

Tracking Dependency Updates & Security: Do Bots Make a Difference?

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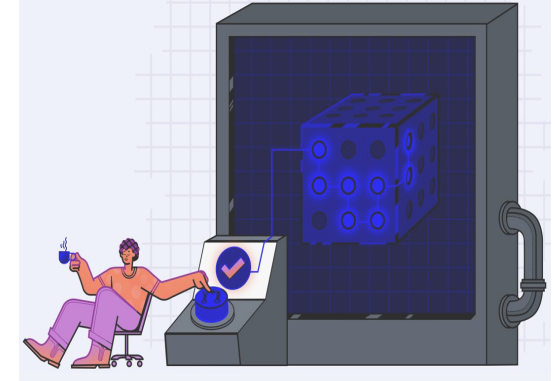
Background

Risks in dependency management

- Security risks
- Technical debt
- Compliance issues

2021 Log4j vulnerability exposed millions of systems

74% open-source codebases contain at least one open-source vulnerability (2024 Synopsys report)



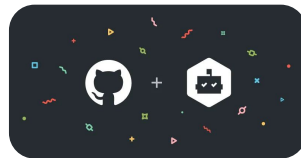
Automated Dependency Management Tools

Automatically open Pull Requests

Update dependencies on a collaborative platform like GitHub

Why?

Dependabot, RenovateBot, SnykBot, Depfu

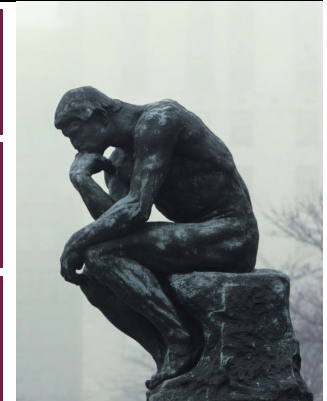


Research Questions

RQ-1: What is the average Dependency Freshness in projects and how does it vary across projects of different sizes, popularity and type?

RQ-2: What are the average Vulnerability Exposure Windows in projects, and how do they vary across projects of different sizes, popularity, and type?

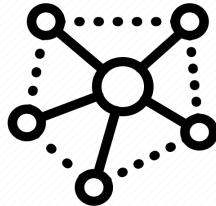
RQ-3: Does the adoption of dependency management bots correlate with reduced Dependency Freshness Vulnerability Exposure Windows?



Dataset & Tools

- 15,117,217 nodes
- 658,078 artifacts
- 14,459,139 releases
- 44,035,495 AddedValue Nodes
- Querid via Cypher
- 77,393 vulnerable releases (1411 artifacts)
- 197,186 artifacts depending on these directly
- GitHub API for metrics and identifying projects
- OSV API

Goblin Framework (represents libraries and releases)



Gathering Project Metrics for Stratified Analysis

- Size: Number of Dependencies + LoC
- Popularity: Number of Dependents + GitHub Stars
- **Project Type**
- Straightforward for numDependencies and numDependents
- **Not so straightforward for the rest!**
- 658,078 artifacts -> reduced set of 38,794 artifacts

Identifying Dependency Freshness

- The time elapsed between the adoption of a dependency by a project and the latest available release of that dependency
- Artifact -> Release -> DependencyVersion (t0) -----> mostRecentVersionOfDependency (t1)
- Aggregate across all releases
- Mean Dependency Freshness + Summary Stats (median+max+min)

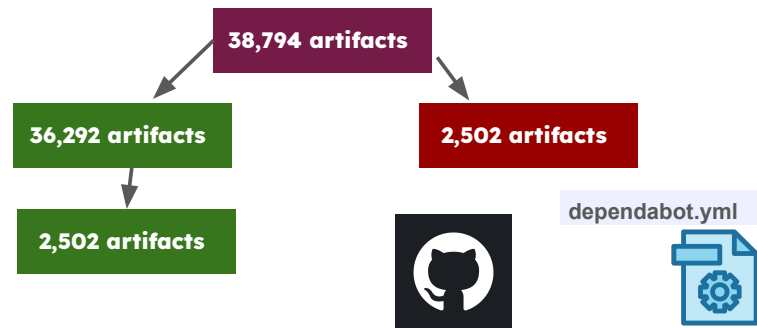
Identifying Vulnerability Exposure Windows

- The windows of time in which an artifact is vulnerable
- ExposureStartTime, ExposureEndTime, cvePublishTime, cveFixTime
- **A ->R -> dep -> B1**
- **A -> R ->dep -> B7 OR A x B**
- No fixes for 29.4% of vulnerable Artifacts (415/1411)

Table 1: Summary of Vulnerability Exposure Windows and Their Descriptions.

Exposure Window Type	Description
True Exposure Window	(<i>exposureEndTime</i> - <i>exposureStartTime</i>), quantifies the total duration an artifact remains vulnerable.
Known Exposure Window	(<i>exposureEndTime</i> - <i>cvePublishTime</i>), captures the period during which the vulnerability was publicly disclosed yet unaddressed.
Patch Lag Window	(<i>exposureEndTime</i> - <i>cveFixTime</i>), reflects the delay between the fix release and the project's adoption of a non-vulnerable version.

Identifying Dependency Freshness and Vulnerability Exposure Windows in Control and Test Group

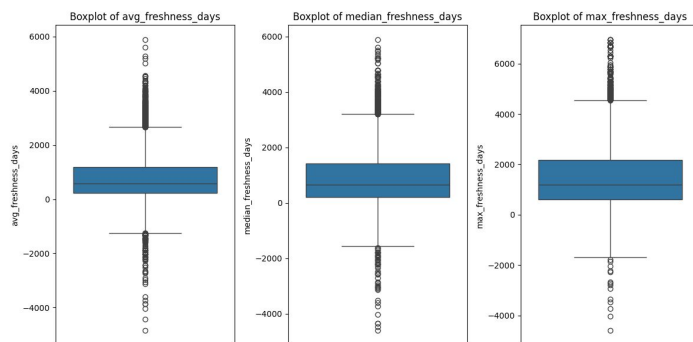


ANALYZING DEPENDABOT IMPACT

- Queried GitHub API for dependabot.yml file
- Control group (36,292) vs treatment group (2502)
- Insufficient representation of bot usage other than Dependabot

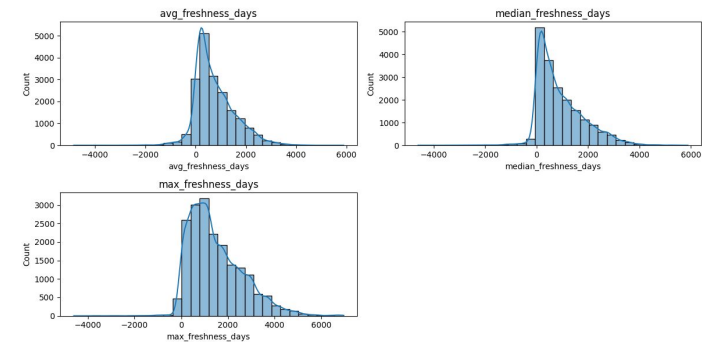
RESULTS

AVERAGE DEPENDENCY FRESHNESS



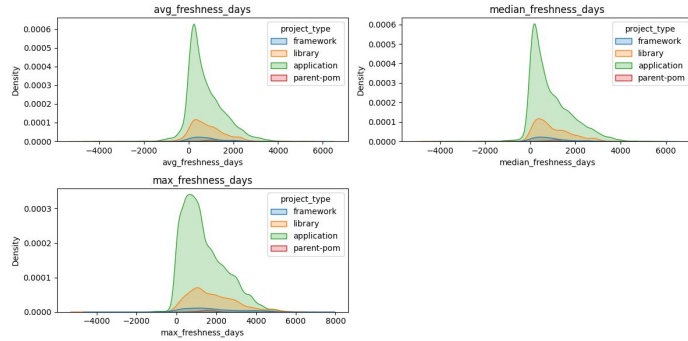
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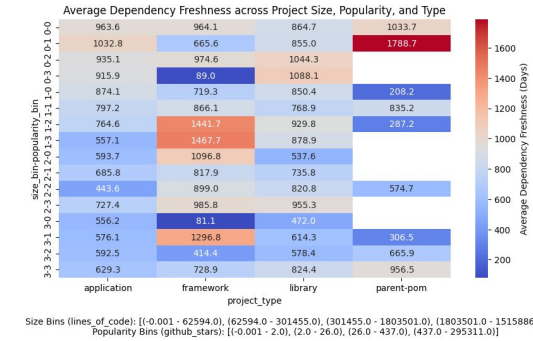
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RESULTS

AVERAGE DEPENDENCY FRESHNESS



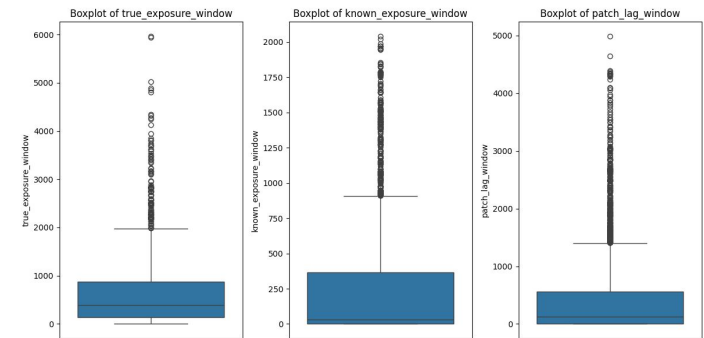
RESULTS

AVERAGE DEPENDENCY FRESHNESS

- Negative correlation for lines of code and number of dependents
- Positive correlation for GitHub stars
- P-values <0.05

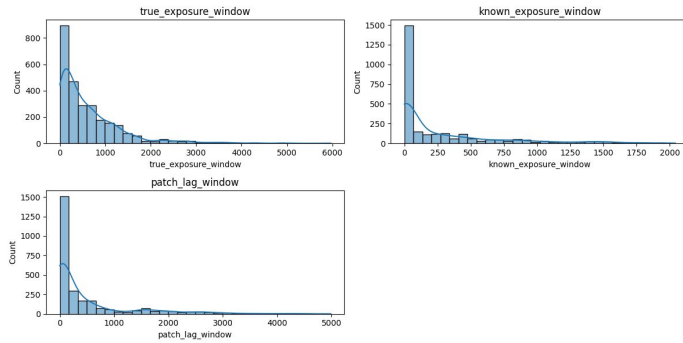
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VULNERABILITY EXPOSURE WINDOW



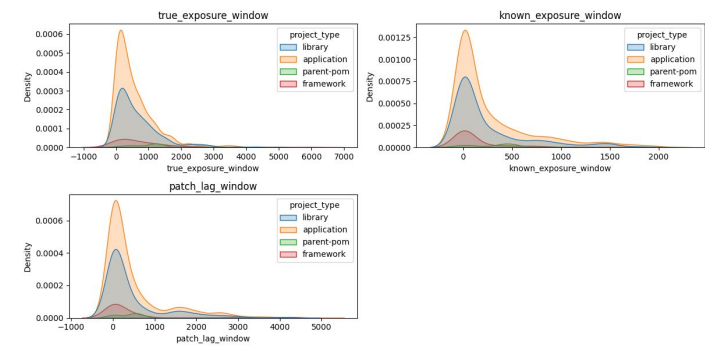
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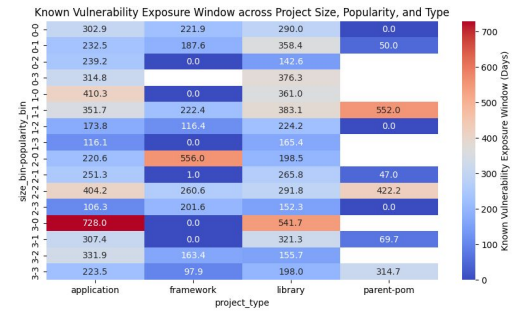
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VULNERABILITY EXPOSURE WINDOW



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VULNERABILITY EXPOSURE WINDOW



Size Bins (lines_of_code): [(-0.001 - 125994.0), (125994.0 - 870243.0), (870243.0 - 4225326.0), (4225326.0 - 1515886467.0)]
Popularity Bins (github_stars): [(-0.001 - 6.0), (6.0 - 53.0), (53.0 - 823.0), (823.0 - 116860.0)]

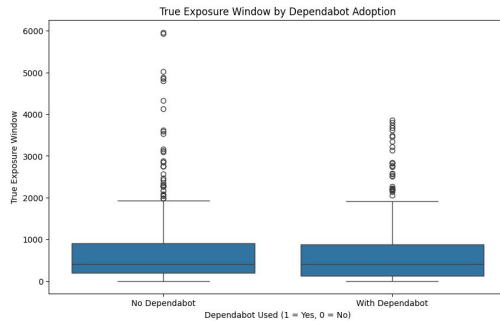
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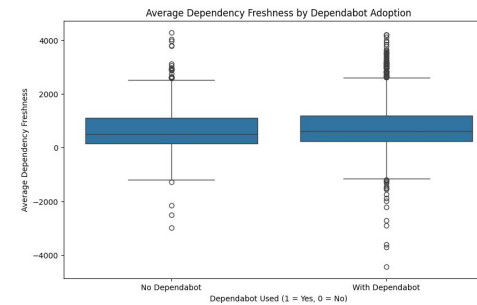
IMPACT OF DEPENDABOT ON VULNERABILITY EXPOSURE WINDOW



- T-statistic: -1.33, P-value: 0.18
- U-statistic: 188886.5, P-value: 0.1655

RESULTS

IMPACT OF DEPENDABOT ON AVERAGE DEPENDENCY FRESHNESS



- T-statistic: 5.93, P-value: 3.3E-9
- U-statistic: 4345206, P-value: 9.47E-9

Threats to Validity

1: Repositories containing a .yaml configuration file (like dependabot.yaml) may not actively use the dependency management bot

Verify active bot usage by checking for pull requests authored by the bot, analyzing commit histories for dependency updates, checking config settings

2: Selection bias could skew results

Use control groups of similar project size, popularity, and domain.

3: Causal ambiguity in the findings

Perform longitudinal study to isolate tool impact.

