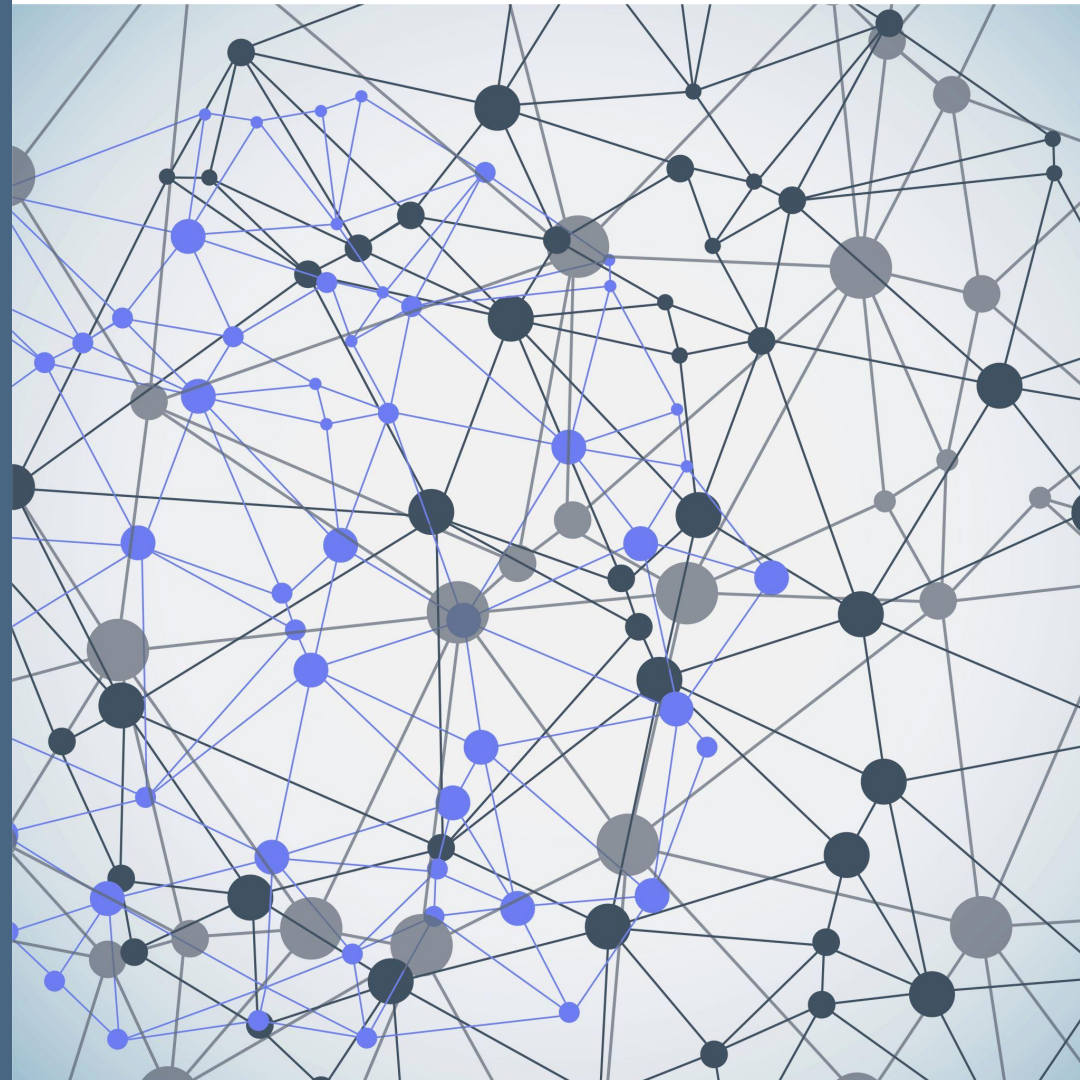


APPLICATION OF DEEP MACHINE LEARNING IN CYBERSECURITY

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INTRODUCTION

Cybercrime is one of the world's fastest growing threats to security.

- Unfortunately corresponds to our growing dependence on computer networks and information technology (banking software, autonomous vehicles, smart assistants - Siri, Alexa)
- An estimated \$6 trillion global cost by 2021 under the 2020 Official Annual Cybercrime Report by Cybersecurity Ventures

Silver Lining: New Deep Machine Learning Cybersecurity Tools...

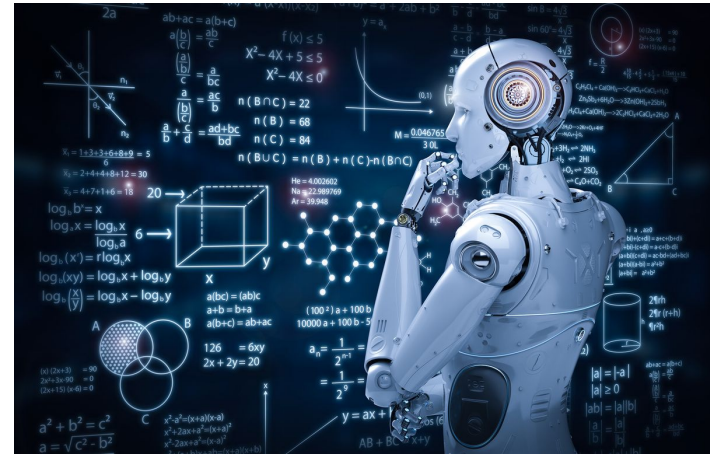


Image Source: [Security Info Watch](#)

WHAT IS CYBERSECURITY?

- Cybersecurity is the practice of maintaining confidentiality, integrity and the availability of the data.
- Cybersecurity comprises of set of tools and techniques to protect the data/ information from various attacks.
- Most common types of cyber threats are Malware, Ransomware, Phishing attacks and Social Engineering.
- Companies spend nearly \$3.92 millions on data breaches.

DATA BREACHES BY THE NUMBERS



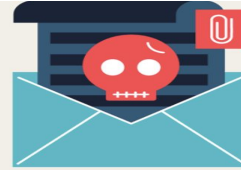
Hackers attack every **39 SECONDS**, on average **2,244** times a day.



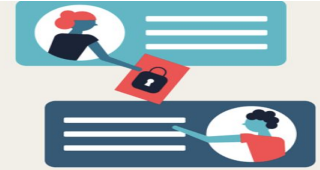
Data breaches exposed **4.1 BILLION** records in the first half of 2019.



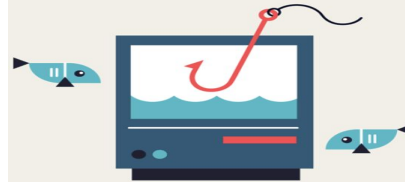
Average cybersecurity spending per employee is **\$1,178**.



48% of malicious email attachments are office files.



34% of data breaches involved internal actors.



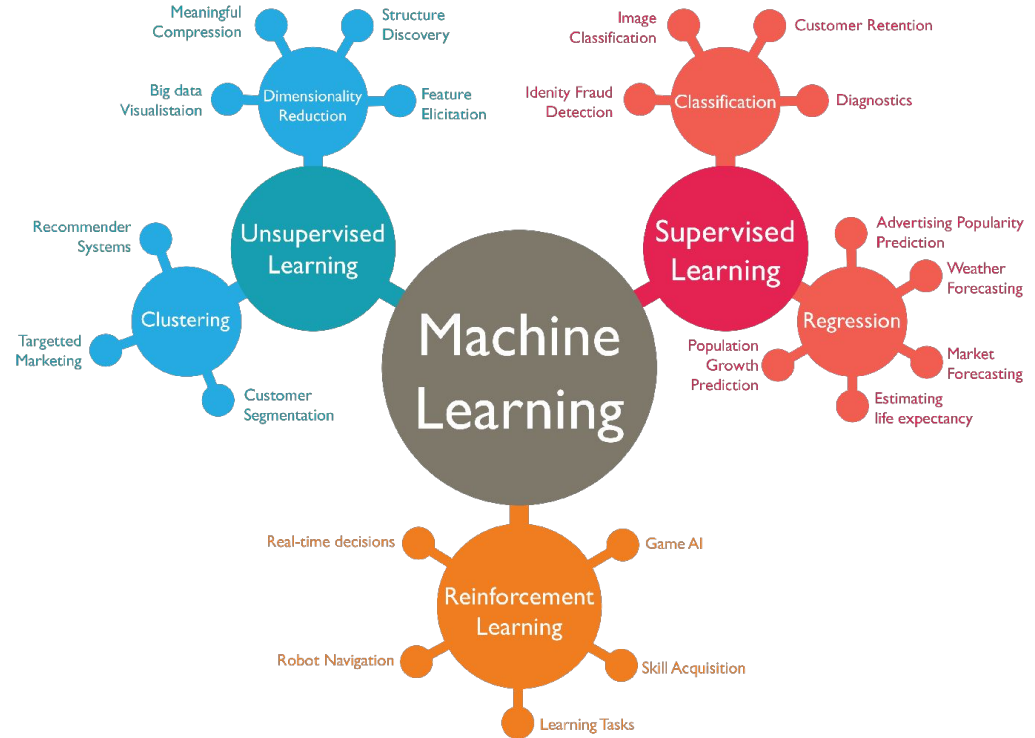
65% of groups used spear-phishing as the



94% of malware was delivered by email.

WHAT IS MACHINE LEARNING?

- It is an application of artificial intelligence and it provides the ability for the system to automatically without any human intervention and trains the system to take decision with any external programming.
- Machine Learning methods are of 3 types:
 - Supervised learning
 - Unsupervised learning
 - Reinforcement learning



WHAT IS DEEP LEARNING?

- Subset of Machine Learning
- Takeaway features:
 - Scalability
 - Feature Training
- *Deep* refers to the number of layers involved
 - *Deep* Learning versus *Shallow* Learning
- Modeling inspired by the brain -> Artificial Neural Networks

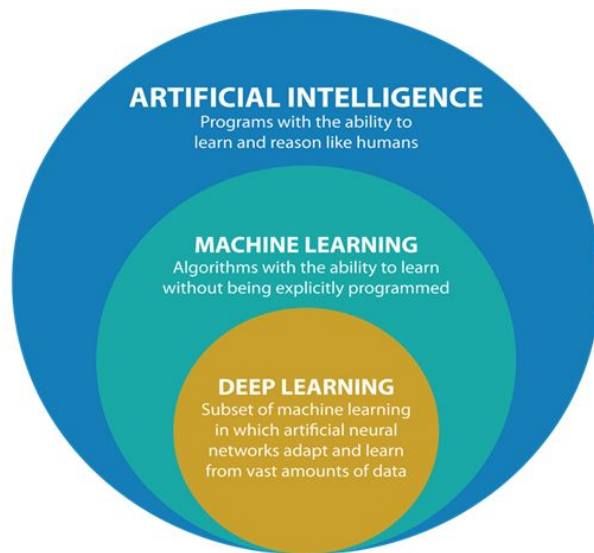


Image Source: [Data Catchup](#)

WHAT ARE ARTIFICIAL NEURAL NETWORKS?

- Type of A.I. technology
- Based on how the neurons in the brain works
 - Consists of nodes
 - Input and output with hidden layers in between
- Can generate nonlinear models
- Requires more time and power as the task becomes more complex

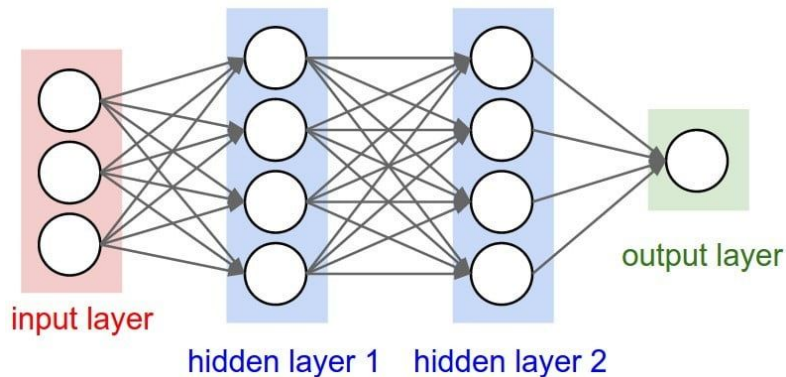


Image Source: [digitaltrends](https://www.digitaltrends.com)

CIA TRIAD



SECURITY GOAL #I: CONFIDENTIALITY

- Confidentiality means protection of data or resources from unauthorized access.
- The hackers usually perform two types of extractions attacks and they are:
 - Model Extraction Attack: extracts the model parameters by sending queries to the model
 - Model Inversion Attack: attacks are carried out by finding the inputs that provide sensitive information from the training datasets when given as input to the models



SECURITY GOAL #1.1: PRIVACY

- Norton defines *privacy* as the rights to information, regarding its access and usage.
- Includes protection of the DL model itself and the training data
- Consider a HIPAA violation...



Image Source: [Mercury News](#)

SECURITY GOAL #2: INTEGRITY

- Ensure that the data has not been tampered or compromised
- Ways to impact integrity
 - Modifying data
 - Unintentionally by using bad data
 - Removing data



Image Source: [cybersecurityglossary](https://www.cybersecurityglossary.com/)

SECURITY GOAL #3: AVAILABILITY

- *Availability* is the ability of authorized users to access the system, network, and data at will. Interruption of this freedom is an availability attack.
- Example: Denial-of-Service Attack
- False Positives/Misclassifications in ML models

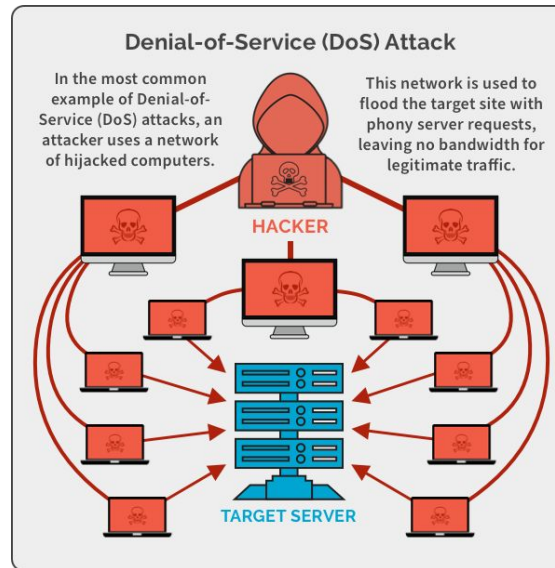


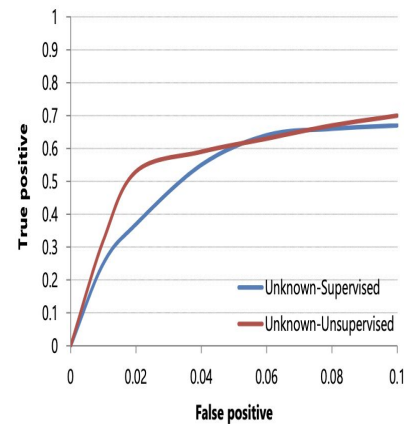
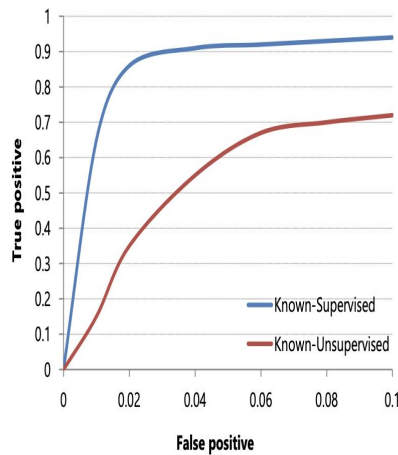
Image Source: [Security Boulevard](#)

DEEP LEARNING APPLICATIONS IN CYBERSECURITY

- Intrusion Detection/Prevention Systems
- Malware Detection
- Spam/Social Engineering Detection
- Network Traffic Analysis
- User Behavior Analysis

INTRUSION DETECTION/PREVENTION SYSTEMS

- Detect any unusual activity in the system.
- Traditional methods are unable to solve the complex problems
- High false positive and false negative detection rates.
- New techniques with artificial intelligence and computer intelligence were proposed.
- Laskov along with his team members developed a technique that gives 95% accurate results.



MALWARE DETECTION

- “Any type of malicious software designed to harm or exploit any programmable device, service or network.”
- Static Analysis versus Dynamic Analysis
- Autoencoders in malware detection applications
 - Li Framework
 - DL4MD Model

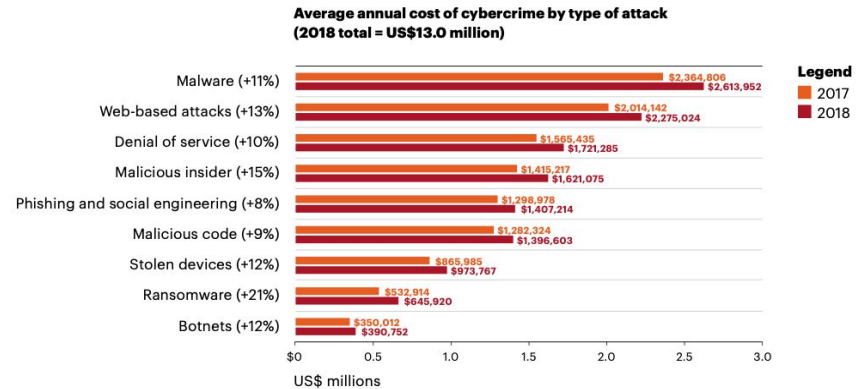
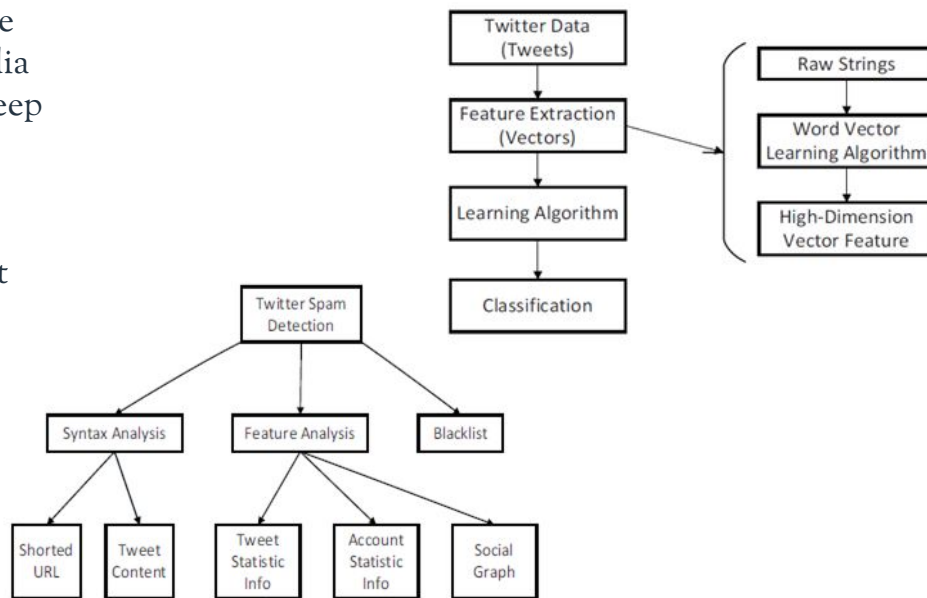


Figure 7: Average annual cost of cybercrime by type of attack, 2018-2019.
From *Accenture Ninth Annual Cost of Cybercrime Study, 2019*.

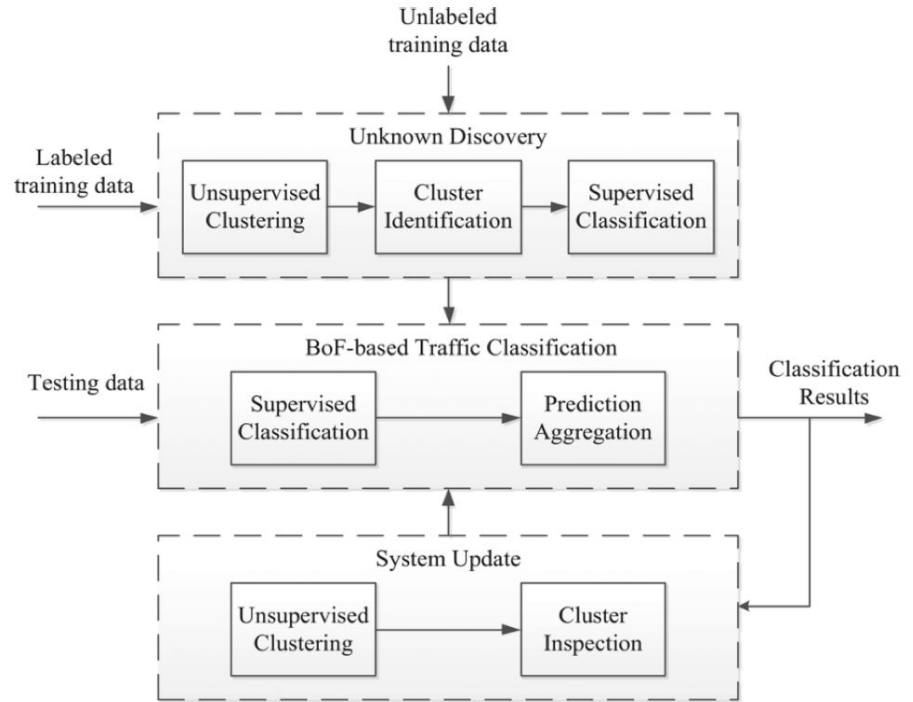
SPAM/SOCIAL ENGINEERING DETECTION

- Detecting spam has become more difficult, especially on social media
- Researchers are starting to use deep learning to detect spam
- Ex: Twitter
- Old method:
 - Characteristic of the tweet
 - Blacklist
 - Time consuming
 - Most would have already visited the site
- New Method:
 - Vectors
 - Deep learning

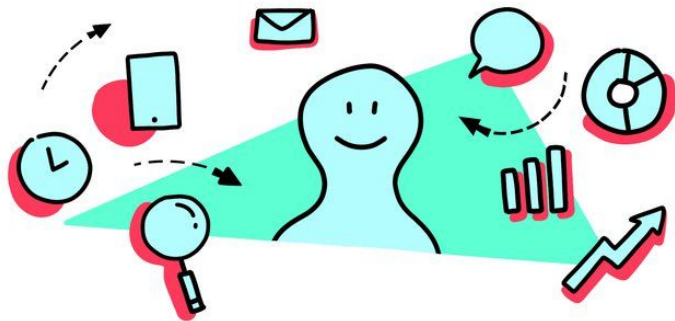


NETWORK TRAFFIC ANALYSIS

- Monitor traffic activity patterns.
- IP address from unusual area is notified.
- Malware behavior is now difficult to detect.
- Especially zero-day attacks are most occurring and have to be prevented.
- A technique to detect the malicious and abnormal behavior was proposed using 3 model framework known as Robust Network Traffic Classification.



USER BEHAVIOR ANALYSIS



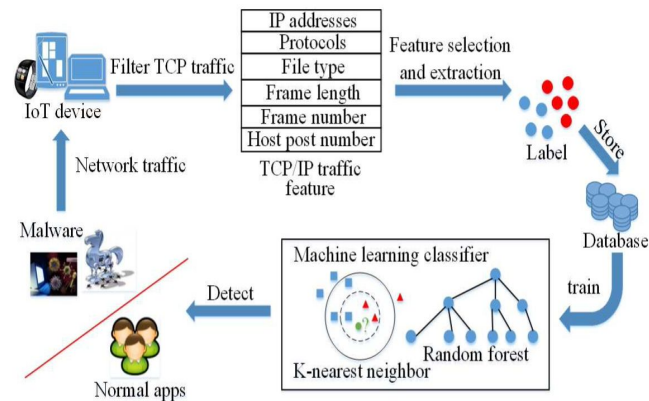
- “Searches for patterns of usage that indicate unusual or anomalous behavior”
- Capable of detecting:
 - Insider threats
 - Targeted attacks
 - Financial fraud
- Recurrent Neural Networks:
 - Tuor model for anomaly detection
 - Fake news detection

USE CASES

- Internet of Things
- Ring Intrusion Detection Devices
- Android Malware Detection

INTERNET OF THINGS (IOT)

- Billions of devices are connected using Internet of Things.
- IoT components are vast and hence the attack surface area is huge and the devices are easily vulnerable.
- Mirai Attack took down thousand of IoT devices.
- Narudin et al. proposed method to detect abnormal behavior in IoT devices.
 - K-NN model
 - Random Forest Model
 - Filtering of network traffic components.



RING INTRUSION DETECTION DEVICES

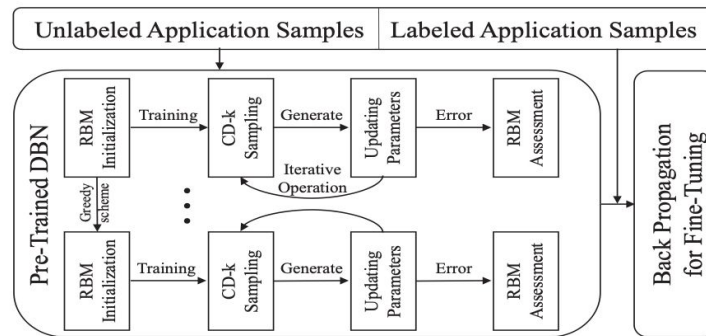
- There's little research in deep learning for smart devices
- Cameras, smart cars, or voice controlled systems have a vulnerability in IDS
- Ex: Ring
 - There are some cases in which people had their Ring devices hacked.
 - Weak IDS
- To improve IDS in smart devices, adversary attacks are developed
 - To study it
 - To learn to defend against it



Image Source: [ring](https://www.ring.com)

ANDROID MALWARE DETECTION

- Malware is one of the most common and dangerous threats to a computer system!
- Android, one of the best-selling operating systems worldwide, is even susceptible.
- Yuan et al. DL model for Android Malware Detection
 - Beyond the risk communication technique often adopted
 - Static and Dynamic analysis of 202 features
 - Max. Accuracy of 96.5%
 - DL model outperformed the following ML models: SVM, Naive Bayes, C4.5, Logistic Regression, and Multi-Layer Perceptron



Droid-Sec Framework Model. From *Droid-Sec: Deep Learning in Android Malware Detection*, 2014.

THANK YOU!
QUESTIONS?