Prometheus a practical workshop vi.i

Tomer Gabel, November 2024

The Pillars of Observability

1. Logs

2. Metrics

3.Traces

The Pillars of Observability

1. Logs

Our focus today.

2. Metrics

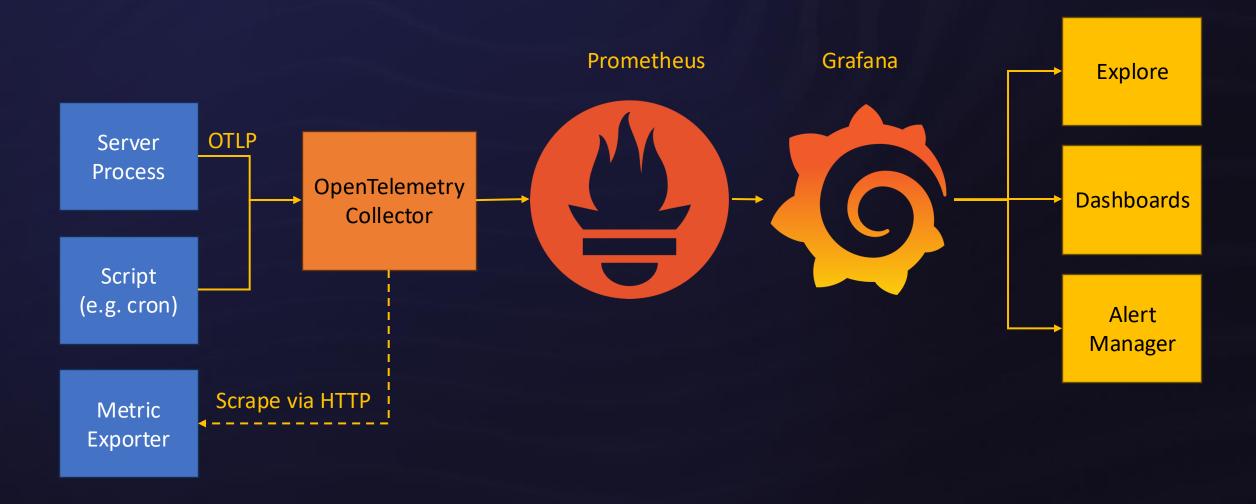
3.Traces

Metrics:

Answer *quantitative* questions

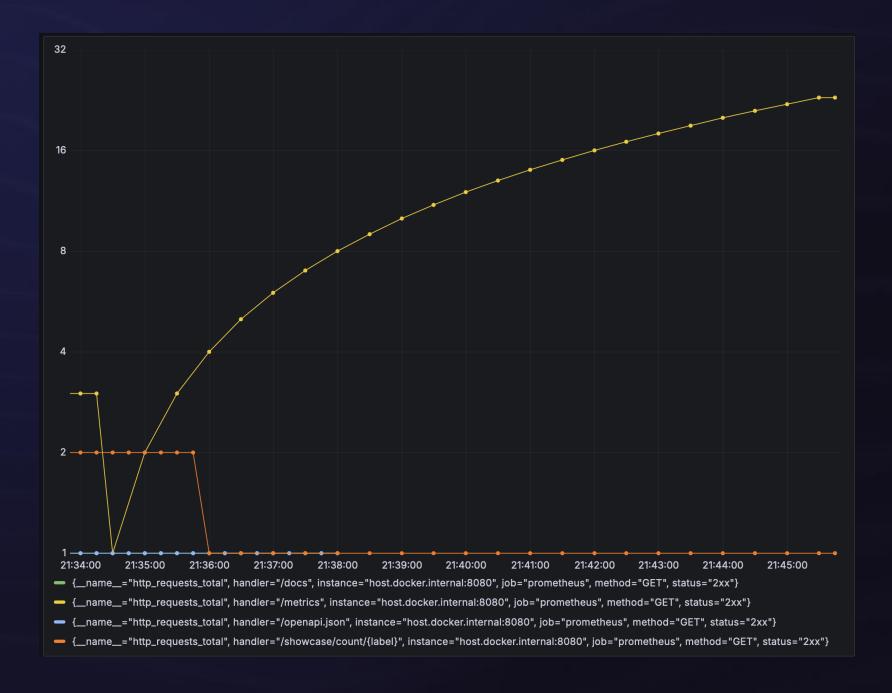
- 1. How many requests did I get?
- 2. How long did they take to process?
- 3. How much memory am I using?
- 4. How much free space is left on the disk?

Enter: Prometheus



Metrics 101

- 1. Name
- 2. Labels
- 3. Time
- 4. Value



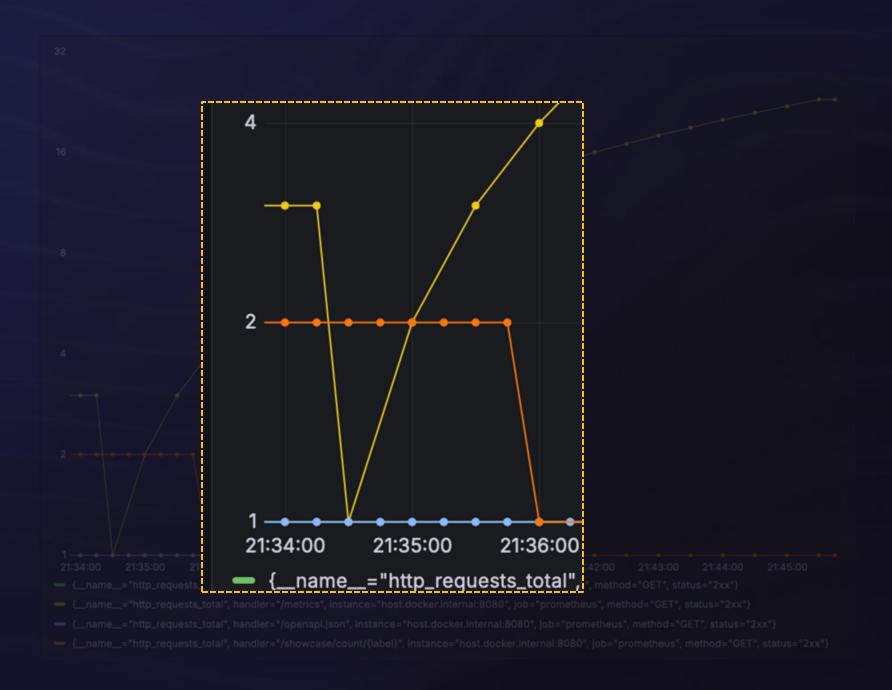
Metrics 101

- 1. Name
- 2. Labels
- 3. Time
- 4. Value

```
"http_requests_total", handler="/docs", instance="host.docker.internal:8080
"http_requests_total", handler="/metrics", instance="host.docker.internal:80
"http_requests_total", handler="/openapi.json", instance="host.docker.interr
"http_requests_total", handler="/showcase/count/{label}", instance="host.de
```

Metrics 101

- 1. Name
- 2. Labels
- 3. Time
- 4. Value



A word on cardinality

- 1. Prometheus is analogous to a *flat database*
 - Partitioned by *time*
 - Labels are like indexed columns
- 2. The DB size is therefore:
 - Timeframe * # labels * # label values

Prometheus metric types: Counter

- As the name implies, a metric that counts
- 2. Can go up or down by any amount
- 3. Stateful value managed by Prometheus
- 4. Examples:
 - # of server requests (by method, path, ...)
 - # of events ("times a user logged in")

Counting the Python way

```
counter = meter.create_counter(
  "my_count",
  description="Event count"
@router.get("/count/{label}")
def count(label: str) -> dict:
  counter.add(1, {"my_label": label})
  return {"status": "ok"}
```



Prometheus metric types: Gauge

- A metric that is sampled on demand
- 2. Each sample is *independent*
- 3. Common examples:
 - CPU core temperature in °C
 - Free space on /dev/sda2

Gauging the Python way

```
gauge = meter.create gauge(
  "my_value",
  description="Some stateful value"
class GaugeData(BaseModel):
  value: int
@router.put("/gauge")
def gauge_value(data: GaugeData) -> dict:
  gauge.set(data.value)
  return {"status": "ok"}
```



Prometheus metric types: Histograms

- 1. Some metrics cannot be represented with one value
 - "How long do requests to /login take?"
 - Do you mean average? Mean? 90th percentile?
- 2. These values prdoduce a *distribution*
- 3. Managed with buckets. Latency, for example:
 - 0-10ms, 10-100ms, 100-1000ms, ...

Histograms the Python way

```
histogram = meter.create_histogram(
  "my_duration",
  unit="ms",
  description="Event duration"
@router.get("/duration/{length_ms}")
def duration(length_ms: int) -> dict:
  histogram.record(length ms)
  return {"status": "ok"}
```



Not all histograms are made equal

Latencies vary wildly (ms vs second), as do sizes (payload in KB, disk in GB).

Views to the rescue!

```
duration_view = View(
   instrument_name="my_duration",
   aggregation=ExplicitBucketHistogramAggregation(range(0, 10000, 500)),
)
meter_provider = MeterProvider(
   metric_readers=[oltp_exporter],
   views=[duration_view],
)
set_meter_provider(meter_provider)
```



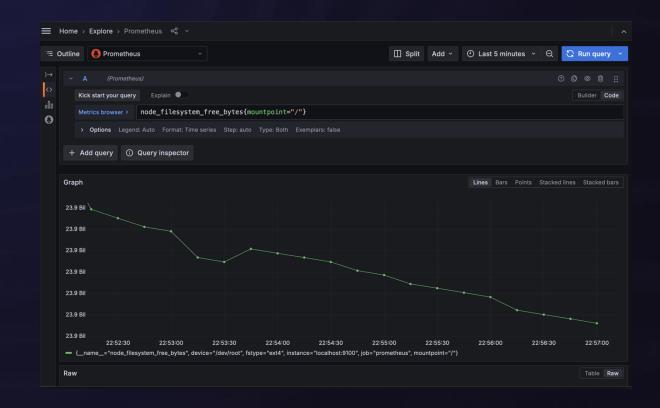
Showcase

- 1. Familiarize yourself with the lab setup
- 2. Get handsy with Grafana
- 3. Open lab-showcase from the class materials
- 4. You have 30 minutes to explore!

Part 2: PromQL

PromQL: Fear, Terror and Ruthless Efficiency

- Extensive, powerful
- 2. ... not entirely trivial



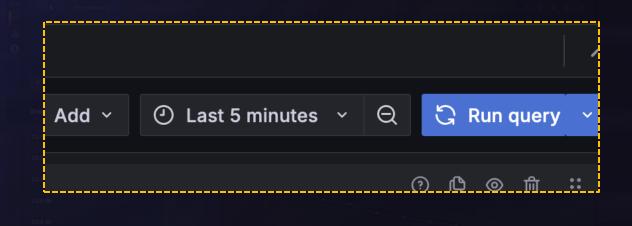
PromQL: Fear, Terror and Ruthless Efficiency

- Extensive, powerful
- 2. ... not entirely trivial
- 3. Basically:
 - A query
 - A timeframe

```
Metrics browser > node_filesystem_free_bytes{mountpoint="/"}
> Options Legend: Auto Format: Time series Step: auto Type: Both Exempla
```

PromQL: Fear, Terror and Ruthless Efficiency

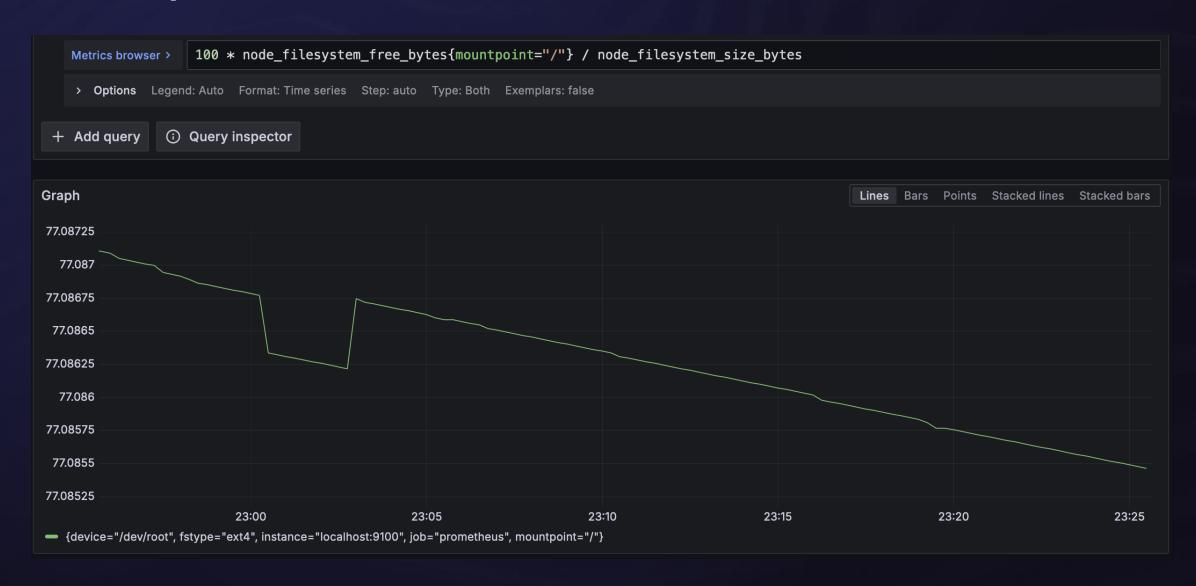
- 1. Extensive, powerful
- 2. ... not entirely trivial
- 3. Basically:
 - A query
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PromQL: Maths!

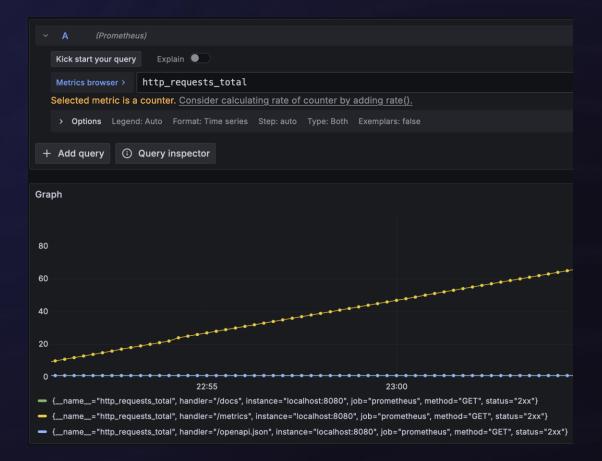
- 1. [Metric, labels, time] yields one value
- 2. Some observations require multiple values
- 3. Expressed with basic arithmetic:
 - % free disk space = *free* disk bytes/*total* disk bytes
 - % CPU = CPU used seconds / (# seconds * # cores)

PromQL: Maths



PromQL: Rates

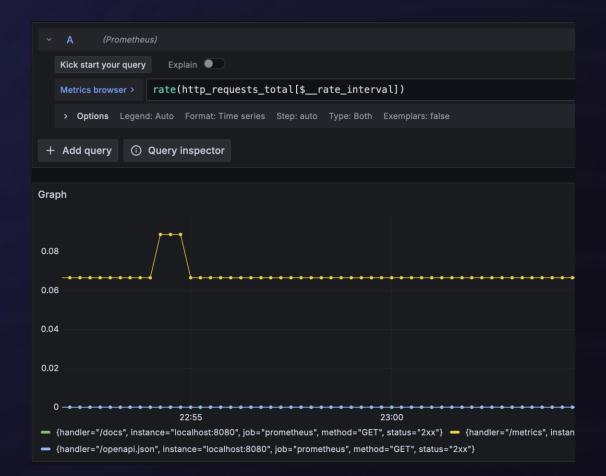
- Counters are very useful
- 2. Can't use them *naïvely*



PromQL: Rates

- Counters are very useful
- 2. Can't use them *naïvely*
- 3. We need a rate derivation
- 4. Magic incantation for the win:

```
rate(metric{...}[$__rate_interval])
```



PromQL: Aggregations

- Consider http_server_duration_milliseconds_count
 - A simple counter
 - Potentially many services, paths
- 2. Suppose you want the grand total?

```
sum(http_server_duration_milliseconds_count)
```

PromQL: Aggregations

- Consider http_server_duration_milliseconds_count
 - A simple counter
 - Potentially many services, paths
- 2. How about the 5 most used paths?

```
topk(5, http_server_duration_milliseconds_count)
```

PromQL: Aggregations

- What are the top 3 most time-consuming endpoints? topk(3, http_server_duration_milliseconds_sum)
- How many processes run which Python version?
 count by (major) (python_info)
- 3. What is the highest rate of pagefaults recorded?

 max(rate(node_vmstat_pgfault[\$__rate_interval]))

And now, what you've all been waiting for

Quantiles (a.k.a. percentiles)

- 1. A histogram is a set of *counters*
 - 0-100ms, 100-200ms, 200-300ms...
- 2. What is the 95th percentile?
 - Linear regression FTW... but maths!



histogram_quantile(0.99, http_server_duration_milliseconds_bucket)





Key Takeaways

1. Instrument early, instrument often

There's no such thing as "too much data"

2. Beware high cardinality

Aggregate, don't transact (that's what logs/traces are for)

3. Learn ye *PromQL*

That's where the real leverage is

Lab time, again!

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- https://github.com/holograph/prometheus-workshop-service-python

Thank you for your attention

Questions?

