



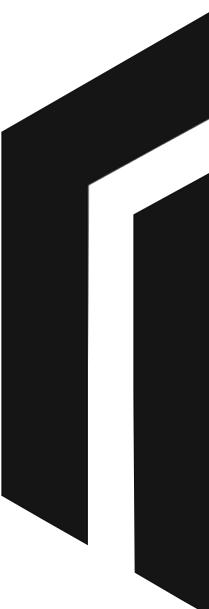




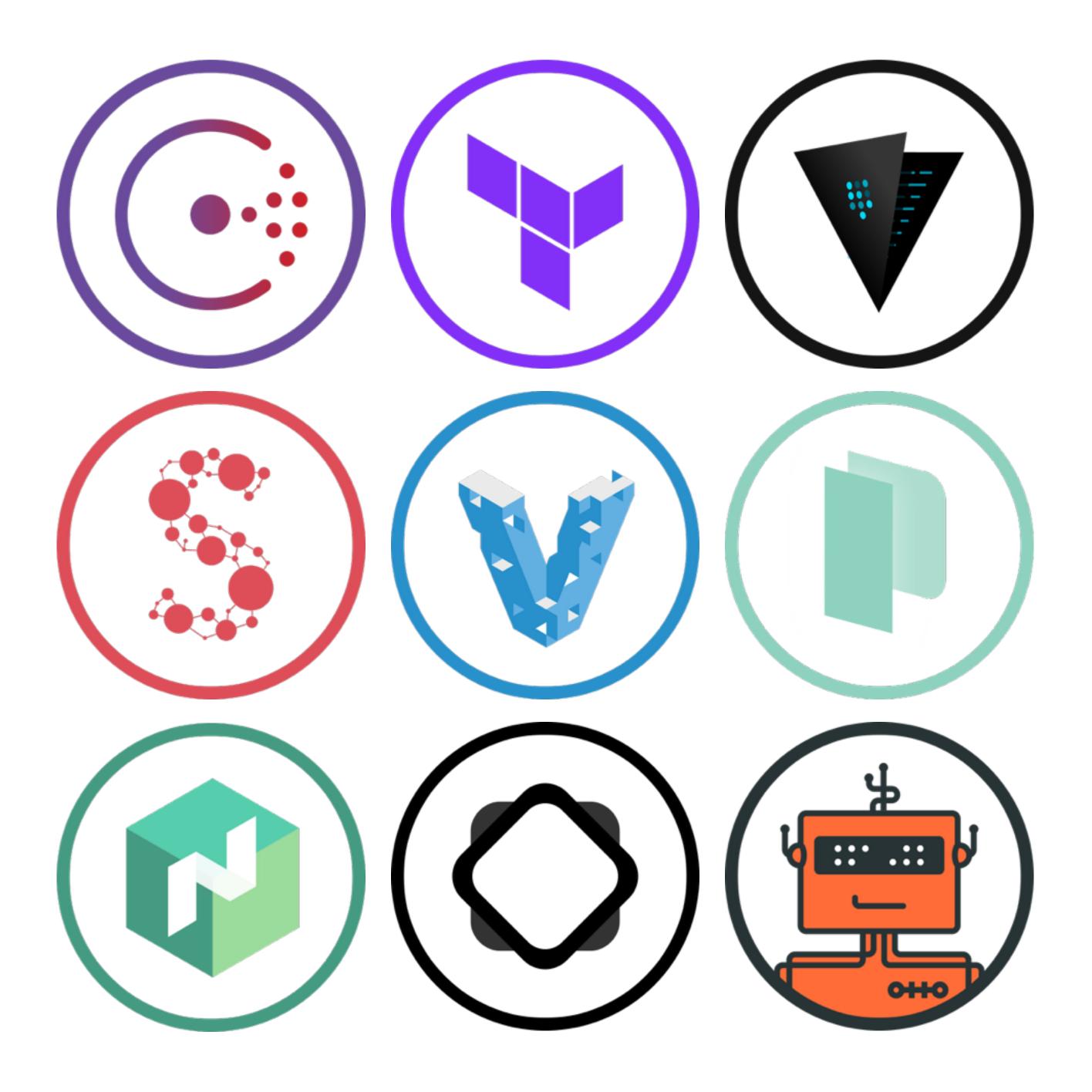


Armon Dadgar @armon













Distributed

Optimistically Concurrent

Scheduler







Distributed

Optimistically Concurrent

Scheduler





Schedulers map a set of work to a set of resources



Work (Input)

Web Server - Thread 1

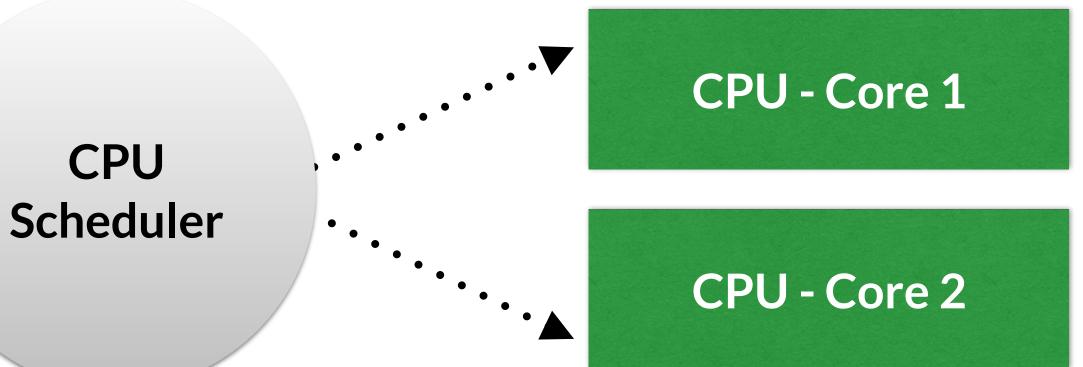
Web Server -Thread 2

Redis - Thread 1

Kernel -Thread 1

CPU Scheduler

Resources







Work (Input)

Web Server -Thread 1

Web Server -Thread 2

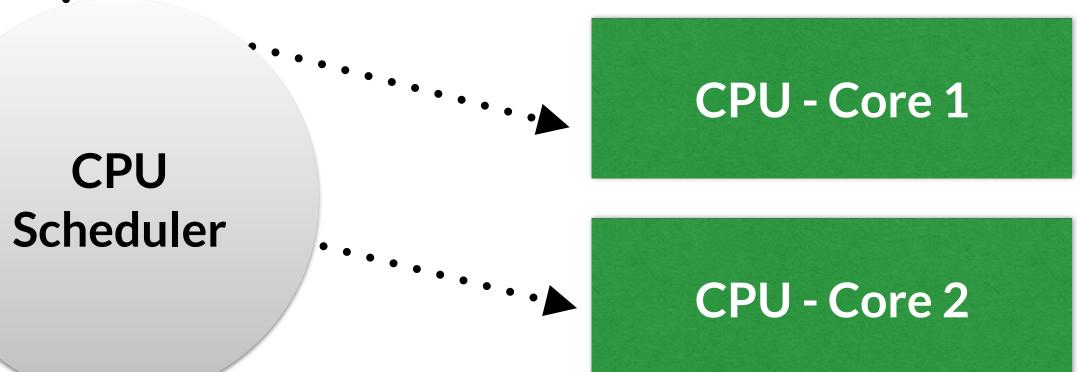
Redis - Thread 1

Kernel -Thread 1

CPU Scheduler

•••••

Resources







Туре

CPU Scheduler

AWS EC2 / OpenStack Nova Virtu

Hadoop YARN Map

Cluster Scheduler Ap

Schedulers In the Wild

Work	Resources
Threads	Physical Cores
ual Machines	Hypervisors
Reduce Jobs	Client Nodes
pplications	Servers





Higher Resource Utilization Decouple Work from Resources Better Quality of Service







Higher Resource Utilization

Decouple Work from Resources

Better Quality of Service



Bin Packing Over-Subscription

Job Queueing



Higher Resource Utilization

Decouple Work from Resources

Better Quality of Service



Abstraction

API Contracts

Standardization



Higher Resource Utilization

Decouple Work from Resources

Better Quality of Service

Advantages

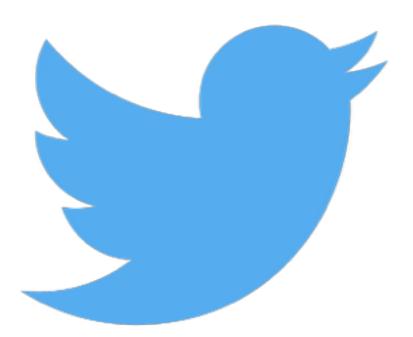
Priorities

Resource Isolation

Pre-emption



amazon webservices™













Cluster Scheduler Easily Deploy Applications **Job Specification**






```
# Define our simple redis job
job "redis" {
  # Run only in us-east-1
  datacenters = ["us-east-1"]
  # Define the single redis task using Docker
  task "redis" {
    driver = "docker"
    config {
      image = "redis:latest"
    }
    resources {
      cpu = 500 \# Mhz
      memory = 256 \# MB
      network {
        mbits = 10
        dynamic_ports = ["redis"]
```





Job Specification

Declares what to run





Nomad determines where and manages how to run

Job Specification



Abstract work from resources

Job Specification

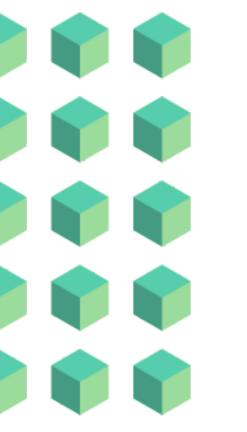


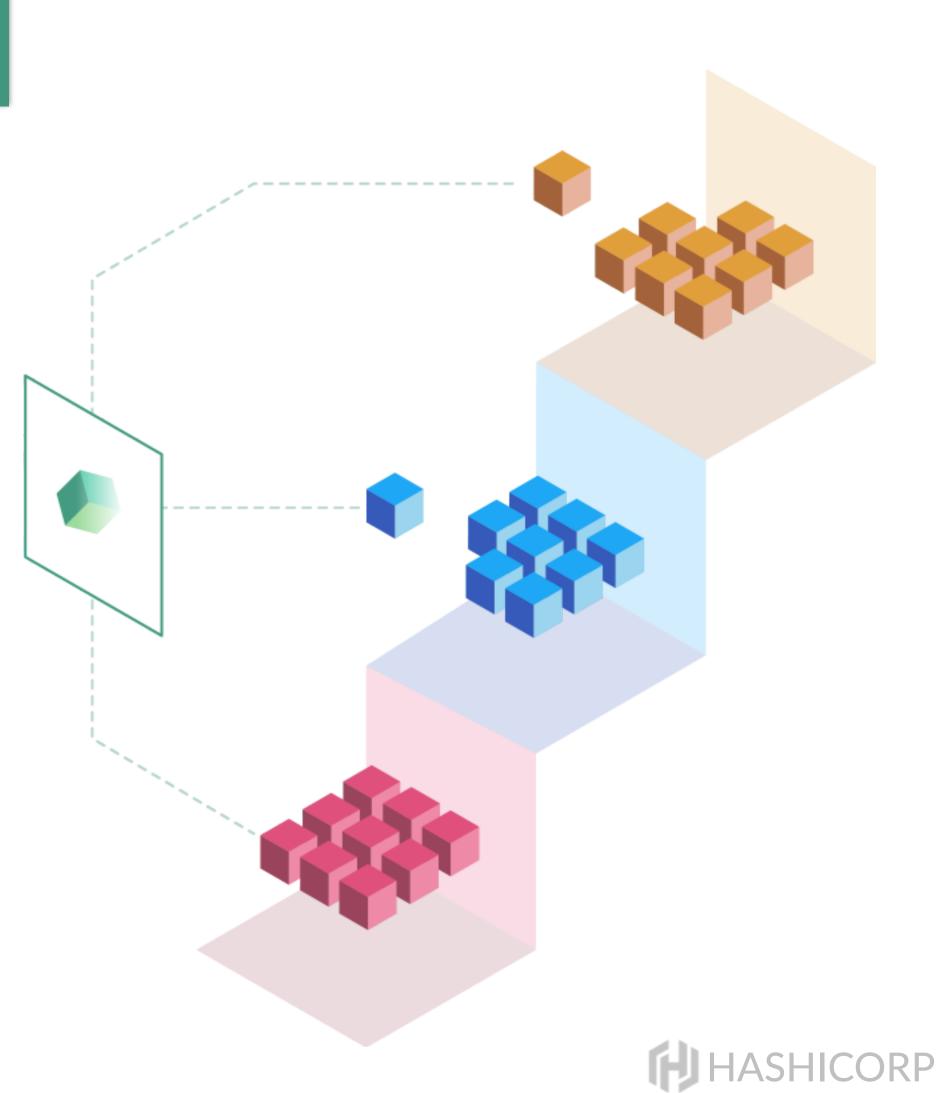
Higher Resource Utilization Decouple Work from Resources Better Quality of Service





Designing Nomad







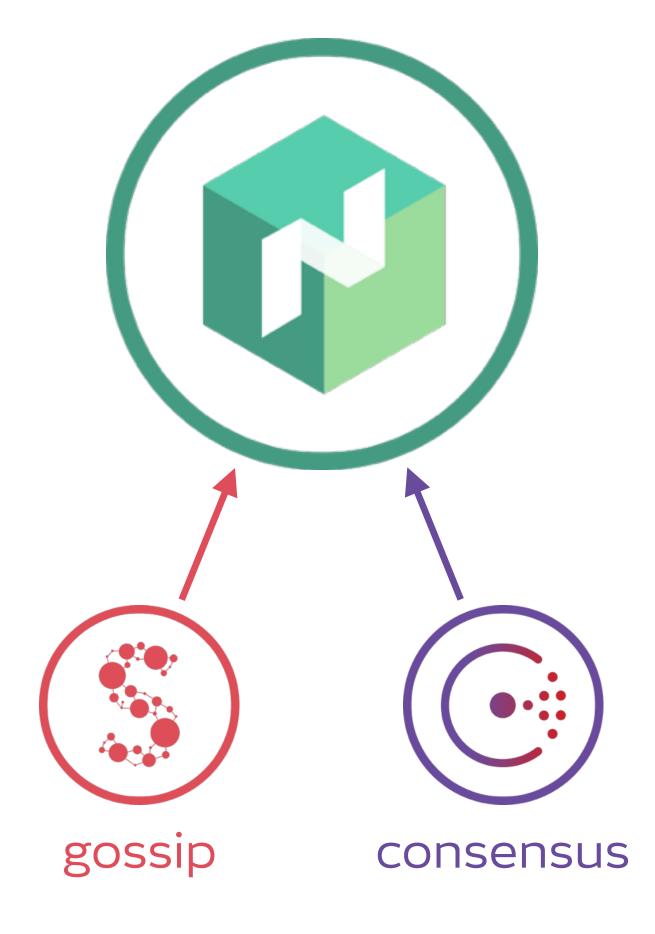
Multi-Datacenter Multi-Region Flexible Workloads Job Priorities **Bin Packing** Large Scale Operationally Simple





Thousands of regions Tens of thousands of clients per region Thousands of jobs per region



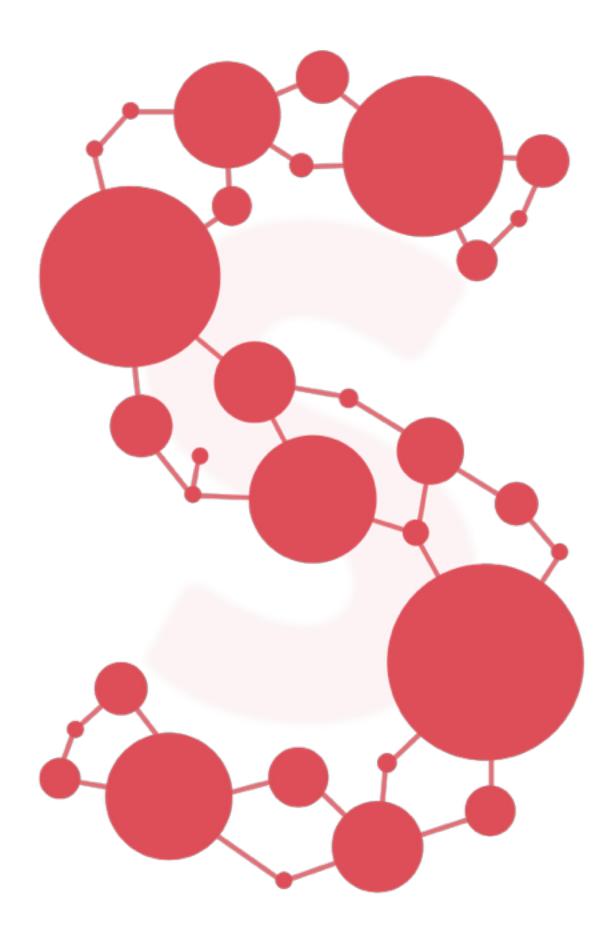


Built on Experience

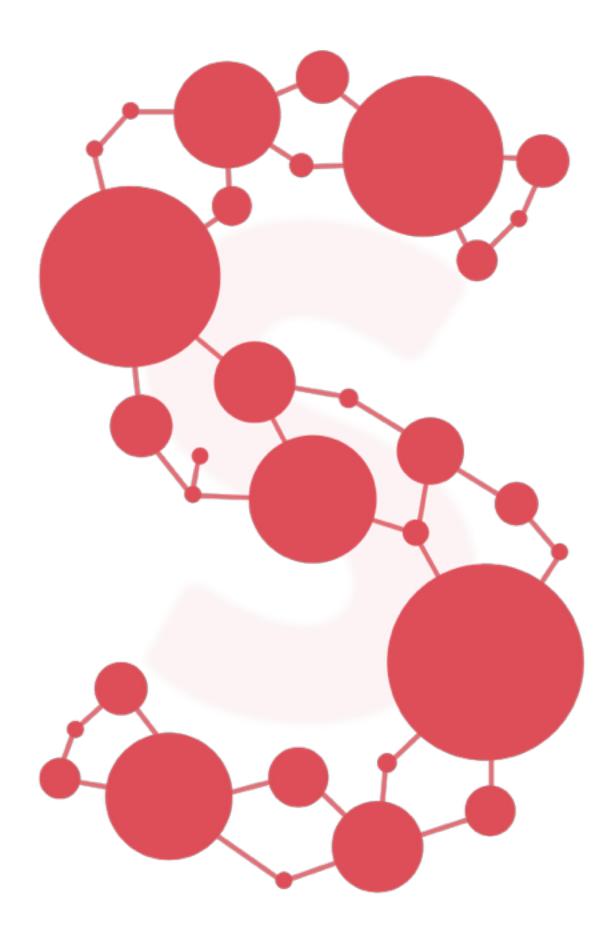




Cluster Management Gossip Based (P2P) Membership Failure Detection Event System



Gossip Protocol Large Scale **Production Hardened** Operationally Simple



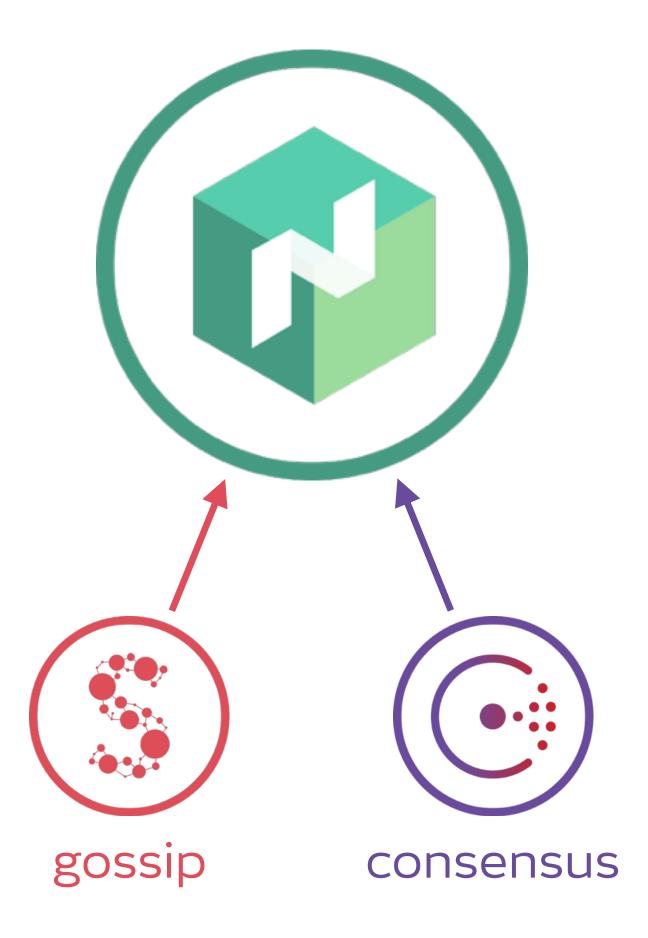
Service Discovery Configuration Coordination (Locking) Central Servers + Distributed Clients



Multi-Datacenter Raft Consensus Large Scale **Production Hardened**



Mature Libraries Design Patterns No Scheduling Logic







Built on Research













Large-scale cluster management at Google with Borg D Q

Abstract: Google's Borg system is a cluster manager that runs hundreds of thousands of jobs, from many thousands of different applications, across a number of clusters each with up to tens of thousands of machines. It achieves high utilization by combining admission control, efficient task-packing, over-commitment, and machine sharing with process-level performance isolation. It supports high-availability applications with runtime features that minimize fault-recovery time, and scheduling policies that reduce the probability of correlated failures. Borg simplifies life for its users by offering a declarative job specification language, name service integration, real-time job monitoring, and tools to analyze and simulate system behavior.

We present a summary of the Borg system architecture and features, important design decisions, a quantitative analysis of some of its policy decisions, and a qualitative examination of lessons learned from a decade of operational experience with it.



Sparrow: Low Latency Scheduling for Interactive Cluster Services

Posted on March 28, 2012 by Patrick Wendell

The Sparrow project introduces a distributed cluster scheduling architecture which supports ultra-high throughput, low latency task scheduling. By supporting very low-latency tasks (and their associated high rate of task turnover), Sparrow enables a new class of cluster applications which analyze data at unprecedented volume and speed. The Sparrow project is under active development and maintained in our public github repository.

Omega: flexible, scalable schedulers for large compute clusters

Abstract: Increasing scale and the need for rapid response to changing requirements are hard to meet with current monolithic cluster scheduler architectures. This restricts the rate at which new features can be deployed, decreases efficiency and utilization, and will eventually limit cluster growth. We present a novel approach to address these needs using parallelism, shared state, and lock-free optimistic concurrency control. We compare this approach to existing cluster scheduler designs, evaluate how much interference between schedulers occurs and how much it matters in practice, present some techniques to alleviate it, and finally discuss a use case highlighting the advantages of our approach -- all driven by real-life Google production workloads.

Mesos – Dynamic Resource Sharing for Clusters

Posted on November 21, 2011 by kilov

Mesos is a cluster manager that provides efficient resource isolation and sharing across distributed applications, or frameworks. It can run Hadoop, MPI, Hypertable, Spark (a new framework for low-latency interactive and iterative jobs), and other applications. Mesos is open source in the Apache Incubator.



Q

Optimistic vs Pessimistic Internal vs External State Single vs Multi Level Fixed vs Pluggable Service vs Batch Oriented



-amplab//

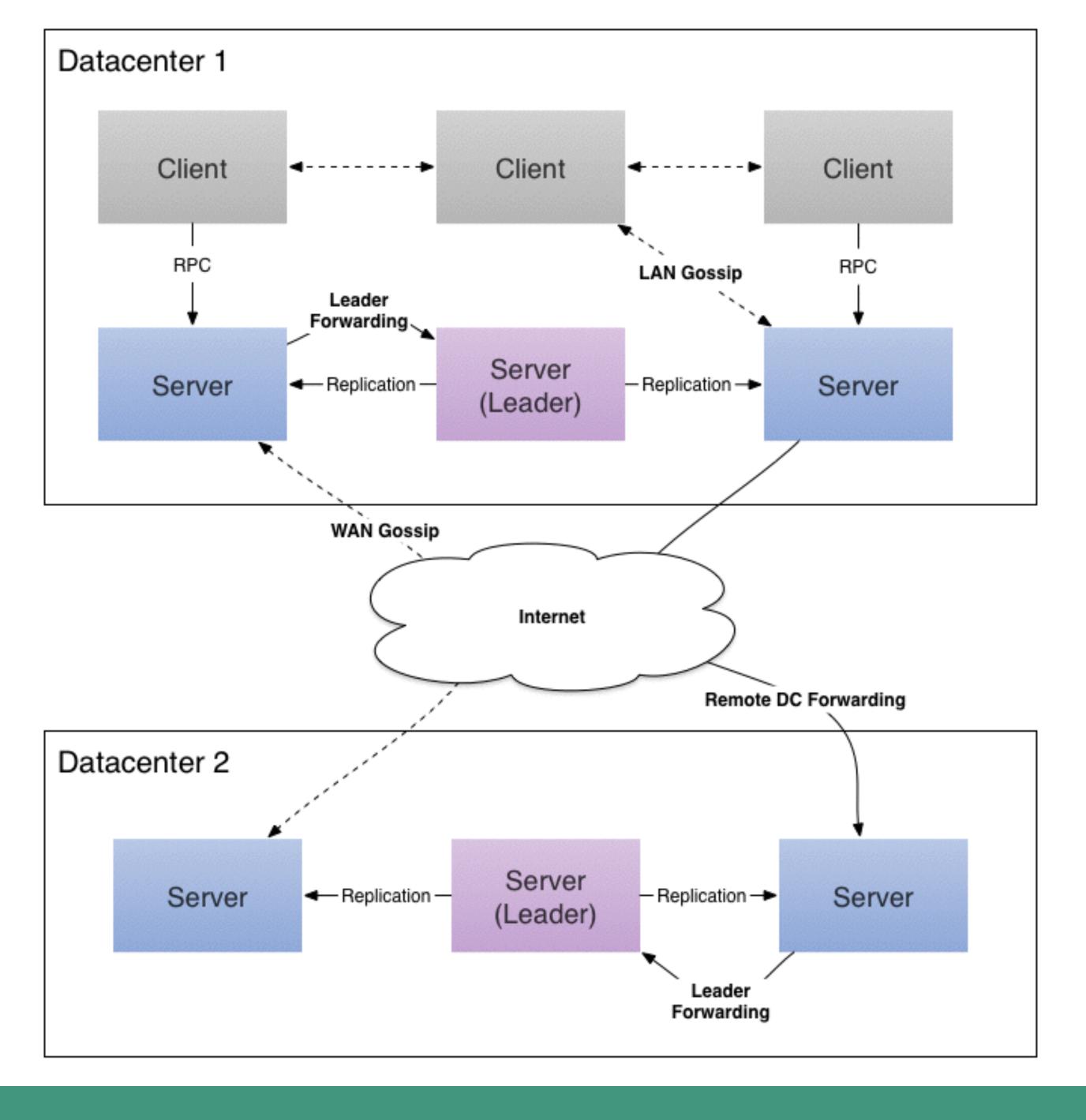




Inspired by Google Omega **Optimistic Concurrency** Internal State and Coordination Service and Batch workloads Pluggable Architecture



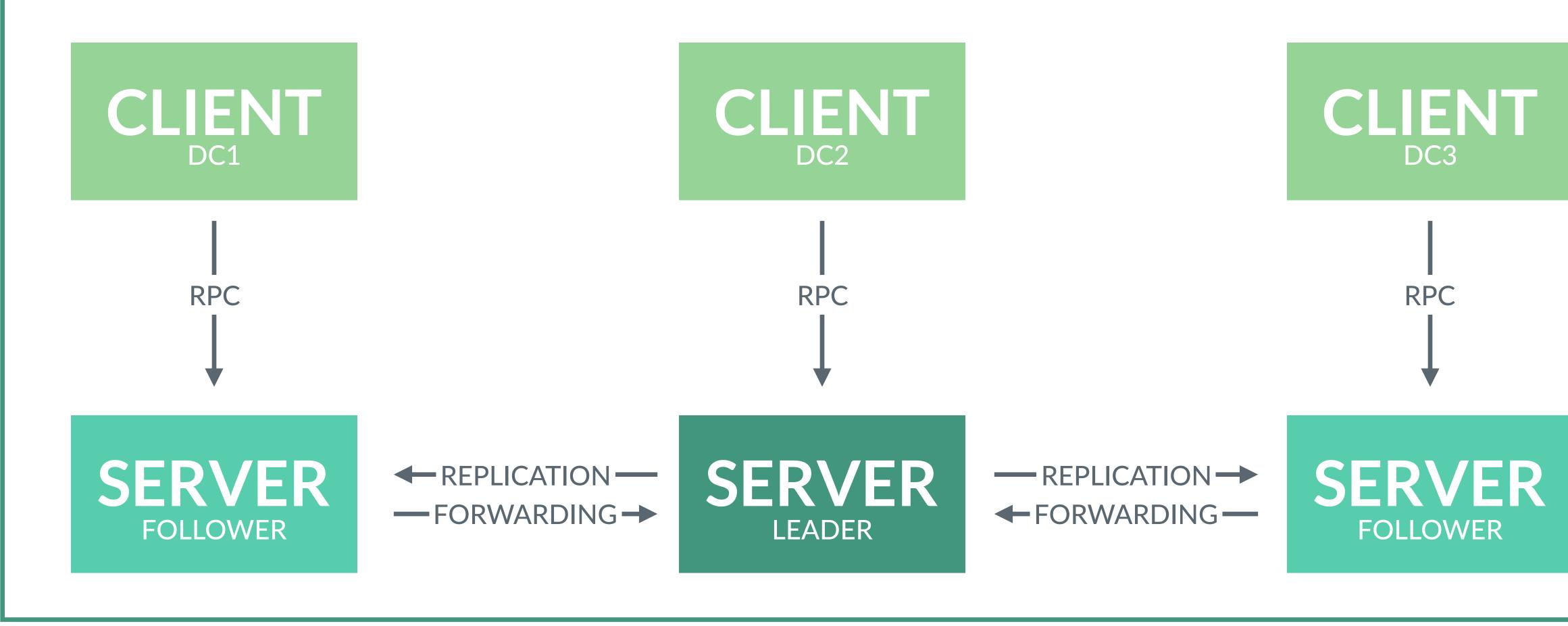




Multi-Datacenter Servers Per DC Failure Isolation Domain is the Datacenter





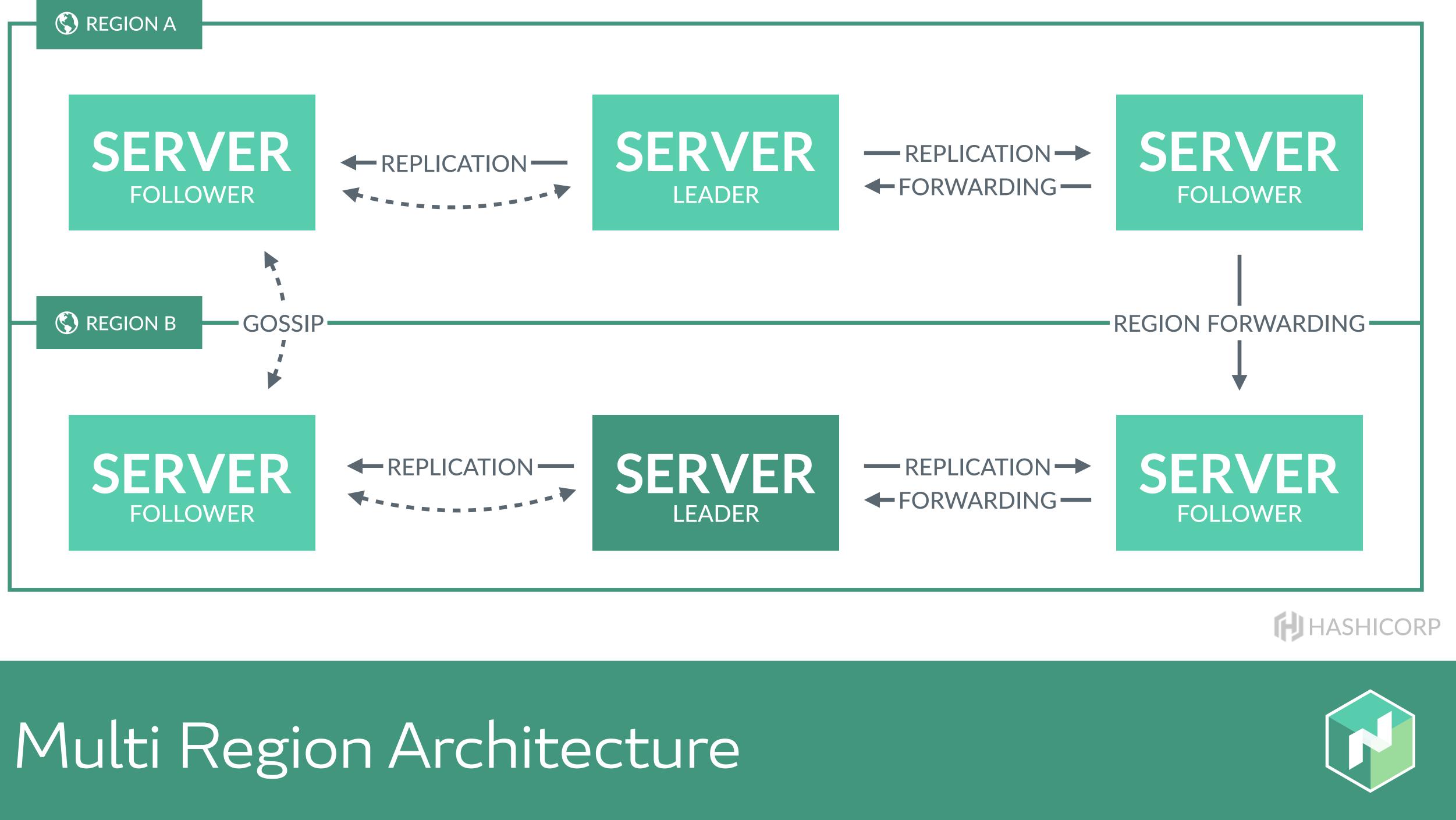


Single Region Architecture











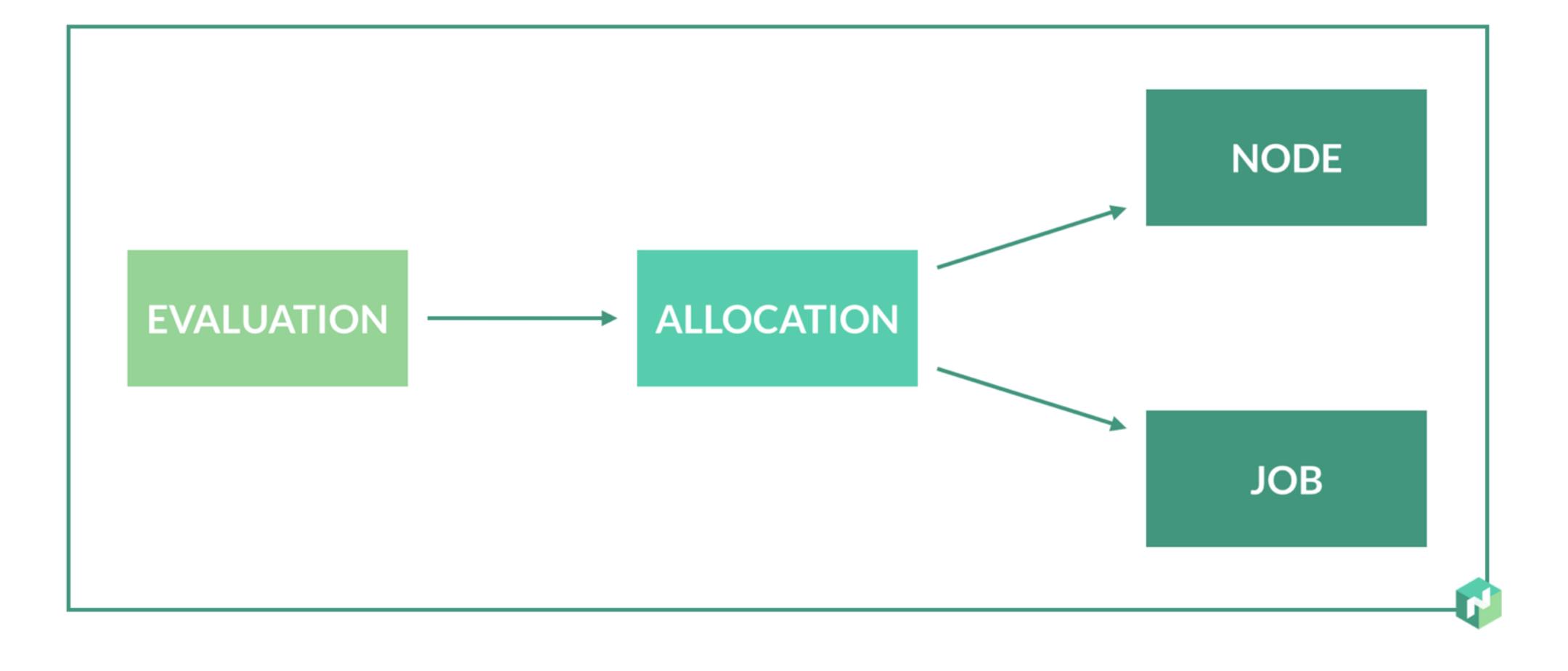


Region is Isolation Domain 1-N Datacenters Per Region Flexibility to do 1:1 (Consul) Scheduling Boundary



HASHICORP





Data Model





Evaluations ~= State Change Event





Create / Update / Delete Job Node Up / Node Down Allocation Failed



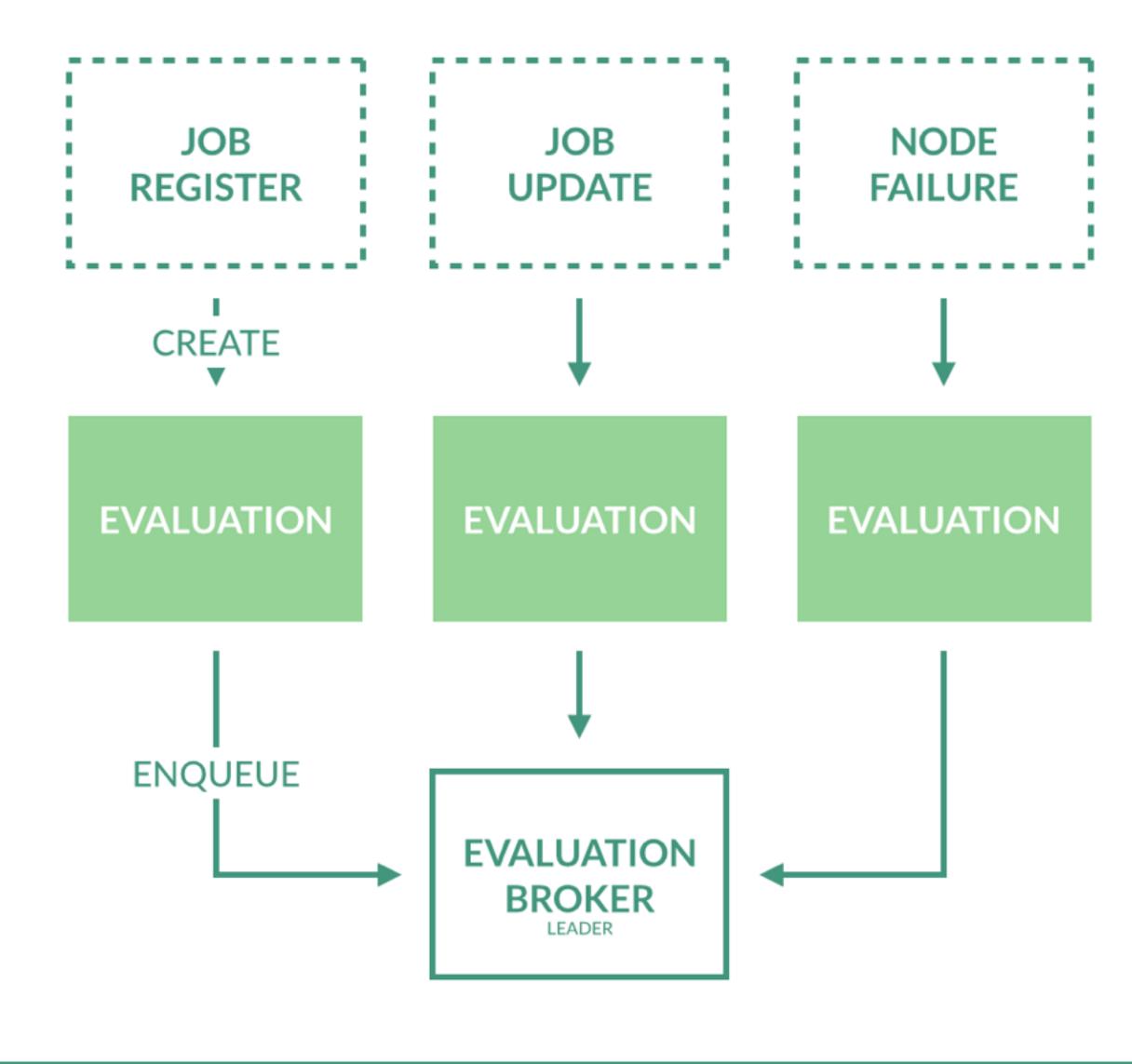
"Scheduler" = func(Eval) => []AllocUpdates





Scheduler func's can specialize (Service, Batch, System, etc)

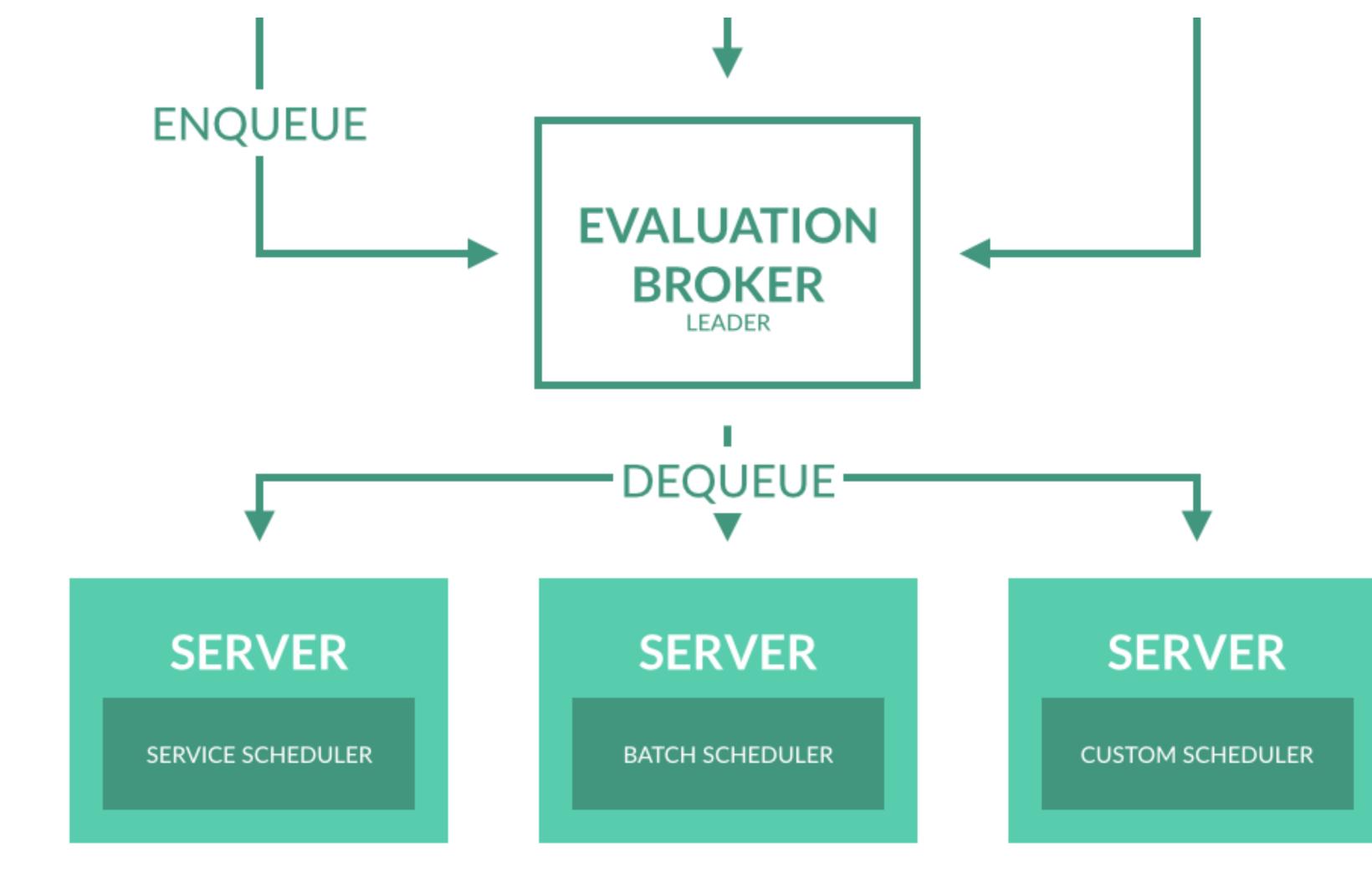




Evaluation Enqueue



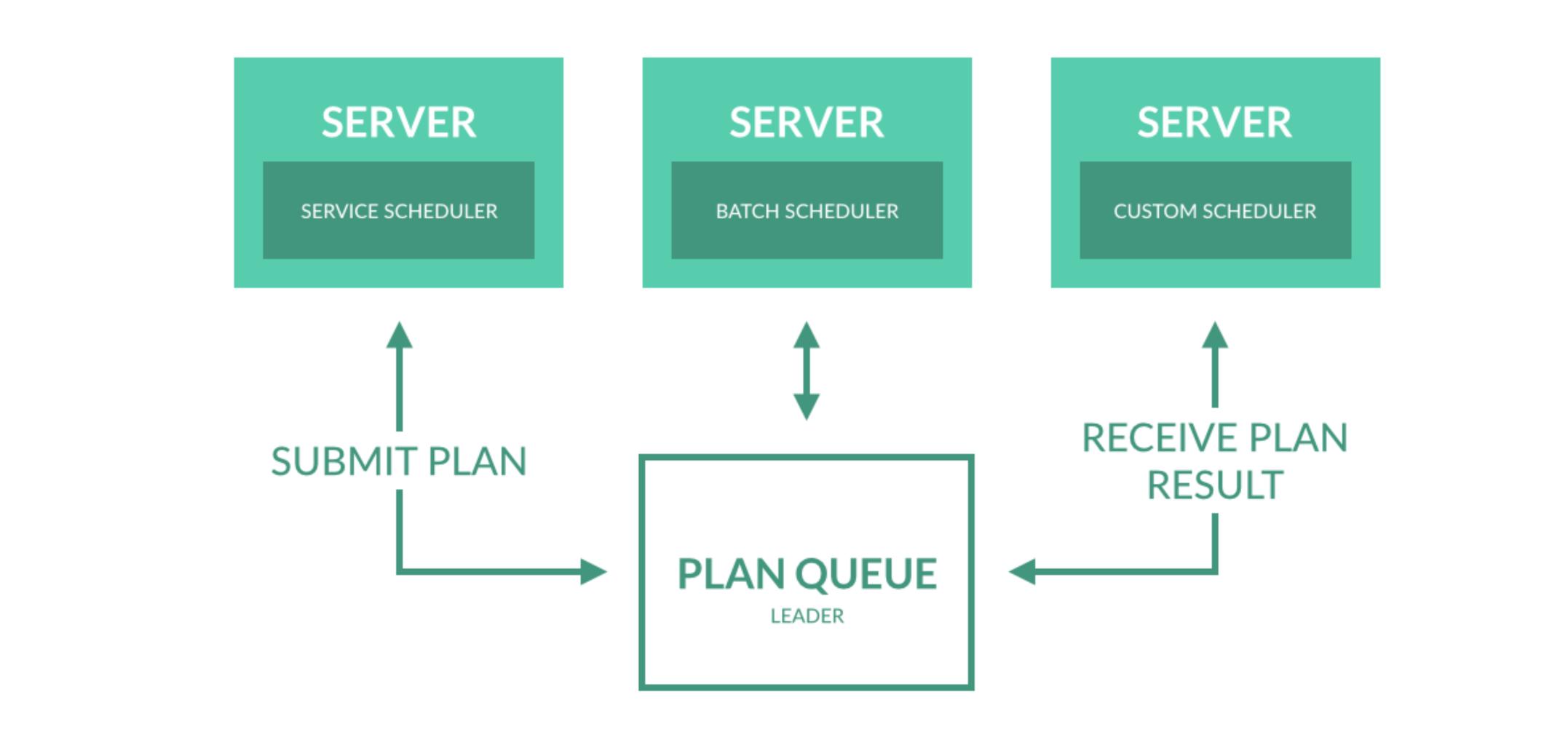




Evaluation Dequeue



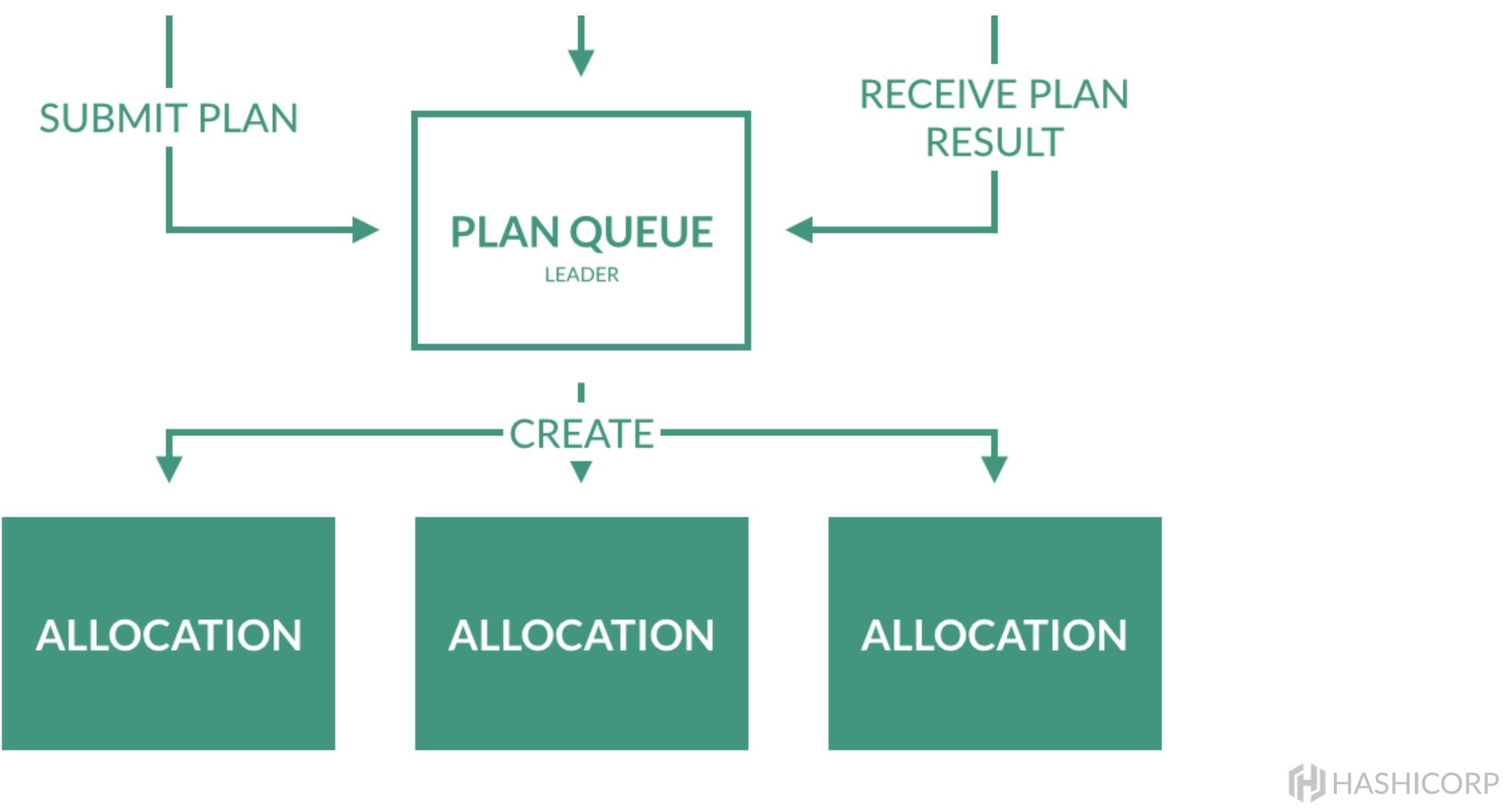




Plan Generation

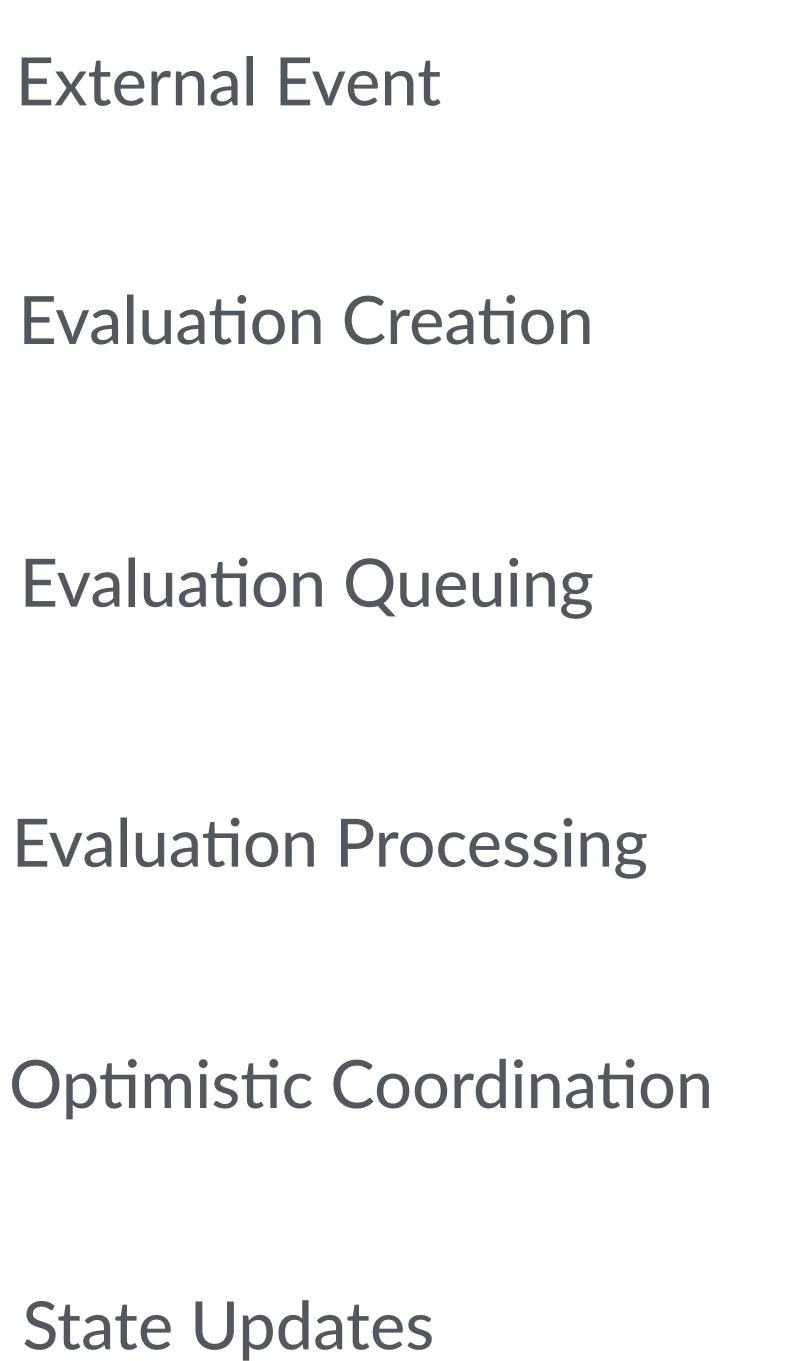
HASHICORP

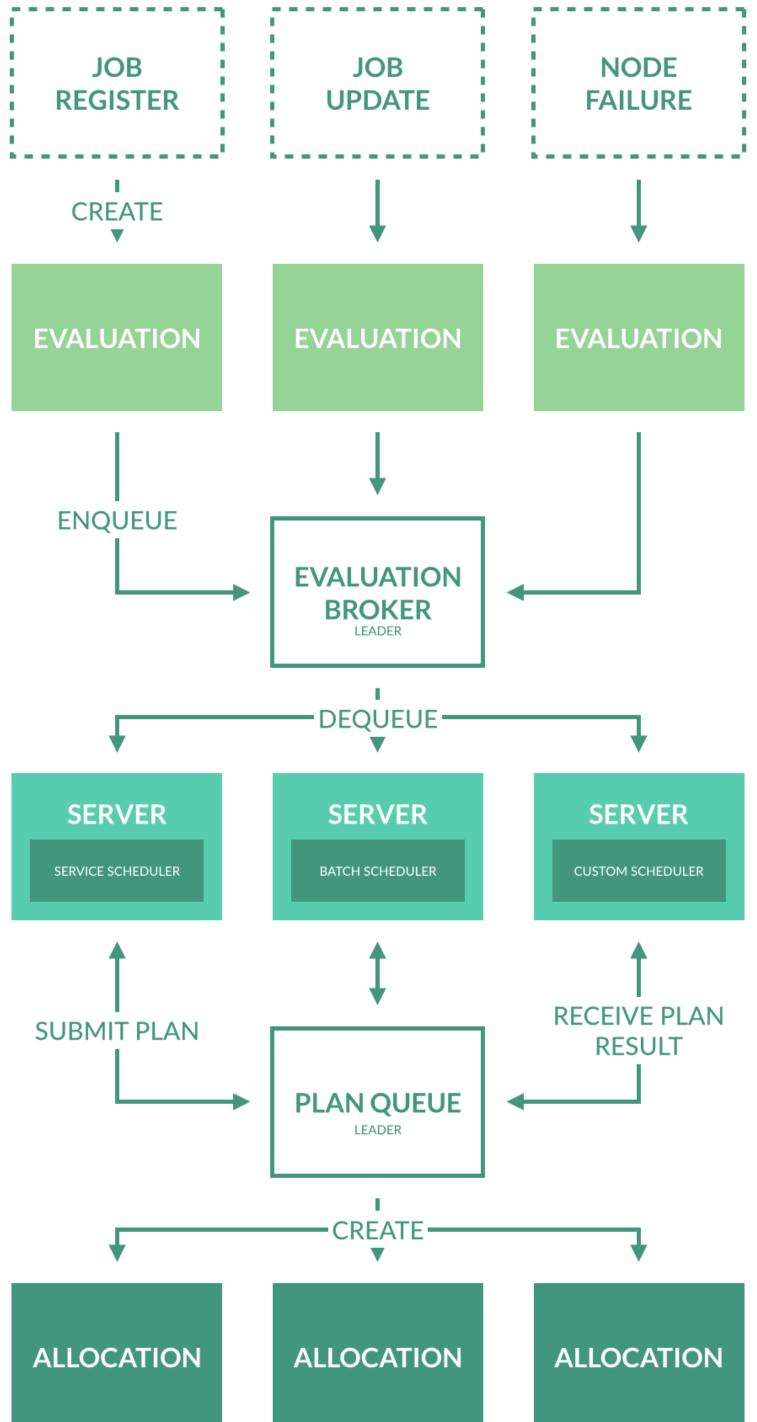




Plan Execution







HASHICORP



Omega Class Scheduler Pluggable Logic Internal Coordination and State Multi-Region / Multi-Datacenter

Server Architecture





Broad OS Support Host Fingerprinting Pluggable Drivers

Client Architecture





Type

Operating System Hardware

Applications

Environment

Fingerprinting

Examples

Kernel, OS, Versions CPU, Memory, Disk Java, Docker, Consul AWS, GCE

HASHICORP



Constrain Placement and Bin Pack

Fingerprinting





"Task Requires Linux, Docker, and PCI-Compliant Hardware" expressed as Constraints

Fingerprinting



"Task needs 512MB RAM and 1 Core" expressed as Resource Ask

Fingerprinting





Execute Tasks Provide Resource Isolation







Containerized

Virtualized

Standalone

Docker Rocket

Qemu / KVM

Java Jar

Static Binaries



Containerized

Virtualized

Standalone

- Docker
- Rocket
- Windows Server Containers
- Qemu / KVM
- Xen
- Hyper-V
- Java Jar
- Static Binaries







Workload Flexibility: Schedulers Fingerprints Drivers Job Specification







Operational Simplicity: Single Binary No Dependencies Highly Available











Released in October Service and Batch Scheduler Docker, Qemu, Exec, Java Drivers

Nomad 0.1





Case Study



DigitalOcean

HASHICORP





Case Study

3 servers in NYC3 100 clients in NYC3, SFO1, AMS2/3 1000 Containers







Case Study

<1s to schedule 1s to first start 6s to 95% 8s to 99%





Service Discovery System Scheduler Restart Policies

Nomad 0.2 - Service Workloads

Enhanced Constraints

HASHICORP



Nomad 0.3 - Batch Workloads

Cron Job Queuing Latency-Aware Scheduling





Production Hardening

Nomad 0.2 in Prod Stress Testing Atlas Integration







Cluster Scheduler Easily Deploy Applications **Job Specification**







Higher Resource Utilization Decouple Work from Resources Better Quality of Service







Thanks! Q/A

