# Security Engineering and Machine Learning (Redacted Version)

**Research Proposal** 



#### Introduction

- Recently, tremendous progress has been made in machine learning (ML) and its use for the Internet of Things (IoT)
- Attacks of ML systems are being developed and released with increased regularity
- Microsoft, Amazon, and Google, are some of the companies that had their ML systems tricked, evaded, or misled





## Motivation

# Gartner: Top 10 strategic technology trends in 2020



"Through 2022, 30% of all AI cyberattacks will leverage trainingdata poisoning, AI model theft, or adversarial samples to attack AIpowered systems" (Gartner, 2019)<sup>1</sup>

There is a broad and pressing call to advance a science of security in ML



<sup>&</sup>lt;sup>1</sup> https://www.gartner.com/en/doc/432920-top-10-strategic-technology-trends-for-2020

#### **Poisoning Attacks**

- Apply ML in IoT environments presents some unique security challenges
- Integrity is essential in ML, and is the centre of attention for most performance metrics, e.g., accuracy
- Poisoning attacks compromise a system's integrity





## **Attack Surface of a Machine Learning System**



<sup>3</sup> N. Papernot, P. McDaniel, A. Sinha and M. P. Wellman, "SoK: Security and Privacy in Machine Learning," *2018 IEEE European Symposium on Security and Privacy (EuroS&P)*, 2018, pp. 399-414, doi: 10.1109/EuroSP.2018.00035.



#### **Attack Surface Applied to a Smart Building Use-Case**





#### **Some Research Questions**

- How can an adversary tamper with or otherwise poison raw input data?
- How can attempts to poison training datasets be detected?
- How can a generalizable approach to detect poisoning attacks at different phases of the data processing pipeline be developed?

These questions can be studied in the context of an IoT application, e.g., smart buildings, and also from the lens of interactive learning

