



Mastering Chaos: Achieving Fault Tolerance with Observability-Driven Prioritized Load Shedding

Building fault-tolerant, performant and cost-efficient applications with the **Aperture** open source project



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Introduction

- **Harjot Gill**

- Co-founder and CEO @ FluxNinja
 - Founded in 2021
 - Based in the San Francisco Bay Area
 - Announced Aperture open source project in late 2022
- Dedicated 10+ years building tooling for DevOps and SREs
- Previously, Co-founder and CEO @ Netsil (Acquired by Nutanix in 2018)
 - Microservices observability start-up, spin-off from University of Pennsylvania
 - Pioneered low-friction API observability: stream-processed packets to reconstruct APIs
 - Mapping complex microservices applications

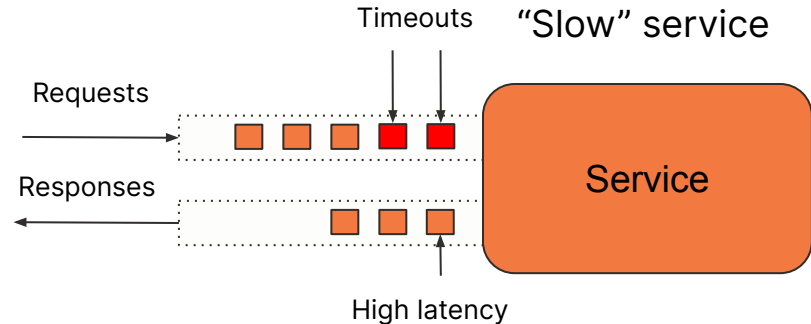
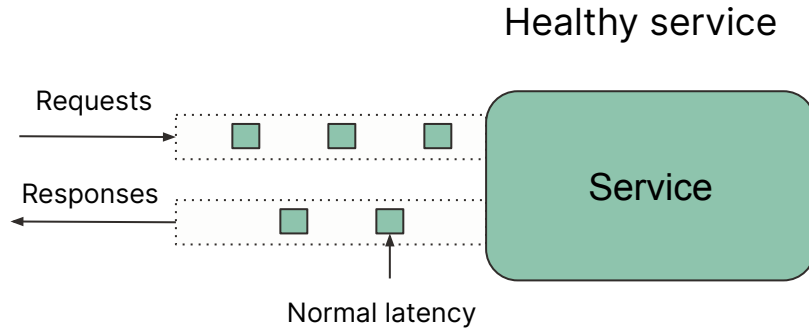
- **Hardik Shingala**

- Software Engineer @ FluxNinja
 - 5+ years of experience in cloud native infrastructure products
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Metastable failures

Little's law conundrum: The inevitability of overloads

Little's law and overloads



Little's law

$$L = \lambda W$$

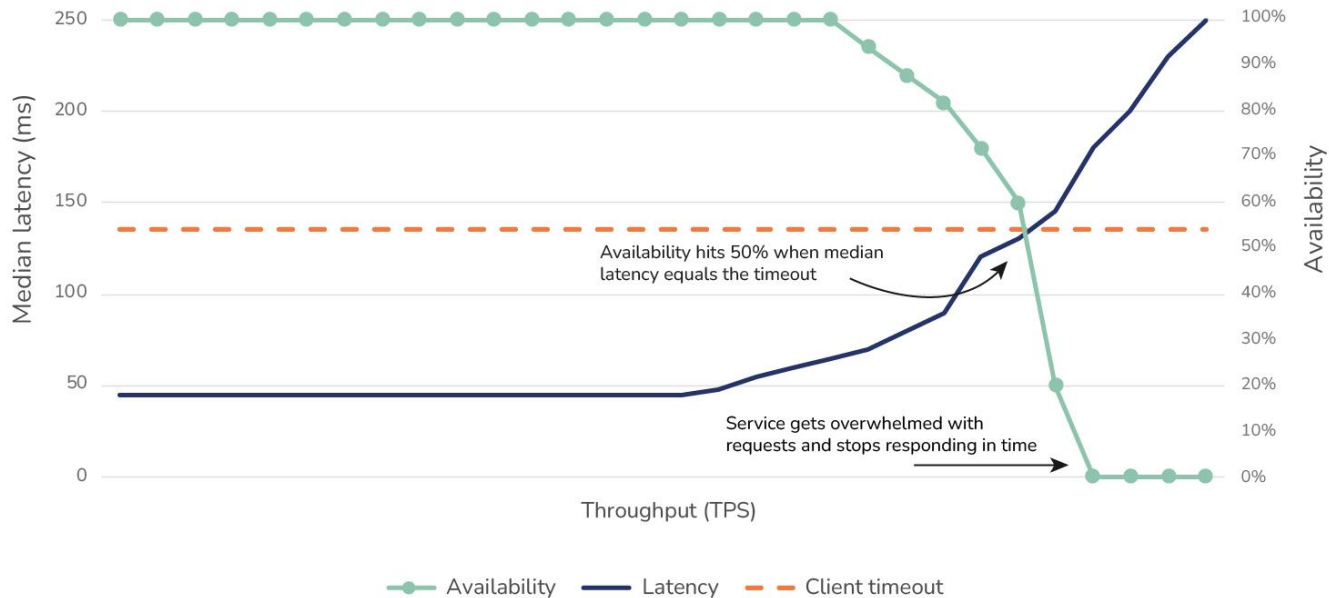
L = Requests in-flight

λ = Average Throughput

W = Average Response time

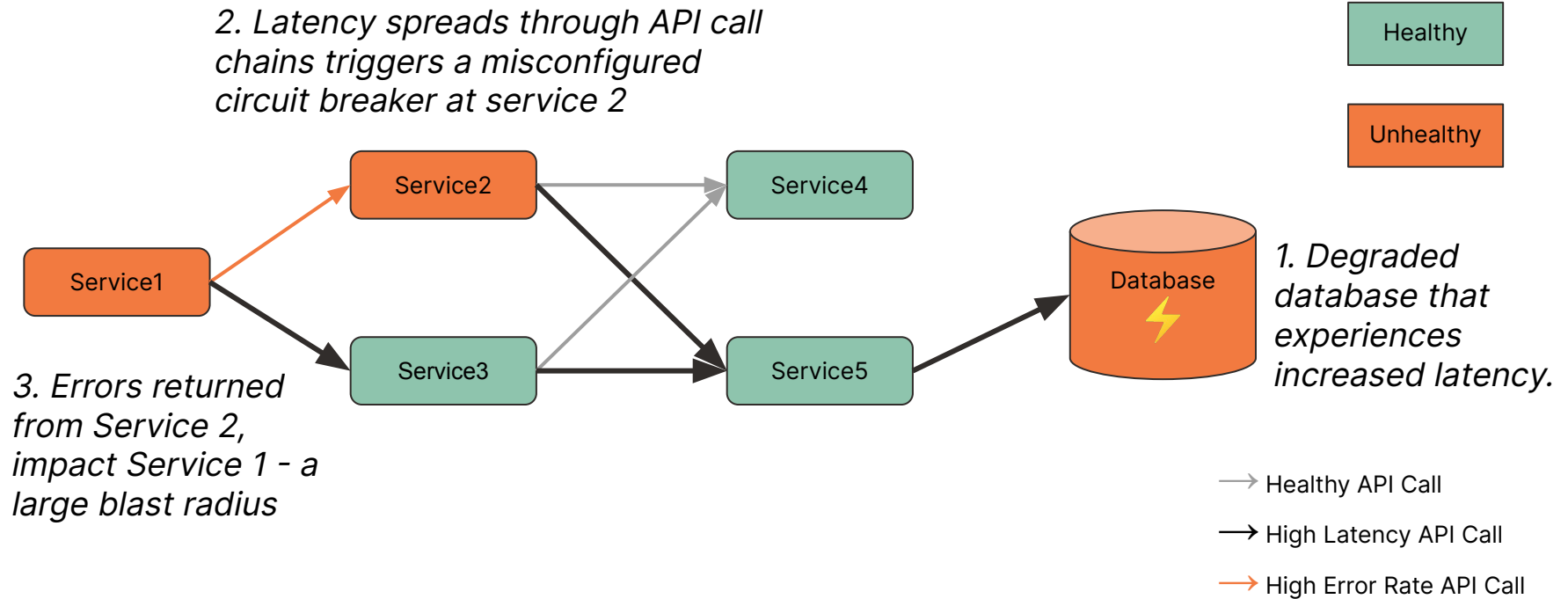
*Every service has an inherent concurrency limit. For a service to remain **stable**, concurrent requests must be limited*

Availability degrades rapidly

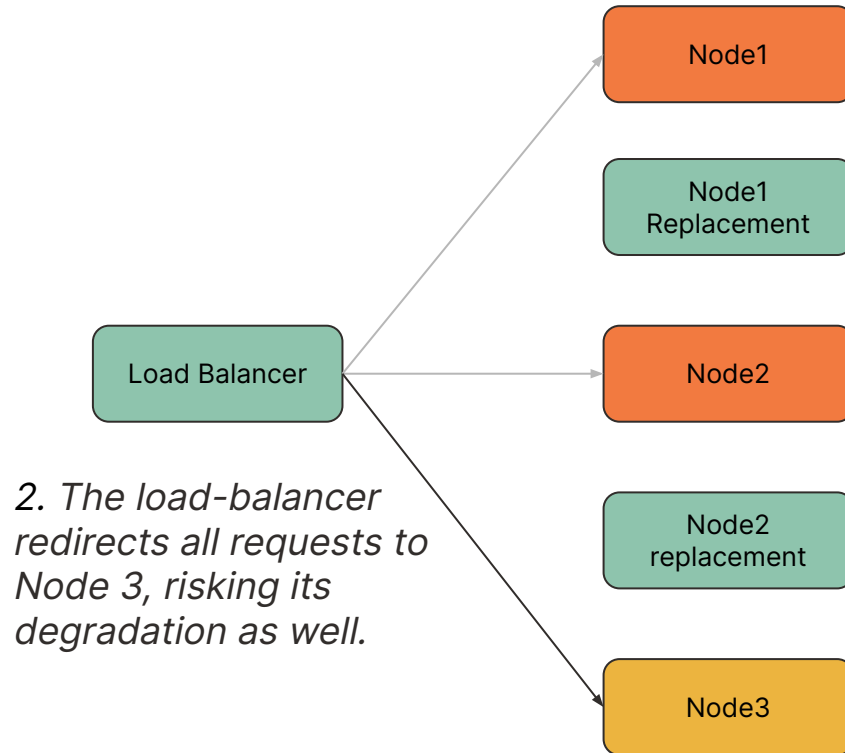


An overload on a service often kicks-off a chain reaction causing an application wide outage...

Cascading failure



Death spiral



1. Nodes 1 and 2 degrade and are replaced by new nodes which are not ready for traffic.

2. The load-balancer redirects all requests to Node 3, risking its degradation as well.

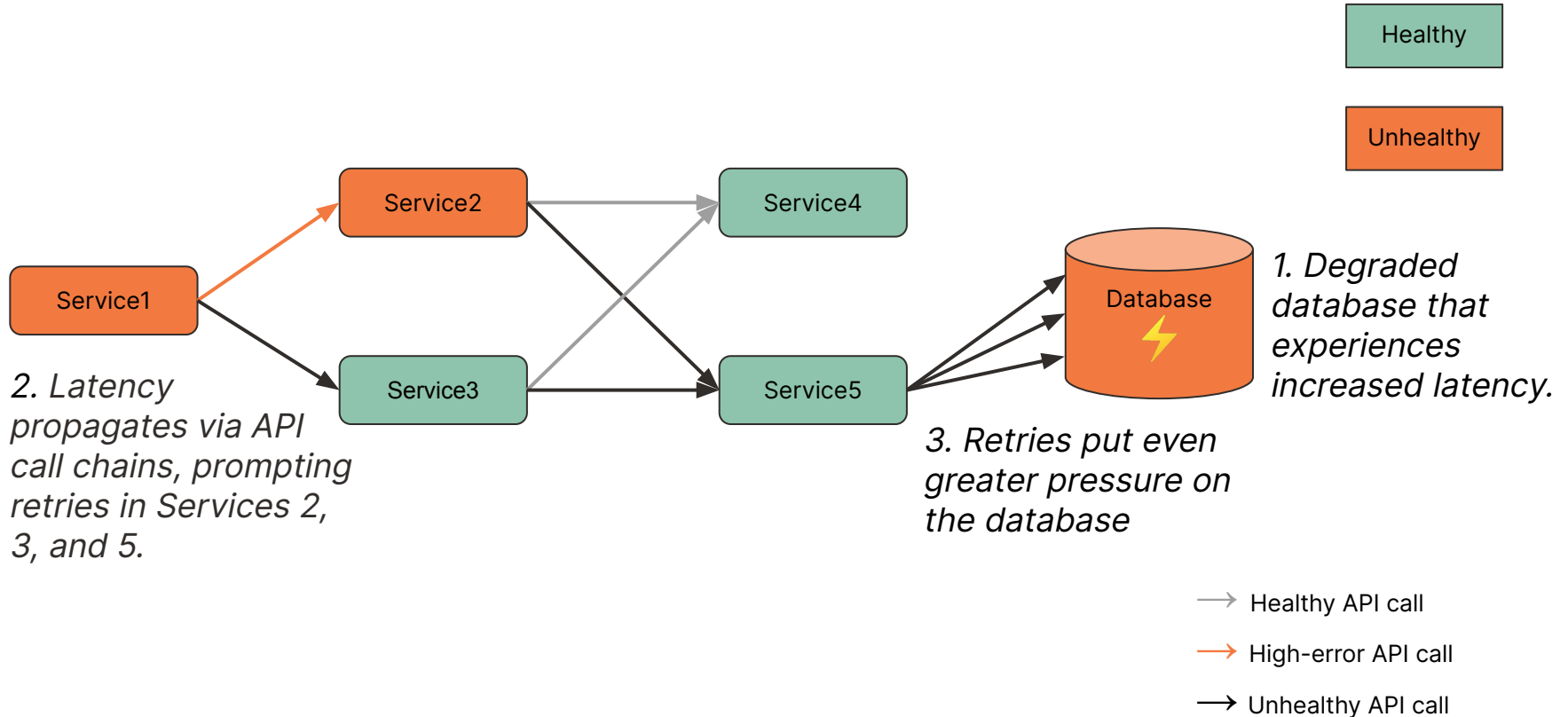
Healthy

Unhealthy

→ API call not in use

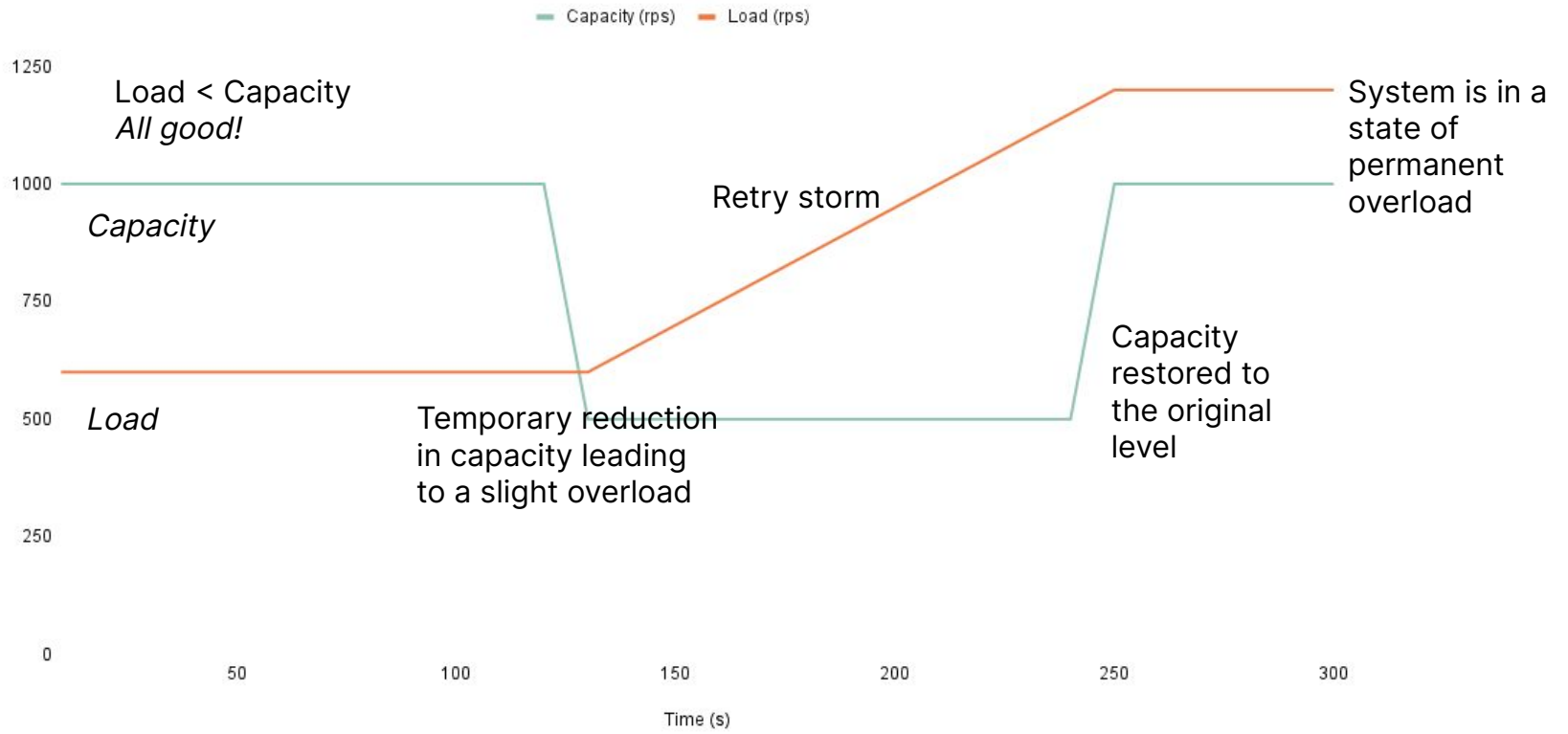
→ High latency API call

Retry storm

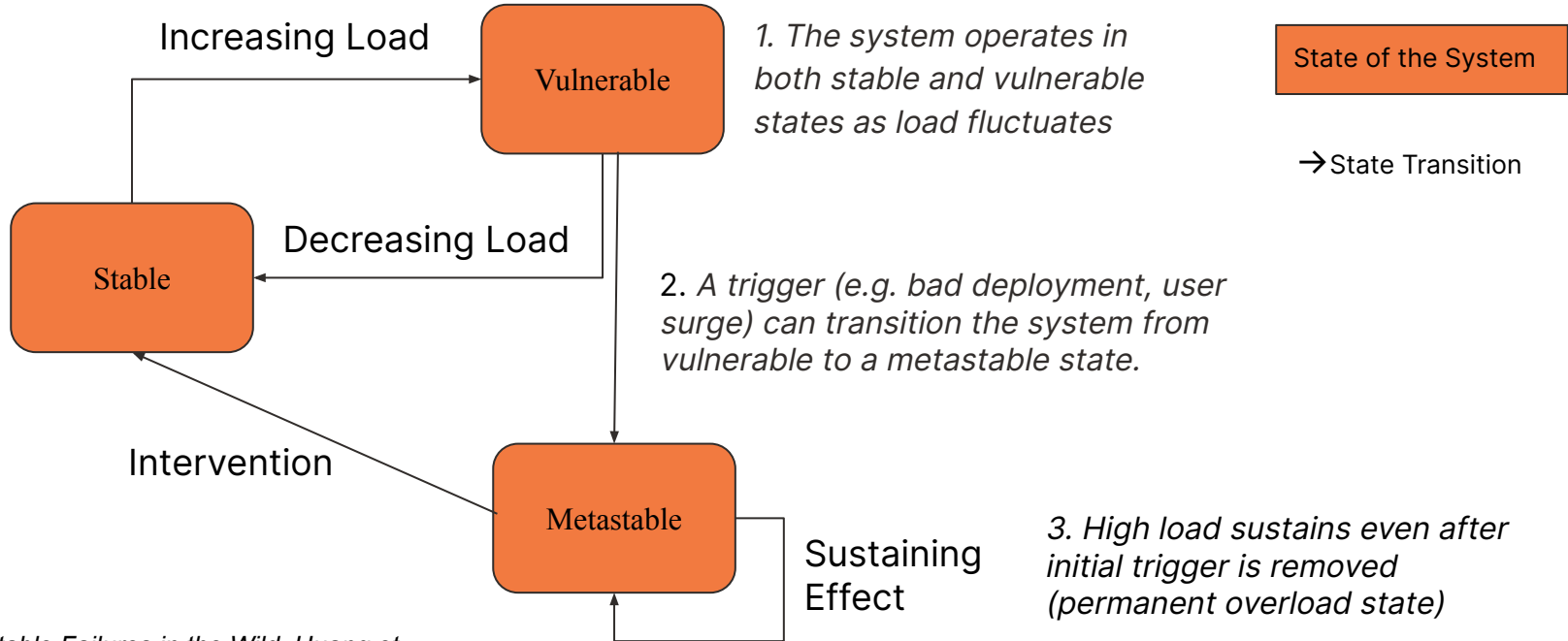


Retry storm: permanent overload

Capacity (rps) and Load (rps)



Metastable failures



Metastable Failures in the Wild, Huang et al.

Common triggers

- Insufficient capacity allocation
- Service upgrades that introduce performance-regressions due to bugs
- Unexpected traffic spikes during new product launches or sales promotions
- Slowdowns in upstream services or third-party dependencies
- Retry storm after a temporary failure
- Cache failure leading to higher load on database
- Subset of servers going offline causing excess load on remaining servers

Metastable failures are unpredictable, yet very common in modern applications

Mitigation strategies

Building indestructible applications

Local countermeasures are ineffective

Circuit breaking

- Typically implemented in service proxy (e.g. Envoy)
- Localized view between service instances (e.g. error rates)
- Rejects all requests when it “trips”
- Hard to configure the “tripping” threshold as some services are more tolerant to errors
- Client-side technique - does not offer service protection

Static rate-limiting

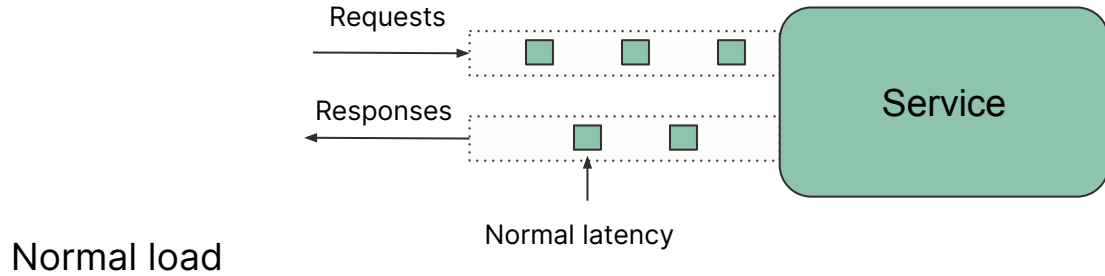
- Typically implemented as a per-user limit
- Does not offer service protection as the per-user limit is not per-service limit

Reactive auto scaling

- Typically scale workers based on resource consumption (e.g. CPU or memory)
- Can be slow as services need time to warm-up, do discovery, establish database connections and so on
- Bottleneck typically shifts elsewhere
- Expensive to absorb transient traffic spikes

Local countermeasures are often slow, inadequate and ineffective

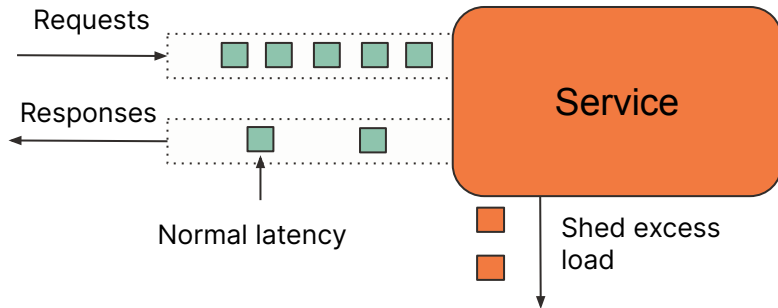
Mitigation with adaptive load shedding



Little's law

$$L = \lambda W$$

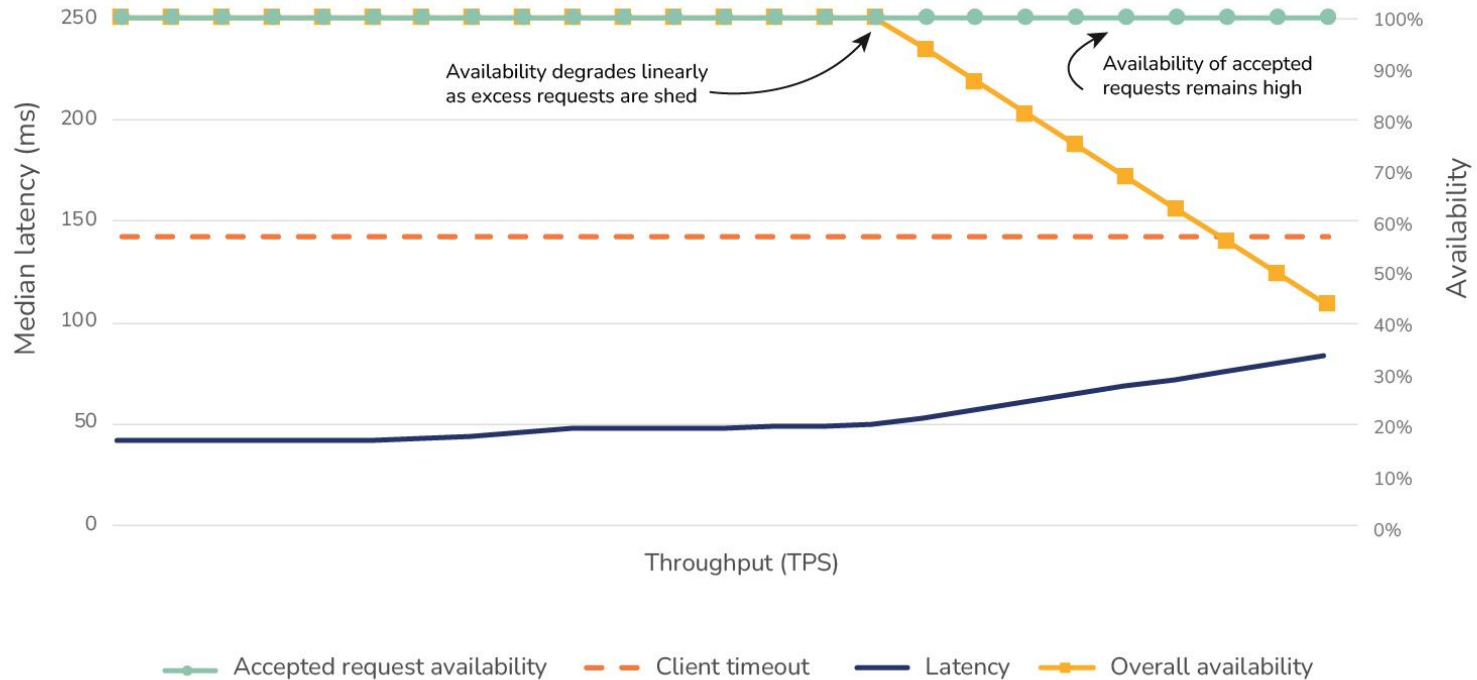
Overload



L = Requests in-flight
 λ = Average Throughput
 W = Average Response time

Service remains **stable** by shedding excess load

Availability degrades gracefully



Requirements for adaptive load shedding

- Determining the ideal load in a constantly changing environment
 - Setting the limit too low can result in rejected requests and wasted capacity
 - Setting the limit too high can lead to slow and unresponsive servers
- Observability: Real-time, global visibility into the state of the entire system
 - Detect overload at databases but load shed at the gateway services
- Controllability: Continuously tracking and correcting system state variables
 - PID controller based closed-loop system
 - Congestion control and active queue management algorithms: TCP BBR, AIMD (Additive increase, multiplicative decrease), CoDel
- Interaction with other control systems with similar goals:
 - Auto scaling
 - Load balancing

Requirements for prioritization

- Optimize user experience and business value: prioritize on attributes such as API endpoints, user types, origin service
- Prioritization and fairness algorithms
 - Token and leaky buckets
 - Network schedulers: weighted-fair queueing
 - Probabilistic dropping
- Estimating the cost (tokens) of admitting different types of requests
 - Tokens = Estimated latency?
 - Tokens = Query complexity?

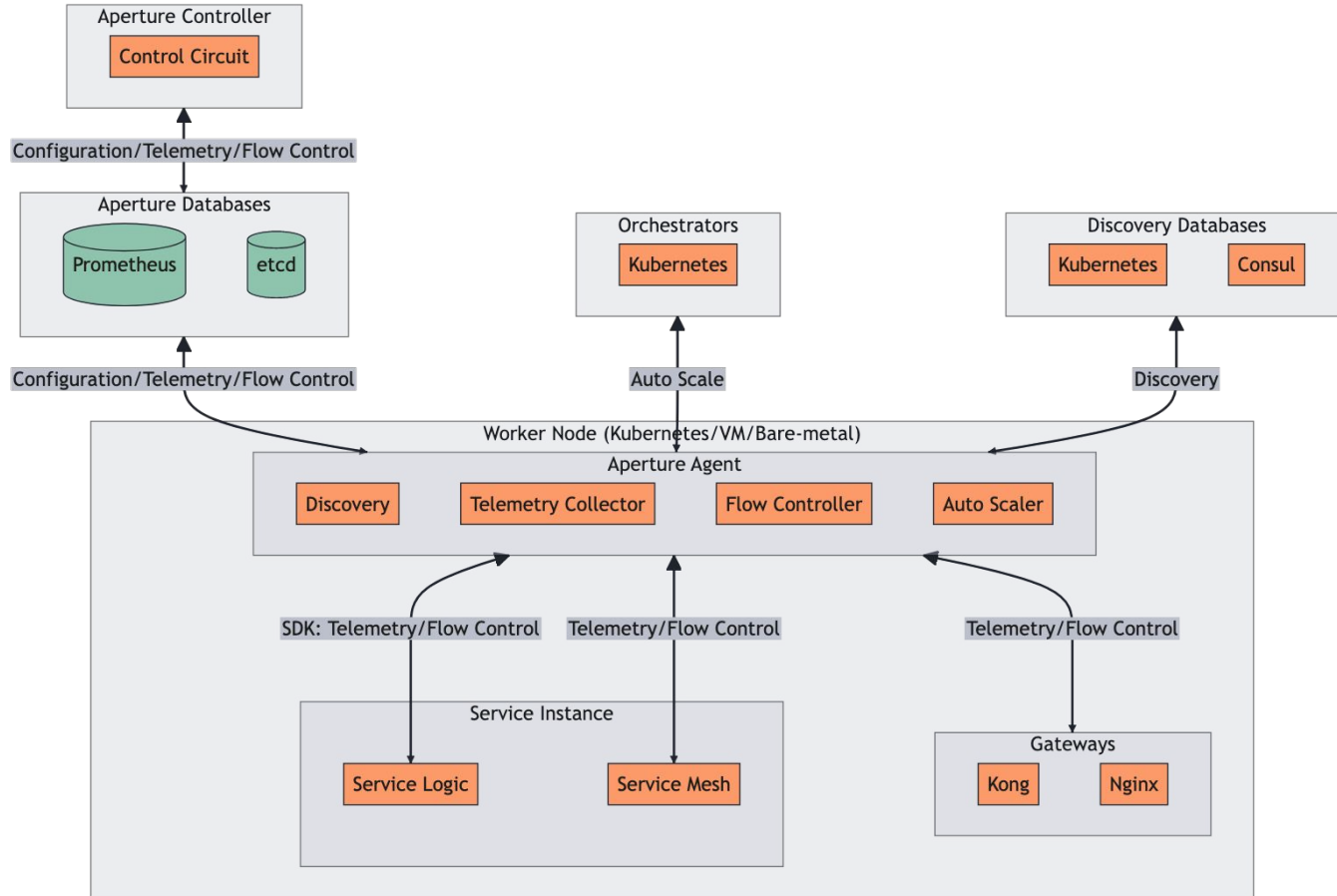
Global load management with Aperture

Controlling the flux: Observability meets Controllability

Aperture overview

- Open source platform for observability-driven load management
- Programmable through declarative policy language expressed as a control circuit graph
- Common policies are packaged as high-level “blueprints”
 - Load scheduling & workload prioritization
 - Quota enforcement
 - Load ramping
 - Auto scaling
- Layered on top of existing stack
 - SDKs: Java, Go, Python etc.
 - Service Mesh: Istio etc.
 - API Gateways and proxies: Nginx, Kong

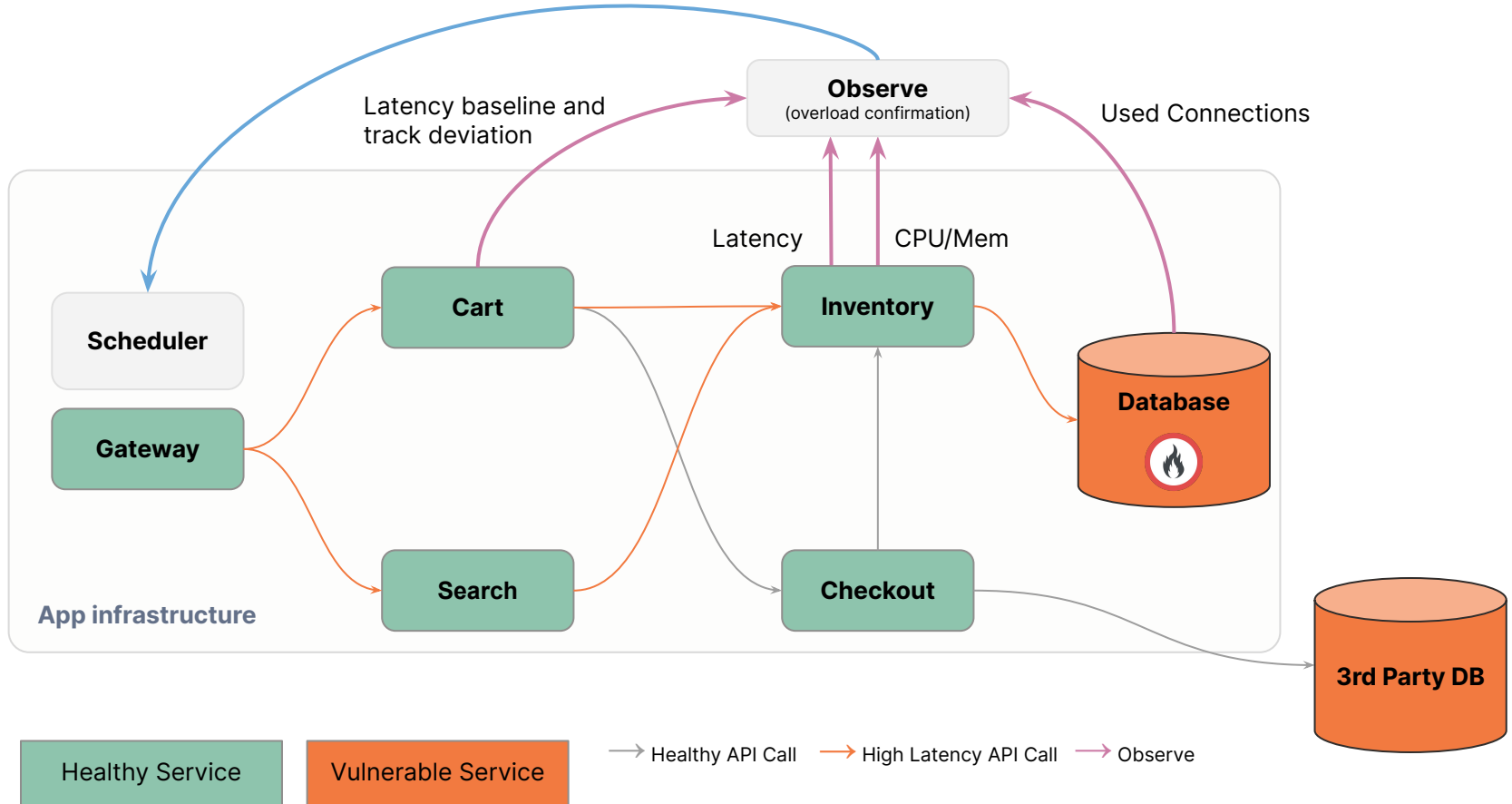
Aperture architecture



Adaptive load scheduler

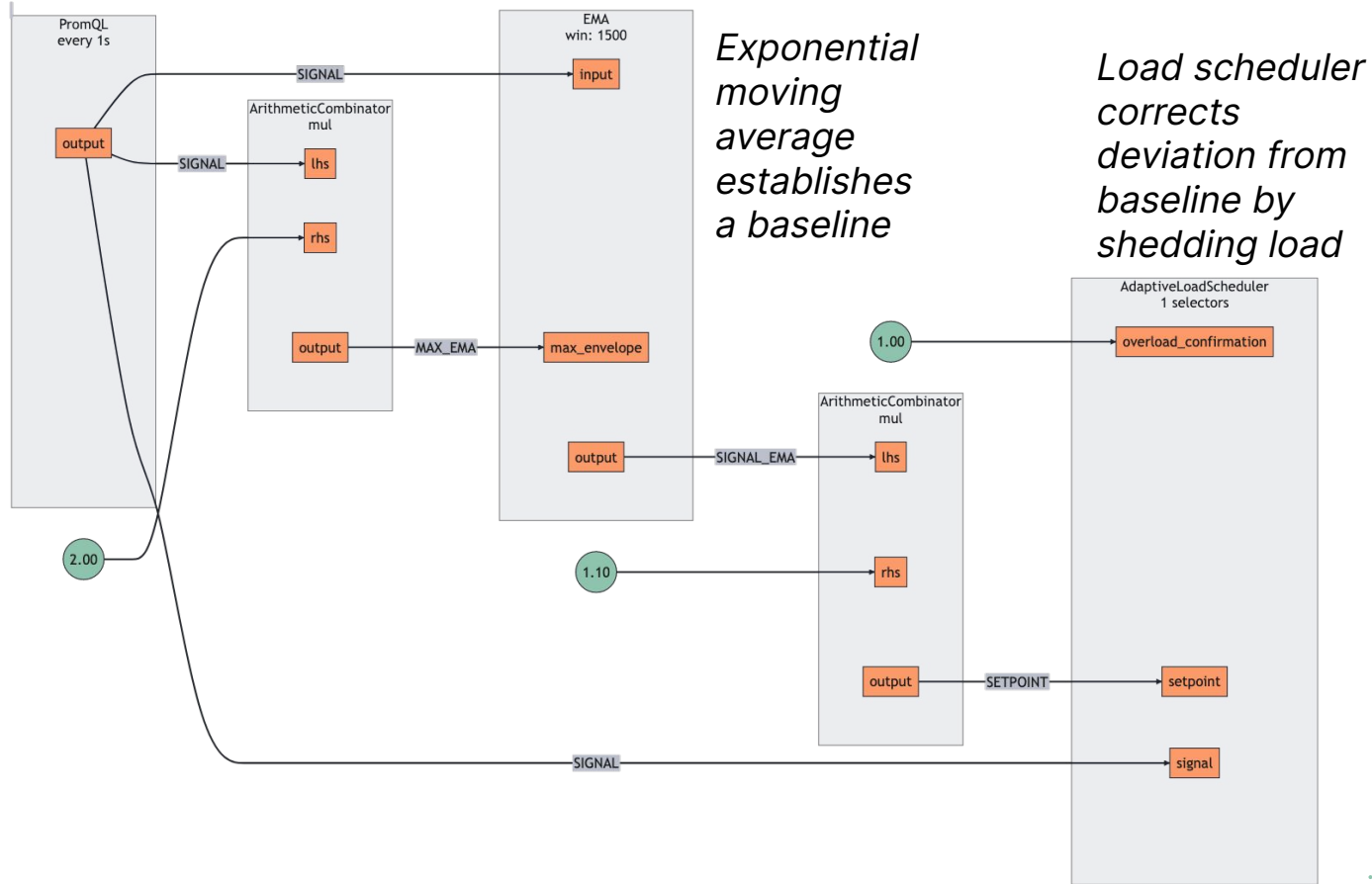
Service protection based on feedback loop

Observability-driven approach



Adaptive load scheduling policy

Service latency is queried periodically



Exponential moving average establishes a baseline

Load scheduler corrects deviation from baseline by shedding load

Load scheduler policy component

```
circuit:  
  components:  
  - flow_control:  
      adaptive_load_scheduler:  
        in_ports:  
          setpoint:  
            signal_name: SETPOINT  
          signal:  
            signal_name: SIGNAL  
        out_ports:  
          desired_load_multiplier:  
            signal_name: DESIRED_LOAD_MULTIPLIER  
          observed_load_multiplier:  
            signal_name: OBSERVED_LOAD_MULTIPLIER  
        parameters:  
          load_scheduler:  
            scheduler:  
              workloads:  
                - label_matcher:  
                    match_labels:  
                      user_type: guest  
                parameters:  
                  priority: 50  
                - label_matcher:  
                    match_labels:  
                      user_type: subscriber  
                parameters:  
                  priority: 200  
          selectors:  
            - control_point: ingress  
              service: service1-demo-app.demoapp.svc.cluster.local
```

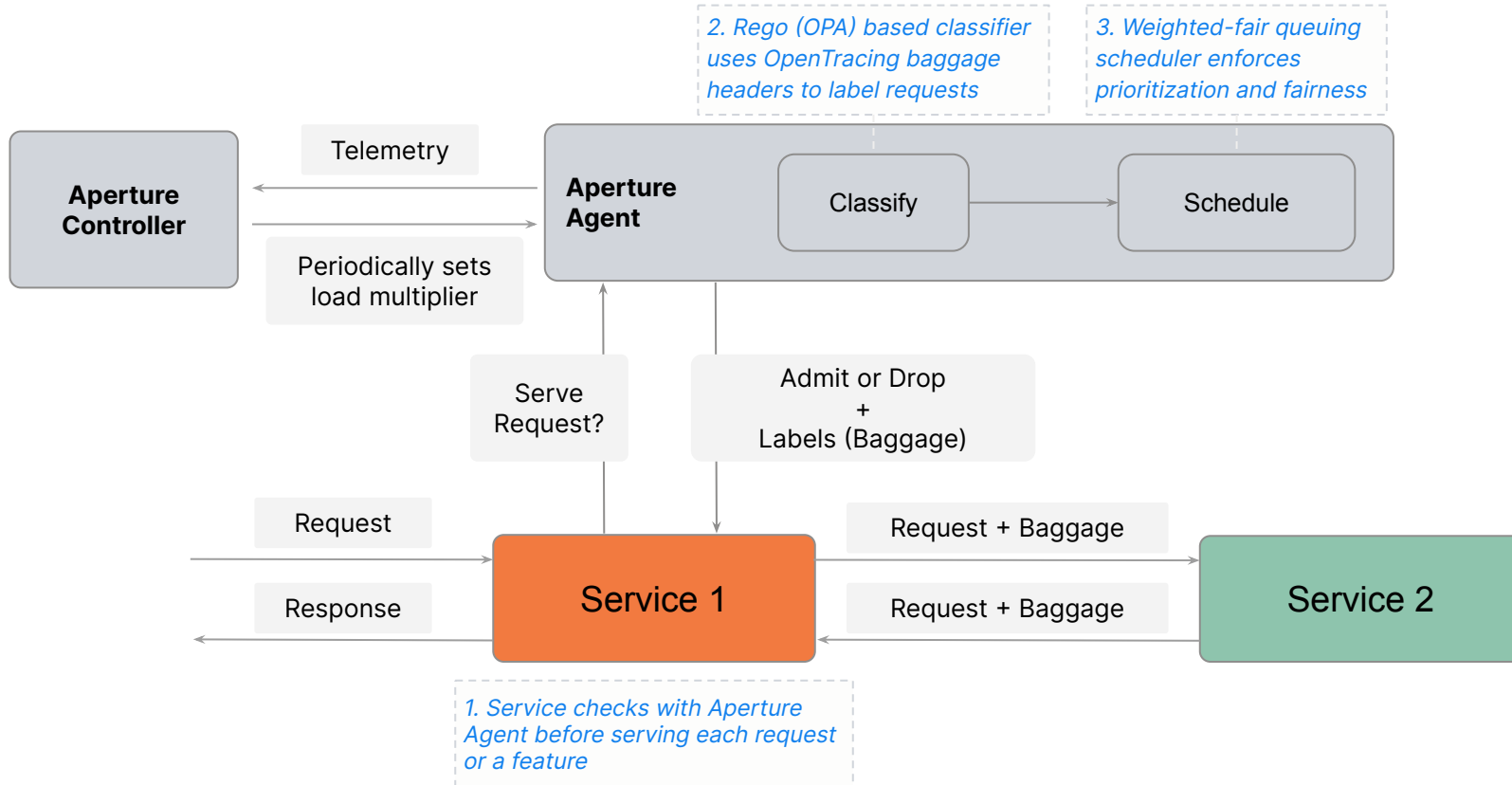
Policy is expressed as a control “circuit” composed of components

Signals flow between components through ports and the circuit is evaluated periodically

Workloads are defined by matching labels and assigning priorities

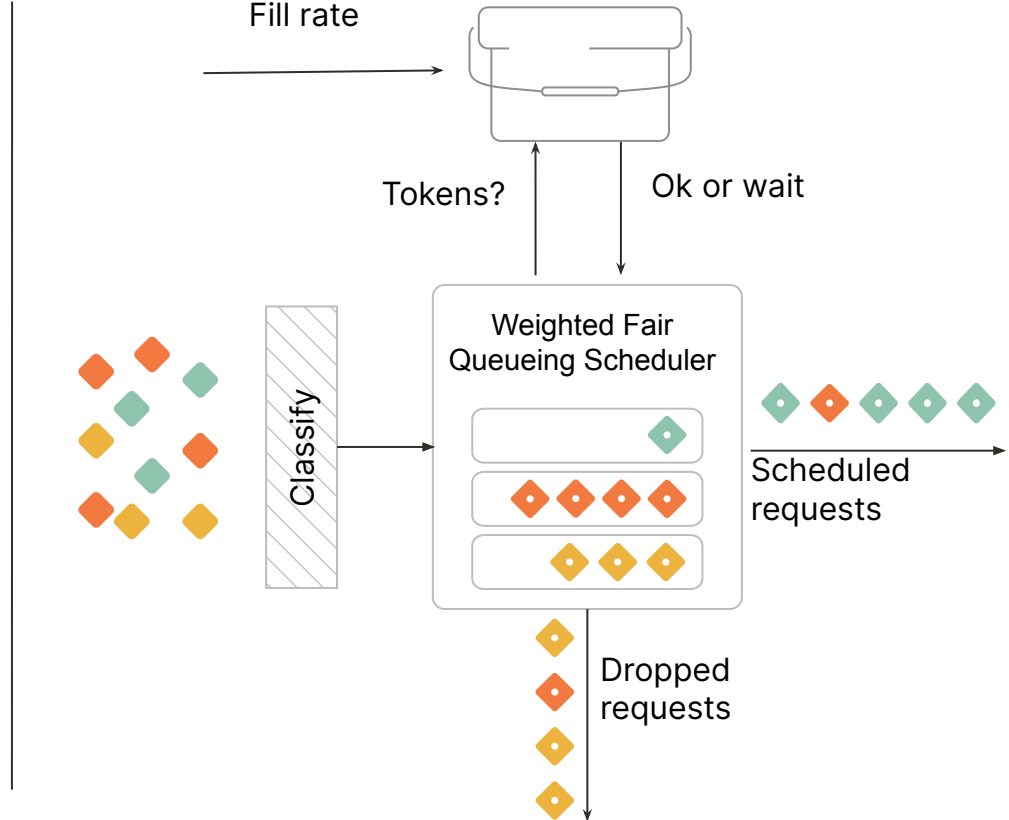
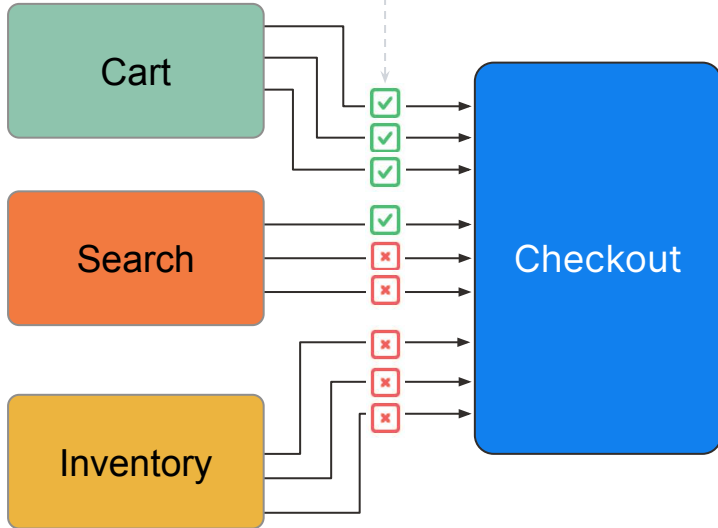
Selectors determine agents where this scheduler will be configured

Adaptive load scheduler insertion



Workload prioritization with Aperture

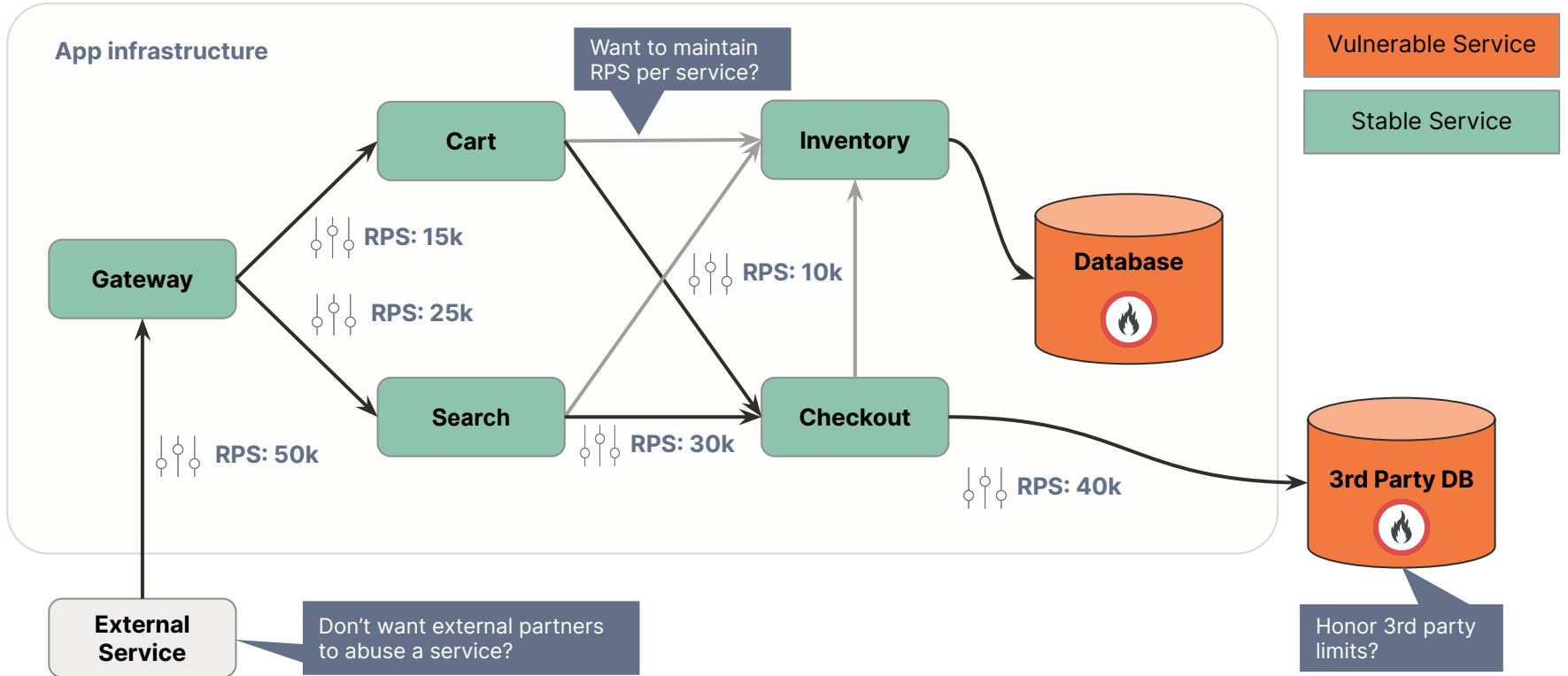
Client Service	Priority
Cart	255
Search	150
Recommendations	10



Global quotas

Enforcing precise limits

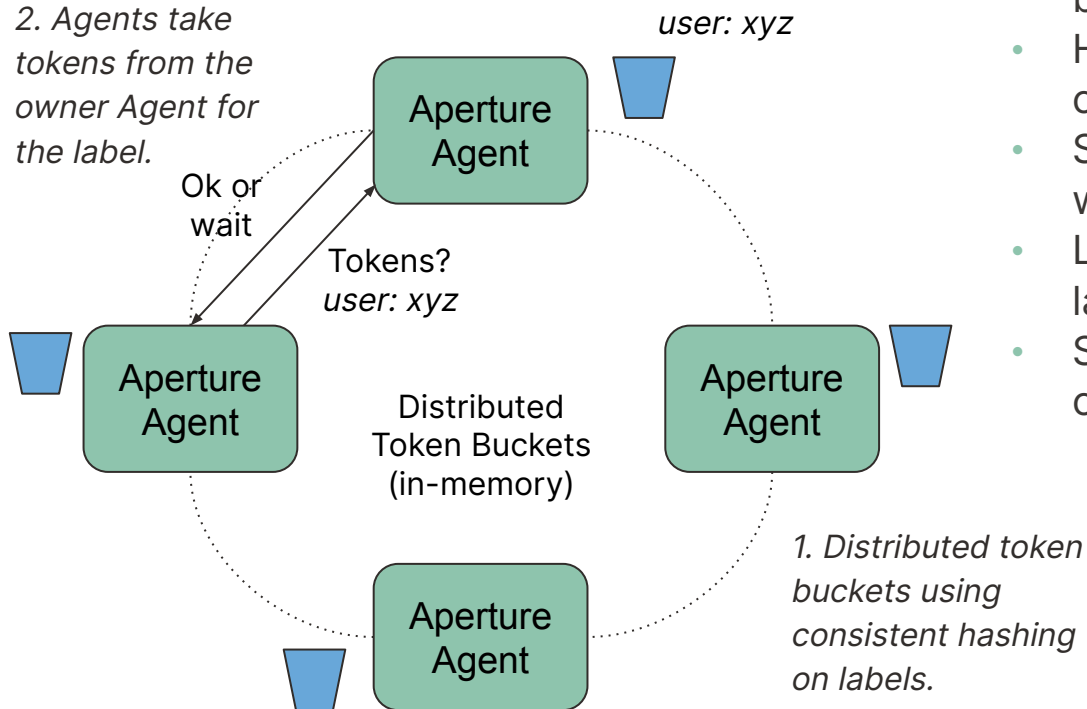
Global quotas



Global quotas

- Service protection
 - When max capacity is known (load testing)
 - Allocate/enforce exact quotas (rps) with other services
- Managing external API rate limits
 - External services such as OpenAI, GitHub, DynamoDB etc. have rate limits. Clients must honor the limit in order to prioritize requests
 - Control costs by preventing accidental overuse
- Preventing abuse
 - Rate-limit external clients based on per-user or per-device quotas

Global quotas in Aperture



- Aperture provides consistent-hashing based global token buckets
- High performance compared to centralized Redis based system
- Smooth load compared to fixed window rate limiting
- Lazy sync (optional) for even lower latencies
- Schedule (prioritize) requests when capacity is reached

Quota scheduler policy component

```
circuit:  
  components:  
    - flow_control:  
      quota_scheduler:  
        in_ports:  
          bucket_capacity:  
            constant_signal:  
              value: 500  
          fill_amount:  
            constant_signal:  
              value: 25  
        rate_limiter:  
          interval: 1s  
          label_key: http.request.header.api_key  
        scheduler:  
          workloads:  
            - label_matcher:  
              match_labels:  
                http.request.header.user_type: guest  
              parameters:  
                priority: 50  
            - label_matcher:  
              match_labels:  
                http.request.header.user_type: subscriber  
              parameters:  
                priority: 200  
          selectors:  
            - control_point: ingress  
              service: service1-demo-app.demoapp.svc.cluster.local
```

Quotas are expressed as -

- *Bucket capacity (for allowing bursts) - e.g. 500 requests*
- *Fill amount and interval - e.g. 25 request per second*
- *Label key - Buckets are created for each key/value pair, e.g. users, services, API keys*

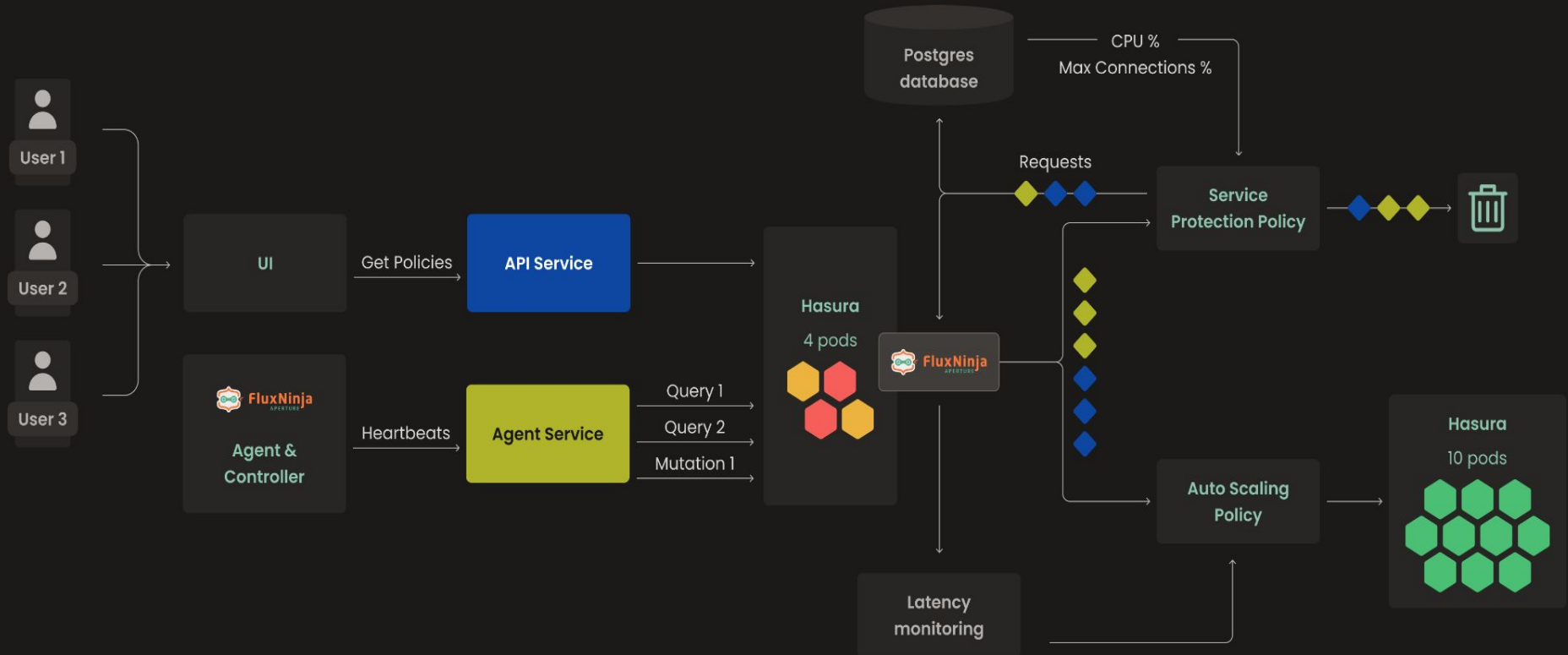
Workloads are defined by matching labels and assigning priorities

Selectors determine agents where this scheduler will be configured

Aperture in FluxNinja ARC

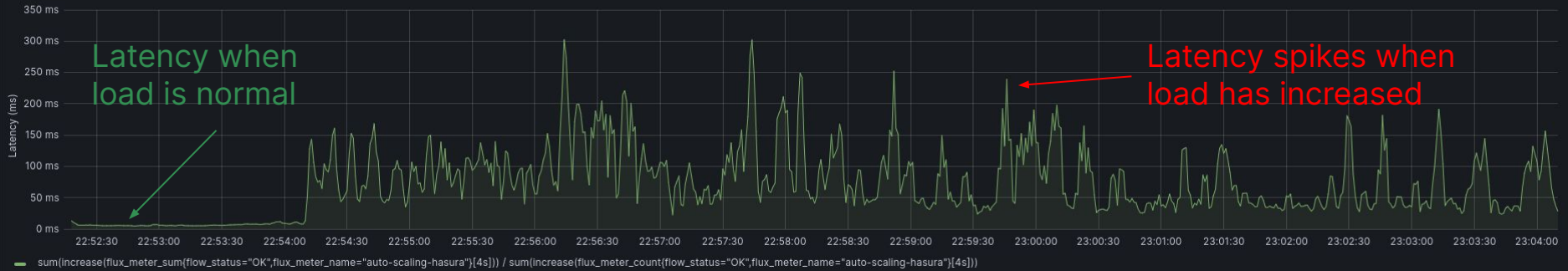
Protecting PostgreSQL by scheduling GraphQL APIs

Protecting PostgreSQL

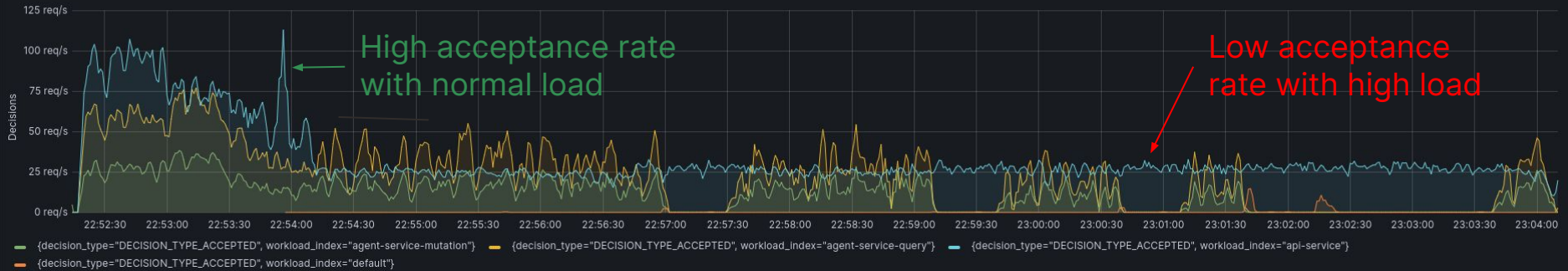


Without Aperture

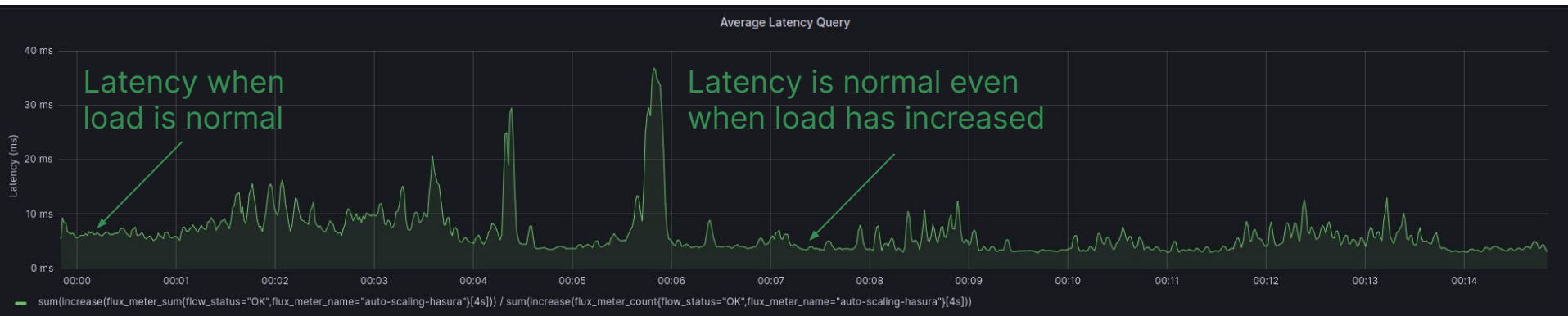
Average Latency Query



Workload Decisions (accepted)



With Aperture



Q & A

- Aperture project on GitHub: <https://github.com/fluxninja/aperture>
- Aperture Docs: <https://docs.fluxninja.com/docs>
- Early access to FluxNinja ARC: <https://app.fluxninja.com/sign-in>

