HODOR: Shrinking Attack Surface on Node.js via System Call Limitation

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ABSTRACT

Node.js applications are becoming more and more widely adopted in the server side, partly due to the convenience of building these applications on top of the runtime provided by popular Node.js engines and the large number of third-party packages provided by the Node Package Management (npm) registry. Node.js provides Node.js applications with system interaction capabilities using system calls. However, such convenience comes with a price, i.e., the attack surface of JavaScript arbitrary code execution (ACE) vulnerabilities is expanded to the system call level.

There lies a noticeable gap between existing protection techniques in the JavaScript code level (either by code debloating or read-writeexecute permission restriction) and a targeted defense for emerging critical system call level exploitation. To fill the gap, we design and implement HODOR¹, a lightweight runtime protection system based on enforcing precise system call restrictions when running a Node.js application. HODOR achieved this by addressing several nontrivial technical challenges. First, HODOR requires to construct highquality call graphs for both the Node.js application (in JavaScript) and its underlying Node.js framework (in JavaScript and C/C++). Specifically, HODOR incorporates several important optimizations in both the JavaScript and C/C++ level to improve the state-of-the-art tools for building more precise call graphs. Then, HODOR creates the main-thread whitelist and the thread-pool whitelist respectively containing the identified necessary system calls based on the call graphs mappings. Finally, with the whitelists, HODOR implements lightweight system call restriction using the Linux kernel feature Secure Computing Mode (seccomp) to shrink the attack surface. We utilize HODOR to protect 168 real-world Node.js applications compromised by arbitrary code/command execution attacks. HODOR could reduce the attack surface to 19.42% on average with negligible runtime overhead (i.e., <3%).

1 INTRODUCTION

Node.js is an open-source, cross-platform JavaScript runtime environment [1], which allows JavaScript code to be executed on the server-side. Currently, some well-known websites use Node.js in their web product, such as Paypal [2], LinkedIn [3], Microsoft [4], and Netflix [5]. The success of Node.js largely owes to the libraries that Node.js depends on. Moreover, Node.js application developers can easily invoke and manage third-party libraries through Node Package Manager (npm) [6]. For example, the libuv [7] library provides Node.js with asynchronous I/O capabilities and V8 [8] library provides Node.js with a JavaScript engine.

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These libraries enable the Node.js applications to interact with the low level system kernel, which however may pose severe security threats besides convenience. Specifically, Node.js applications are likely to depend on obsolete third-party libraries with different kinds of security issues, which are expanded into the Node.js ecosystem [9-12]. As observed in [10], one-quarter of the versions of packages, which make up 19.63% of the npm ecosystem, have dependencies on vulnerable packages. such as gadget chain attacks (prototype pollution attacks) [13, 14], inject-related attacks [15], and supply chain attack [11]. Most of them may lead to ACE attacks [16] (i.e., arbitrary code execution attacks). Take the prototype pollution attack as an example. When the attacker pollutes Object.prototype.someattr and the application runs the code snippet eval(someobject.someattr), s/he can perform the ACE attacks [13, 17]. When armed with the ability of ACE attacks, attackers can further perform critical operations such as reading private files, creating scheduled tasks, and reversing shells.

Existing protection techniques against the above security implication mostly focus on the JavaScript code, either by code debloating [18, 19] or read-write-execute permission restriction [20]. For instance, Mininode [18] and stubbifier [19] aim to identify and remove useless JavaScript code to reduce the attack surface so that the attackers cannot easily exploit the compromised third-party modules. MIR [20] designs a fine-grained read-write-execute permission model and wraps the privilege reduction over every module

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¹HODOR refers to the role of Hodor in the Game of Thrones, who sacrificed his life to resist the attack of the White Walkers by holding the door. We name our tool HODOR to signify its protection of the Node is applications against malicious payloads attack by restricting the use of system calls.

based on the function closure mechanism. Although effective in different ways, existing protections targeting the JavaScript code level have several limitations faced with diverse exploitation possibilities. First, they cannot limit the attack surface of arbitrary command execution. For example, when the attacker can pass any malicious data to exec method, s/he can execute any command without utilizing any third-party modules or even the JavaScript language. Moreover, existing protections [18–20] require to modify the source code of the Node.js application which may influence the application's normal operation to an uncertain degree.



Figure 1: HODOR protection birdview.

In this work, we provide a complementary perspective to shrink the attack surface on Node.js applications at the lower system call level to prevent critical operations from attackers as shown in Fig.1. We notice some attack surface reduction works at the system call level for application scenarios other than Node.js [21-29]. However, such low level protection is currently missing for Node.js and several unique challenges are to be properly addressed. • We need to first precisely identify those system calls used by the Node.js application during execution, which require cross-language mapping from the JavaScript source code to the underlying system calls. Previous approaches are not enough as they mostly generate system call lists over a single language [21-28]. Another work Saphire [29] scans the AST to identify built-in function calls and builds call graphs over the compiled binary from Debian repositories (to address crosslanguage challenge), resulting in some loss of accuracy. 2 Moreover, the highly dynamic nature of JavaScript language makes call graphs construction approaches solely relying on static or dynamic program analysis [18–20] imprecise. ⁽³⁾ The restriction needs to be transparent to the Node.js framework: the restriction needs to be easily integrated into the Node.js framework and will not influence the normal operation of the application with acceptable overhead. Previous restrictions are mostly performed on the process granularity [21-29], which are inefficient for Node.js which is multi-threaded.

To address the challenges, we develop a lightweight system call level protection system HODOR specifically designed for Node.js applications. HODOR is equipped with three core techniques to tackle each challenge: **1** and **2** To generate a precise system call list, HODOR performs cross-language and combined static-dynamic call graphs analysis for the Node.js applications and Node.js framework over JavaScript code and C/C++ code. In particular, HODOR incorporates several significant optimizations to improve state-of-the-art call graphs building methods for both the JavaScript code [30, 31] and the C/C++ code [32]. For JavaScript code, we propose combined static-dynamic analysis to build call graphs for Node.js and we consider the code pattern related to the execution pattern of builtin method features. As for C/C++ code, we propose new partial context-sensitive mechanisms to generate more precise call graphs. Our experimental results show that these optimizations help generate a much more precise set of required system calls for Node.js applications during execution compared to traditional methods. Note that to adapt to the Node.js framework, HODOR distinguishes different system calls executed by different threads. HODOR then create the main-thread whitelist and the thread-pool whitelist respectively containing the identified necessary system calls based on the call graphs mapping. HODOR finally enforces efficient system call limitation on Node.js with the whitelists using the seccomp mechanism in the Linux kernel. Specifically, HODOR implements the restrictions at carefully selected moments to ensure that the protection of different threads does not affect each other.

We implement a prototype of HODOR, and evaluate the effectiveness and efficiency of HODOR in defending 168 real-world Node.js applications compromised by arbitrary code/command execution attacks. The results show that HODOR could reduce the attack surface to 19.42% on average with < 3% runtime overhead. Overall, we make the following contributions:

- We design and implement a system call level protection system HODOR for Node.js applications. HODOR could accurately infer the system calls used by the applications in the thread granularity at runtime and provide system call restrictions for different threads respectively.
- We propose several significant optimizations in the crosslanguage analysis to build more precise call graphs for identifying more accurate system call lists. Specifically, we present dynamic/static combined and new **server-side related mechanisms** to improve JavaScript call graphs construction, and new **partial context-sensitive mechanisms** to improve C/C++ call graphs construction.
- We evaluated HODOR by applying it to defend 168 realworld Node.js packages suffering from arbitrary code/command execution attacks. Experimental results confirm the effectiveness and efficiency of HODOR in shrinking the attack surface significantly (**80.57**% reduction on average) with negligible runtime overhead (i.e., <**3**%).
- We released the implementation, evaluation dataset and constructed attack payloads of HODOR² to facilitate future research in the Node.js attack and defense area.

2 BACKGROUND & MOTIVATION

2.1 Node.js Architecture

As shown in Figure 2, Node.js framework consists of builtin module layer, binding module layer, and dependency module layer. Compared with client-side JavaScript runtime, Node.js has two major important features: non-blocking and interaction with OS kernel. We utilize a Node.js application example in Figure 2 to illustrate the working mechanism of Node.js framework. Figure 2 is a code fragment of the version 0.1.0 of dns-sync [33] package. By utilizing the method exec of the builtin module child_process, it could execute the value of cmd parameter.

Builtin module Layer: Builtin module layer provides Node.js application with builtin modules [34]. Different from JavaScript runtime on browser-side, Node.js runtime provides more functionalities

²https://github.com/NodeHodor/Hodor



Figure 2: The overview of Node.js framework.

suitable for the server side, such as file system access, asynchronous communication and system command execution. Developers can further build packages on the basis of these builtin modules. Specifically, the application can use the require function to invoke builtin modules (Line 1), and these modules use module.exports or exports to export public API of a module (Line 11).

Binding Module Layer: The binding module layer encapsulates Node.js dependencies as binding modules for builtin modules to invoke. The binding module layer consists of several binding modules. When Node.js engine is running, RegisterBuiltinModules will traverse all the node_module structures and link them into a global linked list of Node.js, so that the binding modules can be accessed by the builtin module layer through the internalBinding method. Each binding module will implement the Initialize function (Line 17), and bind the C++ function implemented in this module with the corresponding builtin module layer JavaScript method name. The binding module name and the Initialize function will be stored in a node_module structure.

Dependency Module Layer: Node.js relies on the dependency module layer to achieve this functionality of builtin modules. The dependency module layer consists of several dependencies written in C/C++ code. For example, V8 provides Node.js with JavaScript bytecode execution environment, libuv provides Node.js with non-blocking i/o ability and OpenSSL provides Node.js with cryptographic operations. The Node.js application can then use the internalBinding function to utilize these dependencies. As shown at Line 5, the builtin modules import builtin module process_wrap through internalBinding function call.

Non-Blocking Ability: As Node.js is non-blocking, it can handle thousands of concurrent connections. Node.js runs in a single process

that creates two kinds of threads: *main thread* is used to execute the tasks on the event loop; *threads in the thread pool* are used to handle asynchronous I/O operations. Developers can handle asynchronous I/O operations by calling specific functions. For example, method readFile of the module fs is executed by the threads in the thread pool and it allows the program to read a file in a non-blocking asynchronous way. And the method readFileSync is executed by the main thread and it would block execution until finished. The thread pool would be loaded only when the asynchronous I/O methods are required. Therefore, Node.js applications can be divided into applications that use a thread pool and applications that do not use a thread pool.

2.2 Seccomp BPF

Seccomp BPF [35] provides a defense mechanism to limit the system call set available to a given application. It used the Berkeley Packet Filter language [36] to allow developers to implement system call filtering with configurable policies. It's worth noting that Seccomp filters could be applied to multi-threaded programs. The threads would inherit the filters of their parents when they are created. In this work, we aim to apply the Seccomp mechanism to limit the system call set of Node.js application and engine at the thread level, in order to implement a thread-aware defense mechanism.

2.3 Motivation

We use an example here to illustrate the motivation of our work. Figure 3 shows the growl function defined in growl library (v1.8.0) with more than three million weekly downloads [37]. In this example, users can use the growl function to create a notification message (Line 4). The message is passed to the exec method of child_process module (Line 15), and is executed as the parameter of notify-send command. As the message string is not validated and restricted, command injections can happen. Attackers can inject malicious commands into the message string such as "You have a mail`echo Hacker`", and the Hacker string is being printed. When the attacker is armed with the same permissions as the process that runs the Node.js application, s/he could expand the attack surface and perform critical operations. For example, she can read the sensitive file (e.g., "mail`cat /etc/passwd`"), create a remote connection (e.g., "mail`nc -l -e /bin/bash 8001`"), and change root permission (e.g., "mail`su root`"). Existing works restrict the attacker's ability at the JavaScript source code level, including application reducer [18] and context wrapper [20]. For attackers who have the same permissions as the process running the application, these protections are not effective.

A recent study [38] evaluated the packages in the npm repository and observed that 192,585(31.9%) packages in the npm repository do not need access to security relevant resources, such as file systems or network APIs. The inspiration is that **we do not need to grant a running Node.js application access to unnecessarily many system calls**. In this work, HODOR focuses on shrinking the ACE attack surface significantly by system call limitation. For the attack example described above, we identify that the application does not use other system calls except exec-related and notify-send-related system calls. When running the application, we only allow the application to access exec-related and notify-send-related system calls while

```
// main.js
var growl = require("growl");
1
 2
 3
    var message = 'You have mail!';
 4
    growl(message);
    // ./lib/growl.js
 6
 7
     exports = module.exports = growl; ←
 8
    var exec = require('child_process').exec
9
    cmd = { pkg: "notify-send" };
10
11
12
    function growl(msg, *) {
13
      args = [cmd.pkg];
14
      args.push(quote(msg));
exec(args.join(''),...);
15
   };
16
```

Figure 3: An example of ACE attacks.

disabling its use of file-related, permission-related, and networkrelated system calls. Therefore, even if an attacker has the ability to execute arbitrary commands, s/he cannot perform many critical operations as s/he wishes.

3 THREAT MODEL

We consider the runtime protection of a Node.js application which suffers from two popular types of highly risky attacks: • arbitrary command execution vulnerability where an attacker could execute arbitrary system command, • arbitrary code execution vulnerability where an attacker could execute arbitrary JavaScript code. In practice, the attackers can gain the ability of ACE attacks through injection attack [15], gadget chaining attacks (prototype pollution) [13, 14], and supply chain attack [11]. In general, the functionality of the malicious payloads may include file operation, permission modification, network interaction, and process operation, where a system call is highly likely to be invoked [28].

This work aims to shrink the attack surface of both arbitrary command execution and arbitrary code execution. More precisely, the considered attack surface is the sum of arbitrary code/command execution attack vectors, which can be explicitly quantified as the number of all system calls provided by the system. Our goal is to implement the principle of least privilege for a running Node.js application and minimize the system calls that an attacker can utilize. That is, we only grant a Node.js application access to the minimum set of system calls it requires to run properly. Implementing such system call restrictions does not need additional privileges. It is worth mentioning that our goal is not to prevent ACE attacks but to shrink the attack surface and minimize the critical operations caused by ACE attacks.

This work does not focus on the arbitrary command execution or arbitrary code execution vulnerabilities in the low layer of Node.js (i.e., binding module layer and dependency module layer). Race conditions [39], DOS attacks (such as regular expression DoS) [40, 41], hidden property abuse [42] in Node.js applications, and global variable alterations are also out of the scope of this work, as these attacks are usually triggered without invoking system calls. Installtime attacks [43] is out of the scope of this work since the attacks are triggered during the installation. Moreover, these can be addressed by complementary techniques [20, 39–44].

4 SYSTEM DESIGN

The goal of HODOR is to generate system call whitelists for Node.js applications and apply a runtime protection mechanism based on the whitelists to applications to shrink the attack surface of ACE attacks.

4.1 Overview

Figure 4 shows the overall framework of HODOR containing four main steps. **①** We first adopt the **call graph constructor** for both the source code of Node.js application and the Node.js engine. **②** Next, based on the call graphs, we **build mapping relationships** between the builtin module APIs and their system call list. **③** Then, by analyzing the call graphs of Node.js application, we identify the builtin modules used by the Node.js application. And based on the mappings between methods of builtin modules and system calls, we **generate whitelists** (of system calls) for the application. **④** Finally, we apply the **whitelist restriction** of system calls protection mechanism to the threads on which the Node.js application runs.

4.2 Call Graph Constructor

The system calls used by a Node.js application are encapsulated layer by layer in Node.js engine. In order to generate a system call lists for a Node.js application, we improve the precision of traditional call graphs constructors (for both JavaScript and C/C++ language) with multiple optimizations, and construct modular call graphs for both the application and the dependent Node.js engine.

JavaScript Language: For Node. is applications and builtin modules written in JavaScript, we build their modular call graphs respectively. The state-of-art JavaScript call graphs constructors [30, 31] do not fully consider the execution pattern of the methods in the builtin modules of JavaScript (e.g., Promise) and builtin modules of Node.js (e.g., fs), as the source code of builtin methods is not included in the applications. However, the execution of these functions triggers the invocation of functions or the creation of objects that appear in their arguments. For example, as shown in Figure 2 on Line 2, the execution of builtin method child_process.exec triggers the execution of the callback function in the last argument, and the execution of builtin method Function.binds creates a new function. Not considering such cases can result in a significant number of missing edges from the callback function invocation to the callback function definition of the function nodes creations. In this work, we summarize the execution pattern of the builtin methods provided by JavaScript [45] and Node.js [34] and add the function nodes and edges related to the function execution.

The highly dynamic nature of JavaScript code leads to unsound call graph construction [30]. For example, the JavaScript code can utilize dynamic constructors to create new functions represented as a string or execute JavaScript code represented as a string and can perform dynamic addition and deletion of object properties. These cannot be accurately analyzed by static call graph construction and lead to a significant number of missing edges from the function invocation represented as a string to the function definition represented as a string. In this work, we utilize dynamic call graph construction to identify the missing nodes and missing edges and generate combined static-dynamic call graphs for Node.js applications.



Figure 4: The HODOR pipeline.

C/C++ Language: When analyzing Node.js binding and dependency modules written in C/C++ language, the traditional contextinsensitive call graphs construction method brings considerable false positives. Using this result directly will make the subsequent syscall whitelist overly broad. However, conducting context-sensitive analysis for all the source code will bring extremely large memory and time overhead. Therefore, we propose a **partial context-sensitive** call graphs construction algorithm to handle the above challenges.

We summarize two code patterns leading to false positives and implement context-sensitive analysis for these patterns. The first pattern is context-insensitive for switch-case statement. Taking Figure 5 as an example, uv_fs_access (Line 3) and uv_fs_write (Line 7) call the same uv_fs_work function (Line 14). uv_fs_work implements diverse logic using switch-case statements and only one case-branch can be reached depending on the function parameters. The contextinsensitive algorithm considers all the branches of the switch-case statement to be reached, which causes false positives. We locate functions like uv_fs_work, collect the context information on each callsite, and eliminate unreachable case-branches of these functions. The second pattern is context-insensitive for function-pointer parameter. Taking Figure 6 as an example, Read(Line 1), Unlink and RMDir functions call Call (Line 20) functions through different call chains. The function pointer fn will be called finally, and its value changes depending on different call chains. The context-insensitive algorithm considers all the possible functions (like uv_fs_read, uv_fs_unlink and uv_fs_rmdir) to be called through fn, which causes false positive. We collect all the functions that call function pointers and perform a backward data-flow analysis to determine the assignment location of function pointers and record the call chains. We traverse the call chain and clone each function in it to make sure that every different value of the function pointer pass to different cloned functions (like Call).

4.3 Mappings Builder

After obtaining call graphs of each layer, we identify the function call relationships between layers at this stage. We finally generate a mapping set between builtin modules and system calls.

Builtin Module Layer: Starting with the builtin methods as entry points, we traverse the call graphs of builtin module layer and collect the binding methods used by builtin methods. As shown in Figure 2, we generate the mapping between builtin method exec method on Line 12 and the spawn method of binding method spawn on Line 9.

Binding Module Layer: Referring to the code pattern of registering the C++ modules and methods, we implement an LLVM Pass to collect the mappings between binding methods and C++ functions. As shown in Figure 2, we record the binding module process_wrap and its method spawn on Line 20. Furthermore, on Line 18, we generate the mapping between the method and C++ function Spawn defined on Line 14.

Dependency Module Layer: We generate the mapping between C/C++ functions and libc functions based on the call graphs of the binding module layer and dependency module layer. For example, we generate the mapping between C++ function uv_spawn on Line 24 and the libc function execvp on Line 28 in Figure 2.

Finally, combining the mappings between libc functions and system calls, with the mappings of three layers, we build the mappings between builtin modules and system calls.

4.4 Whitelist Generator

At this stage, we generate system call whitelists for applications. We first identify which builtin modules and builtin methods are used by the application at runtime based on the call graphs generated in Section 4.2. Furthermore, based on the mapping relationships generated in Section 4.3, we link the builtin modules and methods with their system call whitelist. As discussed in Section 2.1, the tasks of the main thread and the thread pool are different, and the required system calls are also different. Therefore, to achieve a thread granularity protection mechanism, we divide the system call whitelist into **main-thread-executed** whitelist and **thread-pool-executed** whitelist based on the code pattern of two kinds of threads.

4.5 HODOR Builder

We generate the corresponding system call filters based on system call whitelists and use the seccomp facility to restrict the actions available within the threads.

A challenge in loading the filters is the timing of loading the filters for the main thread and the child threads in the thread pool separately. Child threads in the thread pool will only be created by the main thread when the tasks arrive. Considering the inheritance feature of seccomp, i.e., the child threads in the thread pool inherit filters of the parent (main) thread when they are created by the parent (main) thread. However, the main thread and the thread pool perform different tasks, inherited filters will block the execution of child threads in the tread pool. To overcome this challenge, we design a fine-grained filtering mechanism, where for different types

of applications, we adopt different loading mechanisms. Especially, for thread pool required applications, we add the filters at two carefully selected moments. We illustrate the detail implementation in Section 5.5.1.

5 IMPLEMENTATION

In this section, we elaborate the core implementation details of HODOR to realize the design goals. To generate a more complete system call list for Node.js applications, we adopt static analysis and dynamic analysis for Node.js applications and Node.js engine.

5.1 JavaScript Call Graph Construction

We first perform static analysis for JavaScript code. Referring to JAM [30], we build modular call graphs for Node.js applications and Node.js builtin modules. The call graph constructor also considers JavaScript promise-chain [46] and execution pattern of builtin methods [34, 45]. JAM is the state-of-art modular call graphs building methods for Node.js applications. However, we could not access the source code of JAM. We reproduce the methodology of constructing modular JavaScript call graphs based on the rules of JAM.

Moreover, different from the previous work, our work also considers the execution pattern of builtin methods of Node.js engine and JavaScript. JAM [30] and js-callgraph [31] did not record the functions that act as a parameter of the builtin method. For Node.js builtin methods, functions can be executed as a callback function. For example, as shown in Figure 2, the anonymous function on Line 2 will be invoked after executing the command. In our work, we first manually collect 32 builtin methods whose parameters are callback functions and record the location of the callback function based on the Node.js API document [34]. Then, when building call graphs, we match these builtin methods and add the edge pointing from the domain where such a method is invoked to the point where the callback function is defined. The state-of-art call graphs building tools also omit the execution pattern of builtin methods of JavaScript. For example, when the function wrapped by the Promise object is finished, the callback function of then method will be invoked. And when a builtin method map of Array Object is invoked (e.g., [1,2,3].map $(x \Rightarrow x * 2)$, its first argument would be executed as a function. Moreover, different from builtin methods provided by Node.js engine, developers could also invoke builtin methods (e.g., Object.create) to create objects, which would also affect the construction of the call graphs. The method properties of the object may be invoked. In our work, we refer to the Standard built-in objects [45]. We collect ten builtin methods with function parameters including the methods of Promise and Array. When building call graphs, we match these builtin methods and add the edges pointing from the domain where such a method is invoked to the point where the callback function is defined. Moreover, we also record the nine builtin methods including the methods of Function and Objects and add nodes when these methods are invoked.

5.1.1 Dynamic Analysis Refiner. The highly dynamic nature of JavaScript leads to unsound call graph construction [30]. JavaScript code can use eval, new Function, and the method of vm to execute JavaScript code represented as a string and returns its completion value. For example, eval("console.log('hello')") prints hello string, and let sum = new Function('a', 'b', 'return a+b'); creates a function that sums two arguments. However, existing work generally does not analyze the string of these functions accurately, thus the definition of a new function or function call represented as a string cannot be included in the call graphs. To solve this deficiency, we introduce dynamic call graphs analysis. We use the dynamic call graphs tool Nodeprof [47]. We first collect the test suites from the packages installed by npm or github.com and build test suites for the application manually. Then we execute test suites and build dynamic call graphs for the application through the dynamic call graphs construction tool. Finally, we combine the nodes and edges of call graphs conducted by the static analysis and call graphs conducted by the dynamic analysis and generate the combined static-dynamic call graphs for the application.

5.1.2 Dynamic Command Execution. Node.js applications can utilize exec and spawn methods of child_process module to execute dynamic commands. For example, as shown in Figure3, notify-send command also requires system calls. However, existing protection work does not consider the system calls needed by the commands. Disallowing the execution of the system calls required by the commands can lead to program failure. In this work, we first identify the location of command execution methods at the static analysis phase. We then extract the commands executed by the command execution methods at the dynamic analysis phase. Then we use Linux strace utility [48] to record system calls utilized by the commands.

5.2 C/C++ Call Graph Construction

The C/C++ call graphs construction method of HODOR is based on SVF [32], which is a static program analysis tool for LLVMbased languages. We wrapped clang [49] with w11vm[50] to compile all the Node.js engine C/C++ source code to LLVM bitcode. By monitoring the command outputs of the compilation process, we utilize 11vm-1ink [51] to link all the source code of binding and dependency modules into a single LLVM bitcode file as our call graphs analysis target.

As described in Section 4.2, the traditional context-insensitive call graphs construction algorithm leads to considerable false positives for switch-case statements and function-pointer parameters. SVF conducts its default call graphs construction algorithm based on context-insensitive analysis. To balance precision and the overhead of context-sensitive analysis, we design a partial context-sensitive algorithm to eliminate false positive cases.

5.2.1 Unreachable Branch Elimination for Switch-Case Statement. We summarize the code patterns to locate functions for switch-case statement context-sensitive analysis: **①** There exists a switch-case statement in the function and different branches of the switch-case statement would call different functions; **②** The parameters of the function influence the branch selection variable.

For each call-site of functions that match the pattern: **①** Performing intra-procedural data-flow analysis to record the parameter or the member field offset within the structure parameter that propagates to the branch selection variable of the switch-case statement. **②** Performing backward data-flow analysis from the caller function to locate the assignment of the above variables. If the assigned constant can be determined, $\langle func_{caller}, func_{switch-case}, \rangle$

	1	<pre>#define INIT(subtype) req->fs_type = UV_FS_ ## subtype;</pre>
	3	<pre>int uv_fs_access(*) { INIT(ACCESS);</pre>
c	- 5	POST;
	6	}
1	7	int uv_fs_write(*) {
	8	INIT(WRITE);
1	- 9	POST;
11	10	}
11	11	
11	12	<pre>#define POST uv_fs_work(&req->work_req);</pre>
11	13	
1	14	<pre>static void uv_fs_work(struct uv_work* w) { uv_fs_work.uv_fs_access </pre>
	15	<pre>req = container_of(w, uv_fs_t, work_req);</pre>
	16	<pre>switch (req->fs_type) {</pre>
	17	X(ACCESS, access(req->path, req->flags)); +
	18	X(WRITE, uv fs write all(req));
	19	
	20	<pre>default: abort();</pre>
	21	}
	22	}

Figure 5: Switch-case statement context-insensitive.

constant_{switch-case}) would be recorded and func_{switch-case} would be cloned using LLVM. O For each cloned func_{switch-case}, case-branches except that belonging to constant_{switch-case} would be deleted.

For example, the call-site of uv_fs_work in uv_fs_access would be replaced to uv_fs_access.12 (12 is ACCESS's enum value).uv_fs_ access.12 contains only the access case-branch and other casebranchs like uv_fs_write_all would be deleted. Therefore, this kind of false positives would be eliminated.



Figure 6: Function-pointer parameter context-insensitive.

5.2.2 Function-Pointer Parameter Resolution. We summarize the code patterns to locate functions for function-pointer parameter context-sensitive analysis: **1** There exists more than one function pointer parameter in the function; **2** The function pointer parameter would be called within the function. In Figure 6, Call satisfies these code patterns.

For each call-site of functions that match the pattern: **①** HODOR performs inter-procedural and context-sensitive data-flow analysis for each call-site to trace the source of the function pointer object. **②** If the depth of the call chain that locates the function pointer object is 1, $\langle \text{func}_{caller}, \text{func}_{callee}, \text{func}_{fp} \rangle$ are recorded. func_{callee} is the function that contains function pointer parameters, and func_{fp} is the function pointer by the function pointer

parameter. We perform function clone on func_{callee}. The cloned function is named func_{callee}.func_{fp}, and the call-site in the cloned function is replaced to call the resolved function pointer. ⁽³⁾ If the call chain depth is greater than 1, $\langle \text{func}_{caller}, \text{func}_{callee}, * \rangle$ are recorded layer by layer backward until the function pointer object is determined to specific funcfp point. Then we perform function clone to the functions of entire call chain from top to down. func_{callee} will be cloned to func_{callee}.func_{fp} in each layer, and the call-site in the cloned function is replaced to invoke the corresponding modified callee function except the last layer, which is replaced to call the resolved function pointer. For example, AsyncCall is cloned to three versions: AsyncCall.uv_fs_read, AsyncCall.uv_fs_unlink and AsyncCall.uv_fs_rmdir. The similar logic is performed in AsyncDestCall, Dispatch and Call. Eventually uv_fs_read is called through the chain: Read \rightarrow AsyncCall. $uv_fs_read \rightarrow AsyncDestCall.uv_fs_read \rightarrow Dispatch.$ is resolved to uv_fs_read. Therefore, for each different value of the function pointer parameter, the call chain is different, which can eliminate the false positives.

We perform partial context-sensitive data-flow analysis to determine the assignment of the variables propagating to the branch selection variables and the source of the function pointer objects for the above two cases. The backward data-flow analysis starts with the branch selection variable or function pointer parameter, then it stops until the variable or the parameter is assigned with constants or actual function. When we achieve the endpoint of backward data-flow analysis, we perform a use-define chain analysis based on LLVM pass to determine whether the variable is assigned with one or more than one constant or actual function in the intra-procedural data flow analysis. When the variable is assigned with a single constant or actual function through different calling contexts, we perform context-sensitive analysis. In contrast, for the variable assigned with different constants or actual functions, we use context-insensitive analysis. As for the inter-procedural data-flow analysis, we additionally locate the starting point of the analysis at the callsite, record the context of different callsites, recursively perform intra-procedural analysis to complete inter-procedural data-flow analysis.

5.2.3 Libc Analysis. Almost all the dependency modules depend on libc library to interact with OS kernel. The popular glibc [52] library cannot be compiled into LLVM IR, thus we cannot utilize SVF to analyze it.

Previous works [28, 29] proposed methods for glibc analysis, but these analysis methods will introduce false positives and false negatives. temporal-specialization [28] utilizes GCC's RTL IR to analyze glibc. This method will lose some important system calls, such as CLONE system call in pthread_create and the EXECVE system call in system, etc.; It also connects some unrelated system calls to libc functions: read contains 41 system calls including VFORK, but actually the read function only contains one syscall READ. Saphire [29] utilizes objdump to generate the assembly of libc.so and analyzes the assemble text to collect system calls of libc functions. This method cannot recognize the CLONE system call in pthread_create and the EXECVE system call in system due to the lack of pointer analysis. Musl libc [53] is a lightweight libc library compatible with Node.js and can be compiled by LLVM. We replaced glibc with musl to construct more precise call graphs. Musl libc uses the inlineAsm statement to set the register numbers including the syscall number and then call the syscall instruction. We implement an LLVM pass to conduct data-flow analysis to identify each inlineAsm statement related to syscall and figure out the syscall number. Musl libc also encapsulates syscall in C function. We compiled Musl with -03 optimization to inline these functions at their call-sites, so that all the encapsulated functions are expanded to inlineAsm statement and our data-flow analysis can gain the correct result.

5.3 Building Mappings

In this stage, we build API mappings to infer the system call whitelist between builtin modules, binding modules, and dependency modules. Algorithm 1 illustrates how HODOR generates mappings of the three layers.

A	lgoi	rithm	1: N	Ларј	ping	generation
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	Data: Call graphs of builtin modules cg.builtin, call graphs of binding modules and
	dependencies cg_bottom, call graphs of libc cg.libc, LLVM IR of binding
	modules <i>ir_bind</i>
	Result: Output mapping dict M
1	$M.builtin \leftarrow \{\}; /* Mappings of builtin modules */$
2	forall $cg.Module \in cg_builtin$ do
3	forall method \in module.exports do
4	$M.builtin.module.method \leftarrow \{\};$
5	Callers C invoked by the method by traversing cg.Module;
6	forall $c \in C$ do
7	if c == internalBinding then
8	$M.builtin.module.method \leftarrow (module, method);$
9	$M.binding \leftarrow \{\};$ /* Mappings of binding modules */
10	forall $ir.Module \in ir_bind$ do
11	forall $method_bind \in ir.Module$ do
12	$M.binding.module.method \leftarrow func;$
13	$M.depend \leftarrow \{\};$ /* Mappings of dependencies */
14	forall $func \in M.bindings.module.method$ do
15	$M_depend.module.method \leftarrow \{\};$
16	Callers C invoked by the function by traversing cg.bottom;
17	forall $c \in C$ do
18	if $c == libc$ then
19	$M.depend.module.method \leftarrow libc;$
20	for all $libc \in M.depend.module.method$ do
21	$M.depend.module.method \leftarrow \{\};$
22	Callers C invoked by the function by traversing cg.libc;
23	forall $c \in C$ do
24	if $c \in syscall$ then
25	$M.depend.module.method.libc \leftarrow$ syscall;
26	return M.builtin, M.binding, M.depend;

5.3.1 Building Mappings for Builtin Modules. Builtin modules utilize module.export or export object to export the builtin modules to Node.js applications and utilize internalBinding method to import the binding modules (i.e., binding method). To map the builtin methods and binding methods, we traverse the call graph (i.e., cg.Module) of each builtin module (Line 2-12). The entry point of the call graph is the method (i.e., M.builtin.module.method) list in the module.export object or export object. When the method invokes a method of an object which is imported by the internalBinding method, we record the method as the method of a binding module (Line 8). The name of the binding module is the argument of the internalBinding, and the name of the binding method is the property of the binding module.

5.3.2 Building Mappings for Binding Modules. As illustrated in Section 2.1, node_module structure is used to register binding modules to the builtin module layer. We develop a LLVM Pass to locate and analyze node_module structures (Line 13-18). We first traverse all global variables in the LLVM IR (i.e., ir.Module) to locate of lib node and locate the node_module structures that are used to register binding modules by matching the variable names and structure types. For each node_module structure, we record the binding module (i.e., M. bindings.module) and the Initialize function. As Initialize function utilizes set-method functions (including SetMethod, SetMethodNoSideEffect, etc [54]) to bind the binding methods (i.e., M.bindings.module.method) and C++ functions defined in binding layer, we traverse the Initialize function and extract set-method functions (Line 16). By analyzing the parameters of the call-sites of the set-method functions, we record mappings between binding methods and C++ functions. Finally, we collect the binding modules and the mapping between the builtin methods and the C++ functions. And the binding modules can be used on builtin module layer.

5.3.3 Building Mappings for Dependency Modules. We identify the system calls utilized by the C++ functions collected in the previous stage (Line 20-25). We first traverse the call graph of dependency modules, starting with the C++ functions that map binding methods. Then we collect libc functions invoked by the C++ functions at Line 22. For libc functions, we collect the system calls utilized by the libc functions by traversing the libc library call graph. Finally, we collect the mappings between the C++ function and system calls.

Moreover, we also identify the builtin methods that submit tasks to the thread pool, and the tasks they submit. As the main thread can submit tasks to the thread pool via uv__work_submit or uv_queue_work function, we first transverse the call graphs of Node.js framework to identify which methods use the thread pool. Furthermore, as uv__work_submit passes the function pointer in the fourth parameter to the thread pool for execution, we collect the tasks that are submitted to the thread pool by analyzing the call-site of uv__work_submit.

5.4 Whitelist Generation

Algorithm 2 illustrates how HODOR generate the whitelist the for Node.js application. We need to identify the method used by the applications. Since the application use require method to load builtin modules and dependent packages, we first traverse the call graph of the application, and identify require method call (Line 3). We identify the builtin methods and modules (i.e., *M.bindings.b*), and generate system call list for the application by linking builtin methods with the mappings constructed in the Building Mappings stage (on Line 4-9). Furthermore, to provide a fine-grained protection, we divide the system call list into system call list of main thread and system call list of thread pool. In Section 5.3, we collect the builtin methods that utilize the thread pool (builtin_threadpool). When generating the system call, if the builtin method is in the builtin thread pool (i.e., builtin_threadpool (Line 9), we add the system calls used by the tasks (i.e., dependency function M.depend.f) in thread-pool whitelist (Line 10), otherwise we add the system calls to the main thread whitelist (Line 12). Finally, we get the main thread whitelist and thread-pool whitelist.

5.5 HODOR Building

5.5.1 Seccomp Implementation. We propose a system-call level protection mechanism HODOR. As discussed in Section 2.1, Node.js applications can be divided into the applications that use the thread-pool thread and applications that do not use the thread-pool thread. For different types of applications, we adopt different loading mechanisms. For thread pool required applications, we first install the filter for the thread-pool thread based on the thread-pool whitelist and then install the filter for the main thread to prevent the thread pool thread from inheriting the main thread filter. Specifically, before the program is loaded, the thread pool will be executed first. We utilize libseccomp [55] library and add the code to the load filter after thread pool initialization. Then, before the main thread loads the applications, we load the main thread filter by using the node-seccomp. node-seccomp is a Node.js package that wraps around the libseccomp C library [56]. For thread pool disrequired applications, we only load the main thread filter. Moreover, HODOR collects system calls, which we refer to as engine-required system calls. These system calls are required by the pure Node.js engine without Node.js applications running on it. We collect these system calls by the dynamic tracing (strace tool [48]), and we add them to the whitelists.

5.5.2 Read/write Permission Restrictions. The state-of-art protection mechanisms based on seccomp do not restrict read and write system calls due to the requirements of application and process [21-29]. However, the lack of fine-grained restriction of read and write system calls allows the attackers to read or write sensitive files. Due to the wide use of read and write system calls in the Node is engine (file operations or lower-level inter-thread communication), system call limitation mechanism can not shrink the attack surface related to these two system calls. In this work, we isolate the file system through chroot mechanism of linux to temporarily limit the root directory to the application directory when running the Node.js application. In this way, the attacker will not be able to modify the files outside the application directory for conducting corresponding attacks. Furthermore, we also switch the ownership of the files in the application directory to a highprivileged user and then set the files to read-only to prevent attackers from modifying the JavaScript code of the application. As the syscall for setting file permissions or switching the uid or gid are restricted

by HODOR in most cases, the ability of attackers to tamper files can be further limited.

5.5.3 Attack Surface Limitation. Next, we will introduce the ability of HODOR to reduce the attack surface of several kinds of popular real-world ACE attacks referring to threat list of Node.js [57, 58]. **O** Gadget chain attacks [13]. Prototype pollution attack is one of gadget chain attacks and is one of the most popular attacks of Node.js applications. The attacker gain the ability of ACE by polluting the arguments of eval or child_process.exec method. HODOR could mitigate gadget chain attacks in two aspects. If the application requires child_process.exec method, the attacker will not gain ACE ability as EXECVE system call is restricted. And if exec method or eval is allowed, HODOR protects the applications against the critical operations by system call limitation. **2** Injection attacks [15]. Injection attacks include command injection, template injection, and code injection [58]. An attacker can achieve arbitrary code execution by injecting malicious code. By system call limitations, some malicious injection code cannot be executed. ^(a) Improper file access attacks [59]. HODOR limits the file permissions that the application can access, and the attackers cannot read/write sensitive files. If the attacker can overwrite the permission allowed and executable files, the malicious code in the executable file cannot compromise the system further as HODOR applies system call limitation on the executable files. **4** Supply chain attacks [11]. Supply chain attacks happen when the dependency chain exists vulnerabilities [11]. HODOR only provides the system calls used by the application. Therefore, although the attackers could exploit the code to obtain the ability of ACE through obsolete packages of the dependency chain, s/he cannot do any critical operations s/he wants due to the system calls limitations.

5.5.4 Quantifying Attack Surface in Syscall Level. We quantify the extent of syscall-related attack surface reduction. The base permissions of the application are the number of all system calls provided by the system.

$$S_{base} = |SYSCALL_{system}| \tag{1}$$

We generate whitelists for the main thread and thread pool and HODOR implements system call restriction on main thread the threads of thread pool:

$$S_{app} = |SYSCALL_{main-thread}| \cup |SYSCALL_{thread-pool}|$$
(2)

Finally we can quantify the extent of attack surface reduction:

$$SR = \frac{S_{app}}{S_{base}} \tag{3}$$

6 EVALUATION

We extensively evaluate HODOR by answering the following four research questions.

- **RQ1:** Can HODOR construct **sufficiently more precise call graphs** to achieve fine-grained system call level protection?
- **RQ2:** Can HODOR effectively **reduce the privileges of attackers** with arbitrary code/command execution attack ability?
- RQ3: How does HODOR compare with state-of-the-art tools?
- **RQ4:** What is the **runtime overhead** of the protection mechanism provided by HODOR?



Figure 7: System call for Node.js applications (RQ1).

 Table 1: HODOR granularity at system call level and thread level (RQ1 ans RQ2).

Type	# of Package	Node.js	w/ Musl	Ho	dor
Type Arb Comm Exec Arb Code Exec Total Type Arb Comm Exec Arb Code Exec Total	" of I ackage	# of CS	# of TS	# of CS	# of TS
Arb Comm Exec	110	1,161	10,636	910	7,617
Arb Code Exec	58	243	3,352	110	2,294
Total	168	1,404	13,988	1,020	9,911
Туре	# of Package	# of MT	# of TP	# of MT	# of TP
Arb Comm Exec	110	11,797	2,958	8,444	1,212
Arb Code Exec	58	3,595	1,218	2,362	496
Total	168	15,392	4,176	10,806	1,708

CS: Critical syscalls invocation; TS: Trivial syscalls invocation; MT: Main Thread system calls invocation; TP: Thread Pool system calls invocation;

Our experiments were conducted on a machine running 64-bit Ubuntu 18.04 with 64 AMD Ryzen Threadripper 3970X 28-Core Processor and 256GB RAM. The call graph building component for JavaScript code amounts to about 3.5K LoC and the call graph building component for C/C++ code amounts to about 2K LoC. The runtime enforcement component amounts to about 100 LoC.

Result overview We use Eq.3, i.e., *SR*, as a criterion to quantify the attack surface reduction ability. The smaller is the criterion, the better protection (attack surface reduction) can HODOR achieve. As shown in Figure 7 and Table 1, HODOR grants an average of 66 system calls each for the 168 packages in our evaluated dataset. For 58 packages with arbitrary code execution, among the 335 system calls provided by Linux kernel (i.e., *S*_{base} is 335), the average *SR* is 12.37%. For 110 packages with arbitrary command execution, the average *SR* is 23.14%. on average, **HODOR can reduce the attack surface of Node.js applications to 19.42%**.

6.1 Dataset

We extensively collect those packages which are affected by arbitrary code execution or arbitrary command execution attacks from the GitHub Advisory Database [60] as follows. Firstly, we extracted the vulnerabilities with "arbitrary code execution" or "command execution" in the description. Then, we filtered out the vulnerabilities which are exploited from the client-side (e.g., Cross-Site Scripting attack). For the remaining vulnerabilities, we selected those readily available proof-of-concept exploits. In total, our experimental dataset consists of 169 packages. Among the 169 packages, HODOR crashes on one package. 58 packages suffer from code injection attacks , sandbox breakout attacks , file overwrite attacks , prototype pollution attacks , and template injection attacks . The left 111 packages suffer from arbitrary command execution attacks. The dataset fully covers the evaluated packages in MIR [20] and packages suffering

from arbitrary code execution in SecBench.js [61]. Table 5 in Appendix shows the detailed descriptions of the vulnerable packages. We utilize this dataset as experiment targets for **RQ1**, **RQ2**, **RQ3** and **RQ4**. To evaluate the **RQ1**, we also take three large-scale realworld applications as extra datasets: koa [62], express [63] and json-server [64], whose github stars were no less than 30,000. Furthermore, to extensively evaluate **RQ4** (Runtime Overhead), we take two extra datasets: Node.js core tests [65] and 4 well-known web frameworks including koa [62], fastify [66], express [63], and connect [67], whose weekly downloads were no less than 500,000.

6.2 Call Graph Construction and Resulting Protection (RQ1)

We evaluate the improvement of HODOR in terms of call graphs construction and the resulting protection.

Call Graphs Analysis We first evaluate the achieved precision improvement of our call graphs construction optimizations, which is critical to accurately identify the system calls utilize by the Node.js application. In total, for 168 Node.js applications, **O** Our static call graphs building method adds 3,555 edges that are triggered by the execution of builtin methods of JavaScript and 547 edges that are triggered by builtin methods of Node.js engine. In particular, 37 out of 168 packages (22.02%) invoke system calls that are triggered by the execution of builtin methods and would be undetectable without optimization. For example, npm-lockfile utilizes method exec of child_process module in the builtin objection Promise. A lack of analysis of them will lead to a too-restrictive whitelist, and the protection system will affect the normal operation of the application. 2 As for dynamic call graph building methods, we build five test suites for the tested application on average and build combined static-dynamic call graphs for the application. In total, the combined static-dynamic call graphs building method adds and 1,271 nodes and 2,474 edges. These nodes and edges are generated due to the dynamic nature of JavaScript code, which cannot be captured by static call graphs analysis. ⁽¹⁾ For C/C++ code in the Node.js engine, HODOR totally cloned and replaced 36 functions for switch-statement and 284 functions for function-pointer parameter to implement partial context-aware analysis.

For the optimization of static call graphs construction for JavaScript code, we compared our method with the state-of-the-art JavaScript call graphs construction tools JAM [30] and js-callgraph [31]. Since we could not access the source code of JAM, we reproduce their comparison experiments and utilize the results they have done on six Node.js packages. We measured the precision (measured by

comparing against dynamic call graphs building tool NodeProf [47]) according to JAM test method. As shown in Table 2, the result shows that HODOR finds that the coverage based on dynamically created edges is **92.39%**, compared to 79.77% for JAM and 52.49% for js-callgraph. The missing edges are triggered by the dynamic nature of JavaScript code. The results suggest that our optimizations for JavaScript code result in substantially higher coverage than the state-of-art JavaScript call graphs construction tools.

 Table 2: Experimental results for JavaScript call graph constructions (RQ1).

	HODOR	Jam	js-callgraph
makeappicon 1.2.2	95.02%	86.05%	/
npm-git-snapshot 0.1.1	86.63%	82.45%	43.86%
nodetree 0.0.3	83.33%	70.65%	/
jwtnoneify 1.0.1	93.43%	71.43%	/
npmgenerate 0.0.1	100%	97.42%	59.81%
smrti 1.0.3	97.29%	80.80%	66.20%
openbadges-issuer 0.4.0	91.05%	75.85%	40.08%
Average	92.39%	79.77%	52.49%

We choose 7 of 12 Node.js packages as NodeProf successfully builds call graph for these packages;

Recall: The coverage based on dynamically created edges;

For call graphs construction for C/C++ code, we compare our partial context-sensitive algorithm with state-of-art call graphs building tool SVF. Under the premise of running the program properly, we compared the size of the system call list generated by HODOR and SVF. In total, the size of the system call whitelist constructed by HODOR is 71.02% (10,931/15,392) of the size of the whitelist constructed by SVF. In other words, the optimization for C/C++ code significantly reduces the attack surface to 71.02% of SVF-built call graphs. Furthermore, we divided the 335 system calls into 17 critical system calls (in Table 10 in Appendix) and 318 trivial system calls referring to temporal-specialization [28]. The critical system calls include command-execution-related system calls, network-related system calls, and permission-related system calls³. These critical system calls are not many but can cause enormous consequences such as creating scheduled tasks and reversing shells. As shown in Table 1, for critical system calls, our method could reduce the whitelist to 72.65% (1,020/1,404) of whitelist generated by SVF-built call graphs. For trivial system calls, our method could reduce the whitelist to 70.85% (9,911/13,988) of whitelist generated by SVF-built call graphs. More details are listed in Table 5 in Appendix. These results show that our optimization for C/C++ code help generate more precise system call lists for better shrinking the attack surface of ACE attacks.

System Call Level Protection. For 58 packages with arbitrary code execution, among the 335 system calls provided by Linux kernel (i.e., S_{base} is 335), the average *SR* is 12.37For 110 packages with arbitrary command execution, the average *SR* is 23.14%. HODOR can shrink the attack surface to 19.42% of the Node, is applications.

We divided the system calls into engine-required system call type and app-required system call type. **Engine-required** system calls are a list of system calls required for the pure Node.js engine on which no application runs, and we record these system calls with the help of strace [48] utility. The engine-required system call set includes 28 system calls for the main thread and 15 system calls for the thread pool. The full list of the engine-required system calls is shown in Table 9 in Appendix. Other system calls are considered as apprequired system calls, which are only used when the application calls the methods of builtin modules. Among the 168 packages, we observed 129 of 168 (76.79%) packages use builtin modules and methods. Therefore, in addition to the engine-required system call, HODOR also needs to use the mappings between builtin modules and system calls to infer the app-required system calls. 39 of 168 (23.21%) packages do not utilize the builtin modules, that is, do not need app-required system calls. All of them are packages that are compromised by arbitrary code attacks. Since no additional builtin methods are used, HODOR can reduce the attack surface to the maximum extent, i.e., the attacker could only use 28 system calls.

Thread Level Protection. HODOR adds filters on both the main threads and the threads of thread pool. As shown in Table 1, for the main thread, the attack surface can be reduced to 19.20% by HODOR on average. Among the 129 builtin-method-needed packages, we observe 66 packages (51.16%) invoke the system calls that are executed in the thread pool. For these packages, HODOR also adds filters on the threads of the thread pool, and the attack surface thread pool can be reduced to 7.73% by HODOR on average. For example, HODOR can identify npm-lockfile package utilize method readFile, writeFile, stat, and mkdir belongs to the builtin module fs, which are accomplished by the thread pool. Armed with HODOR, attackers cannot perform thread pool related exploits apart from readFile, writeFile, stat, and mkdir.

Dynamic Command Execution We extract and analyze the command arguments of command execution methods (i.e., exec and spawn methods of child_process module) and add the system calls required by these commands into the system call list of the main thread. For 98 packages, we add 4,887 system calls to the system calls, including 61 critical system calls and 4,826 trivial system calls. Code Coverage and HODOR's Effectiveness. We reported the coverage of each of the evaluated benchmark in the test suite in the last three columns of Table 5 in the Appendix to show its effect on HODOR's effectiveness. The key observation here is that there is little correlation between the code coverage and the number of system calls in the whitelist of HODOR, i.e., the Pearson productmoment correlation coefficient between the package module covered line number and the whitelist size is -0.038, coefficient between the code coverage of package and the whitelist size is -0.027, and coefficient between the code coverage of package module and the whitelist size is -0.066 (more details in Figure 9 in the Appendix). The implication is that HODOR is able to precisely infer the system calls required regardless of the code coverage, which thus alone does not have a significant effect on HODOR's effectiveness.

HODOR's Effect on Benign Applications. We used the above test suites and three more real-world applications to evaluate HODOR's effect on the application's normal operation. Among the 169 packages, 168 packages run correctly, while HODOR crashed on the package extra-asciinema [68]. Note that as shown in Table 5 in Appendix, the average covered line of the package module is 69.79%, and the average cover line of the package including the dependent

³Note that read, write, and unlink are carefully addressed by limiting file permissions by HODOR as discussed in Section 5.5.2

module is 33.98%. The crashed package contains Python scripts, and there was an interaction between Python3 and JavaScript scripts at runtime. HODOR does not support to analyze Python scripts leading to the syscall list missing required by the Python scripts at runtime. Therefore, the package suffered from a too-restrictive list of system calls, causing it to crash with the protection of HODOR. For realworld applications testing, we deployed and accessed the websites. As shown in Table 3, the average covered line of the package module is 56.11%, and the average cover line of the package is 11.15%. The result shows that HODOR can be integrated into well-known applications and does not affect the normal functionalities.

Takeaway: The optimization of JavaScript call graph construction helps identify hidden required system calls for 23.21% packages. And the optimization in C/C++ level further reduces the system call permissions by 71.02% on average. They together enable HODOR to reduce the attack surface for the main thread to 19.20%, and reduce the attack surface of the thread pool thread to 7.73%, while not affecting the application's normal operation.

Table 3: Experimental results for correctness (RQ1).

	Stars	LoC	# Syscall	% CL-1	% CL-2	Pass Ratio
json-server	66.5K	258,094	61	53.36%	3.34%	32/32
express	60.6K	58,997	63	66.05%	26.12%	75/75
koa	33.9K	291,226	64	48.92%	3.98%	28/28
Average	53.7K	202,772	62.67	56.11%	11.15%	135/135
CL-1: Covered	line of the pac	skage module: C	I -2: Covered line	of the package:		

6.3 **Exploit Mitigation (RQ2)**

We further evaluate the effectiveness of HODOR from the exploition perspective. Traditional runtime protection work usually uses the attack payload provided by the proof of concept when evaluating their effectiveness. However, the payload provided by proof of concept is far from enough, as the operations performed by attackers in reality can be sophisticated and require system calls, while the proof of concept is generally simple, such as print 123 or execute 1s. Therefore, to extensively verify the effectiveness of our tool, we construct different advanced attack payloads to simulate various dangerous behaviors of attackers, where a variety of critical system calls can be invoked. Specifically, according to different vulnerability types, we construct seven payloads written in bash language for arbitrary code execution and seven payloads written in JavaScript language for arbitrary command execution. The details are shown in Table 8 in Appendix.

For the packages with arbitrary command execution, HODOR could successfully defend against 62.08% of the critical exploits and defend against 99.09% of real-world exploits. HODOR cannot block the execution of functions associated with arbitrary command execution (e.g., child_process.exec), but it can reduce the ability of the code that can be executed. For the packages with arbitrary code execution, HODOR could defend against all of the attacks with arbitrary code execution. On average, HODOR could effectively mitigate the execution of 91.63% critical exploits and defend against 51.72% of real-world exploits.

We demonstrate two case studies to further show the exploit mitigation capability of HODOR. O For arbitrary command execution vulnerabilities, attackers can inject shell code to achieve arbitrary code execution. This type of vulnerability utilizes the exec method or spawn method of the child process module. Such vulnerabilities utilize exec system calls, while due to the inheritance feature of seccomp, the system calls that exec can invoke are restricted. Package dns-sync use child_process.execSync method to execute the bash command. By command injections, the attackers could inject the bash code into the applications. HODOR limits the attack surface by applying the system call restrictions on the commands. Thus although the attacker can inject arbitrary command into the execSync method of child_process module, the critical exploits cannot execute successfully as HODOR disables the system calls required by critical exploits. 2 For arbitrary code execution vulnerabilities, the attacker can inject JavaScript code to achieve the purpose of arbitrary code execution. Package access-policy does not use builtin methods, and the vulnerabilities are caused by the parameter injection of the eval method. HODOR provide access-policy with minimum permissions. Therefore although the attacker can inject arbitrary JavaScript code into the eval method, the exploits cannot execute successfully as HODOR disables the system calls required by critical exploits.

Takeaway: HODOR could effectively mitigate the execution of 73.59% exploits.

Comparison with Other Techniques (RQ3) 6.4

Table 4: Comparison with state-of-art protection mechanisms.

Property	Mininode [18]	Mir [20]	HODOR
Security mechanism Runtime protection	Code debloating	Privilege reduction JavaScript level	Syscall limitation Kernel level
Source code change	Debloating	Function Closure	Non-change
Protection granularity	/	Module	Application

Qualitative Comparison. As shown in Table 4, we perform a qualitative comparison between HODOR and the state-of-art works [18, 20] for attack surface shrinking from four aspects. Security Mechanism. Mininode removes the unused code and MIR designs privilege reduction through context wrappers for the application, which both targets the JavaScript source code level (and may bring noticeable overhead). HODOR takes a complementary perspective to target the system call level for shrinking the attack surface. Mininode does not provide runtime protection. For attackers with ACE attack ability, s/he can still perform relevant critical operations. MIR provides runtime protections mainly at the JavaScript application level while HODOR provides runtime protection at the Kernel level via Syscall limitation. Source Code change. Mininode reduces the packages and MIR wraps the module with a function closure, which both need source code modification. Nevertheless, considering the dynamic nature of JavaScript, the source code level modifications might invalidate the protection mechanism and even disable the application to run properly. Protection Granularity. MIR is module-level protection where different module has different privilege. Mininode and HODOR are application-level protection.

Quantitative Comparison. We compare HODOR with state-of-art attack surface shrinking tool for Node.js applications, i.e., MIR, a permission inference and restriction system, which restricts the arbitrary code attack surface by adding a read-write-execute (RWX) permission model. MIR has two configurations, MIR-sa and MIR-da. MIR-sa generate default permissions for the package and MIR-da enforce the protection on the Node.js applications at runtime.

We utilize MIR-sa and MIR-da to run the 58 packages that suffer from arbitrary code attacks and check the permissions against the payload we constructed in Section 6.3. We manually audit the default permission generated by MIR-sa and run the tested application with MIR-da. MIR-sa failed to analyze 15 of 58 (25.86%) packages, as it can not generate proper permissions for these packages. Therefore, MIR-sa loads the most strict filter for them, which blocks their proper execution. MIR-da failed to run 17 of 58 (29.31%) packages. Among them, three failures were caused by program errors of MIR-da. 14 failures were caused by changes of MIR to the global variables of the applications. For the rest 41 packages, 99.39% of the critical exploits could be blocked under the permission models generated by MIR-sa. MIR-da introduced an overhead of 6.98%. Since MIR-da checks the permissions of the module every time methods or global variables are invoked. while the overhead of HODOR is 0.72% (more details in Sec. 6.5).

Takeaway: Compared to MIR and mininode, HODOR does not require source code modification and is capable of defending against a wider spectrum of attacks (additionally covering arbitrary command execution) with less runtime overhead.

6.5 Runtime Overhead (RQ4)

To evaluate the overhead introduced by libc library changes and HODOR additions, we ran these tests using Node.js in three states: Node.js with glibc, Node.js with Musl libc, and with HODOR Node.js (i.e. protected Node.js with Musl libc). Since Node.js is developed based on glibc libc, we standardize the results on when Node.js with glibc runs code.

For the Node.js core tests, the test case will run a statement containing a builtin method execution multiple times (200 in our experiments) and record the running time. Figure 8a shows that the total time of running the tests. The results show that the overhead introduced by libc library replacement is 1.27%, and the protection mechanism introduces 0.61% overhead further.

For the web frameworks of koa, fastify, express, and connect, we measure the overhead of HODOR by observing the response of the index page provided by the the example code in the framework tutorial. We conducted 2000 requests of index page visits and record the total response time of the server. As shown in Figure 8b, the replacement of libc library introduces 0.76% overhead on average and the protection mechanism introduces 2.80% overhead.

Figure 8c shows the runtime overhead of different attack type of the 168 packages. We find that libc replacement introduces 1.90% overhead on average, and the protection mechanism introduces 0.39% on average. Not surprisingly, we find packages that use fewest builtin methods (i.e., the packages effected by arbitrary code execution attack) introduce minimal overhead (0.22%), while packages

that use more system call related operations (i.e., the packages effected by arbitrary command execution attack) introduce higher overhead (0.42%).

Takeaway: On average, the runtime overhead of HODOR is 0.61% for Node.js core tests, 2.80% for the web framework, and 0.39% for all the 168 packages, which is in general acceptable.

7 RELATED WORK

System Call Limitation: Many studies reduced the attack surface by limiting the system call set that the attacker can invoke in other application domains including software [25, 43], android applications [69], container applications [24, 28], linux applications [27] and PKU-based memory isolation systems [70], etc. For instance, the closest and recent work to HODOR proposed by Wyss et al. [43] introduces Latch for mediating the install-time capabilities of npm packages. It generates a system call manifest of install script and enforces it to prevent undesirable install-time behavior. Bulekov et al. [29] proposed an automatic approach for generating and applying system-call limitations to interpreted PHP applications. They performed static and dynamic analysis to build call graphs over the interpreter for the binary of the PHP application and interpreter. Compared with HODOR, Saphire scanned the AST to identify all built-in function calls, built call graphs over the compiled binary from Debian repositories, and applied protection to the entire process, which are coarse-grained for Node.js applications. Moreover, in addition to including the system calls used by the application, the whitelist also preserved the system calls that Node.js invokes at runtime. Ghavamnia et al. [28] was tailored for server applications. They presented an approach that could restrict the system call set further based on the execution phase. Ghavamnia et al. [24] proposed a solution for automatically generating limited system call policies for Docker containers including server application containers. Their work is orthogonal as Node.js provides capabilities to create a web server. Nevertheless, the system calls invoked by Node, is are related to the application. What's more, Node.js is a multi-threaded application, which needs thread granularity filtered limitation.

Code Debloating, Privilege Reduction, and Security Isolation: In this line of research, some studies removed the unused API or unused code of the application. Koishybayev et al. [18] leveraged static analysis to remove the unused code and dependencies of Node.js applications. Azad et al. [71] obtained the dead code of PHP applications by using the by dynamic analysis. Snyder et al. [72] evaluated the Web API of modern browsers and proposed a Web-API access extension for client-side users. Qian et al. [73] utilized a hybrid approach to determine the bloated units of Chromium and removed them. In the comparison of our work, these technologies could prevent attackers from further exploiting malicious code. However, as mentioned in Section 6.4, these techniques cannot limit arbitrary code execution attacks. Others implement privilege reduction. Bittau et al. [74] presented the system Wedge to splitting the application into fine-grained, least-privilege compartments. Gudka et al. [75] proposed the Security-Oriented Analysis of Application Programs (SOAAP) that could create isolated components for complex applications to limit privileges leaked. Vasilakis et al. [20] focused on



Figure 8: Runtime overhead of Node.js core tests, web framework and applications under the protection of HODOR (RQ4).

the dynamic compromise on Node.js. They proposed a read-writeexecute (RWX) permission module MIR for the Node.js application. MIR is close and the most recent to our work, however, as shown in section 6.4, this work cannot reduce the attack surface of some arbitrary code execution attacks. Others implement security isolation mechanism. For instance, Vasilakis *et al.* [76] introduced BreakApp that transforms a module into its own compartment and enforces security policies on module communication automatically based on the users' configuration. While the policies are generated manually. Ahmadpanah *et al.* [77] performed security analysis on Node-RED applications and proposed sandbox system SandTrap for the applications. SandTrap can enforce fine-grained access control policies on third-party applications.

JavaScript Program Analysis: In recent years, researchers have developed static analyzers (such as JSAI [78], TAJS [79], WALA [80], and SAFE [81]) and dynamic analysis (such as Jalangi [82] and Nodeprof [47]) to understand behaviors of JavaScript programs and to detect their bugs in a sound manner. Analysis precision and performance are being increased. Andreasen et al. [83] presented a static dataflow analysis for JavaScript with high degrees of context sensitivity. Park et al. [84] presented Loop-Sensitive Analysis (LSA) approach to enhance the static analysis precision in loops. Stein et al. [85] presented a conventional non-relational static dataflow analysis with a value refinement mechanism to increase precision on demand at critical locations. Park et al. [86] proposed dynamic shortcuts to switch between abstract and concrete execution during JavaScript static analysis in a sound way. By employing more efficient static and dynamic analysis, we can generate more fine-grained system call whitelist for the Node.js applications.

Node.js Application Analysis: The past two years have seen many researches on Node.js security. Nielsen *et al.* [30] proposed an approach to build call graphs for Node.js applications, which can be used for security scanning. Li and Kang *et al.* [58] leveraged a static analysis to generate a graph structures, called Object Dependence Graph (ODG), for detecting Node.js vulnerabilities. staticu *et al.* [87] presented an automatically extracting taint specifications for Node.js applications. Xiao *et al.* [42] found the hidden property abusing (HPA) attack, and designed a tool to detect and verify the vulnerability. Compared to the previous work focusing on mining Node.js application vulnerabilities, our work aims to provide runtime protection for Node.js applications.

8 DISCUSSION & CONCLUSION

We now conclude by considering potential limitations of HODOR as well as opportunities for future work: • HODOR does not analyze

native extensions [88] of Node.js applications yet, as the evaluated applications do not use them. They will be considered for future through LLVM pass extensions (in Section 5.3.2). **@** HODOR provides application-level enforcement mechanisms that is more coarse-grained than module-level mechanisms (e.g., mir [20]). Moreover, HODOR cannot reduce the attack surface to zero due to system call requirements of pure Node.js engine. In the future, it can be combined with other protection mechanisms at different levels to complement each other. **@** As now, our static analysis for Node.js application only supports the module system CommonJS, which lacks portability on ES6 and AMD module systems. In the future, we will consider supporting ES6 and AMD module systems.

In this paper, we propose HODOR, a novel permission restriction system in the lower system call level for Node.js applications to shrink the attack surface. HODOR first generates fine-grained system call permissions for the Node.js applications, benefiting from multiple significant optimizations to improve state-of-the-art methods in call graph construction for both JavaScript code and C/C++ code. A more precise CFG can also benefit a broad range of applications including supply chain attack detection, program security analysis, code navigation and refactoring, etc. Then, based on the identified required permissions, HODOR creates the whitelists and implements the runtime restriction in thread granularity level with the seccomp mechanism. Armed with HODOR, we apply HODOR on 168 real-world Node.js applications suffering from arbitrary code/command execution attacks. Extensive experiments show that the attack surfaces of the vulnerable packages reduce to 19.42%, while introducing negligible runtime overhead, i.e., less than 3%.

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Table 5: HODOR granularity of packages at system call level and thread level (RQ1).

			Node.js wi	th Musl Libc	Ho	dor	Node.js wit	h Musl Libc	Ho	dor			
Attack Type	CVE	Package Name	# of CS	# of TS	# of CS	# of TS	# of MT	# of TP	# of MT	# of TP	# of CL	% of CL-1	% of CL-2
Arbitrary Command Injection	/ CVE 2021 22363	command-exists	10	95	7	74	105	87	80	34	47	74.60%	74.60%
Arbitrary Command Injection	CVE-2021-23360	killport‡	11	99	11	72	110	0	83	0	237	18.53%	95.83%
Arbitrary Command Injection Arbitrary Command Injection	CVE-2021-23356 CVE-2018-13797	kill-process-by-name macaddress	11	99 93	11	72 58	110	0	83 63	0	6 52	75.00%	75.00%
Arbitrary Command Injection	CVE-2022-25973	mc-kill-port	10	95	7	74	105	87	80	33	405	34.40%	72.72%
Arbitrary Command Injection Arbitrary Command Injection	CVE-2021-23377	onion-oled-js open	10	93 93	3	57 57	103	0	60 60	0	145	10.11% 47.82%	82.35% 47.82%
Arbitrary Command Injection	CVE-2018-3757	pdf-image‡	11	99	11	75	110	87	85	34	194	25.19%	86.15%
Arbitrary Command Injection Arbitrary Command Injection	CVE-2018-3746 CVE-2017-1000220	pdfinfojs nidusage‡	10	93 93	4	60 57	103	87	62 60	34 0	535 68	15.06%	100.00%
Arbitrary Command Injection	CVE-2021-23379	portkiller‡	10	93	3	57	103	0	60	0	12	88.23%	88.23%
Arbitrary Command Injection Arbitrary Command Injection	CVE-2021-23359 CVE-2021-23348	port-killer portprocesses:	10	95 93	7	72 57	105	0	79 60	0	8 19	100.00% 88.57%	100.00% 88.57%
Arbitrary Command Injection	CVE-2018-16460	ps	10	93	3	57	103	0	60	0	28	75.00%	75.00%
Arbitrary Command Injection Arbitrary Command Injection	CVE-2021-23355 CVE-2021-23374	ps-kill ps-visitor	11	99 99	11	72 72	110	0	83	0	3 29	12.50%	100.00%
Arbitrary Command Injection	CVE-2021-23380	roar-pidusage‡	10	93	3	57	103	0	60	0	62	50.81%	70.81%
Arbitrary Command Injection Arbitrary Command Injection	1	samsung-remote scp	11	99	11	75	110	0	83	54 0	14	56.52% 87.50%	87.50%
Arbitrary Command Injection	CVE-2018-3772	whereis‡	11	99	11	72	110	0	83	0	15	28.30%	83.33%
Arbitrary Command Injection Argument Injection	CVE-2021-23399 CVE-2022-24437	git-pull-or-clone	10	99	9	72	100	87	83	36	10	31.72%	92.85%
Command Injection	CVE-2020-7636	adb-driver	11	99	11	73	110	87	83	33	66 720	39.75%	91.30%
Command Injection Command Injection	CVE-2018-16462	apex-publish-static-files	10	95	9	73	105	0	82	54 0	8	2.18%	7.76%
Command Injection	CVE-2020-7633	apiconnect-cli-plugins‡	10	93	3	57	103	0	60	0	19,230	8.83%	11.43%
Command Injection	CVE-2021-3190 CVE-2020-7730	bestzip‡	11	99	11	82	110	87	92	35	4,204	12.60%	89.04%
Command Injection	CVE-2019-10807	blamer‡	10	93	3	57	103	0	60	0	1,465	13.88%	57.57%
Command Injection Command Injection	CVE-2020-7795 CVE-2020-7613	clamscan‡	10	95	6	56	105	87	57	35	2,493	8.16%	21.28%
Command Injection	/ CVE 2020 15122	cocos-utils	10	93	4	64 76	103	87	66	34	95	9.17%	9.17%
Command Injection	CVE-2020-7635	compass-compile	10	93	3	57	103	0	60	0	202	13.96%	95.65%
Command Injection	CVE-2020-7781TAB	connection-tester	10	95	9	72	105	0	81	0	43	68.25%	67.74%
Command Injection	CVE-2019-10/89 CVE-2020-28425	curljs	11	99	11	72	110	0	83	0	75	66.37%	78.12%
Command Injection	CVE-2020-28438	deferred-exec	11	99 03	11	72	110	0	83	0	1,339	9.93%	87.75%
Command Injection	CVE-2020-7631	diskusage-ng	10	99	11	72	110	0	83	0	23	13.93%	32.39%
Command Injection	CVE-2020-7606	docker-compose-remote-api	11	99	9	75	110	87	83	35	1,143	10.99%	88.88%
Command Injection	CVE-2013-10301 CVE-2021-26275	eslint-fixer‡	11	99	11	72	110	0	83	0	5,629	10.41%	66.66%
Command Injection	CVE-2021-23376 CVE-2021-23376	ffmpegdotjs ‡	11	99 00	11	73	110	0	84	0	43	41.74%	36.95%
Command Injection	/	find-process‡	10	93	3	57	103	0	60	0	110	5.56%	42.96%
Command Injection	/ CVF-2020-28429	freespace‡	11	101	11	80 57	112	0	91	0	27	44.26%	59.09%
Command Injection	CVE-2020-7630	git-add-remote	11	101	11	80	112	0	91	0	13	65.00%	65.00%
Command Injection Command Injection	CVE-2020-28434 CVE-2018-3785	gitblame git-dummy-commit	11	99 101	11	72	110	0	83	0	9 490	18.00%	75.00%
Command Injection	CVE-2019-10802	giting	11	99	11	73	110	87	83	34	197	20.02%	64.56%
Command Injection Command Injection	CVE-2022-1440	git-interface git-tags-remote:	10	93 99	9	60 72	103	0	69 83	0	94 236	68.11% 32.14%	68.11% 100.00%
Command Injection	CVE-2020-28436	google-cloudstorage-commands:	10	93	3	57	103	0	60	0	13	52.00%	52.00%
Command Injection Command Injection	CVE-2017-16042 CVE-2020-36650	growl gry	10	93 99	9	61 74	103	0 87	70 83	0 34	35 190	45.45% 59.00%	45.45% 88.23%
Command Injection	CVE-2020-7601	gulp-scss-lint‡	11	99	9	86	110	87	89	42	5,065	9.60%	40.54%
Command Injection Command Injection	CVE-2020-7607 CVE-2020-7605	gulp-styledocco gulp-tape	11	99 99	11	86 86	110	87	91	44 44	664 32	5.08%	84.21%
Command Injection	CVE-2020-28437	heroku-env	11	99	11	75	110	87	85	34	25	33.33%	42.00%
Command Injection Command Injection	CVE-2019-10788 CVE-2019-10787	im-resize	10	99	3	57	100	0	60	0	85	19.63%	85.72%
Command Injection	CVE-2020-7629	install-package‡	11	99	11	72	110	0	83	0	42	93.33%	93.33%
Command Injection	CVE-2020-8178 CVE-2021-23381	killing	10	95	11	72	110	0	83	0	38	9.01%	90.24%
Command Injection	CVE-2019-15609 CVE-2018-16461	kill-port-process	11	101	11	81	112	87 87	91	33	259	20.45%	56.75%
Command Injection	/	local-devices	10	93	8	79	103	87	66	57	37	37.37%	55.73%
Command Injection	CVE-2019-10783	lsof lycwed-spritesheetis	11	99 93	11	72	110	0 87	83 63	0	54	91.30% 34.02%	91.30% 63.54%
Command Injection	CVE-2020-7786	macfromip	11	99	11	75	110	87	85	34	39	45.88%	45.88%
Command Injection Command Injection	CVE-2020-28434 CVE-2019-10786	monorepo-build network-manager	10	95 95	7	72	105	0	79	0	7,810	7.58%	85.18% 93.00%
Command Injection	CVE-2019-15597	node-df	11	99	11	72	110	0	83	Õ	84	1.71%	91.30%
Command Injection Command Injection	CVE-2020-7627 CVE-2020-28433	node-key-sender: node-latex-pdf	11	99 99	11	72	110	0	83	0	93	77.50%	77.50%
Command Injection	CVE-2020-7632	node-mpv	10	95	7	72	105	0	79	Õ	36	14.63%	14.63%
Command Injection Command Injection	CVE-2020-7602 CVE-2020-7785	node-prompt-here node-ps	10	95 99	7	72	105	0	79	0	8 37	6.66% 75.51%	57.14% 75.51%
Command Injection	/	node-unrar	11	99	11	72	110	0	83	0	17	8.21%	100.00%
Command Injection Command Injection	CVE-2022-0841 CVE-2021-23375	npm-lockfile‡ psnode	10	93 99	3	57 72	103	0	60 83	0	1,963 63	6.12%	70.96% 35.13%
Command Injection	CVE-2020-7604	pulverizr	11	99	11	77	110	87	87	33	614	6.56%	87.02%
Command Injection	CVE-2021-24033 CVE-2019-10796	react-dev-utils	10	95 99	11	72	105	0	81	0	0,425 28	65.11%	65.00%
Command Injection	CVE-2019-10804	serial-number	11	99	11	73	110	87	83	33	45	57.69%	57.69%
Command Injection	CVE-2020-7621	strider-git strong-nginx-controller	10	99	3	57	103	87 0	60	0C 0	404 17,994	24.82% 6.37%	42.40% 49.70%
Command Injection	CVE-2020-28432	theme-core:	10	93	4	58	103	87	60	33	1,721	15.46%	81.35%
Command Injection	CVE-2020-7/84 CVE-2020-7628	umount	11	99 99	11	72	110	0	83	0	2,376 965	8.20%	71.73%
Command Injection	/ CVF_2020_28431	vboxmanage.js‡	11	99 99	11	75 72	110	87	85 83	34	55 31	25.82%	25.82%
Command Injection	CVE-2020-15362	wifiscanner	11	99	11	73	110	87	83	33	215	9.10%	83.33%

Command Injection	CVE-2020-28447	xopen±	1 11	99	11	72	110	0	83	0	9	90.00%	90.00%
Command Injection	1	xps	11	99	11	72	110	0	83	0	51	4.26%	54.54%
Remote Code Execution	CVE-2020-36378	aaptjs	11	101	11	87	112	0	98	0	457	13.00%	97.05%
Remote Code Execution	/	arpping	11	99	11	72	110	0	83	0	98	27.68%	78.40%
Remote Code Execution	CVE-2020-11079 CVE 2021 23632	dns-sync‡	10	101		83	102	8/	93	35	544	27.80%	97.22%
Remote Code Execution	CVE-2021-23032	git git_lib‡	10	95	3	57	103	0	60	0	048	11.35%	59.02%
Remote Code Execution	, ',	git-narse	10	93	4	60	103	87	62	34	704	2.96%	39.51%
Remote Code Execution	,	gity	11	99	11	72	110	0	83	0	33	13.25%	55.93%
Remote Code Execution	1	imagickal‡	11	99	11	75	110	87	85	35	1,183	13.13%	83.16%
Remote Code Execution	1	meta-git‡	10	93	3	57	103	0	60	0	355	8.32%	89.02%
Remote Code Execution	/	node-os-utils‡	10	95	7	72	105	0	79	0	172	47.77%	47.77%
Remote Code Execution	CVE-2020-7620	pomelo-monitor	10	93	3	58	103	0	61	0	90	86.53%	86.53%
Remote Shell Command Injection	CVE-2015-7982	gm‡	10	93	3	5/	103	0	60	0	6/5	49.52%	58.50%
Arbitrary Code Execution	CVE-2020-7729	grunt	11	98	5	67	109	87	70	33	6,747	7.69%	47.13%
Arbitrary Code Execution	/	is-my-json-valid	1	28	1	28	29	0	29	0	3,560	9.92%	93.84%
Arbitrary Code Execution	CVE-2020-7777	jsen	11	99	11	72	110	0	83	0	766	34.56%	95.15%
Arbitrary Code Execution	CVE-2020-7673	node-extend	1	28	1	28	29	0	29	0	24	82.75%	82.75%
Arbitrary Code Execution	CVE-2017-16082	pg	1	28	1	28	29	0	29	0	260	12.26%	27.54%
Arbitrary Code Execution	CVE-2020-7640	pixi-class	1	28	1	28	29	0	29	0	38	/9.16%	/9.16%
Arbitrary Code Execution	CVE-2022-0748	post-toader	0	83	1	40	89	0	41	0	900	8.04% 5.10%	09.23% 86.33%
Arbitrary Code Execution	CVE-2021-23389	total is	6	83	2	46	89	87	44	35	2 072	6.55%	6 55%
Arbitrary Code Execution	CVE-2021-23390	total4	10	94	5	72	104	87	75	38	1.874	6.79%	6.79%
Arbitrary Code Execution	CVE-2017-1001004	typed-function	1	28	1	28	29	0	29	0	397	74.62%	74.62%
Arbitrary Code Injection	1	kmc	6	83	1	45	89	0	46	0	2,267	6.02%	55.81%
Arbitrary Code Injection	1	marsdb	1	28	1	28	29	0	29	0	1,825	18.20%	54.06%
Arbitrary Code Injection	1	mixin-pro	1	28	1	28	29	0	29	0	78	82.10%	82.10%
Arbitrary Code Injection	/	m-log	1	28	1	28	29	0	29	0	289	0.88%	100.00%
Arbitrary Code Injection	1	mobile-icon-resizer	10	93	4	59	103	87	61	35	54	33.75%	65.51%
Arbitrary Code Injection		mock2easy	10	93	5	70	103	87	73	35	/	/	/
Arbitrary Code Injection		modjs	10	93	4	62	103	8/	64	34	167	2.11%	6.77%
Arbitrary Code Injection		modulify mol proto	0	83	1	40	20	0	41	0	1,155	1.97%	80.70%
Arbitrary Code Injection		mongoosemask	1	28	1	28	29	0	29	0	40.5	10.02%	67.56%
Arbitrary Code Injection	,	protois	1	28	i	28	29	0	29	0	132	10.09%	90.76%
Arbitrary Code Injection	CVE-2020-7660	serialize-javascript	1	28	1	28	29	õ	29	õ	71	5.10%	93.42%
Arbitrary File Overwrite	CVE-2021-32803	tar	6	83	3	56	89	87	57	48	7,830	11.69%	83.14%
Arbitrary File Write	CVE-2018-1002204	adm-zip‡	6	83	1	41	89	0	42	0	516	26.81%	26.84%
Code Execution	CVE-2017-5941	node-serialize	1	28	1	28	29	0	29	0	52	92.85%	92.85%
Code Injection	CVE-2022-25760	accesslog	6	83	1	43	89	0	44	0	95	41.48%	83.33%
Code Injection	CVE-2020-7674	access-policy	1	28	1	28	29	0	29	0	81	0.26%	96.42%
Code Injection	CVE-2021-21277	angular-expressions	1	28	1	28	29	0	29	0	/13	48.01%	48.01%
Code Injection	CVE-2020-7075 CVE 2018 3784	cu-messenger	1	20	1	28	29	0	29	0	374	50.77% 84.11%	90.09%
Code Injection	CVE-2018-5784 CVE-2019-15657	eslint-utils	6	83	2	50	89	87	48	35	436	8.92%	80.89%
Code Injection	CVE-2021-23639	front-matter	6	83	2	42	89	87	36	33	1.096	11.35%	80.95%
Code Injection	CVE-2020-6836	hot-formula-parser	1	28	1	28	29	0	29	0	1,650	6.25%	89.94%
Code Injection	1	js-yaml	1	28	1	28	29	0	29	0	1,753	19.62%	70.66%
Code Injection	CVE-2022-21122	metacalc	7	90	2	52	97	0	54	0	59	54.12%	86.95%
Code Injection	CVE-2019-5413	morgan	6	83	2	48	89	87	46	35	2,135	11.69%	81.48%
Code Injection	CVE-2022-25921	morgan-json	6	83	1	32	89	0	33	0	261	12.59%	94.11%
Code Injection	CVE-2020-7672	mosc.js	1	28	1	28	29	0	29	0	47	0.93%	88.67%
Code Injection	CVE-2020-7609 CVE 2016 10548	node-rules.js	1	28	1	28	29	0	29	0	939	40.57%	91.58%
Code Injection	CVE-2010-10348	thenify	1	28	1	28	29	0	29	0	50	56.81%	82 14%
Prototype Pollution	CVE-2020-7743	mathis	1	28	i	28	29	Ő	2.9	Ő	2.650	9.83%	8.52%
Prototype Pollution	CVE-2021-23594	realms-shim	6	83	1	38	89	0	39	Õ	1,118	3.59%	78.18%
Remote Code Execution	1	djv	1	28	1	28	29	0	29	0	300	33.44%	55.35%
Remote Code Execution	1	mongodb-query-parser	7	90	3	60	97	87	59	35	2,688	5.32%	28.49%
Remote Code Execution	CVE-2019-10758	mongo-express	6	83	2	43	89	87	42	33	9,225	11.46%	23.58%
Remote Code Execution	CVE-2020-24391	mongo-parse	1	28	1	28	29	0	29	0	342	83.82%	83.82%
Sandbox Breakout	/	notevil	1	28	1	28	29	0	29	0	2,828	9.68%	89.58%
Sandbox Breakout	CVE-2019-10/69	sater-eval	10	90	2	52	9/	0	54	22	119	35.31%	80.76%
Sandboy Breakout		static_eval	10	95	4		1105	87	82	33	2 801	40.00%	40.0070
Sandbox Breakout	í í	value-censorship	7	90	2	52	97	0	54	0	3.089	8.43%	95.00%
Sandbox Bypass	CVE-2019-10761	vm2	7	90	2	52	97	0	54	Ő	50	0.09%	53.33%
Sandbox Escape	CVE-2020-7710	safe-eval	6	83	1	38	89	0	39	0	11	95.15%	95.15%
Sandbox Escape	CVE-2020-7710	zhaoyao91-eval-in-vm	6	83	1	38	89	0	39	0	7	100.00%	100.00%
Template Injection	CVE-2022-29078	ejs	6	83	1	40	89	0	41	0	32	5.22%	87.33%
Template Injection	CVE-2021-23358	underscore	1	28	1	28	29	0	29	0	239	48.38%	48.38%
Ari	oitrary Command Executi	ion	1,161	10,636	910	7,617	11,797	2,958	8,444	1,212	936	35.61%	69.14%
	Arbitrary Code Execution		243	3.352	110	2.294	3.595	1.218	2.362	496	1.178	30.83%	71.03%
-	and a second consecution							-,			-,		

CS: Critical system calls invocation; TS: Trivial system calls invocation; MT: Main Thread system calls invocation; TP: Thread Pool system calls invocation; ‡: Packages that invoke system calls that are triggered by the execution of builtin methods; CL: Covered line num of the package; CL-1: Covered line of the package; CL-2: Covered line of the package;

					HOD	OR					
De la construcción de la	T. 14-1 A.M. J.		Ci	md	ad Permission				Net		
Package Name	Initial Attack	Initial Attack	Exec	Fork	Setgid	Setuid	Connect	Listen	Bind		
command-exists	Write command-exists	~	×	×	~	~	×	~	~		
kill-by-port	Write success	~	×	×	~	~	~	~	~		
killport	Write success		ž	×	~	~	~	~	~		
macaddress	Write /tmp/poof		Ŷ	ŝ	~	~	~	~	~		
mc-kill-port	Write newFile.txt	~	×	×	~	~	×	~	~		
onion-oled-js	Write success	~	×	×	~	~	~	~	~		
open ndf-image	Write /tmp/tada Write /tmp/backed	2	Ŷ	×	2	~	~	2	2		
pdfinfojs	Write a	~	×	×	~	~	~	~	~		
pidusage	Execute /usr/local/bin/python	~	×	×	~	~	~	~	~		
portkiller port killer	Write success Write success	~	×	×	~	~	ý	~	~		
portprocesses	Write success		Ŷ	ŝ	~	~	î	~	~		
ps	Write success.txt	~	×	×	~	~	~	~	~		
ps-kill	Write success	~	×	×	~	~	~	~	~		
ps-visitor roor-pidusage	Write success Write success	~	×	×	~	~	~	~	~		
samsung-remote	Write /tmp/malicious;		x	x	~	×	~	~	~		
scp	nc localhost 4444;	~	×	×	~	~	~	~	~		
whereis	Write /tmp/tada	~	×	×	~	~	~	~	~		
wincred sit pull or slope	Write success		×	×	~	~	č.	~	~		
adb-driver	Write a	2	Ŷ	ŝ	~	~	Ŷ	~	~		
alfred-workflow-nodejs	Write hacked	~	×	×	~	~	~	~	~		
apex-publish-static-files	Write apex-publish-static-files	~	×	×	~	~	×	~	~		
apiconnect-cli-plugins	Write Song	~	×	×	~	<i>.</i>	<i>.</i>	<i>.</i>	~		
async-git bestzin	Write HACKED # Write bestzin	2	Ŷ	×	ĉ	ĉ	2	ĉ	ĉ		
blamer	Write vulnerable	~	×	×	~	~	v	v	~		
cd-messenger	Write JHU	~	×	×	~	~	×	~	~		
clamscan	Write create.txt	~	×	×	~	~	×	~	~		
cocos-utils	Write hacked Write codecov	~	×	×	2	2	×	2	2		
compass-compile	Write JHU		Ŷ	ŝ	~	~	Ŷ	~	~		
connection-tester	Write 111	~	×	×	~	~	×	~	~		
curling	Write JHU	~	×	×	~	~	~	~	~		
curljs	Write JHU		ž	×	~	~	~	~	~		
diskstats	Write HACKED	2	Ŷ	ŝ	~	~	~	~	~		
diskusage-ng	Write Song	~	×	×	~	~	~	~	~		
docker-compose-remote-api	Write vulnerable.txt	~	×	×	~	~	×	×	×		
enpeem orlint fixor	Write create.txt Write colint fixer		×	×	~	~	~	~	~		
ffmpegdotis	Write success		Ŷ	ŝ	Ŷ	Ŷ	~	~	~		
ffmpeg-sdk	Write success	~	×	×	~	~	~	~	~		
find-process	Write /tmp/semicolon_file	~	×	×	~	~	~	~	~		
freespace	Write /tmp/semicolon_file	~	×	×	~	~	×	~	~		
git-add-remote	Write Song	2	Ŷ	ŝ	~	~	×	~	~		
gitblame	Write JHU	~	×	×	~	~	v	v	~		
git-dummy-commit	Write git-dummy-commit	~	×	×	~	~	×	~	~		
giting	Write create.txt	~	×	×	~	~	~	~	~		
git-interface git-tags-remote	Write /tmp/pwned Write /tmp/command-injection test	2	Ŷ	×	2	~	~	2	2		
google-cloudstorage-commands	Write JHU	~	x	x	~	~	~	~	~		
growl	Write aaaa	~	×	×	~	~	~	~	~		
gry	Write HACKED		ž	×	~	~	ý	Č.	<i>.</i>		
guip-scss-lint guip-styledocco	Write Create.txt Write Vulnerable	2	Ŷ	×	2	~	×	x	×		
gulp-tape	Write JHU.txt	~	×	×	×	×	×	×	×		
heroku-env	Write JHU	~	×	×	×	×	~	~	~		
im-metadata	Write im-metadata		×	×	×	×	×	×	×		
install-nackage	Write Song	2	Ŷ	ŝ	~	~	~	~	~		
jison	Write pwned	~	×	×	~	~	~	~	~		
killing	Write success	~	×	×	~	~	~	~	~		
kill-port-process	Write kill-port-process		ž	×	~	~	×	~	~		
local-devices	Makek directory attacker	2	Ŷ	ŝ	~	~	~	~	~		
lsof	Write create.txt	~	×	×	×	×	v	v	~		
lycwed-spritesheetjs	Write 111233 #	~	×	×	~	~	~	~	~		
macfromip	Write JHU2	~	×	×	~	×	<i>.</i>	~	~		
network-manager	Write JHU Write create.txt	2	Ŷ	x	~	~	x	2	2		
node-df	Write HACKED	~	×	×	~	~	v	v	~		
node-key-sender	Write Song	~	×	×	~	~	~	~	~		
node-latex-pdf	Write JHU		ž	×	×	×	×	×	×		
node-mpv	Write create txt		Ŷ	Ŷ	~		ŝ	2	2		
node-ps	Write JHU	~	×	×	~	~	v	~	~		
node-unrar	Write node-unrar	~	×	×	~	~	~	~	~		
npm-lockfile	Write rce	~	×	×	~	~	~	~	~		
psnode pulverize	Write Success Write Song	2	×	×	ž	~	ř	~	~		
react-dev-utils	Write react-dev-utils	-	x	x	~	~	×	v	~		
rpi	Write vulnerable.txt	~	×	×	~	~	~	~	~		
serial-number	Write create.txt	~	×	×	×	×	~	~	~		
strider-git strong-nginy controllor	Write HACKED; Write Song		×	×	×	×	×	×	×		
theme-core	Write JHU	-	ŝ	ŝ	~	~	~	ž	~		
ts-process-promises	Write JHU	~	×	×	~	~	V	~	~		
umount	Write Song	· .	×	×	×	×	<i>v</i> .	×.	~		
vboxmanage.js	Write HACKED	~	x	×	~	~	~	~	~		

Table 6: Exploit execution for packages with arbitrary command execution attacks.

wc-cmd	Write JHU	~	×	×	~	~	~	~	
wifiscanner	Write /tmp/exploit.txt	~	×	×	~	~	~	~	
xopen	Write JHU	~	×	×	~	~	~	~	
xps	Write HACKED	~	×	×	~	~	~	~	
aaptjs	Write aaptjs	~	×	×	×	×	×	×	
arpping	Write HACKED	~	×	×	~	~	~	~	
dns-sync	Write pwned	~	×	×	~	~	×	~	
git	date	×	×	×	~	~	×	~	
git-lib	Write HACKED;	~	×	×	~	~	~	~	
git-parse	Write HACKED	~	×	×	~	~	~	~	
gity	Write HACKED	~	×	×	~	~	~	~	
imagickal	Write HACKED	~	×	×	~	~	~	~	
meta-git	Write HACKED	~	×	×	~	~	~	~	
node-os-utils	Write DUMMY_FILE	~	×	×	~	~	×	~	
pomelo-monitor	Write Song	~	×	×	~	~	~	~	
gm	Write gm	~	×	×	~	~	~	~	

★: Exploits are executed; ✓: Exploits are blocked;

Table 7: Exploit execution for packages with arbitrary code execution attacks.

	HODOR							MIR									
	T-14-1 444-1		Cmd Permission Net				Cmd Permission Net										
Package Name	Initial Attack	Initial Attack	Exec	Fork	Setgid	Setuid	Connect	Listen	Bind	Initial Attack	Exec	Fork	Setgid	Setuid	Connect	Listen	Bind
accesslog	Print xSS	×	~	~	~	~	~	~	~				/				
access-policy	Print 123	×	~	~	~	~	~	~	~	· ·	~	~	~	~	~	~	~
adm-zip	Path traversal	~	~	~	~	~	~	~	~				/				
angular-expressions	Write file angular-expressions-success	~	~	~	~	<i>.</i>	~	~	~				. /				
cd-messenger	Print JHU	×															
cryop	Print defconrussia	×									~	~	· ·	~	~	~	~
ajv	touch HACKED Write file ais suggests																
ejs oslint utils	Write file estint utile success																
front-matter	Print 1	×		2						•			• ,	•	•	•	•
grunt	Returns Date now	2	x	×		2		2	2	· ·	~	~	v '	~	~	~	~
hot-formula-parser	Write file test		2	2		2		2	2		2	2	2				
is-my-ison-yalid	Execute cat /etc/nasswd	~	, v	v	~	~	~	, v	v		•	•	• /	•	•	•	•
isen	Write file malicious	~	x	×	×	x	×	x	×								
is-vaml	Returns Date.now	~	v	v	v	v	~	V	V								
kmc	Write file kmc-success	~	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
marsdb	Write file marsdb-success	~	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
metacalc	Print process	×	~	~	~	~	~	~	~				/				
mixin-pro	Print hacked	×	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
m-log	Print injected	×	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
mock2easy	Write mock2easy-success	×	~	~	~	~	×	×	×	 ✓ 	~	~	~	~	~	~	~
modjs	Write modjs-success.txt	×	×	×	~	~	×	×	×	 ✓ 	~	~	~	~	~	~	~
modulify	Print hacked	~	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
mol-proto	Write file mol-proto-success	×	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
mongodb-query-parser	touch test-file	~	~	~	~	~	×	×	×				/				
mongo-express	exec calculator	~	~	~	~	~	~	~	~				/				
mongoosemask	Print "my evil code was run"	×	~	~	~	~	~	~	~				/				
mongo-parse	Write file hacked	~	~	~	~	~	~	~	~	· ·	~	~	~	~	~	~	~
morgan	Write file mongui-success	×	~	~	~	~	~	~	~	· ·	~	~	~	~	~	~	~
morgan-json	Print GLOBAL CTF HIT	~	~	~	~	~	~	~	~	~	~	~	~	~	~	~	~
mosc.js	Write file Song	~	~	~	<i>v</i> .	~	~	~	~	· ·	~	<i>v</i> .	<i>v</i>	<i>v</i> .	~	<i>v</i> .	~
node-extend	Print 123	×	<i>.</i>	<i>v</i>	<i>.</i>	<i>.</i>	~	<i>v</i>	~		<i>v</i> .	<i>v</i> .	~	<i>v</i>	~	· ·	~
node-extend	Print 123	×															
node-rules.js	Print 123	*															
node-serialize	Execute is																
notevii	Print pwned	, in the second s															
Pg nint along	Print process.env	÷															
pixi-class	Print ree	2															
post-toauer	Write file protois success																
realme_chim	Messed with Object toString			2				2									
reduce-css-cale	Read /etc/passwd			2				2									
safe-eval	Return proces		2	2		2		2	2		2	2	2				2
safer-eval	Print id	×	2	2		2		2	2		2	2	2				2
sandbox	Print process pid	×	x	x	2		2	2	2		x	x	~	2	2	2	2
serialize-javascrint	Print 1	×	2	~	2		2	2	2		2	2	~	2	2	2	2
serialize-to-is	Execute Is	×	~	~	~	~	~	~	~	· ·	~	~	~	~	~	~	~
static-eval	Print hacked	~	×	×	~	~	~	~	~	-	~	~	~	~	~	~	~
tar	Overwrite file	~	~	~	~	~	~	~	~	-	~	~	~	~	~	~	~
thenify	Write file Song	~	~	~	~	~	~	~	~				1				
total.js	Touch HACKED	×	~	~	~	~	×	×	×				/				
total4	Touch HACKED	~	×	×	~	~	×	×	×				/				
typed-function	Execute whoami	~	~	~	~	~	~	~	~				/				
underscore	touch HELLO	×	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
value-censorship	Access the Function constructor	~	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
vm2	return process.env	×	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
zhaoyao91-eval-in-vm	return process.env	×	~	~	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~
mahila isan madaan	Drint hookod	×	×	×	~	~	~	~	~	 ✓ 	~	~	~	~	~	~	~

★: Exploits are executed; ✓: Exploits are blocked;

Table 8: Vulnerability payloads.

Critical Sys	scall	JavaScript	C*		
Cmd Execution	exec	child_process.exec	execve		
Cinu Execution	fork	child_process.fork	fork		
Permission	setgit	process.setuid	setuid		
1 crimission	setuid	process.setgid	setgid		
	bind	net.connect	bind		
Network	connect	dgram.createSocket.bind	connect		
	listen	server.listen	listen		

*: We compile the C code into binary and execute the binary as a shell.

Table 9: Engine-required system calls.

Thread Type	Engine-required System Call
Main Thread	mprotect; futex; rt_sigaction; munmap; read; fstat; getpid; open; ioctl; rt_sigprocmask; stat; fcntl; writev; epoll_pwait; pread64; dup3; close; write; getcwd; getdents64; rt_sigreturn; brk; shutdown; statx; readlink; madvise; exit_group; epoll_ctl; mmap;
Thread Pool	read; futex; openat; socket; statx; open; exit; close; write; rt_sigprocmask; fcntl; getcwd; madvise; munmap; mmap;

Table 10: Critical system calls.





Figure 9: The relationships between line coverage and syscall number.