Assignment 2

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Section 1: Objectives, Questions, and Metrics (GQM Approach)

Objective:

To investigate the impact of code smells on the modularity of software.

Questions:

- How does a bad code smell affect the coupling of classes in Java projects?
- How does a bad code smell affect the cohesion of classes in Java projects?
- What trends can be observed in modularity metrics for classes with bad smells compared to those without?

Metrics:

- 1. Lack of Cohesion in Methods (LCOM): LCOM quantifies the absence of coherence among the methods within a class. More LCOM values indicate inferior organization and potential difficulties in maintaining the code, as classes with more cohesion are typically more understandable and adaptable.
- 2. Coupling Between Objects (CBO): The CBO metric quantifies the degree of interdependence between a class and other classes within the system. More significant interdependencies are shown by higher CBO values. This can lead to more complexity and less maintainability since changes in one class may require changes in related classes.

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Section 2: Description of Subject Programs (Dataset)

Criteria for Selection

• Lines of Code (LOC): Projects should have at least 10,000 lines of code.

• Age: Projects should be at least 3 years old.

• **Number of Developers:** Projects should have at least 3 developers.

Justification:

Lines of Code (LOC): Projects with at least 10,000 lines of code are likely to possess sufficient complexity and variety to exhibit various bad smells and modularity issues. Smaller projects may not provide enough data to enable a meaningful analysis of these factors. Therefore, selecting projects with a substantial amount of code ensures that the study can effectively identify and evaluate the presence and impact of bad smells.

Age: Projects that are at least 3 years old are more suitable for this study because they are likely to have gone through multiple phases of development and maintenance. These phases can introduce and evolve bad smells over time. Analyzing older projects allows for the examination of the long-term effects of bad smells on modularity, providing insights into how these issues develop and persist in mature codebases.

Number of Developers: Projects with at least 3 developers are more likely to involve collaborative development practices. Such collaboration often results in varied coding styles and a higher potential for the introduction of bad smells. This diversity in development practices creates a richer dataset for analysis, enabling a comprehensive study of the correlation between bad smells and modularity across different development environments.

Table:

| Project Name | LOC (Lines | Age | Number of | Description | |
|---------------------|------------|---------|------------|------------------------------------|--|
| | of Code) | (Years) | Developers | | |
| Apache | 81,000 | 20 | 124 | Helper utilities for the java.lang | |
| Commons Lang | | | | API. | |
| JUnit5 | 68,000 | 6 | 90 | Testing framework for Java. | |
| Spring | 2,000,000 | 17 | 1,270 | Infrastructure support for Java | |
| Framework | | | | applications. | |
| Apache Hadoop | 3,000,000 | 15 | 2,300 | Framework for distributed | |
| | | | | processing of large data sets. | |
| Java Design | 28,000 | 8 | 268 | Design patterns implemented in | |
| Patterns | | | | Java | |
| Google Guava | 200,000 | 13 | 275 | Core libraries with new collection | |
| | | | | types and utilities. | |
| Apache Tomcat | 1,300,000 | 24 | 750 | Java Servlet, JSP, EL, and | |
| | | | | WebSocket implementation. | |
| Dubbo | 39,000 | 10 | 393 | Apache Dubbo is a high- | |
| | | | | performance, java based, open- | |
| | | | | source RPC framework. | |
| RxJava | 67,000 | 10 | 328 | Library for composing | |
| | | | | asynchronous programs. | |

| Apache | 1,400,000 | 14 | 1,100 | Highly scalable, distributed | |
|-----------|-----------|----|-------|------------------------------|--|
| Cassandra | | | | NoSQL database. | |

Section 3: Tool Description

CK Metrics Tool:

The CK Metrics tool is a widely used static code analysis tool that calculates various metrics to assess the quality and maintainability of Java code. Developed by Mauricio Aniche, this tool specifically measures several key object-oriented metrics, including Coupling Between Objects (CBO) and Lack of Cohesion in Methods (LCOM), among others. These metrics provide insights into the modularity and complexity of the software. The CK Metrics tool was chosen for this study because it provides comprehensive and relevant metrics, such as CBO and LCOM, which are essential for analyzing the modularity of software. Its ease of use and robust reporting capabilities make it an ideal choice for this empirical study.

JDeodorant:

JDeodorant is an Eclipse plug-in designed to detect code bad smells in Java applications. Developed by Theodoros Tsantalis, it identifies common bad smells such as God Class, Long Method, and Feature Envy, and suggests possible refactoring solutions. JDeodorant's detection algorithms are based on well-established research in the field of software engineering. JDeodorant was selected for this study due to its effectiveness in identifying a wide range of code bad smells, which are critical for our analysis. Its integration with Eclipse and user-friendly interface make it accessible for both novice and experienced developers, ensuring that bad smells are detected accurately and efficiently.

Section 4: Results

| Project | Avg CBO | Avg LCOM | With Smell | Without Smell |
|----------------------|----------|------------|------------|---------------|
| apache-cassandra | 6.630418 | 56.188151 | 6364 | 4202 |
| apache-hadoop | 6.836977 | 54.394456 | 8611 | 6577 |
| apache-tomcat | 4.485937 | 151.157508 | 2424 | 2198 |
| Apahce-commons-lang | 3.120930 | 273.753488 | 518 | 557 |
| dubbo | 5.177738 | 41.552289 | 2557 | 1899 |
| google-guava | 3.872595 | 47.358163 | 2488 | 3956 |
| java-design-patterns | 3.374439 | 1.904709 | 651 | 1133 |
| Junit-5 | 4.365096 | 37.137295 | 1399 | 1529 |
| rxjava | 2.524062 | 47.416603 | 2485 | 9380 |
| spring-framework | 5.249460 | 30.516154 | 8664 | 8453 |

Table 1: Average Coupling and Cohesion Metrics for Projects with and without Code Bad Smells

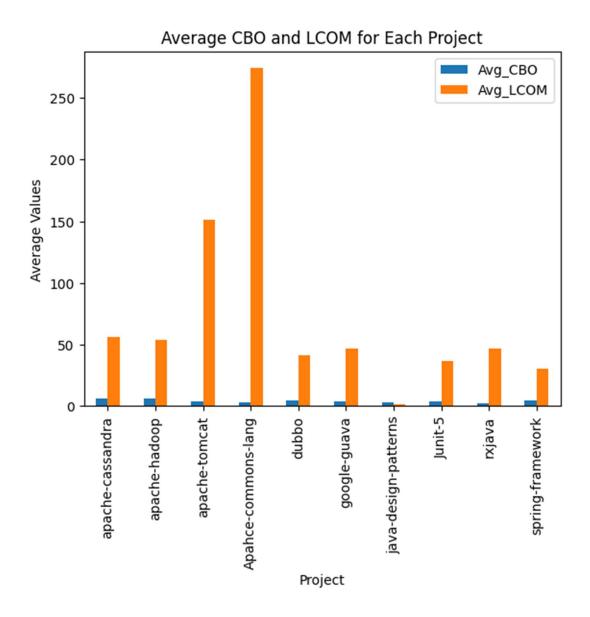


Fig 1: Average Coupling (CBO) and Cohesion (LCOM) for Each Project

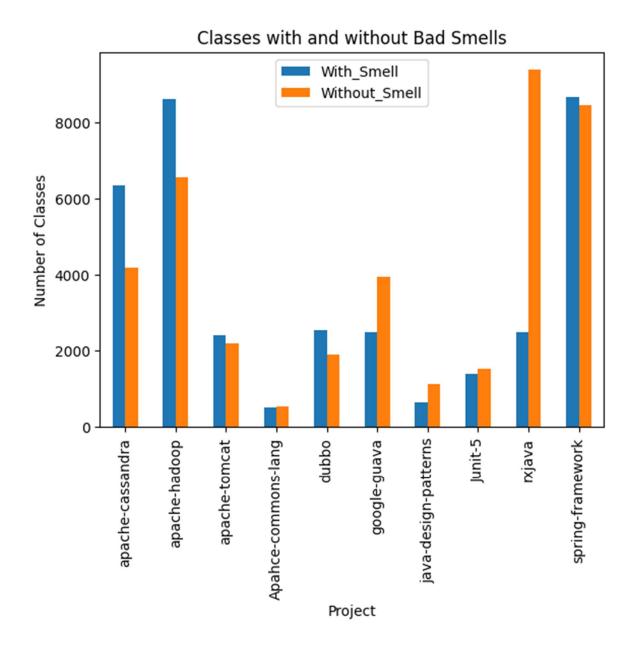


Fig 2: Number of Classes with and without Code Bad Smells in Each Project

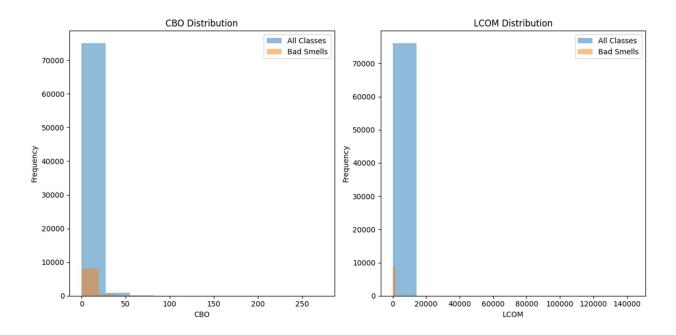


Fig 3: Distribution of CBO and LCOM Metrics for Classes with and without Code Bad Smells

Key Observations:

CBO (Coupling Between Objects):

- Projects like apache-hadoop and apache-cassandra have higher average CBO values,
 indicating a higher level of coupling between classes when code smells are present.
- For some projects like google-guava, the average CBO is lower with code smells, which
 might indicate specific architectural or coding practices that mitigate the impact of smells
 on coupling.

LCOM (Lack of Cohesion in Methods):

Projects such as apache-tomcat and apache-commons-lang show extremely high
 LCOM values with code smells, suggesting that methods in classes are not well-related,
 thereby reducing modularity.

• Interestingly, **rxjava** has a significantly lower LCOM without smells, indicating better method cohesion in the absence of code smells.

Section 5: Conclusion

The analysis of the effect of code smells on modularity in various software projects reveals significant insights into the impact of poor coding practices on system design and maintenance. The study examined two key metrics: Coupling Between Objects (CBO) and Lack of Cohesion in Methods (LCOM), which serve as indicators of modularity and the presence of code smells in the projects. The findings suggest that code smells have a detrimental effect on modularity, influencing both the coupling and cohesion of the codebase.

Firstly, the CBO values highlight the extent of coupling between different classes within the projects. High CBO values are indicative of tightly coupled systems where classes are highly dependent on one another. Projects such as apache-hadoop and apache-cassandra exhibit high average CBO values, especially when code smells are present. This suggests that code smells exacerbate the coupling between classes, making the code more interconnected and harder to maintain. On the contrary, projects like google-guava show lower CBO values even with the presence of code smells, which could be attributed to specific architectural or coding practices that help in mitigating the impact of smells on coupling.

Secondly, the LCOM values shed light on the cohesion within the classes. High LCOM values imply that the methods within a class are not well-related, reducing the overall modularity of the system. The analysis reveals that projects such as apache-tomcat and apache-commons-lang have extremely high LCOM values when code smells are present, indicating a significant reduction in cohesion and, consequently, modularity. In contrast, rxjava shows a markedly lower LCOM in the absence of code smells, pointing to better method cohesion and improved modularity when code smells are minimized.

The presence of code smells in a codebase tends to correlate with higher coupling and lower cohesion, leading to reduced modularity. This reduction in modularity can have several adverse effects on software projects. High coupling makes the system more brittle and less adaptable to changes, as modifications in one class may require changes in several other dependent classes. This interdependency complicates maintenance and increases the risk of introducing bugs. Furthermore, low cohesion within classes means that the functionality of the class is not well-defined, making it harder to understand, test, and reuse.

Moreover, the varied impact of code smells across different projects underscores the importance of context and the specific coding practices employed in each project. Some projects may adopt architectural patterns or coding standards that mitigate the adverse effects of code smells, leading to better modularity despite their presence. For instance, the lower CBO values in google-guava suggest that certain practices can help maintain modularity even when code smells exist.

In conclusion, the effect of code smells on modularity is evident across the studied projects. Higher coupling and lower cohesion are significant indicators of the negative impact of code smells, leading to less modular and more difficult-to-maintain codebases. Addressing code smells through refactoring and adhering to good coding practices is essential for improving modularity and ensuring the long-term maintainability and flexibility of software systems. The study highlights the need for continuous monitoring and management of code quality to mitigate the detrimental effects of code smells on modularity.

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