

Deploy NVIDIA Triton Inference Server (Automated Deployment)

NetApp Solutions

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Deploy NVIDIA Triton Inference Server (Automated Deployment)

To set up automated deployment for the Triton Inference Server, complete the following steps:

1. Open a VI editor and create a PVC yaml file vi pvc-triton-model- repo.yaml.

```
kind: PersistentVolumeClaim
apiVersion: v1
metadata:
  name: triton-pvc namespace: triton
spec:
  accessModes:
   - ReadWriteMany
  resources:
    requests:
    storage: 10Gi
  storageClassName: ontap-flexvol
```

2. Create the PVC.

```
kubectl create -f pvc-triton-model-repo.yaml
```

3. Open a VI editor, create a deployment for the Triton Inference Server, and call the file triton_deployment.yaml.

```
apiVersion: v1
kind: Service
metadata:
 labels:
   app: triton-3gpu
 name: triton-3gpu
  namespace: triton
spec:
  ports:
  - name: grpc-trtis-serving
   port: 8001
   targetPort: 8001
  - name: http-trtis-serving
   port: 8000
   targetPort: 8000
  - name: prometheus-metrics
```

```
port: 8002
    targetPort: 8002
  selector:
   app: triton-3gpu
  type: LoadBalancer
apiVersion: v1
kind: Service
metadata:
 labels:
    app: triton-1gpu
 name: triton-1gpu
 namespace: triton
spec:
 ports:
  - name: grpc-trtis-serving
   port: 8001
   targetPort: 8001
  - name: http-trtis-serving
   port: 8000
   targetPort: 8000
  - name: prometheus-metrics
   port: 8002
   targetPort: 8002
  selector:
    app: triton-1gpu
  type: LoadBalancer
apiVersion: apps/v1
kind: Deployment
metadata:
  labels:
    app: triton-3gpu
  name: triton-3gpu
  namespace: triton
spec:
 replicas: 1
  selector:
   matchLabels:
      app: triton-3gpu version: v1
  template:
    metadata:
      labels:
        app: triton-3gpu
       version: v1
    spec:
```

```
containers:
      - image: nvcr.io/nvidia/tritonserver:20.07-v1-py3
        command: ["/bin/sh", "-c"]
       args: ["trtserver --model-store=/mnt/model-repo"]
       imagePullPolicy: IfNotPresent
       name: triton-3gpu
       ports:
        - containerPort: 8000
        - containerPort: 8001
        - containerPort: 8002
       resources:
         limits:
           cpu: "2"
           memory: 4Gi
           nvidia.com/gpu: 3
         requests:
           cpu: "2"
           memory: 4Gi
           nvidia.com/gpu: 3
       volumeMounts:
        - name: triton-model-repo
         gpu-count: "3"
     volumes:
      - name: triton-model-repo
       persistentVolumeClaim:
         claimName: triton-pvc---
apiVersion: apps/v1
kind: Deployment
metadata:
 labels:
    app: triton-1gpu
 name: triton-1gpu
 namespace: triton
spec:
 replicas: 3
 selector:
   matchLabels:
     app: triton-1gpu
     version: v1
  template:
   metadata:
     labels:
       app: triton-1gpu
       version: v1
    spec:
```

```
containers:
- image: nvcr.io/nvidia/tritonserver:20.07-v1-py3
 command: ["/bin/sh", "-c", "sleep 1000"]
 args: ["trtserver --model-store=/mnt/model-repo"]
 imagePullPolicy: IfNotPresent
 name: triton-1qpu
 ports:
 - containerPort: 8000
  - containerPort: 8001
  - containerPort: 8002
 resources:
   limits:
     cpu: "2"
     memory: 4Gi
     nvidia.com/qpu: 1
   requests:
     cpu: "2"
     memory: 4Gi
     nvidia.com/qpu: 1
 volumeMounts:
  - name: triton-model-repo
   gpu-count: "1"
volumes:
- name: triton-model-repo
 persistentVolumeClaim:
   claimName: triton-pvc
```

Two deployments are created here as an example. The first deployment spins up a pod that uses three GPUs and has replicas set to 1. The other deployment spins up three pods each using one GPU while the replica is set to 3. Depending on your requirements, you can change the GPU allocation and replica counts.

Both of the deployments use the PVC created earlier and this persistent storage is provided to the Triton inference servers as the model repository.

For each deployment, a service of type LoadBalancer is created. The Triton Inference Server can be accessed by using the LoadBalancer IP which is in the application network.

A nodeSelector is used to ensure that both deployments get the required number of GPUs without any issues.

4. Label the K8 worker nodes.

```
kubectl label nodes hci-ai-k8-worker-01 gpu-count=3
kubectl label nodes hci-ai-k8-worker-02 gpu-count=1
```

5. Create the deployment.

```
kubectl apply -f triton_deployment.yaml
```

6. Make a note of the LoadBalancer service external LPS.

```
kubectl get services -n triton
```

The expected sample output is as follows:

rarvind@deployment-jump:~/triton-inference-server\$ kubectl get services -n triton									
NAME	TYPE	CLUSTER-IP	EXTERNAL-IP	PORT (S)	AGE				
triton-1gpu-v20-07-v1	LoadBalancer	10.233.21.185	172.21.231.133	8001:31238/TCP,8000:30171/TCP,8002:32348/TCP	10h				
triton-3gpu-v20-07-v1	LoadBalancer	10.233.13.17	172.21.231.132	8001:31549/TCP,8000:30220/TCP,8002:31517/TCP	10h				

7. Connect to any one of the pods that were created from the deployment.

```
kubectl exec -n triton --stdin --tty triton-1gpu-86c4c8dd64-545lx --/bin/bash
```

8. Set up the model repository by using the example model repository.

```
git clone
cd triton-inference-server
git checkout r20.07
```

9. Fetch any missing model definition files.

```
cd docs/examples
./fetch_models.sh
```

10. Copy all the models to the model repository location or just a specific model that you wish to use.

```
cp -r model_repository/resnet50_netdef/ /mnt/model-repo/
```

In this solution, only the resnet50_netdef model is copied over to the model repository as an example.

11. Check the status of the Triton Inference Server.

```
curl -v <<LoadBalancer_IP_recorded earlier>>:8000/api/status
```

The expected sample output is as follows:

```
curl -v 172.21.231.132:8000/api/status
* Trying 172.21.231.132...
* TCP NODELAY set
* Connected to 172.21.231.132 (172.21.231.132) port 8000 (#0)
> GET /api/status HTTP/1.1
> Host: 172.21.231.132:8000
> User-Agent: curl/7.58.0
> Accept: */*
< HTTP/1.1 200 OK
< NV-Status: code: SUCCESS server id: "inference:0" request id: 9
< Content-Length: 1124
< Content-Type: text/plain
<
id: "inference:0"
version: "1.15.0"
uptime ns: 377890294368
model status {
 key: "resnet50 netdef"
 value {
   config {
      name: "resnet50 netdef"
      platform: "caffe2 netdef"
      version policy {
        latest {
         num versions: 1
        }
      }
      max batch size: 128
      input {
       name: "gpu 0/data"
        data type: TYPE FP32
        format: FORMAT NCHW
        dims: 3
        dims: 224
        dims: 224
      }
      output {
        name: "gpu 0/softmax"
        data type: TYPE FP32
        dims: 1000
        label_filename: "resnet50_labels.txt"
      instance group {
        name: "resnet50 netdef"
        count: 1
```

```
gpus: 0
        gpus: 1
       gpus: 2
       kind: KIND GPU
     default model filename: "model.netdef"
     optimization {
       input pinned memory {
         enable: true
       output pinned memory {
        enable: true
      }
   version status {
     key: 1
     value {
       ready state: MODEL READY
       ready state reason {
ready_state: SERVER_READY
* Connection #0 to host 172.21.231.132 left intact
```

Next: Deploy the Client for Triton Inference Server (Automated Deployment)

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