

F5 NGINX

# Adding OpenTelemetry to Modern Adaptive Apps

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NGINX

# Modern Adaptive Apps



*Are defined by their capabilities not their implementation.*



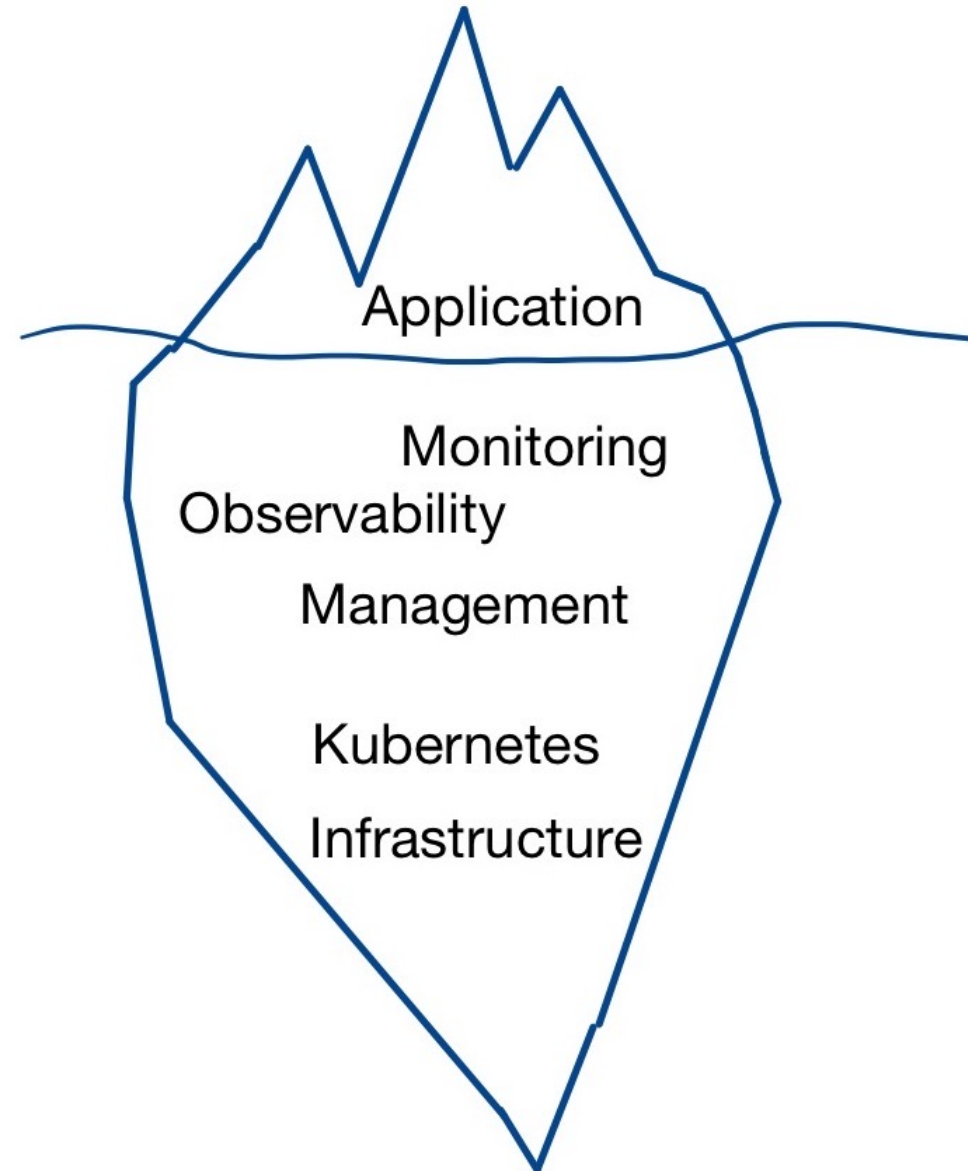
*We speak about them using terms like portable, scalable, observable, reproducible, and debuggable.*

- Deploy in minutes
- **Quickly find performance bottlenecks**
- Easily change hosting provider
- Scale up/down quickly
- Gracefully degrade upon failure
- Serve the customers' needs
- **Provide answers to platform engineers' questions**
- Protect itself from attacks
- Manage state in a knowable way
- **Provide context on errors and crashes**
- Costs scale reasonable with consumption

# What is our target?

A microservices architected app using Kubernetes that aims to be as production ready as possible.

*The application is only one part of the modern application deployment*



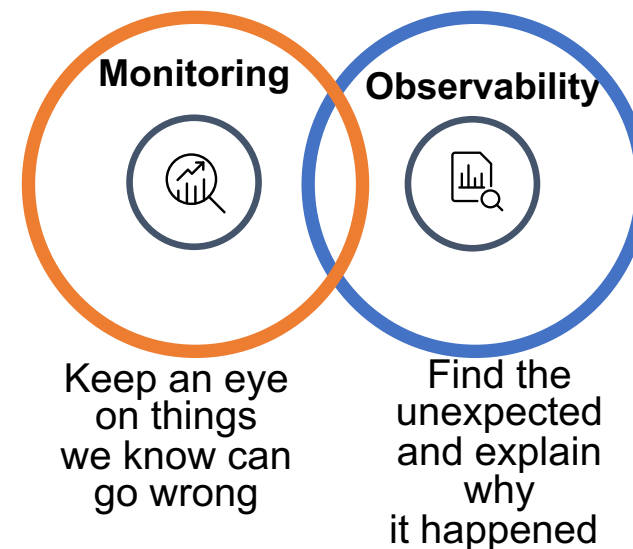
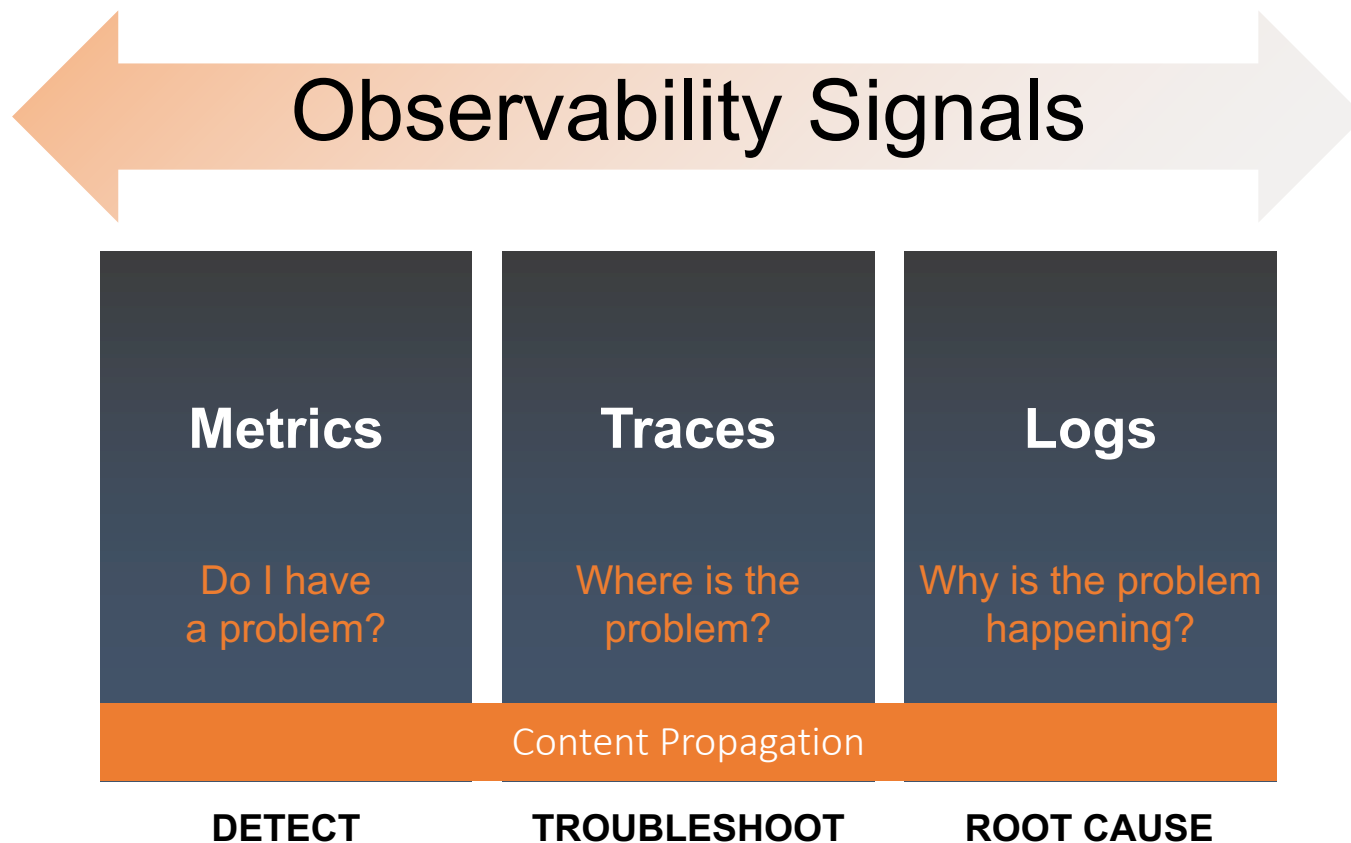
# Observability is a data problem

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*The more observable a system, the quicker we can understand why it's acting up and fix it*

# Observability

Observability helps detect, investigate and resolve the *unknown unknowns* – FAST



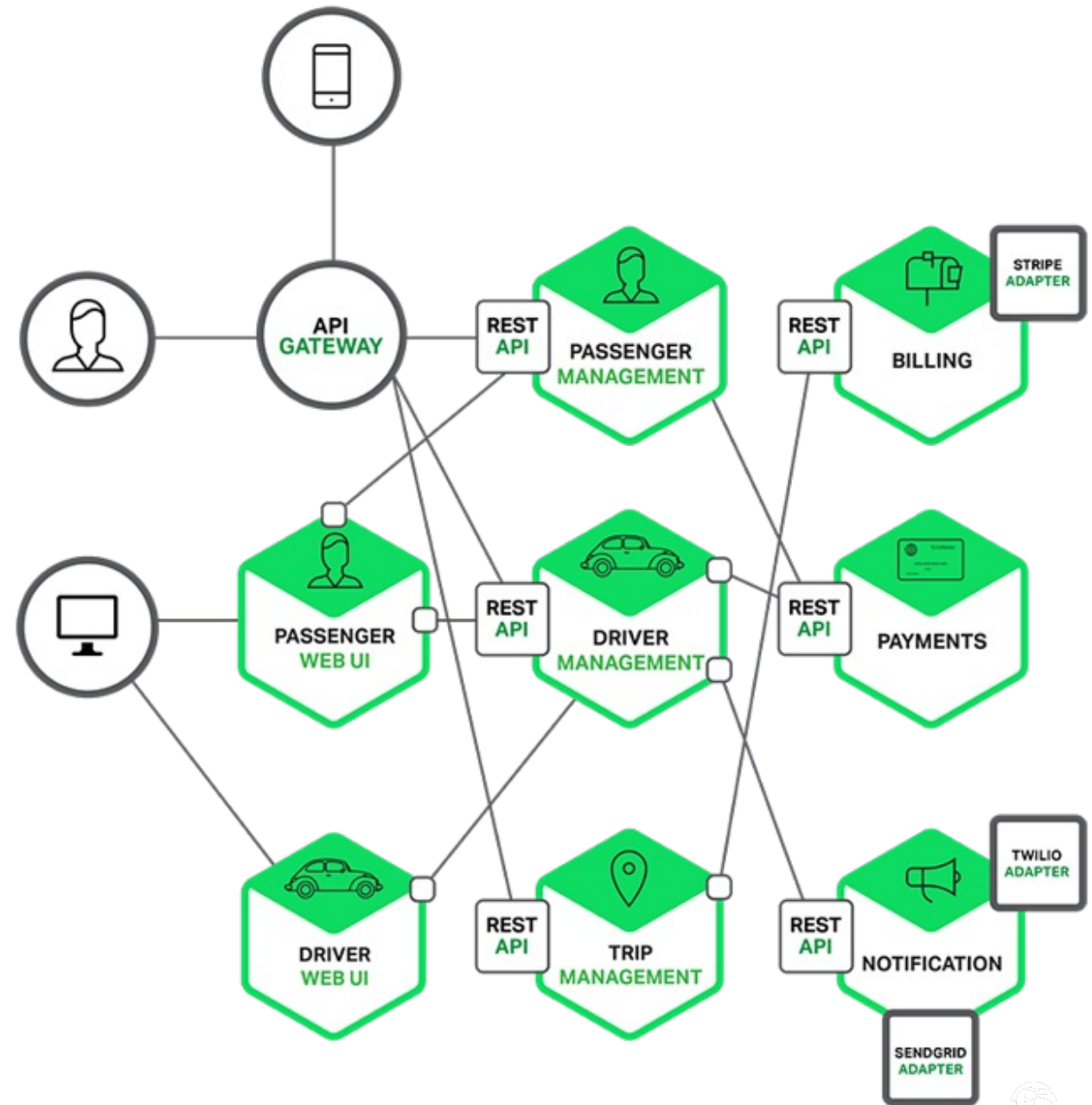
- Better visibility to the state of the system
- Precise and predictive alerting
- Reduces Mean Time to Clue (MTTC) and Mean Time to Resolution (MTTR)

# So why Observability?

## Microservices!

Single application composed of many loosely coupled and independently deployable smaller services

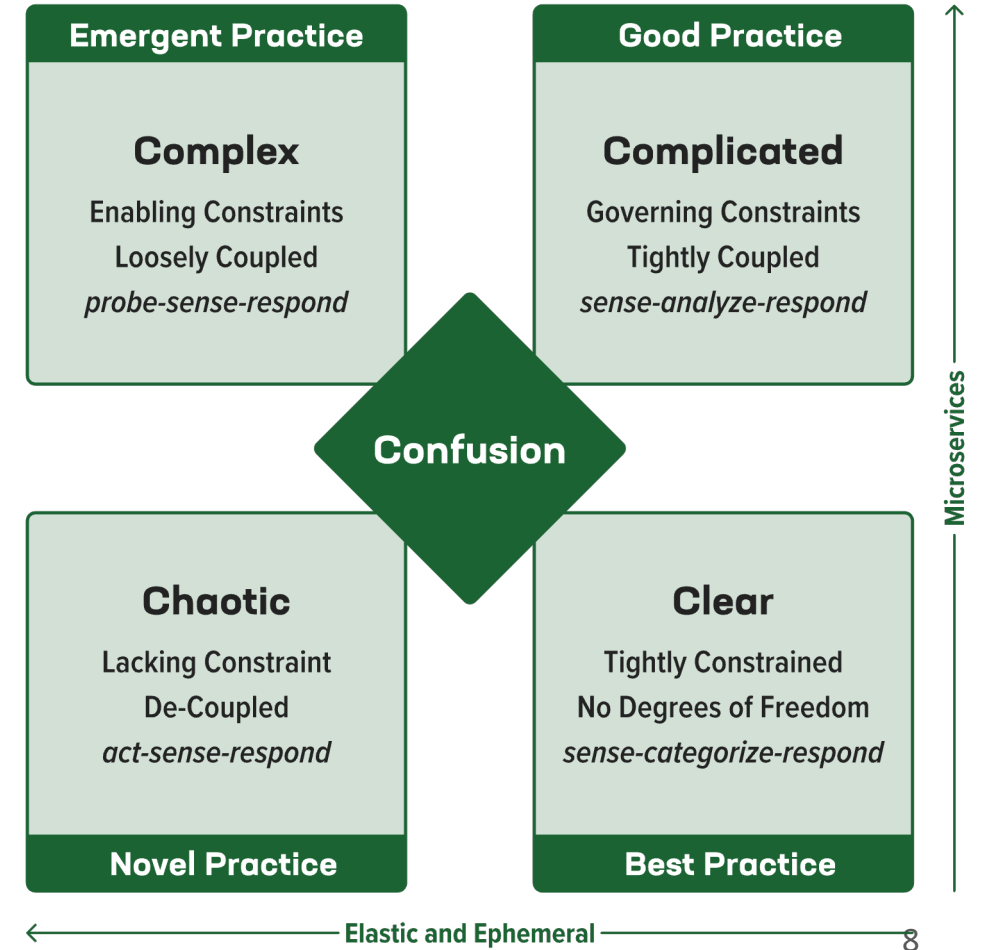
- Often polyglot in nature
- Highly maintainable and testable
- Loosely coupled
- Independently deployable
- Often in Cloud environments
- Organized around business capabilities
- Each potentially owned by a small team



# But They Add Challenges

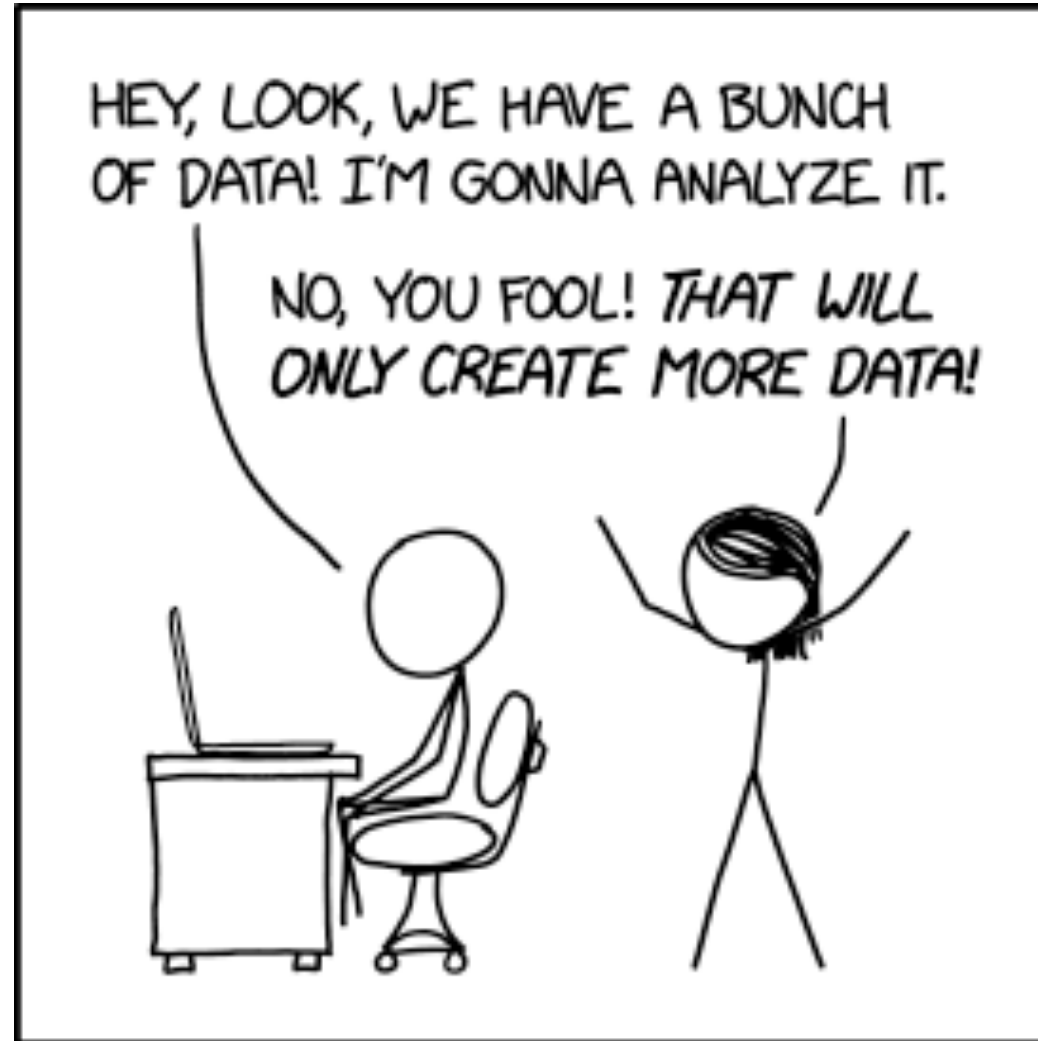
Especially when we consider this in a cloud

- Microservices create complex interactions.
- Failures don't exactly repeat.
- Debugging multitenancy is painful.
- So much data!





# The problem data - there's so much of it



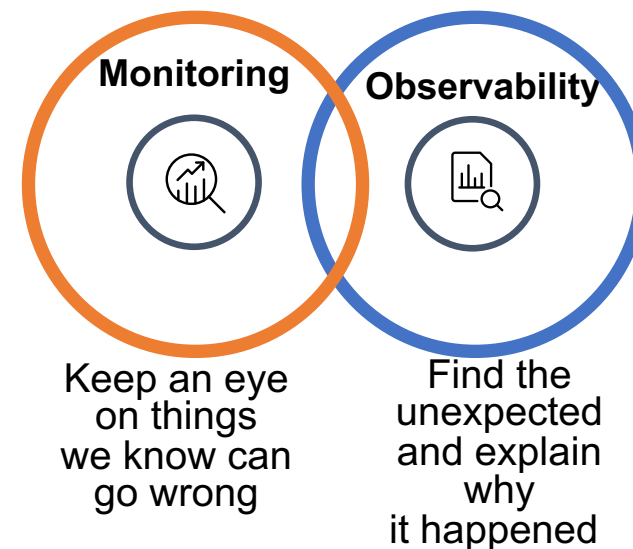
CC-2.5-BY-NC

XKCD

[xkcd: Data Trap](#)

# Observability -

Observability helps detect, investigate and resolve the *unknown unknowns* – FAST



- Better visibility to the state of the system
- Precise and predictive alerting
- Reduces Mean Time to Clue (MTTC) and Mean Time to Resolution (MTTR)

# Identifying the needs : Wishlist

Technology	Logging	Tracing	Metrics	Error Agg	Health Checks	Runtime Intro	Heap/Core Dumps
Elastic APM	Yes	Yes	Yes	Yes	Yes	No	No
Grafana	Yes	Yes	Yes	Yes	Yes	No	No
Graylog	Yes	No	No	No	No	No	No
Jaeger	No	Yes	No	Yes	No	No	No
OpenCensus	No	Yes	Yes	No	No	No	No
OTel	Beta	Yes	Yes	Yes	Yes	No	No
Prometheus	No	No	Yes	No	No	No	No
Statsd	No	No	Yes	No	No	No	No
Zipkin	No	Yes	No	Yes	No	No	No

# Time to get Qualitative

The OpenTracing project is *archived*. [Learn more.](#)  
[Migrate to OpenTelemetry](#) today!

OpenCensus and OpenTracing have merged into OpenTelemetry!

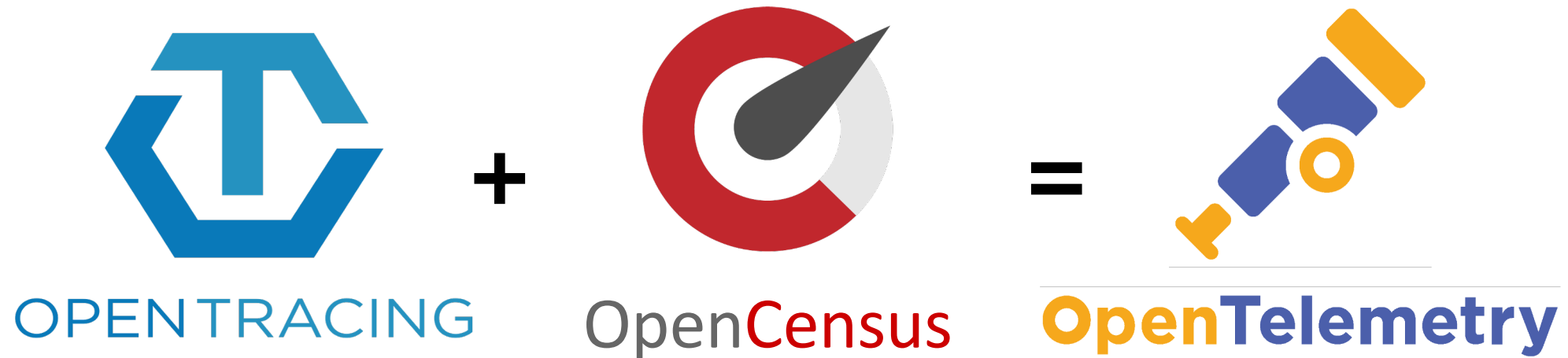


## OpenCensus

*Easily collect telemetry like metrics and distributed traces from your services*

# What is OpenTelemetry?

- Standards-based agents, cloud-integration
- Automated code instrumentation
- Support for developer frameworks
- Any code, any time



# Cloud Native Telemetry

Telemetry “verticals”

Telemetry  
“layers”

	Tracing	Metrics	Logs, etc
Instrumentation APIs	foreach(language)		
Canonical implementations	foreach(language)		
Data infrastructure	collectors, sidecars, etc		
Interop formats	w3c trace-context, wire formats for trace data, metrics, logs, etc		

# Why Open Telemetry Matters

OpenTelemetry users build and own their collection strategies, without vendor lock-in.

OpenTelemetry puts the focus on **analytics** not **collection**.



# MARA and Open Telemetry: The Vision

1

Integration of  
OTEL tracing  
into MARA

2

Convert  
Metrics to use  
OTEL

3

Convert Log  
Management  
to use OTEL

4

Operationalize  
OTEL



# Our Considerations

- Avoid Lock In
  - Ability to switch between observability technologies
- Ease of Use
  - Reduction in friction for implementation
  - Automated instrumentation when possible
- Visualization Tooling
  - Ability to use and correlate data to make decisions
- Resource Use

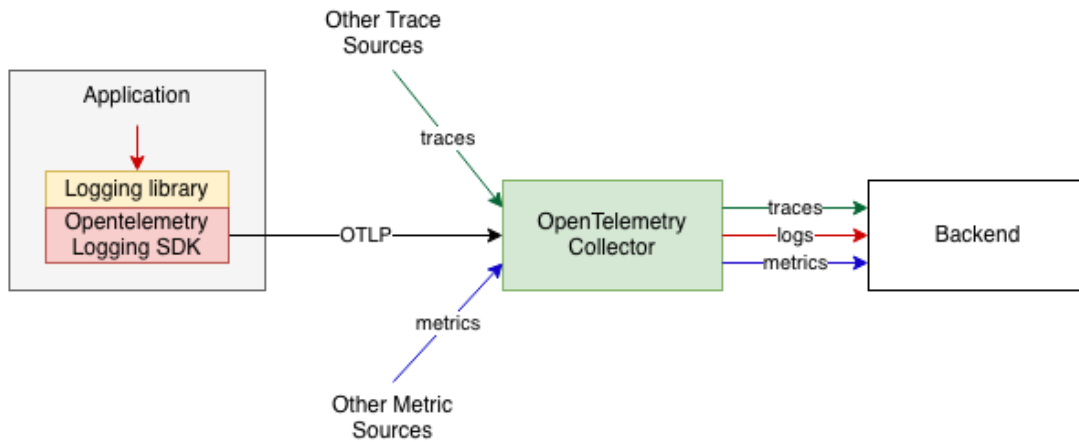
# Let's start with logs

We all know what logs are but their simplicity rapidly leads to some complex decisions

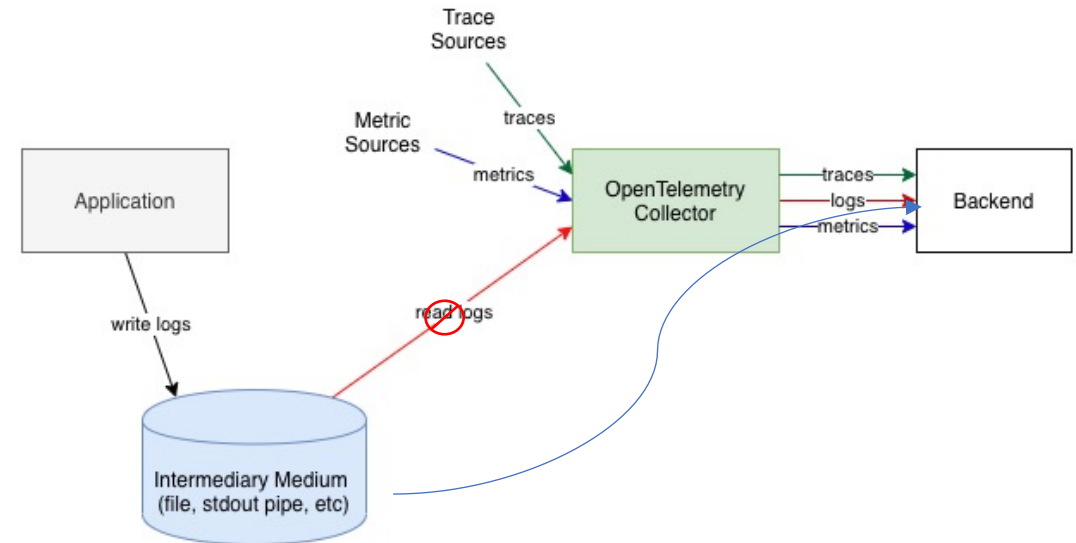
- Simple - Harvest the log output
- Complicated – Transport, Storage, indexing (?), lifetime

To be useful, our log files must be easily searchable based on varying criteria

# What we wanted

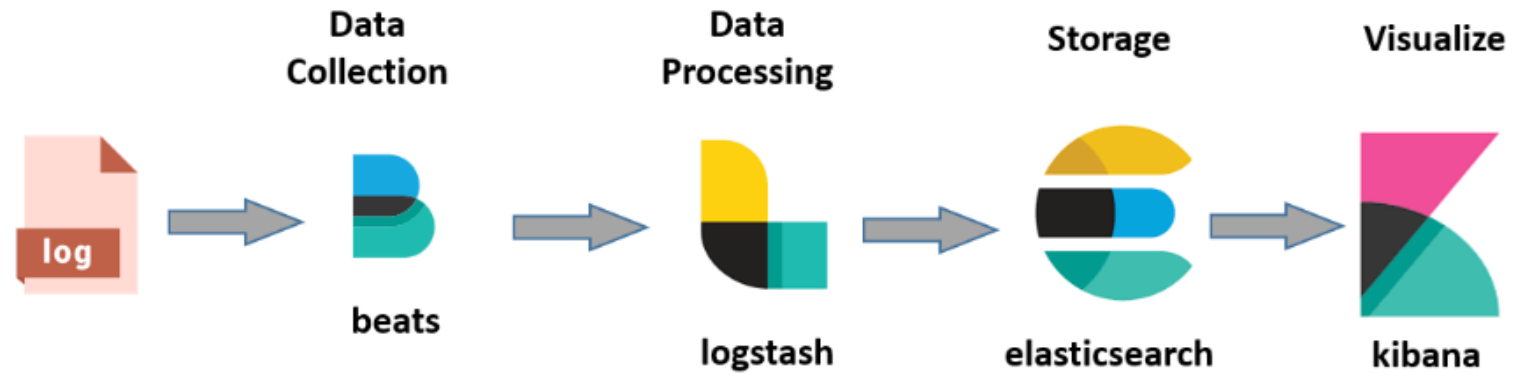


# What we got



# Elastic Stack, for now

- Filebeat – data transport in Kubernetes DaemonSet
- Bitnami – chart to split the deployment into ingest, co-ordinating, master and data nodes
- Kibana – search + pre-loaded stuff

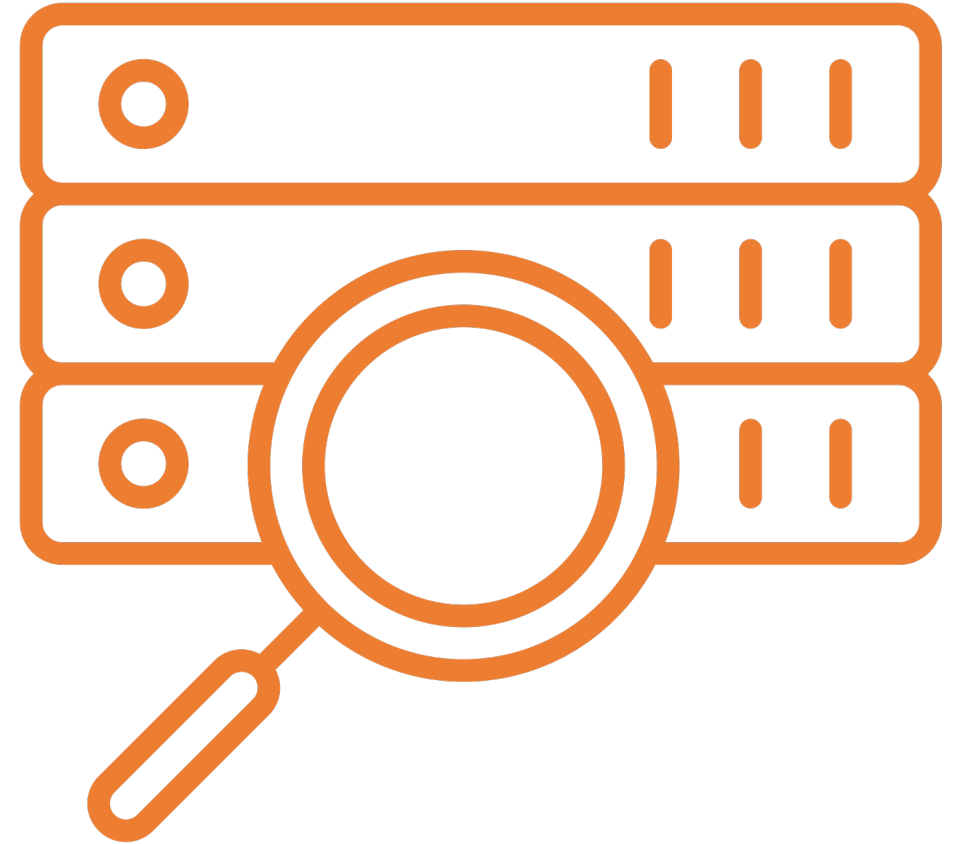


It works, but

- Extremely resource hungry
- Query variance is okay

# Metrics: Prometheus and Friends

- Deployed via helm
  - kube-prometheus-stack
- Chart Deploys
  - Prometheus Operator
  - Prometheus Rules
  - Grafana deployment and dashboards
  - Does not deploy IC dashboards
- Uses service monitors for dynamic scraping
- Configured to work with statsd
  - Deployed to monitor Prometheus
- Available to all MARA components



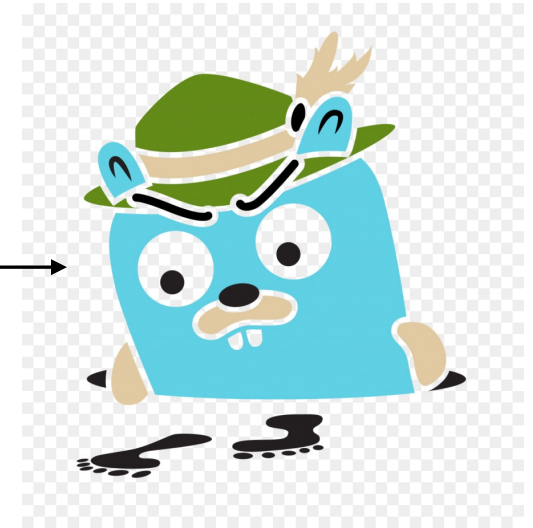
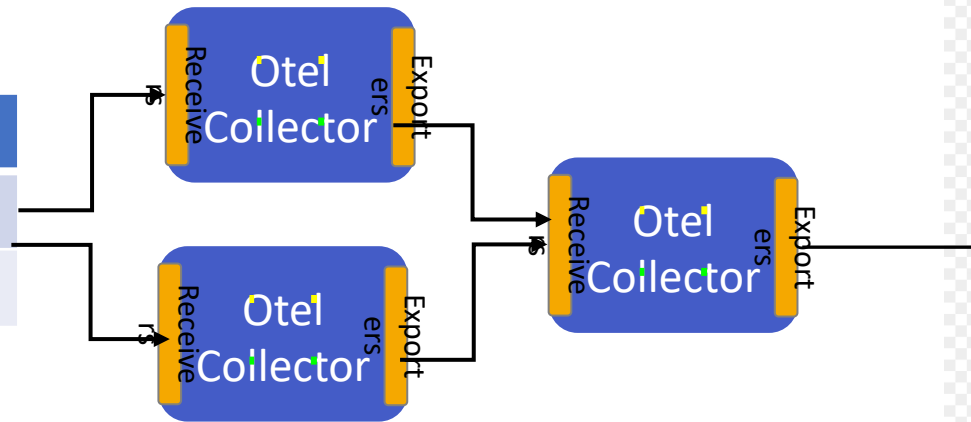
# Distributed Tracing

Complex and semi-chaotic

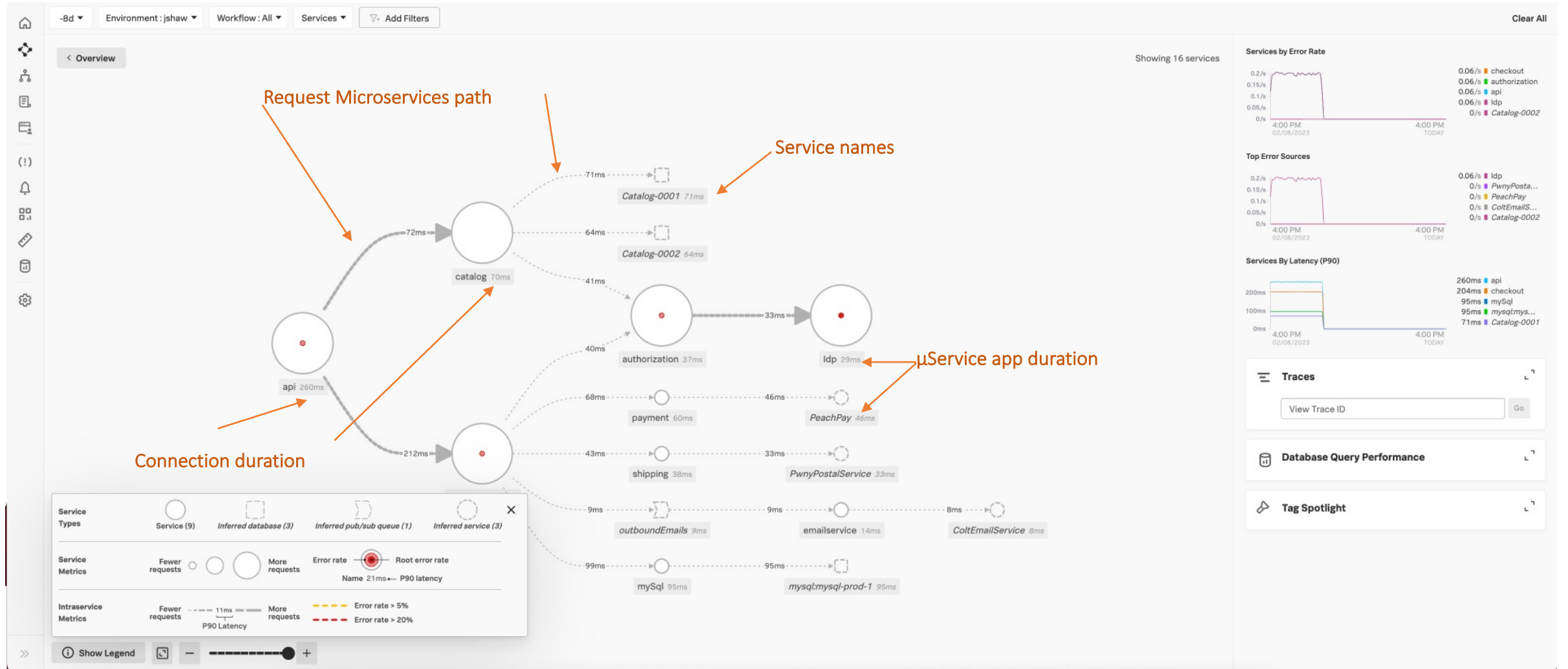
- Must not impact QoS of the application
- Must support all desired languages

- OTEL operator is deployed
  - Several options for collector deployment
  - Includes log to console, Lightstep, Sumo Logic
  - All POC
- App (s) instrumented for tracing
- OTEL is still developing

Language	Framework	Services
Java	Sprint Boot	3
Python	Flask	3



# Let's look at a trace



The screenshot displays the Jaeger UI interface for analyzing a distributed trace. The main view is a waterfall chart showing the sequence of spans in a trace. The trace is for the service 'api' and the operation '/catalog'. The spans are as follows:

- api: /catalog** (Total duration: 113ms)
  - api: http.Request** (33ms)
    - catalog: /auth** (32ms)
      - catalog: http.Request** (31ms)
        - authorization: /auth/valid** (28ms)
          - authorization: http.Request** (25ms)
            - Idp: authRequest** (22ms)
              - api: http.Request** (76ms)
                - catalog: /catalog/{itemId}** (74ms)
                  - catalog: http.Request** (71ms)
                    - Catalog-0001: http.Request** (71ms)
                      - catalog: http.Request** (61ms)
                        - Catalog-0002: http.Request** (61ms)


The right sidebar provides a 'Performance Summary' with a bar chart showing the distribution of time across different components:

    - Application:** 19.0% (represented by a red bar)
    - Database:** 76.1% (represented by a blue bar)
    - Network:** 4.9% (represented by a green bar)

Global Tags for the trace include 'Workflow Name' and 'api:/catalog'.

The diagram illustrates the relationship between three variables: Service names, μService app duration, and μService total performance. Service names is shown as a box on the left, μService app duration as a box in the middle, and μService total performance as a box on the right. Arrows point from Service names to μService app duration and from μService app duration to μService total performance, indicating a flow or dependency.

Note the 2 spans makes up the trace duration (almost)



Note the 2 spans makes up the trace duration (almost)





contend /home

108.78ms

776.07ms

775.22ms

445.81ms

| Category   | Value    |
|------------|----------|
| Category 1 | 144.42ms |

36.74ms

| Method   | Time (ms) |
|----------|-----------|
| Method 1 | 452.11ms  |

| Task   | Time (ms) |
|--------|-----------|
| Task 1 | 451.26ms  |

| Method   | Time (ms) |
|----------|-----------|
| Method 1 | 326.23ms  |

61.43ms

67.91ms

51.27ms





83.8ms

83.18ms

# But it's all about the language

Python

- Pretty straightforward
  - Added 2 files
  - Updated requirements.txt to include dependencies

|   |  |  |
|---|--|--|
|  <b>dekobon</b> feat: use bunyan output for json logs with python apps ... |  |  |
| ..  |  |  |
|  Dockerfile  | feat: add trace ids to Python logs                     |  |
|  gunicorn.conf.py  | feat: use bunyan output for json logs with python apps |  |
|  tracing.py  | feat: add trace ids to Python logs                     |  |

# And some were harder than others

## Java

- Straight Java / Greenfield = Not too bad
  - Import the libraries and use the APIs
- With Spring, life looked easy
  - Use Spring Cloud Sleuth
    - Adds trace/span IDs to Slf4j
    - Instruments common ingress and egress points
    - Adds traces to scheduled tasks
    - Can generate Zipkin traces
- But at that time:
  - Autoconfig was a milestone release, and supported old OTel versions
  - We needed to pull from Spring Snapshot due to coded dependency references

# Answer: a common telemetry module

- Extended tracing functionality
  - Provided Spring enabled autoconfiguration classes, adding more trace resource attributes
  - Built a NoOp implementation to let us disable tracing
  - Added a trace name interceptor to standardize our trace names
  - Added an error handler to output errors both to logs and traces
  - Enhanced the implementation of tracing attributes (service name, instance id, machine id, etc)
  - Built a tracing statement inspector to put trace ids into comments that precede SQL statements

We also extended the reach to Apache by creating a Spring compatible HTTP client

All this was integrated using the OpenTelemetry NGINX module (beta)

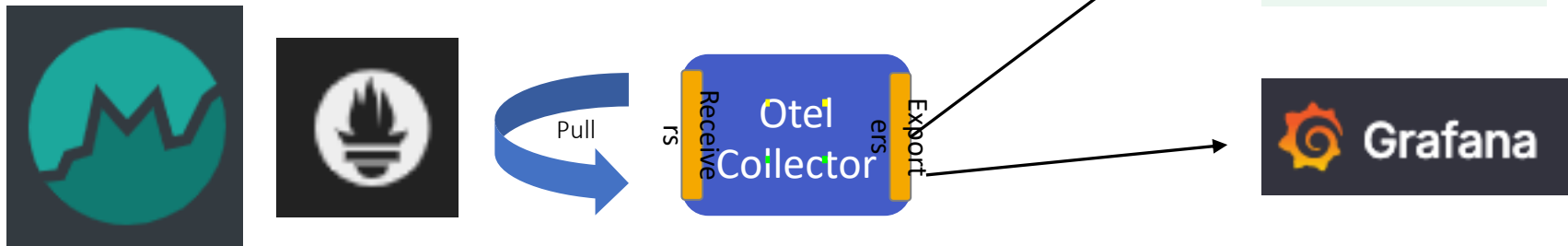
[nginxinc/nginx-unsupported-modules: Container builds of unsupported NGINX modules \(github.com\)](https://github.com/nginxinc/nginx-unsupported-modules)

# Counting Metrics

- We skipped Python
- Java required some study and decisions.
  - Original code used Micrometer/Stackdriver
- We found there were some significant limits to OTel for metrics in JVM
- Micrometer is a mature metrics layer for JVM
  - It is the default metrics API in Spring

# OpenTelemetry and Micrometer

- OTEL Collector can use metrics from *anything*
  - Prometheus, Statsd, OTLP, etc
- Micrometer supports a lot of back ends
  - Prometheus, Statsd, etc
- OTEL Metrics SDK / JVM supports sending metrics via OTLP



# Known Challenges

- Maturity of components
  - Some components still in beta
- Early in life cycle of other components
  - Instrumentation type
  - Manual versus Auto-instrumentation
- Operational concerns
  - Resource concerns (cpu / mem)
  - Performance impact
  - Long term data storage concerns
- Standards / Best Practices



# What's Next

- Load and Performance testing
- Track and adopt OpenTelemetry enhancements (LOGS!)
  - Log data model is pretty settled
  - Stanza work within the Collector means can handle logs.
- Extend OpenTelemetry to other open source NGINX projects
- Test OpenTelemetry solution versus various analyzers



# TL;DR

## Metrics, Traces and Logs

- All took different approaches to get what we wanted
- The Collector was our friend

Metrics and Traces had some interesting gotchas

This is a snapshot in time; things have changed

Auto-instrumentation is great

- When it works
- And gives you what you want



<https://github.com/nginxinc/kic-reference-architectures>

Try it for yourself



[nginxinc/bank-of-sirius: Bank of Sirius \(github.com\)](https://github.com/nginxinc/bank-of-sirius)

Questions?

Thanks for attending

<https://www.linkedin.com/in/davemc>

