

Microarchitectural Leakage Templates and Their Application to Cache-Based Side Channels CCS 2022

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Previously...

Security of proprietary CPUs

Discovering new side channels

Generic Description of a Side Channel

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- P(A): A **code** template
	- B: Distinct **behaviors**
		- e.g. timing: $\mathcal{B} = \{ \bullet \text{ fast}, \circ \text{slow} \}$
- $R(A, b)$: **Relations** between inputs, leading to a certain behavior
	- "When inputs X and Y are in relation, then behavior •"

Leakage Template

Template: Cache-Timing Side Channel

Figure: Leakage Template: Cache-Timing Side Channel

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Case Studies

Figure 6: Case studies' LTs with selected relations. In (a) a^b means b times inlining repetition of instruction a . In (b), π_{li} is inlining n_i simple arithmetic, logical or nop instructions. For (a) and (b) triggering and not triggering behavior are denoted by • and ◦, respectively. In (c), P denotes distinguishing behaviors and is the number of prefetched lines. Relations must be checked in order, the first matching relation determines the number of expected prefetches.

Table 3: Example permutation outcome. Each number represents an instruction from the initial testcase. Underlined numbers are loads from addresses that have the same tag.

ases show the same behavior. Thus, the exact values of tags and sets do not matter. E4: Word Offset Behavior. In E2, we observed that the byte

offsets of loaded addresses affect previction. To broaden our understanding, in this experiment, we leveraged GTSes as shown in Table 4. They generate testcases for 5-load programs with all possible combinations of tags and sets (for loads targeting up to two Microarchitectural Leakage Templates and Their Application to Cache-Based Side Channels CCS '22, November 7–11, 2022, Los Angeles, CA, USA.

Table 6: Transmission and error rates of sota. covert channels.

8.2 Previction w/o Shared Memory (**PR_PP**)

Based on experiment E5 in § 6.2, previction may target preloaded memory addresses and leak information in the absence of shared memory, e.g., through Prime+Probe. The sender code of our previction-based Prime+Probe primitive PR_PP is similar to that of PR_FR. However, in PR_PP, the receiver first loads two memory lines into the targeted cache set before the execution of the sender code. The receiver then probes the lines to determine the leaked bits.

8.3 Prefetching Control-Flow Leakage (**PRF_CF**)

PRF_CF allows leaking the control flow of a program based on prefetching. It is based on the results of E7 in § 6.3. Fig. 8 shows an example code of PRF_CF. The sender code has a 4-load prefetching sequence with a fixed stride (lines 2, 5, 8, and 15). The loads are separated by a number of arithmetic instructions. The instruction at line 12 is conditionally executed depending on one bit of a secret that is stored in x20 (lines 9 through 12). According to E7, the number of executed instructions within a prefetching sequence affects the number of prefetched cache lines. By measuring the time required to reload a (possibly prefetched) address x1+512, the receiver can determine whether an instruction was executed and consequently learn the secret bit.

8.4 Prefetching on an Interrupted Seq. (**PRF_IS**)

Inspired by E7, we tested the effect of intermediate memory operations on prefetching. We observed that an intermediate load from a different page leads to prefetching of additional cache lines by a 3-load stream. PRF_IS is based on this outcome. It also allows

In the paper: 3 Leakage Templates, 4 Covert Channels

Searching for Instances of a Leakage Template

Recall:

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Leakage Template

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1. **Static Analysis**

Search for candidate code sections matching $P(A)$

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2. **Dynamic Analysis**

For each candidate section:

Check whether different inputs fulfill relations for different behaviors

(= are distinguishable based on behavior)

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et al. [40]. Data-dependent loads from a lookup table may or may not trigger the prefetcher to load certain cache lines into the cache, depending on the resulting memory access pattern. Therefore, the cache state of potentially prefetched cache lines indicates the existence of relations between the accessed lookup table elements and, by extension, the processed data. Shin et al. exploit these relations to leak the scalar of a scalar point multiplication on an elliptic curve. In Elliptic Curve Diffie-Hellman (ECDH), a scalar represents the private key. The attack recovers the key incrementally. The same computation is applied to both the target scalar and a candidate scalar. By changing the candidate scalar such that the prefetching behavior assimilates, both scalars assimilate as well. Even though this vulnerability is no longer present in recent OpenSSL versions, we still consider it a reasonable case study to demonstrate that LTs can be used to identify real-world vulnerabilities in binaries.

Approach: Combining Static and Dynamic Analysis. Shin et al. [40] limit the scope of their search to a specific cryptographic operation. In contrast, our starting point is the whole OpenSSL binary. We combine static and dynamic binary analysis techniques to search it for instances of the prefetching LT (see Fig. 6.c). First, we scan the binary for code sections that match the code pattern $P(A)$ of the LT. This results in a list of candidate code sections that potentially contain a prefetching side-channel. Second, we need to check whether a candidate section satisfies different relations $\mathcal{R}(A, b)$ for different input values. If this is the case, we expect the section to show input-dependent behavior, indicating a side channel. Not all relations can be resolved statically, especially if they refer to addresses in instruction operands. To overcome this, we dynamically analyze the target code to learn its concrete addresses.

Performing Static Analysis. We use asmregex [5] to statically

analysis step to check a candidate code section for input-dependent

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Table 5: Confusion matrix, comparing prefetching behavior classification based on relations with the actual behavior.

prefetching behavior based on the relations $\mathcal{R}(A,b)$ from the LT.
Second, we use a Flush+Reload side channel to record a *cache trace*. This trace contains the cache state of the memory lines around SOR tb after execution. It is captured for evaluation purposes and indicates the actual prefetching behavior of the CPU.

In order to show that the LT accurately represents the prefetching behavior, we recorded traces for 100 random input values to the library function. For each input value, we determined the expected prefetching behavior using the access trace² and compared it with the actual behavior using the corresponding cache trace.

Evaluation. Table 5 illustrates the classification performance. For all 66 cases where the load instructions satisfy the relations for P₀, the cache traces show that no prefetching occurred. In six cases, the relations for behavior P₅ are satisfied. The three relevant load instructions load data from three consecutive cache lines and the number of instructions between the load instructions (n) and 2) is within the specified bounds. In all six cases, the cache trace shows that prefetching of three additional cache lines occurred. In the remaining 28 cases, the relations for none of the behaviors from the LT are satisfied. The reason is that the distances n_1 and n_2 between the relevant load instructions are outside the parameter

range we tested when the LT was created. We denote these cases

In the paper: Re-identifying a known vulnerability *(Shin et al.¹, CCS'18)*: Prefetching-based side channel in Elliptic Curve Diffie-Hellman (ECDH) code in OpenSSL 1.1.0g

PLUMBER USE Cases and Limitations

Additional Use Cases

- Facilitate reverse engineering of microarchitectural components
	- Examples in the paper: branch predictor, cache slice mapping

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Limitations

- Focus on cache-based side channels
- Implemented for ARM architecture

Leakage Template

• Code

- Behaviors
- Relations

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Matching Binaries

1. F Static 2. Dynamic

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Generative Testcase Specification (GTS)

[Backup](#page-40-0) [References](#page-50-2)

哈 Generative Testcase Specification (GTS)

A **Classifier**

- Classifies test cases based on the observed behavior
- For each behavior: produce a bit table
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Classifier

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- For each behavior: produce a bit table
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Analyzer

- For each bit table (= behavior): Identify common features
- \Rightarrow Extracts relations that trigger a certain behavior

Prefetching on ARM Cortex-A53

• Loads cache lines in advance that are likely to be needed soon

Figure: Leakage Template: Prefetching. P_1 means prefetching I lines.

Prefetching on ARM Cortex-A53

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Steps to Create the Leakage Template

- 1. Number of sequential loads
- 2. Intermediate instructions
- 3. Respecting page boundary
- 4. Multiple prefetching sequences
-

5. Cache hits Figure: Leakage Template: Prefetching. P_1 means prefetching *l* lines.

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Re-Identifying a Prefetching-Based Vulnerability in OpenSSL

Vulnerability (Shin et al.² **CCS'18):**

Prefetching-based attack on Elliptic Curve Diffie-Hellman (ECDH) in OpenSSL 1.1.0g

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- Search for the code template from the prefetching Leakage Template
- \Rightarrow Identified 429 matching sequences across 18 OpenSSL modules (including the target code section)

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- Run code with different inputs
- Evaluate register contents against relations
- \Rightarrow Different inputs satisfy relations for different behaviors

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Conclusion: Different classes of inputs are distinguishable based on prefetching behavior.

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