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HawkEDA: A Tool for Quantifying Data Integrity Violations in Event-driven Microservices

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ABSTRACT

A microservice architecture advocates for subdividing an application into small and independent components, each communicating via well-defined APIs or asynchronous events, to allow for higher scalability, availability, and fault isolation. However, the implementation of substantial amount of data management logic at the application-tier and the existence of functional dependencies cutting across microservices create a great barrier for developers to reason about application safety and performance trade-offs.

To fill this gap, this work presents HawkEDA, the first data management tool that allows practitioners to experiment their microservice applications with different real-world workloads to quantify the amount of data integrity anomalies. In our demonstration, we present a case study of a popular open-source event-driven microservice to showcase the interface through which developers specify application semantics and the flexibility of HawkEDA.

CCS CONCEPTS

• Information systems → Database management system engines.

KEYWORDS

microservice, event-driven architecture, data integrity

ACM Reference Format:


1 INTRODUCTION

The emergence of cloud computing as a paradigm for large-scale deployments has made a case for practitioners to rethink how traditional applications, those following the so-called monolithic architecture, are designed and deployed. Particularly, we are witnessing the growing adoption of the microservice architectural style [10].

In contrast to a monolithic architecture, where application modules are coupled and communicate with each other via function calls, a microservice architecture promotes designing modules as independent applications decoupled from each other. The interaction between microservices occurs via lightweight mechanisms like HTTP-based protocols or asynchronous messages. As these modules are independent, changing, testing, and redeploying a functionality is often less complex since only the affected microservice is involved instead of the entire application. Microservices can also be polyglot, using different programming languages and underlying databases (the so-called decentralized data management principle).

Besides, microservices promote fault isolation, where faults can be constrained to an individual microservice boundary (e.g., through a container-based deployment), curtailing failures from propagating to other architecture components.

1.1 Challenges and Motivation

Although microservice architectures present several benefits, such as fault isolation and scalability of individual components, designing microservices is, however, a non-trivial task.

C1. Asynchronous event streams. Even though microservices are supposed to operate independently, they often present functionality dependencies among each other [7].

Practice #1: To avoid coupling microservices through their defined APIs, practitioners are increasingly relying on asynchronous events [6], often backed up by message queues like Kafka [5], to communicate a need or trigger a computation in another microservice (e.g., as part of a workflow).

Practice #2: In addition, it is often the case where substantial data management logic is implemented at the application-level in microservices [7]. In other words, as the microservice state is oblivious to the underlying databases, the microservice becomes a stateful application.

Asynchronous events in microservice architectures pose challenges to maintaining application safety, particularly on enforcing ordering constraints on processing distinct events generated by different sources. As these events trigger updates in a microservice’s private state, the event processing order plays a significant role in ensuring data integrity.

Ensuring event-based constraints at the applications-level is a complex and error-prone task. Therefore, developers tend to rely on weaker models, such as eventual consistency, that fall apart on enforcing event-based constraints. This first observation leads us to the question #1: How harmful the lack of event-based constraint enforcement is to application safety?

C2. Cross-microservice coordination. Besides, it is usually the case where a microservice needs to coordinate with another microservice to carry out a cross-microservice business transaction.

This may be in two forms: (i) Querying data from a different microservice. For example, in an online shopping application, the order microservice may retrieve the available discounts for a given
group of users from the user microservice. (ii) Locking items belonging to a different microservice. For example, the cart microservice may request the stock microservice to reserve a stock item.

**Practice #3**: Because of Practice #2, application-level concurrency control is a prevalent practice in microservices, creating a great barrier to ensure application integrity.

To achieve higher performance, practitioners end up implementing cross-microservice requests under a weak isolation level and this may lead to data integrity violations subject to the contention level of workloads. The widespread implementation of cross-microservice requests takes us to the question **#2**: What are the trade-offs of encoding weak isolation requests across microservices regarding data integrity and performance?

**Motivation.** It is difficult for microservice developers to experiment with different design decisions (e.g., how to co-locate data and logic across microservices), not to mention measuring their impact on performance and correctness in their applications. In microservices, this becomes more pressing since, instead of only reasoning about the chain of function calls and such implications in a single database state (i.e., a monolithic application), developers have to reason about the complex interplay of events and requests spanning multiple microservices. As a result, they encounter challenges in identifying the most appropriate application safety and performance trade-offs and resorting to suboptimal designs.

### 1.2 Contributions

While existing work has focused on investigating application-level concurrency control [2] and performance inefficiencies [9] in monolithic web applications, no work has focused on supporting microservice developers to reason about the trade-offs on designing microservices. To fill this gap, we present HawkEDA \(^1\)\(^2\), a data management tool to support microservice developers to experiment with and reason about application safety and performance trade-offs of event-driven microservices. To meet this goal, a microservice developer specifies its application through two building blocks: (i) Application invariants – An invariant expresses the application correctness criteria. For instance, a developer may specify that an order containing a product item experiencing unavailability in stock should not observe a payment processed event. The observation of such an event under the specified condition incurs an anomaly. (ii) Application interfaces – Express the interfaces through which the tool can interact with and reason about the application, including the events spanning multiple microservices, their schemas, the APIs to interact with, and the message broker’s (that enables the microservice’s event-based communication) to connect to.

To the best of our knowledge, HawkEDA is the first online monitoring tool targeted for capturing data integrity anomalies in event-driven microservice architectures, allowing for prompt identification of anomalies arisen from concurrency control handled at the application-level. The demonstration of HawkEDA will guide users on the lifecycle of monitoring the complex arrangement of event-driven microservices. We will allow participants to modify the workload characteristics, e.g., data skewness and amount of

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\(^1\)Hawks usually surprise their victims with unexpected behavior, bringing to light the weaknesses and vulnerabilities of the chosen victims in their environment

\(^2\)Available at: https://github.com/LennoxAlexion/HawkEDA
A checkout_cart event is issued by Cart microservice after an user’s checkout request and includes the catalog items requested. A payment_processed event confirms that the order was correctly processed by the Payment microservice. The invariants define that (1) no more than one payment_processed event should be observed per user checkout and (2) overselling catalog items should not be observed. It is noteworthy that invariant 1 is independent and, without a dependency relation, overselling may be incorrectly detected by the tool. According to eShopContainer’s semantics, one can only assert whether an order was placed (and therefore its items) after the payment has been issued. Therefore, to capture such dependency relation, a precedence relation can be established as an invariant, as described by invariant #2, which shows that the tool can only reason about oversold catalog items after a payment is issued.

### Listing 1: Scenario definition example

#### # Workload definition
(i) Load products to Catalog microservice
(ii) Load users to User microservice

#### # Execution definition
Define skewness of requests over data items
Define distribution of users submitting requests
Define amount of time of scenario execution

#### # Events definition
A. checkout_cart ( checkout_id, items[( product_id, ... )] )
B. payment_processed ( payment_id, checkout_id, ... )

#### # Invariants definition
1. SUM ( payment_processed ) = ( 0, 1 ) FOR EACH checkout_id
2. SUM ( checkout_cart.items[i.product_id] where i ∈ items )
   <= <product amount> (ref. to (i))
   \[ \text{AFTER payment_processed WHERE checkout_cart.checkout_id} = \text{payment_processed.checkout_id} \]

### Complex Event Processing
The case for monitoring microservices’ behavior over data streams brings the opportunity to take advantage of a well-established approach for detecting patterns from streams: complex event processing (CEP) [8].

After receiving the user’s input, the scenario definition module sets up a contextual information of the application, including the message broker server information, queues to connect to, and necessary APIs for communicating to. Then, the module builds the CEP queries to detect the constraints specified. The application context is packaged and forwarded to the CEP component.

The queries are then instantiated through the CEP query engine, responsible for continuous processing of application-defined streams. To enable CEP queries in our tool, we use Esper [4], a Java-based library that allows embedding full-featured CEP capabilities in a Java application. The queries are defined via an Event Processing Language, making a sufficient abstraction for expressing events and their relations over time as found in our case.

The stream receiver module is responsible for connecting to a queue service (e.g., message broker), subscribing to specified application streams, and ingesting those streams into the CEP query engine. For last, an event listener module is responsible for receiving the output streams of the defined CEP queries and forward these to the data analysis component. More specifically, the output reflects data integrity anomalies observed in microservices behavior.

### Data Analysis
The resulting streams from the CEP query engine are shipped to the data analysis component, comprised of an event processor module and a data store that accumulates the observed anomalies from experiments ran over time. The event processor module becomes necessary because some detected patterns over the streams must be combined with data retrieved from microservices’ APIs to reason about the existence of anomalies. For instance, when

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**Figure 2: Architecture of HawkEDA**

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3 DESIGN AND IMPLEMENTATION

HawkEDA’s design reflects the characteristics found in event-driven microservice architectures: (i) It should offer appropriate flexibility to adapt to and process the distinct set of event streams exhibited by microservices; (ii) It should allow users to specify what invariants are expected to be enforced by the microservices, such as event ordering constraints, and; (iii) It should allow real-time detection of data integrity violations. HawkEDA is implemented in Java and presents an interactive shell interface so developers can input their application’s specifications. The tool presents a component-based design, exhibited in Figure 2, which we discuss next.

**Scenario Definition.** To support users specifying their application invariants and characteristics in the tool, a fundamental building block is a scenario. For instance, a user may specify that event1 is expected to precede any occurrence of event2 given a specified key present in both event schemas, thus specifying a dependency relation. Besides, a scenario also specifies whether the tool runs continuously or not, an important feature to support users testing their event-driven microservices in staging environments before deployment in "production."

Every scenario class must implement the ScenarioInterface to instantiate the queries operating over the application event streams, set up the required scenario environment, including but not limited to cleaning memory buffers from past executions and creating synthetic data for session-based testing. For instance, a scenario should define the characteristics of the workload such as data skewness, dataset to operate over, and scale factor of the underlying microservices databases. Such fine-tuning of parameters is allowed through the interface definition, enabling users to create an artificial load that reflects their real-world scenarios.

An example scenario definition for C2 is shown in Listing 1. A checkout_cart event is issued by Cart microservice after an user’s checkout request and includes the catalog items requested. A payment_processed event confirms that the order was correctly processed by the Payment microservice. The invariants define that (1) no more than one payment_processed event should be observed per user checkout and (2) overselling catalog items should not be
is invited to specify that the system should prevent items from being oversold, an interaction taking place between Order and Catalog microservices. In the end, the users will be able to respond to question #2. In this sense, it is noteworthy that depending on the concurrency and skewness experienced by the application, handling concurrency control at the application-level may be a suitable trade-off for many microservice applications. Figure 3 exhibits that, up to 500 concurrent requests, a uniform distribution does not face a substantial amount of anomalies.

5 STATUS AND FUTURE WORK

At the time of this writing, HawkEDA only supports the specification of application semantics through the implementation of a Java-based interface. The interface exposes generic methods that are used to instantiate a scenario. In the future, we intend to extend the tool to support the specification of an application scenario through loosely structured file formats, like JSON, to achieve higher flexibility and reduced complexity for the developers.

Besides, to make HawkEDA more appropriate for a broader audience, we intend to integrate the tool with a graphical user interface while providing more advanced features such as historical analysis of anomalies observed in each application version. Lastly, although different workloads can be specified through a scenario in our tool, we intend to support the specification of volatile workloads within a single scenario to cope with applications with higher fluctuation in data distributions.

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REFERENCES


