Singularity GPU Containers Options

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GPU Singularity Containers

NGC Catalog

- NGC (NVIDIA GPU-Accelerated Containers) offers a comprehensive catalog of GPU-accelerated software for Deep Learning, HPC and Visualization applications.
- Consists of containers, pre-trained models, Helm charts for Kubernetes deployments and industry specific AI toolkits with SDK
- Guest Access and Authenticated Access
- GPUs compatible with NGC: V100, A100, T4, Jetson, RTX Quadro.

Docker Hub

- Official Docker images for the machine learning framework
- Containers with GPU support marked "-gpu"
- Usually, Guest Access

For Future reading: NVIDIA Cloud Documentation, List of Nvidia graphics processing units

GPU Compute Capability (NGC)

Compute Capability

- The compute capability of a device is represented by a version number, also sometimes called its "SM version". This version number identifies the features supported by the GPU hardware. The compute capability comprises a major revision number X and a minor revision number Y and is denoted by X.Y
- Note: The compute capability version of a particular GPU should not be confused with the CUDA version (e.g., CUDA 7.5, CUDA 8, CUDA 10.2 etc), which is the version of the CUDA software platform. The CUDA platform is used by application developers to create applications that run on many generations of GPU architectures. New versions of the CUDA platform typically also include software features that are independent of hardware generation.

	Fermi ⁺	Kepler ⁺	Maxwell [‡]	Pascal	Volta	
	sm_20	sm_30	sm_50	sm_60	sm_70 ===	===> Tesla V100
Tesla K20, K40 =	=====>	sm_35	sm_52	sm_61	sm_72	
		sm_37	sm_53	sm_62		

For Future reading: <u>CUDA Programming_Guide</u>, <u>CUDA-Enabled GPUs</u>, <u>NVIDIA Deep Learning Frameworks</u>

Running NGC Containers with Singularity (1)

- Register and Activate Your NGC Account
- Generate Your NGC API Key
- Load singularity module:
- \$ module load singularity/3.2.1
- \$ module list
- Currently Loaded Modules:
- singularity/3.2.1
- Setting NGC container registry authentication credentials:
- \$ export SINGULARITY_DOCKER_USERNAME='\$oauthtoken'
- \$ export SINGULARITY_DOCKER_PASSWORD=<NVIDIA NGC API key>
- Pulling container to a local Singularity image:
- \$ export SINGULARITY_TMPDIR=/home/TMP/\$USER
- \$ mkdir -p \$SINGULARITY_TMPDIR
- \$ singularity pull tensorflow_19.06-py3.sif docker://nvcr.io/nvidia/tensorflow:19.06-py3
- For containers from Docker Hub
- \$ singularity pull docker://tensorflow/tensorflow:2.3.1-gpu
- GPU use:

If your host system has an NVIDIA GPU card and a driver installed, you can leverage the card with the --nv option

Running NGC and GPU Containers with Singularity (2)

Checking TensorFlow version

\$ singularity exec --nv tensorflow_19.06-py3.sif python -c 'import tensorflow as tf; print(tf.__version__)'
1.13.1
\$ singularity exec --nv tensorflow_2.3.1-gpu.sif python -c 'import tensorflow as tf; print(tf.__version__)'
2.3.1

Checking available GPUs (Interactive shell)

@ui-1\$ qsub -I -I nodes=1:ppn=2:gpus=1 ==> For getting node with GPU
@ui-1\$ qsub -I -I nodes=1:ppn=2:gpus=1 -I feature=v100 ===> For getting node with Tesla V100 GPU
qsub: waiting for job 1028957.rudens to start
qsub: job 1028957.rudens ready
@wn60 \$ cd \$PBS_O_WORKDIR
@wn60 \$ module load singularity
\$ singularity exec --nv tensorflow_19.06-py3.sif tf_gpu_avail_01.py

['/device:CPU:0', '/device:XLA_CPU:0', '/device:XLA_GPU:0', '/device:XLA_GPU:1', '/device:GPU:0', '/device:GPU:1']

Python script tf_gpu_avail_01.py

...

import tensorflow as tf
import os
from tensorflow.python.client import device_lib
print([device.name for device in device_lib.list_local_devices() if device.name != None])

Running NGC and GPU Containers with Singularity(3)

Running simple TensorFlow script (tf2_gpu_cpu.py)

import tensorflow as tf
import os
import time
cpu_slot = 0
gpu_slot = 0

Using CPU at slot 0
with tf.device('/CPU:' + str(cpu_slot)):
 # Starting a timer
 start = time.time()
 A = tf.constant([[3, 2], [5, 2]])
 print(tf.eye(2,2))
Printing how long it took with CPU
 end = time.time() - start
 print(end)

Using the GPU at slot 0
with tf.device('/GPU:' + str(gpu_slot)):
 # Starting a timer
 start = time.time()
 # Doing operations on CPU
 A = tf.constant([[3, 2], [5, 2]])
 print(tf.eye(2,2))

Printing how long it took with GPU
end = time.time() - start
print(end)
print(('Your TensorFlow version: {0}').format(tf.__version__))

This script is comparing time spent using the CPU versus GPU. The simple operation here is creating a constant and an identity matrix by defining a tensor A, getting the rows and columns and making an identity matrix.

@ui-2 tf_gpu]\$ qsub -I -I nodes=1:ppn=2:gpus=1
qsub: waiting for job 1045525.rudens to start
qsub: job 1045525.rudens ready
@wn56 ~]\$ cd \$PBS_O_WORKDIR
@wn56 ~]\$ module load singularity
@wn56 tf_gpu]\$ singularity exec --nv tensorflow_2.3.1-gpu.sif python
tf2_gpu_cpu.py

...
tf.Tensor(
[[1. 0.]
[0. 1.]], shape=(2, 2), dtype=float32)
0.006877899169921875 ===> GPU time
tf.Tensor(
[[1. 0.]
[0. 1.]], shape=(2, 2), dtype=float32)
0.13370490074157715 ===> CPU time

Notes on Running NGC Containers with Singularity

- If your host system has an NVIDIA GPU card and a driver installed, you can leverage the card with the --nv option
- There is currently a bug in Singularity 3.1.x and 3.2.x causing the LD_LIBRARY_PATH to be incorrectly set within <u>some</u> <u>container's</u> environment. As a workaround the LD_LIBRARY_PATH must be unset before invoking Singularity:

\$ LD_LIBRARY_PATH="" singularity exec

• There is not necessary to load cuda module due to containers already has installed CUDA environment. In some cases, may be necessary to set for compatibility issues \$CUDA_HOME variable by loading module cuda, or searching for other versions:

\$ module spider cuda
Versions:
 cuda/cuda-7.5
 cuda/cuda-8.0
 cuda/cuda-9.2
 cuda/cuda-10.1
 cuda/cuda-10.2

Main Singularity Commands

Main online container registries and commands to load

Singularity Library: Docker Hub: Singularity Hub:

https://cloud.sylabs.io/library : \$ singularity pull library:// https://hub.docker.com : https://singularity-hub.org : NVIDIA GPU Cloud: https://ngc.nvidia.com:

\$ singularity pull docker:// \$ singularity pull shub:// \$ singularity pull docker://nvcr.io/

\$ singularity pull docker://gcc:5.3.0 \$ singularity pull library://godlovedc/demo/lolcow \$ singularity pull docker://nvcr.io/nvidia/tensorflow:19.06-py3

Image examinations

\$ singularity verify lolcow latest.sif \$ singularity inspect lolcow latest.sif \$ singularity inspect -d lolcow latest.sif \$ singularity inspect -r lolcow latest.sif

Verify signatures # To show labels # To show the Singularity recipe file (deffile) # To inspect the runscript

Singularity cache

\$ HOME/.singularity/cache/ \$ singularity cache list \$ singularity cache clean -a \$ singularity cache clean --type=library,oc

User's Singularity cache directory # List and size of cached files # clean all the cache # clean only library, and oci cache

Run Command

Run defined set of commands from a definition file's runscript. It is only available when using an image that was built from a definition file that specified a runscript.

\$ singularity run library://godlovedc/demo/lolcow \$ singularity run ./lolcow latest.sif

Exec command Executes command from container

\$ singularity exec ./lolcow latest.sif fortune \$ singularity exec library://godlovedc/demo/lolcow fortune

Shell command

The shell command allows you to spawn a new shell within your container and interact with it as though it were a small virtual machine

\$ singularity shell library://godlovedc/demo/lolcow \$ singularity shell container.sif

User-defined bind paths: \$ singularity shell --bind /scratch, src[:dest[:opts]], container.sif **GPU use:** If your host system has an NVIDIA GPU card and a driver installed, you can leverage the card with the --nv option : \$ singularity run --nv container.sif

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