

Resource Optimization in Kubernetes

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Adobe Experience Manager (AEM)

- An existing distributed Java OSGi application
- Using OSS components from Apache Software Foundation
- A huge market of extension developers
- Writing modules that run in-process on AEM

AEM on Kubernetes

- Running on Azure
- 25+ clusters and growing
- Multiple regions: US, Europe, Australia, Japan, more coming
- Customers can run their own code
- Cluster permissions are limited for security

AEM on Kubernetes

Using namespaces to provide a scope

- network isolation
- quotas
- permissions

AEM on Kubernetes

Each customer environment is a micro-monolith ™

Multiple teams building services

Need ways to scale that are orthogonal to the dev teams

Resources in Kubernetes

Kubernetes workloads must set resource requests and limits:

- Requests: how many resources are guaranteed
- Limits: how many resources can be consumed

Resources in Kubernetes

And are applied to

- CPU: may result in CPU throttling
- Memory: limit enforced, results in Kernel OOM killed
- Ephemeral storage: limit enforced, results in pod eviction

Resources in Kubernetes

- AEM is a Java application
- JVM takes all the memory on startup and manages it
- JVM memory use is hidden from Kubernetes, which sees all of it as used
- JDKs >11 will detect the available memory in the container, not the host

Automatically increase and reduce the cluster size

Based on CPU/memory requests

Some head room for spikes

Multiple scale sets in different availability zones

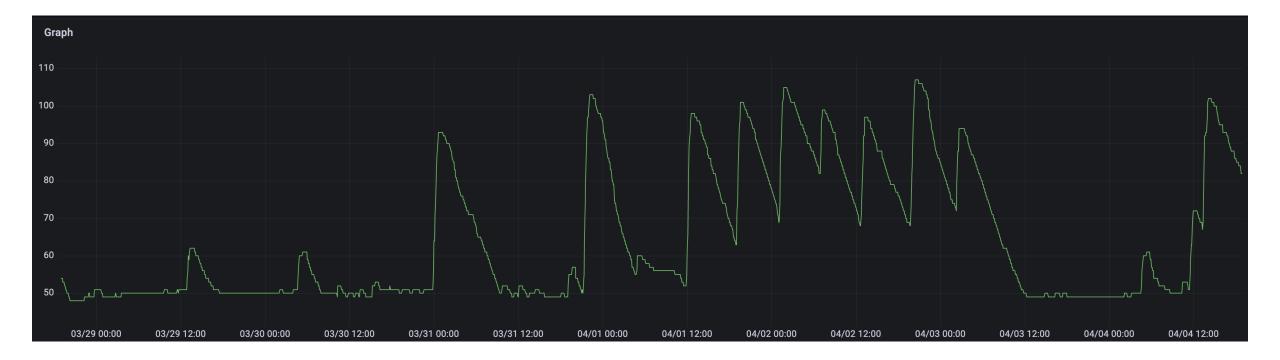
Multiple worker tiers defined as node groups

Max nodes managed at the cluster level

Least waste Scaling Strategy

Selects the node group with the least idle CPU after scaleup

Savings: 30-50%





Horizontal Pod Autoscaler

Creating more pods when needed

AEM scales on CPU and http requests per minute (rpm) metrics

CPU autoscaling is problematic

Periodic tasks can spike the CPU, more pods do not help

Spikes on startup can trigger a cascading effect

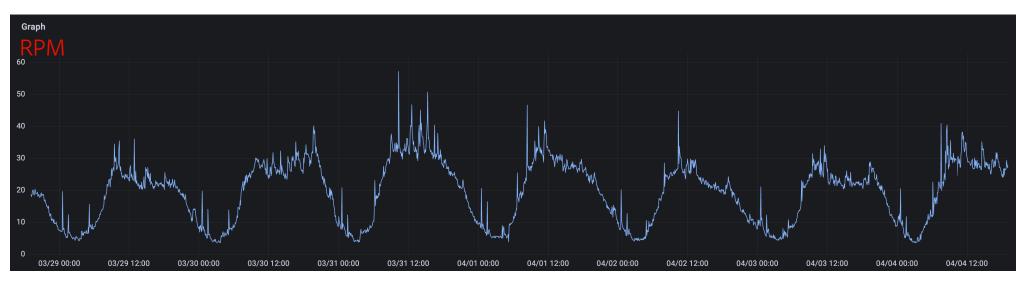
AEM needs to be warmed up on startup

rpm autoscaling is better suited

As long as customers don't have expensive requests Savings: 50-75%

Horizontal Pod Autoscaler





Vertical Pod Autoscaler

Increasing/decreasing the resources for each pod

Allows scaling resources up and down for a deployment

Requires restart of pods (automatic or on next start)

Makes it slow to respond, can exhaust resources in busy nodes

Vertical Pod Autoscaler

Only used in AEM dev environments to scale down if unused

And only for some containers

Savings: 5-15%

Hibernation

Scaling to zero environments not used

Hibernation

Scaling down multiple deployments associated to one "AEM environment"

Deleting ingress routes and other objects that may limit cluster scale

Hibernation

Cronjob that periodically checks for last access data in Prometheus

- UI for user to dehibernate
- Savings: 60-80%

Automatic Resource Configuration

In most clusters services request more cpu/memory than used

ARC can transparently reduce cpu/memory requirements

Limits are not affected, so side effects are limited, would not trigger OOM Killer (likely)

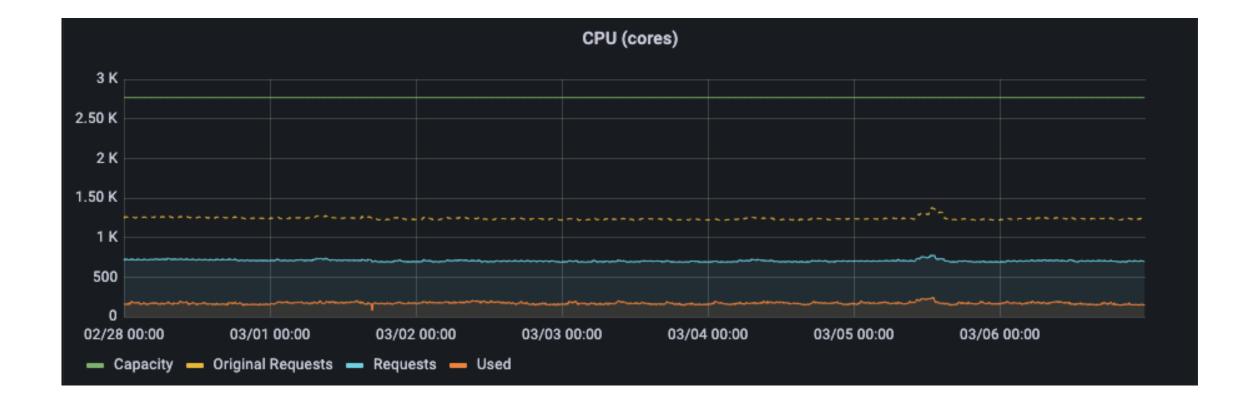
ARC recommender leverages historical metrics at the deployment level

Can provide recommendations about optimization at deployment level based on actual usage

ARC Cluster Ratios

ARC can dial down resource requests at cluster/namespace level

Savings: 10-15%



ARC Recommender

Why ARC and not VPA recommender?

- Full control over recommendation algorithm
- Implementation at more global cluster level with deployment level recommendations

Resource Optimization in Kubernetes

From Kubernetes ecosystem: Cluster autoscaler, HPA, VPA

Internal:

Hibernation, ARC

At application and infrastructure levels

A combination of them will help you optimize and reduce resources used

