DISCOVERING DIGITAL SIBLINGS QUANTIFYING INTER-REPOSITORY SIMILARITY THROUGH GITHUB DEPENDENCY STRUCTURES

MOTIVATION

Digital siblings - Repositories with similar goals or problem domains.

- The Open Source software ecosystem is extensive, allowing developers to collaborate, share and learn from the code of others.
- Finding repositories similar to one's own allows for better collaboration, knowledge transfer, and code reuse.
- With more than 100 million repositories on GitHub, finding similar projects manually is a very difficult task.
- Dependencies, such as libraries and frameworks used by software projects, provide insight into the project's topic/problem domain, making an automated approach for finding digital siblings possible.

RESEARCH QUESTIONS

How can the dependency structures of GitHub repositories be leveraged to find their digital siblings?

- RQ2: What metrics, derived from analyzing dependency structures, most accurately quantify the similarity between GitHub repositories?
- RQ3: Which clustering methods are most effective in grouping GitHub repositories into clusters mirroring similar problem domains?
- RQ4: How can dependency structures as a similarity metric be composed with the similarity metrics investigated by the other RP group members?

DATA COLLECTION

minecraft plugins https://github.com/filoghost/HolographicDisp https://github.com/NoCheatPlus/NoCheatPlus https://github.com/ViaVersion/ViaBackwards https://github.com/games647/FastLogin https://github.com/TownyAdvanced/Towny https://github.com/SkinsRestorer/SkinsRestor

Dataset: Selecting a list of GitHub repositories to analyse. Currently: Minecraft modifications vs. server plugins.

Extraction: Extracting dependency tree

REPOSITORY VECTORIZATION



org.spigotmc:spigot-api

com.github.seeseemel.... junit:junit github.scarsz:configur... com.vdurmont:emoji-...

DBSCAN

Algorithm Design & Implementation: Creating a similarity metric based on shared dep. structures.

2. CLUSTERING TECHNIQUES EVALUATION



Rand Index / Normalized Mutual Information between Repository Clustering Techniques



3. COMPOSABLE SIMILARITY METRIC

Combines Similarity Metrics with Clustering Techniques.

A model *M* is trained by computing *k* clustering using the chosen training method. In inference, a characteristic vector C is computed for each unseen vector by calculating the distance of the vector to each cluster in pre-trained M. Similarity is then calculated using Euclidean distance on characteristic vectors.

METHODOLOGY

Performance Evaluation

- Similarity Metrics Euclidean Distance, XOR Sim., and AND Sim. • Clustering Techniques - Agglomerative, K-Means, and DBSCAN.

CONCLUSIONS

RQ2: Similarity Metrics **RQ3: Clustering Techniques RQ4:** Composable Approach siblings.

Main question:

There are effective similarity metrics and clustering techniques which can identify groups of digital siblings in a set of GitHub repositories.

4. COMPOSABLE SIMILARITY METRIC EVALUATION



1. SIMILARITY METRICS EVALUATION

EUCLIDEAN DISTANCE

Calculates the Euclidean distance between repository dependency vectors.

XOR SIMILARITY Performs XOR between repository dependency vectors and uses the proportion of dependencies as a similarity matching dependencies as a



metric.

ifference Factor between Repository Similarity Metrics

similarity metric.



"provider": "github.com" "repo": "AuthMeReloaded" "AuthMe ": "javax.xml.bind:jaxb-api",

DEPENDENCY EXTRACTION

ovider": "github.com" 'repo": "Slimefun4"

information for each repository in the data set.



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- Data Collection We collect a list of repositories to analyse.
- Dependency Extraction Direct and transitive dependencies are extracted automatically from each repository in our dataset.
- **Repository Vectorization** Each repository is transformed into a binary vector representing the dependencies used by the project.

- Composable Approach Most effective similarity metric and
- clustering technique combined for a hybrid, composable metric.

The AND Similarity metric proved the most efficient. On average, it assigns $\sim 3x$ higher similarity values when repositories are *digital siblings*, as opposed to dissimilar repositories.

The *DBSCAN* clustering proved the most efficient. ~79% of clusters produced by DBSCAN aligned with the reference clusters.

Training the model using K-Means (k = 5) clustering proved the most efficient. The metric assigns $\sim 2x$ higher similarity values to digital