

MAYALOK: A Cyber-Deception Hardware Using Runtime Instruction Infusion

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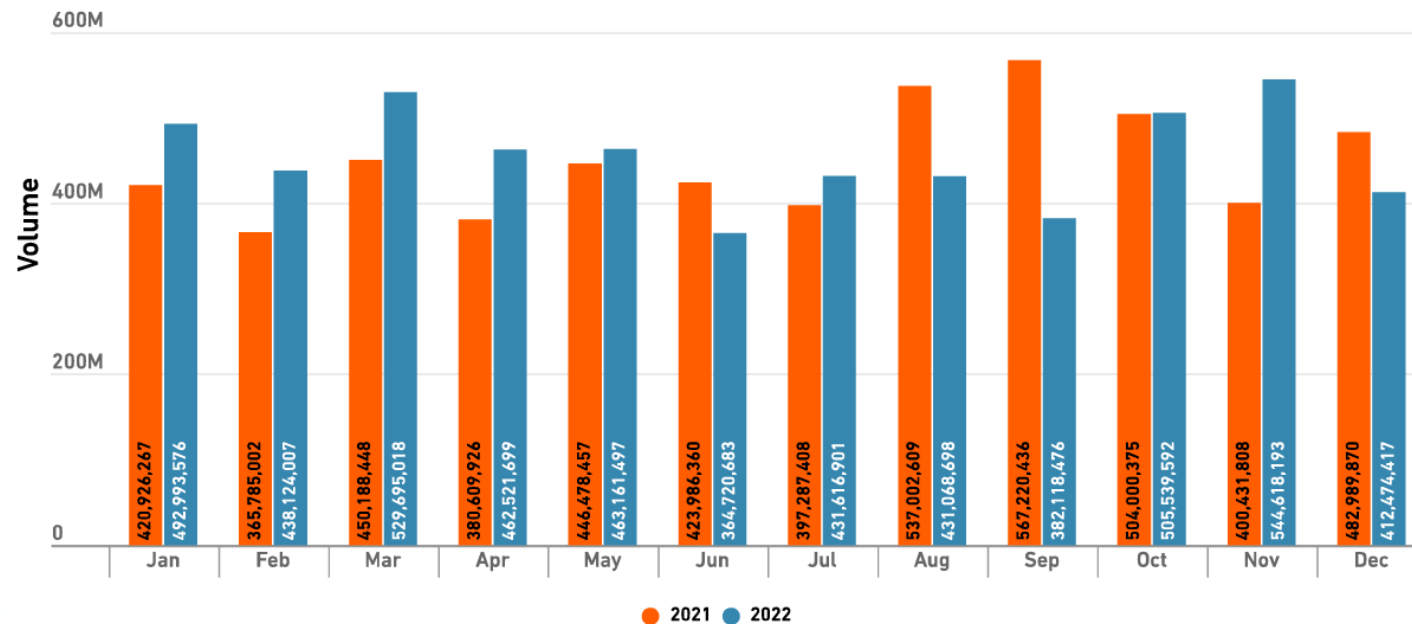
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Evolving Digital Threats

Malware facts of 2022^[1]:

- A total of **5.5 billion** attacks attempted
- **465,501** were never-before-seen malware variants
- Ransomware held majority at **493.3 million**
- Other prominent threats: cryptojacking, IoT malware, malicious PDFs, encrypted attacks



Total volume of attacks in 2021 and 2022 is **10.9 billion!**

[1] "2023 SonicWall Cyber Threat Report", SonicWall Whitepaper, 2023.

Reactive Defenses Against Malware

Detect-and-Stop

- *Signature-based* → binary analysis, hash-check, fingerprinting
- *Trace-based* → runtime logs of syscalls, instructions, network packets
- × Fails against unknown malware due to static nature
- × Unfeasible runtime cost

System Hardening

- *Software* → compiler optimizations, feature reduction, digital signatures
- *Hardware* → isolated enclaves, memory protections, cryptographic modules
- × Patching after deployment can be impossible
- × Risks exposing defense mechanism to attackers

Cyber-Deception

Proactive in protecting sensitive assets

Interact with adversaries using honey-resources

Profile attackers to build robust defenses

Why Cyber-Deception?

Prevention

Protect sensitive system assets

- **Slow down** attackers with bloated system assets
- **Redirect** malicious actions to honey-resources
- *e.g., honey-pot servers, moving target defenses*

Engage

Support

- **Confuse** attackers with manipulation
- **Influence** the attack path towards
- *e.g., instrumenting malware binaries, software honey-patches*

Current techniques are underwhelming in their ability to provide **efficient** and **effective** cyber deception!

Research

Preserve attack forensics

- **Study** the collected data to create robust system designs
- **Reinforce** security measures with high-fidelity attacker details
- *e.g., honey-VMs used as sandboxes*

Related Efforts

- ❑ Qassrawi and Hongli, “Deception methodology in virtual honeypots,” IEEE NSWCTC, 2010.
 - Honeypot VMs for attacker interaction
 - Vulnerable to artifact discovery techniques
- ❑ Sajid et al., “Soda: A system for cyber deception orchestration and automation,” ACM ACSAC, 2021.
 - API hooking-based malware instrumentation
 - High execution time overheads and limited to API calls
- ❑ Gallagher et al., “Morpheus: A vulnerability-tolerant secure architecture based on ensembles of moving target defenses with churn,” ACM ASPLOS, 2019.
 - Hardware-based obfuscation using MTD ensembles
 - High impact on performance of all applications

Supporting Cyber-Deception

Dynamic

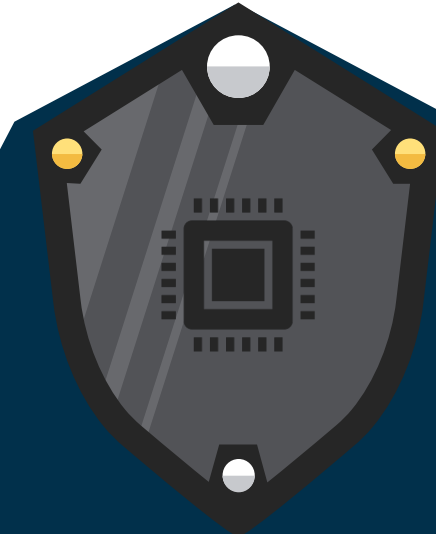
Implement deception tactics during malware execution

Coverage

Target multiple applications with the same deception platform

Independent

Modify malware execution path with minimal external dependencies



Hardware provides these features!

Low-overhead

Sustain a nominal runtime cost for implementing deception tactics

Transparent

Deception activity should be invisible to suspecting adversaries

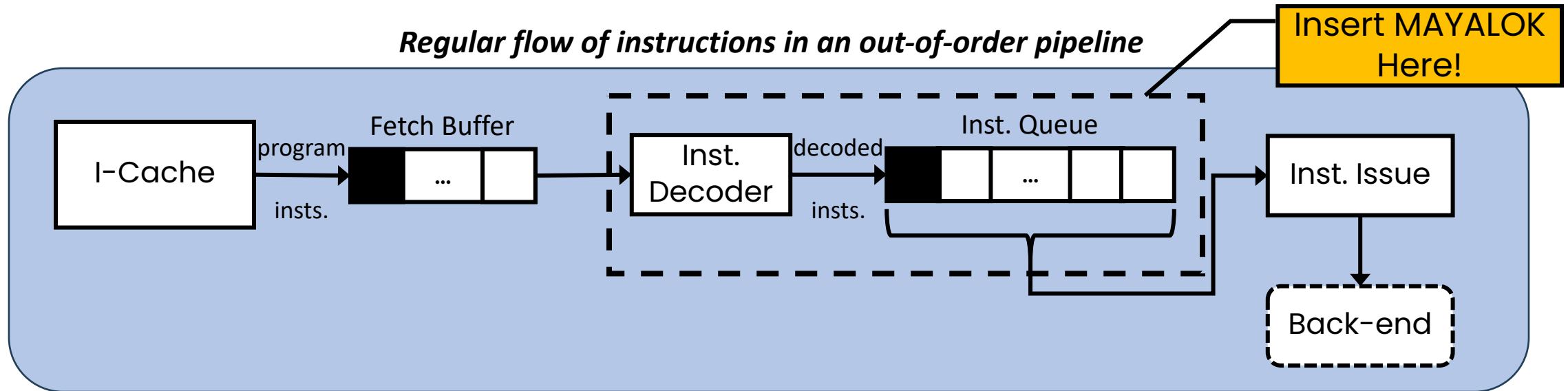
Customizable

Provide the ability to alter deception tactics depending on malware type

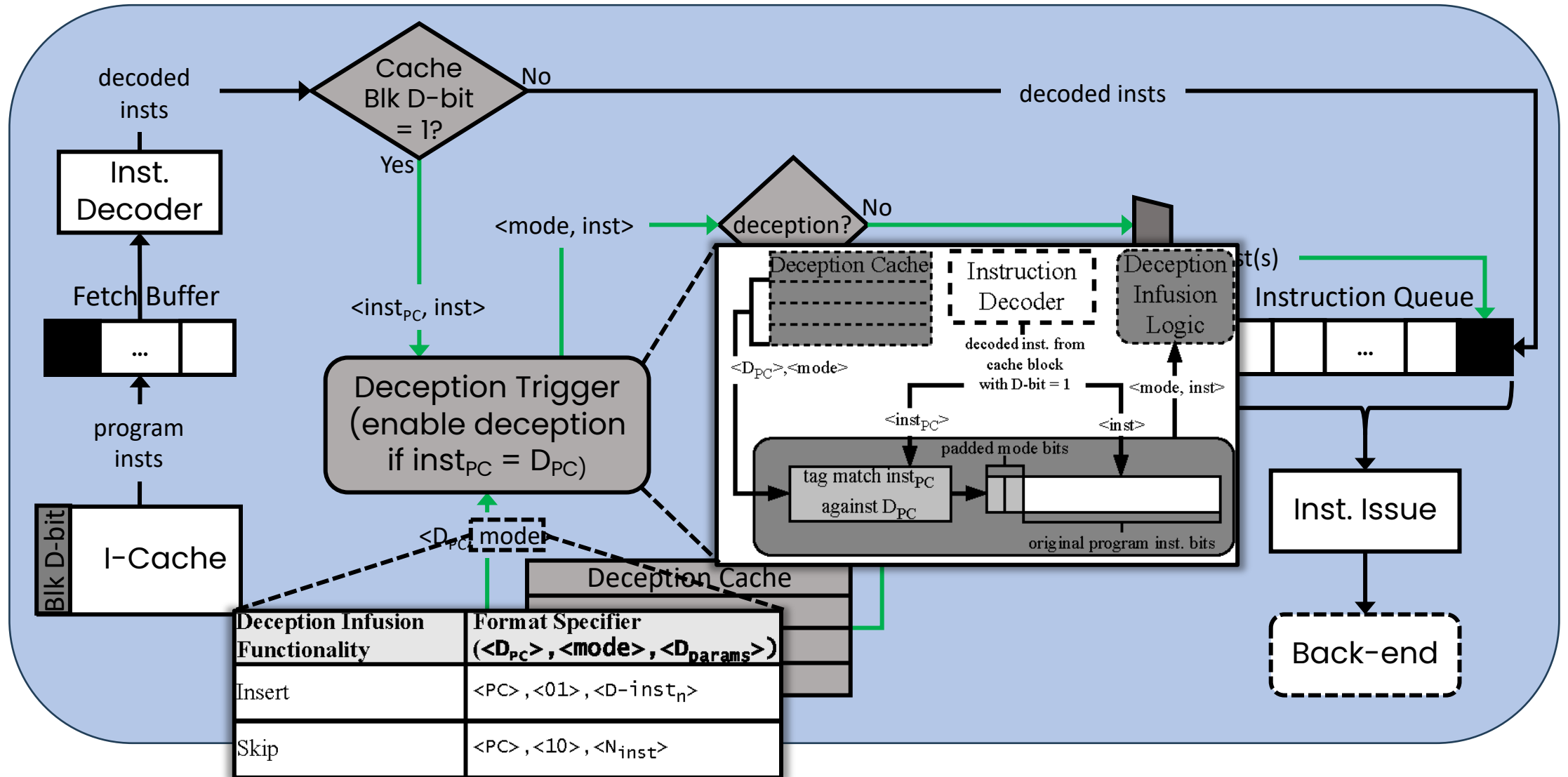
The MAYALOK* Cyber-Deception Hardware

□ A **front-end** implementation in the processor pipeline:

- ✓ Each application executing **malicious instructions** is tagged
- ✓ Instructions are **dynamically** inserted/removed
- ✓ At the **fetch/decode** stage
- ✓ Preserves the **efficiency** of processor pipeline for non-deception instructions



**MAYALOK in Sanskrit means an imaginary realm*



MAYALOK Hardware Overhead

- ❑ MAYALOK hardware structure with the highest overhead is the Deception Cache:
 - Requires a **2KB** cache to store relevant **D_{PC}** and **D_{params}**
 - Supports **16 entries** of **128 bytes** each for a malware process
 - Sufficient to track most malicious **$inst_{PC}$** based on our analysis of popular malware samples
- ❑ Instructions targeted for infusion are tagged pre-runtime
 - Using binary analysis tools like Angr
 - Optimizes the deception trigger mechanism

(All experiments were conducted on an x86 build of the Gem5 simulator clocked at 2GHz)

Security Analysis on Malware

Malware Sample (Type)	Security Outcome	Infusion Mode	Attacker Penalty
Return-to-libc (Buffer Overflow)	Pointer Protected	Insert	Failed overflow
RSA Timing Attack (Side Channel)	Keys Protected	Insert	No Side Channel
Petya (Ransomware)	Files Protected	Insert & Skip	∞ time
Credential Stealer (InfoStealer)	Credentials Protected	Insert	Fake data supply

Conclusion

- ❑ Showed the promise of a new security paradigm i.e., **cyber-deception**
- ❑ Presented an **efficient** hardware support for runtime **instruction infusion**
- ❑ Demonstrated the **effectiveness** of MAYALOK against multiple malware samples

Future of hardware in cyber-deception:

- ✓ **Additional primitives** for improved cyber-deception support
- ✓ **Automating** cyber-deception to adapt itself with new attack vectors

“MAYAVI: A Cyber-Deception Hardware for Memory Load-Stores”, ACM GLSVLSI.
Knoxville, Tennessee. June 5 – 7, 2023.

Thank You

Preet Derasari, Kailash Gogineni, and Guru Venkataramani

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This research was supported by the U.S. Office of Naval Research