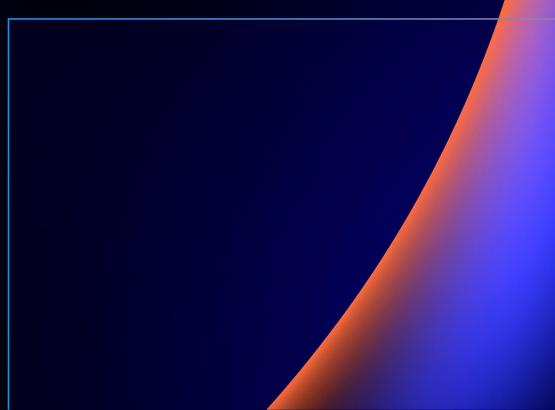
aws re: Invent

NOV. 29 - DEC. 3, 2021 | LAS VEGAS, NV



A N T 3 2 2

Amazon EMR on EKS

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aws

Agenda

- Overview of Amazon EMR
- Overview of Amazon EMR on EKS
- Walkthrough of how a job runs
- Advanced Amazon EMR on EKS usage
- Discussion 🙂

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Amazon EMR

Easily run Spark, Hive, Presto, HBase, Flink, and more big data apps on AWS

Latest versions



Updated with latest open-source frameworks within 30 days

Support for popular OSS like Flink, Hudi Fast performance at low cost



Spark workloads run 2.4x faster compared to open source

50%–80% reduction in costs with Amazon EC2 Spot and Reserved Instances

> Per-second billing for flexibility

Use Amazon S3 storage



Process data in Amazon S3 securely with high performance using the EMRFS connector

Scale compute and storage independent of each other Easy and scalable



Fully managed; no cluster setup, node provisioning, or cluster tuning

Vertical and horizontal Auto Scaling to suit workload demands

Amazon EMR on EKS



Run Apache Spark jobs on demand on **Amazon Elastic Kubernetes Service (EKS)** – without provisioning EMR clusters – to improve resource utilization and simplify infrastructure management

How Amazon EMR on EKS solves the problem

DevOps team

- Auto scaling considerations
- Patching and upgrading
- Integration with the rest of the platform

Data engineers/ data scientists

- Performance optimization
- Cold startup time
- Data isolation



- Cost optimization
- Security
- Access control



Amazon EMR on EKS as a managed service is integrated with AWS natively Amazon EMR on EKS provides 2.4x better performance; k8s multi-tenancy design Amazon EMR on EKS charges per job and achieves higher resource utilization with other microservices

Amazon EMR helps accelerate move to EKS



Provide managed distribution Spark on Kubernetes (2.4 and 3.1) Manage job execution on your behalf

APACHE Spark Simplify secure execution using granular access control

Native integration with Amazon S3, AWS Glue Data Catalog, and more

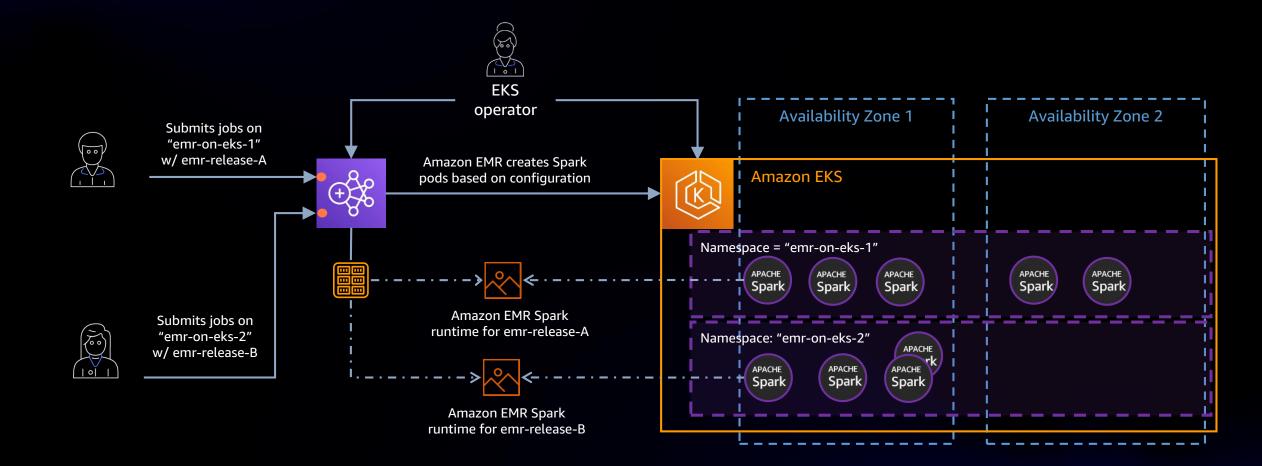
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Simplify debugging with Spark History Server

Support integration with Apache Airflow

Differentiated performance with Amazon EMR runtime for Apache Spark

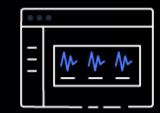
Job-centric workflow



Job submission options









AWS Command Line Interface (AWS CLI)

AWS Tools and AWS SDK

Apache Airflow AWS Step Functions





Job debugging options









Amazon CloudWatch

Amazon Simple Storage Service (Amazon S3)

AWS Management Console

Amazon EMR Studio



How to run a Spark job on Amazon EMR on EKS

```
aws emr-containers start-job-run \
    --virtual-cluster-id cluster_id \
    --name sample-job-name \
    --execution-role-arn execution-role-arn \
    --release-label emr-6.3.0-latest \
    --job-driver '{
        "sparkSubmitJobDriver": {
            "entryPoint": "local:///usr/lib/spark/examples/src/main/python/pi.py",
            "sparkSubmitParameters": "--conf spark.executor.instances=2 --conf spark.executor.memory=2G --conf
spark.executor.cores=2 --conf spark.driver.cores=1"
        }
    }'
```

Pod templates

- Schedule Spark executors to run on Amazon EC2 Spot Instances
- Run a separate "sidecar" container next to the Spark driver or executor – logging, additional monitoring
- Run an "init" container that prepares the environment, e.g., downloading jars

•••

```
apiVersion v1
kind Pod
spec
 volumes
     name source-data-volume
     emptyDir: {}
     name metrics-files-volume
     emptvDir {}
 nodeSelector
   eks.amazonaws.com/nodegroup: emr-containers-nodegroup
 containers
   name: custom-side-car-container # Sidecar container
   image: <side_car_container_image>
       name RANDOM SIDECAR
       value random
   volumeMounts
       name metrics-files-volume
       mountPath /var/metrics/data
   command
       /bin/sh
       '-C'
        <command-to-upload-metrics-files>
  initContainers.
   name: spark-init-container-driver # Init container
   image: <spark-pre-step-image>
   volumeMounts
       name: source-data-volume # Use EMR predefined volumes
       mountPath /var/data
   command
       /bin/sh
        <command-to-download-dependency-jars>
```



Custom containers

- Install and configure packages specific to your workload
- Set environment variables
- Incorporate data pipeline into CI/CD

•••

FROM 895885662937.dkr.ecr.us-west-2.amazonaws.com/spark/emr-6.3.0-latest

USER root

Install Chrome

Install bokeh and sampledata

RUN pip3 install \
 bokeh>=2.3.2 \
 chromedriver-py>=91.0.4472.19.0 \
 selenium>=3.141.0
RUN bokeh sampledata

USER hadoop:hadoop

Resources

EMR Containers Best Practices Guide https://aws.github.io/aws-emr-containers-best-practices/

Amazon EMR on Amazon EKS Demos https://youtube.com/playlist?list=PLUe6KRx8LhLpJ8CyNHewFYukWm7sQyQrM

EMR on EKS Workshop (Advanced) https://www.eksworkshop.com/advanced/430_emr_on_eks/

EMR on EKS Workshop https://emr-on-eks.workshop.aws/

GitHub: AWS CDK Example https://github.com/aws-samples/aws-cdk-for-emr-on-eks

GitHub: Custom Image Validation Tool https://github.com/awslabs/amazon-emr-on-eks-custom-image-cli



Workload isolation

- Spark workloads are IO intensive (disk and network) but Kubernetes/Amazon EKS doesn't provide IO isolation
- How to isolate Amazon EMR workloads?
- Label nodes dedicated to Amazon EMR
- Use node selectors with these labels
- Use Taints and Tolerations to explicitely reject other workloads

Workload isolation code example

s–east–1b

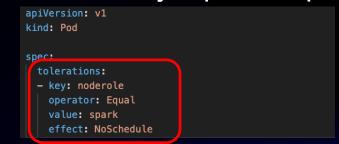
Eksctl nodegroup definition

name: spot-4xl-arm-nvme-1b	
availabilityZones: ["us-east-1b"]	
minSize: Ø	
maxSize: 4	
privateNetworking: true	
instancesDistribution:	
<pre>instanceTypes: ["r6gd.4xlarge"]</pre>	
onDemandBaseCapacity: 0	
onDemandPercentageAboveBaseCapacity: 0	
<pre>spotAllocationStrategy: capacity-optimized</pre>	
labels:	
arch: arm	
disk: nvme	
noderole: spark	
tags.	
k8s.io/cluster-autoscaler/node-template/label/arch: arm	
k8s.io/cluster-autoscaler/node-template/label/kubernetes.io/os: linux	
k8s.io/cluster-autoscaler/node-template/label/noderole: spark	
k8s.io/cluster-autoscaler/node-template/label/disk: nvme	
k8s.io/cluster-autoscaler/node-template/label/node-lifecycle: spot	
k8s.io/cluster-autoscaler/node-template/taint/spot: "true:NoSchedule"	
k8s.io/cluster-autoscaler/node-template/label/topology.kubernetes.io/zor	ie: u
k8s.io/cluster-autoscaler/experiments: owned	
k Os.ic/eluster_autoscaler/enab led: "true"	
taints:	
noderole: "spark:NoSchedule"	
Lam:	
withAddonPolicies:	
ehs: true	

Amazon EMR job definition

"name": "nyTaxiAnalytics",				
"virtualClusterId": "awlrqupqxuy8ijws3c0q2ik81",				
<pre>"executionRoleArn": "arn:aws:iam::152324983118:role/gromav",</pre>				
"releaseLabel": "emr-6.3.0-latest",				
"jobDriver": {				
"sparkSubmitJobDriver": {				
<pre>"entryPoint": "s3://aws-data-lake-workshop/emr-eks-demo/emr-eks-demo-assembly-6.3.0.jar",</pre>				
"sparkSubmitParameters": "class ValueZones"				
}				
},				
"configurationOverrides": {				
"applicationConfiguration": [
{				
"classification": "spark-defaults",				
"properties": 🛛				
"spark.kubernetes.node.selector.noderole": "spark",				
"spark.kubernetes.driver.podTemplateFile": "s3://myBucket/emr-eks-demo/driver-pod-template.yml",				
"spark.kubernetes.executor.podTemplateFile": "s3://myBucket/emr-eks-demo/executor-pod-template.yml"				

Amazon EMR job pod template



Single AZ deployments

- Spark workloads generate lots of network transfer in shuffle phases. Cross AZ network can be costly and degrade performance.
- How to deploy Amazon EMR jobs in single AZ?
- Amazon EKS managed nodegroups automatically add AZ labels (topology.kubernetes.io/zone: us-east-1a)
- Amazon EKS Cluster Autoscaler choose one AZ based on the costs for the Spark driver
- Use AZ level affinity to co-locate Spark executors in the same AZ



Single AZ code example

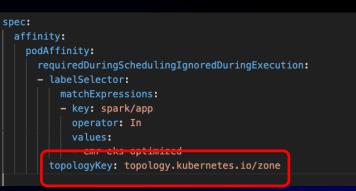
Eksctl nodegroup definition

<pre>- name: spot-4xl-arm-nvme-1b</pre>
availabilityZones: ["us-east-1b"]
minSize: 0
maxSize: 4
privateNetworking: true
instancesDistribution:
instanceTypes: ["r6gd.4xlarge"]
onDemandBaseCapacity: 0
onDemandPercentageAboveBaseCapacity: 0
spotAllocationStrategy: capacity-optimized
labels:
arch: arm
disk: nvme
noderole: spark
tags:
k8s.io/cluster-autoscaler/node-template/label/arch: arm
k8s.io/cluster-autoscaler/node-template/label/kubernetes.io/os: linux
k8s.io/cluster-autoscaler/node-template/label/noderole: spark
k8s.io/cluster-autoscaler/node-template/label/disk: nvme
k8s.io/cluster-autoscaler/node-template/label/node-lifecycle: spot
k8s_io/cluster_autoscaler/node_template/taint/spots_"true:NoSchedule"
k8s.io/cluster-autoscaler/node-template/label/topology.kubernetes.io/zone: us-east-1b
k0s.is/cluster_autoscaler/experiments: owned
k8s.io/cluster-autoscaler/enabled: "true"
taints:
noderole: "spark:NoSchedule"
iam:
withAddonPolicies:
ebs: true

Amazon EMR job definition

name": "nyTaxiAnalytics",					
virtualClusterId": "awlrgupgxuv8ijws3c0g2ik81",					
executionRoleArn": "arn:aws:iam::152324983118:role/gromav",					
releaseLabel": "emr-6.3.0-latest",					
jobDriver": {					
"sparkSubmitJobDriver": 🛛					
"entryPoint": "s3://aws-data-lake-workshop/emr-eks-demo/emr-eks-demo-assembly-6.3.0.jar",					
"sparkSubmitParameters": "class ValueZones"					
configurationOverrides": {					
"applicationConfiguration": [
"classification": "spark-defaults",					
"properties": {					
"spark.kubernetes.node.selector.noderole": "spark",					
"spark.kubernetes.node.selector.arch": "arm".					
"spark.kubernetes.driver.label.spark/app": "emr-eks-optimized",					
"spark.kubernetes.executor.label.spark/app": "emr-eks-optimized",					
"spark.kubernetes.driver.podTemplateFile": "s3://my-bucket/emr-eks-demo/driver-pod-template.yml",					
"spark.kubernetes.executor.podTemplateFile": "s3://my-bucket/emr-eks-demo/executor-pod-template.yml"					

Amazon EMR job pod template



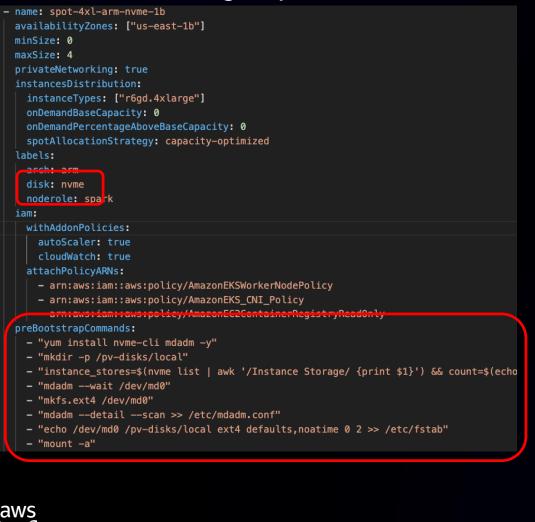
Kubernetes local volumes for tmp data

- Spark exchanges data between nodes (shuffle) and write this tmp data to disk before sending over the network
- Default configuration is using Amazon EC2 root volume used by the kubelet and limited in space and performance
- For shuffle intensive workloads, recommended to use fast local storage
- How to leverage fast local storage?
- Use nodegroups with NVMe instance store (r5d, r5ad, r5dn...) and label them (disk: nvme)
- Configure the bootstrap action to initialize a RAIDO on top of instance stores
- Use node selector with the NVME disk label
- Configure Amazon EMR job to mount the local storage as HostPath volume
- Set the *spark.tmp.dir* to this volume



Kubernetes local volumes code example

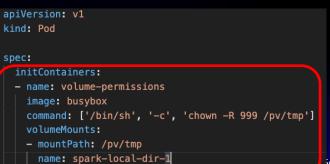
Eksctl nodegroup definition



Amazon EMR job definition

"spark.kubernetes.driver.podTemplateFile": "s3://my-bucket/emr-eks-demo/driver-pod-template.yml",
"spark.kubernetes.executor.podTemplateFile": "s3://my-bucket/emr-eks-demo/executor-pod-template[.yml"

Amazon EMR job pod template



Autoscaling

Amazon EMR Spark application can autoscale using Dynamic Ressource Allocation but requires Amazon EKS resources to start new Pods

Amazon EKS is compatible with Kubernetes Cluster Autoscaler

Scaling in can impact Amazon EMR job if shuffle data are lost

How to autoscale Amazon EMR Spark?

- Install the K8S Cluster Autoscaler
- Configure the nodegroups to be managed by the Cluster Autoscaler
- Configure Dynamic Resource Allocation in EMR job
- Enable beta feature to avoid removing executors with shuffle data spark.dynamicAllocation.shuffleTracking.enabled spark.dynamicAllocation.shuffleTracking.timeout

Optional: Amazon FSX Lustre can be used to persist shuffle data outside executors

Autoscaling code example

-1b

Eksctl nodegroup definition

	- name: spot-4xl-arm-nvme-10
	availabilityZones: ["us-east-1b"]
	minSize: 0
	maxSize: 4
	privateNetworking: true
	instancesDistribution:
	instanceTypes: ["r6gd.4xlarge"]
	onDemandBaseCapacity: 0
	onDemandPercentageAboveBaseCapacity: 0
	spotAllocationStrategy: capacity-optimized
1	tags:
	k8s.io/cluster-autoscaler/node-template/label/arch: arm
	k8s.io/cluster-autoscaler/node-template/label/kubernetes.io/os: linux
	k8s.io/cluster-autoscaler/node-template/label/noderole: spark
	k8s.io/cluster-autoscaler/node-template/label/disk: nvme
	k8s.io/cluster-autoscaler/node-template/label/node-lifecycle: <pre>spot</pre>
	k8s.io/cluster-autoscaler/node-template/taint/spot: "true:NoSchedule"
	k8s.io/cluster-autoscaler/node-template/label/topology.kubernetes.io/zone: us-east-
	k8s.io/cluster-autoscaler/experiments: owned
	k8s.io/cluster-autoscaler/enabled: "true"
	withAddonPolicies:
	ebs: true
	fsx: true

Amazon EMR job definition

```
"name": "nyTaxiAnalytics",
"virtualClusterId":
"executionRoleArn":
"releaseLabel": "emr-6.3.0-latest",
"jobDriver": {
  "sparkSubmitJobDriver": {
    "entryPoint": "s3://aws-data-lake-workshop/emr-eks-demo/emr-eks-demo-assembly-6.3.0.jar",
    "sparkSubmitParameters": "--class ValueZones"
},
"configurationOverrides": {
  "applicationConfiguration": [
      "classification": "spark-defaults",
      "properties": 1
        "spark.dynamicAllocation.enabled":"true",
        "spark.dynamicAllocation.minExecutors": "10",
        "spark.dynamicAllocation.maxExecutors": "20",
        "spark.kubernetes.allocation.batch.size": "10",
        "spark.dynamicAllocation.executorAllocationRatio": "1",
        "spark.dynamicAllocation.shuffleTracking.enabled": "true",
        "spark.dynamicAllocation.shuffleTracking.timeout": "300s",
       "spark_executor_memory": "126"
        "spark.driver.memory": "4G",
        "spark.executor.cores": "3",
```

autoScaler: true

Using Amazon EC2 Spot instances

- Amazon EC2 Spot is a good fit for Spark workloads because Spark has internal mechanism to recover an executor loss
- The Spark driver is the critical component containing the processing logic to recover, missing data is recomputed
- How to use EC2 Spot with EMR?

aws

- Amazon EKS managed nodegroups automatically add lifecycle labels (eksCapacity=SPOT)
- Schedule driver on on-demand and executor on Spot
- Use Taints and Tolerations to explicitly reject non-tolerant workload from EC2 Spot
- External shuffle service to improve resiliency

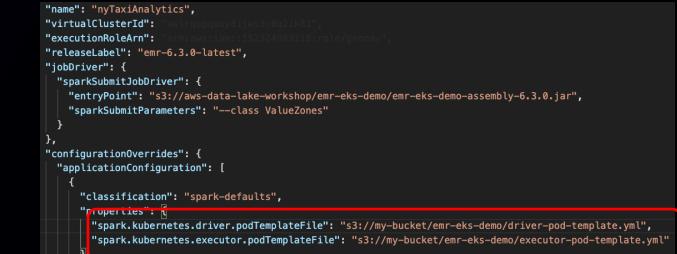
Using Amazon EC2 Spot code example

spot nedule"

Eksctl nodegroup definition

-	- name: spot-4xl-arm-nvme-1b			
	availabilityZones: ["us-east-1b"]			
	minSize: 0			
	maxSize: 4			
	privateNetworking: true			
	instancesDistribution:			
	instanceTypes: ["r6gd.4xlarge", "r6g.4xlarge"]			
	onDemandBaseCapacity: 0			
	onDemandPercentageAboveBaseCapacity: 0			
	spotAllocationStrategy: capacity-optimized			
	Lags.			
	k8s.io/cluster-autoscaler/node-template/label/node-lifecycle: s			
	k8s.io/cluster-autoscaler/node-template/taint/spot: "true:No			
	k8s.io/cluster-autoscaler/experiments: owned			
	kes io/cluster_autoscaler/enabled: "true"			
	taints:			
	spot: "true:NoSchedule"			

Amazon EMR job definition



Amazon EMR job pod template

apiVersion: v1 kind: Pod	
spect	
tolerations:	
– key: spot	
operator: Equal	
value: true	
effect: NoSchedule	
nodeSelector:	
node-lifecycle: spot	

Using AWS Fargate

- AWS Fargate provides serverless compute engine for Amazon EKS
- EMR on EKS can leverage Fargate Pods for both Spark drivers and executors
- How to use Fargate Pods with EMR?
- Create a Fargate profile for the EMR Virtual Cluster namespace
- The Fargate profile can use executors labels to schedule executors only (speed up the scheduling)
- Limitations:
- Approximately 4 minutes to launch Fargate Pods
- Limited disk space for tmp data (20 GB)



Using AWS Fargate code example

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Eksctl fargate profile definition

fargateProfiles:

- name: emr-serverless
- selectors:
- # All workloads in the "emr-serverless" Kubernetes namespace matching the following
- # label selectors will be scheduled onto Fargate:
- namespace: emr-serverless
- # Only Spark executors (Pods with this label) will run on Fargate
- labels
- spark-role: executor

Amazon EMR job definition

```
"name": "nyTaxiAnalytics",
"virtualClusterId":
"executionRoleArn":
"releaseLabel": "emr-6.3.0-latest",
"jobDriver": {
 "sparkSubmitJobDriver": {
    "entryPoint": "s3://aws-data-lake-workshop/emr-eks-demo/emr-eks-demo-assembly-6.3.0.jar",
    "entryPointArguments": ["s3://nyc-tlc/trip data", "2017","s3://nyc-tlc/misc/taxi _zone_looku
    "sparkSubmitParameters": "--class ValueZones"
"configurationOverrides": {
 "applicationConfiguration": [
     "classification": "spark-defaults",
      "properties": {
        "spark.hadoop.hive.metastore.client.factory.class":"com.amazonaws.glue.catalog.metastore
        "spark.dynamicAllocation.enabled":"true",
        "spark.dynamicAllocation.minExecutors": "10",
        "spark.dynamicAllocation.maxExecutors": "20",
        "spark.kubernetes.allocation.batch.size": "10",
       "spark.dynamicAllocation.executorAllocationRatio": "1",
        "spark.dynamicAllocation.shuffleTracking.enabled": "true",
        "spark.dynamicAllocation.shuffleTracking.timeout": "300s",
        "spark.executorEnv.AWS_REGION": "us-east-1",
       "spark.kubernetes.driverEnv.AWS_REGION": "us-east-1",
        "spark.executor.memory": "12G",
        "spark.driver.memory": "4G",
       "snark evecutor cores", "2"
       "spark.kubernetes.driver.label.spark/role": "driver",
        "spark.kubernetes.executor.label.spark/role": "executor"
```

Thank you.

Vincent Gromakowski

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