

Improving the Functionality and Performance of a Text Privatization Benchmarking Platform

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Outline



- Motivation
- **Problem Statement**
- Analysis
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- Demonstration
- Evaluation
- Conclusion

Motivation – The Privacy-Utility Trade-off



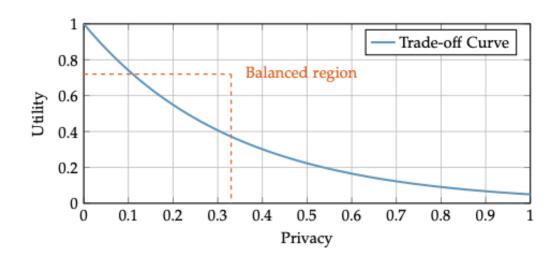


Figure: Illustration of the privacy—utility trade-off.

Source: Adapted from Dwork & Roth (2014).

Context:

- Digitalization (e.g., healthcare, social media) generates massive amounts of sensitive text data [1].
- Regulatory compliance, including guidelines such as GDPR, requires the protection of data.

The Challenge:

- Privacy: Masking identifiers (such as names and ages) and quasi-identifiers to prevent re-identification [2].
- Utility: Preserving semantic meaning for downstream NLP tasks.
- The Trade-off: Increasing privacy typically degrades utility; finding the "Balanced Region" is difficult.

Research Gap:

 Lack of standardized benchmarking platforms that measure both dimensions effectively.

[2] Pilán et al., Applied Soft Computing (2025).

Problem Statement – Limitations of PrivBench



The Starting Point: An initial modular prototype existed, but faced three scaling barriers:

1. Performance Bottlenecks:

- High latency due to "Cold Start" container initialization for every task.
- No resource reuse; heavy models for some modules are reloaded constantly.
- Benchmarking requires efficiency to be practical in production [3].

2. Reproducibility Issues:

- No version control for modules; updating an algorithm overwrote previous results.
- It is not possible to compare results over time (historical data loss).

3. Usability:

- No visibility into task position (queue).
- Manual entry of metadata is required for every submission.

Research Questions



RQ1 (Reproducibility & Standardization):

• How can an existing modular evaluation framework for text privatization be extended to enable the standardized, reproducible, and interoperable measurement of privacy, utility, and performance?

RQ2 (Engineering & Performance):

 How can the platform be designed and engineered to fulfill key software quality attributes such as performance, maintainability, extensibility, and fairness in the benchmarking process?

Analysis of the Existing System



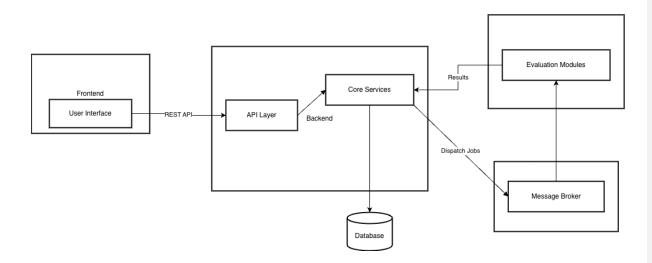


Figure: Component-level architecture of the legacy PrivBench system.

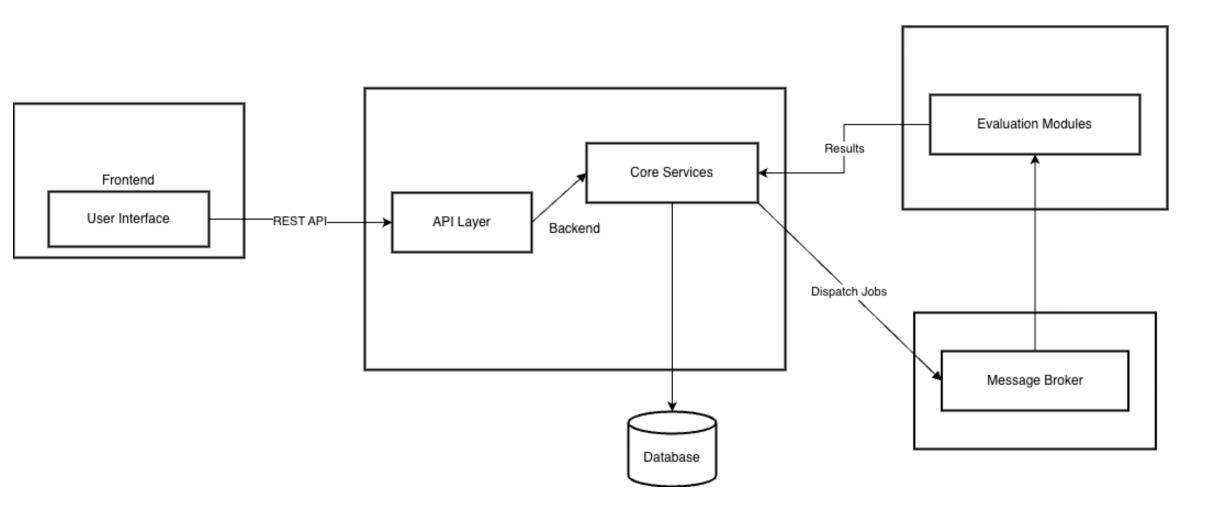
Workflow:

- 1. User uploads submission.
- 2. System spins up a *new* Docker container.
- Module runs.
- 4. The container is destroyed.

Identified Bottlenecks:

- **Operational Complexity:** Repeated container startup/shutdown costs time.
- **Black Box:** No queue visibility for users.
- **Data Traceability (Versioning):** Previous results were overwritten with each module update.





Improved System Architecture



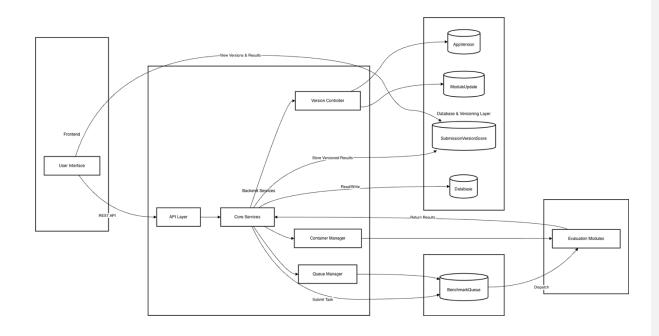


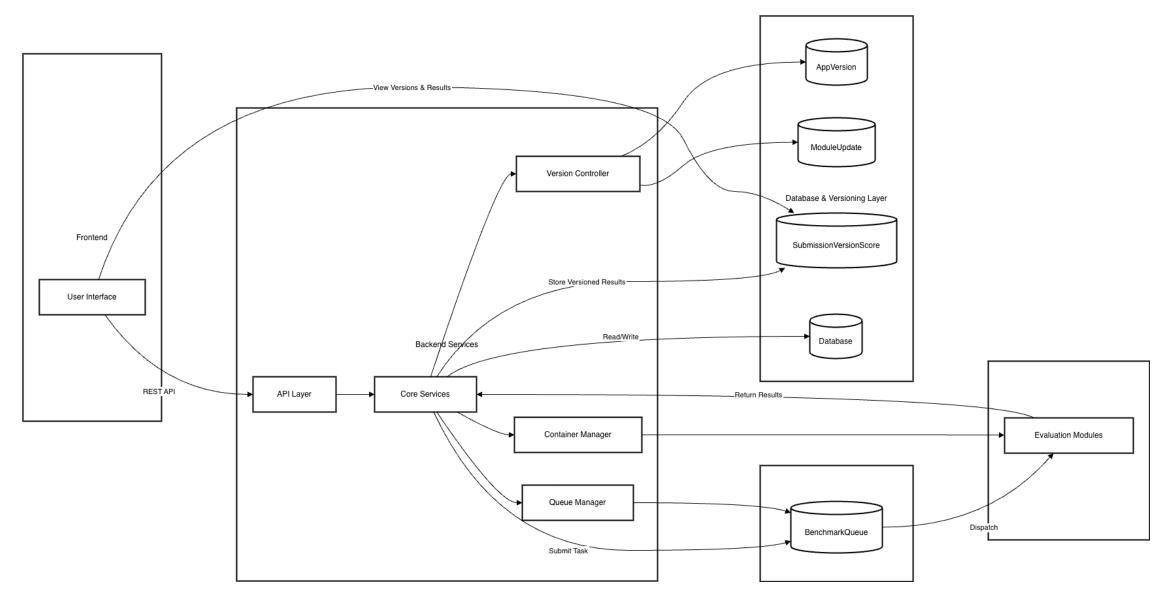
Figure: Component-level architecture of the extended PrivBench system.

New Core Components:

- Container Manager: Orchestrates Docker lifecycles and resource reuse.
- Queue Manager: Implements FCFS scheduling and handles fairness.
- Version Controller: Manages app version and module updates.

Database Extension: Added entities for version tracking and templates.





Performance Optimization (Container Manager)



- Objective: Eliminate initialization latency ("Cold Starts").
- Implementation Strategy:
 - Warm Starts: Containers are initialized once at application startup.
 - Pool Management: Running containers have an "idle" state with models pre-loaded in memory.
 - Execution Flow: Submission -> Assign to Idle Container -> Execute -> Return to Pool.

Benefit and Result: Eliminates initialization latency and model loading overhead.

Scalability & Fairness (Queue Management)



- Objective: Prevent server overload and provide user transparency.
- The Solution: A benchmark submission queue.
- Features:
 - Serialization: Validates and enqueues submissions to prevent overload.
 - **Transparency:** Users see "Position: X" and its corresponding status (Waiting, Processing, or Completed).
 - Fairness: Ensures First-Come-First-Serve execution.
 - User Experience: Enables cancellation of long-running tasks.

Reproducibility (The Versioning System)



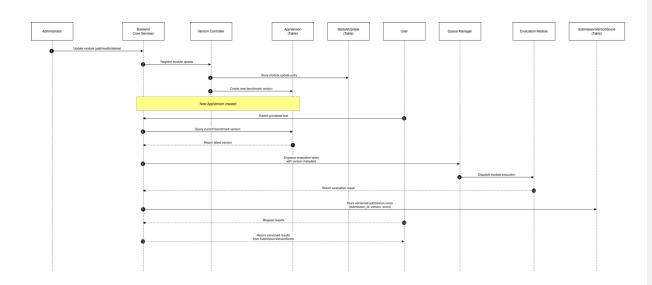


Figure: Sequence diagram illustrating the versioning workflow.

Objective: Ensure results remain valid even as the system evolves.

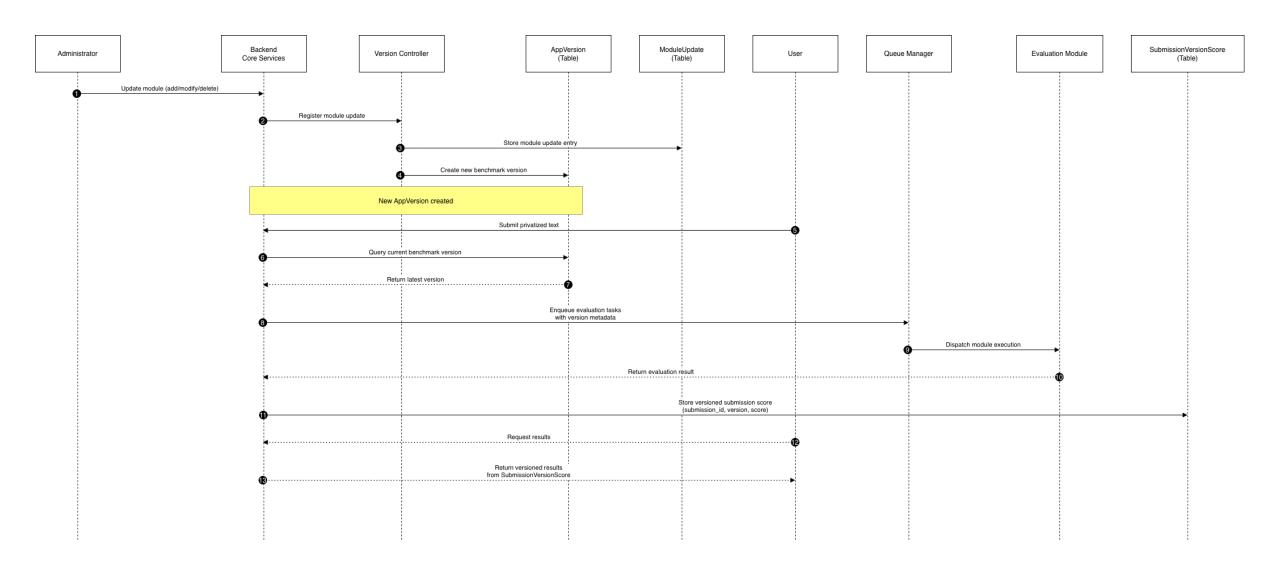
New Data Models:

- AppVersion: Snapshots of the global system state.
- ModuleUpdate: Tracks specific changes (e.g., "v1.1:
 Updated algorithm").
- SubmissionVersionScore links a result to the specific version used during evaluation.

Outcome:

Enables historical comparison even after system updates.

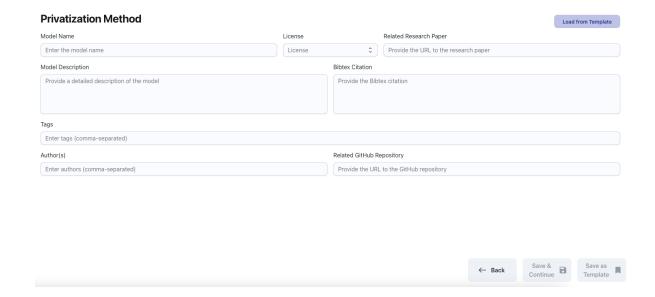


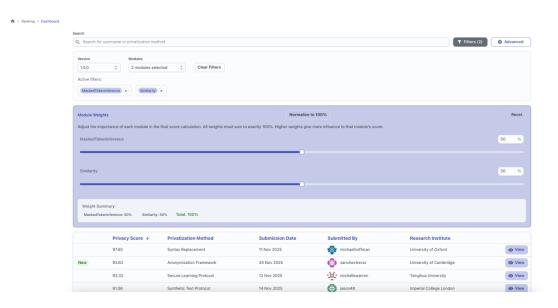


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Usability Enhancements







Metadata Templates:

- Users define model details (such as authors and citations) once and reuse them.
- Reduces friction for repeated experiments.

Version-Aware Leaderboard:

- Filters rankings by benchmark version.
- Ensures fairness by only comparing scores generated under identical module configurations.

System Demonstration



Workflow to demonstrate:

- 1. Submission: Loading a saved metadata template.
- 2. Queuing: Submitting and pointing out the "Position: 1" status indicator.
- **3. Results:** Viewing the final score tagged with "Version 1.1.0".
- 4. Comparison: Filtering the leaderboard to show historical data.

Evaluation Setup



Hardware Environment:

- MacBook Air M1 (8-core CPU, 8GB RAM).
- Represents a standard developer environment.

Workload:

- Dataset: Lightweight dataset (100 text rows) to isolate architecture behavior.
- Repetitions: 10 independent runs per module type.

Test Modules:

- Similarity Module: Lightweight (Transformer-based embeddings).
- MaskedTokenInference: Computationally intensive (Masked Language Model).

Performance Results



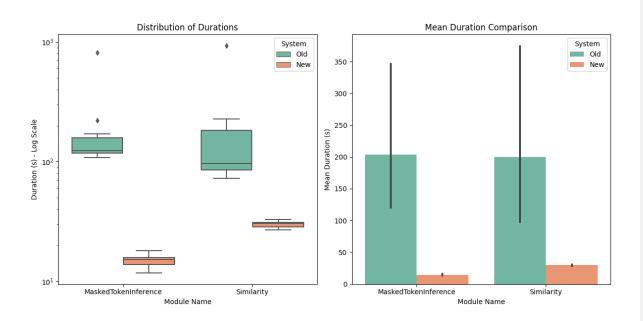


Figure: Performance comparison between the Old and New systems.

Quantitative Findings (Mean Duration): Massive reduction in execution time thanks to warm starts.

- MaskedTokenInference: Dropped from 204s (Old) to 15s (New).
- Similarity: Dropped from 200s (Old) to 30s (New).

Speedup: 6x to 13x improvement depending on module complexity.

Conclusion: The Container Manager successfully eliminated the initialization bottleneck.

Usability Results



Transparency: Verified that users can track real-time queue positions.

Efficiency: Template metadata significantly simplified the submission flow.

Clarity: Versioned leaderboards successfully grouped comparable results, solving the data traceability issue.

Limitations & Future Work



1. Evaluation Environment

- Limitation: Performance benchmarking was restricted to a local CPU environment.
- Future Work: Conduct full-scale stress testing on the GPU-enabled VM architecture.

2. Versioning Scope

- Limitation: System versions module configurations, but not yet the datasets or internal logic files.
- Future Work: Extend the data model to include dataset versioning and module logic tracking for total reproducibility.

3. Scalability

- Limitation: The current architecture utilizes a single Celery worker pool.
- Future Work: Implement distributed worker nodes or Kubernetes orchestration for massive parallel loads.

Conclusion



Summary of Contributions:

- **Performance:** Achieved 6x-13x speedup by shifting from Cold Starts to Warm Starts via Container Management.
- Reproducibility: Built a Versioning System for standardized comparison.
- Fairness & Usability: Implemented Queue Management to handle concurrent submissions.
- Final Status: Transformed PrivBench from a prototype to a production-ready platform.

Q&A



Thank you for your attention.

References



- 1. [Yadav et al., 2023] Data Privacy in Healthcare: In the Era of Artificial Intelligence. *Indian Dermatology Online Journal*.
- 2. [Pilán et al., 2025] Truthful text sanitization guided by inference attacks. Applied Soft Computing.
- 3. [Dwork & Roth, 2014] The Algorithmic Foundations of Differential Privacy.
- 4. [Liang et al., 2022] HELM: Holistic Evaluation of Language Models.

