

SSF CLAS

Cyber-security for learning and control
systems

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- Machine-Learning based systems under attack

1. Assess the vulnerability of ML-based systems
2. Detect attacks, and devise secure ML algorithms
3. Illustrate the concepts in a smart building testbed



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- Machine Learning under attack?

At training time



- Machine Learning under attack?

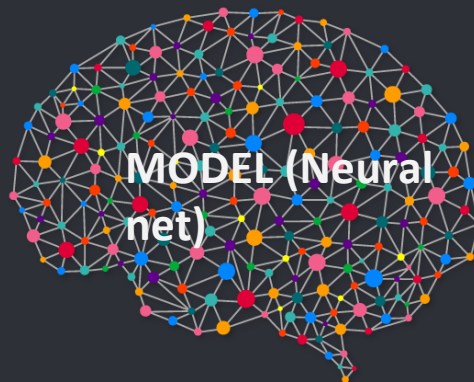
At training time



Slightly modify the data in an *adversarial* manner

- Machine Learning under attack?

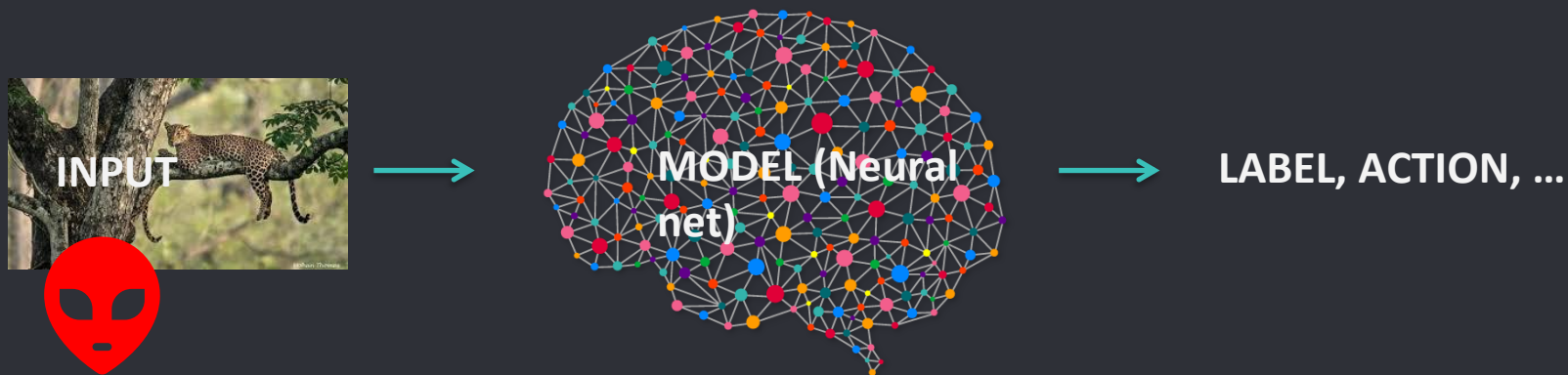
At test time



LABEL, ACTION, ...




- Machine Learning under attack?

At test time



Slightly modify the input to the model in an *adversarial* manner

- A first alarm at Google: Goodfellow et al. ICML 2015

	+ 0.07		=	
X		δX		$X + \delta X$
True input classified as a "Panda" with 57% confidence		Imperceptible adversarial "noise"		Adversarial example classified as a "Gibbon" with 99% confidence

A heuristic test-time attack (not optimized), and w/o detection consideration

- CLAS expected results in ML-based controlled systems

Optimal attacks



Risk management



KTH live-in lab



Optimal detection algo



Privacy attack



?



RESULTS SO FAR

A generic optimization framework for attack design / detection

Example: attacking at test-time a control policy obtained through deep RL

1. Maximal detection rate of an attack : $\pi \quad \mathbb{P}[\text{detect}] = e^{-I(\pi)}$
2. Optimal attack :

$$\min_{\pi} R(\pi) \quad s.t. \quad I(\pi) \geq \gamma$$

Atari games (here pong)

Game without adversary



Main Agent Score	2nd Agent Score
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0	0
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Original Frame



Change



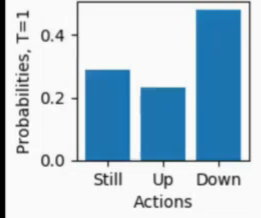
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Final image

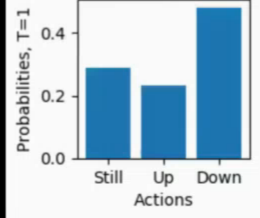


0	0
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Original actions



Adversarial actions



● Other results / current activities

- Optimization framework applied to
 - Various types of attacks
 - Multi-sensor systems
 - Attacks at training time ...
- Securing ML-algorithms
- +50 academic publications

Demonstrator: KTH Live-in lab (privacy and security)

Windows opening
Light



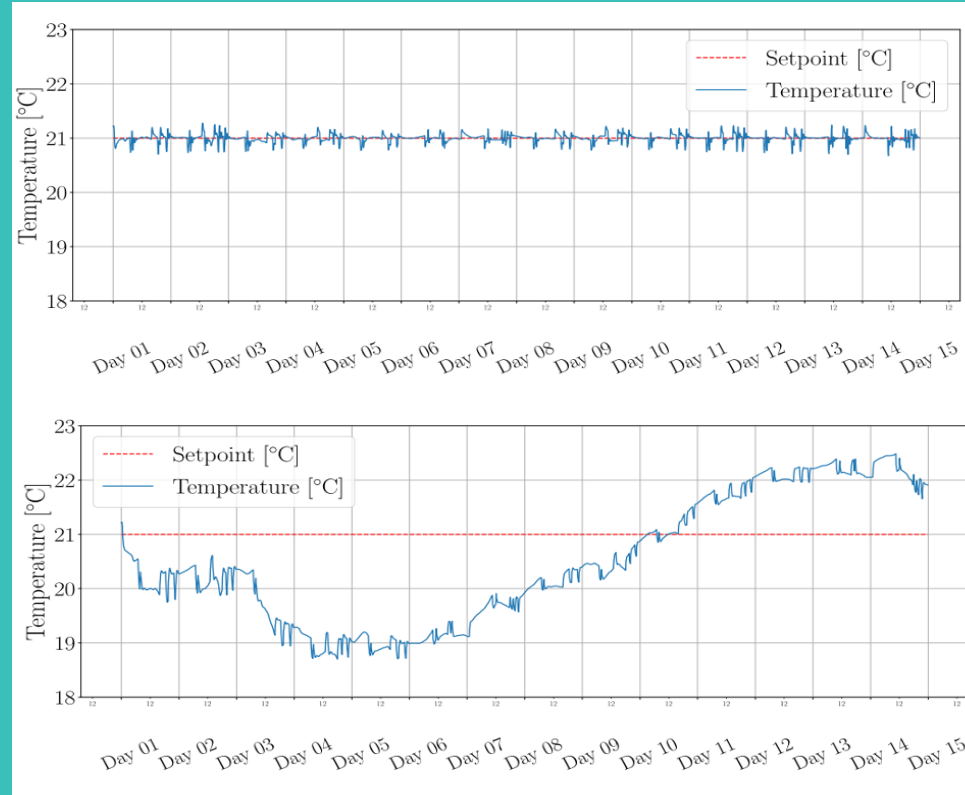
Temperatures
CO₂ concentration

Air flows



Boreholes
monitoring

Heat pump control





Desired collaborations

We can assess the **security-level** of ML algorithms and models

... anyone running ML-based systems is welcome to contact us