Modelling Risk in Open-Source Software

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Course: MCM Computing Supervisor: Darragh O'Brien

Introduction

- → Popularity of Open-Source Software has shown a steady increase over time
 - ◆ 3.6 million repos depend on the top 50 open-source projects
- → The use comes with the risk of software vulnerabilities
 - Attackers can exploit these
- → Prediction of risk is important for this
 - ◆ Software metrics
 - Project Activity
 - Vulnerability Data

Research Question

Can we provide developers with an effective risk measure when choosing between multiple open-source components?

Key Terms

- → Maven
 - Open-source build automation & project management tool
- → Time series analysis
 - Fitted model based on past data can predict future values
- → ARIMA
 - Autoregressive integrated moving average
- → AutoARIMA
 - Tests parameter combinations to determine the best model

Background Research

Vulnerability Propagation

- → Only 1.2% directly use vulnerable code
- → Small packages can affect 375, 607 packages in the Maven ecosystem
 - jackson-databind, netty-codec-http
 - CVEs can affect a large number of projects
- → The more dependencies the more complex it is to find vulnerabilities in dependencies

Project Metadata Analysis

- → Project activity level is an indicator of survival
 - ♦ Number of commits, time-to-fix
 - ◆ Long-time Contributors
- → Less than 50% of abandoned projects find new contributors
- → Standard measurement for commits is:
 - Number of commits per month

Vulnerability CVE Data Analysis

- → Predicting number of vulnerabilities per month is also important
- → ARIMA is a very popular method of prediction
 - Seasonality not a factor
- → Datasets from the NVD were used in every study we analysed
 - Prediction of CVEs
- → Most models fall flat at the three month mark

Case Studies

Log4j Vulnerability

- → Discovered in November 2021 in popular logging library
- → Gateway to gain control of the machine
- Many projects unaware they were affected

Heartbleed Vulnerability

- → Discovered in April 2014 in popular cryptography library
- → Receive private information after crafting similar messages
- Many servers unaware they were affected

Methodology

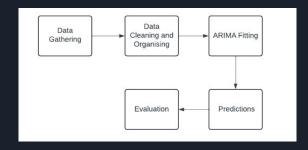
Data & Dependencies

- → Different GitHub projects to test the algorithm
- → GitHub API for project activity
- → NVD API for vulnerability
 Data

- Each Maven Dependency as a node
- → NetworkX graph
- → Keywords and libraries extracted for prediction
- → Colour-coded

Data Source	Purpose	Type Maven-dependency trees API	
GitHub projects	Find Dependencies		
GitHub API	Project Meta-data		
NVD API	Vulnerability Prediction	CVE API	

Predictions



Project Activity

- → GitHub API
 - ◆ All available data gathered
- Commits, average time-to-fix issues or both
- → AutoARIMA prediction
- → Predicted value for the next month returned

Vulnerabilities

- → NVD API
 - Call made for each keyword
- → Number of vulnerabilities per month
- → AutoARIMA Prediction
- → Predicted value for the next month returned

Risk Calculations

$$projectActivityScore = (x/numDaysToFixIssues)*10$$

$$projectActivityScore = (numCommits/x) * 10$$

$$vulnerabilityScore = (x/vulsCountPerMonth)*10$$

$$overallScore = (projectActivityScore + \\ vulnerabilityScore)/2$$

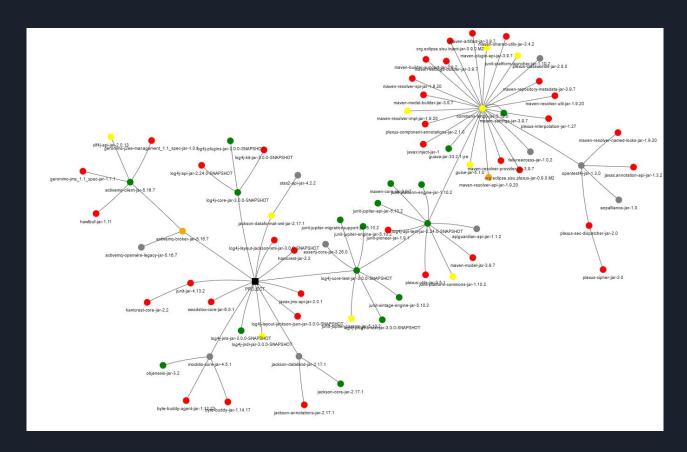
- → User-decided acceptable levels
- → Options for project activity score
- → Combination of project activity and vulnerability for overall score

Score	Risk Level	
<0	Not Enough Data	
0 - 2.5	Low Risk	
2.5-5	Medium Risk	
5-7.5	High Risk	
>7.5	Severe Risk	

Results

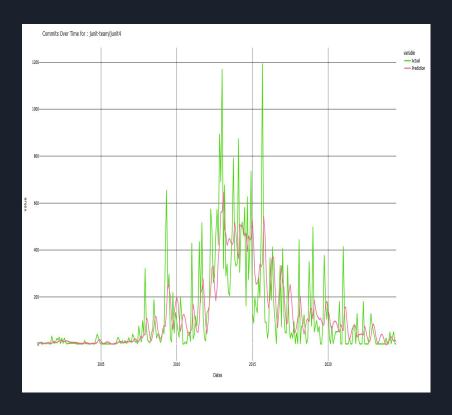
Example Final Graph





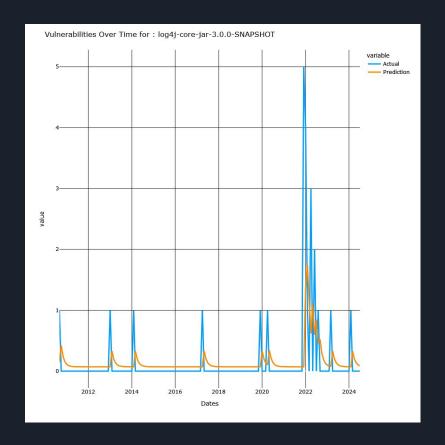
Example Graph of Project Activity Model

- → JUnit4 package commits over time
- Clear to see that activity has declined in recent years
- → 2012 to 2017 were high commit years
 - ◆ At its peak over 200 commits per month
- → Prediction is fairly accurate as can be seen in the graph



Example Graph of Vulnerability Data Model

- → Log4j vulnerabilities over time
- Clear where many were released in 2021
 - 5 released in one month
- → Reasonably accurate predictions in subsequent months



Evaluations

Prediction	MAE	RMSE
Commits per month	123.18	213.96
Vulnerabilities per month	0.145	0.476

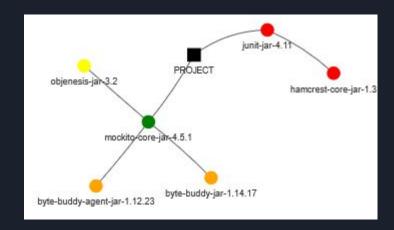
- → Log4j sample project
- → Project Activity Predictions > CVE Data Predictions
- → Commonly used metrics
 - Mean Absolute Error (MAE)
 - Root Mean Squared Error (RMSE)
- → Large fluctuations in commits
- → Small fluctuations in vulnerability prediction

Technical Challenges

- → Graphing dependencies and dependencies of dependencies etc.
- → Choosing of ARIMA parameters
 - Decided on AutoARIMA
 - PACF and ACF graphs
 - ◆ Differencing data until ADF test below 0.05
- → Connecting dependencies to GitHub URLs
 - Done using a mapping file

Tool Uses

- → Comparing similar dependencies
 - ◆ JUnit vs Mockito
- → Exploring risk in current dependencies
 - Finding risky modules



Student Contribution

Róisín Ní Bhriain

- → Gitlab setup + Assigning Issues
- → Literature Review
- → Software
- → Practicum Paper First + Second Draft

Sneha Dechamma Mallengada Suresh

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Conclusion

- → Growing reliance on OSS
 - Challenges regarding security
- → Better overview of dependency risk
- → Focused on Maven dependency trees
- Compromised on accuracy to predict many dependencies
- → Tool could aid developers
 - Choosing less risky modules
 - Discovering already used risky dependencies

References

Thank You For Listening!

Questions?