



# Cognitive and Self-Adaptive System for Effective Distributed-Tracing (Using Jaeger, Open Tracing)

Presenters - Akash Gusain & Mitul Tandon

# **Observability in Software Development**



# In a world where software complexity is the new norm, observability is our superpower, turning chaos into clarity

### What is Observability?

- Provides end-to-end visibility into complex software environments.
- ◆ Enables teams to diagnose issues and enhance performance.

### Why is Observability Important?

- Helps solve problems faster.
- Supports better decision-making and smarter workflows.
- Improves overall customer experiences.

### **Key Types of Observability Data**

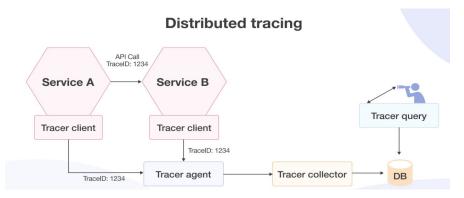
- Metrics
- Events
- ◆ Logs
- Traces

## The Role of Tracing in Observability



### What are Distributed Traces?

Distributed tracing is a technique that tracks the journey of requests as they flow through microservices, providing a detailed view of interactions and performance.



### Why Do We Need Tracing?

- ◆ Essential for understanding performance in cloud-based and microservices architectures.
- Provides insights into how requests are processed, helping to quickly identify bottlenecks and errors.
- ♦ Enhances the ability to maintain and optimize complex distributed applications.

# Tracing the History of Distributed Tracing



**2010**: *Dapper* - Google releases a paper on Dapper, its internal distributed tracing infrastructure, emphasizing its value for developer and operations teams.

**2012:** *Zipkin* - Inspired by Dapper, Twitter launches Zipkin as an open-source tool for distributed tracing to address latency challenges.

**2014:** *Kubernetes* - Although not a tracing tool, Kubernetes transforms cloud computing and accelerates the growth of cloud-native projects, making tracing essential for distributed systems.

**2015**: *Jaeger* - Uber introduces Jaeger, an open-source distributed tracing system for monitoring and profiling microservices, later accepted as a hosted project by CNCF.

**2015:** *OpenTracing* - CNCF welcomes OpenTracing, aimed at providing consistent, vendor-agnostic APIs for managing loosely-coupled microservices.

**2017:** *OpenCensus* - Google unveils OpenCensus, a library set for gathering metrics and distributed traces, allowing real-time data analysis for application health.

**2019:** OpenTelemetry - OpenCensus and OpenTracing combine to create OpenTelemetry, offering a unified framework for distributed tracing, with V1.0.0 launched in 2021 to ensure stability.



# **Working of Dapper**

### **Key Terms:**

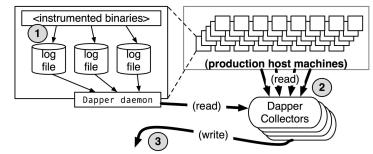
- **Span**: A unit of work within a trace that records details like timestamps, IDs, and annotations.
- **Trace**: The path of a single request through a distributed system, capturing its operations and interactions.
- **Sampling** A technique used to collect a subset of traces to reduce overhead and manage data volume, while still providing insights into performance.
- Latency Time it takes trace data to propagate from instrumented application binaries to the central repository.

### **Dapper Sampling Strategy**

**Purpose of Sampling**: Control overhead and manage trace data size in high-throughput environments.

#### Points considered:

- Head based sampling approach.
- Aggressive sampling frequency.
- Balance trace collection with performance management.
- Control size of trace data.



trace id	span 12	span 23	span 34	span 45	span 56	
123456	nil	nil	<data></data>	<data></data>	nil	
246802	<data></data>	nil	nil	nil	<data></data>	
357913	nil	<data></data>	nil	nil	nil	

(Central Bigtable repository for trace data)

### **Current Problems**



- → DAPPER based tracing implementations such as Zipkin and Jaeger implement a trace-sampling strategy which randomly samples only 1-5% of the traces.
- → Distribution is heavily skewed towards the normal/consistent execution-traces missing out on "unusual/interesting" execution-traces.
- → "Head-Based" sampling policies that adopt random sampling decision at the start of the execution flow.
- → Storage overhead if sampling rate is increased.
  - ◆ Increase Sampling Rate -> Need more storage
  - ◆ Lower Sampling Rate -> Lose out on capturing Error traces
- → From our analysis, over 98.1% of the saved traces returned an HTTP status code of 2xx (OK). The remaining 18 status codes collectively accounted for just 1.9% of the traced executions significantly underrepresenting the "unusual traces".

# Cognitive and Self-adaptive Solution



A "tail-based sampling" approach which makes it feasible for the tracing system to use intelligent filtering techniques in order to persist interesting and meaningful traces.

### → 100% sampling rate

Microservices publishes all the span events to Kafka. This is done asynchronously to minimize performance overhead.

### → Span-aggregator

Obtains spans from kafka and stitches all matching spans to re-generate the execution trace.

### → Adaptive-Sampler

Each trace is assigned a specific sampling rates using the adaptive sampler engine.

Receives the sampling rate from the Trace-Clusterer and decides whether to persist the trace in the Trace-Storage Server based on this rate.

# Cognitive and Self-adaptive Solution



### → Trace-Clusterer

Utilizes a density-based clustering algorithm (DBSCAN) to cluster incoming traces in real-time. It groups and categorizes traces while calculating a Sampling-Rate for each cluster. This rate helps ensure meaningful traces are captured, and clusters are dynamically adjusted based on sampled traces.

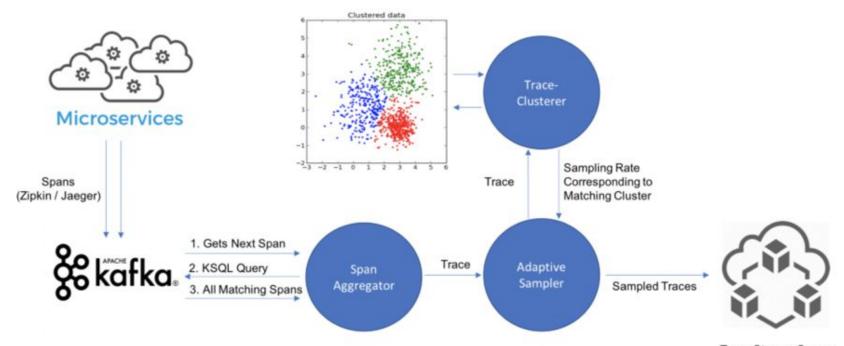
Cluster data points - Maintains a list of feature-vectors and data points corresponding to the 'N' recent sampled Spans and assign cluster for each new datapoint using spatial clustering.

Calculate Weighted Sampling Rate (WSR) – formula which gives inverse weight to less collected data points based on the clustering due to feature vectors

Operation	Operation	Service	Span-1	Span-1	Span-1	Span-2	
Method	Path	Name	hostnam	duration	status	hostname	
			е		code		

# Visualizing the solution





Trace Storage Server (Zipkin / Jaeger Server)

### Conclusion



- The proposed solution has the potential to significantly enhance trace quality by effectively sampling relevant data, thereby improving the representation of interesting and unusual traces.
- This solution is especially advantageous for services with high Service Level Objectives (SLOs), where traditional sampling methods frequently introduce biases towards usual/normal executions.
- Due to the loosely coupled nature of this approach, it can be further developed as a standalone service or plugin.
  This flexibility allows organizations to integrate and instrument their systems with OpenTelemetry-compliant architectures seamlessly.

# Thank You for the time and attention!



Leave a feedback

Looking forward to connect with you to exchange more ideas, collaborations and opportunities!

#### Akash Gusain

- <u>aakashgusain11@gmail.com</u>
- in <u>linkedin.com/in/akash-gusain-2b375119b</u>

### Mitul Tandon

- mitultandon@gmail.com
  - linkedin.com/in/mitul-tandon-240842177