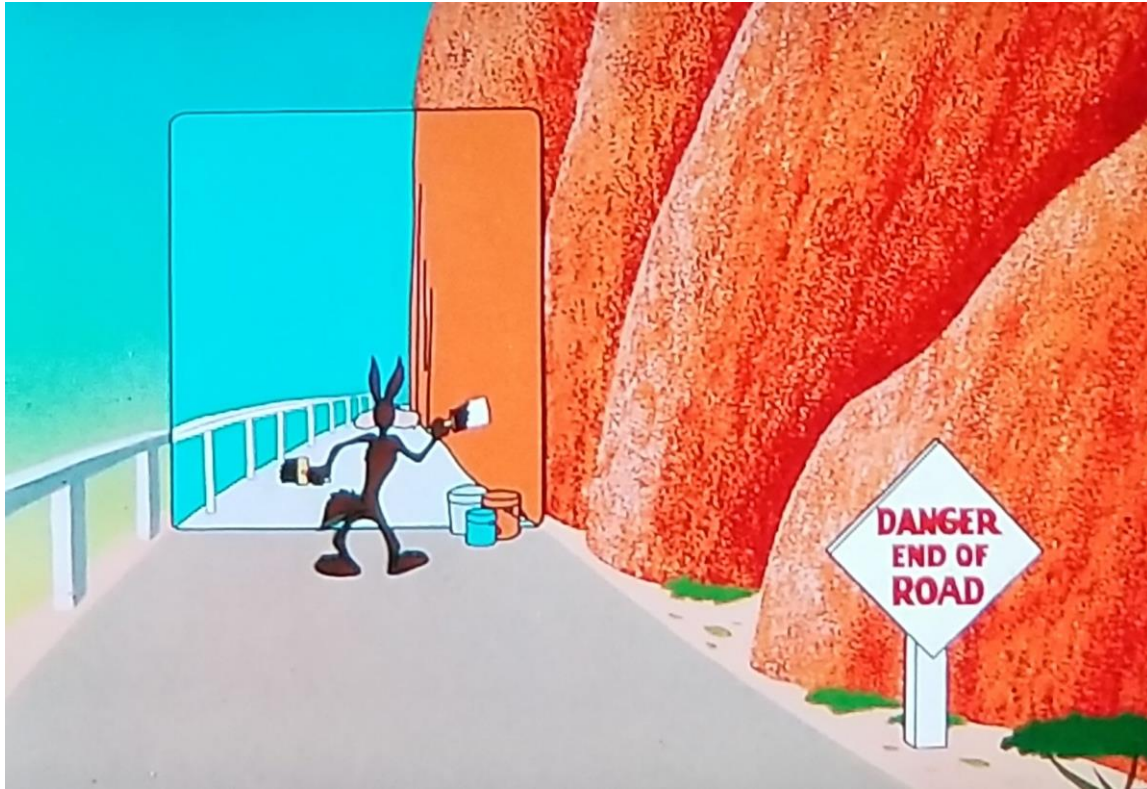
- Automotive safety since 1953
- Vision systems since 2005
- Passenger cars up to L3
- Currently:
  - Delivering camera systems to MB, VCC, Geely, GAC, BYD, Chery, BTET, and GM
  - In development agreements with BMW, Renault, and others

# Agenda

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- What, when, why?
- Technology overview
- Risk management

# Adversarial image attacks against automotive systems



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Disrupting ADAS or AD vehicle operation

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Fools vision perception

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Manipulation of the environment

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Not disruptive to humans

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# Scenarios

## Adhesive markings on street signs

- Disrupts sign detections
- Misclassifying street signs

## Painting road surfaces

- Misleading lane markings

## Markings or designs on vehicles

- Disrupts vehicle detections
- Misclassifying vehicle

## Patches placed near road

- Appears as street sign or traffic participant
- Disrupts detections of signs, traffic, or lanes



Imaged by Heritage Auctions, HA.com

# Motivations and Actors



## Economic

- Competitors

## Political

- Terrorists
- State actors

## Environmental

- Activists



# Technology

## Misclassification

Adversarial Machine Learning in Image Classification: A Survey Towards the Defender's Perspective

7

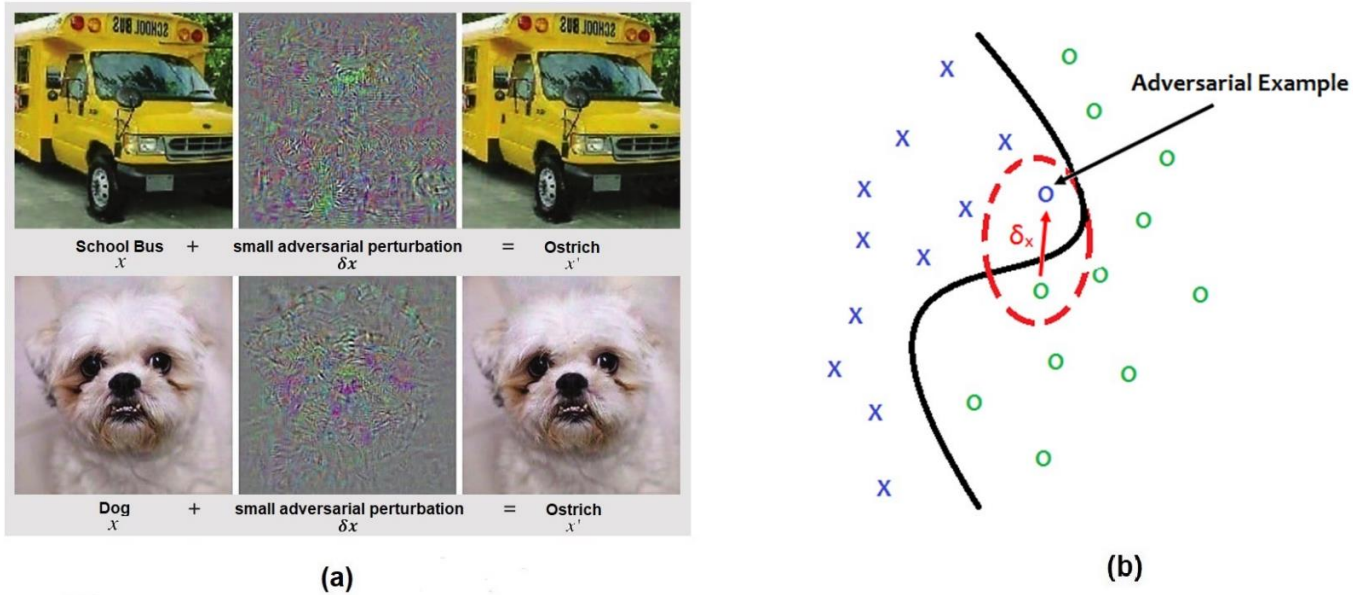


Fig. 4. (a): Malicious and usually imperceptible perturbations present in a input image can induce trained models to misclassification. Adapted from Klarreich [93]. (b): The objective of an adversarial attack is to generate a perturbation  $\delta x$  and insert it into a legitimate image  $x$  in order to make the resulting adversarial image  $x' = x + \delta x$  cross the decision boundary. Adapted from Bakhti et al. [8].

# Technology

## Disrupting detections

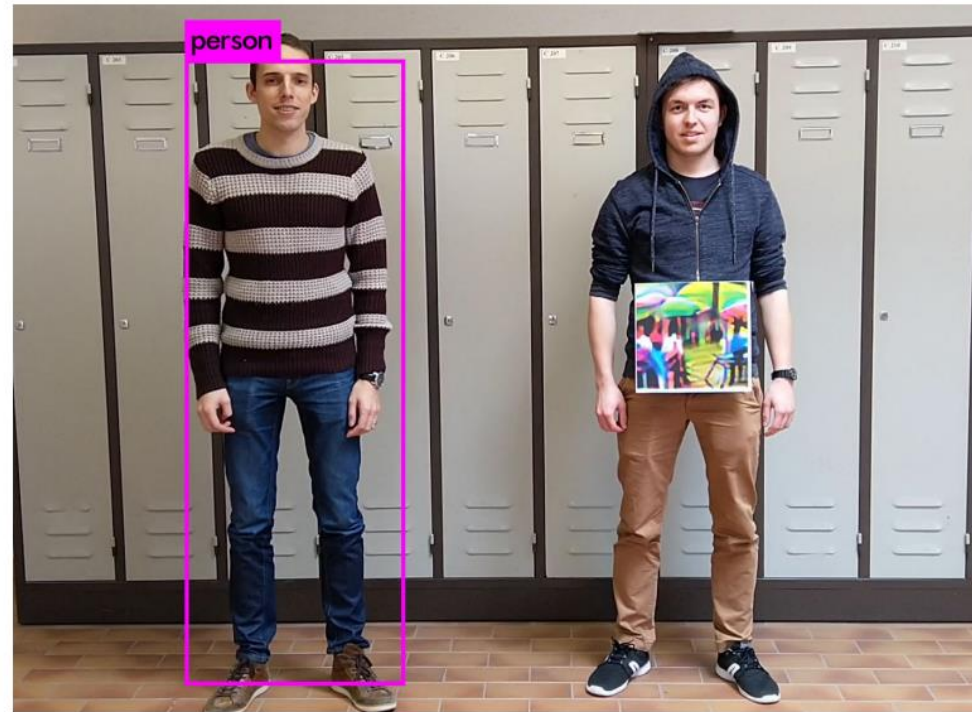


Figure 1: We create an adversarial patch that is successfully able to hide persons from a person detector. Left: The person without a patch is successfully detected. Right: The person holding the patch is ignored.



# Technology


## Billboards



**Fig. 1: The top subfigure shows an example customizable roadside billboard. The bottom two subfigures show an adversarial billboard example, where the Dave [3] steering model diverges under our proposed approach.**

# Technology

Physical attacks in the environment (not requiring access to the perception system)



Blackbox attacks (not requiring access to the algorithms)



Feasibility/realism (possible to implement in the real world)



Attack robustness (robust against variation in lighting, size, perspective, etc.)

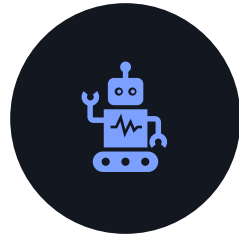
Source sample text

Focus on attack methods with these features

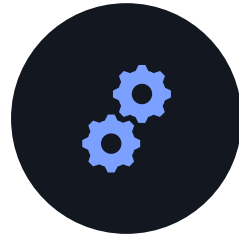
# Risk Management



SOTIF  
ISO 21488



SAFETY AND AI  
ISO/PAS 8800



SAFETY FOR  
DRIVING  
AUTOMATION  
SYSTEMS  
ISO/TS 5083



CYBERSECURITY  
ISO 21434,  
ISO/TR 4804

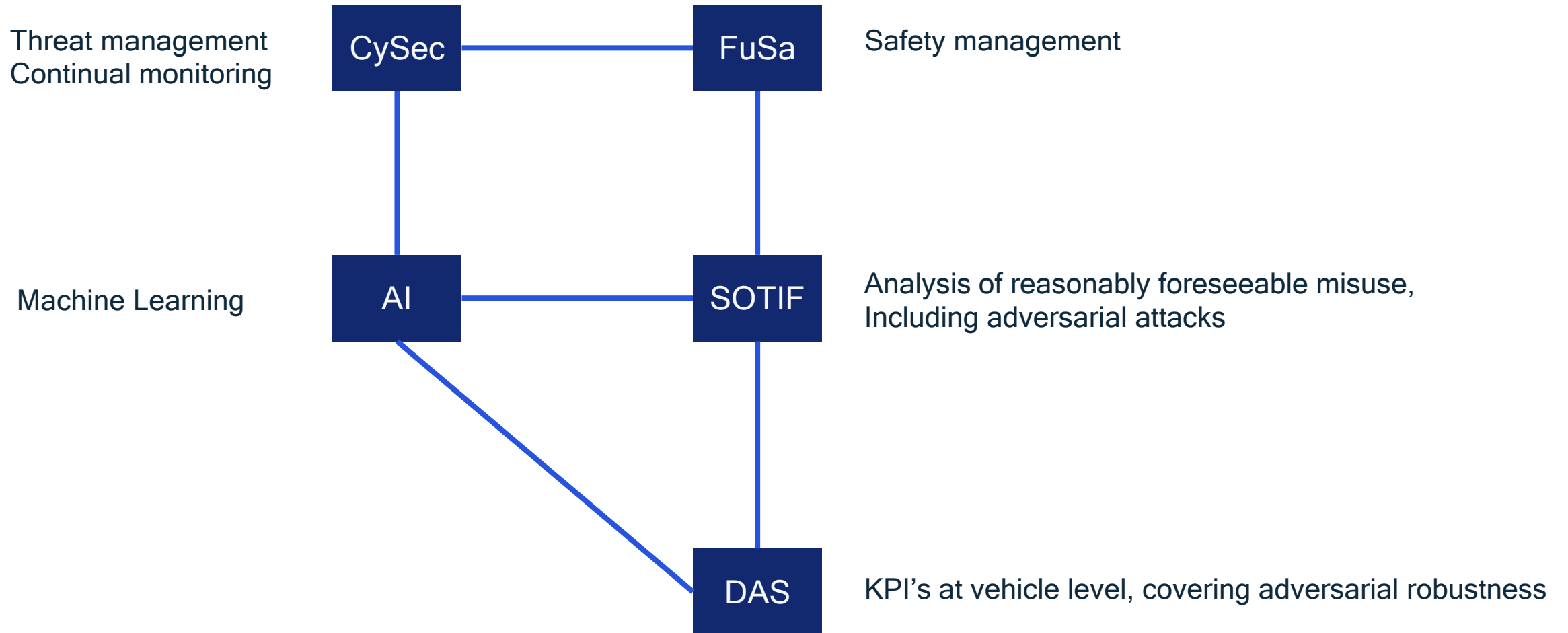


FUNCTIONAL  
SAFETY  
ISO 26262

Source sample text

## Which standards apply?

# Standards Landscape



# Risk Mitigation

## Cybersecurity

- To prevent white-box attacks: Attack path analysis, vulnerability analysis, risk treatment
- Continual monitoring for both cybersecurity but also adversarial image attacks

## ISO/PAS 8800 and Assurance of Machine Learning for use in Autonomous Systems (AMLAS)

- To prevent ground-truth attacks and poisoning attacks
- To prevent black-box attacks

## ISO/TS 5083

- For robustness against black-box attacks

## ISO 21488

- For resilience
- Analysis of hazards, triggering conditions, etc



# Defenses



## Proactive

Adversarial training  
Defensive distillation  
Model ensemble  
Network regularization  
Certified robustness  
...



## Reactive

Adversarial detection  
Adversarial transformation  
...





# Risk Mitigation

## Stay updated on the literature

- [https://www.researchgate.net/profile/Pan-He-9/publication/321936593\\_Adversarial\\_Examples\\_Attacks\\_and\\_Defenses\\_for\\_Deep\\_Learning/links/5a5cc59e0f7e9b4f7839614f/Adversarial-Examples-Attacks-and-Defenses-for-Deep-Learning.pdf](https://www.researchgate.net/profile/Pan-He-9/publication/321936593_Adversarial_Examples_Attacks_and_Defenses_for_Deep_Learning/links/5a5cc59e0f7e9b4f7839614f/Adversarial-Examples-Attacks-and-Defenses-for-Deep-Learning.pdf)
- <https://arxiv.org/pdf/2009.03728.pdf>
- <https://arxiv.org/pdf/2104.01789.pdf>
- <https://arxiv.org/pdf/1707.02476.pdf>
- <https://arxiv.org/pdf/1801.09344.pdf>
- ...



# Stay Safe

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