Multi-level Observation and Understanding of Program Behaviors

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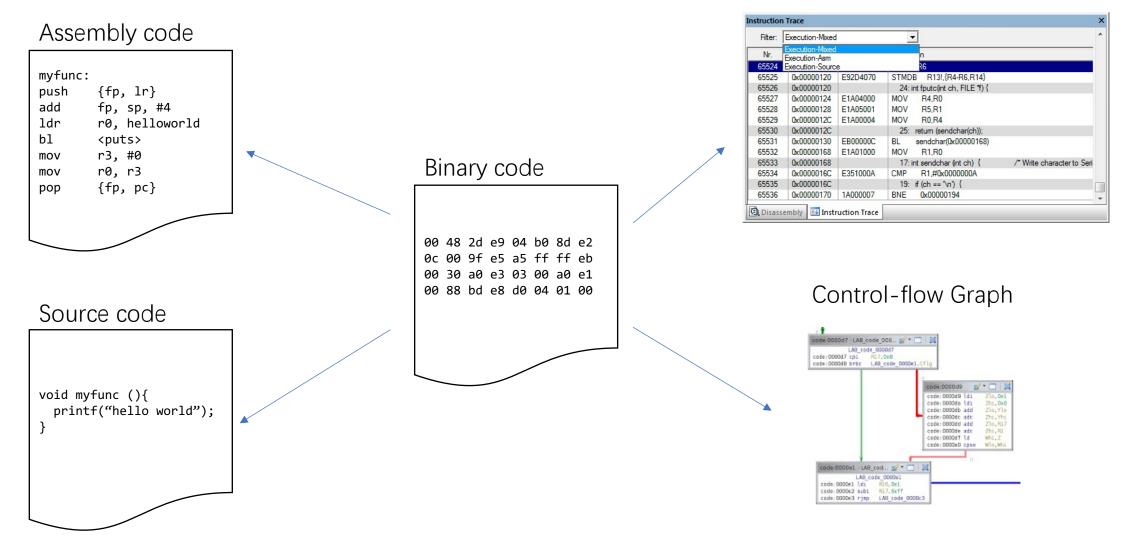


Security Incidents Are on The Rise



Binary-Level View

Instruction Trace



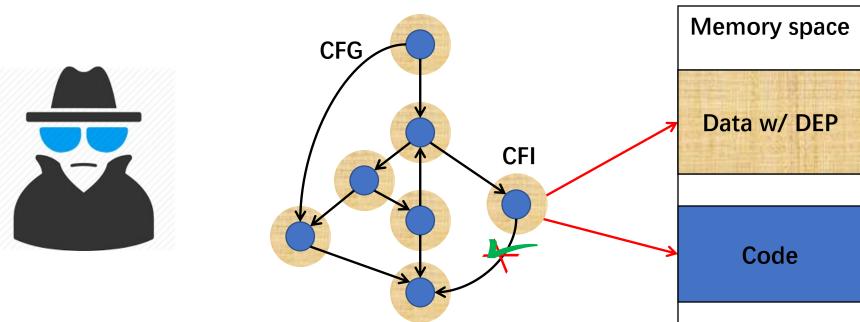
Binary-Leve Vulnerability, Attacks and Defenses

- Code injection
- Data Execution Prevention

• Code reuse



- return-to-libc
 return-oriented programming (ROP)
- Data-oriented Programming (DOP)





Audit-Log-Level View

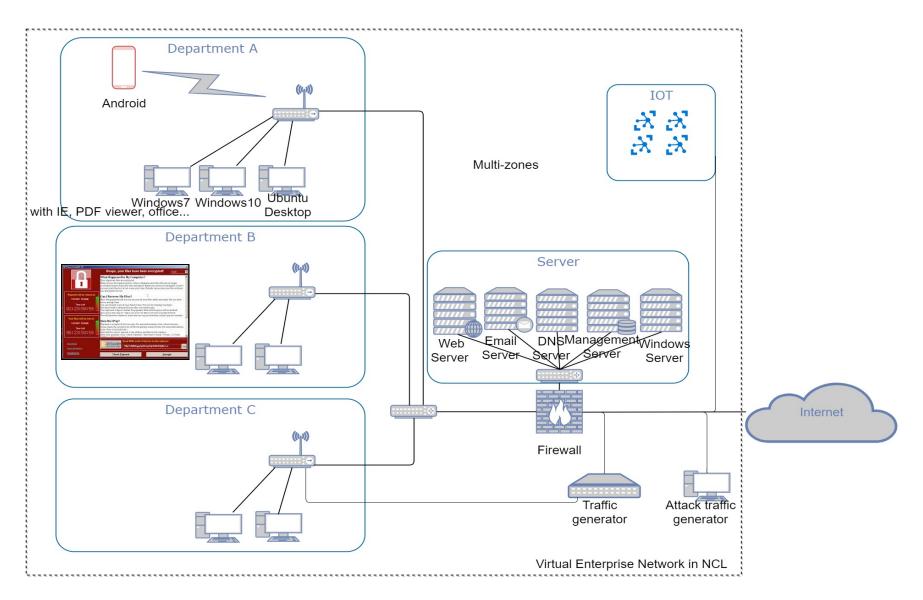
- User-space utilities (e.g., Auditd) collect system call records from kernel space through Netlink and write them to a log file under /var/log/audit
 - An Example of a read log entry in Auditd

type=PROCTITLE msg=audit(15/08/2019 14:37:30.522:61916019) : proctitle=sshd: junzeng [priv]
type=SYSCALL msg=audit(15/08/2019 14:37:30.522:61916019) : arch=x86_64 syscall=read
success=yes exit=52 a0=0x3 a1=0x7ffd69eecad0 a2=0x4000 a3=0x7ffd69ef0a60 items=0 ppid=5512
pid=5542 auid=junzeng uid=junzeng gid=junzeng euid=junzeng suid=junzeng fsuid=junzeng
egid=junzeng sgid=junzeng fsgid=junzeng ses=1805 comm=sshd exe=/usr/sbin/sshd key=(null)
_----

• An Example of a read log entry in Auditbeat

{"@timestamp":"2020-11-04T14:27:14.666Z","@metadata":{"beat":"auditbeat","type":"doc",
"version":"6.8.12"},"auditd":{"sequence":989996,"result":"success","session":"1402","data":
{"a3":"20656c706f657020","tty":"(none)","a2":"1000","arch":"x86_64","syscall":"read",
"exit":"4096","a1":"5583baa77f70","a0":"5"}},"user":{"name_map":{"suid":"root",
"auid":"junzeng","egid":"root","euid":"root","fsuid":"root","gid":"root","sgid":"1000","arch":"x86_64","syscall":"root",
"sgid":"1000","egid":"0","auid":"1000","uid":"0","process":{"exe":"/usr/sbin/sshd",
"pid":"7959","ppid":"1689","name":"sshd"}

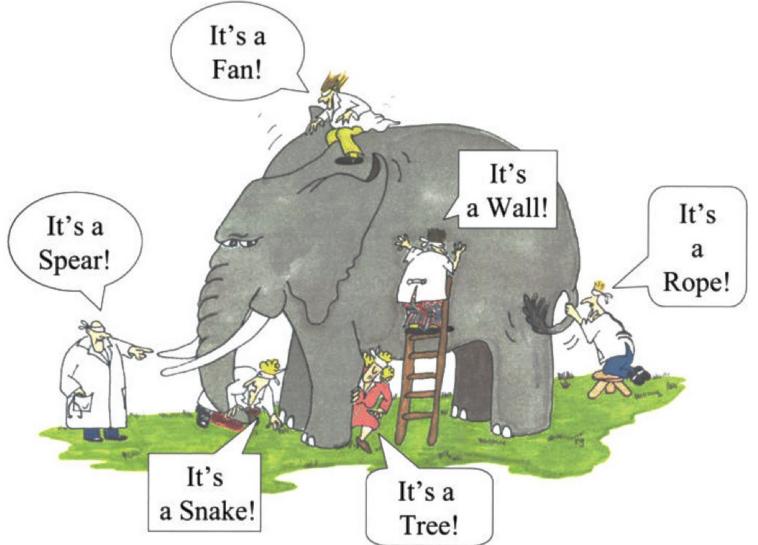
Enterprise Network View



High-level Report



Understanding of Cyber Security Events



Endpoint Monitoring Solutions

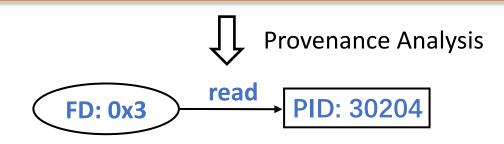
Endpoint monitoring solutions record audit logs for attack investigation



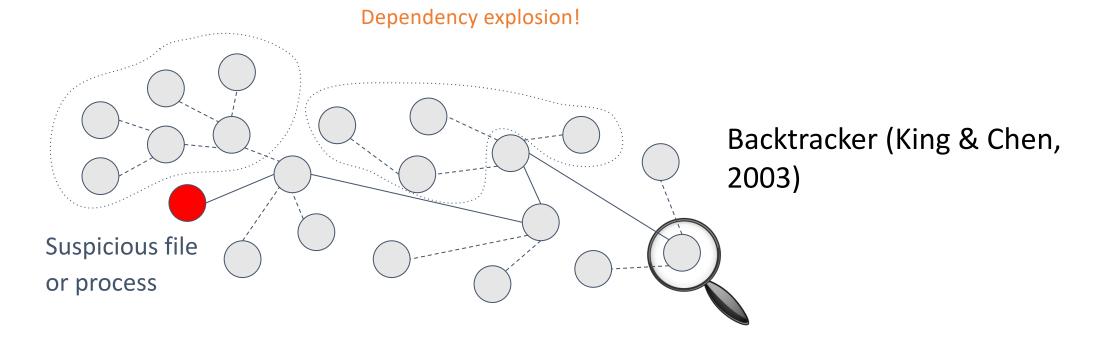
Audit logs:

- A history of events representing OS-level activities
- Provide visibility into security incidents with data provenance

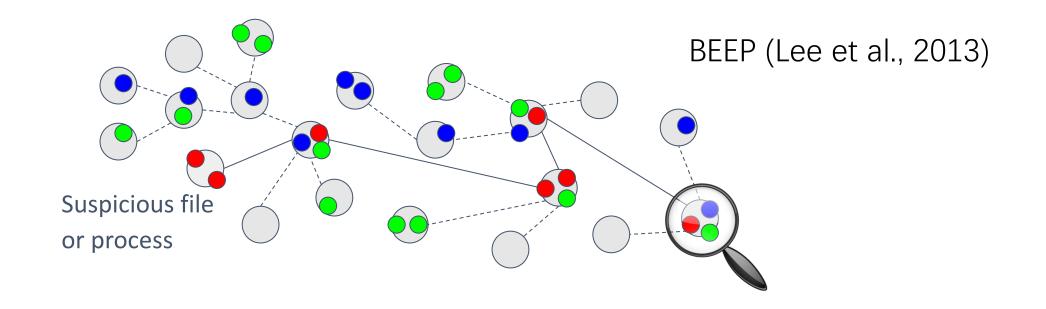
type=SYSCALL msg=audit(**30/09/19 20:34:53**.383:98866813) : arch=x86_64 **syscall=read** exit=25 **a0=0x3** ppid=15757 **pid=30204** auid=junzeng sess=6309



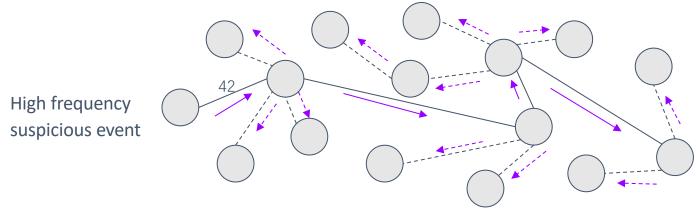
- Starting from a detection point, *Backtracker* does:
 - Events & objects identification related detection point
 - Generate dependency graph
 - Use rules to prune unrelated nodes in the dependency graph



- Resolve *dependency explosion* problem in a long running application
 - Fine-grained provenance tracing technique
 - Identifying unit boundaries & dependences
 - Partition into individual unit
 - Code instrumentation

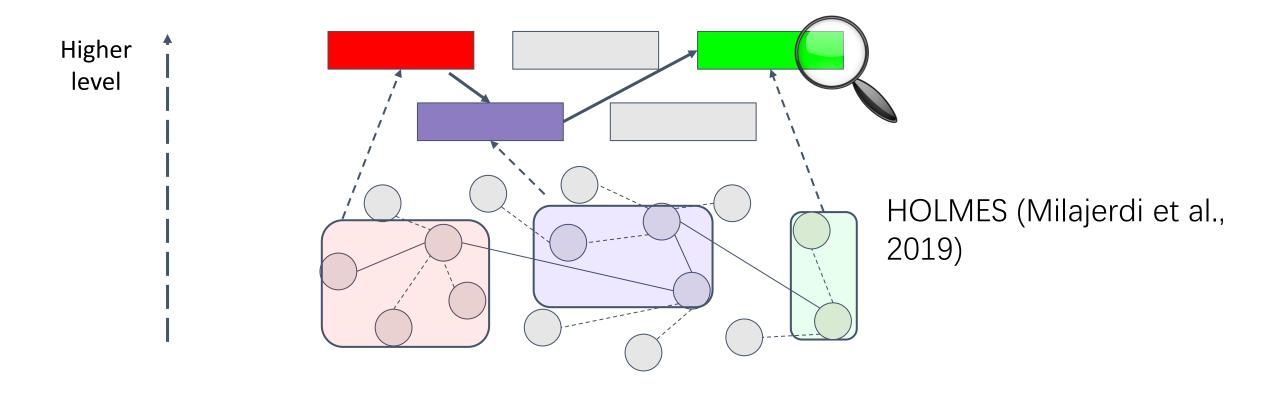


- Address *threat alert fatigue* during threat investigation
 - Sssign anomaly scores to every edge in dependency graph
 - Based on frequency of events that have occured (historical & contextual information)
 - Propagated score through edges in the graph
 - Generate aggregated anomaly score for triaging



NoDoze (Hassan et al., 2019)

- Generate high-level graph during threat investigation
 - Develop robust & reliable detection signal
 - Correlate between suspicious information flow



Related Work

• Scale up provenance analysis:

• Data reduction [NDSS'16, 18 ...] & Query system [Security'18, ATC'18 ...]

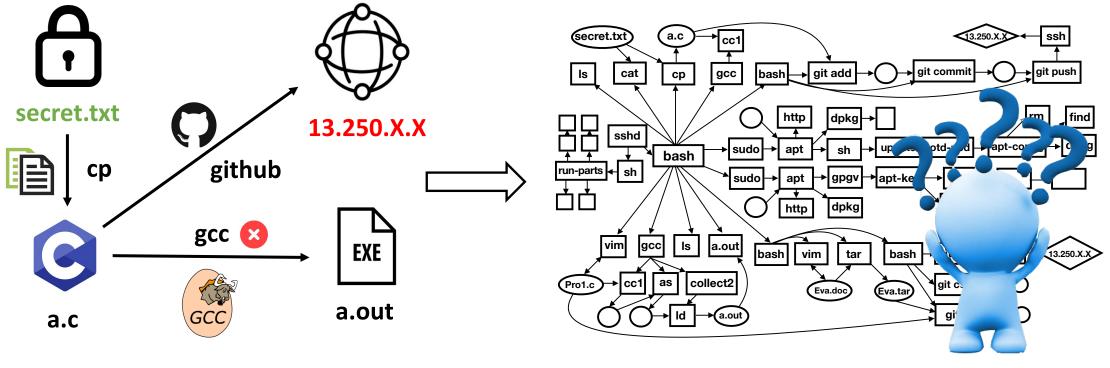
Can we automatically **abstract** high-level behaviors from low-level audit logs and **cluster** semantically similar behaviors before human inspection?

 Query graph [VLDB'15, CCS'19], Tactics Techniques Procedures (TTPs) specification [SP'19,20], and Tag policy [Security'17,18]

Behavior-specific rules heavily rely on domain knowledge (time-consuming)

Motivating Example

Attack Scenario: A software tester exfiltrates sensitive data that he has access to

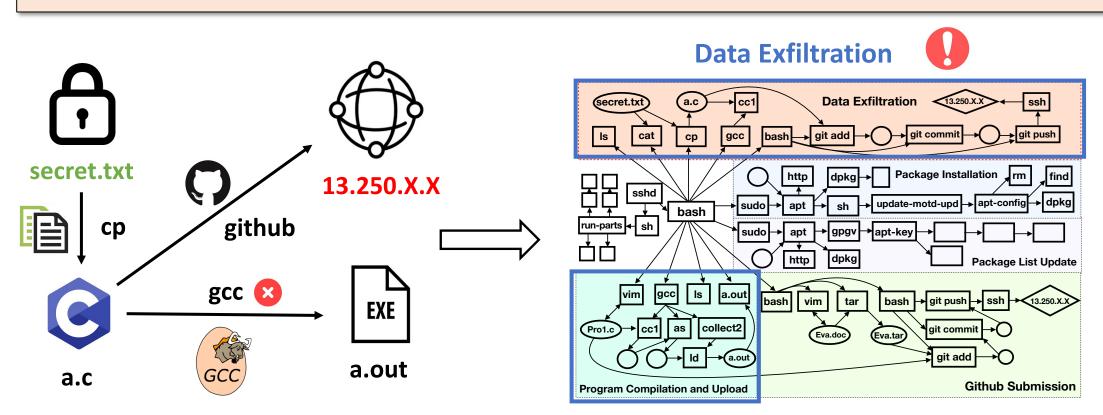


Motivating Example Logs

Data Exfiltration Steps

Motivating Example

Attack Scenario: A software tester exfiltrates sensitive data that he has access to



Data Exfiltration Steps

Program Compiling and Upload (cluster)

Motivating Example Logs

Challenges for Behavior Abstraction

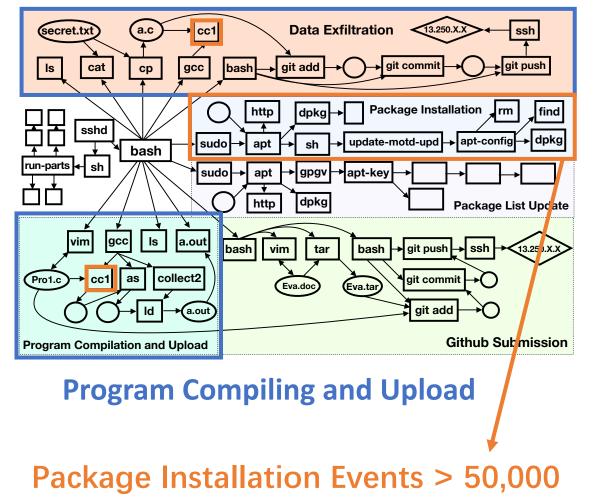
Data Exfiltration

Event Semantics Inference:

 Logs record general-purpose system activities but lack knowledge of high-level semantics

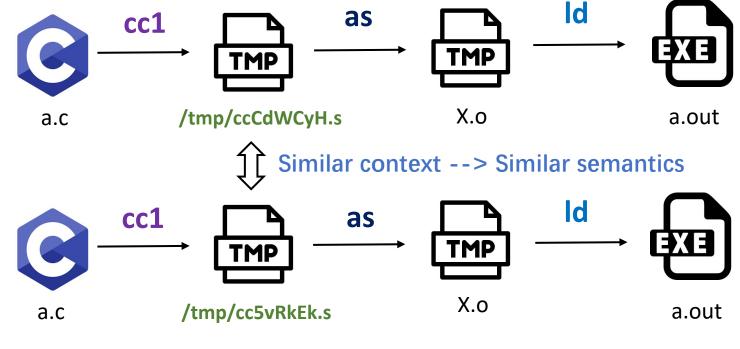
Individual Behavior Identification:

- The volume of audit logs is **overwhelming**
- Audit events are highly interleaving



Our Insights

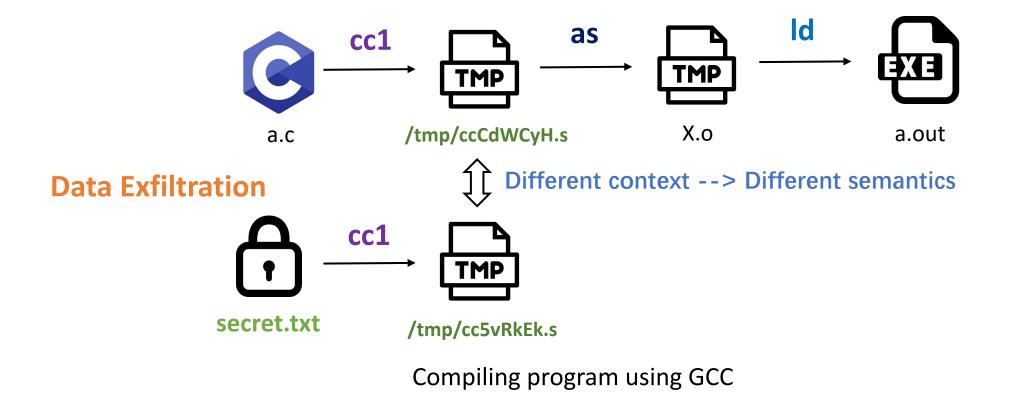
How do analysts manually interpret the semantics of audit events?



Compiling program using GCC

Our Insights

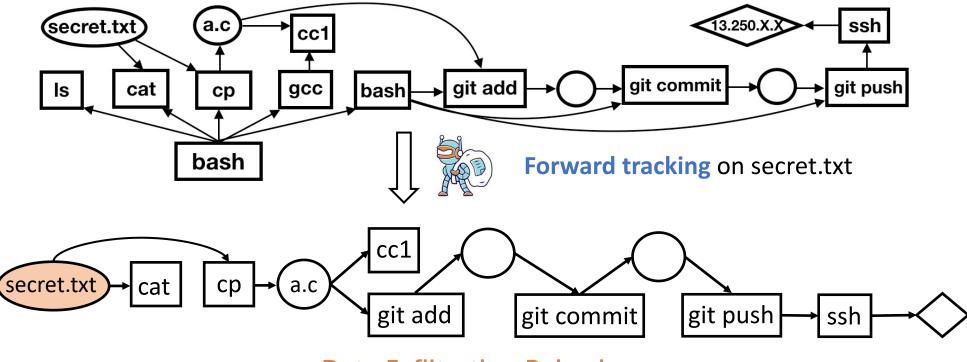
How do analysts manually interpret the semantics of audit events?



Reveal the semantics of audit events from their usage contexts in logs

Our Insights

How do analysts manually identify behaviors from audit events?



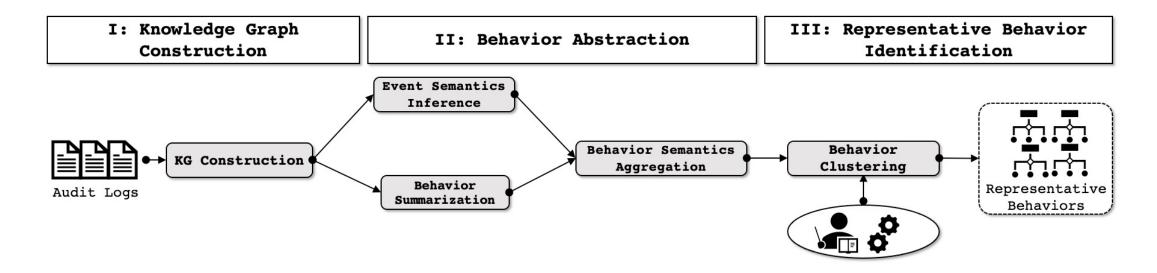
Data Exfiltration Behavior

Summarize behaviors by tracking information flows rooted at data objects

WATSON

An automated behavior abstraction approach that aggregates the semantics of audit logs to model behavioral patterns

- Input: audit logs (e.g., Linux Audit^[1])
- Output: representative behaviors



Knowledge Graph Construction

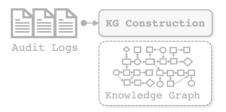
We propose to use a **knowledge graph** (KG) to represent audit logs:

- KG is a directed acyclic graph built upon triples
- Each triple, corresponding to an audit event, consists of three elements (head, relation, and tail):

 $\mathcal{KG} = \{(h, r, t) | h, t \in \{Process, File, Socket\}, r \in \{Syscall\}\}$

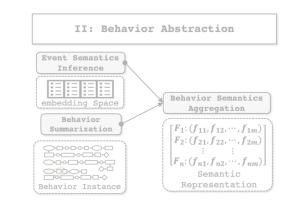
• KG unifies heterogeneous events in a homogeneous manner

I: Knowledge Graph Construction



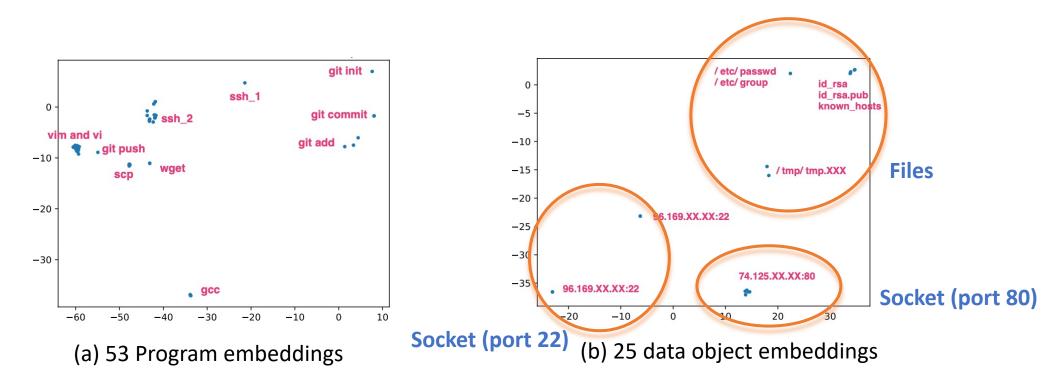
Event Semantics Inference

- Suitable granularity to capture contextual semantics
 - Prior work [CCS'17] studies log semantics using events as basic units.
 - Lose contextual information within events
 - Working on **Elements** (head, relation, and tail) preserves more contexts
- Employ an embedding model to extract contexts
 - Map elements into a vector space
 - Spatial distance represents semantic similarities
 - TransE: a translation-based embedding model
 - Head + Relation ≈ Tail → Context decides semantics



Event Semantics Explicability

Use t-SNE to project the embedding space (64 dimensional in our case) into a 2D-plane, giving us an intuition of embedding distribution

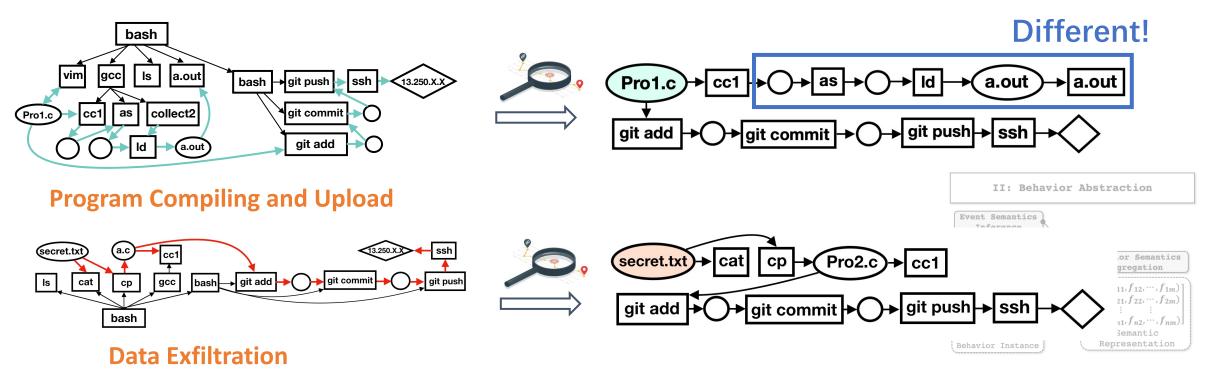


Semantically similar system entities are clustered in the embedding space

Behavior Summarization

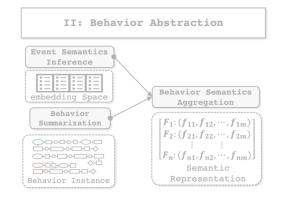
Individual behavior identification: Apply an adapted depth-first search (DFS) to track information flows rooted at a data object:

- Perform the DFS on every data object except libraries
- Two behaviors are merged if one is the subset of another



Behavior Semantics Aggregation

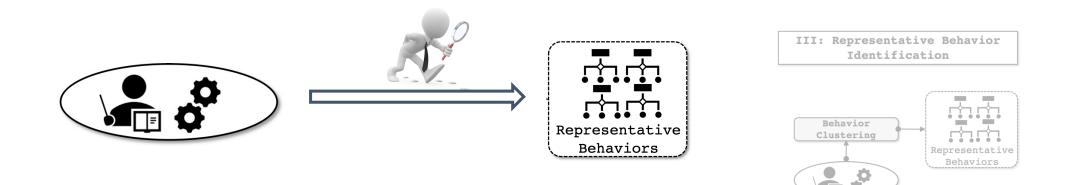
- How to aggregate event semantics to represent behavior semantics?
 - Naïve approach: Add up the semantics of a behavior's constituent events
 - Assumption: audit events equally contribute to behavior semantics
- Relative event importance
 - Observation: behavior-related events are common across behaviors, while behavior-unrelated events the opposite
 - Apply frequency as a metric to define event importance
 - Quantify the frequency: Inverse Document Frequency (IDF)
- The presence of **noisy events**
 - Redundant events [CCS'16] & Mundane events





Representative Behavior Identification

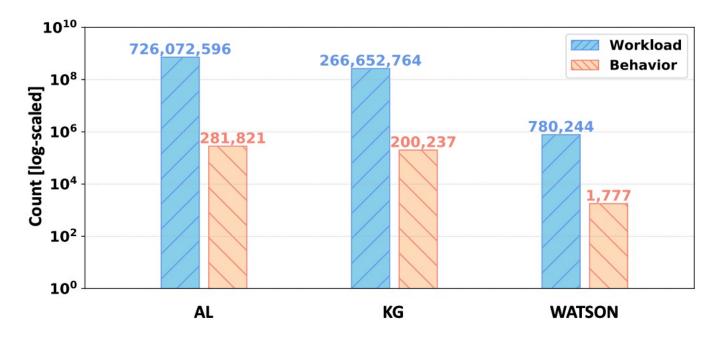
- Cluster semantically similar behaviors: Agglomerative Hierarchical Clustering analysis (HCA)
- Extract the most representative behaviors
 - Representativeness: Behavior's average similarity with other behaviors in a cluster
 - Analysis workload reduction: Do not go through the whole behavior space



Efficacy in Attack Investigation

Measure the **analysis workload reduction** of APT attack investigation in the DARPA TRACE dataset:

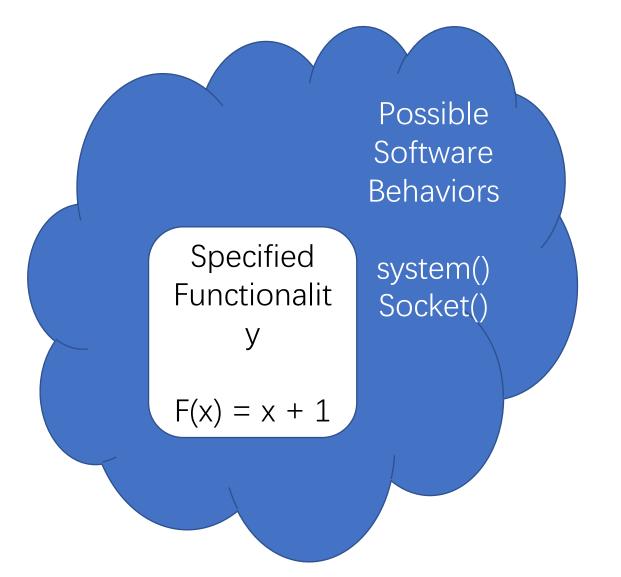
• Analysis workload: the number of events to recognize all behaviors



Two orders of magnitude reduction in analysis workload and behaviors

Functionality, Flexibility, and Security

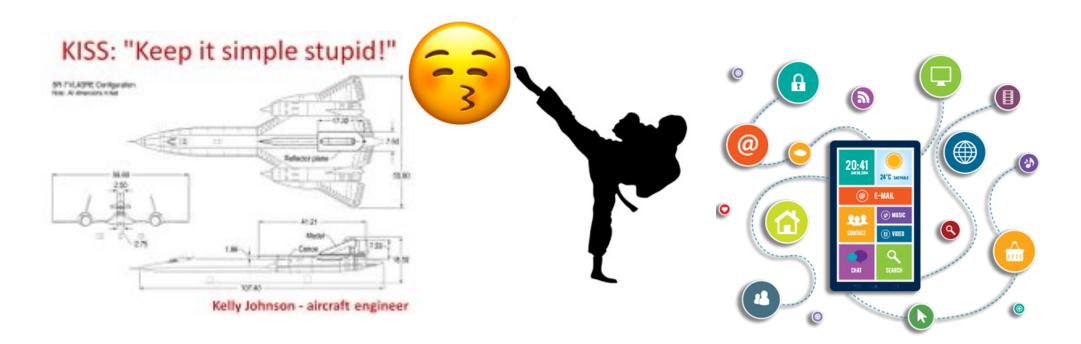
- Security is about "nothing else"
 - Specified functionality and **only** specified functionality
- Flexibility is the root of many security problems



Simplicity in System Design

• KISS (Keep It Simple, Stupid)

• KICS (Keep It Complex & Smart)



Dimensions of System Research

Human	Human factor, social engineering	Law, policy, politics	
System	lsolation, memory exploit, provenance analysis, …	Psychology, cognition, responsibility	

Human

System

Objective for Graduate Education

- Deep knowledge and skills in technology domain
 - Abstraction and presentation of thoughts
- Ability to think and analyze broadly
 - Especially in challenging times, calling for independent minds
- Understanding systems better
 - Order, flexibility, force, …
- Form and live with a philosophy
 - Embrace trust
 - Life with minimal dependency

Research Areas and Range of Development

DANGER-KEEP OUT! 危險,請避開! BAHAYA-JANGAN DEKAT! Aunwia-A௫சில்வராதர்கள்!

Views from different perspectives Culture, social system, etc.



Top-ranked degree program and open culture Top researchers in various fields



Economics, business, and technology Fintech Institute

Trusted environment in Web/Mobile

System and security: National Cybersecurity R&D Lab Behavior/Psychology and security: CFPR in Arts and Social Science

Binary and System Analysis Attack diagnosis and attribution

Human behavior in cyber experimentation

- Multiple-level views of cyber incidents
- Our Insights in log analysis
 - Infer audit event semantics by usage contexts
 - Identify behaviors with information flows rooted at data objects
- On system research

Understand the movement of the sun and moon from traces of shadows under the roof.

审堂下之阴,而知日月之行,阴阳之变也



Understanding systems 理解系统 Abstracting knowledge 提炼知识 Connecting facts 参悟规律

