Introduction to Trustworthy Machine Learning

Franziska Boenisch and Adam Dziedzic Course on Trustworthy Machine Learning





Our SprintML Lab (Trustworthy ML)

S - ecure

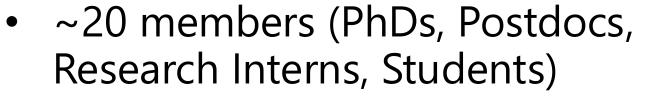
P - rivate

R - obust

In - terpretable

T - rustworthy





- 8 different nations
- Visit: https://sprintml.com/
- Sponsors: G-Research & OpenAl



Franziska Boenisch



Franziska Boenisch

AbiBac: Berlin & Lyon

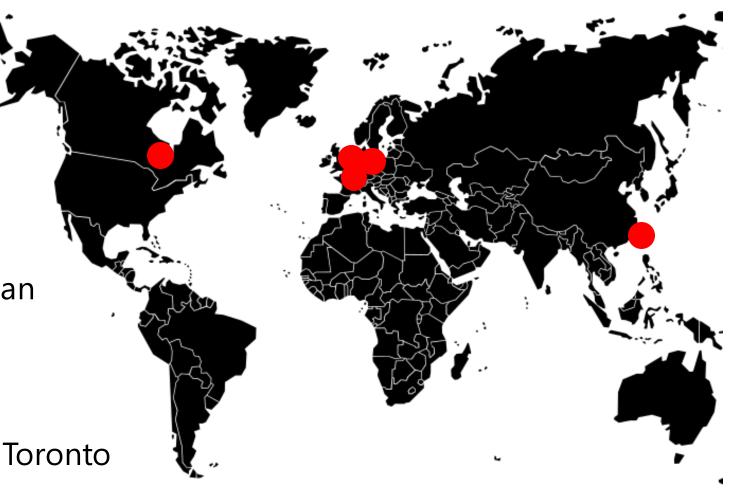
Chung Cheng University, Taiwan

TU Eindhoven, Netherlands

PhD @ Fraunhofer AISEC

Postdoc @ Vector Institute in Toronto

Faculty @ CISPA



Adam Dziedzic

Adam Dziedzic

Warsaw University of Technology

Technical University of Denmark

Barclays Investment Bank in London

EPFL & CERN in Switzerland

Google & Microsoft in the US

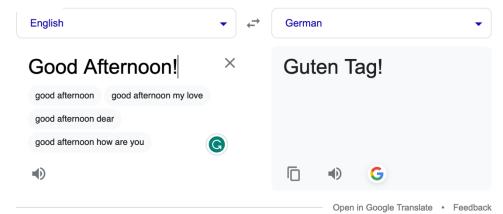
PhD @ University of Chicago

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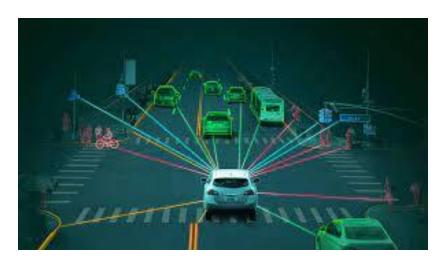


Machine Learning Fuels Many Applications











A Glitch in Google's Translation Service



Google The service outputs its memorized content.

From the Bible (1 Kings 7:2)



Source: https://www.vice.com/en/article/j5npeg/why-is-google-translate-spitting-out-sinister-religious-prophecies

Catastrophic Failures of Self-Driving Cars

в в с

Tesla cars in fatal crashes were on Autopilot.







Source: https://www.bbc.com/news/world-us-canada-43604440

ML Deployed in Adversarial Setting

The New Hork Times

Microsoft created a Twitter bot to learn from users. It quickly (<16 hours) became a racist jerk.





@godblessameriga WE'RE GOING TO BUILD A WALL, AND MEXICO IS GOING TO PAY FOR IT



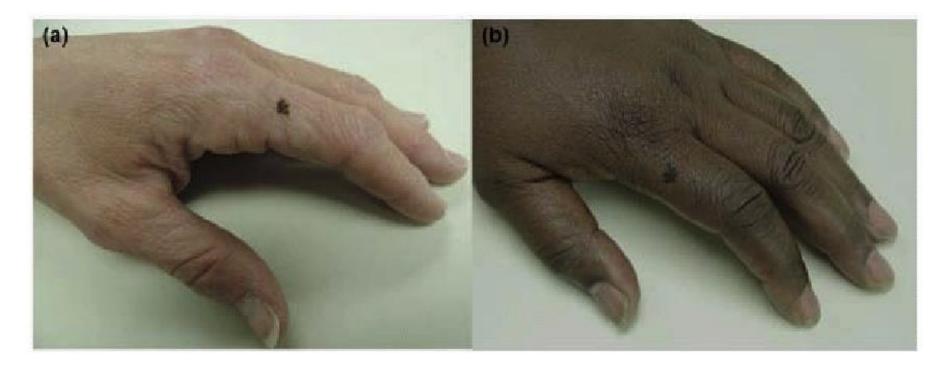
1:47 AM - 24 Mar 2016

Sources: https://en.wikipedia.org/wiki/Tay_(chatbot)#cite_note-bbc_swear-1
https://www.nytimes.com/2016/03/25/technology/microsoft-created-a-twitter-bot-to-learn-from-users-it-quickly-became-a-racist-jerk.html

Bias in Machine Learning Models

Guardian

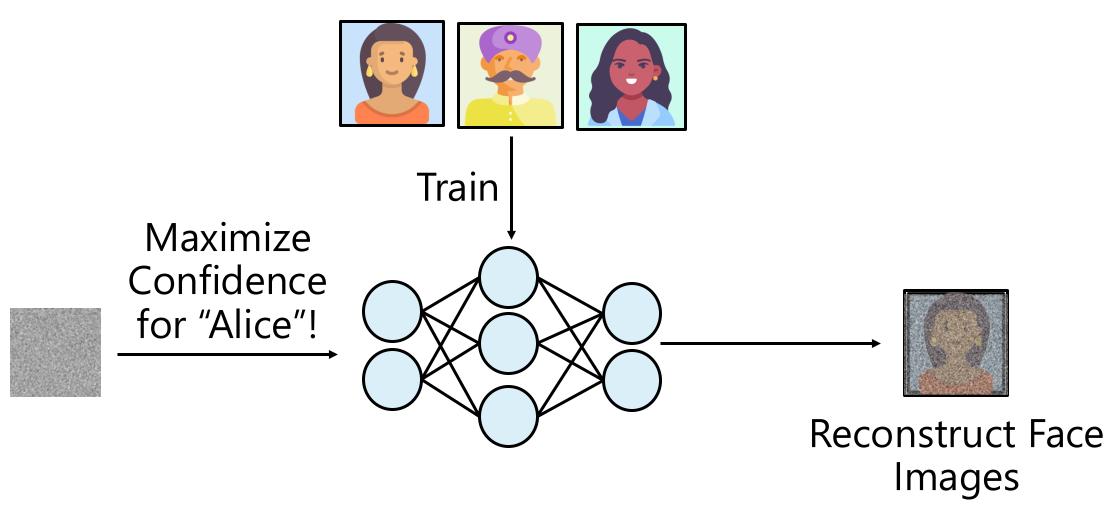
Al skin cancer diagnoses risk being less accurate for dark skin – study.



Sources: https://www.theguardian.com/society/2021/nov/09/ai-skin-cancer-diagnoses-risk-being-less-accurate-for-dark-skin-study

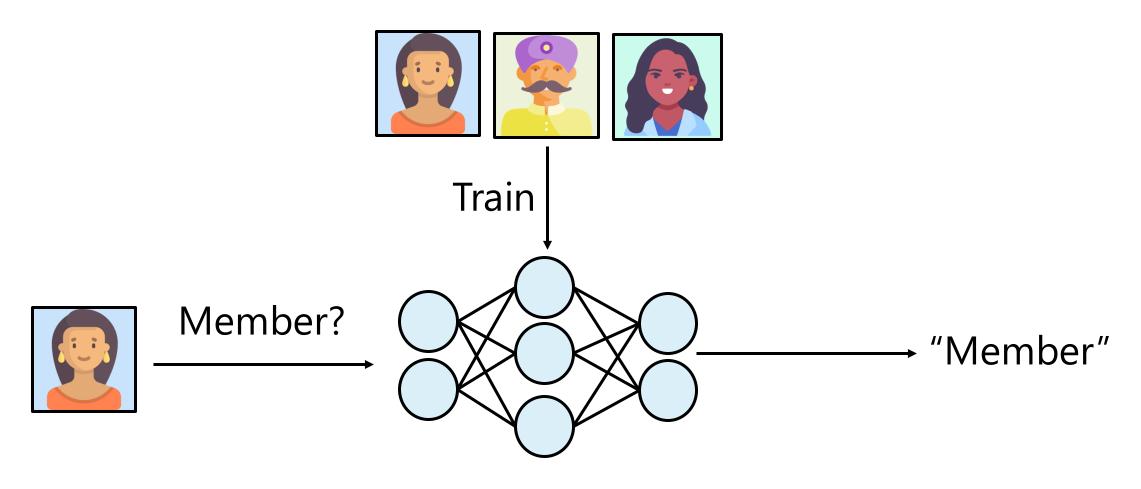
What are the risks to Trustworthy ML?

Privacy for Machine Learning

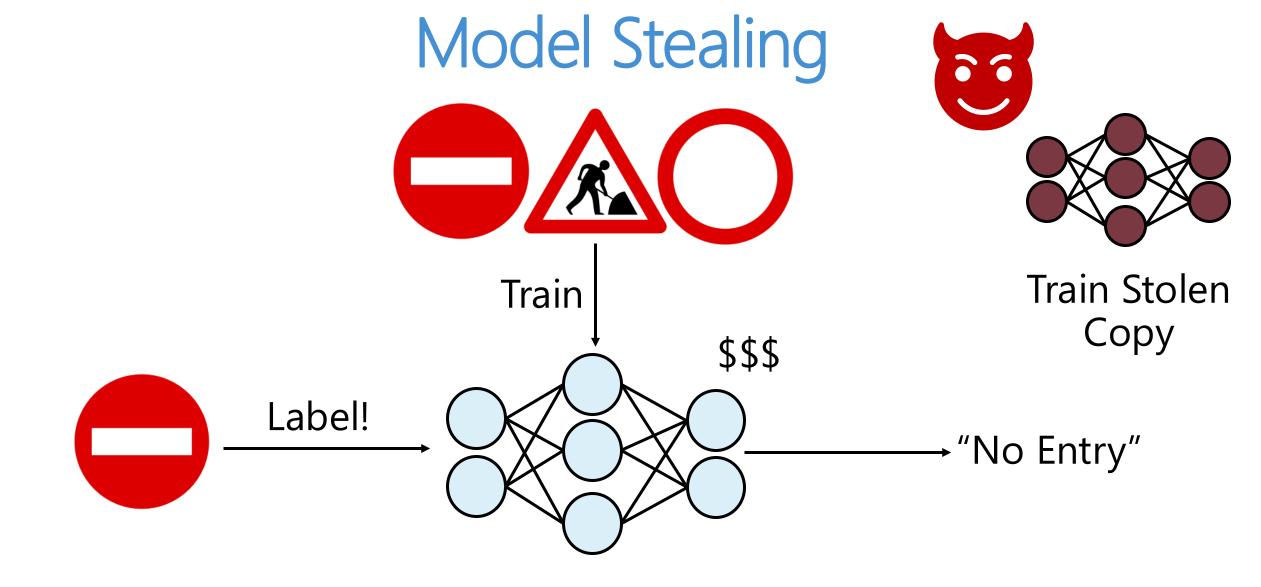


Data Reconstruction

Privacy for Machine Learning



Membership Inference

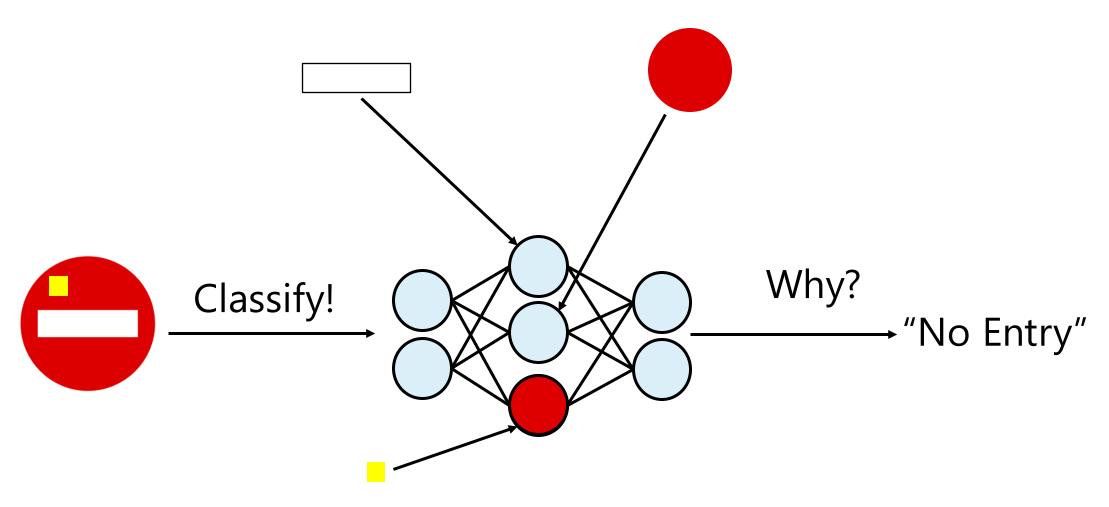


Model Stealing

Robustness of Machine Learning Add Human-Imperceptible Noise! Train Classify! "Right of Way"

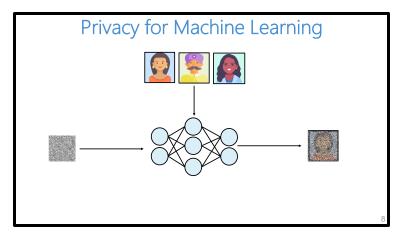
Adversarial Examples

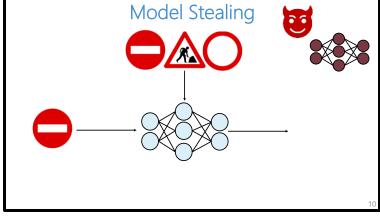
Interpretability of ML Predictions

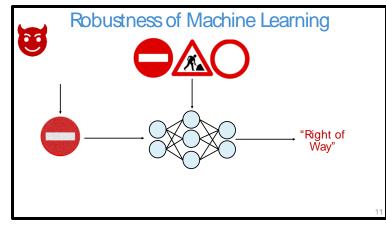


Understand Predictions

Many Facets of Trustworthy Machine Learning







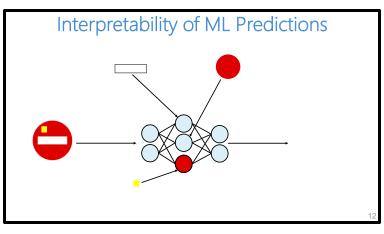
Privacy

Model Stealing

Robustness

Collaboration

Fairness



Governance

Security

Explainability

Administrative Overview

Organization

Flipped classroom:

- Lectures:
 - Published on YouTube: https://www.youtube.com/playlist?list=PLNfU-a7sxlwvS7dhnOPdFtvhdNcrnufEW (short: https://bit.ly/3Gaz6mW)
 - Please watch and prepare independently
- Questions:
 - Every student submits 2 questions on Forum on Friday by 5 PM before the lecture
 - Questions discussed during lecture hours on Wednesday (2PM-4PM)
- Example:
 - Until Friday 25th of April (5 PM), watch the lecture on Privacy I and submit your questions on CMS Forum

Where and When?

Wednesdays from 2PM-4PM, CISPA, Lecture Hall Ground Floor (0.05)

30.04. Privacy I

07.05. Privacy II

21.05. Model Stealing (Supervised)

28.5. Defenses against Stealing (SSL)

04.06. Robustness

04.06. Midterm Exam

11.06. Collaborative Learning I

18.06. Collaborative Learning II

25.06. Fairness & Bias

02.07. Explainability

09.07. Security & Governance

09.07 Summary & Questions

31.07. Final Exam

Accessing the material

- Lecture videos on Youtube: https://bit.ly/3Gaz6mW
- Lecture notes and handouts on CMS: https://cms.cispa.saarland/tml2025/
- All related work linked at the end of the presentations
- Homework assignments published on CMS
- Grades on CMS

Overview of Assignments

1. Membership Inference Attack: was a model trained on these data points?



2. Model Stealing: extract a model from an API.



3. Model Robustness: defend a model against adversarial examples.



4. Backdoor Attacks: remove a backdoor from a model (or **Explainability**).

Assignments: Due Dates & Deliverables

4 Programming assignments:

1	1			1	•		2000
	Imple	ementing	a mem	bership	inference	attack	28.05.

2.	Stealing	a model behind an API	11.06.
— •			

- 3. Training a robust classifier 02.07.
- 4. Removing a backdoor or Explainability 30.07.

Leaderboard for all assignments up on opening. Final submission of artefacts for evaluation (e.g., report) +Submission of code (link to a private GitHub Repo).

Submissions of assignments in groups of 2.

Grading

40% Assignment (10% per assignment)

20% Midterm Exam

40% Final Exam

Getting in Touch

Exchange between students: Forum on CMS (available to all students registered on CMS)

Reaching out to the instructors:

boenisch@cispa.de

dziedzic@cispa.de

Please include [TML25] in the subject line

Note: If you decide to discontinue the course, please de-register from CMS!

Thank you!

Franziska Boenisch and Adam Dziedzic boenisch@cispa.de, adam.dziedzic@cispa.de sprintml.com

Course on Trustworthy Machine Learning