

From Noise to Signal: Precisely Identify Affected Packages of Known Vulnerabilities in npm Ecosystem

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The npm Supply Chain Security Challenge

Background: npm is The world's largest software ecosystem, 3+ million packages

The Problem: *Complex Dependency Chains*

- Deep nesting, intricate dependencies
- Single vulnerability → massive downstream impact
- Example: pac-resolver CVE → 285,000+ GitHub repos at risk

Current Reality:

- ~25% of package versions depend on vulnerable packages

Current State - The Alert Fatigue Crisis

Package-Level Analysis:

- SCA tools (npm audit, Dependabot) report vulnerable dependencies
- Developers face critical question: *"Am I really affected?"*

The Alert Fatigue Crisis:

- 95% of vulnerabilities have fixes available
- Yet 80% of enterprise dependencies remain unpatched for >1 year
- Root cause: "Patching paralysis" from alert overload
- Uncertainty about true impact → costly update process

Package-Level Analysis Limitation

The Current Standard: Package-Level Analysis

- SCA tools report vulnerable dependencies present in dependency graph

Critical Limitation: Coarse-grained analysis

- Presence \neq Reachability
- Cannot determine if vulnerable code is actually called
- Leads to massive false positive rates

Why Not Function-Level Analysis? (Scalability)

The Logical Solution: *Function-Level Reachability*

- Determine if vulnerable code is actually called
- This is the necessary first step before *exploitability*

Challenge 1: Scalability

- Existing approaches face critical computational barriers
- Whole-program analysis is computationally prohibitive
- **Jelly/JAM:** Only 37% success rate under 4GB memory limit
- Must re-analyze entire dependency graph for each project

Why Not Function-Level Analysis? (Dynamic Nature)

Challenge 2: JavaScript's Dynamic Nature

- First-class functions: passed as arguments, returned from functions
- Callbacks and higher-order functions obscure call targets
- Dynamic property access: `obj[variable]()` → statically undecidable

```
1 // Dynamic property access
2 obj[propName](); // propName is a variable, static analysis cannot determine the call target
3
4 // Higher-order function
5 function process(callback) {
6   callback(); // Unknown which function is passed in
7 }
8
9 // Dynamic prototype chain modification increases analysis uncertainty
10 Object.prototype.newMethod = function() { ... }; javajavasc
```

Impact: Simple syntactic analysis is highly imprecise

Why Not Function-Level Analysis? (Module Systems)

Challenge 3: JavaScript's Module Systems

- CommonJS (CJS): `require()`, `module.exports`
 - Mutable exports modified at runtime
 - Dynamic require expressions: `require(variable)`
- ECMAScript Modules (ESM): static imports
- **Challenge:** CJS + ESM interoperability

Our Goal: Precise function-level analysis at *ecosystem scale*

Methodology - Key Insight

Key Insight: *npm package versions are immutable*

- Once published, (package, version) pair never changes
- **Opportunity:** Pre-compute once, reuse many times
- **Our Strategy:** "Analyze-once, reuse-many-times" model
- Enables *ecosystem-scale* analysis with practical performance

VulTracer Overview

- **Key Insight:** *Pre-compute once, reuse many times*

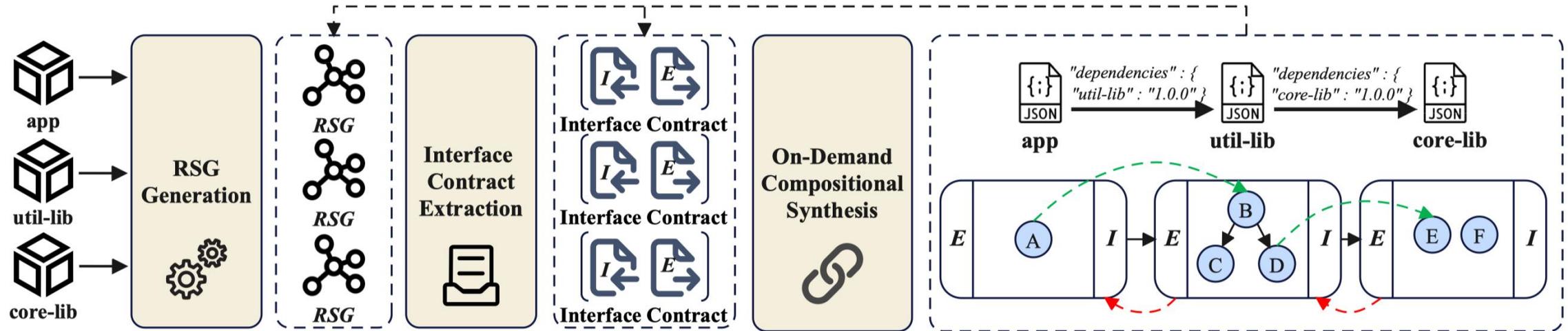


Fig. 1: Overview of VULTRACER.

Phase 1 - Rich Semantic Graph (RSG)

Rich Semantic Graph (RSG): Beyond Standard Call Graphs

Traditional CG Problem:

- function A() → function B()
- Loses boundary information needed for composition
- Cannot model external dependencies or public APIs

RSG Solution: *Reifies boundaries as vertices:*

Vertex Types:

- Programmatic entities (modules, functions)
- Invocation points (reified as vertices)
- Export anchors (public API)

Edge Types:

- Lexical nesting (contains)
- Call resolution (internal/external)
- Export resolution (standard/reference)

Result: Self-contained, composable package representations

DEF 1: Formal structure of RSG

$G = (V, E)$ where for a given package P :

The vertex set: $V \triangleq V_{\text{ent}} \cup V_{\text{invk}} \cup V_{\text{export}}$

• $V_{\text{ent}} \triangleq V_{\text{mod}} \cup V_{\text{func}}$

* $V_{\text{mod}} \triangleq V_{\text{mod_int}} \cup V_{\text{mod_ext}}$, where:

– $V_{\text{mod_int}} \triangleq \{v \mid v \text{ is a module defined in } P\}$

– $V_{\text{mod_ext}} \triangleq \{v \mid v \text{ is a module imported by } P\}$

* $V_{\text{func}} \triangleq V_{\text{func_int}} \cup V_{\text{func_ext}}$, where:

– $V_{\text{func_int}} \triangleq \{v \mid v \text{ is a function defined in } P\}$

– $V_{\text{func_ext}} \triangleq \{v \mid v \text{ is a function imported by } P\}$

• $V_{\text{invk}} \triangleq V_{\text{invk_int}} \cup V_{\text{invk_ext}}$, where:

* $\rho : V_{\text{invk}} \rightarrow V_{\text{ent}}$

* $V_{\text{invk_int}} \triangleq \{v \in V_{\text{invk}} \mid \rho(v) \in V_{\text{mod_int}} \cup V_{\text{func_int}}\}$

* $V_{\text{invk_ext}} \triangleq \{v \in V_{\text{invk}} \mid \rho(v) \in V_{\text{mod_ext}} \cup V_{\text{func_ext}}\}$

• $V_{\text{export}} \triangleq \{v \mid v \text{ is an export anchor in } P\}$

The edge set: $E \triangleq E_{\text{call}} \cup E_{\text{export}} \cup E_{\text{contains}}$

• $E_{\text{call}} \triangleq \{(v, \rho(v)) \mid v \in V_{\text{invk}}\}$

* $E_{\text{int_call}} \triangleq \{(v, \rho(v)) \in E_{\text{call}} \mid \rho(v) \in V_{\text{mod_int}} \cup V_{\text{func_int}}\}$

* $E_{\text{ext_call}} \triangleq \{(v, \rho(v)) \in E_{\text{call}} \mid \rho(v) \in V_{\text{mod_ext}} \cup V_{\text{func_ext}}\}$

• $E_{\text{export}} \triangleq E_{\text{std_exp}} \cup E_{\text{ref_exp}}$

* $E_{\text{std_exp}} \subseteq V_{\text{export}} \times (V_{\text{mod_int}} \cup V_{\text{func_int}})$

* $E_{\text{ref_exp}} \subseteq V_{\text{export}} \times (V_{\text{mod_ext}} \cup V_{\text{func_ext}})$

• $E_{\text{contains}} \triangleq \{(v, u) \mid v \in V_{\text{ent}}, u \in V, \text{contains}(v, u)\}$

• $\text{contains} \subseteq V \times V$, denotes the lexical-nesting relation in AST.

Phase 2 - Interface Contracts

Formal Abstraction for Composition: Interface Contracts

Contract = \langle Export Manifold, Import Manifest \rangle

Export Manifold (\mathcal{M}_E):

- API Path \rightarrow Set of Functions
- Example: \langle moduleExport, getMember('process') $\rangle \rightarrow \{\text{process_func}\}$
- Example: \langle moduleExport, getMember('parser') $\rangle \rightarrow \{\text{external_parse_func}\}$

Import Manifest (\mathcal{M}_I):

- Use Path \rightarrow Set of Invocation Points
- Example: \langle moduleImport('util-lib'), getMember('process') $\rangle \rightarrow \{\text{invk_node_1}\}$
- Catalogs external dependencies

DEF 2: Interface Contract

$C(P) \triangleq \langle \mathcal{M}_E, \mathcal{M}_I \rangle$ where:

- Export Manifold (\mathcal{M}_E) : $\Pi_{\text{def}} \rightarrow 2^{V_{\text{func_int}} \cup V_{\text{func_ext}}}$
- Import Manifest (\mathcal{M}_I) : $\Pi_{\text{use}} \rightarrow 2^{V_{\text{invk_ext}}}$
- $\Pi_{\text{def}} \triangleq \{p \in \text{Op}^+ \mid p[0] = \text{moduleExport}\}$
- $\Pi_{\text{use}} \triangleq \{p \in \text{Op}^+ \mid p[0] = \text{moduleImport}(\text{pkg})\}$
- $\text{Op}^+ \triangleq \{\langle op_1, \dots, op_n \rangle \mid op_n \in \text{Op}\}$ where:
 - * $\text{Op} \triangleq \{\text{moduleImport}(\text{pkg}), \text{moduleExport},$
 $\text{getMember}(\text{prop}), \text{getReturn}(),$
 $\text{getParameter}(\text{idx}), \text{getInstance}()\}$

Phase 2 - Interface Contracts

Benefits:

- Machine-readable API specification
- Enables semantic matching during composition
- Decouples implementation from interface

DEF 2: Interface Contract

$C(P) \triangleq \langle \mathcal{M}_E, \mathcal{M}_I \rangle$ where:

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 $\text{getParameter}(\text{idx}), \text{getInstance}()\}$

Phase 3 - Compositional Synthesis

Building Ecosystem-Scale Call Graphs: *Compositional Synthesis*

Strategy:

1. Topological sort of dependency graph
2. Bottom-up processing: leaves first
3. For each package:
 - Load pre-computed RSG from cache
 - Compose with already-resolved dependency graphs

Interface stitching:

- Match use paths (Import Manifest) to definition paths (Export Manifest)
- Create cross-package call edges
- Handle both direct calls and transitive resolution

Result: Precise, on-demand call graphs at ecosystem scale

Algorithm 1 The compositional synthesis algorithm.

```
1: Input:  $P_{app}$ : The target application package;  $D$ : The set of dependencies.
2: Output: The resolved synthesis graph for  $P_{app}$ .
3: procedure SYNTHESIZEECG( $P_{app}, D$ )
4:    $L \leftarrow \text{ReverseTopologicalSort}(D \cup \{P_{app}\})$ 
5:    $\text{ResolvedGraphs} \leftarrow \text{new Map}()$ 
6:   for all package  $P_i$  in  $L$  do
7:      $G_i \leftarrow \text{GetOriginalRSG}(P_i)$ 
8:     for all dependency  $P_j$  of  $P_i$  do
9:        $G_j \leftarrow \text{ResolvedGraphs.get}(P_j)$ 
10:       $G_i \leftarrow \text{COMPOSE}(G_i, G_j)$ 
11:    end for
12:     $\text{ResolvedGraphs.put}(P_i, G_i)$ 
13:  end for
14:  return  $\text{ResolvedGraphs.get}(P_{app})$ 
15: end procedure

16: function COMPOSE( $G_{caller}, G_{callee}$ )
17:    $G_{new} \leftarrow G_{caller} \cup G_{callee}$ 
18:   Let  $\langle \mathcal{M}_{E,c}, \mathcal{M}_{I,c} \rangle \leftarrow \text{GetContract}(G_{caller})$ 
19:   Let  $\langle \mathcal{M}_{E,d}, \mathcal{M}_{I,d} \rangle \leftarrow \text{GetContract}(G_{callee})$ 
20:   for all path  $u \in \text{domain}(\mathcal{M}_{I,c})$  do
21:     if  $u$  targets package of  $G_{callee}$  then
22:       MATCHANDRESOLVE( $u, \mathcal{M}_{E,d}, G_{new}$ )
23:     end if
24:   end for
25:   return  $G_{new}$ 
26: end function
```

RQ1 - Call Graph Accuracy

Evaluation Setup:

- 7 projects with 100% test coverage
- Dynamic ground truth (NodeProf + GraalVM)
- Compare against Jelly (state-of-the-art)

Key Results:

- VulTracer: F1 score of 0.905
- Perfect precision: 1.000 (zero false positives)
- Superior recall: 0.841 vs Jelly's 0.806
- Inter-package coverage: 65.08% vs Jelly's 58.67%

RQ1 - Call Graph Accuracy

Evaluation Setup:

TABLE I: Evaluation of intra- and inter-package call resolution accuracy across different tools. VT denotes VULTRACER.

Project	Stars	Intra-package						Inter-package	
		Jelly(R)	VT(R)	Jelly(P)	VT(P)	Jelly(F ₁)	VT(F ₁)	Jelly(Coverage)	VT(Coverage)
gulpjs/gulp	33.1K	0.884	0.884	0.884	1.000	0.884	0.938	75.84% (113/149)	83.22% (124/149)
markdown-it/markdown-it	19.2K	0.484	0.491	0.737	1.000	0.584	0.658	100% (1/1)	100% (1/1)
tj/co	11.9K	0.993	0.907	0.168	1.000	0.287	0.951	37.50% (3/8)	37.50% (3/8)
woorm/franc	4.2K	0.720	1.000	0.947	1.000	0.818	1.000	4.44% (2/45)	31.11% (14/45)
primus/eventemitter3	3.4K	0.825	0.819	0.886	1.000	0.854	0.900	44.58% (42/94)	45.74% (43/94)
bcoe/c8	2K	0.970	0.921	0.867	1.000	0.916	0.959	69.35% (86/124)	71.77% (89/124)
cosmicanant/recursive-diff	153	0.765	0.863	0.780	1.000	0.772	0.926	-	-
Average		0.806	0.841	0.753	1.000	0.731	0.905	58.67% (247/421)	65.08% (274/421)

Bold values indicate the superior result in each comparison pair. **R** denotes Recall, and **P** denotes Precision. Coverage is shown as (covered items/total items) percentage.

RQ2 - Scalability Wins

Performance Comparison on 99 Real Projects (CVE-2023-32314):

Jelly (Monolithic):

- Success rate: **37.37%** (37/99 packages)
- Majority fail: out-of-memory errors
- Must re-analyze **26,653 dependencies** for each project
- Average time: 31.34 minutes (for successful cases only)

VulTracer (Compositional):

- Success rate: **99.41%** (503/506 packages)
- **98% reduction** in analysis scope (26,653 → 506 unique packages)
- One-time pre-computation: 174 minutes (cached for reuse)
- On-demand synthesis: **41.87 seconds** per project

Key Advantage: Time scales with unique packages, not project complexity

RQ2 - Scalability Wins

Performance Comparison on 99 Real Projects (CVE-2023-32314):

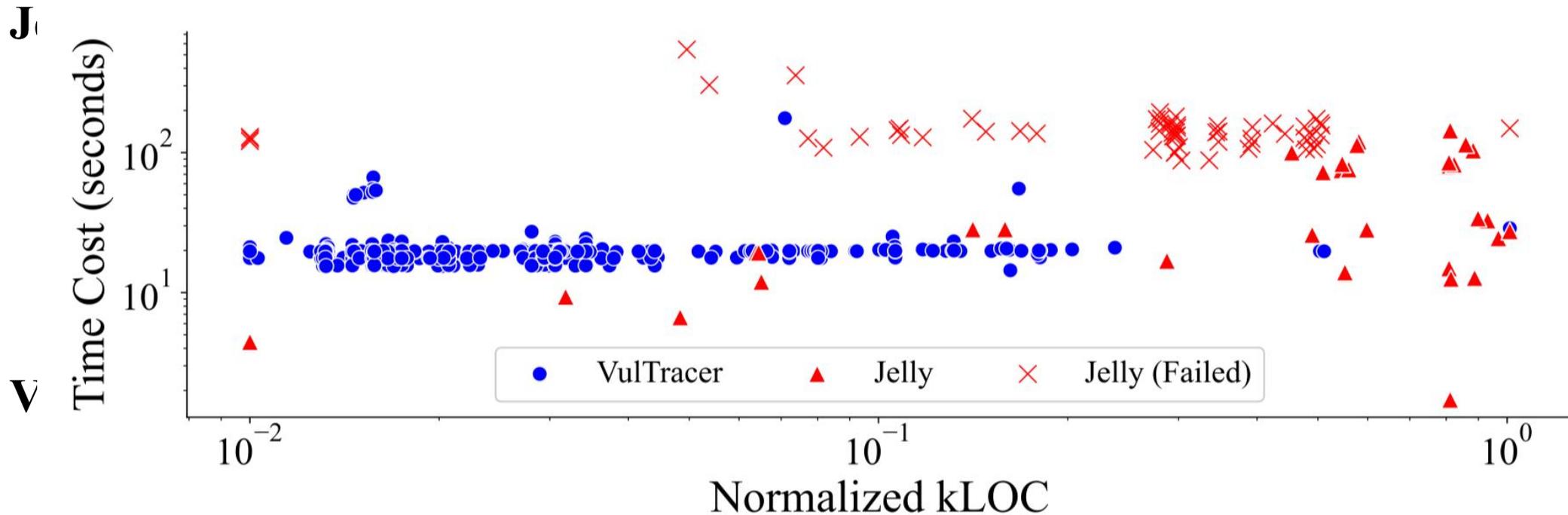


Fig. 3: Time consumption comparison of different tools.

- On-demand synthesis: **41.87 seconds** per project

Key Advantage: Time scales with unique packages, not project complexity

RQ3 - False Positive Reduction

Real-World Vulnerability Auditing

Dataset: 12 applications (from JAM benchmark)

npm audit baseline:

- 532 vulnerable packages reported
- 75 vulnerability propagation paths

VulTracer Results:

- 532 packages → 53 alarms (49 unique vulnerabilities)
- Manual verification: Only **21 True Positives**
- **94%** false positive reduction VS npm audit
- **Better than prior work:** JAM achieved 81% reduction

Impact: Dramatically reduces alert fatigue while maintaining accuracy

TABLE II: Comparison of vulnerability audit results of VULTRACER and npm audit.

Target	# Pkgs	# P_{vul}	# Alarm	npm audit		VULTRACER	
				TP	FP	TP	FP
makeappicon@1.2.2	14	6	2	0	2	0	
touch@0.0.1	25	5	4	4	0	0	
spotify-terminal@0.1.2	85	16	6	3	3	0	
ragan-module@1.3.0	56	1	3	0	3	0	
npm-git-snapshot@0.1.1	36	2	4	0	4	0	
nodetree@0.0.3	5	6	2	0	2	0	
jsonwebtoken@1.0.1	79	7	4	0	4	0	
foxx-framework@0.3.6	61	1	3	0	3	0	
npmgenerate@0.0.1	23	5	4	4	0	4	
smrti@1.0.3	59	1	3	0	3	0	
writex@1.0.4	46	16	8	6	2	6	
openbadges-issuer@0.4.0	43	9	10	6	4	6	
Total	532	75	53 (49)	21	32	21	2

Pkgs refers to the count of packages within the dependency graph. P_{vul} indicates the number of vulnerability propagation path. **Alarm** denotes all vulnerable packages in the dependency graph.

RQ4 - Ablation Study

Variants:

- **VulTracer-Full:** Complete system
- **VT-NoContract:** Removes formal contracts (uses name matching)
- **VT-SimpleAPI:** Restricts API vocabulary (removes getReturn, getParameter, getInstance)
- **VT-NoRTS:** Disables Reverse Topological Sort

TABLE III: Ablation study results for VULTRACER.

Variant	Intra F_1	Inter Cov.	Resolved
VULTRACER (Full)	0.905	65.08%	274
VT-NoContract	0.905	19.95%	84
VT-SimpleAPI	0.905	48.22%	203
VT-NoRTS	0.905	58.19%	245

Note: **Intra F_1 :** Avg. Intra-Package F_1 score; **Inter Cov.:** Inter-Package Coverage; **Resolved:** Number of calls resolved (out of 421 total).

Ecosystem-Scale Study Datasets

DS_{npm}: *Complete npm ecosystem*

- 3,267,273 unique packages
- 34,685,976 distinct versions
- 900+ million dependency links
- Data collected through December 31, 2024

DS_{CVE}: *Two-dimensional vulnerability selection*

1. High-impact: 6 CVEs from top 10 most downloaded packages
 - lodash, debug, semver, minimatch
2. Diversity: 21 CVEs aligned with 2024 CWE Top-25
 - Covering injection, prototype pollution, path traversal, etc.

Total: 27 unique CVEs with precisely identified vulnerable functions

Ecosystem-Scale Study Dataset

DS_{npm}: Complete npm ecosystem

- 3,267,273 unique packages

TABLE VII: List of selected high-impact vulnerabilities. #Vul_{func} denotes the number of vulnerable functions.

CVE ID	Package Name	Downloads (2024)	# Vul _{func}
CVE-2021-23337	lodash	2.68B	1
CVE-2022-25883	semver	16.57B	14
CVE-2017-16137	debug	13.61B	1
CVE-2017-20165	debug	13.61B	1
CVE-2022-3517	minimatch	9.78B	7
CVE-2016-10540	minimatch	9.78B	5

2. Diversity: 21 CVEs aligned with 2024 CWE Top-25

- Covering injection, prototype pollution, path traversal

Total: 27 unique CVEs with precisely identified

TABLE VIII: Detailed list of selected vulnerabilities for diversity evaluation (CWE-Top-25). #Vul_{func} denotes the number of vulnerable functions.

CWE ID	Package Name	CVE ID	# Vul _{func}
CWE-79	happy-dom	CVE-2024-51757	4
CWE-787	electron	CVE-2022-4135	2
CWE-89	parse-server	CVE-2024-27298	4
CWE-352	whistle	CVE-2024-55500	5
CWE-22	@vendure/ asset-server-plugin	CVE-2024-48914	7
CWE-125	@openzeppelin/contracts	CVE-2024-27094	-
CWE-78	find-exec	CVE-2023-40582	3
CWE-416	@fastly/js-compute	CVE-2024-38375	7
CWE-862	snarkjs	CVE-2023-33252	3
CWE-434	strapi	CVE-2022-27263	3
CWE-94	angular-expressions	CVE-2024-54152	1
CWE-20	@vendure/ asset-server-plugin	CVE-2024-48914	7
CWE-77	openssl	CVE-2023-49210	1
CWE-287	isolated-vm	CVE-2022-39266	5
CWE-269	@aws-amplify/cli	CVE-2024-28056	4
CWE-502	gatsby-plugin-mdx	CVE-2022-25863	4
CWE-200	eventsource	CVE-2022-1650	1
CWE-863	next-auth	CVE-2022-35924	3
CWE-918	parse-url	CVE-2022-2900	1
CWE-119	@solana/web3.js	CVE-2024-30253	5
CWE-476	ws	CVE-2024-37890	2
CWE-798	-	-	-
CWE-190	@chainsafe/lodestar	CVE-2022-29219	1
CWE-400	@stryker-mutator/util	CVE-2024-57085	1
CWE-306	-	-	-

RQ5 - Over-Approximation

RQ5: How Much Do Package-Level Alerts Over-Approximate?

Analysis: 27 CVEs, 703,896 Direct Dependents (d1)

Key Finding:

- **Single-hop:** *67.51%* global over-approximation
- **Multi-hop (transitive):** *68.28%* of package-level alerts are false positives
- Only **32.49%** of flagged packages actually reach vulnerable functions

Implication: *Package-level tools create massive noise*

- 2 out of 3 alerts don't represent real threats
- Alert fatigue is a solvable technical problem

TABLE IX: Comprehensive single-hop reachability analysis merging High-Impact and Diversity datasets. The global average is weighted based on the number of CVEs in each dataset.

Dimension	CVE ID	Package	# d_0	# d_1	# C_{mod}	# C_{func}	# C_{vuln_func}
High-Impact	CVE-2021-23337	lodash	100	396,112	264,179 (66.69%)	244,130 (61.63%)	11,574 (2.92%)
	CVE-2022-3517	minimatch	26	38,112	28,667 (75.22%)	15,791 (41.43%)	15,791 (41.43%)
	CVE-2016-10540	minimatch	23	10,341	9,211 (89.07%)	3,528 (34.12%)	3,528 (34.12%)
	CVE-2022-25883	semver	74	139,257	111,138 (79.81%)	102,209 (73.40%)	73,314 (52.65%)
	CVE-2017-16137	debug	55	70,297	54,098 (76.96%)	51,425 (73.15%)	50,454 (71.77%)
	CVE-2017-20165	debug	42	39,365	29,702 (75.45%)	29,583 (75.15%)	29,576 (75.13%)
Diversity	CVE-2022-1650	eventsourcing	17	167	109 (65.27%)	100 (59.88%)	100 (59.88%)
	CVE-2022-25863	gatsby-plugin-mdx	125	610	286 (46.89%)	0 (0.00%)	0 (0.00%)
	CVE-2022-27263	strapi	16	30	3 (10.00%)	3 (10.00%)	3 (10.00%)
	CVE-2022-2900	parse-url	11	204	67 (32.84%)	63 (30.88%)	63 (30.88%)
	CVE-2022-29219	@chainsafe/lodestar	23	23	17 (73.91%)	11 (47.83%)	0 (0.00%)
	CVE-2022-35924	next-auth	17	58	34 (58.62%)	10 (17.24%)	9 (15.52%)
	CVE-2022-39266	isolated-vm	15	38	25 (65.79%)	25 (65.79%)	25 (65.79%)
	CVE-2022-4135	electron	504	2,453	1,978 (80.64%)	1,816 (74.03%)	1,816 (74.03%)
	CVE-2023-33252	snarkjs	27	309	243 (78.64%)	220 (71.20%)	148 (47.90%)
	CVE-2023-40582	find-exec	8	11	11 (100.00%)	11 (100.00%)	11 (100.00%)
	CVE-2023-49210	openssl	2	56	0 (0.00%)	0 (0.00%)	0 (0.00%)
	CVE-2024-27298	parse-server	9	27	13 (48.15%)	8 (29.63%)	8 (29.63%)
	CVE-2024-28056	@aws-amplify/cli	5	13	0 (0.00%)	0 (0.00%)	0 (0.00%)
	CVE-2024-30253	@solana/web3.js	109	428	400 (93.46%)	381 (89.02%)	168 (39.25%)
	CVE-2024-37890	ws	86	4,080	3,163 (77.52%)	2,389 (58.55%)	1,561 (38.26%)
	CVE-2024-38375	@fastly/js-compute	24	24	0 (0.00%)	0 (0.00%)	0 (0.00%)
	CVE-2024-48914	@vendure/asset-server-plugin	38	41	0 (0.00%)	5 (12.20%)	5 (12.20%)
	CVE-2024-51757	happy-dom	150	452	252 (55.75%)	257 (56.86%)	13 (2.88%)
	CVE-2024-54152	angular-expressions	8	65	44 (67.69%)	42 (64.62%)	41 (63.08%)
	CVE-2024-55500	whistle	7	27	13 (48.15%)	13 (48.15%)	0 (0.00%)
CVE-2024-57085	@stryker-mutator/util	80	640	588 (91.88%)	546 (85.31%)	64 (10.00%)	
Average			-	-	57.72%	45.71%	32.49%

Understanding Attenuation - API Surface

Why Such Different Propagation Rates?

Factor 1: API Breadth

- Broad API (lodash):

- 242 functions available
- template (vulnerable): rank #49, only 0.30% of calls
- Top functions: forEach (10.66%), isFunction (9.99%) are safe
- **Result:** 2.92% propagation (shallow usage)
- Long-tail distribution limits impact

- Narrow API (debug):

- Focused on core debugging functionality
- 98% of importers use the vulnerable function
- **Result:** 71.77% propagation (deep usage)
- Limited alternative functions drive high usage

TABLE V: Frequency and dependency analysis of lodash functions. N_{call} represents the total call count. D_{total} is the number of downstream packages including all versions, while D_{uniq} is the count of unique downstream package names.

No.	F_{name}	# N_{call} (%)	Downstream Dependencies (d_1)	
			# D_{total} (%)	# D_{uniq} (%)
1	forEach	834,950 (10.66%)	48,224 (19.75%)	4,363 (20.00%)
2	isFunction	782,490 (9.99%)	41,857 (17.15%)	3,741 (17.15%)
3	get	753,685 (9.62%)	37,695 (15.44%)	2,066 (9.47%)
4	map	457,311 (5.84%)	60,801 (24.91%)	5,513 (25.27%)
5	isEmpty	423,905 (5.41%)	48,974 (20.06%)	3,084 (14.14%)
6	isObject	335,162 (4.28%)	44,844 (18.37%)	3,650 (16.73%)
7	isString	326,310 (4.17%)	59,572 (24.40%)	4,542 (20.82%)
8	cloneDeep	283,015 (3.61%)	44,765 (18.34%)	2,829 (12.97%)
9	isUndefined	271,190 (3.46%)	27,684 (11.34%)	2,551 (11.69%)
10	filter	213,382 (2.73%)	36,464 (14.94%)	2,919 (13.38%)
48	contains	23,728 (0.30%)	7,413 (3.04%)	801 (3.67%)
49	template	23,148 (0.30%)	11,574 (4.74%)	1,150 (5.27%)
50	isNaN	22,903 (0.29%)	8,475 (3.47%)	437 (2.00%)

Insight: Vulnerability impact correlates with API specificity

Understanding Attenuation - Unused Dependencies

Factor 2: Unused Dependencies

Critical Finding:

- 22.80% of high-impact direct dependents (d_1), 42.28% globally
- Never import the dependency at all (fail C_{mod} condition)

Examples:

- CVE-2021-23337 (lodash): 131,933 packages declare but never use
- Declared in package.json but no actual import statements

Implication: Instant false positive

- Presence in dependency graph \neq actual usage

TABLE IV: Single-hop reachability analysis and attenuation for High-impact vulnerabilities. The full detailed results for all 27 CVEs are provided in Table IX in the Appendix.

CVE ID	Package Name	# Vul _{func}	# d_0	# d_1	# C_{mod}	# C_{func}	# C_{vuln_func}
CVE-2021-23337	lodash	1	100	396,112	264,179 (66.69%)	244,130 (61.63%)	11,574 (2.92%)
CVE-2022-3517	minimatch	7	26	38,112	28,667 (75.22%)	15,791 (41.43%)	15,791 (41.43%)
CVE-2016-10540	minimatch	5	23	10,341	9,211 (89.07%)	3,528 (34.12%)	3,528 (34.12%)
CVE-2022-25883	semver	14	74	139,257	111,138 (79.81%)	102,209 (73.40%)	73,314 (52.65%)
CVE-2017-16137	debug	1	55	70,297	54,098 (76.96%)	51,425 (73.15%)	50,454 (71.77%)
CVE-2017-20165	debug	1	42	39,365	29,702 (75.45%)	29,583 (75.15%)	29,576 (75.13%)
Average	-	-	-	-	77.20%	59.81%	46.34%

Interpretation

Interpretation: *True vulnerability impact is shallow and localized*

Key Insights:

- Initial attenuation factors (unused deps, shallow API usage) compound at each hop
- Real vulnerabilities rarely propagate deep through dependency chains
- Most transitive propagation paths are noise, not signal
- Function-level analysis reveals true attack surface

Practical Impact:

- Focus remediation on nearby dependencies (≤ 4 hops)
- Prioritize direct and close transitive dependencies
- Long dependency chains rarely represent real threats

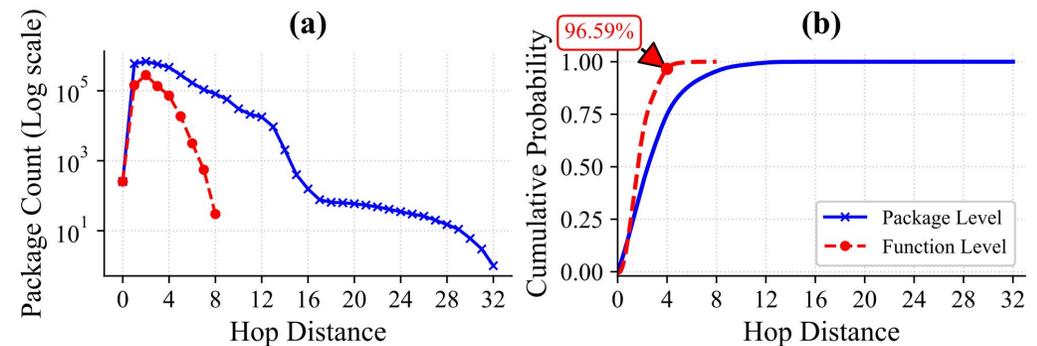


Fig. 4: Package- vs. Function-level propagation decay: (a) Count per hop; (b) CDF.

Practical Implications

For Developers and Security Teams:

- **Reachability-driven triage:** Prioritize by evidence of reachability, not just presence
- **Direct dependencies first:** Reachable vulnerabilities in direct deps warrant immediate action
- **Dependency hygiene:** Remove unused dependencies to reduce attack surface and noise
- **Strategic allocation:** Focus resources on genuine, reachable threats

For SCA Tool Vendors:

- **Move beyond package-level:** Integrate function-level call-graph analysis
- **Reduce alert fatigue:** Distinguish presence from reachability
- **Actionable intelligence:** Provide precise, prioritized threat information
- **Transform value proposition:** From overwhelming reports to targeted insights

Conclusions

VulTracer bridges the gap between precision and scalability:

- **Addresses critical problem:** Alert fatigue from imprecise tools
- **Enables practical solution:** Function-level analysis at CI/CD speed
- **Provides actionable insights:** Focus on reachable, genuine threats
- **Transforms understanding:** Vulnerability propagation is shallower than believed

From Noise to Signal:

- Package-level analysis: **68.28% noise**
- Function-level analysis: Reveals true signal
- **Impact:** Developers can finally answer *"Am I really affected?"*

Future: Path to more effective software supply chain security

From Noise to Signal: Precisely Identify Affected Packages of Known Vulnerabilities in npm Ecosystem

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