

**Westfälische
Hochschule**

Gelsenkirchen Bocholt Recklinghausen
University of Applied Sciences

AI for IT security *and* IT security for AI

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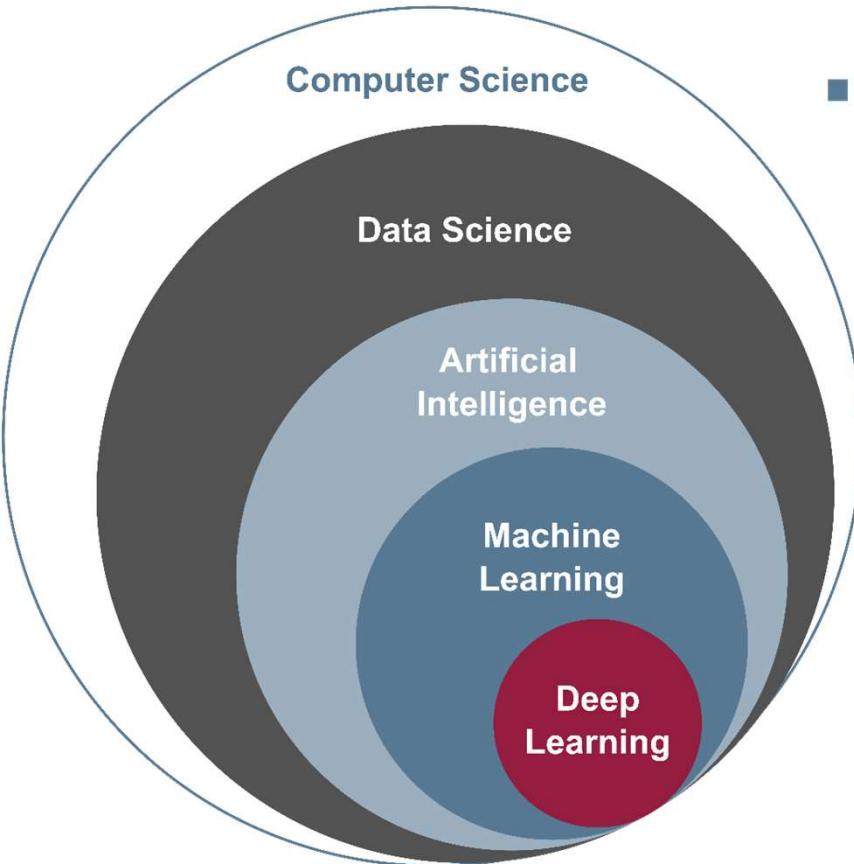
Chairman of the board of the IT Security Association TeleTrusT

Member of the board of the Internet industry association eco.



Classification

→ Artificial intelligence (AI)

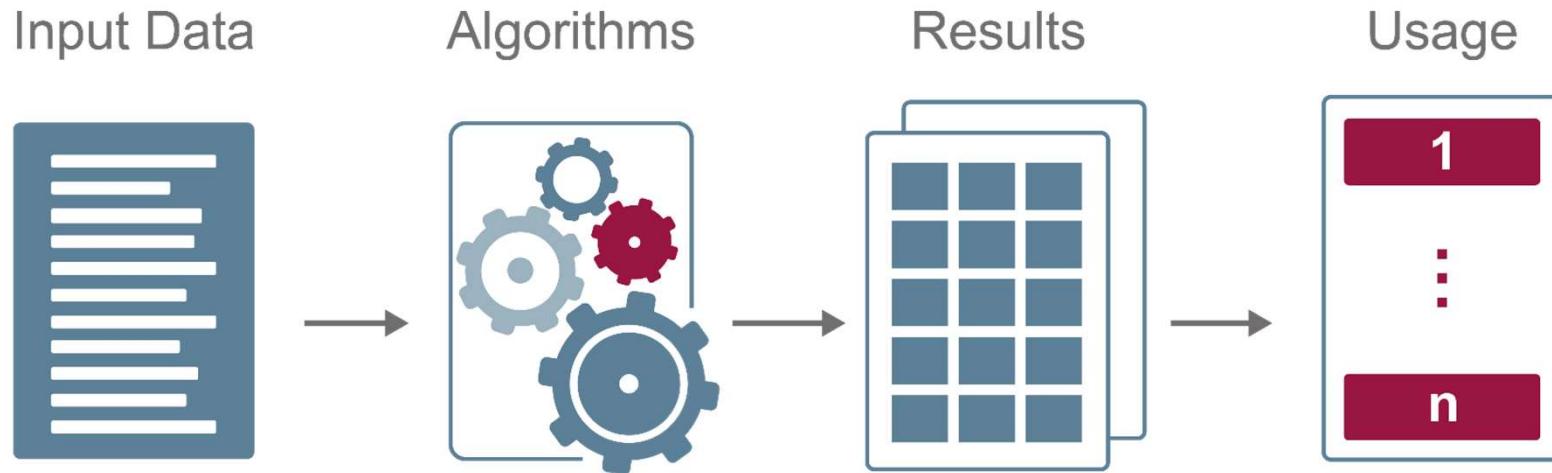


- Data science generally refers to the **extraction of knowledge from data**.
- Artificial intelligence translates intelligent behavior into algorithms.
 - **Strong "Artificial Intelligence"** *automatically replicate „human-like intelligence“.*
 - Superintelligence, **Singularity** (*“Machine” improves itself, is more intelligent than humans ... future*)
 - **Weak “artificial intelligence”** (*machine learning – successfully implemented today*)
 - **Machine learning** is "artificial" **generation of knowledge from experience** (**in data**) by computer.
 - **Deep learning** is an important **improvement** of machine learning

Large Language Model (LLM) like ChatGPT

Machine learning

→ Workflow



Input Data

Quality of the input data: completeness, representativeness, timeliness ...

Algorithms (ML)

Support Vector Machine (SVM), k-Nearest Neighbor (kNN) ... Deep Learning

Results

Results from the processing (AI algorithm) of the input data.

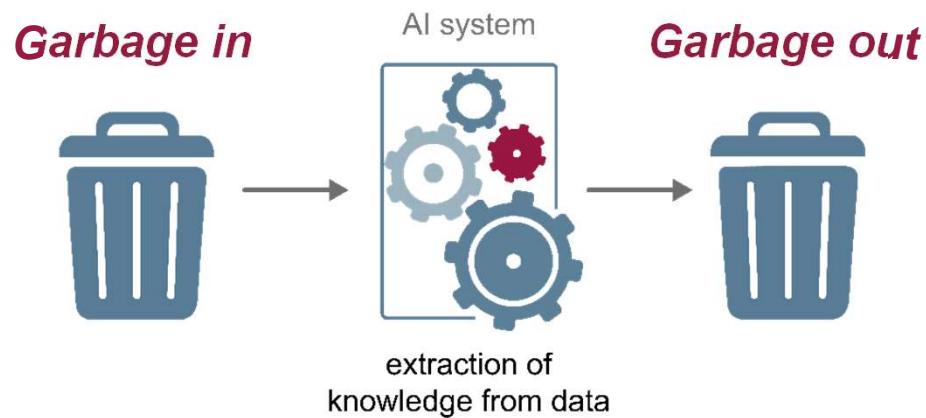
Usage

The application decides how to use results (trust).

Trustworthiness of AI

→ Quality of the data

Paradigm

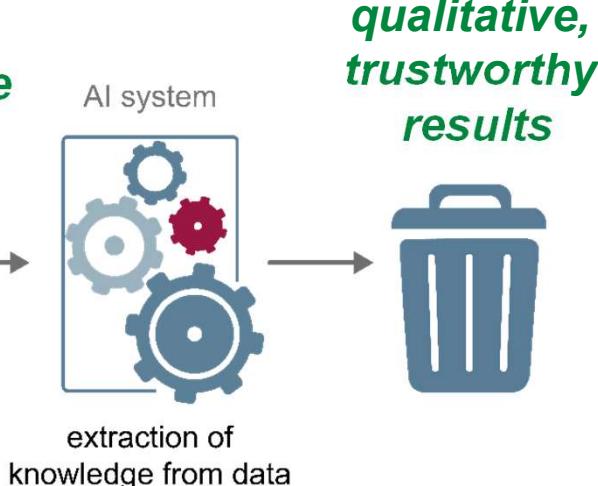


Standards for data quality:

- Content of the data and correctness
- Traceability of data (including data sources)
- Completeness and representativeness
- Availability and timeliness

Motivate high quality and
secure **sensors**

*high
data quality of the
input data*



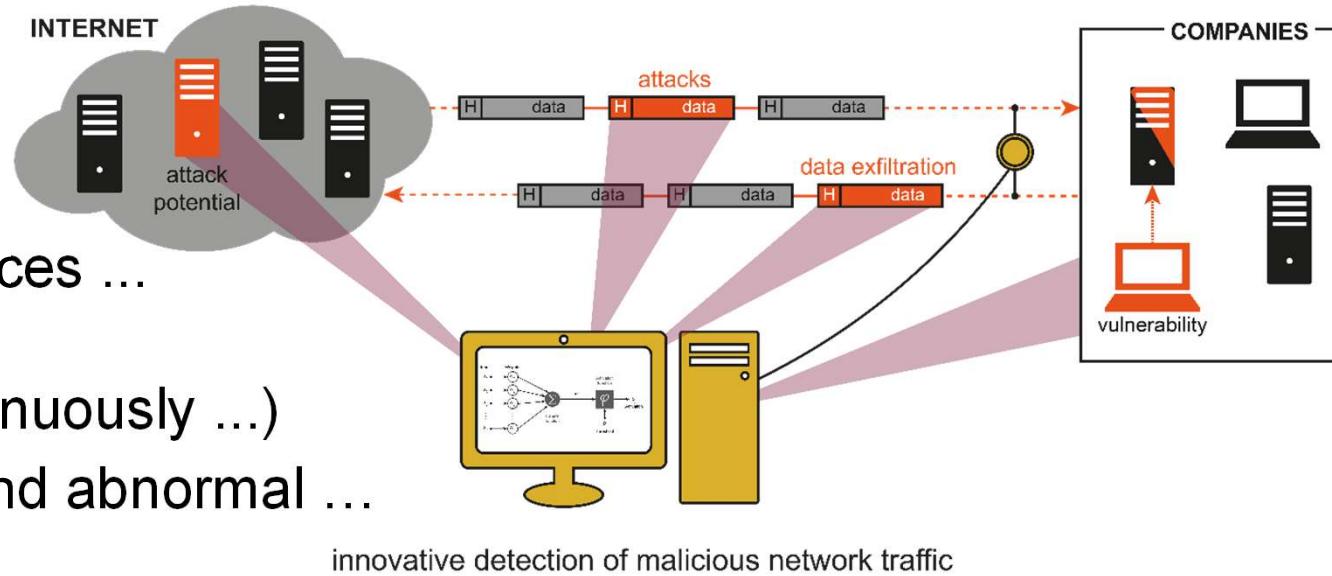
Other Ideas:

- Establish data pools
- Promote exchange of data
- Create interoperability
- Push open data strategy

Artificial intelligence → for IT security

- Increasing the **detection rate** of attacks

- Network, IT end devices ...
- adaptive models
(independently, continuously ...)
- Difference: normal and abnormal ...



- **Support / Relief from IT security experts**
(of whom we do not have enough)
 - Finding **important** security-relevant events (prioritization)
 - **(Partial) autonomy** in response ... resilience ...

- **Improvements to existing IT security solutions**
 - AI contributes to increased impact and robustness
 - For example: risk-based and adaptive authentication



Research projects

→ Alert-System for online banking

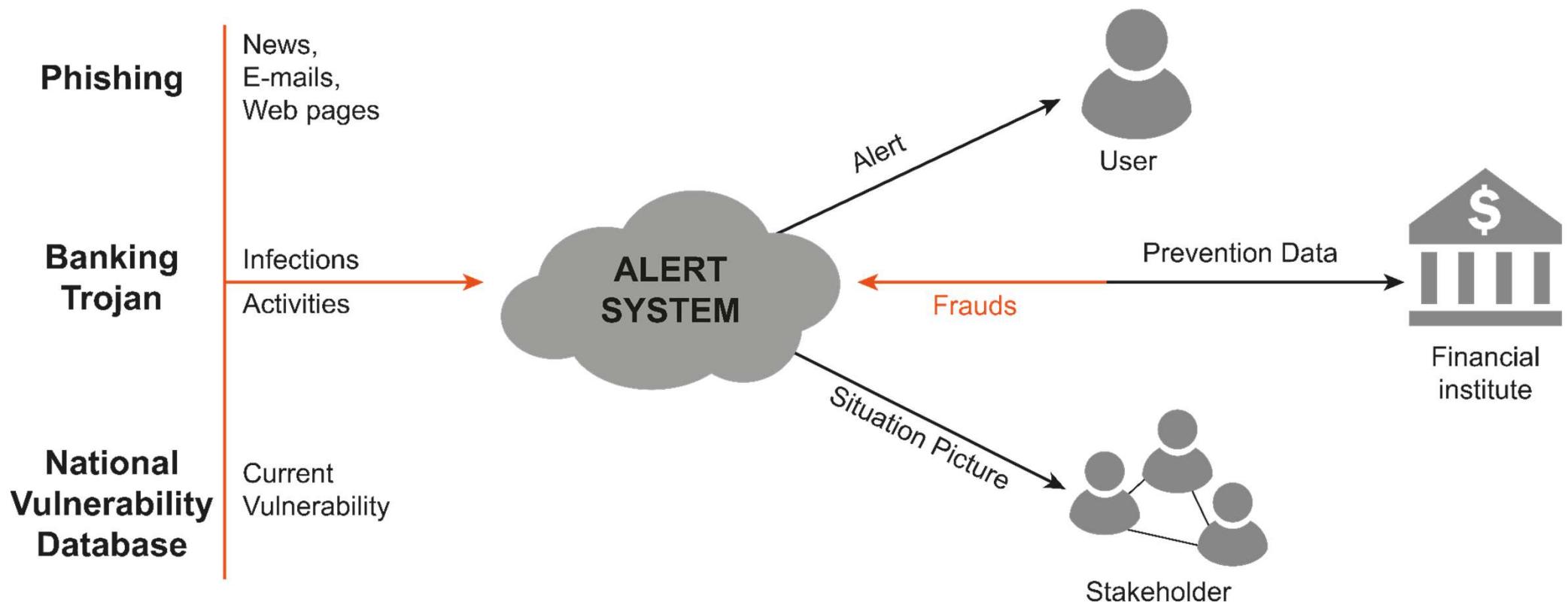
- **How could a solution look like?**
 - Daily warnings in the event of an increased risk situation (online banking)
→ enable the bank customer and the bank to react quickly and appropriately
 - Instruct the users when there are dangers
→ so that the bank customer can behave "correctly"

- **Approach of the alert system**
 - Identify **security metrics** for fraud
 - Determine **danger situation** with AI
 - **Warn** users and banks



Alert-System for online banking

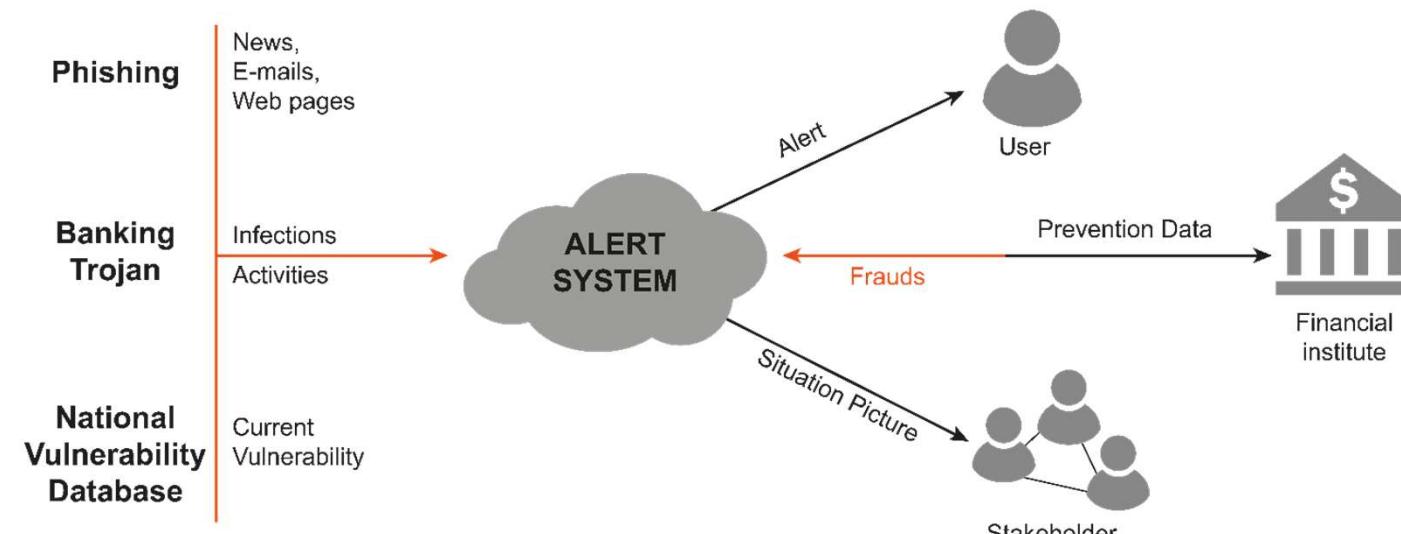
→ Basic concept



Alert-System for online banking

→ Numbers for the test period of 456 days

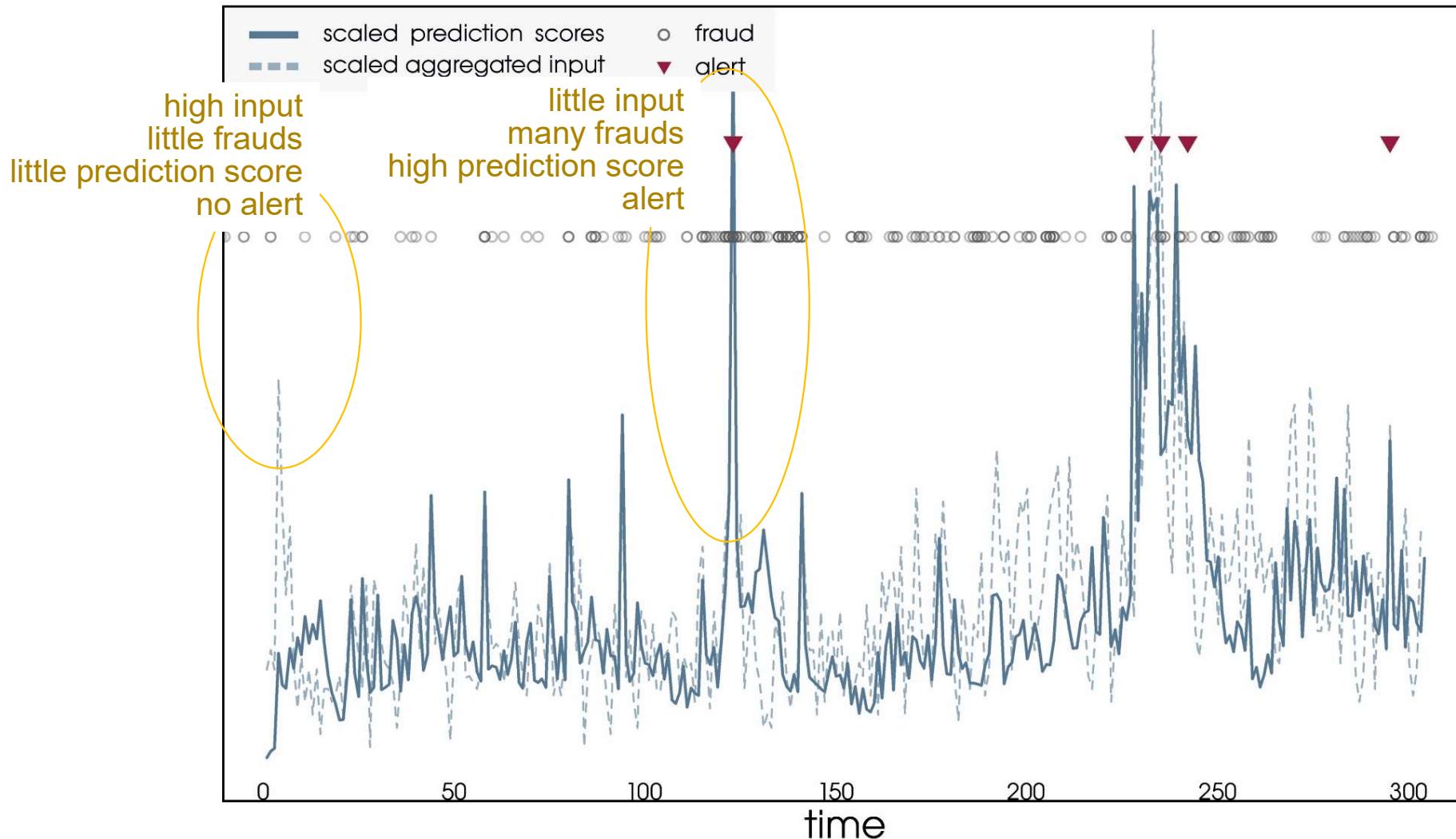
- 1.904 News (phishing attack) – “Stackoverflow Network”
- 5.589 **E-mail** (phishing attack) – „Spam Archive“
- 2.776 Phishing **websites** – „PhishTank“
- 23.184 **infections** of banking Trojans (malware) - Anti-malware companies
- 875 relevant **vulnerabilities** (NVD)
- 459 successful **fraud cases** in online banking - banking group



1/3 for the training period (152 days) 2/3 for evaluation period (304 days)

Assess the result → k-Nearest Neighbor

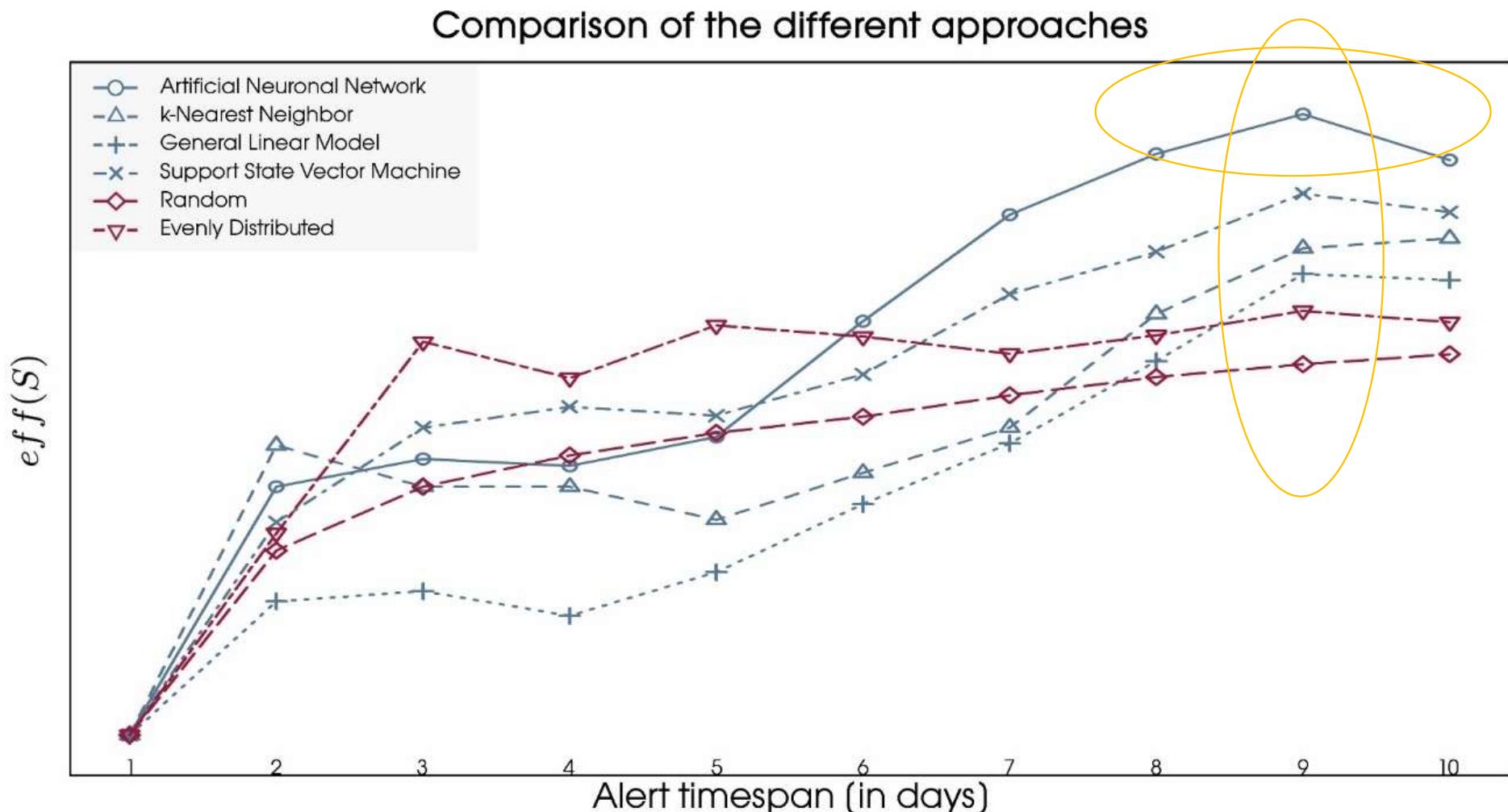
k-Nearest Neighbor



Results

→ Comparison of the different methods

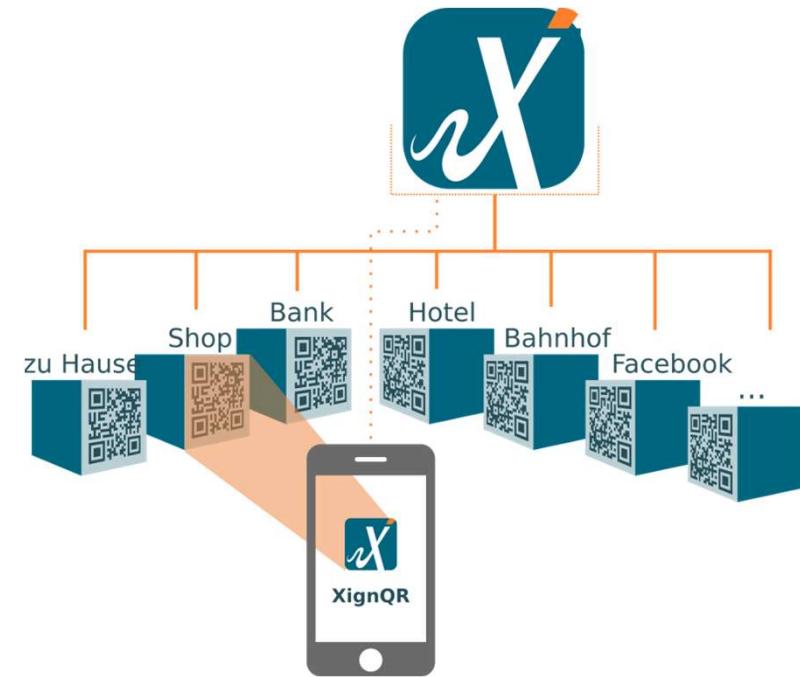
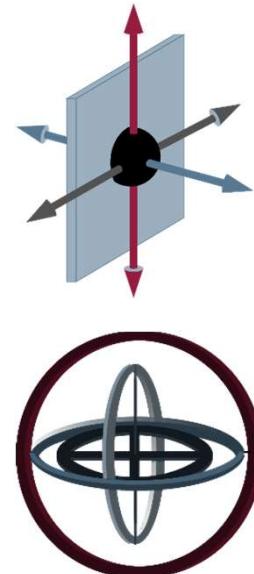
„But three times as much time for training
for Artificial Neural Networks“



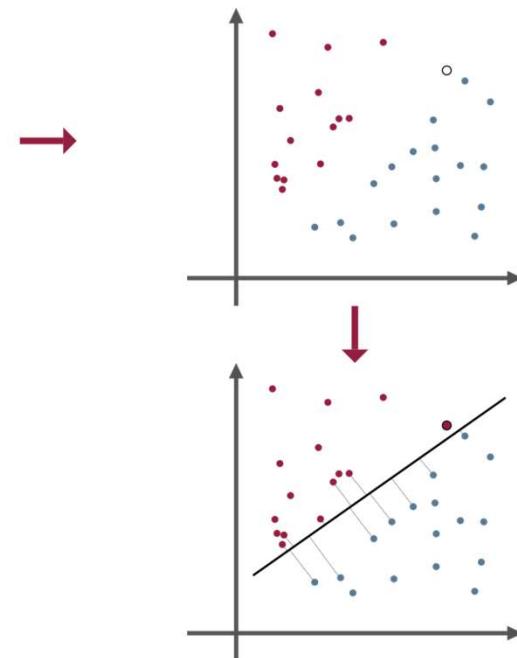
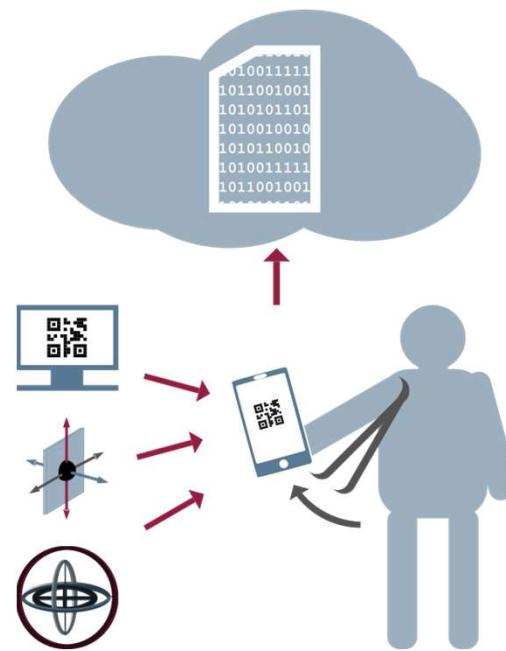
Research projects

→ Passive Authentication

- A user is automatically detected by the way of scanning the QR code.
- Throughout the process, passive biometric movement data is measured.
- Data collection by
 - Accelerometer
 - Position sensor

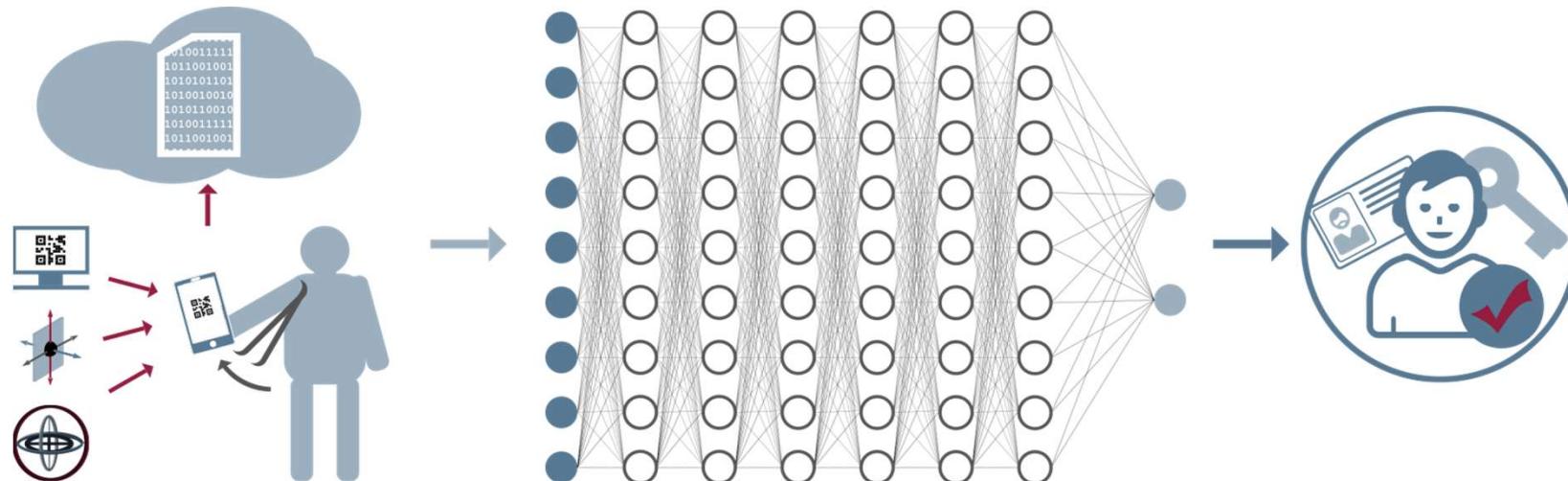


Passive Authentication → Support-Vector-Machine (SVM)



- **Input data:**
 - User takes the smartphone from pocket
 - Measure **location** and **acceleration** of the smartphone
- **ML algorithm:**
 - Data is classified by a model
 - red match is **positive** classification
 - blue a **negative** classification (e.g. other users)
- **Output:**
 - Authentication is either successful or fails (95 %)

Passive Authentication → Artificial Neural Networks



■ Input data: ■ ML algorithm: ■ Output:

- Location and acceleration data of the user are generated

■ ML algorithm:

- Input data is processed in the artificial neurons in the layers

■ Output:

User	Accordance
0	0,059 %
1	99,85 %
2	0,087 %

time, type, x, y, z
271, Accelerometer, -0.07606506, 9.173798, 3.6333618
277, Accelerometer, 1.0681152E-4, 9.146423, 3.5619507
279, Gyroscope, 0.027664185, 0.06774902, 0.02182006
...

```
[[5.9110398e-04 9.9853361e-01 8.7528664e-04]]  
Predicted Class [1]  
Predicted Person: Sandra Kreis
```

AI for IT Security

→ Further examples

- Log analysis
- Malware detection
- Security Information and Event Management (SIEM)
- Threat Intelligence
- Voice recognition
- Image recognition (ID card, Videoldent ...)
- Authentication method
- Fake News
- IT Forensics
- Secure software development (new possibilities with ChatGPT)
- ...

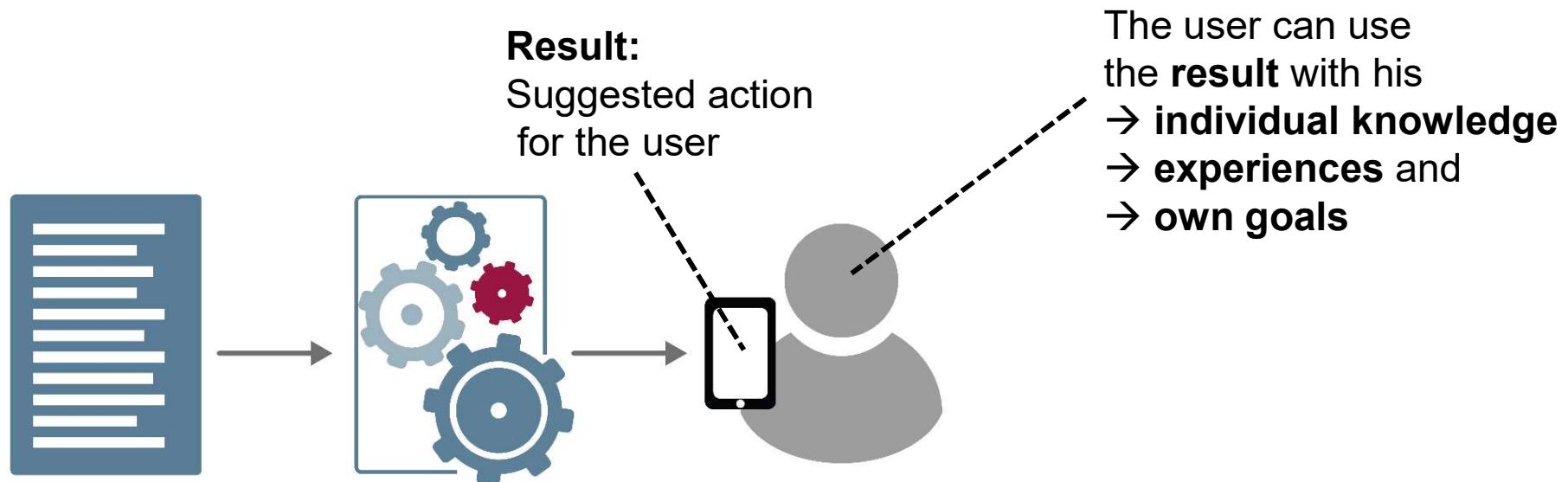
Artificial intelligence → Opportunities and risks

- When it comes to **individual knowledge** and **complexity of thinking**, humans are still superior to algorithms! +
- **Algorithms can more quickly generate knowledge** from existing data! +
- Individual knowledge + algorithms knowledge = +++
- **Practical Problem Medicine / Watson**
 - Diagnostics (machine)
 - Liability (human)

Trustworthiness

→ types of validation of results

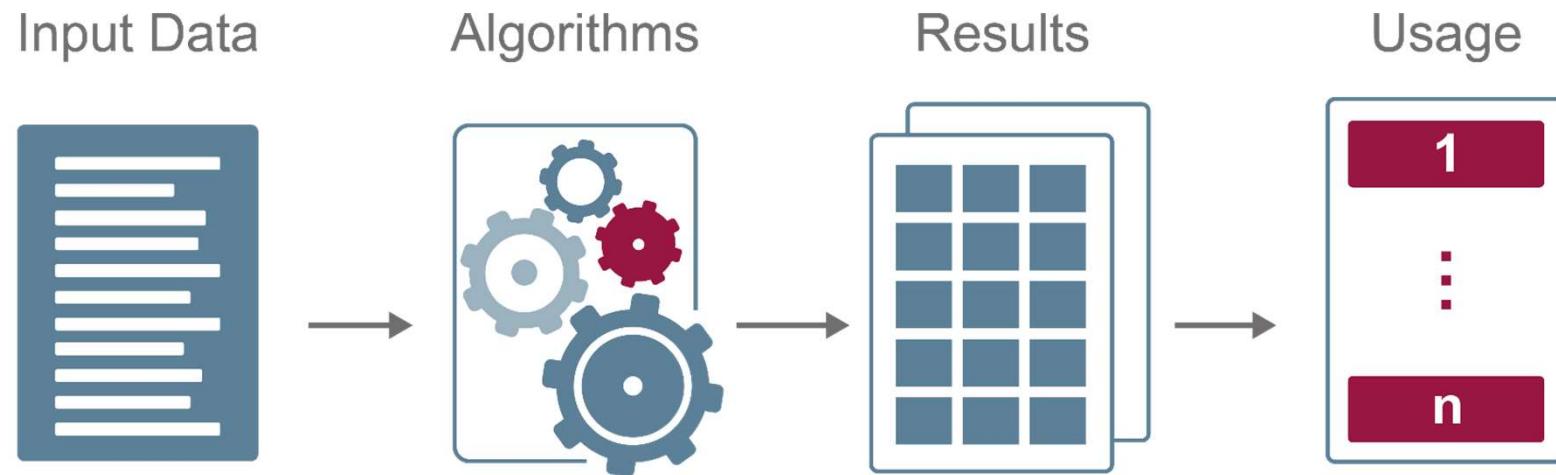
- „Keep the human in the loop“
 - AI result must be understood as a **recommendation for the user**.
 - This promotes the **self-determination** of users and increases their trustworthiness.



- **Automated applications** (e.g., autonomous driving)
 - Simulation, test and **validation**
 - Responsibility, **liability** and insurance

Attacks → on machine learning (AI)

Hackers attack and manipulate the workflow (“result”)

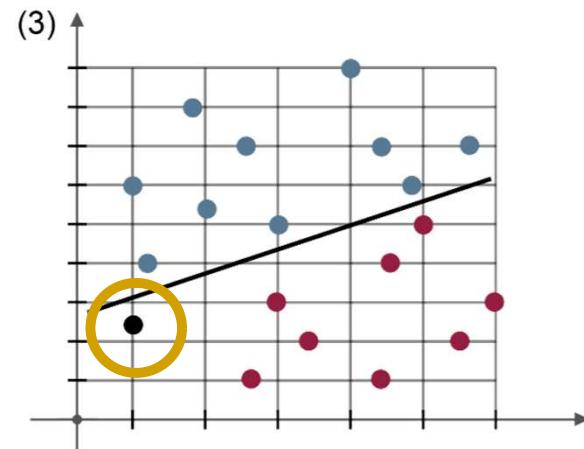
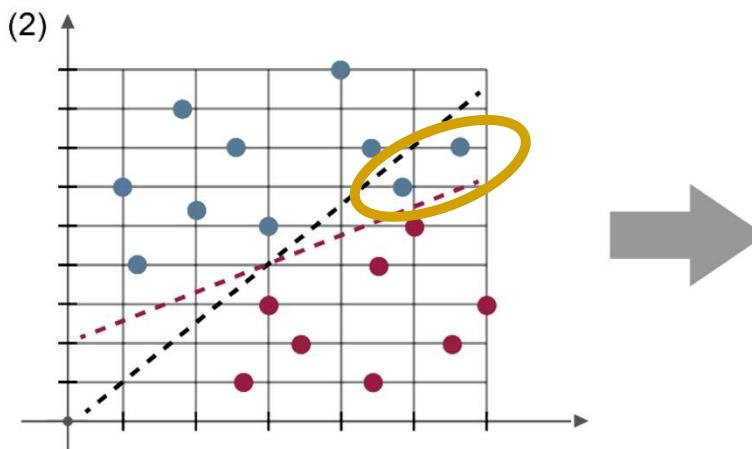
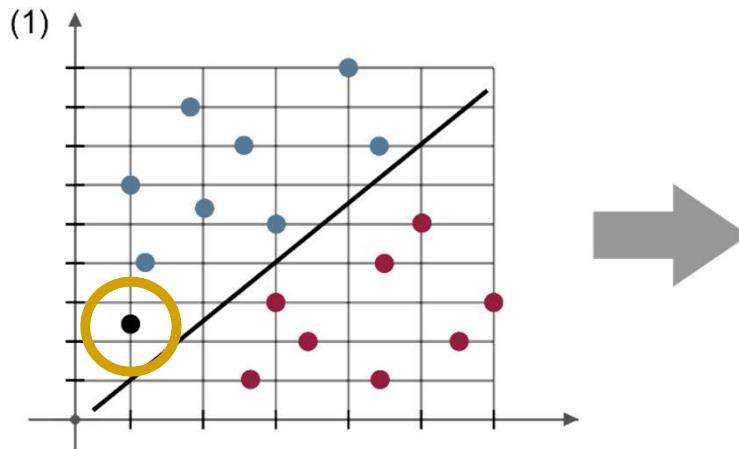


- ***Input data (input)***
- ***Algorithms / Models***
- ***Results (output)***
- ***Usage***

Attacks on machine learning

→ Manipulation of training data (Poisoning Attack)

- (1) Normal classification of a new input.
(new black dot belongs to the blue class)
- (2) Example: manipulation of training data
 - Incorrectly classified data will be injected into the training phase as an attack (two more blue dots).
 - This manipulates the straight line of the model for classification (straight line becomes flatter).
- (3) This can be used by an attacker to create wrong classifications.
(now the new black dot belongs to the red class)



Attacks on machine learning

→ Manipulation of Input Data (Evasion Attack)

Normal traffic sign



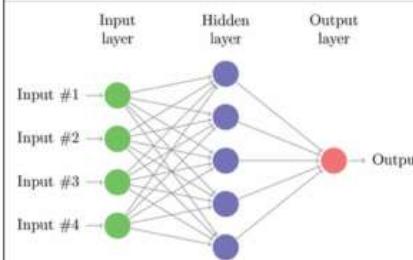
Self-driving car's front camera



Benign sign



Neural network classifier



Classification output:
Speed limit (80)



Correct

(a) Operation of the computer vision subsystem of an AV under *benign conditions*

Fake traffic sign



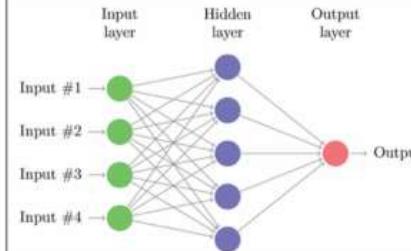
Self-driving car's front camera



Adversarial sign



Neural network classifier



Classification output:
Stop



Incorrect

Car unexpectedly
stops on a highway



(b) Operation of the computer vision subsystem of an AV under *adversarial conditions*

Fig. 1. **Difference in operation of autonomous cars under benign and adversarial conditions.** Figure 1b shows the classification result for a drive-by test for a physically robust adversarial example generated using our Adversarial Traffic Sign attack.

Secure AI

→ Protection of the implementation and data

State of the art

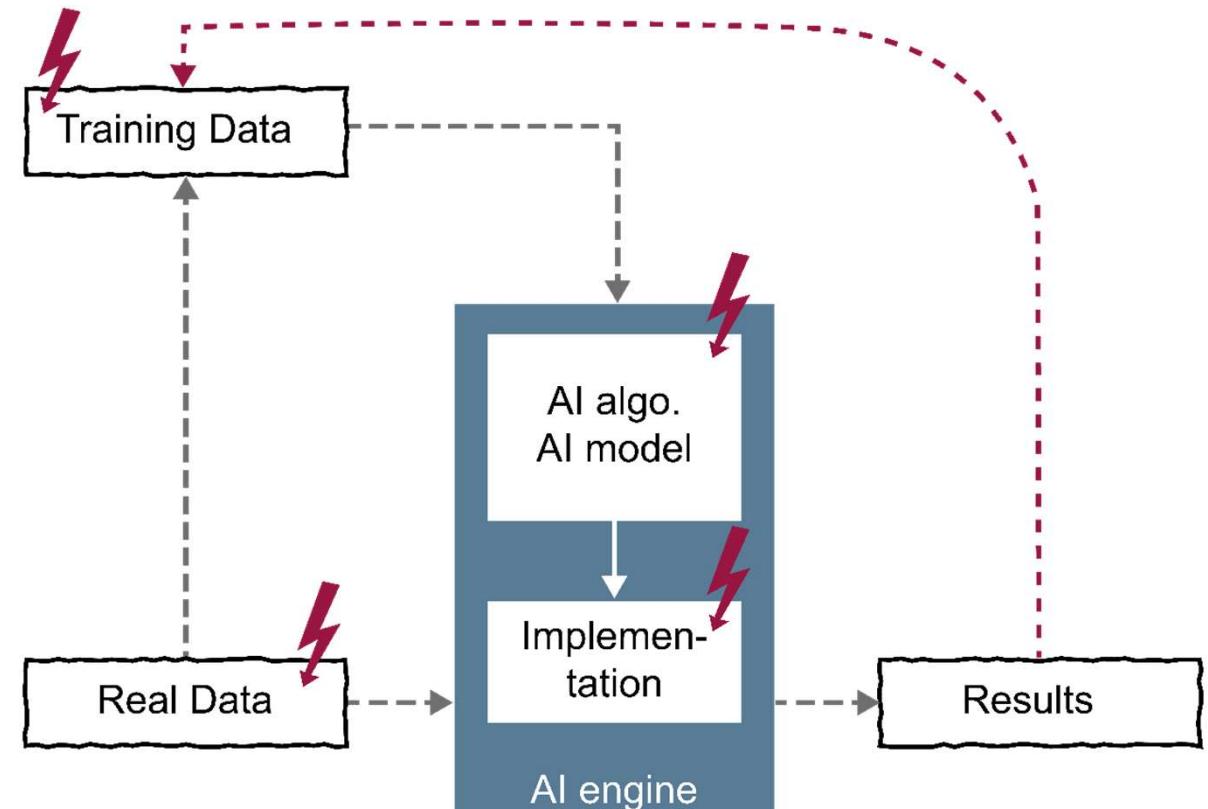
IT security measures

for protection

- the ***data*** (training, real, result),
- the ***AI engine*** and
- the ***application***

Security goals:

- ***Integrity***
(detection of data manipulation)
- ***Confidentiality***
(protection of business secrets)
- ***Data protection***
(protection of personal data)
- ***Availability***
(of the application and results)



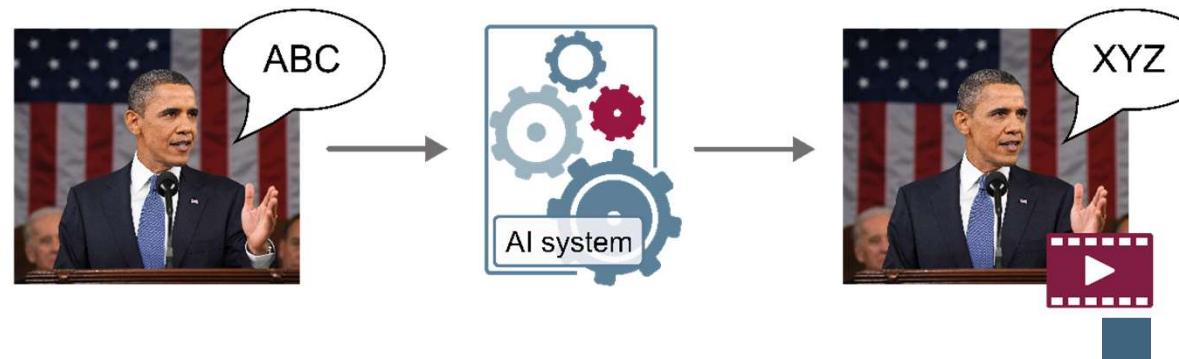
***Use of a
high quality
AI technology***

Artificial intelligence

→ Attackers use AI

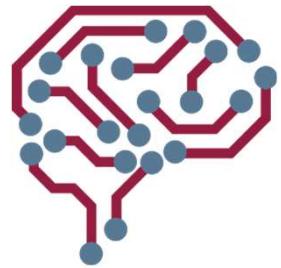
Hacker also uses AI for their own purposes (dual-use)

- Vulnerability search (faster attack, new attack vectors ...)
- Social engineering (chat bots like ChatGPT ...)
- Password cracker
- Polymorphic malware (programming with ChatGPT)
- New attack structures and procedures
- Video manipulation (deep fake)
 - "Fake Obama Video,"
 - "Make Putin Smile Video"



AI for IT security - IT security for AI → Result and outlook

- **AI / ML is an important technology in the field of IT security**
 - Detect threats, vulnerabilities, attacks ...
 - Support of IT security experts
 - Secure software development
 - ...
- **We need to secure our AI to be able to produce trustworthy results**
 - Hackers attack and manipulate data, algorithm/models and results
 - ...
- **Balance of power for the future between attacker and defender**
 - The **attackers** use AI for their attacks **very successfully**
 - The defenders should do this more and also together
 - ...



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AI for IT security and IT security for AI

*With **secured Artificial Intelligence**
into a **more secure and trustworthy** digital future!*

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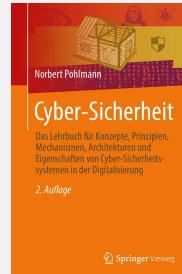


Anhang / Credits

Wir empfehlen

Cyber-Sicherheit

Das **Lehrbuch** für Konzepte, Mechanismen, Architekturen und Eigenschaften von Cyber-Sicherheitssystemen in der Digitalisierung“, Springer Vieweg Verlag, Wiesbaden 2022
<https://norbert-pohlmann.com/cyber-sicherheit/>



7. Sinn im Internet (Cyberschutzraum)

<https://www.youtube.com/cyberschutzraum>



Master Internet-Sicherheit

<https://it-sicherheit.de/master-studieren/>



Glossar Cyber-Sicherheit

<https://norbert-pohlmann.com/category/glossar-cyber-sicherheit/>



It's all about Trust!

<https://vertrauenswürdigkeit.com/>



Quellen Bildmaterial

Eingebettete Piktogramme: Institut für Internet-Sicherheit – if(is)

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WWW

<https://www.internet-sicherheit.de>

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<https://twitter.com/ProfPohlmann>

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<https://www.youtube.com/user/InternetSicherheitDE/>

Prof. Norbert Pohlmann

<https://norbert-pohlmann.com/>

Der Marktplatz IT-Sicherheit

(IT-Sicherheits-) Anbieter, Lösungen, Jobs, Veranstaltungen und Hilfestellungen (Ratgeber, IT-Sicherheitstipps, Glossar, u.v.m.) leicht & einfach finden.
<https://www.it-sicherheit.de/>

Literature

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- N. Pohlmann: Lehrbuch „Cyber-Sicherheit“, Springer Vieweg Verlag, Wiesbaden 2022
Druckausgabe (ISBN 978-3-658-36242-3) und eBook (ISBN 978-3-658-36243-0).

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