CSIT
CENTRE FOR SECURE INFORMATION TECHNOLOGIES

SECURING OUR DIGITAL TOMORROW

MALWARE AND MACHINE LEARNING

QUEEN'S UNIVERSITY BELFAST

Dr. BooJoong Kang
Research Fellow , CSIT

23 Sept. 2020

Who am I?

- BooJoong Kang (강부중)
- Research Fellow
- Centre of Secure Information Technologies
- Queen's University Belfast, UK
- BS, MS, PhD from Hanyang University

- Cybersecurity
  - Network Security and Malware Analysis

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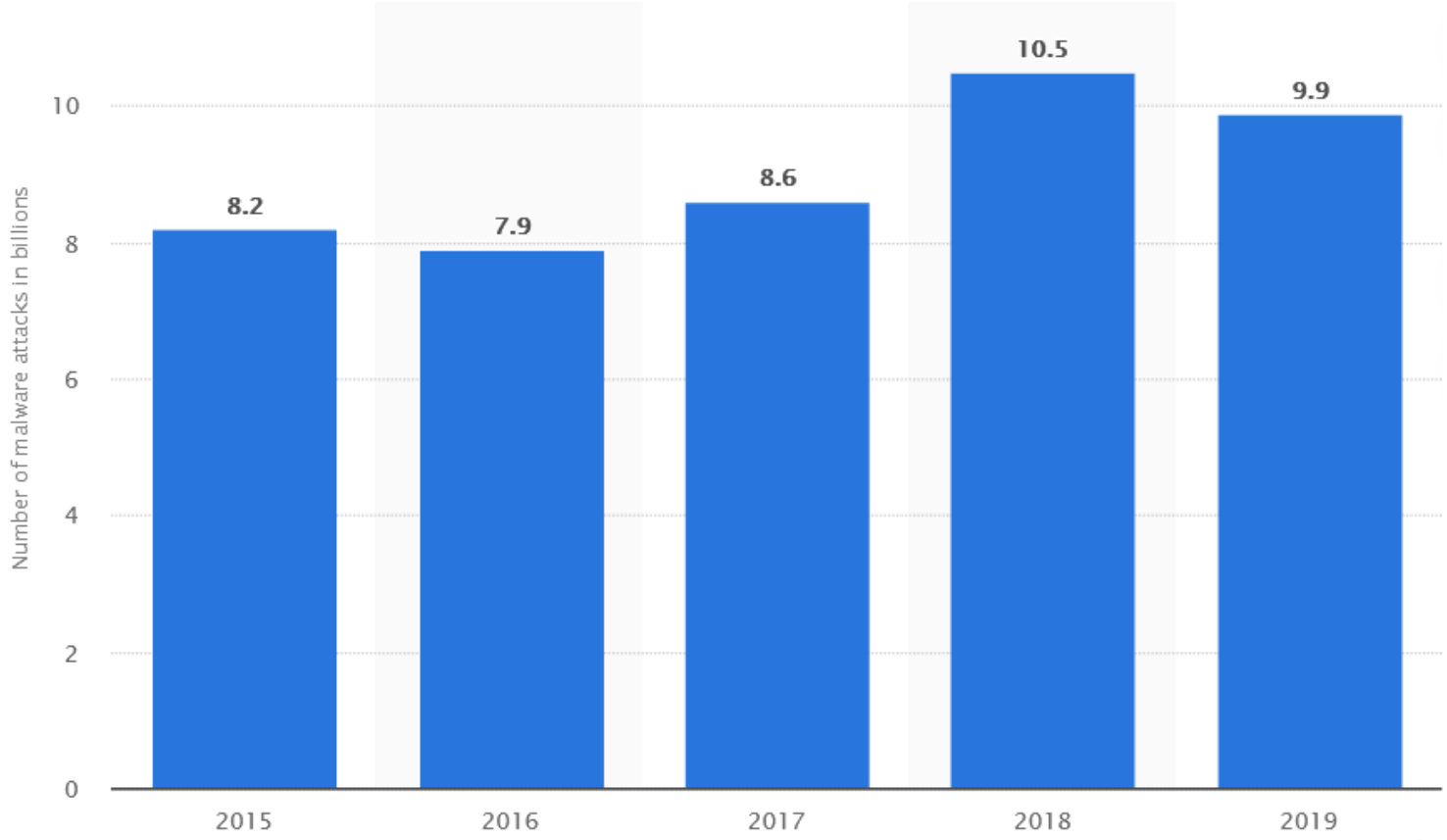
Outline

- Malware
- Malware Analysis
- Malware Detection
- ML-driven Malware Detection
- Adversarial Examples

Malware

- Malware (a portmanteau for malicious software) is any software intentionally designed to cause damage to a computer, server, client, or computer network



The first of many: Creeper

- Bob Thomas at BBN Technologies in 1971
- As a demonstration of mobile applications
  - Software that could automatically hop between computers on a network
- Find a computer, hop over to it and display “I’M THE CREEPER : CATCH ME IF YOU CAN.”

- Later, Thomas’s colleague Ray Tomlinson
- An update to the Creeper
  - Leave a copy of itself before moving onto the next one (self-replication)
- Invented another virus called the Reaper
  - Find infected computers and delete the Creeper

- Creeper: Worm
- Updated Creeper: Virus
- Reaper: Anti-Virus

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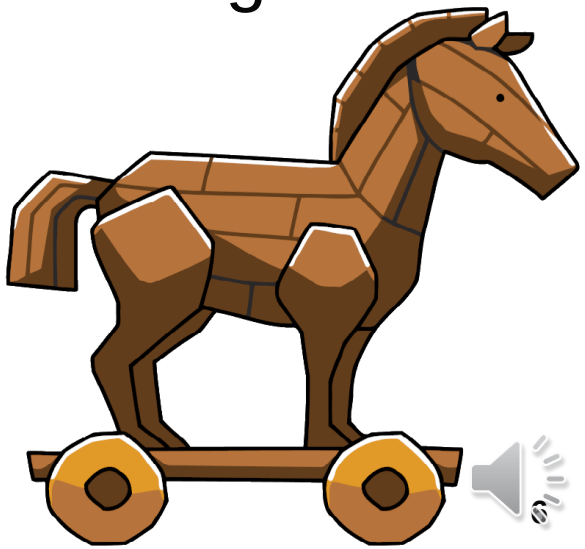
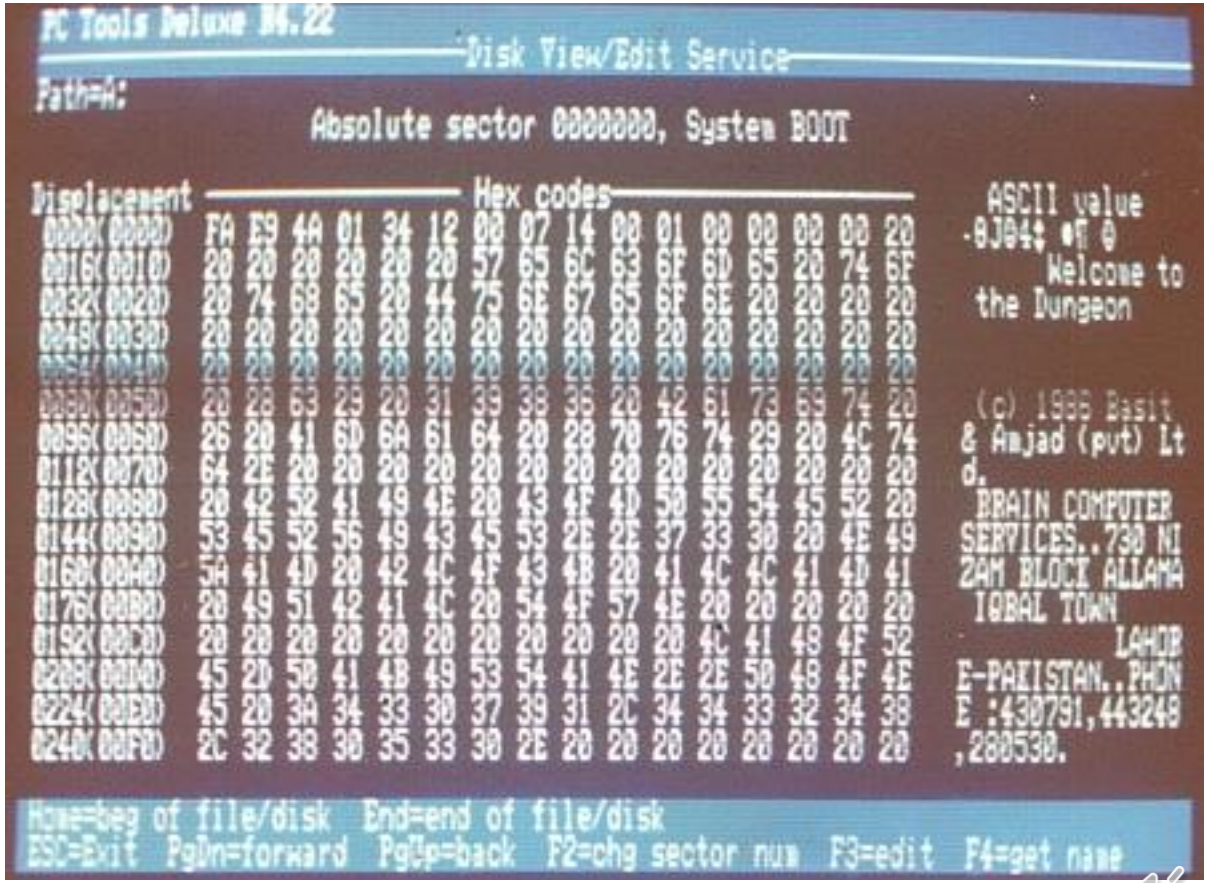
The first Trojan: ANIMAL

- John Walker, 1975
- "20 questions" program that guess the animal
- Become popular and start to occupy a lot of his time to copy it
- Subroutine, called PERVADE, saves copies of ANIMAL to any user-accessible directories that would, while the user was answering the questions
- Many of these directories were shared between offices
  - so the Trojan spread via that vector as well



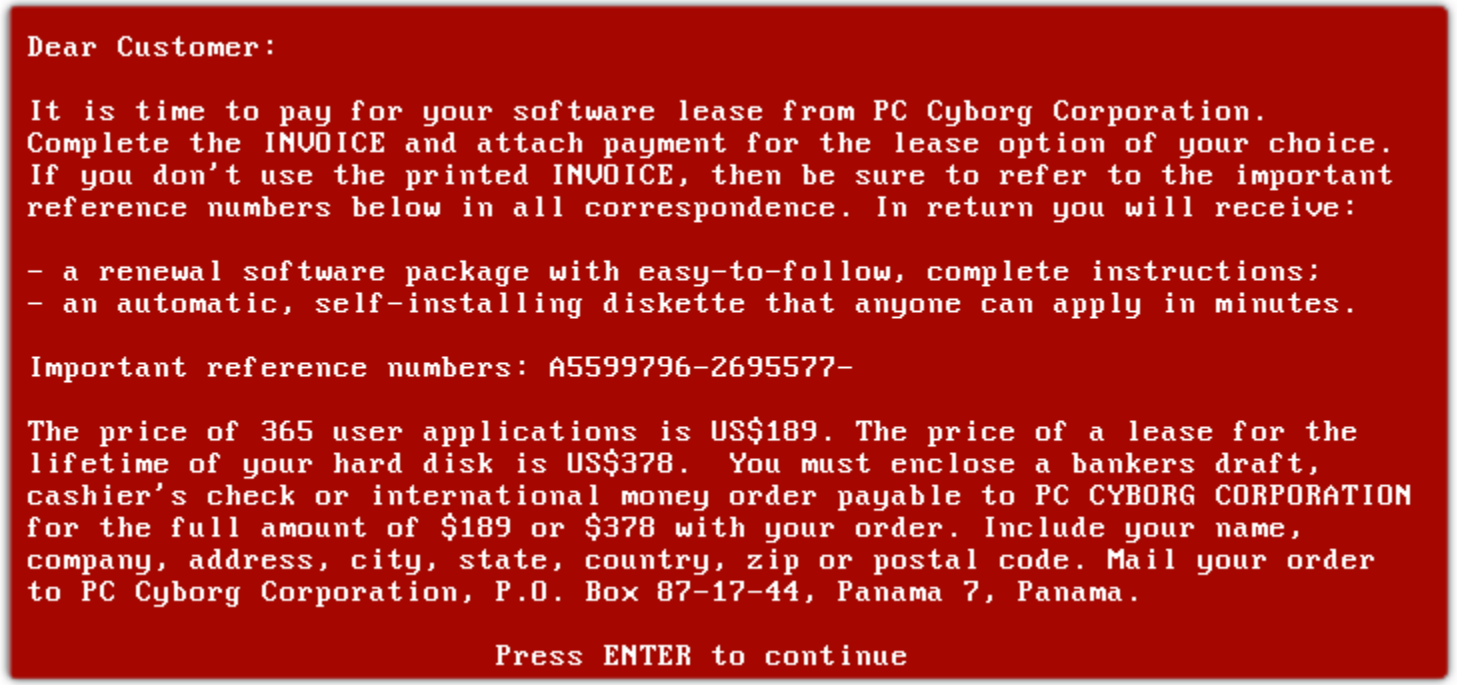
The first PC virus: Brain

- Basit Farooq Alvi and Amjad Farooq Alvi, two brothers, in 1986
- Infect IBM PC, slow down floppy disks and take up a good chunk of memory
- Designed to protect their medical software from piracy
- A message in the code that included their address and phone numbers



The first ransomware: AIDS Trojan

- Joseph Popp in 1989
- Once installed, it counts all the times you booted up the computer
- Once you'd restarted 90 times
  - hide all files, rendering them inaccessible
  - demand you send a letter to an address with $189 to "renew your license."
- He was eventually caught but let off the hook after being declared mentally unfit and agreeing to donate the profits to AIDS research



The first social engineering attack: Melissa

- In 1999, the first email-based viruses and the first to ever use social engineering
- Spread by sending emails to email contacts
  - The headline: "Important Message From ABC“
  - The body text: "Here is that document you asked for ... don't show anyone else ;-)“
- Infected DOC file includes a list of 80 pornographic websites as well as usernames and passwords
- Also other DOC files found on the computer, which have also been infected
  - classified or private files would be shared with friends, family, or work associates

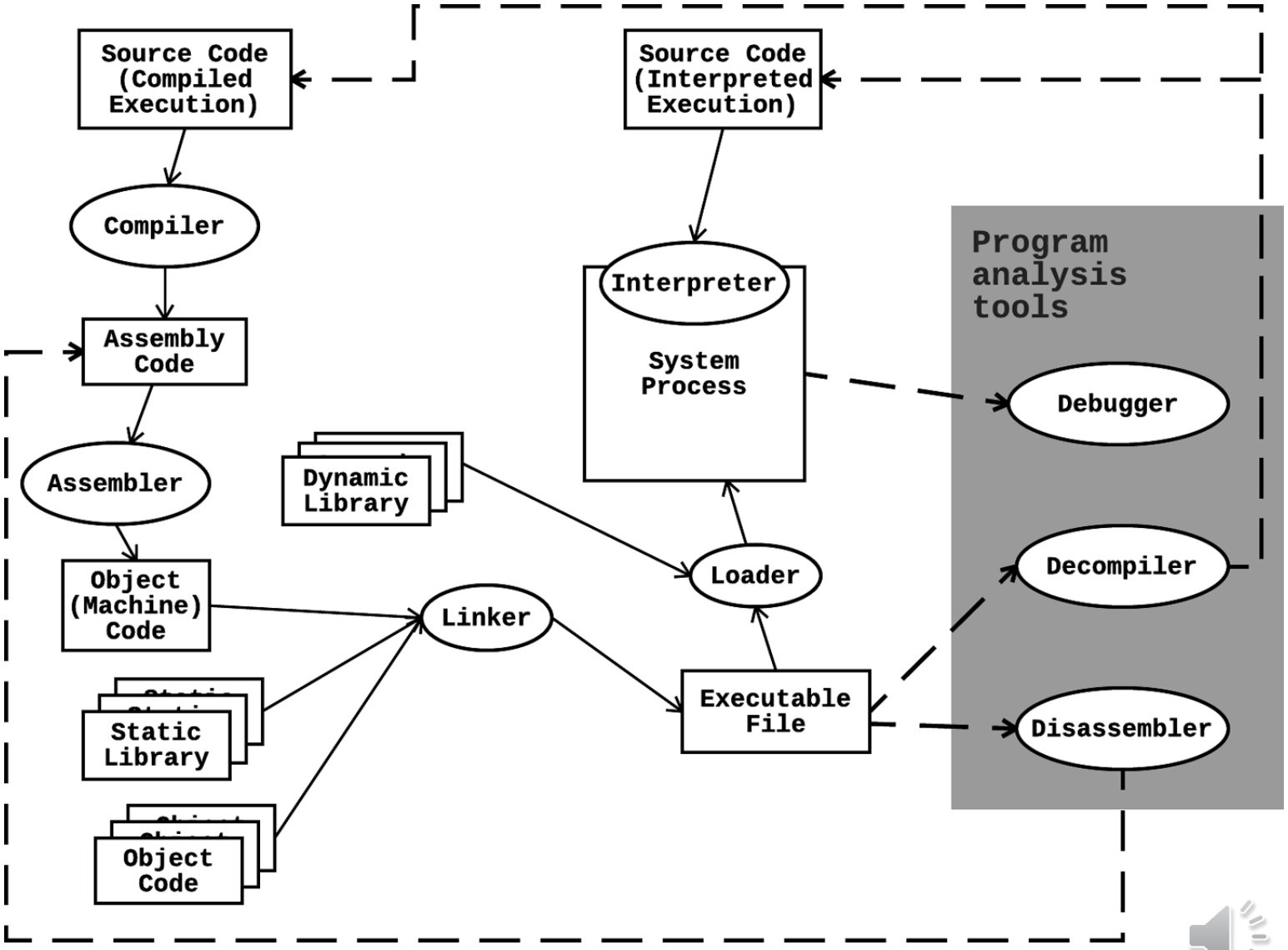
- ILOVEYOU (Love Letter Worm) in 2000
- An email disguised as a love letter
- The attached text file overwrite files, steal user data such, then send itself to everyone on your email contact list
- Compromised an estimated 45 million computers around the world (about 10% of all connected computers) and caused over 8 billion dollars in damages

Malware in the 21st Century

- Between 2000 and 2010, malware grew significantly, both in number and in how fast infections spread
  - A dramatic increase in malware toolkits, including the now infamous Sony rootkit and Crimeware kits
  - SQL injection attacks became a leading threat, claiming popular victims
    - SQL Slammer (2003) infected nearly 75,000 computers in ten minutes
  - Conficker Worm (2008) caused some of the worst damage seen since Slammer appeared in 2003
- Between 2010 and the present time, significant evolution in the sophistication of malware
  - Organized crime and state sponsors upped the game dramatically with large, well-funded development teams
  - Developing advanced malware with evasion tactics that outsmart conventional anti-malware systems
  - Infiltrating factories and military systems became a common reality, and the monetization of malware grew rapidly with dramatic growth in ransomware and other illegal schemes
  - Stuxnet Worm (2010) was designed with the express purpose of attacking Iran's nuclear program and included the ability to impact hardware as well as software
    - one of the most resource-intensive bits of malware created to date
  - Zeus Trojan (2011) is one of the most successful pieces of botnet software in the world, impacting millions of machines
  - WannaCry Ransomware (2017) locked people out of their data and demanded they pay a ransom or lose everything
    - Affected at least 150 countries, including hospitals, banks, telecommunications companies, warehouses, and many other industries

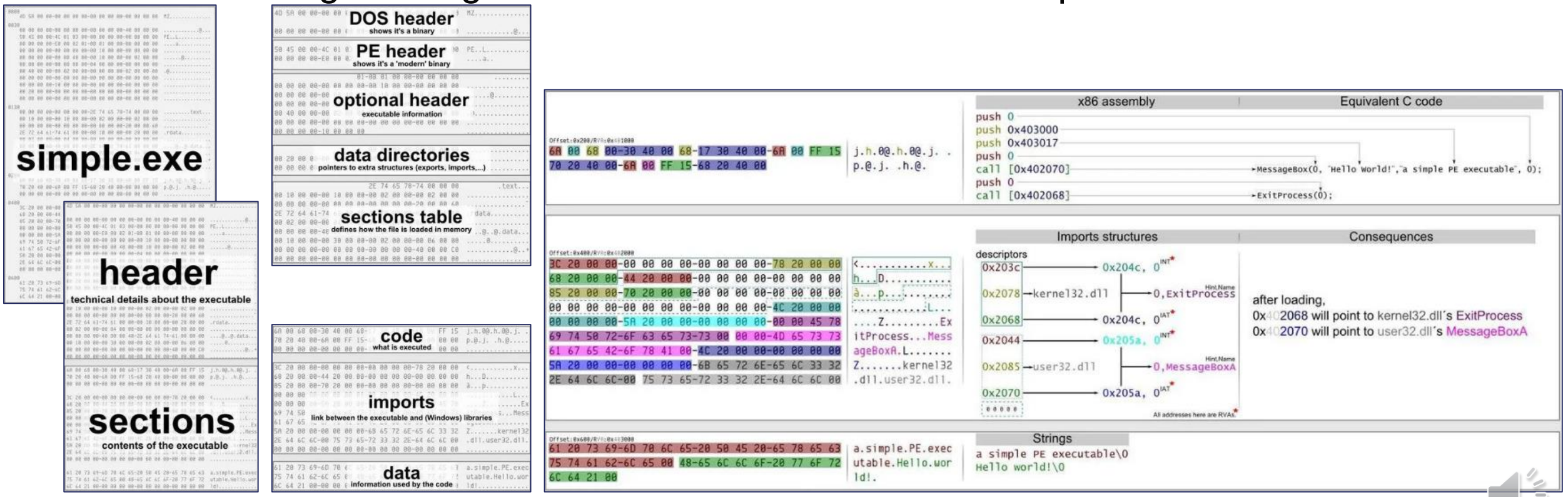
Malware Analysis

- To understand
  - How it works
  - How to identify it
  - How to defeat or eliminate it



Static Analysis

- Static Analysis examines the file itself
  - Reverse-engineering with a disassembler or a decompiler



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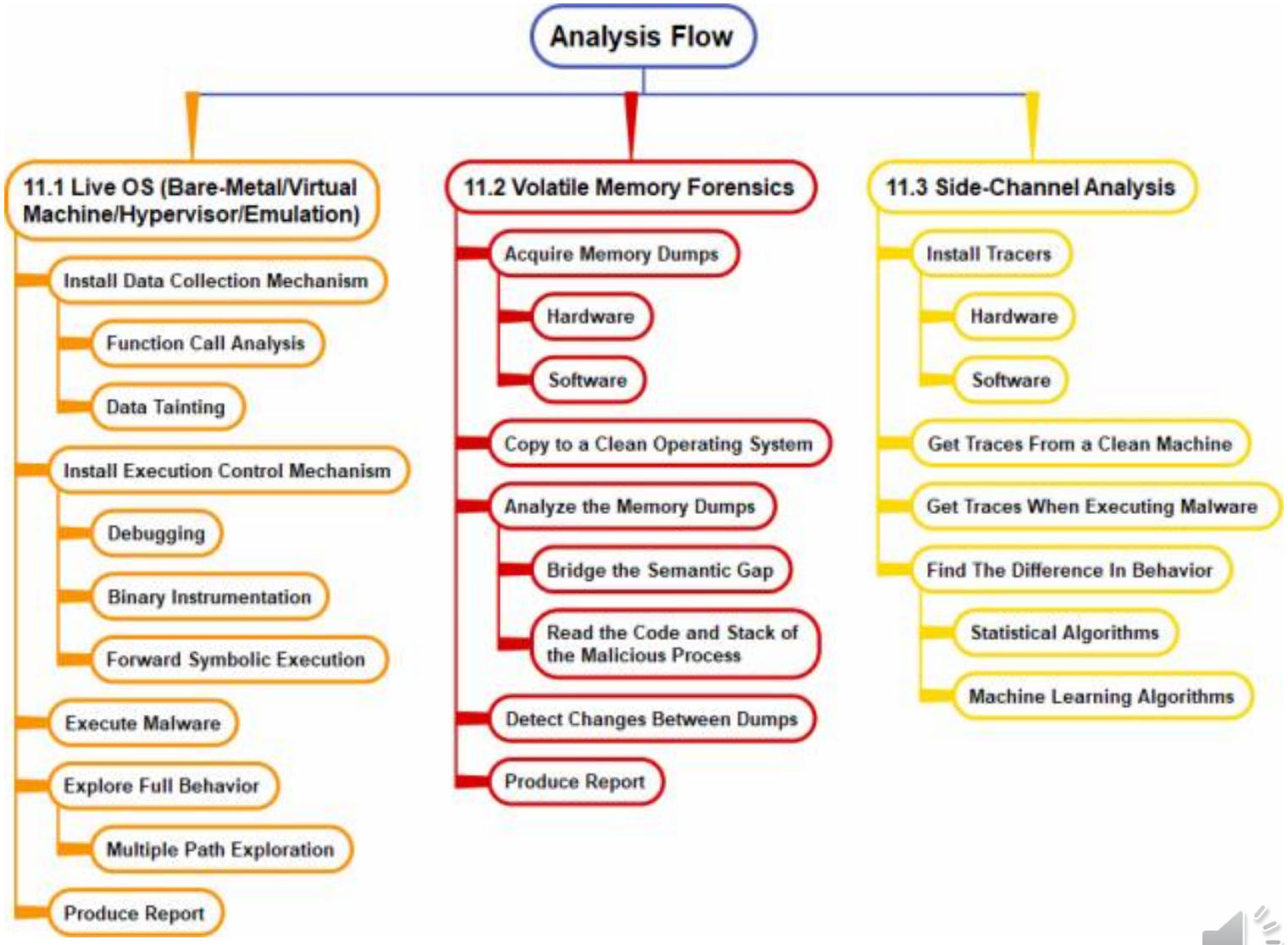
Features from Static Analysis

- Strings
- Byte-patterns
- Code
  - Functions
    - Function Call Graph
  - Instructions
    - Opcode
    - Control Flow Graph
  - APIs
- Combinations of Features can characterise programs

Dynamic Analysis

- Examine the runtime behaviour of programs
  - the execution and effects of programs with VMs or Debuggers



Features from Dynamic Analysis

- Executed Code
  - Function Calls
    - Called FCG
  - Executed Instructions
    - Executed CFG
  - API Calls
- Memory/Register Usages
- Network Traffic
- Any other behaviours

Malware Detection

- Identify Unique Properties of Malware from Static/Dynamic Analysis
- Confusion Matrix in Malware Detection
  - Low False Positive is more important in Malware Detection

	Malware	Benign
Detected as Malware	True Positive	False Positive
Detected as Benign	False Negative	True Negative

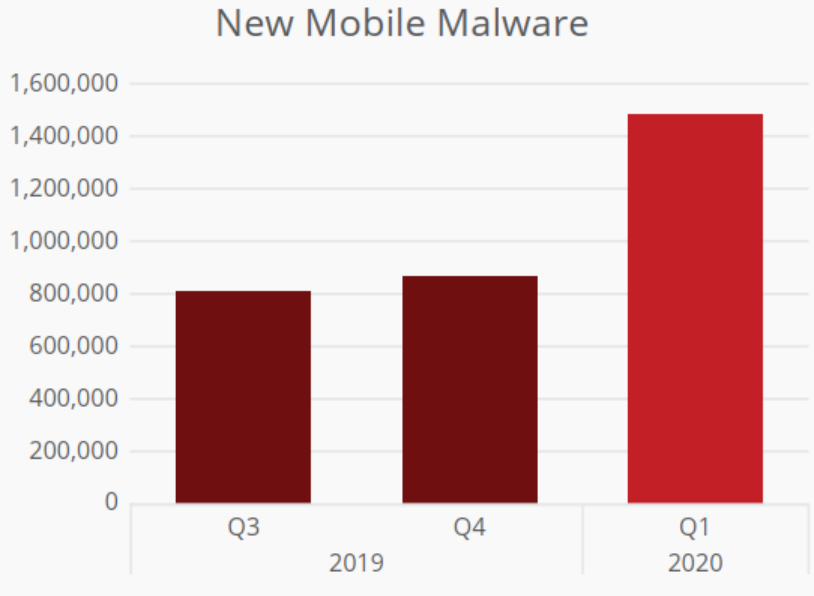
Malware Signatures

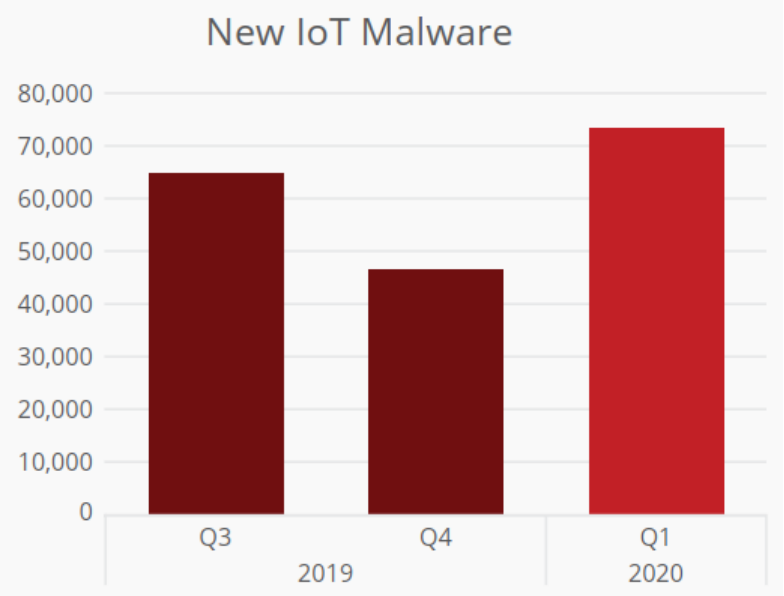
- Unique Properties of Malware
  - Hash
  - Strings
  - Byte-patterns
- Fast & Low False Positives
- Easy to Evade
  - a padding yields a different hash
  - variants, zero-day malware
    - continuous signature updates

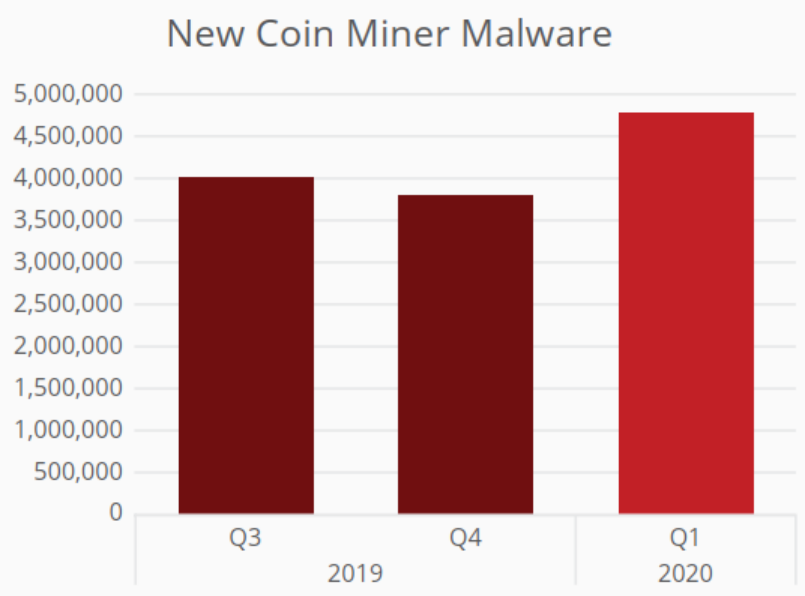
```
rule Enfal_Malware_Backdoor {
meta:
description = "Generic Rule to detect the Enfal Malware"
author = "Florian Roth"
date = "2015/02/10"
super_rule = 1
hash0 = "6d484daba3927fc0744b1bbd7981a56ebef95790"
hash1 = "d4071272cc1bf944e3867db299b3f5dce126f82b"
hash2 = "6c7c8b804cc76e2c208c6e3b6453cb134d01fa41"
strings:
$mz = { 4d 5a }
$s1 = "Micorsoft Corportation" fullword wide
$s2 = "IM Monnitor Service" fullword wide
$x1 = "imemonsvc.dll" fullword wide
$x2 = "iphlpsvc.tmp" fullword
$x3 = "{53A4988C-F91F-4054-9076-220AC5EC03F3}" fullword
$z1 = "urlmon" fullword
$z2 = "Registered trademarks and service marks are the property of their" wide
$z3 = "XpsUnregisterServer" fullword
$z4 = "XpsRegisterServer" fullword
condition:
( $mz at 0 ) and
(
( 1 of ($s*) ) or
( 2 of ($x*) and all of ($z*) )
)
and filesize &lt; 40000
}
```

Why ML?

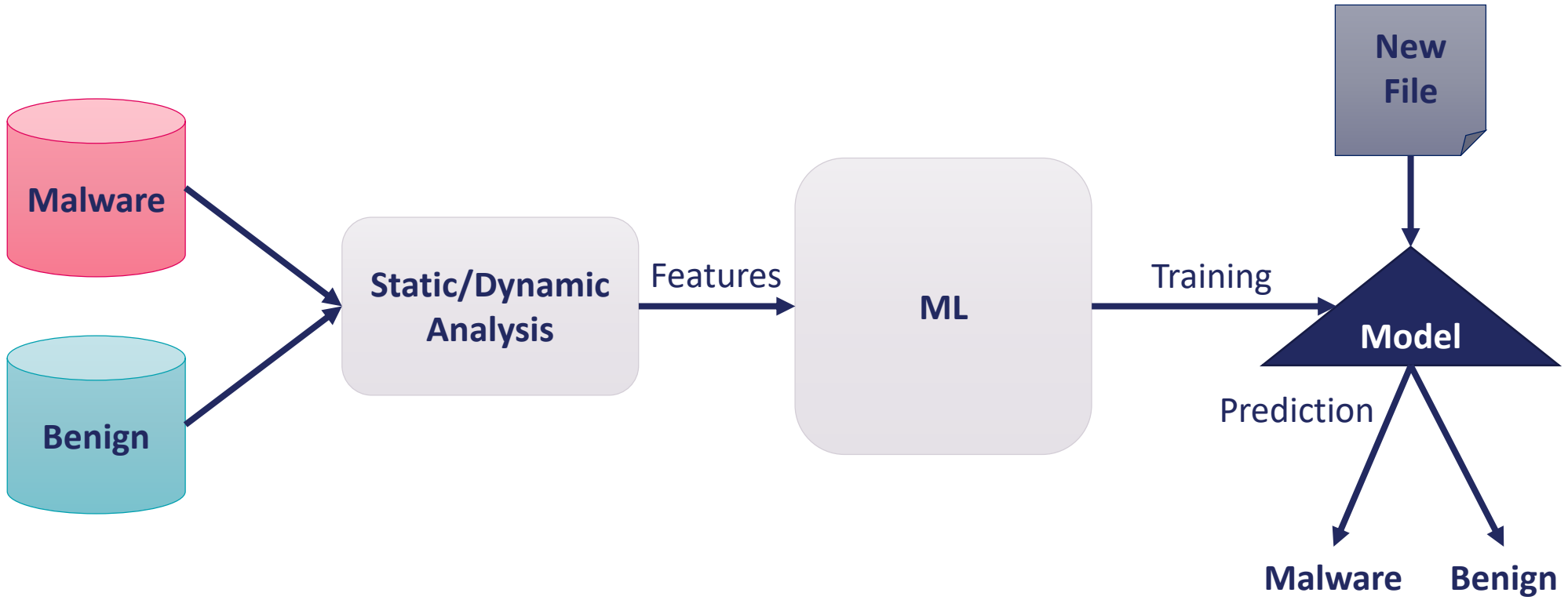
- Manual analysis for signatures can’t cope with the number of new malware





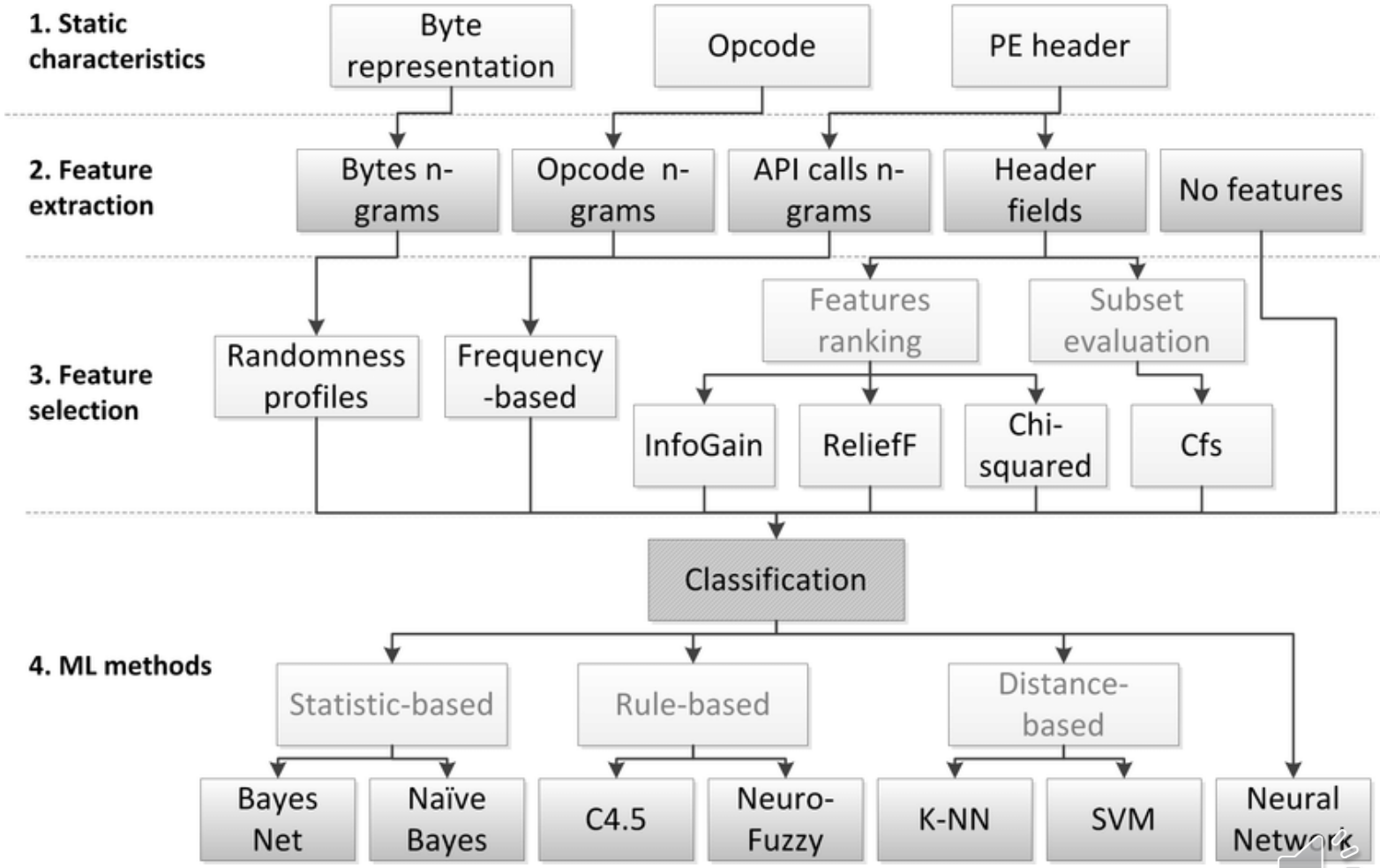


ML-driven Malware Detection



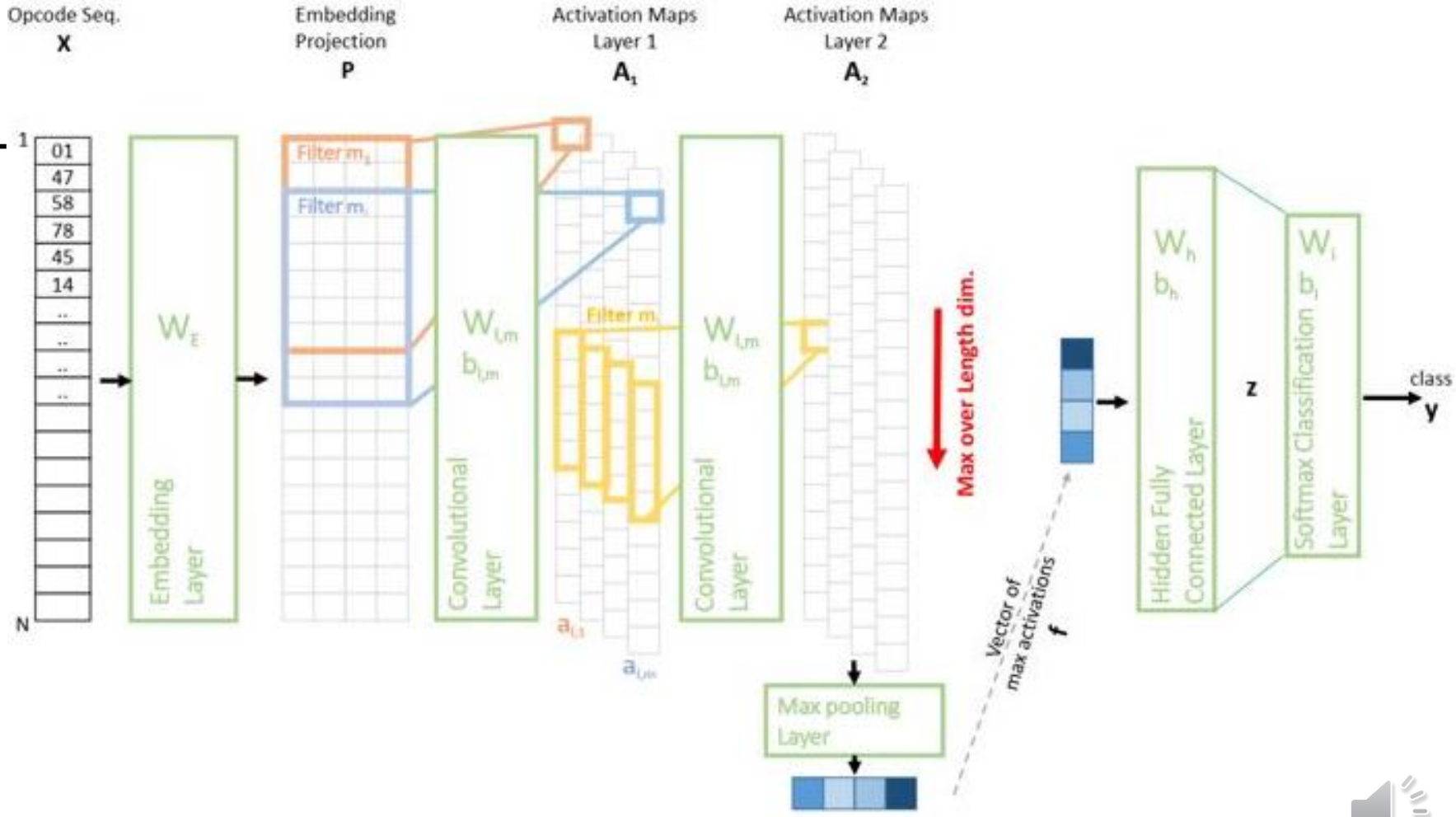
ML with Sequence Data

- N-grams
  - unigram of “COLD”
    - C, O, L, D
  - bigram of “COLD”
    - CO, OL, LD
  - n-gram (n=4) of “COLD”
    - COLD
- Word2vec
  - n-grams into a numerical vector
  - existence or frequency



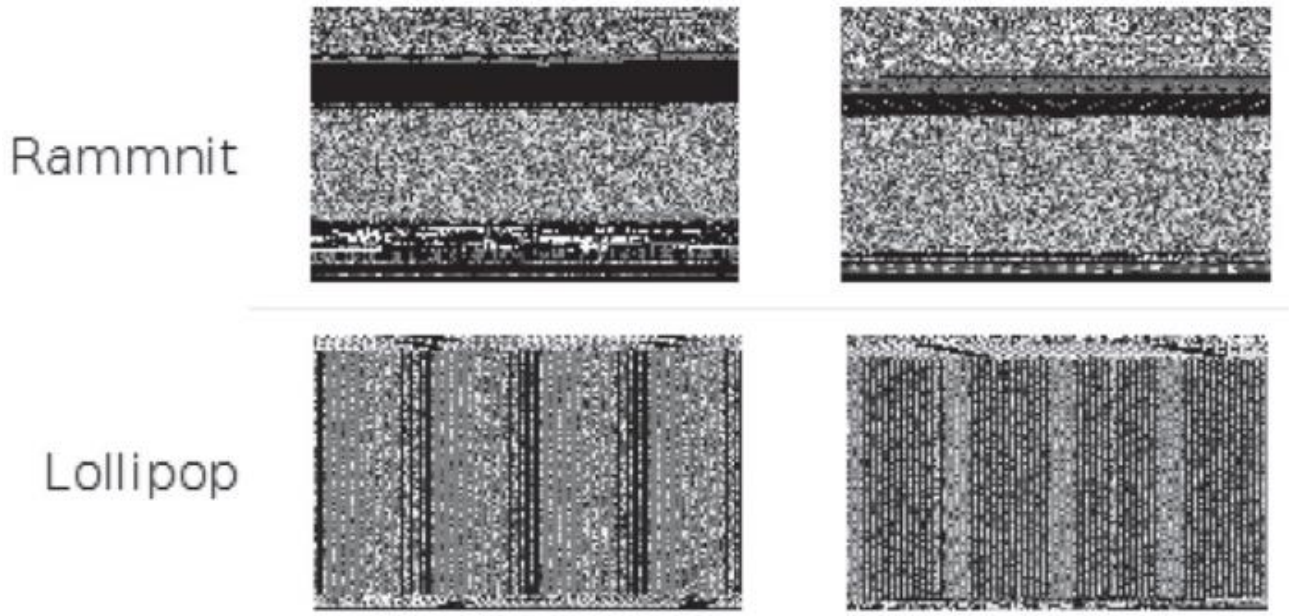
Deep Learning in Malware Detection

- Feature Extraction and Selection by DL
- No need to enumerate n-grams of Opcode sequences



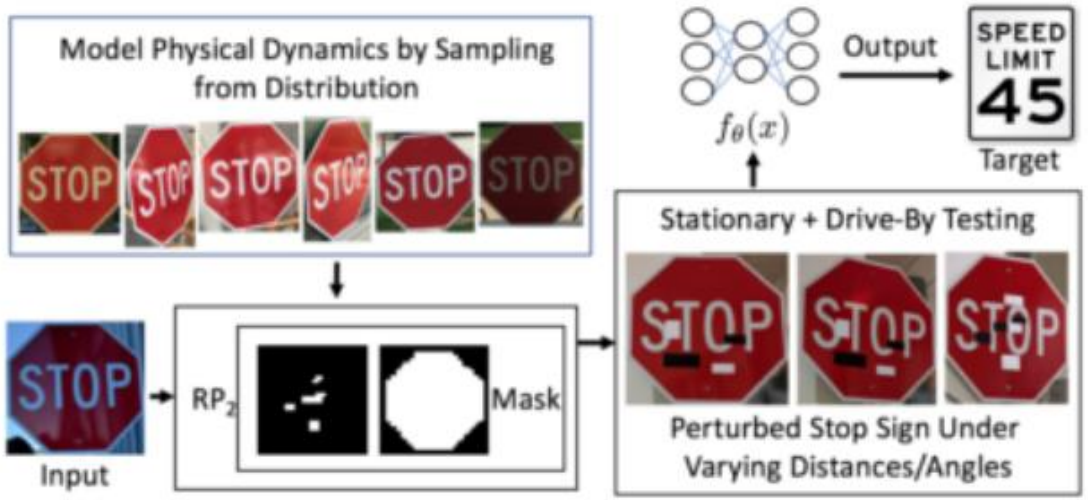
Malware Visualisation

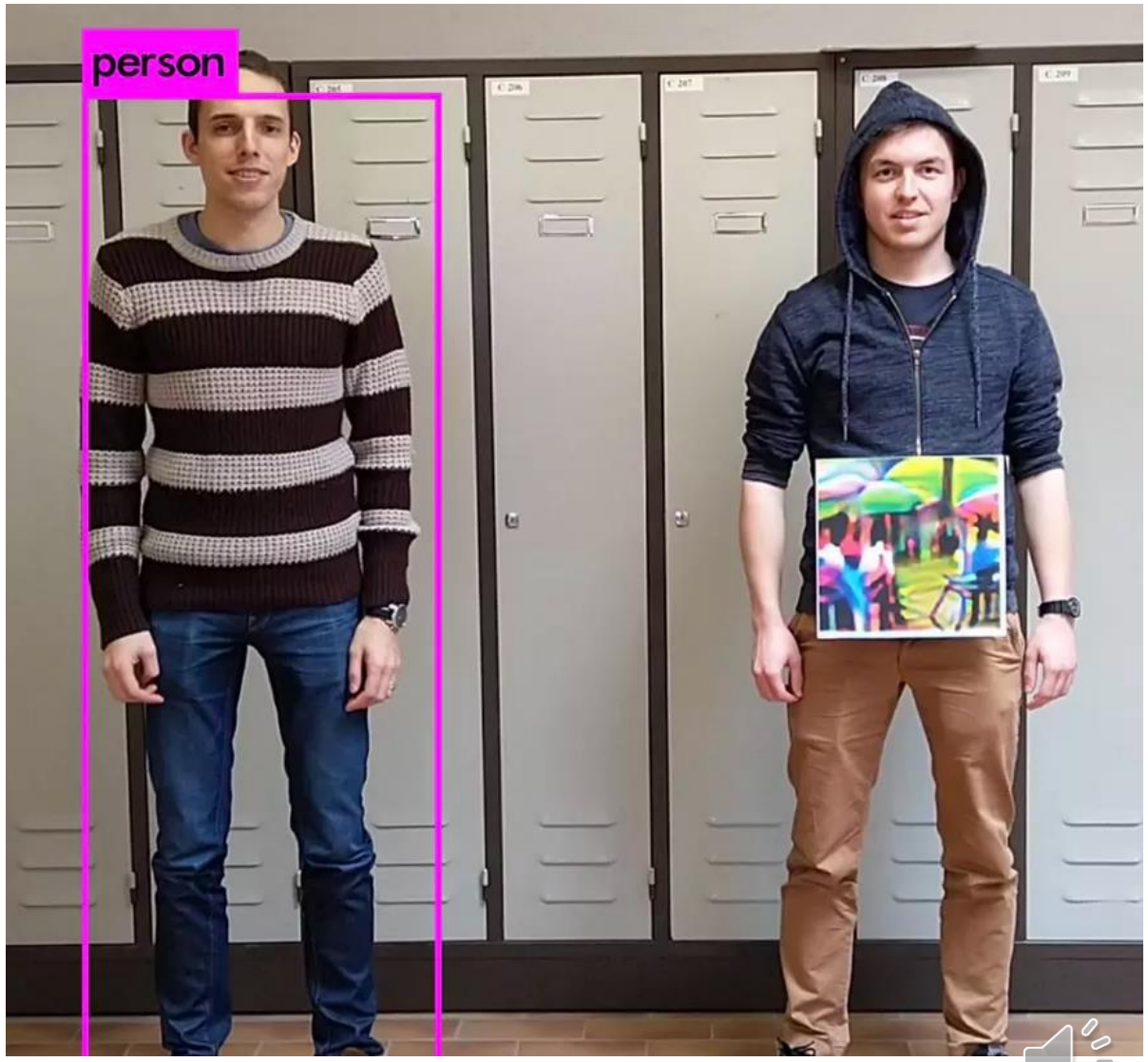
- Malware's binary content as a gray-scale image
  - every byte as one pixel in an image
  - reused samples
- Some Drawbacks
  - need to select an image width
  - non-existing spatial correlations between pixels in different rows
  - suffer from code obfuscation techniques
    - encryption/compression will change the bytes



Adversarial Examples

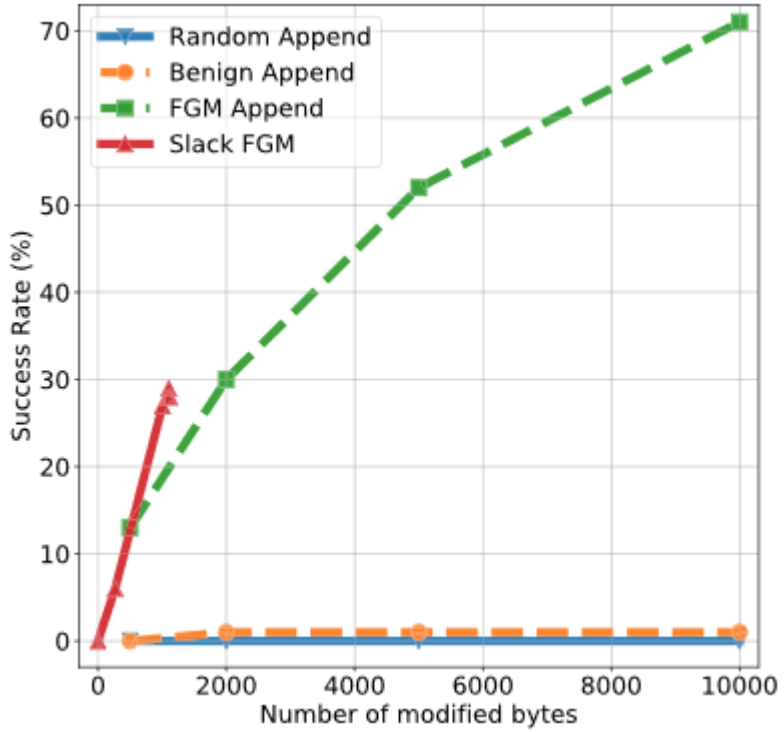
- Attempt to fool models





Adversarial Examples in Malware Detection

- Append Random, Benign, FGM bytes
  - might exceed the model's maximum size
- Slack FGM
  - unused space between sections
    - misalignments between the virtual addresses and the multipliers over the block sizes on disk



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Summary

- Malware
- Static/Dynamic Analysis
- Malware Detection
- ML-driven Malware Detection
- Adversarial Examples

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B.KANG@QUB.AC.UK