Unveiling Method-Level Performance Trends

An Empirical Analysis of Java Code Performance Evolution in Open-Source Projects



Polytechnique Montréal Summer/Fall 2024





What is "Performance Evolution"?

```
# Version 1 (Commit 1)
def calculate_total_price(items: list[Item]) -> float:
  total : float = 0
  for item in items:
     total += get_price_from_database(item)
  return total
```

```
# Version 2 (Commit 2)
def calculate_total_price(items: list[Item]) -> float:
    price_cache : dict[Item, float] = {}
    total : float = 0
    for item in items:
        if item not in price_cache:
            price_cache[item] = get_price_from_database(item)
        total += price_cache[item]
    return total
```

And of course, in Java

```
# Version 1 (Commit 1)
public double calculateTotalPrice(List<Item> items) {
    double total = 0;
    for (Item item : items) {
        total += getPriceFromDatabase(item);
    }
    return total;
}
```

Version 1 (Commit 2)
public double calculateTotalPrice(List<Item> items) {
 Map<String, Double> priceCache = new HashMap<>();
 double total = 0;
 for (Item item : items) {
 total += priceCache.computeIfAbsent(item,
 this::getPriceFromDatabase);
 }

return total;

Current Works/Research We're Doing?

"Insights on Method-Level Performance Changes" "Performance-Oriented Software Refactoring"

Upcoming

"Automated Generation of Performance Regression Unit Tests Using Adaptive Instrumentation and Code Analysis"

Primary components in this work



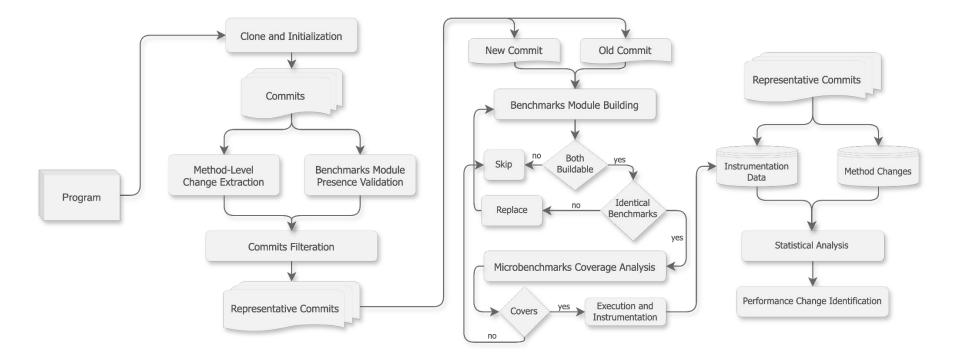
Java Performance Evolution Buddy

The primary **pipeline** for mining, analysis, and benchmarking

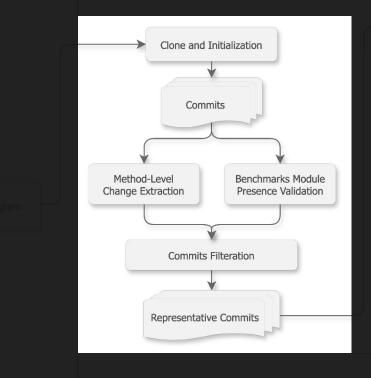


Java Instrumentation Buddy A lightweight instrumenting agent for Java

Overview



Step 1: Project Initialization and Data Collection



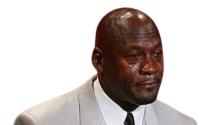
1. Clone the project 2. Iterate through its commits i. Should have at least one valid method-level code change ii. Should have JMH module iii. Should **not** be a **merge** commit iv. Should have valid pom.xml v. CI build should be successful 3. Extract method-level changes -> for before/after the commit 4. Save as Representative Commits

Which **projects** are we analyzing?

Project	Commits	KLoC	Commits with Method Changes	Commits with Benchmark	Representative Commits	Executed Commits	Collected Changed Methods
jetty.project	30,160	339.06	2,472	12,720	2,470	56	124
netty	11,604	216.98	4,241	7,669	4,240	57	97
jdbi	5,709	28.49	1,266	1,919	313	90	136
fastjson2	4,372	178.5	1,726	3,752	1,726	220	615
Chronicle-Core	3,911	13.25	780	3,170	585	2	3
SimpleFlatMapper	3,433	51.79	911	1,969	485	45	68
apm-agent-java	3,066	80.22	891	2,984	889	86	176
zipkin	2,955	23.51	656	2,726	615	46	93
feign	2,063	17.42	351	1,384	229	54	114
protostuff	1,603	42.29	448	1,354	448	4	4
JCTools	1,043	31.48	339	1,042	339	26	52
objenesis	1,049	2.69	107	784	72	12	14
client_java	866	27.38	155	667	154	9	11

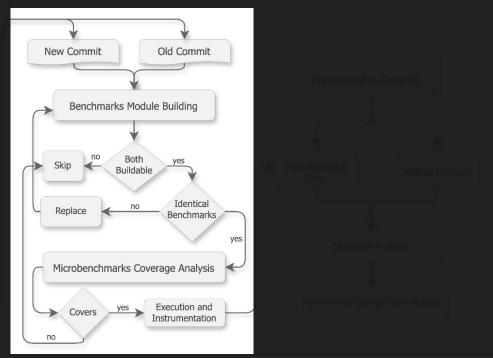
I know it sounds weird, but project building procedure is not an easy task folks...

SPECIALLY FOR JAVA



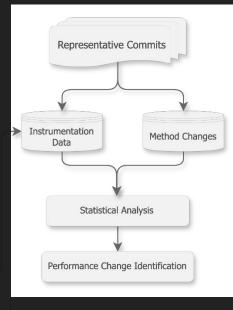
Step 2: Benchmarking and Instrumentation

1. Build JMH benchmarks (bef/aft) -> if any fails, skip! 2. Check for identical benchmarks -> replace with old if not same -> if not compatible, use newer -> if fails again, skip! 3. Get microbenchmarks coverage -> if no coverage, skip! 4. Execute and instrument microbenchmarks



Step 3: Performance Change Analysis

1. With before/after trace data i. Mann–Whitney U Test to check significance ii. Cliff's Delta Effect Size to get significance size 2. Indicate performance change -> improvement? regression? neutral? 3. (Exclusive) Label code change type -> algorithmic? data structure? other? 4. (Exclusive) Analyze the performance trend

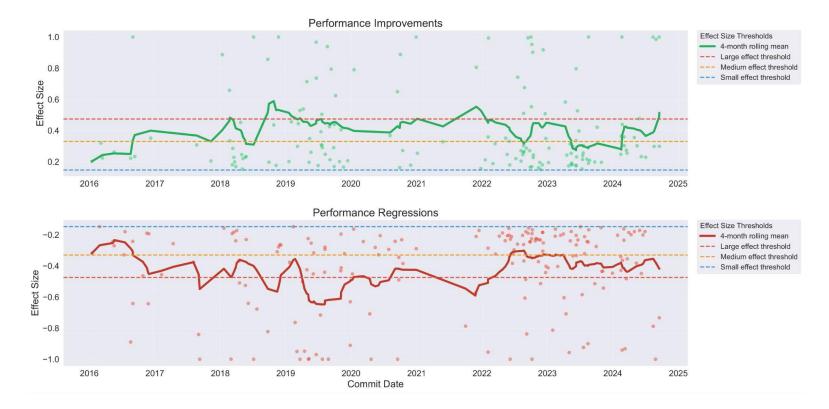


This pipeline (JPerfEvo) is submitted to "International Conference on Mining Software Repositories (MSR) 2025 - Data and Tool Showcase Track"



What Are the Patterns of Performance Changes in Java Projects Over Time?

The distribution of performance changes (i.e., effect size) over time across all projects



The distribution of performance changes (i.e., effect size) over time across all projects.

Performance Improvements

Performance regressions show slightly larger effect sizes than improvements, with both demonstrating increased stability over time

4-month rolling mean
 --- Large effect threshold
 --- Medium effect threshold
 --- Small effect threshold

The distribution of performance change effect size categories based on the performance change type



The distribution of performance change effect size categories based on the performance change type



Distribution of code change impacts on performance across project lifecycle stages

Droiget Stage	Change Type					
Project Stage	Improvement	Regression	Unchanged			
Early	19.20%	21.43%	59.38%			
Middle	14.84%	18.13%	67.03%			
Late	12.76%	16.87%	70.37%			

Distribution of code change impacts on performance across project lifecycle stages

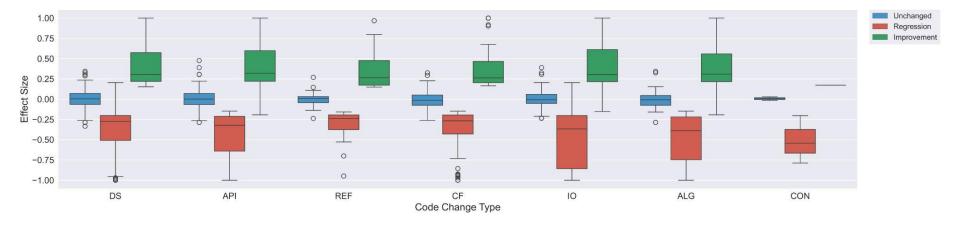
The evolution across life stages shows a trend toward increased stability as projects mature

Late 12.76% 16.87% 70.37%



What is the Correlation Between Code Changes and Performance Impacts, and What Defines Commits with Significant Performance Shifts?

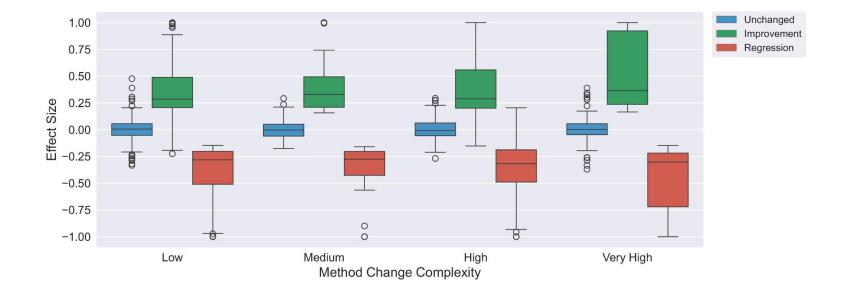
How each change type contributes to performance improvements, regressions, and neutral changes



The distribution of performance changes (i.e., effect size) over time across all projects

API/Library Call and Algorithm Change modifications tend to have the greatest positive impact on performance, while Exception and I/O Handling changes contribute the most to performance regressions

Effect size distribution by method change complexity



The distribution of performance changes (i.e., effect size) over time across all projects

While complex changes are indeed riskier, carefully planned complex modifications may be more likely to yield significant performance gains than cause severe degradations

> Medium High Method Change Complexity

Very High

Performance change distribution (in percentage) based on commiter's experience

Author	Change Type					
Experience	Improvement	Regression	Unchanged			
Junior	13.73%	19.41%	66.86%			
Mid	13.97%	16.44%	69.59%			
Senior	17.22%	17.78%	65.00%			

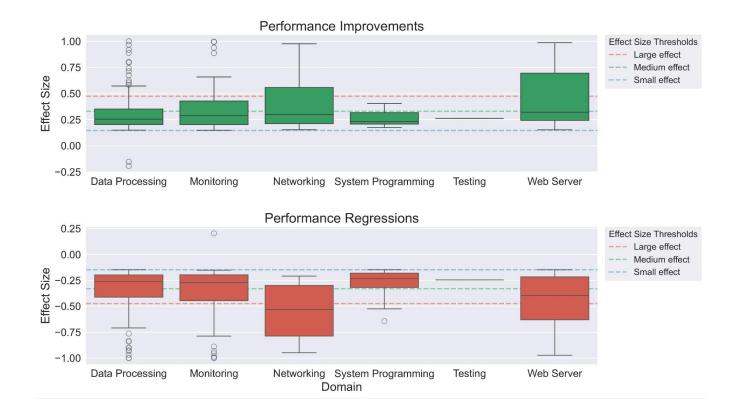
Performance change distribution (in percentage) based on commiter's experience

Balanced performance maintenance may be better achieved through the collaborative work of mid-level developers' careful approach and seniors' optimization expertise

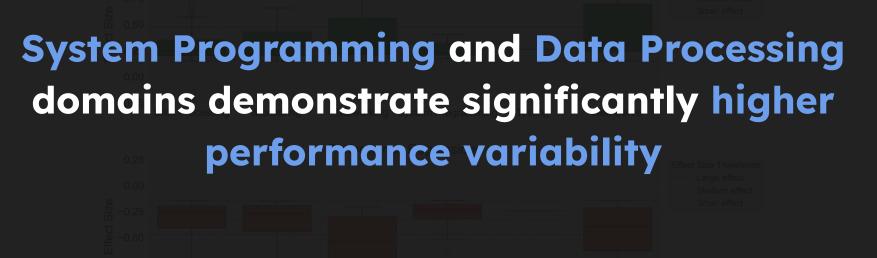


Are There Significant Differences in Performance Evolution Patterns Across Different Domains or Project Sizes?

Performance change effect size in each project's domain

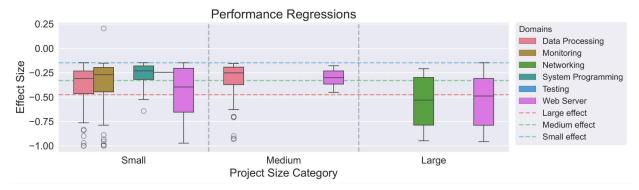


Performance change effect size in each project's domain



Comparison of performance change affected by project's size, indicated for each domain





Comparison of performance change affected by project's size, indicated for each domain

So, small projects need careful handling due to high variability, medium-sized projects benefit from stability, and large projects require proactive measures to prevent regressions. The complete study is also submitted to "Can't say it, it's double-blind"

Java's all grown up—focus on tiny tweaks and watch out for sneaky regressions!

Big algorithm and I/O changes are like juggling chainsaws—great rewards but great risks!

One size doesn't fit all—tailor your performance strategy to your project's quirks, and remember: small projects can cause big surprises! **Thanks!**