ECE 473/573 Cloud Computing and Cloud Native Systems Lecture 22 Observability

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Observability

Tracing

Metrics

Logging

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This lecture: 11

Next two lectures: batch and stream processing

- MapReduce: Simplified Data Processing on Large Clusters https://research.google/pubs/pub62/
- Resilient Distributed Datasets: A Fault-Tolerant Abstraction for In-Memory Cluster Computing http://people.csail. mit.edu/matei/papers/2012/nsdi_spark.pdf

Outline

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Observability

The need to understand our systems better.

- Complexity of software comes from complex requirements.
- Good software design needs good visibility into systems.
- No amount of fancy frameworks or protocols can solve the problem of bad software.

 Observability: the ability to infer system's internal states from knowledge of its external outputs. E.g.

- What does that error message mean and what triggers it?
- Why the performance is not as expected?
- While logging may be available for general troubleshooting purposes, is it possible to answer specific questions that the developers haven't thought of yet?

Evolution of Traditional Monitoring

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Traditional monitoring focuses on the "known unknowns"

- Identify/predict expected or previously observed failure modes.
- Work well for simple systems through trial and error.
- Require code updates that is not flexible.
- However, understanding and monitoring all possible failure (or non-failure) states in a complex system is impossible.
 - Scale of data is beyond human brain power and attention span.
 - Non-deterministic behaviors are difficult to reason with.
 - Interactions between component faults and system failures are very complicated.
- Monitoring shows that system is not working and observability answers why it is not working.

Three Pillars of Observability

Tracing: details from one request to its response consisting of all functions called and messages communicated.

E.g. arguments and return values and time spent.

 Metrics: numerical data points representing system states at specific points in time.

E.g. CPU/memory/disk/network usage.

- Logging: appending records of noteworthy events to the log for later review or analysis.
 - But how can we manage and search many log files for specific information?
- A synergy of the three leads to better observability.

When? What? Where? Why?

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- Track requests as they propagate through the system.
 - Not limited to function calls within a specific process or thread.
 - Need to consider queues and communications across process, network, and even security boundaries.
 - Help to pinpoint component failures, identify performance bottlenecks, and analyze service dependencies.
- Model of requests: spans and traces
 - A request may consist of many works and addition requests, that are running recursively and parallelly.
- Span: a unit of work from beginning to end.
 - Identified by a name with start/end times.
 - Model heirarchy of works and requests as nested spans.
 - Model causal relationships as ordered spans.
- Trace: collection of spans and their relationships.

Tracing with OpenTelemetry

```
const serviceName = "foo"
func main() {
 setupTracerProvider()
 tr := otel.GetTracerProvider().Tracer(serviceName)
 ctx, sp := tr.Start(context.Background(), "main") // Start the root span
 defer sp.End() // End completes the span
 SomeFunction(ctx)
7
func SomeFunction(ctx context.Context) {
 tr := otel.GetTracerProvider().Tracer(serviceName)
 _, sp := tr.Start(ctx, "SomeFunction")
 defer sp.End()
  ... // Do something MAGICAL here!
}
```

- Record begin of span at the beginning of a function.
 - Usually the function name is used for the span.
- Make use of defer to record end of span.

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Tracing with OpenTelemetry (cont.)

```
func setupTracerProvider() {
  stdExporter, err := stdout.NewExporter(
    stdout.WithPrettyPrint(),
  )
  jaegerEndpoint := "http://localhost:14268/api/traces"
  serviceName := "fibonacci"
  jaegerExporter, err := jaeger.NewRawExporter(
    jaeger.WithCollectorEndpoint(jaegerEndpoint),
    jaeger.WithProcess(jaeger.Process{
      ServiceName: serviceName,
    }).
  tp := sdktrace.NewTracerProvider(
    sdktrace.WithSyncer(stdExporter),
    sdktrace.WithSyncer(jaegerExporter))
  otel.SetTracerProvider(tp)
}
```

- Use a remote exporter to collect spans for a single request across multiple servers.
- Use multiple exporters so the information can be found in convenient locations like local logs.

```
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```

Additional Tracing Features

```
Attributes and events can be added to spans.
   span.AddEvent("Canceled by external signal",
     label.Int("pid", 1234),
     label.String("signal", "SIGHUP"))
     Attributes are key-value pairs.
     Events are points in time.
Autoinstrumentation is available as wrappers for many
   popular libraries.
   func main() {
     // http.HandleFunc("/", helloGoHandler)
     http.Handle("/", otelhttp.NewHandler(
       http.HandlerFunc(helloGoHandler), "root"))
     log.Fatal(http.ListenAndServe(":3000", nil))
   }
```

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Metrics

- Collection of numerical data about a component, process, or activity over time. E.g.
 - Computing resources: CPU, memory used, disk/network I/O
 - Infrastructure: instance replica count, autoscaling events
 - Applications: request count, error count
 - Business metrics: revenue, customer sign-ups
- Metrics consist of data points as samples.
 - Sample should have a name, a value, and a timestamp, possibly annotated with labels as key-value pairs.
 - A set of samples form a time series that can be visualized and analyzed, e.g. for anomaly detection.
- Push vs. Pull metric collection.
 - Applications push metrics to collector: simple but needs scaling mechanisms like message queues.
 - Collector contact applications to pull metrics back: more flexible and allow ad-hoc inspections, but less friendly for service discovery, load balancer, and ephemeral services.

Metrics with OpenTelemetry

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```
func main() {
  . . .
 prometheusExporter, err := prometheus.NewExportPipeline(prometheus.Config{})
 mp := prometheusExporter.MeterProvider()
 otel.SetMeterProvider(mp)
 http.Handle("/metrics", prometheusExporter)
 log.Fatal(http.ListenAndServe(":3000", nil))
}
 Prometheus is an open source monitoring and alerting toolkit
```

Use a pull model over HTTP to scrape metric data Manage them in its time series database.

Synchronous Instruments

```
var requests metric.Int64Counter
func buildRequestsCounter() error {
 meter := otel.GetMeterProvider().Meter(serviceName)
 requests, err := meter.NewInt64Counter("fibonacci_requests_total",
   metric.WithDescription("Total number of Fibonacci requests."),
 return err
7
var labels = []label.KeyValue{
 label.Key("application").String(serviceName),
 label.Key("container_id").String(os.Getenv("HOSTNAME")),
7
func Fibonacci(ctx context.Context, n int) chan int {
 requests.Add(ctx, 1, labels...)
 // The rest of the function...
}
```

Call buildRequestsCounter in main to initialize the counter requests that is used later.

Synchronous Instruments (cont.)

```
func updateMetrics(ctx context.Context) {
 meter := otel.GetMeterProvider().Meter(serviceName)
 mem, _ := meter.NewInt64UpDownCounter("memory_usage_bytes",
   metric.WithDescription("Amount of memory used."),
 goroutines, _ := meter.NewInt64UpDownCounter("num_goroutines",
   metric.WithDescription("Number of running goroutines."),
  )
 var m runtime.MemStats
 for {
   runtime.ReadMemStats(&m)
   mMem := mem.Measurement(int64(m.Sys))
   mGoroutines := goroutines.Measurement(int64(runtime.NumGoroutine()))
   meter.RecordBatch(ctx, labels, mMem, mGoroutines)
    time.Sleep(5 * time.Second)
 }
}
```

- Metrics can be measured and recorded in a periodic manner.
- Int64UpDownCounter allows to record metrics that can increase or decrease.

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Logging

- Why can't we just use fmt.Printf (and so on)?
 - Easy to provide lots of context-rich data for a component.
- It is difficult to extract information from verbose and unstructured logs.
 - In particular at scale, when you are interested in logs from many, but not one, components.
- ► To generate and store logs consumes CPU and I/O resources.
 - Without careful planning, could easily consume significant amount of resources.
- How to store logs for access at scale?
 - Many services we have discussed and will discuss are created for processing logs!

Structured Logging

- Treat logs as streams of events.
 - Instead of lines in files that should be read by humans.
 - Applications generate events for logs.
 - Underlying infrastructure takes care of routing, storage, indexing, and analysis.
 - Developers could still read logs as lines in files for development but will access them differently, e.g. through database queries, in production.
- Structured logging: describe events as key-value pairs.
 - There is no need to generate human friendly lines that need to be parsed later.
 - Instead of,

```
2020/11/09 02:15:10AM User 12345: GET /help in 23ms
2020/11/09 02:15:11AM Database error: connection reset by peer
Store logs in JSON,
```

```
{"time":1604888110, "level":"info", "method":"GET", "path":"/help", '
{"time":1604888111, "level":"error", "error":"connection reset by peed
```

Logging with Zap

```
logger, err := zap.NewProduction()
if err != nil {
    log.Fatalf("can't initialize zap logger: %v", err)
}
logger.Info("failed to fetch URL",
    zap.String("url", url),
    zap.Int("attempt", 3),
    zap.Duration("backoff", time.Second),
)
```

Zap is a popular open source logging library.

Known for its speed and low memory usage.

- In particular with strong typing, through a bit awkward to use.
- The Sugar method provides a easier but slower interface.

```
logger, _ := zap.NewProduction()
sugar := logger.Sugar()
sugar.Infof("failed to fetch URL: %s", url)
```

 Make your application observable by integrating OpenTelemetry solutions.