NVIDIA Magnum IO GPUDirect Storage

Configuration and Benchmarking Guide
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Chapter 1. Introduction

NVIDIA® GPUDirect® Storage (GDS) is the newest addition to the GPUDirect family. GDS enables a direct data path for direct memory access (DMA) transfers between GPU memory and storage, which avoids a bounce buffer through the CPU. This direct path increases system bandwidth and decreases the latency and utilization load on the CPU.

The purpose of this guide is to help the user evaluate and test GDS functionality and performance by using sample applications. These applications can be run after you set up and install GDS and before you run the custom applications that have been modified to take advantage of GDS.

Refer to the following guides for more information about GDS:

- GPUDirect Storage Design Guide
- GPUDirect Storage Overview Guide
- cuFile API Reference Guide
- GPUDirect Storage Release Notes
- GPUDirect Storage Best Practices Guide
- GPUDirect Storage Troubleshooting Guide
- GPUDirect Storage O_DIRECT Requirements Guide

To learn more about GDS, refer to the following posts:

- GPUDirect Storage: A Direct Path Between Storage and GPU Memory.
- The Magnum IO series.
Chapter 2. About this Guide

Configuration and benchmarking are very tightly coupled activities. Benchmarking provides the ability to determine the potential performance based on the current system configuration, and the impact of configuration changes. Configuration changes are sometimes required to achieve optimal benchmark results, which will potentially translate into increased performance of production workloads.

This guide provides information and examples of the various system configuration attributes, both hardware and software, and how they factor into the delivered performance of GPUDirect Storage. Local drive configurations (Direct Attached Storage - DAS) and Network storage (Network Attached Storage - NAS) are covered. The benchmarking tool included when GDS is installed, gdsio, is covered and its use demonstrated.

Appendix A covers benchmarking and performance in general, along with considerations when benchmarking storage systems.
Chapter 3. Benchmarking GPUDirect Storage

GDS enables high throughput and low latency data transfer between storage and GPU memory, which allows you to program the DMA engine of a PCIe device with the correct mappings to move data in and out of a target GPU's memory. As such, it becomes clear the path between the GPU and the network card or storage device/controller factors significantly into delivered performance, both throughput and latency. The PCIe topology, PCIe root complex, switches and the physical location of the GPU and network and storage devices need to be examined and factored into the configuration details when benchmarking GDS.

Achieving optimal performance with GDS benchmarking requires working through the PCIe topology and determining:

- which IO devices and GPUs are on the same PCIe switch or root complex
- which device communication paths require traversing multiple PCIe ports and possibly crossing CPU socket boundaries

The diagram in the following section illustrates an example of PCIe topology, showing different devices across multiple PCIe switches.

Determining PCIe device proximity is not necessarily an easy task, as it requires using multiple Linux utilities to correlate device names and numbers to the hierarchical numbering scheme used to identify PCIe devices, referred to as BDF notation (bus:device.func) or extended BDF notation, which adds a PCIe domain identifier to the notation, as in domain:bus:device.func.

```
$ lspci | grep -i nvidia
36:00.0 3D controller: NVIDIA Corporation Device 20b0 (rev a1)
```

```
$ lspci -D | grep -i nvidia
0000:36:00.0 3D controller: NVIDIA Corporation Device 20b0 (rev a1)
```

In the first example, note the standard PCIe BDF notation for the first NVIDIA GPU, 36:00.0. In the second example, the -D flag was added to show the PCIe domain (extended BDF), 0000:36:00.0.
3.1. Determining PCIe Device Affinity

The examples in this section were performed on an NVIDIA® DGX-2™ system. The figure below shows a subset of the DGX-2 system architecture, illustrating the PCIe topology:

A DGX-2 system has two CPU sockets, and each socket has two PCIe trees. Each of the four PCIe trees (only one is shown above) has two levels of switches. Up to four NVMe drives hang off of the first level of switches. Each second-level switch has a connection to the first level switch, a PCIe slot that can be populated with a NIC or RAID card, and two GPUs.

The commands and methodology in the following sample output apply to any system that runs Linux. The goal is to associate GPUs and NVMe drives in the PCIe hierarchy and determine which device names to use for GPUs and NVMe drives that share the same upstream PCIe switch. To resolve this issue, you must correlate Linux device names with PCIe BDF values. For the locally attached NVMe disks, here is an example that uses Linux /dev/disk/by-path directory entries:

dgx2> ls -l /dev/disk/by-path
lrwxrwxrwx 1 root root  9 Nov 19 12:08 pci-0000:00:14.0-usb-0:8.1:1.0-scsi-0:0:0:0 -> ../../../sr0

Figure 1. PCIe Topology
Since the current system configuration has nvme0 and nvme1 devices configured into a RAID0 device (/dev/md0 not shown here), the focus is on the remaining available nvme devices, nvme2 through nvme9. You can get the same PCIe-to-device information for the GPUs that use the nvidia-smi utility and specify the GPU attributes to query:

```
dgx2> nvidia-smi --query-gpu=index,name,pci.domain,pci.bus,pci.device,pci.device_id,pci.sub_device_id --format=csv
index, name, pci.domain, pci.bus, pci.device, pci.device_id, pci.sub_device_id
0, Tesla V100-SXM3-32GB, 0x0000, 0x34, 0x00, 0x1DB810DE, 0x12AB10DE
1, Tesla V100-SXM3-32GB, 0x0000, 0x36, 0x00, 0x1DB810DE, 0x12AB10DE
2, Tesla V100-SXM3-32GB, 0x0000, 0x39, 0x00, 0x1DB810DE, 0x12AB10DE
3, Tesla V100-SXM3-32GB, 0x0000, 0x3B, 0x00, 0x1DB810DE, 0x12AB10DE
4, Tesla V100-SXM3-32GB, 0x0000, 0x57, 0x00, 0x1DB810DE, 0x12AB10DE
5, Tesla V100-SXM3-32GB, 0x0000, 0x59, 0x00, 0x1DB810DE, 0x12AB10DE
6, Tesla V100-SXM3-32GB, 0x0000, 0x5C, 0x00, 0x1DB810DE, 0x12AB10DE
7, Tesla V100-SXM3-32GB, 0x0000, 0x5E, 0x00, 0x1DB810DE, 0x12AB10DE
8, Tesla V100-SXM3-32GB, 0x0000, 0xB7, 0x00, 0x1DB810DE, 0x12AB10DE
9, Tesla V100-SXM3-32GB, 0x0000, 0xB9, 0x00, 0x1DB810DE, 0x12AB10DE
10, Tesla V100-SXM3-32GB, 0x0000, 0xBC, 0x00, 0x1DB810DE, 0x12AB10DE
11, Tesla V100-SXM3-32GB, 0x0000, 0xBE, 0x00, 0x1DB810DE, 0x12AB10DE
12, Tesla V100-SXM3-32GB, 0x0000, 0xE0, 0x00, 0x1DB810DE, 0x12AB10DE
13, Tesla V100-SXM3-32GB, 0x0000, 0xE2, 0x00, 0x1DB810DE, 0x12AB10DE
14, Tesla V100-SXM3-32GB, 0x0000, 0xE5, 0x00, 0x1DB810DE, 0x12AB10DE
15, Tesla V100-SXM3-32GB, 0x0000, 0xE7, 0x00, 0x1DB810DE, 0x12AB10DE
```

Use the Linux lspci command to tie it all together:

```
dgx2> lspci -tv | egrep -i "nvidia | micron"
+-------[0000:d7]-+-00.0-[d8-e7]----00.0-[d9-e7]----00.0-[da]----00.0 Micron Technology
         |                  +---0.0-[db]----0.0 Micron Technology
         |                  +---0.0-[de-e2]----00.0-[df-e2]----0.0-
         |                  +---0.0-[e0]----0.0 NVIDIA Corporation GV100GL [Tesla V100 SXM3 32GB]
         |                  |                  +---0.0-[e2]----0.0 NVIDIA Corporation GV100GL [Tesla V100 SXM3 32GB]
         |                  |                  +---0.0-[e5]----0.0 NVIDIA Corporation GV100GL [Tesla V100 SXM3 32GB]
         |                  |                  +---0.0-[e7]----0.0 NVIDIA Corporation GV100GL [Tesla V100 SXM3 32GB]
         |                  |                  +---0.0-[af-c7]----0.0-[b0-c7]----0.0-[b1]----0.0 Micron Technology
         |                  +---0.0-[b2]----0.0 Micron Technology
         +-------[0000:ae]-+-00.0-[af-c7]----0.0-[b0-c7]----0.0-[b1]----0.0 Micron Technology
         |                  +---0.0-[b2]----0.0 Micron Technology
         +-------[0000:bf]-+-00.0-[b0-c7]----0.0-[b1]----0.0 Micron Technology
         |                  +---0.0-[b2]----0.0 Micron Technology
```

NVIDIA Magnum IO GPUDirect Storage
In the above example, we explicitly searched for SSDs from the given vendor. To determine the manufacturer of the NVMe SSD devices on your system, simply run `lsblk -o NAME,MODEL`. Alternatively, use `nvme` as the string to match with `nvidia`.

A few things to note here. First, the NVME SSD devices are grouped in pairs on each of the PCIe upstream switches, as shown in the displayed extended BDF format (left most column), showing domain zero, and Bus IDs 0xd7 (0000:d7), 0xae, 0x4e and 0x2b. Also, two distinct NVIDIA device IDs are revealed (right-most column) - 0x1db8 and 0x1ac2.
The 0x1db8 devices are the Tesla V100 SXM3 32GB GPUs, and the 0x1ac2 devices are NVSwitches. Our interest here is in the GPU devices, and the topology shows that there will be an optimal performance path between a pair of NVMe SSDs and four possible V100 GPUs. Given this information, we can create a RAID0 device comprised of two NVMe SSDs on the same PCIe switch, and determine which GPUs are on the same PCIe upstream switch.

Starting at the top of the `lspci` output, note two NVMe drives at PCIe bus 0xda and 0xdb. The disk-by-path data indicates these are nvme8 and nvme9 devices. The four GPUs on the same segment, 0xe0, 0xe2, 0xe5 and 0xe7, are GPUs 12, 13, 14 and 15 respectively, as determined from the `nvidia-smi` output. The following table shows the PCIe GPU-to-NVMe affinity for all installed GPUs and corresponding NVMe SSD pairs.

### Table 1. DGX-2 GPU / NVMe Affinity (example)

<table>
<thead>
<tr>
<th>GPU #</th>
<th>GPU PCIe</th>
<th>NVMe #</th>
<th>NVMe PCIe</th>
</tr>
</thead>
<tbody>
<tr>
<td>0, 1, 2, 3</td>
<td>0x34, 0x36, 0x39, 0x3b</td>
<td>nvme2, nvme3</td>
<td>0x2e, 0x2f</td>
</tr>
<tr>
<td>4, 5, 6, 7</td>
<td>0x57, 0x59, 0x5c, 0x5e</td>
<td>nvme4, nvme5</td>
<td>0x51, 0x52</td>
</tr>
<tr>
<td>8, 9, 10, 11</td>
<td>0xb7, 0xb9, 0xbc, 0xbe</td>
<td>nvme6, nvme7</td>
<td>0xb1, 0xb2</td>
</tr>
<tr>
<td>12, 13, 14, 15</td>
<td>0xe0, 0xe2, 0xe5, 0xe7</td>
<td>nvme8, nvme9</td>
<td>0xda, 0xdb</td>
</tr>
</tbody>
</table>

With this information, we can configure a target workload for optimal throughput and latency, leveraging PCIe topology and device proximity of the GPUs and NVMe SSDs. This will be demonstrated in the next couple sections. Note that it is not guaranteed the actual PCIe BDF values will be the same for every NVIDIA DGX-2. This is because enumeration of the PCIe topology is based on specific configuration details and determined at boot time.

The same logic applies to storage that is network attached (NAS). The network interface (NIC) becomes the “storage controller”, in terms of the data flow between the GPUs and storage. Fortunately, determining PCIe topology is a much easier task for GPUs and NICs, as the `nvidia-smi` utility includes options for generating this information. Specifically, `nvidia-smi topo -mp` generates a simple topology map in the form of a matrix showing the connection(s) at the intersection of the installed GPUs and network interfaces.

For readability, the sample output below from a DGX-2 system shows the first eight columns, and the first four Mellanox device rows, not the entire table generated when executing `nvidia-smi topo -mp`.

```
dgx2> nvidia-smi topo -mp

<table>
<thead>
<tr>
<th>GPU0</th>
<th>GPU01</th>
<th>GPU02</th>
<th>GPU03</th>
<th>GPU04</th>
<th>GPU05</th>
<th>GPU06</th>
<th>GPU07</th>
</tr>
</thead>
<tbody>
<tr>
<td>GPU0</td>
<td>X</td>
<td>PIX</td>
<td>PXB</td>
<td>PXB</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
</tr>
<tr>
<td>GPU1</td>
<td>PIX</td>
<td>X</td>
<td>PXB</td>
<td>PXB</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
</tr>
<tr>
<td>GPU2</td>
<td>PXB</td>
<td>PXB</td>
<td>X</td>
<td>PIX</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
</tr>
<tr>
<td>GPU3</td>
<td>PXB</td>
<td>PXB</td>
<td>X</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
</tr>
<tr>
<td>GPU4</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>X</td>
<td>PIX</td>
<td>PXB</td>
</tr>
<tr>
<td>GPU5</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>PXB</td>
<td>PXB</td>
<td>X</td>
</tr>
<tr>
<td>GPU6</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>PXB</td>
<td>PXB</td>
<td>X</td>
</tr>
<tr>
<td>GPU7</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>NODE</td>
<td>PXB</td>
<td>PXB</td>
<td>X</td>
</tr>
<tr>
<td>GPU8</td>
<td>SYS</td>
<td>SYS</td>
<td>SYS</td>
<td>SYS</td>
<td>SYS</td>
<td>SYS</td>
<td>SYS</td>
</tr>
</tbody>
</table>
```
The optimal path between a GPU and NIC will be one PCIe switch path designated as PIX. The least optimal path is designated as SYS, which indicates that the data path requires traversing the CPU-to-CPU interconnect (NUMA nodes).

If you use this data when you configure and test GDS performance, the ideal setup would be, for example, a data flow from mlx5_0 to/from GPUs 0 and 1, mlx5_1 to/from GPUs 1 and 2, and so on.

### 3.2. GPUDirect Storage Configuration Parameters

There are various parameters and settings that will factor into delivered performance. In addition to storage/filesystem-specific parameters, there are system settings and GDS-specific parameters defined in `/etc/cufile.json`.

#### 3.2.1. System Parameters

On the system side, the following should be checked:

- **PCIe Access Control Service (ACS)**
  
  PCIe ACS is a security feature for peer-to-peer transactions. Each transaction is checked to determine whether peer-to-peer communication is allowed between the source and destination devices. Each such transaction must be routed through the root complex, which induces latency and impacts sustainable throughput. The best GDS performance is obtained when PCIe ACS is disabled.

- **IOMMU**
The PCIe Input/Output Memory Management Unit (IOMMU) is a facility for handling address translations for IO devices, and requires routing through the PCIe root complex. Optimal GDS performance is achieved when the IOMMU is disabled.

### 3.2.2. GPUDirect Storage Parameters

This section describes the JSON configuration parameters used by GDS.

When GDS is installed, the `/etc/cufile.json` parameter file is installed with default values. The implementation allows for generic GDS settings and parameters specific to a file system or storage partner.

Note: Consider `compat_mode` for systems or mounts that are not yet set up with GDS support.

#### Table 2. GPUDirect Storage cufile.json Variables

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Default Value</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>logging:dir</td>
<td>CWD</td>
<td>Location of the GDS log file.</td>
</tr>
<tr>
<td>logging:level</td>
<td>ERROR</td>
<td>Verbosity of logging.</td>
</tr>
<tr>
<td>profile:nvtx</td>
<td>false</td>
<td>Boolean which if set to true, generates NVTX traces for profiling.</td>
</tr>
<tr>
<td>profile:cufile_stats</td>
<td>0</td>
<td>Enable cuFile IO stats. Level 0 means no cuFile statistics.</td>
</tr>
<tr>
<td>profile:io_batchsize</td>
<td>128</td>
<td>Maximum size of the batch allowed.</td>
</tr>
<tr>
<td>properties:max_direct_io_size_kb</td>
<td>16384</td>
<td>Maximum IO chunk size (4K aligned) used by cuFile for each IO request (in KB).</td>
</tr>
<tr>
<td>properties:max_device_cache_size_kb</td>
<td>131072</td>
<td>Maximum device memory size (4K aligned) for reserving bounce buffers for the entire GPU (in KB).</td>
</tr>
<tr>
<td>properties:max_device_pinned_mem_size_kb</td>
<td>33554432</td>
<td>Maximum per-GPU memory size in KB, including the memory for the internal bounce buffers, that can be pinned.</td>
</tr>
<tr>
<td>properties:use_poll_mode</td>
<td>false</td>
<td>Boolean that indicates whether the cuFile library uses polling or synchronous wait for the storage to complete IO. Polling might be useful for small IO.</td>
</tr>
<tr>
<td>Parameter</td>
<td>Default Value</td>
<td>Description</td>
</tr>
<tr>
<td>---------------------------------------------</td>
<td>---------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>properties:poll_mode_max_size_kb</td>
<td>4</td>
<td>Maximum IO request size (4K aligned) in or equal to which library will be polled (in KB). Refer to Poll Mode below.</td>
</tr>
<tr>
<td>properties.force_compat_mode</td>
<td>false</td>
<td>If true, this option can be used to force all IO to use compatibility mode. Alternatively the admin can unload the nvidia_fs.ko or not expose the character devices in the docker container environment.</td>
</tr>
<tr>
<td>properties:allow_compat_mode</td>
<td>false</td>
<td>If true, enables the compatibility mode, which allows cuFile to issue POSIX read/write. To switch to GDS-enabled I/O, set this to false. Refer to Compatibility Mode below.</td>
</tr>
<tr>
<td>properties:rdma_dev_addr_list</td>
<td>[]</td>
<td>Provides the list of relevant client IPv4 addresses for all the interfaces that can be used for RDMA.</td>
</tr>
<tr>
<td>properties:rdma_load_balancing_policy</td>
<td>RoundRobin</td>
<td>Specifies the load balancing policy for RDMA memory registration. By default, this value is set to RoundRobin. Here are the valid values that can be used for this property:</td>
</tr>
<tr>
<td></td>
<td></td>
<td><strong>FirstFit</strong> - Suitable for cases where numGpus matches numPeers and GPU PCIe lane width is greater or equal to the peer PCIe lane width.</td>
</tr>
<tr>
<td></td>
<td></td>
<td><strong>MaxMinFit</strong> - This will try to assign peers in a manner that there is least sharing. Suitable for cases, where all GPUs are loaded uniformly.</td>
</tr>
<tr>
<td></td>
<td></td>
<td><strong>RoundRobin</strong> - This parameter uses only the NICs that are the closest to the GPU for memory registration in a round robin fashion.</td>
</tr>
<tr>
<td>Parameter</td>
<td>Default Value</td>
<td>Description</td>
</tr>
<tr>
<td>-------------------------------</td>
<td>---------------</td>
<td>-----------------------------------------------------------------------------</td>
</tr>
<tr>
<td>properties:rdma_dynamic_routing</td>
<td>false</td>
<td>Boolean parameter applicable only to Network Based File Systems. This could be enabled for platforms where GPUs and NICs do not share a common PCIe-root port.</td>
</tr>
<tr>
<td>properties:rdma_dynamic_routing_order</td>
<td>[ &quot;GPU_MEM_NVLINKS&quot;, &quot;GPU_MEM&quot;, &quot;SYS_MEM&quot;, &quot;P2P&quot; ]</td>
<td>The routing order applies only if rdma_dynamic_routing is enabled. Users can specify an ordered list of routing policies selected when routing an IO on a first-fit basis.</td>
</tr>
<tr>
<td>properties:io_batchsize</td>
<td>128</td>
<td>The max number of IO operations per batch.</td>
</tr>
<tr>
<td>properties:gds_rdma_write_support</td>
<td>true</td>
<td>Enable GDS write support for RDMA based storage.</td>
</tr>
<tr>
<td>properties:io_priority</td>
<td>default</td>
<td>Enable io priority w.r.t. compute streams. Valid options are &quot;default&quot;, &quot;low&quot;, &quot;med&quot;, &quot;high&quot;. Tuning this might be helpful in cases where cudaMemcpy is not performing as expected because of the GPU being consumed by the compute.</td>
</tr>
<tr>
<td>fs:generic:posix_unaligned_writes</td>
<td>false</td>
<td>Setting to true forces the use of a POSIX write instead of cuFileWrite for unaligned writes.</td>
</tr>
<tr>
<td>fs:lustre:posix_gds_min_kb</td>
<td>4KB</td>
<td>Applicable only for the EXAScaler filesystem. This is applicable for reads and writes. IO threshold for read/write (4K aligned) that is equal to or below the</td>
</tr>
<tr>
<td>Parameter</td>
<td>Default Value</td>
<td>Description</td>
</tr>
<tr>
<td>-------------------------</td>
<td>---------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>threshold that cufile will use for a POSIX read/write.</td>
<td>fs:lustre:rdma_dev_addr_list</td>
<td>Provides the list of relevant client IPv4 addresses for all the interfaces that can be used by a single lustre mount. This property is used by the cuFile dynamic routing feature to infer preferred RDMA devices.</td>
</tr>
<tr>
<td></td>
<td>fs:lustre:mount_table</td>
<td>Specifies a dictionary of IPv4 mount addresses against a Lustre mount point. This property is used by the cuFile dynamic routing feature. Refer to the default cufile.json for sample usage.</td>
</tr>
<tr>
<td></td>
<td>fs:nfs:rdma_dev_addr_list</td>
<td>Provides the list of IPv4 addresses for all the interfaces a single NFS mount can use. This property is used by the cuFile dynamic routing feature to infer preferred RDMA devices.</td>
</tr>
<tr>
<td></td>
<td>fs:nfs:mount_table</td>
<td>Specifies a dictionary of IPv4 mount addresses against a Lustre mount point. This property is used by the cuFile dynamic routing feature. Refer to the default cufile.json for sample usage.</td>
</tr>
<tr>
<td></td>
<td>fs:weka:rdma_write_support</td>
<td>If set to true, cuFileWrite will use RDMA writes instead of falling back to posix writes for a WekaFs mount.</td>
</tr>
<tr>
<td></td>
<td>fs:weka:&lt;rdma_dev_addr_list&gt;</td>
<td>Provides the list of relevant client IPv4 addresses for all the interfaces a single WekaFS mount can use. This property is also used by the cuFile dynamic routing feature to infer preferred rdma devices.</td>
</tr>
<tr>
<td></td>
<td>fs:weka:mount_table</td>
<td>Specifies a dictionary of IPv4 mount addresses against a WekaFS mount point.</td>
</tr>
<tr>
<td>Parameter</td>
<td>Default Value</td>
<td>Description</td>
</tr>
<tr>
<td>-------------------------------</td>
<td>---------------</td>
<td>-------------------------------------------------------------------------------</td>
</tr>
<tr>
<td></td>
<td></td>
<td>This property is used by the cuFile dynamic routing feature. Refer to the default cufile.json for sample usage.</td>
</tr>
<tr>
<td>denylist:drivers</td>
<td>[]</td>
<td>Administrative setting that disables supported storage drivers on the node.</td>
</tr>
<tr>
<td>denylist:devices</td>
<td>[]</td>
<td>Administrative setting that disables specific supported block devices on the node. Not applicable for DFS.</td>
</tr>
<tr>
<td>denylist:mounts</td>
<td>[]</td>
<td>Administrative setting that disables specific mounts in the supported GDS-enabled filesystems on the node.</td>
</tr>
<tr>
<td>denylist:filesystems</td>
<td>[]</td>
<td>Administrative setting that disables specific supported GDS-ready filesystems on the node.</td>
</tr>
<tr>
<td>miscellaneous:skip_topology_detection</td>
<td>false</td>
<td>Setting this to true will skip topology detection in compat mode. This will reduce the high startup latency seen in compat mode on systems with multiple PCI devices.</td>
</tr>
<tr>
<td>execution::max_io_queue_depth</td>
<td>128</td>
<td>This specifies the maximum number of pending work items that can be held by the cuFile library’s internal threadpool sub-system.</td>
</tr>
<tr>
<td>execution::max_io_threads</td>
<td>4</td>
<td>This specifies the number of threadpool threads that can process work items produced into a work queue corresponding to a single GPU on the system.</td>
</tr>
<tr>
<td>execution::parallel_io</td>
<td>true</td>
<td>Setting this to true will allow parallel processing of work items by enqueuing into the threadpool subsystem provided by the cuFile library.</td>
</tr>
<tr>
<td>execution::min_io_threshold_size_kb</td>
<td>8192</td>
<td>This option specifies the size in KB that the I/O work item submitted by the application would be split into, when</td>
</tr>
<tr>
<td>Parameter</td>
<td>Default Value</td>
<td>Description</td>
</tr>
<tr>
<td>-----------------------------------------------</td>
<td>---------------</td>
<td>---------------------------------------------------------------------------------------------------------------------------------------------</td>
</tr>
<tr>
<td>execution::max_request_parallelism</td>
<td>4</td>
<td>This number specifies the maximum number of parallel buffers available, that the original I/O work item buffer can be split into, when enqueuing into the threadpool sub-system.</td>
</tr>
</tbody>
</table>

Note: Workload/application-specific parameters can be set by using the environment variable that is set to point to an alternate cufile.json file, for example, CUFILE_ENV_PATH_JSON=/home/gds_user/my_cufile.json.

There are two mode types that you can set in the cufile.json configuration file:

- **Poll Mode**
  The cuFile API set includes an interface to put the driver in polling mode. Refer to cuFileDriverSetPollMode() in the cuFile API Reference Guide for more information. When the poll mode is set, a read or write issued that is less than or equal to properties:poll_mode_max_size_kb (4KB by default) will result in the library polling for IO completion, rather than blocking (sleep). For small IO size workloads, enabling poll mode may reduce latency.

- **Compatibility Mode**
  There are several possible scenarios where GDS might not be available or supported, for example, when the GDS software is not installed, the target file system is not GDS supported, O_DIRECT cannot be enabled on the target file, and so on. When you enable compatibility mode, and GDS is not functional for the IO target, the code that uses the cuFile APIs fall backs to the standard POSIX read/write path. To learn more about compatibility mode, refer to cuFile Compatibility Mode.

From a benchmarking and performance perspective, the default settings work very well across a variety of IO loads and use cases. We recommended that you use the default values for max_direct_io_size_kb, max_device_cache_size_kb, and max_device_pinned_mem_size_kb unless a storage provider has a specific recommendation, or analysis and testing show better performance after you change one or more of the defaults.

The cufile.json file has been designed to be extensible such that parameters can be set that are either generic and apply to all supported file systems (fs:generic), or file system specific (fs:lustre). The fs:generic:posix_unaligned_writes parameter enables the use of the POSIX write path when unaligned writes are encountered. Unaligned writes are generally sub-optimal, as they can require read-modify-write operations.
If the target workload generates unaligned writes, you might want to set `posix_unaligned_writes` to true, as the POSIX path for handling unaligned writes might be more performant, depending on the target filesystem and underlying storage. Also, in this case, the POSIX path will write to the page cache (system memory).

When the IO size is less than or equal to `posix_gds_min_kb`, the `fs:lustre:posix_gds_min_kb` setting invokes the POSIX read/write path rather than cuFile path. When using Lustre, for small IO sizes, the POSIX path can have better (lower) latency.

The GDS parameters are among several elements that factor into delivered storage IO performance. It is advisable to start with the defaults and only make changes based on recommendations from a storage vendor or based on empirical data obtained during testing and measurements of the target workload.

The following is the JSON schema:

```json
# /etc/cufile.json
{
  "logging": {
    // log directory, if not enabled will create log file
    // under current working directory
    // "dir": "/home/<xxxx>", // ERROR|WARN|INFO|DEBUG|TRACE (in decreasing order of priority)
    "level": "ERROR"
  },
  "profile": {
    // nvtx profiling on/off
    "nvtx": false,
    // cufile stats level(0-3)
    "cufile_stats": 0
  },
  "execution": {
    // max number of workitems in the queue;
    "max_io_queue_depth": 128,
    // max number of host threads per gpu to spawn for parallel IO
    "max_io_threads": 4,
    // enable support for parallel IO
    "parallel_io": true,
    // minimum IO threshold before splitting the IO
    "min_io_threshold_size_kb": 8192,
    // maximum parallelism for a single request
    "max_request_parallelism": 4
  },
  "properties": {
    // max IO size (4K aligned) issued by cuFile to nvidia-fs driver(in KB)
    "max_direct_io_size_kb": 16384,
    // device memory size (4K aligned) for reserving bounce buffers
    // for the entire GPU (in KB)
    "max_device_cache_size_kb": 131072,
    // limit on maximum memory (4K aligned) that can be pinned
    // for a given process (in KB)
    "max_device_pinned_mem_size_kb": 33554432,
    // true or false (true will enable asynchronous io submission to nvidia-fs driver)
    "use_poll_mode": false,
    // maximum IO request size (4K aligned) within or equal
    // to which library will poll (in KB)
    "poll_mode_max_size_kb": 4,
    // allow compat mode, this will enable use of cufile posix read/writes
```

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3.3. GPUDirect Storage Benchmarking Tools

There are several storage benchmarking tools and utilities for Linux systems, with varying degrees of features and functionality. The fio utility is one of the more popular and powerful tools that is used to generate storage IO loads and offers significant flexibility for tuning IO generation based on the desired IO load characteristics. For those familiar with fio on Linux systems, the use of gdsio will be very intuitive.

Since GDS is relatively new technology, with support dependencies and a specific set of libraries and APIs that fall outside standard POSIX IO APIs, none of the existing storage IO load generation utilities include GDS support. As a result, the installation of GDS includes the gdsio load generator which provides several command line options that enable generating various storage IO load characteristics via both the traditional CPU and the GDS data path.
3.3.1. **gdsio Utility**

The *gdsio* utility is similar to a number of disk/storage IO load generating tools. It supports a series of command line arguments to specify the target files, file sizes, IO sizes, number of IO threads, etc. Additionally, *gdsio* includes built-in support for using the traditional IO path (CPU), as well as the GDS path - storage to/from GPU memory.

Starting 12.2, the tool also supports three new memory (-m <2, 3, 4>) types to exercise the host memory support option using cuFile APIs. The new memory type option is only supported with certain transfer modes i.e. -x (0, 5, 6, 7). Additionally support for non O_DIRECT file descriptors is also introduced. It can be specified by the option -O 1. By default the *gdsio* utility works with O_DIRECT file descriptors which is represented by -O 0, although that need not be specified explicitly.

```
dgx2> ./gdsio --help

gdsio version :1.1
Usage [using config file]: gdsio rw-sample.gdsio
Usage [using cmd line options]: ./gdsio
  -f <file name>
  -D <directory name>
  -d <gpu_index (refer nvidia-smi)>
  -n <numa node>
  -m <memory type(0 - (cudaMalloc), 1 - (cuMem), 2 - (cudaMallocHost), 3 - (malloc) 4 - (mmap))>
  -w <number of threads for a job>
  -s <file size(K|M|G)>
  -o <start offset(K|M|G)>
  -i <io_size(K|M|G) > <min_size:max_size:step_size>
  -p <enable nvlinks>
  -b <skip bufregister>
  -o <start file offset>
  -V <verify IO>
  -x <xfer_type>
  -I <(read) 0|(write)1| (randread) 2| (randwrite) 3>
  -T <duration in seconds>
  -k <random_seed> (number e.g. 3456) to be used with random read/write>
  -U <use unaligned(4K) random offsets>
  -R <fill io buffer with random data>
  -F <refill io buffer with random data during each write>
  -B

xfer_type:
0 - Storage->GPU (GDS)
1 - Storage->CPU
2 - Storage->CPU->GPU
3 - Storage->CPU->GPU_ASYNC
4 - Storage->PAGE_CACHE->CPU->GPU
5 - Storage->GPU_ASYNC_STREAM
6 - Storage->GPU_BATCH
7 - Storage->GPU_BATCH_STREAM

Note:
read test (-I 0) with verify option (-V) should be used with files written (-I 1)
with -V option
read test (-I 2) with verify option (-V) should be used with files written (-I 3)
with -V option, using same random seed (-k),
same number of threads(-w), offset(-o), and data size(-s)
write test (-I 1/3) with verify option (-V) will perform writes followed by read

These *gdsio* options provide the necessary flexibility to construct IO tests based on a specific set of requirements, and/or simply to assess performance for several different
load types. Important to note that when using the \(-D\) flag to specify a target directory, \texttt{gdsio} must first execute write loads (\(-I\) 1 or \(-I\) 3) to create the files. The number of files created is based on the thread count (\(-w\) flag); 1 file is created for each thread. This is an alternative to using the \(-f\) flag where file pathnames are specified. The \(-D\) and \(-f\) flags cannot be used together.

The transfer types (\(-x\) flag) are further defined in the following table:

<table>
<thead>
<tr>
<th>x</th>
<th>Transfer Type</th>
<th>File Open O_DIRECT?</th>
<th>Host Memory Allocation Type</th>
<th>Device Memory Allocation Type</th>
<th>Copies</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>XFER_GPU_DIRECT</td>
<td>Yes</td>
<td>cudaMallocHost, malloc or mmap</td>
<td>cudaMalloc() / cuMemMap()</td>
<td>Zero copy</td>
</tr>
<tr>
<td>1</td>
<td>XFER_CPU_ONLY</td>
<td>Yes</td>
<td>Posix_mem_align (4k)</td>
<td>N/A</td>
<td>Zero copy</td>
</tr>
<tr>
<td>2</td>
<td>XFER_CPU_GPU</td>
<td>Yes</td>
<td>cudaMallocHost()</td>
<td>cudaMalloc() (use of multiple CUDA streams for cuMemcpyAsync())</td>
<td>One copy</td>
</tr>
<tr>
<td>3</td>
<td>XFER_CPU_ASYNC_GPU</td>
<td>Yes</td>
<td>cudaMallocManaged()</td>
<td>cudaMallocManaged() (use cuMemAdvise() to set hints for managed memory + cuMemcpyPrefetchAsync())</td>
<td>One copy (streaming buffer)</td>
</tr>
<tr>
<td>4</td>
<td>XFER_CPU_CACHED_GPU</td>
<td>No (use page cache)</td>
<td>cudaMallocHost()</td>
<td>cudaMalloc() (use of multiple CUDA streams for cuMemcpyAsync())</td>
<td>Two copies</td>
</tr>
<tr>
<td>5</td>
<td>XFER_GPU_DIRECT_ASYNC</td>
<td>Yes</td>
<td>cudaMallocHost, malloc or mmap</td>
<td>cudaMalloc() / cuMemMap()</td>
<td>Zero copy</td>
</tr>
<tr>
<td>6</td>
<td>XFER_GPU_BATCH</td>
<td>Yes</td>
<td>cudaMallocHost, malloc or mmap</td>
<td>cudaMalloc() / cuMemMap()</td>
<td>Zero copy</td>
</tr>
<tr>
<td>7</td>
<td>XFER_GPU_BATCH_STREAM</td>
<td>Yes</td>
<td>cudaMallocHost, malloc or mmap</td>
<td>cudaMalloc() / cuMemMap()</td>
<td>Zero copy</td>
</tr>
</tbody>
</table>

* Starting from GDS 1.7 release, if the poll mode is enabled through \texttt{cufile.json} then it will use poll mode. Otherwise, the \texttt{XFER_GPU_DIRECT_ASYNC} option would exercise stream-based async I/O mechanism.

Similar to the Linux \texttt{fio} storage load generator, \texttt{gdsio} supports the use of config files that contain the parameter values to use for a \texttt{gdsio} execution. This offers an alternative to lengthy command line strings, and the ability to build a collection of config files that can easily be reused for testing different configurations and workloads. The \texttt{gdsio} config file syntax supports global parameters, as well as individual job parameters. There are sample \texttt{gdsio} config files installed with GDS in /usr/local/cuda/gds/tools. The files
Benchmarking GPUDirect Storage

with the .gdsio extension are sample gdsio config files, and the README included in the same directory provides additional information on the command line and config file syntax for gdsio.

With these options and the support of parameter config files, it is a relatively simple process to run gdsio and assess performance using different data paths to/from GPU/CPU memory.

3.3.2. gds-stats Tool

The gds_stats tool is used to extract per-process statistics on the GDS IO. It can be used in conjunction with other generic tools (Linux iostat), and GPU-specific tools (nvidia-smi, the Data Center GPU Manager (DCGM) command line tool, dcgmi) to get a complete picture of data flow on the target system.

To use gds_stats, the profile:cufile_stats attribute in /etc/cufile.json must be set to 1, 2 or 3.

Note: The default value of 0 disables statistics collection.

The different levels provide an increasing amount of statistical data. When profile:cufile_stats is set to 3 (max level), the gds_stats utility provides a -l (level) CLI flag. Even when GDS is collecting level 3 stats, only level 1 or level 2 stats can be displayed.

In the example below, a gdsio job is started in the background, and level 3 gds_stats are extracted:

dgx2> gdsio -D /nvme23/gds_dir -d 2 -w 8 -s 1G -i 1M -x 0 -I 0 -T 300 &
[1] 850272
dgx2> gds_stats -p 850272 -l 3

 cuFile STATS VERSION : 3
  GLOBAL STATS:
  Total Files: 8
  Total Read Errors : 0
  Total Read Size (MiB): 78193
  Read BandWidth (GiB/s): 6.32129
  Avg Read Latency (us): 1044
  Total Write Errors : 0
  Total Write Size (MiB): 0
  Write BandWidth (GiB/s): 0
  Avg Write Latency (us): 0
  READ-WRITE SIZE HISTOGRAM :
  0-4(KiB): 0  0
  4-8(KiB): 0  0
  8-16(KiB): 0  0
  16-32(KiB): 0  0
  32-64(KiB): 0  0
  64-128(KiB): 0  0
  128-256(KiB): 0  0
  256-512(KiB): 0  0
  512-1024(KiB): 0  0
  1024-2048(KiB): 78193  0
  2048-4096(KiB): 0  0
  4096-8192(KiB): 0  0
  8192-16384(KiB): 0  0
  16384-32768(KiB): 0  0
  32768-65536(KiB): 0  0
  65536-... (KiB): 0  0
  PER_GPU STATS:
Here are the levels of `gds_stats` that are captured and displayed:

- **Level 3.**
  Shown above, includes (tarting at the top), a summary section, GLOBAL STATS, followed by a READ-WRITE SIZE HISTOGRAM section, PER_GPU STATS, PER_GPU POOL BUFFER STATS, PER_GPU POSIX POOL BUFFER STATS, PER_GPU RDMA STATS and RDMA MRSTATS.

- **Level 2**
  The GLOBAL STATS and READ-WRITE SIZE HISTOGRAM sections.

- **Level 1**
 GLOBAL STATS.

These are described as:

- **GLOBAL STATS** - Summary data including read/write throughput and latency.

- **READ-WRITE SIZE HISTOGRAM** - Distribution of the size of read and write IOs.

- **PER_GPU STATS** - Various statistics for each GPU, including read and write throughput, counters for sparse IOs, POSIX IOs, errors, unaligned IOs and data on registered buffers.

The next two stats provide information on the buffer pool used for bounce buffers for both GDS IO and POSIX IO. These pools use fixed size 1MB buffers in a 128MB pool (See
“max_device_cache_size_kb”: 131072 in the /etc/cufile.json parameters. This pool is used when buffers are not registered, unaligned buffer or file offsets, and when the storage and GPU cross NUMA nodes (typically CPU sockets).

- **PER_GPU POOL BUFFER STATS** - Bounce buffer stats when GDS is in use.
- **PER_GPU POSIX POOL BUFFER STATS** - System memory bounce buffer stats when compat mode (POSIX IO) is used.

These last two stats provide data related to RDMA traffic when GDS is configured with Network Attached Storage (NAS).

- **PER_GPU RDMA STATS** - RDMA traffic.
- **PER_GPU RDMA MRSTATS** - RDMA memory registration data.

The gds_stats are very useful for understanding important aspects of the IO load. Not just performance (BandWidth and Latency), but also the IO size distribution for understanding an important attribute of the workload, and PER_GPU STATS enable a view into which GPUs are reading/writing data to/from the storage.

There are various methods that you can use to monitor gds_stats data at regular intervals, such as shell wrappers that define intervals and extract the data of interest. Additionally, the Linux watch command can be used to monitor gds_stats data at regular intervals:

```
Every 1.0s: gds_stats -p 951816 | grep 'BandWidth\|Latency'
```

```
psg-dgx2-g02: Fri Nov 20 13:16:36 2020
Read BandWidth (GiB/s): 6.38327
Avg Read Latency (us): 1261
Write BandWidth (GiB/s): 0
Avg Write Latency (us): 0
```

In the above example, gds_stats was started using the Linux watch command:

```
watch -n 1 "gds_stats -p 31470 | grep ‘BandWidth\|Latency’"
```

This command results in the bandwidth and latency stats being updated in your Command Prompt window every second.
Chapter 4. GPUDirect Storage Benchmarking on Direct Attached Storage

This section covers benchmarking GDS on storage directly attached to the server, typically in the form of NVMe SSD devices on the PCIe bus. The specific examples on DGX-2 and DGX A100 can be used as guidelines for any server configuration. Note that in the following examples, the output of various command line tools and utilities is included. In some cases, rows or columns are deleted to improve readability and clarity.

4.1. GPUDirect Storage Performance on DGX-2 System

Currently, GDS supports NVMe devices as direct attached storage, where NVMe SSDs are plugged directly into the PCIe bus. The DGX-2 system comes configured with up to 16 of these devices that are typically configured as a large RAID metadevice. As per the previous section, the DGX-2 system used to execute these examples was very specifically configured, such that pairs of NVMe SSDs on the same PCIe switch are in a RAID0 group, and the gdsio command line intentionally selects GPUs that share the same upstream PCIe switch.

A Simple Example: Writing to large files with a large IO size using the GDS path.

This example uses a RAID0 device configured with nvme2 and nvme3 with an ext4 file system (mounted as /nvme23, with a gds_dir subdirectory to hold the generated files).

dgx2> gdsio -D /nvme23/gds_dir -d 2 -w 8 -s 500M -i 1M -x 0 -I 0 -T 120

IoType: READ XferType: GPUD Threads: 8 DataSetSize: 818796544/4096000(KiB) IOSize: 1024(KiB) Throughput: 6.524658 GiB/sec, Avg_Latency: 1197.370995 usecs ops: 799606 total_time 119.679102 secs

Here is some additional information about the options in the example:

- `-D /nvme23/gds_dir`, the target directory.
- `-d 2`, selects GPU # 2 for data target/destination.
- `-w 8`, 8 workers (8 IO threads)
- `-s 500M`, the target file size.
- **i** 1M, IO size (important for assessing throughput).
- **-x** 0, the IO data path, in this case GDS.

See Table 2 in *gdsio Utility* for more information.

- **-I** 0, writes the IO load (0 is for reads, 1 is for writes)
- **-T** 120, runs for 120 seconds.

The results generated by gdsio show expected performance, given that the storage IO target is a RAID 0 configuration of the two NVMe SSDs, where each SSD is configured with around 3.4GB/sec large read performance. The average sustained throughput was 6.5GB/sec, with a 1.2ms average latency. We can look at system data during the gdsio execution for additional data points on data rates and movement. This is often useful for validating results reported by load generators, as well as ensuring the data path is as expected. Using the Linux *iostat* utility (*iostat -cxzk 1*):

```
avg-cpu:  %user  %nice  %system %iowait  %steal  %idle
0.03    0.00    0.42    7.87    0.00   91.68
```

<table>
<thead>
<tr>
<th>Device</th>
<th>r/s</th>
<th>rkB/s</th>
<th>r_await</th>
<th>rareq-sz</th>
<th>w/s</th>
<th>wkB/s</th>
<th>. . .</th>
<th>%util</th>
</tr>
</thead>
<tbody>
<tr>
<td>md127</td>
<td>54360.00</td>
<td>6958080.00</td>
<td>0.00</td>
<td>128.00</td>
<td>0.00</td>
<td>0.00</td>
<td>. . .</td>
<td>0.00</td>
</tr>
<tr>
<td>nvme2n1</td>
<td>27173.00</td>
<td>3478144.00</td>
<td>1.03</td>
<td>128.00</td>
<td>0.00</td>
<td>0.00</td>
<td>. . .</td>
<td>100.00</td>
</tr>
<tr>
<td>nvme3n1</td>
<td>27179.00</td>
<td>3478912.00</td>
<td>0.95</td>
<td>128.00</td>
<td>0.00</td>
<td>0.00</td>
<td>. . .</td>
<td>100.00</td>
</tr>
</tbody>
</table>

Also, data from the *nvidia-smi dmon* command:

```
dgx2> nvidia-smi dmon -i 2 -s putcm
```

<table>
<thead>
<tr>
<th># gpu</th>
<th>pwr</th>
<th>gtemp</th>
<th>mtemp</th>
<th>sm</th>
<th>mem</th>
<th>enc</th>
<th>dec</th>
<th>rxpci</th>
<th>txpci</th>
<th>mclk</th>
<th>pclk</th>
<th>fb</th>
<th>barl</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>63</td>
<td>37</td>
<td>37</td>
<td>0</td>
<td>4</td>
<td>0</td>
<td>0</td>
<td>8923</td>
<td>0</td>
<td>958</td>
<td>345</td>
<td>326</td>
<td>15</td>
</tr>
<tr>
<td>2</td>
<td>63</td>
<td>37</td>
<td>37</td>
<td>0</td>
<td>4</td>
<td>0</td>
<td>0</td>
<td>8922</td>
<td>0</td>
<td>958</td>
<td>345</td>
<td>326</td>
<td>15</td>
</tr>
<tr>
<td>2</td>
<td>63</td>
<td>37</td>
<td>37</td>
<td>0</td>
<td>4</td>
<td>0</td>
<td>0</td>
<td>8930</td>
<td>0</td>
<td>958</td>
<td>345</td>
<td>326</td>
<td>15</td>
</tr>
<tr>
<td>2</td>
<td>63</td>
<td>37</td>
<td>37</td>
<td>0</td>
<td>4</td>
<td>0</td>
<td>0</td>
<td>8764</td>
<td>0</td>
<td>958</td>
<td>345</td>
<td>326</td>
<td>15</td>
</tr>
</tbody>
</table>

This data is consistent with the results reported by gdsio. The iostat data shows just over 3.4GB/sec from each of the two NVMe drives, and close to 1ms latency per device. Note each drive sustained about 27k writes-per-second (IOPS). The second data set from the dmon subcommand of nvidia-smi, note the rxpci column. Recall our gdsio command line initiated GPUDirect Storage reads, so reads from the storage to the GPU. We see the selected GPU, 2, receiving over 8GB/sec over PCIe. This is GPUDirect Storage in action - the GPU reading (PCIe receive) directly from the NVMe drives over PCIe.

While the preceding information is important to enable an optimal configuration, the GDS software will always attempt to maintain an optimal data path, in some cases via another GPU that has better affinity to the storage targets. By monitoring PCIe traffic with either nvidia-smi or dcgmi (the command line component of DCGM), we can observe data rates in and out of the GPUs.

Using a previous example, running on the same RAID0 metadevice comprised of two NVMe drives on the same PCIe switch, but specifying GPU 12 this time, and capturing GPU PCIe traffic with dcgmi, we'll observe sub-optimal performance as GPU 12 is not on the same downstream PCIe switch as our two NVMe drives.

```
dgx2> gdsio -D /nvme23/gds_dir -d 12 -w 8 -s 500M -i 1M -x 0 -I 0 -T 120
```

IoType: READ XferType: GPUD Threads: 8 DataSetSize: 491438080/4096000(KiB) IOSize: 1024(KiB) Throughput: 3.893747 GiB/sec, Avg_Latency: 2003.091575 usecs ops: 479920 total_time 120.365276 secs

Note throughput dropped from 6.5GB/sec to 3.9GB/sec, and latency almost doubled to 2ms. The PCIe traffic data tells an interesting story:
Note we observe PCIe traffic on GPU 12, but also traffic on GPU 0. This is GDS in action once again. The cuFile library will select a GPU for the data buffers (GPU memory) that is on the same PCIe switch as the storage. In this case, GPU 0 was selected by the library, as it is the first of four GPUs on the same PCIe switch as the NVMe devices. The data is then moved to the target GPU (12).

The net effect of a sub-optimal device selection is an overall decrease in throughput, and increase in latency. With GPU 2, the average throughput was 6.5GB/sec, average latency 1ms. With GPU 12, the average throughput was 3.9GB/sec, average latency 2ms. Thus we observe a 40% decrease in throughput and a 2X increase in latency when a non-optimal configuration is used.

Not all workloads are about throughput. Smaller IO sizes and random IO patterns are an attribute of many production workloads, and assessing IOPS (IO Operations Per Second) performance is a necessary component to the storage benchmarking process.

A critical component to determining what peak performance levels can be achieved is ensuring there is sufficient load. Specifically, for storage benchmarking, the number of processes/threads generating IO is critical to determining maximum performance.

The gdsio tool provides for specifying random reads or random writes (-I flag). In the examples below, once again we're showing an optimal combination of GPU (0) and NVMe devices, generating a small (4k) random read low with an increasing number of threads (-w).

| IoType: RANDREAD XferType: GPUD Threads: 4 DatasetSize: 11736528/4194304(KiB) IOSize: 4(KiB) Throughput: 0.093338 GiB/sec, Avg_Latency: 163.478958 usecs ops: 2934132 total_time 119.917332 secs |
| IoType: RANDREAD XferType: GPUD Threads: 8 DatasetSize: 2345880/8388608(KiB) IOSize: 4(KiB) Throughput: 0.187890 GiB/sec, Avg_Latency: 162.422553 usecs ops: 5863720 total_time 119.049917 secs |
| IoType: RANDREAD XferType: GPUD Threads: 16 DatasetSize: 48209436/16777216(KiB) IOSize: 4(KiB) Throughput: 0.385008 GiB/sec, Avg_Latency: 158.918796 usecs ops: 12052359 total_time 119.415992 secs |
| IoType: RANDREAD XferType: GPUD Threads: 32 DatasetSize: 96418872/33554432(KiB) IOSize: 4(KiB) Throughput: 0.770016 GiB/sec, Avg_Latency: 155.101790 usecs ops: 24104728 total_time 120.010070 secs |
| IoType: RANDREAD XferType: GPUD Threads: 64 DatasetSize: 192837744/67108864(KiB) IOSize: 4(KiB) Throughput: 1.540032 GiB/sec, Avg_Latency: 152.703896 usecs ops: 48209436 total_time 120.020170 secs |
We can compute the IOPS by dividing the ops value by the total time. Note in all cases the total time was just over 119 seconds (120 seconds was specified as the run duration on the command line).

<table>
<thead>
<tr>
<th>Thread Count (-w)</th>
<th>IOPS (ops / total_time)</th>
</tr>
</thead>
<tbody>
<tr>
<td>4</td>
<td>24,468</td>
</tr>
<tr>
<td>8</td>
<td>49,255</td>
</tr>
<tr>
<td>16</td>
<td>100,928</td>
</tr>
<tr>
<td>32</td>
<td>238,245</td>
</tr>
<tr>
<td>64</td>
<td>484,612</td>
</tr>
<tr>
<td>128</td>
<td>850,240</td>
</tr>
</tbody>
</table>

It is interesting to observe the average latency on each run (Avg_Latency) actually gets better as the number of threads and IOPS increases, with 134.5us average latency at 484,612 IOPS running 64 threads. Increasing the thread count to 128, we observe a slight uptick in latency to 151.51us while sustaining 850,240 random reads per second. Tracking latency with throughput (or, in this case, IOPS) is important in characterizing delivered performance. In this example, the specification for the NVMe drives that make up the RAID0 device indicates a random read capability of about 800k IOPS per drive. Thus, even with 128 threads generating load, the latency is excellent as the load is well within drive specifications, as each of the two drives in the RAID0 device sustained about 425,000 IOPS. This was observed with the `iostat` utility:

```
avg-cpu:  %user   %nice %system %iowait  %steal   %idle
        16.03    0.00    6.52   76.97    0.00    0.48

Device            r/s     rkB/s   rrqm/s  %rrqm r_await rareq-sz     . . .  %util
md127         856792.00 3427172.00     0.00   0.00    0.00     4.00  . . .   0.00
nvme2n1       425054.00 1700216.00     0.00   0.00    0.13     4.00  . . .   100.80
nvme3n1       431769.00 1727080.00     0.00   0.00    0.13     4.00  . . .   100.00
```

We observe the row showing the RAID0 metadevice, md127, displaying total reads per second (r/s) reflects the sum of the two underlying NVMe drives.

Extending this example to demonstrate delivered performance when a GPU target is specified that is not part of the same PCIe segment:

```
dgx2> gdsio -D /nvme23/gds_dir -d 10 -w 64 -s 1G -i 4K -x 0 -I 2 -k 0308 -T 120
IoType: RANDREAD XferType: GPUD Threads: 64 DatasetSize: 13268776/67108864(KiB) IOSize: 4(KiB) Throughput: 0.105713 GiB/sec, Avg_Latency: 2301.201214 usecs ops: 3317194 total_time 119.702494 secs
```

In this example we specified GPU 10 as the data read target. Note the dramatic difference in performance. With 64 threads generating random reads, latency went from 151.51us to 2.3ms, and IOPS dropped from 850k IOPS to about 28k IOPS. This is due to the overhead of GDS using a GPU on the same PCIe segment for the primary read buffer, then moving that data to the specified GPU. Again, this can be observed when monitoring GPU PCIe traffic:

```
dgx2> dcmi dmon -e 1009,1010 -d 1000
# Entity   PCITX  PCIRX
Id         
GPU 0   108216883 122481373
GPU 1   185690   61385
GPU 9   183268   60918
```

NVIDIA Magnum IO GPUDirect Storage
We observe PCIe traffic on both GPU 10, which was specified in the gdsio command line, and GPU 0, which was selected by GDS as the primary read buffer due to its proximity to the NVMe devices. Using gds_stats, we can see the buffer allocation on GPU 0:

```
dgx2> gds_stats -p 1545037 -l 3
```

**GLOBAL STATS:**
- Total Files: 64
- Total Read Errors: 0
- Total Read Size (MiB): 4996
- Read BandWidth (GiB/s): 0.126041
- Avg Read Latency (us): 2036
- Total Write Errors: 0
- Total Write Size (MiB): 0
- Write BandWidth (GiB/s): 0
- Avg Write Latency (us): 0

**READ-WRITE SIZE HISTOGRAM:**
- 0-4(KiB): 0 0
- 4-8(KiB): 1279109 0
- 8-16(KiB): 0 0
- ... 65536-...(KiB): 0 0

**PER_GPU STATS:**
- GPU 0 Read: bw=0 util(%)=0 n=0 posix=0 unalign=0 r_sparse=0 r_inline=0 err=0 MiB=0
- Write: bw=0 util(%)=0 n=0 posix=0 unalign=0 err=0 MiB=0 BufRegister: n=0 err=0 free=0 MiB=0
- GPU 9 Read: bw=0 util(%)=0 n=0 posix=0 unalign=0 r_sparse=0 r_inline=0 err=0 MiB=0
- Write: bw=0 util(%)=0 n=0 posix=0 unalign=0 err=0 MiB=0 BufRegister: n=0 err=0 free=0 MiB=0
- GPU 10 Read: bw=0.124332 util(%)=6387 n=1279109 posix=0 unalign=0 r_sparse=0 r_inline=0 err=0 MiB=4996
- Write: bw=0 util(%)=0 n=0 posix=0 unalign=0 err=0 MiB=0 BufRegister: n=64 err=0 free=0 MiB=0
- GPU 11 Read: bw=0 util(%)=0 n=0 posix=0 unalign=0 r_sparse=0 r_inline=0 err=0 MiB=0
- Write: bw=0 util(%)=0 n=0 posix=0 unalign=0 err=0 MiB=0 BufRegister: n=0 err=0 free=0 MiB=0

**PER_GPU POOL BUFFER STATS:**
- GPU 0 pool_size_MiB : 64 usage : 63/64 used_MiB : 63

The output from gds_stats shows Read activity on GPU 10 (specified on the gdsio command line), and POOL BUFFER activity on GPU 0, with 63 of 64 1MB buffers in use. Recall GDS selected GPU 0 because it’s the first GPU on the same PCIe segment as the NVMe drives. This illustrates one of the uses of the GPU POOL BUFFER (see section on gds_stats).

There are two key points to consider based on these results. First, for small, random IO loads, a large number of threads generating load are necessary to assess peak performance capability. Second, for small, random IO loads, the performance penalty of a sub-optimal configuration is much more severe than was observed with large throughput-oriented IO loads.
GPUDirect Storage Benchmarking on Direct Attached Storage

4.2. GPUDirect Storage Performance on a DGX A100 System

GDS is also supported on DGX A100 system, the world's first 5 peta FLOPS AI system built with a new generation of GPUs, NVMe drives and network interfaces. Please refer to the DGX A100 product page for details. In this section, we will use the same test methodology we used on the DGX-2 example to benchmark GDS performance on a DGX A100 system.

First, we map out the GPU and NMVe drive affinity:

- Checking the NVMe drive name and PCIe BFD values:

```bash
dgxuser@dgxa100:~$ ls -l /dev/disk/by-path/
total 0
lrwxrwxrwx 1 root root 13 Oct 26 10:52 pci-0000:08:00.0-nvme-1 -> ../../nvme0n1
lrwxrwxrwx 1 root root 13 Oct 26 10:52 pci-0000:09:00.0-nvme-1 -> ../../nvme1n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:22:00.0-nvme-1 -> ../../nvme2n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:22:00.0-nvme-1-part1 -> ../../nvme2n1p1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:22:00.0-nvme-1-part2 -> ../../nvme2n1p2
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:23:00.0-nvme-1 -> ../../nvme3n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:23:00.0-nvme-1-part1 -> ../../nvme3n1p1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:23:00.0-nvme-1-part2 -> ../../nvme3n1p2
lrwxrwxrwx 1 root root 9 Oct 26 10:51 pci-0000:25:00.3-usb-0:1.1:1.0-scsi-0:0:0:0 -> ../../sr0
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:52:00.0-nvme-1 -> ../../nvme4n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:53:00.0-nvme-1 -> ../../nvme5n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:89:00.0-nvme-1 -> ../../nvme6n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:8a:00.0-nvme-1 -> ../../nvme7n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:c8:00.0-nvme-1 -> ../../nvme8n1
lrwxrwxrwx 1 root root 10 Oct 26 10:51 pci-0000:c9:00.0-nvme-1 -> ../../nvme9n1
```

- Checking the GPU index numbering correlated to the PCIe BFD:

```bash
dgxuser@gpu01:~$ nvidia-smi --query-gpu=index,name,pci.domain,pci.bus, --format=csv
index, name, pci.domain, pci.bus
0, A100-SXM4-40GB, 0x0000, 0x07
1, A100-SXM4-40GB, 0x0000, 0x0F
2, A100-SXM4-40GB, 0x0000, 0x47
3, A100-SXM4-40GB, 0x0000, 0x4E
4, A100-SXM4-40GB, 0x0000, 0x87
5, A100-SXM4-40GB, 0x0000, 0x90
6, A100-SXM4-40GB, 0x0000, 0xB7
7, A100-SXM4-40GB, 0x0000, 0xB0
```

- Checking the NVMe drive and GPU PCIe slot relationship:

```bash
dgxuser@dgxa100:~$ lspci -tv | egrep -i "nvidia|NVMe"
|           +-01.1-[b1-cb]----00.0-[b2-cb]---00.0-[b3-b7]----00.0-[b4-b7]----00.0-[b5-b7]----00.0-[b6-b7]----00.0-[b7]----00.0-NVIDIA Corporation
|               Device 20b0
| |       +01.1-[b1-cb]----00.0-[b2-cb]---00.0-[b3-b7]----00.0-[b4-b7]----00.0-[b5-b7]----00.0-[b6-b7]----00.0-[b7]----00.0-NVIDIA Corporation
| |           Device 20b0
| | |       +08.0-[be-ca]----00.0-[bf-ca]--
| | |       +00.0-[c0-c7]----00.0-[c1-c7]----00.0-[c2]----00.0-NVIDIA Corporation Device
| | | |       laf1
```
Then mapping the NVMe and GPU affinities:

Table 4. Mapping the NVMe and GPU Affinity

<table>
<thead>
<tr>
<th>GPU #</th>
<th>GPU PCIe Bus #</th>
<th>NVMe #</th>
<th>NVMe PCIe Bus #</th>
</tr>
</thead>
<tbody>
<tr>
<td>0, 1</td>
<td>0x07, 0x0F</td>
<td>nvme0, nvme1</td>
<td>0x08, 0x09</td>
</tr>
<tr>
<td>2, 3</td>
<td>0x47, 0x4e</td>
<td>nvme4, nvme5</td>
<td>0x52, 0x53</td>
</tr>
<tr>
<td>4, 5</td>
<td>0x87, 0x90</td>
<td>nvme6, nvme7</td>
<td>0x89, 0x8a</td>
</tr>
<tr>
<td>6, 7</td>
<td>0xb7, 0xbd</td>
<td>nvme8, nvme9</td>
<td>0xc8, 0xc9</td>
</tr>
</tbody>
</table>
By default, all NVMe drives in DGX systems are configured as a RAID0 storage array /dev/md1, but they are connected with different PCIe switches, as shown below.

Figure 2. "DGX-A100 Block Diagram"

This configuration might lead to suboptimal performance when data is moved across PCIe switches to reach the devices configured in /dev/md1. We can reconfigure the raid array and re-group the two NVMe drives attached to the same PCIe switches in a single RAID0 device, and create a testing workload between this RAID0 device and affiliated GPUs attached to the same PCIe switch. The following table provides information about the new mapping of raid array and GPU affinity:

Table 5. New Mapping of the RAID Array and GPU Affinity

<table>
<thead>
<tr>
<th>Raid Array</th>
<th>GPU #</th>
<th>GPU PCIe</th>
<th>NVMe #</th>
<th>NVMe PCIe</th>
</tr>
</thead>
<tbody>
<tr>
<td>/raid1 (/dev/md1)</td>
<td>0, 1</td>
<td>0x07, 0x0F</td>
<td>nvme0, nvme1</td>
<td>0x08, 0x09</td>
</tr>
<tr>
<td>/raid2 (/dev/md2)</td>
<td>2, 3</td>
<td>0x47, 0x4E</td>
<td>nvme4, nvme5</td>
<td>0x52, 0x53</td>
</tr>
<tr>
<td>/raid3 (/dev/md3)</td>
<td>4, 5</td>
<td>0x87, 0x90</td>
<td>nvme6, nvme7</td>
<td>0x89, 0x8A</td>
</tr>
<tr>
<td>/raid4 (/dev/md4)</td>
<td>6, 7</td>
<td>0xB7, 0xbd</td>
<td>nvme8, nvme9</td>
<td>0xc8, 0xc9</td>
</tr>
</tbody>
</table>
Now we can design our tests based on the above mapping. The GPU and raid array with NVMe drives in the same row has the affinity and will produce the optimized results.

Example write operation, NVMe raid 0 device and GPUs are affiliated with the same PCIe switch:

dgxuser@dgxa100:~$ /usr/local/gds/tools/gdsio -D /raid4/gds-io -d 7 -w 8 -s 500M -i 1M -x 0 -I 1 -T 120

IoType: WRITE XferType: GPUD Threads: 8 DataSetSize: 901008384/4096000(KiB) IOSize: 1024(KiB) Throughput: 7.181430 GiB/sec, Avg_Latency: 1087.870030 usecs ops: 879891

Observe the throughput of GDS write operation is 7.18GB/s on /raid4, which is close to the performance limits of 2 NVMe SSDs configured in that /raid4 device. This can be verified with other commands which check GPUs' and NVMes' I/O statistics:

Checking GPU PCIe TX/RX IO from DCGM: this shows data is coming out of GPU7 at 7.6GB/s

dgxuser@dgxa100:~$ dcmi dmon -e 1009,1010 -d 1000

# Entity                  PCITX         PCIRX
Id                    GPU 0  283298  672747
                      GPU 1  286415  701223
                      GPU 2  362656  778953
                      GPU 3  386259  817762
                      GPU 4   305924  787709
                      GPU 5  316632  826865
                      GPU 6  332589  831274
                      GPU 7  816730609 472336198
                      GPU 0  148856  65076
                      GPU 1  149444  66021
                      GPU 2  149408  65493
                      GPU 3  148862  65479
                      GPU 4  147445  65019
                      GPU 5  145363  62444
                      GPU 6  148393  65497
                      GPU 7  8367830609 492336198

Checking the GPU PCIe TX/RX IO from nvidia-smi utility, this shows average 7.67GB/s from all the columns captured here. These are almost identical results from the nvidia-smi and dcmi dmon commands.

dgxuser@dgxa100:~$ nvidia-smi dmon -i 7 -s putcm

# gpu # Idx     pwr     gtemp     mtemp     sm    mem     enc    dec    rxpci     txpci    mclk    pclk    fb    bar1
       W     C      C     %     %     %     %     %     %     MB/s    MB/s   MHz    MHz    MB    MB
 7   66   37    36     0   0    0    0   255  8235  1215  1095   430  14
 7   66   37    36     0   0    0    0   298  7079  1215  1095   430  14
 7   66   37    35     0   0    0    0   323  8444  1215  1095   430  14
 7   66   37    36     0   0    0    0   325  8760  1215  1095   430  14
 7   66   37    36     0   0    0    0   308  7482  1215  1095   430  14
 7   66   37    36     0   0    0    0   274  8774  1215  1095   430  14
 7   66   37    36     0   0    0    0   326  8721  1215  1095   430  14
 7   66   37    35     0   0    0    0   332  8802  1215  1095   430  14
 7   66   37    35     0   0    0    0   336  8160  1215  1095   430  14

Also check the NVMe IO statistics to verify /md4 raid array (includes nvme8n1 and nvme9n1) with the iostat -cxzm 1 command. This command computes the NVMe IO statistics every second so the number may fluctuate somewhat, but the 7.46GB/s reading here matches GDS and GPU readings approximately.

dgxuser@dgxa100:~$ iostat -cxzm 1

Linux 5.4.0-52-generic (dgxa100) 10/25/20  _x86_64_ (256 CPU)
GPUDirect Storage Benchmarking on Direct Attached Storage

NVIDIA Magnum IO GPUDirect Storage

avg-cpu: %user %nice %system %iowait %steal %idle
0.12 0.00 0.32 0.04 0.00 99.51

<table>
<thead>
<tr>
<th>Device</th>
<th>r/s</th>
<th>w/s</th>
<th>rMB/s</th>
<th>wMB/s</th>
<th>rrrm/s</th>
<th>wrqm/s</th>
<th>%rrqm</th>
<th>%wrqm</th>
</tr>
</thead>
<tbody>
<tr>
<td>loo0</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
</tr>
<tr>
<td>nvme0n1</td>
<td>1.04</td>
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<td>104.33</td>
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<td>0.00</td>
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</tr>
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<td>0.03</td>
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<td>29.58</td>
<td>0.00</td>
<td>0.00</td>
</tr>
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<td>0.00</td>
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<td>6.34</td>
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<td>0.00</td>
<td>6.26</td>
<td>0.14</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
</tr>
<tr>
<td>nvme6n1</td>
<td>0.15</td>
<td>0.00</td>
<td>6.26</td>
<td>0.35</td>
<td>0.00</td>
<td>109.22</td>
<td>0.00</td>
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<tr>
<td>nvme7n1</td>
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<td>0.00</td>
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<td>0.00</td>
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<tr>
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<td>0.00</td>
<td>0.00</td>
<td>6.34</td>
<td>0.16</td>
<td>0.00</td>
<td>109.22</td>
<td>0.00</td>
<td>0.00</td>
</tr>
<tr>
<td>md1</td>
<td>0.36</td>
<td>0.90</td>
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<tr>
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<td>0.00</td>
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<td>6.51</td>
<td>0.00</td>
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<td>51.70</td>
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<td>0.00</td>
<td>0.00</td>
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<td>0.00</td>
</tr>
</tbody>
</table>

Test result shows 7.116GB/s throughput, which is lower than the previous 7.181GB/s with NVMe and GPU affinity. The average latency 1096.57 us is also higher than the previous case. The reason is GDS in this case will use the closest GPU (GPU 6 in this example), to the target as an intermediate device to move the data, thus introducing some performance penalty. This behavior can be observed.

Checking GPU PCIe TX/RX IO from “nvidia-smi”: this shows data is moving from GPU 0 -> GPU 6 -> target NVMe drives:

dgxuser@dgxa100:~/gds$ nvidia-smi dmon -i 0 -s putcm
# gpu   pwr gtemp mtemp  sm  mem enc dec rxpci txpci mclk pclk   fb   bar1
# Idx     W    C        C      %     %        %     %  MB/s  MB/s   MHz   MHz   MB
MB 0 65 32 32 39 0 0 0 3 7950 1215 1095 430 14
0 65 32 31 34 0 0 0 3 9213 1215 1095 430 14
0 65 32 32 32 0 0 0 2 9174 1215 1095 430 14

NVIDIA Magnum IO GPUDirect Storage
The iostat command also shows IO performance on /md4 (includes nvme8n1 and nvme9n1) is 7302.01MB/s, or 7.13GB/s, very close to 7.11GB/s reported by gdsio.

It is expected that the examples, from NVIDIA DGX-2 and DGX A100 systems, provide sufficient detail such that the commands and methodology used to achieve maximum performance can be applied to any server system with NVMe devices and NVIDIA GPUs that support GDS.
NAS configurations bring the network element into the storage equation, which of course must be factored in when assessing performance. Throughput, IOPS, and latency are all dependent on the network configuration, such as the number of configured interfaces, interface speed, as well as the backend storage configuration. For example, if the storage backend is NVMe SSDs, 4 such devices doing streaming reads will saturate a single 100GB network. The entire configuration end-to-end needs to be examined and sized appropriately to ensure any potential hardware bottlenecks are identified and resolved.

5.1. GPUDirect Storage Benchmarking on NFS

The Network File System, NFS, was invented by Sun Microsystems in the early 1980s, and became one of the earliest network storage solutions broadly deployed in production environments. An NFS server with direct attached storage can export that storage over the network, making exported file systems available to any number of NFS clients. NFS offers ease of use and administration, a mature code base (it’s been around for decades) which makes it relatively robust. Early implementations relied on UDP or TCP as the protocol for transferring data over the network with NFS. With Remote Direct Memory Access (RDMA) capability, network overhead is significantly reduced, enabling higher throughput and lower latency read/write operations between NFS clients and the server.

With GPUDirect Storage configured for NFS, the transport protocol is RDMA, leveraging the high-speed, low-latency data flow between the client(s) and server. The RDMA operations move data over the network via the IB interfaces, then data is moved to/from the IB cards to the GPUs via DMA operations, bypassing the CPU and system memory.

As discussed previously, the PCIe topology is a factor in determining the optimal configuration on the NFS client. Ideally, the NICs issuing reads/writes to the NFS server should be on the same PCIe switch as the issuing GPU. On NVIDIA DGX systems, the nvidia-smi utility helps to determine the optimal NIC/GPU pairings:
The above example is generated on a DGX-2 system. For brevity, 8 of the 16 GPUs and 4 of the 8 NICs are shown. From the Legend provided with the topology map, we see the optimal path is represented as PIX (Connection traversing at most a single PCIe bridge). Thus, for GPU 0 and 1, the mlx5_0 is the closest NIC, for GPU 2 and 3, it’s mlx5_1, etc. The network device names in Linux use the prefix ib, so we need to see which interface names map to which device names. This is done with the ibdev2netdev utility:

nfs_client> ibdev2netdev
mlx5_0 port 1 ==> ib0 (Up)
mlx5_1 port 1 ==> ib1 (Up)
mlx5_2 port 1 ==> ib2 (Up)
mlx5_3 port 1 ==> ib3 (Up)
mlx5_4 port 1 ==> ib4 (Down)
mlx5_5 port 1 ==> ib5 (Down)
mlx5_6 port 1 ==> ib6 (Up)
mlx5_7 port 1 ==> ib7 (Up)
mlx5_8 port 1 ==> ib8 (Up)
mlx5_9 port 1 ==> ib9 (Up)

The network configuration between the NFS server and client will obviously factor significantly into the delivered performance. It is beyond the scope of this document to
detail the steps involved in configuring networks, subnets, routing tables. An overview covers the basics prior to getting into examples of running gdsio on NFS.

On the server side, there is a single RAID0 device consisting of 8 locally attached NVMe SSDs. The device is configured with an ext4 file system, and made accessible via two mount points. There are two 100GB network interfaces configured on the server for NFS traffic, and each NIC is assigned multiple IP addresses to different subnets to balance network traffic from the client.

On the client side, there are 8 100GB networks configured. The network configuration (subnets, routing tables) determines which interface on the client side will route to which of the two interfaces on the server side when the client issues an NFS mount. The client-side mount paths have been intentionally named to reflect the network path between the client and the server. Specifically, the mount points on the client are decomposed as:

```
/mnt/nfs/[client-side network interface]/data/[server-side network interface]
```

where the client network interface name is the actual device (ib9, ib8, etc.) and the server is either 0 or 1, indicating which one of the two server networks will handle the connection.

```
nfs_client> mount | grep nfs
192.168.0.10:/mnt/nfs_10/10 on /mnt/nfs/ib9/data/0 type nfs
   (rw,relatime,vers=3,rsize=1048576,wsize=1048576,namlen=255,hard,proto=rdma,port=20049,
    timeo=600,retrans=2,sec=sys,mountaddr=192.168.0.10,mountvers=3,mountproto=tcp,
    local_lock=none,addr=192.168.0.10)
192.168.0.11:/mnt/nfs_11/11 on /mnt/nfs/ib9/data/1 type nfs
   (rw,relatime,vers=3,rsize=1048576,wsize=1048576,namlen=255,hard,proto=rdma,port=20049,
    timeo=600,retrans=2,sec=sys,mountaddr=192.168.0.11,mountvers=3,mountproto=tcp,
    local_lock=none,addr=192.168.0.11)
192.168.1.10:/mnt/nfs_10/10 on /mnt/nfs/ib8/data/0 type nfs
   (rw,relatime,vers=3,rsize=1048576,wsize=1048576,namlen=255,hard,proto=rdma,port=20049,
    timeo=600,retrans=2,sec=sys,mountaddr=192.168.1.10,mountvers=3,mountproto=tcp,
    local_lock=none,addr=192.168.1.10)
192.168.1.11:/mnt/nfs_11/11 on /mnt/nfs/ib8/data/1 type nfs
   (rw,relatime,vers=3,rsize=1048576,wsize=1048576,namlen=255,hard,proto=rdma,port=20049,
    timeo=600,retrans=2,sec=sys,mountaddr=192.168.1.11,mountvers=3,mountproto=tcp,
    local_lock=none,addr=192.168.1.11)
...```

In the mount output on the client (partial, shown above), `/mnt/nfs/ib9/data/0` will route through the ib9 interface on the client to the interface assigned 192.168.0.10 on the server (reflected as ‘0’ in the path string), and `/mnt/nfs/ib9/data/1` will also route through ib9 on the client, to the second interface on the server, 192.168.0.11 (reflected as ’1’ in the path string). This convention is used for all the client-side mounts, so IO pathnames can be selected to effectively balance the load across all interfaces.

From a performance/benchmarking perspective, the key aspects of the configuration under test, in addition to the balanced network setup just discussed, are the underlying storage config on the server (8 x NVMe SSD) and number and speed of network interfaces on the server (2 x 100Gb). In terms of throughput, two 100Gb networks can sustain about 12GB/sec each under ideal conditions. The storage servers can support (2 x 12GB/sec) 24GB/sec or so, but keep in mind there are many things that factor into delivered performance - protocol overhead, network MTU size, NFS attributes, software stack, load characteristics, etc.
In this first example, gdsio is used to generate a random write load of small IOs (4k) to one of the NFS mount points, that will traverse ib0 on the client side. The ib0 interface is on the same PCIe segment as GPU 0 and 1.

```
nfs_client> gdsio -D /mnt/nfs/ib0/data/0/gds_dir -d 0 -w 32 -s 500M -i 4K -x 0 -I 3 -T 120
IoType: RANDWRITE XferType: GPUD Threads: 32 DataSetSize: 81017740/16384000 (Kib)
IOSize: 4 (Kib) Throughput: 0.645307 GiB/sec, Avg_Latency: 189.166333 usecs ops: 20254435 total_time 119.732906 secs
```

The first invocation specifies GPU 0, the second GPU 12. Note the difference in ops and latency. In the optimal GPU/IB case, we observed just over 169k IOPS, 189.2us average latency. In the non-optimal case (GPU 12 to the same mount point), we see an increase in latency to 213.3 usec at about 150k IOPS. The performance difference is not huge (12% in latency, 19% in IOPS), but worth noting nonetheless.

For random reads, the difference between the optimal and sub-optimal case is larger:

```
nfs_client> gdsio -D /mnt/nfs/ib0/data/0/gds_dir -d 0 -w 32 -s 500M -i 4K -x 0 -I 2 -T 120
IoType: RANDREAD XferType: GPUD Threads: 32 DataSetSize: 111181604/16384000 (Kib)
IOSize: 4 (Kib) Throughput: 0.890333 GiB/sec, Avg_Latency: 137.105980 usecs ops: 27795401 total_time 119.091425 secs
```

With GPU 0 and ib0, we see about 234k IOPS, 194us average latency. With GPU 10 and ib0, we see about 165k IOPS, 194us latency.

Small, random IOs are all about IOPS and latency. For determining throughput, we use larger files sizes and much larger IO sizes.

```
nfs_client> gdsio -D /mnt/nfs/ib0/data/0/gds_dir -d 0 -w 32 -s 1G -i 1M -x 0 -I 1 -T 120
IoType: WRITE XferType: GPUD Threads: 64 DataSetSize: 876086272/67108864 (Kib)
IOSize: 1024 (Kib) Throughput: 6.962237 GiB/sec, Avg_Latency: 8976.802942 usecs ops: 855553 total_time 120.004668 secs
```

Above a large sequential write, then read, was generated. In both cases, 32 threads were spawned (-w 32) doing 1M IOs. In the write case, we sustained 6.9GB/sec, and in the read case 9.5GB/sec throughput. Both invocations used ib0 and GPU 0. Changing the GPU from 0 to 8:

```
nfs_client> gdsio -D /mnt/nfs/ib0/data/0/gds_dir -d 8 -w 32 -s 1G -i 1M -x 0 -I 0 -T 120
IoType: READ XferType: GPUD Threads: 32 DataSetSize: 1196929024/33554432 (Kib)
IOSize: 1024 (Kib) Throughput: 9.482088 GiB/sec, Avg_Latency: 3295.183817 usecs ops: 1168876 total_time 120.382817 secs
```

Above a large sequential write, then read, was generated. In both cases, 32 threads were spawned (-w 32) doing 1M IOs. In the write case, we sustained 6.9GB/sec, and in the read case 9.5GB/sec throughput. Both invocations used ib0 and GPU 0. Changing the GPU from 0 to 8:

```
nfs_client> gdsio -D /mnt/nfs/ib0/data/0/gds_dir -d 8 -w 32 -s 1G -i 1M -x 0 -I 0 -T 120
IoType: READ XferType: GPUD Threads: 32 DataSetSize: 1053419520/33554432 (Kib)
IOSize: 2048 (Kib) Throughput: 8.352013 GiB/sec, Avg_Latency: 7480.408305 usecs ops: 514365 total_time 120.284676 secs
```
We note again a decrease in throughput (8.3GB/sec from 9.5GB/sec) and increase in latency (3.4ms to 7.5ms).

Using the supported gdsio config files facilitates an easily reusable “tool box” of IO loads and configurations. Also, with multiple jobs running, gdsio will aggregate the results, making it easier to see the complete performance picture.

Here's a sample configuration file to generate 4k random reads to multiple NFS mount points for 4 different GPUs. Note various parameters defined in the global section, then job-specific parameters (GPU, target mount point, number of threads) in each job section.

```
[globals]
name=nfs_random_read
#0,1,2,3,4,5
xfer_type=0
#IO type, rw=read, rw=write, rw=randread, rw=randwrite
rw=randread
#block size, for variable block size can specify range e.g. bs=1M:4M:1M, (1M : start block size, 4M : end block size, 1M : steps in which size is varied)
bs=4k
#file-size
size=500M
#secs
runtime=120

[jobs]
#numa node
numa_node=0
gpu_dev_id=0
num_threads=16
directory=/mnt/nfs/ib0/data/0/gds_dir

[job2]
numa_node=0
gpu_dev_id=2
num_threads=16
directory=/mnt/nfs/ib1/data/0/gds_dir

[job3]
numa_node=0
gpu_dev_id=4
num_threads=16
directory=/mnt/nfs/ib2/data/0/gds_dir

[job4]
numa_node=0
gpu_dev_id=6
num_threads=16
directory=/mnt/nfs/ib3/data/0/gds_dir
```

Executing gdsio using the above config file, simply pass the file name as the only argument:

```
nfs_client> gdsio nfs_rr.gdsio
```

IoType: RANDREAD XferType: GPUD Threads: 64 Data Set Size: 277467756/32768000 (KiB)

IOCsize: 4 (KiB) Throughput: 2.213928 GiB/sec, Avg_Latency: 110.279539 usecs ops: 69366939 total_time 119.522363 secs

The sustained random read rate was about 580k IOPS (69366939 / 119.52).
For throughput testing, the load attributes need to be changed so gdsio issues reads and writes, not random reads and random writes (see `rw=` in the global section). Also, the IO size (`bs=` in the global section) must be increased to maximize throughput.

Using an edited gdsio config file reflecting those changes, we can generate a throughput-oriented workload with multiple GPUs. The configuration file:

```plaintext
[global]
name=nfs_large_read
xfer_type=0
rw=read
bs=1M
size=1G
runtime=120
do_verify=0

[job1]
numa_node=0
gpu_dev_id=0
num_threads=8
directory=/mnt/nfs/ib0/data/0/gds_dir

[job3]
numa_node=0
gpu_dev_id=2
num_threads=8
directory=/mnt/nfs/ib1/data/1/gds_dir

[job5]
numa_node=0
gpu_dev_id=4
num_threads=8
directory=/mnt/nfs/ib2/data/0/gds_dir

[job7]
numa_node=0
gpu_dev_id=6
num_threads=8
directory=/mnt/nfs/ib3/data/1/gds_dir

[job9]
numa_node=1
gpu_dev_id=8
num_threads=8
directory=/mnt/nfs/ib6/data/0/gds_dir

[job11]
numa_node=1
gpu_dev_id=10
num_threads=8
directory=/mnt/nfs/ib7/data/1/gds_dir

[job13]
numa_node=1
gpu_dev_id=12
num_threads=8
directory=/mnt/nfs/ib8/data/0/gds_dir

[job15]
numa_node=1
gpu_dev_id=14
num_threads=8
directory=/mnt/nfs/ib9/data/1/gds_dir

Running the large read load:

nfs_client> gdsio nfs_sr.gdsio
```
GPUDirect Storage Benchmarking on Network Attached Storage

We see 12.76GB/sec sustained throughput with the configuration used.

The examples shown are intended to serve as a starting point. NAS storage environments, NFS or partner solutions, can be complex to configure given the number of variables that come into play once a network is introduced in the storage I/O path. Various configuration options can be changed, both in terms of the load generated (the gdso command line or config file), as well as the system setup (network, NFS, etc) in order to determine the optimal configuration for the target workload.

In order to observe the data rates while under load, there are a few options. Certainly, on the server side, Linux utilities like `iostat` should be used to monitor traffic and capture statistics on the backend storage, as well as `nfsstat` on both the client and server side for NFS specific statistics. For byte and packet rates over the networks, there is a dependency on the actual network devices and software stack, and whether or not any per-interface statistics are maintained. In our configuration, NVIDIA/Mellanox cards are used:

```
nfs_client> lspci -v | grep -i mellanox
35:00.0 Infiniband controller: Mellanox Technologies MT27800 Family [ConnectX-5]  
   Subsystem: Mellanox Technologies MT27800 Family [ConnectX-5]  
   ...  
```

In this environment, various counters are maintained that can be examined and, with a relatively simple script, per-second data and packet rates can be monitored. These counters can be found in `/sys/class/infiniband/[INTERFACE]/ports/[PORT NUMBER]/counters`, e.g. `/sys/class/infiniband/mlx5_19/ports/1/counters`. The counters of interest are:

- `port_rcv_data` - receive bytes
- `port_xmit_data` - transmit bytes
- `port_rcv_packets` - receive packets
- `port_xmit_packets` - transmit packets

Note the `ibdev2netdev` utility should be used to determine the correct interface name that corresponds to the configured device name. And of course these same counters will be available on the client, assuming of course the same network hardware and software.

On the client side, GDS maintains a stats file with useful counters specific to GDS:

```
nfs_client> cat /proc/driver/nvidia-fs/stats  
GDS Version: 0.9.0.743  
NVFS statistics (ver: 2.0)  
NVFS Driver (version: 2:3:1)  
Active Shadow-Buffer (MiB): 128  
Active Process: 1  
Reads: n=424436862 ok=424436734 err=0 readMiB=21518771  
io_state_err=0  
Reads: Bandwidth (MiB/s)=13081 Avg-Latency (usec)=9773  
Sparse Reads: n=19783000 io=0 holes=0 pages=0  
Writes: n=309770912 ok=309770912 err=0 writeMiB=9748727  
io_state_err=0 pg-cache=0 pg-cache-fail=0 pg-cache-eio=0  
Writes: Bandwidth (MiB/s)=6958 Avg-Latency (usec)=18386  
Mmap: n=3584 ok=3584 err=0 munmap=3456  
```

NVIDIA Magnum IO GPUDirect Storage
Bar1-map : n=3584 ok=3584 err=0 free=3456 callbacks=0 active=128
Error : cpu-gpu-pages=0 sg-ext=0 dma-map=0
Ops : Read=128 Write=0
GPU 0000:be:00.0 uuid:87e5c586-88ed-583b-df45-fcee0f1e7917 : Registered MiB=0
       Cache MiB=0 max_pinned MiB=8 cross_root_port(%)=0
GPU 0000:e7:00.0 uuid:029faa3b-cb0d-2718-259c-6dc650c636eb : Registered MiB=0
       Cache MiB=0 max_pinned MiB=8 cross_root_port(%)=0

The GDS stats provide read and write operations counts (for example, Reads : n=[read_count] and Writes : n=[write_count]), as well Bandwidth in MB/sec. Also error counters are maintained that should be monitored periodically.
Chapter 6. Summary

Configuring and benchmarking storage can be a complex task. Fortunately the technology and tools have reached a state of maturity in the industry, reducing complexity and enabling reduced time to set up, measure and deploy a configuration that meets production performance requirements.

With the addition of GPUDirect Storage to the GPUDirect family, we now have the technology and the tools to feed data-hungry GPUs at extremely high throughput and very low latency. This translates into faster execution of computationally intensive workloads leveraging the immense compute power of NVIDIA GPUs.
Benchmarking is the process of running software specifically designed to generate data for the purpose of assessing performance. The scope of what is being utilized/Measured will vary. Many benchmarks are designed to simulate production workloads (MLperf, TPC, SPEC, and so on), utilize the entire system, and require an audit process before a vendor can make the benchmark results publicly available. Some benchmarks target measuring the performance of a specific subsystem; CPUs, GPUs, storage, etc. Such benchmarks are sometimes referred to as microbenchmarks, and are often publicly available software intended to generate load and report performance results specific to the components under test, e.g. iperf for network performance and fio for disk/storage performance. These subsystem-specific benchmarks and tools are extremely useful as they enable a “building-block” approach to overall system capability and performance. They also are good tools for verifying various compute subsystems before moving to a full system workload.

A.1. The Language of Performance

System performance is typically described either in application-specific terms (for example, images-per-second, transactions-per-second, Deep Learning (DL) training time-per-epoch, Inference throughput, and so on) or more generic terms:

**Bandwidth - how much**
- Bandwidth is the theoretical maximum attainable data rate, expressed as bytes-per-second, sometimes bits-per-second.

**Throughput - how fast**
- Throughput can reflect a data rate, e.g. 12GB/sec over a network link, 3.4GB/sec sequential reads from an NVMe SSD. We also express some workload-specific metrics in terms of throughput, for example, words-per-second, images-per-second, and so on.

**IOPS - how many**
- Input/Output operations per second, typically used in the context of disk IO (reads-per-second, writes-per-second) and network IO (packets-per-second, messages-per-second).
Latency - how long

The time required to complete an operation, e.g. 5 milliseconds (ms) to do a disk read on an HDD, 40 microseconds (us) to write to an SSD, 70 nanoseconds (ns) for a CPU read data from system memory.

Specific aspects of GPU and CPU performance may also be expressed in one or more of the above terms, for example, throughput as the number of floating point or integer operations per second, memory bandwidth and latency, interconnect (UPI/QPI, NVswitch, etc.) bandwidth and latency, and so on.

Monitoring performance and capturing metrics that fall into one or more of the above categories is typically done using base operating system utilities. There are a great many tools and utilities available for Linux intended to provide observability into system utilization and performance. A discussion of these tools is beyond the scope of this paper, but we will refer to specific examples as applicable. Also, various benchmark tools generate these detailed metrics, for example, the gdsio utility, which gets installed as part of GDS, generates detailed data points of throughput, latency, and IOPS.

A.2. Benchmarking Storage Performance

Determining expected storage performance (and also network performance) has the advantage of doing the math based on the specifications of the underlying hardware being measured. For example, if the target storage is Direct Attached Storage (DAS) in the form of NVMe SSDs on the system’s PCIe bus, the device specifications provide throughput, IOPS and latency values, and that information, coupled with known PCIe bandwidth (Gen3, Gen4) and number of PCIe lanes (typically 16 lanes) configured for the devices, the maximum theoretical performance can be calculated. This is discussed further in the Storage Performance Basics for Deep Learning blog post.

For example, NVIDIA® DGX-2® systems include NVMe SSDs for local storage. For such devices configured as a RAID0 volume, the expectation for large reads and writes will be four times the specified performance for a single device. The same logic applies to workloads that are more IOPS intensive (small reads and writes). Putting some numbers on this example, the device specification for large reads indicates 3.5GB/sec, so (4 x 3.5GB/sec) 14GB/sec expected throughput on large reads from the RAID0 volume. As these are PCIe devices, depending on PCIe topology, achieving 14GB/sec pushes the theoretical limit of PCI3 Gen3 16 lanes. The key point here is a solid understanding of the configuration details is necessary to establish performance expectations.

The same logic and methodology applies to Network Attached Storage (NAS), for example, four NVMe SSDs doing large reads or writes will potentially saturate a 100Gb NIC, so it is important to understand the entire storage data path to correctly assess the delivered performance.

An important consideration when benchmarking storage is the presence of a file system. There are two components to file systems that will impact performance:

- The operating system’s page cache
The page cache, which is implemented transparently and used by default when reading and writing data to/from a file system, caches data and metadata in the system's main memory. The performance benefit of the page cache can be substantial, as a typical two CPU socket system can read/write system memory at significantly higher throughput and lower latency than storage.

- The file system-specific settings and tuneable parameters

  File system-specific settings/parameters can happen at both creation time (mkfs) and at mount time in the form of mount options. Some mount options are file system independent (for example, atime, async, and so on), while other options are specific to the file system. Any performance evaluation of the storage technology requires bypassing the page cache to ensure that actual storage performance is being measured, and not the page cache read/write performance.

GDS currently supports the ext4 file system on Linux, and requires the use of the O_DIRECT flag on files that are targeted for read/write by the GDS IO APIs. With O_DIRECT set, you can bypass the system's page cache.

Another factor in assessing storage performance results are the IO characteristics, which are historically categorized as:

- Random IO (small IO size)
- Sequential IO (large IO size)

The terms random and sequential refer to the on-disk block layout of the target files and are relevant when assessing storage that uses hard disk drives (HDDs). HDD technology implements spinning disks with read/write heads that perform seek operations to locate the target blocks where the data is to be read or written. This seek time induces IO latency due to electro-mechanical delays, (moving the read/write heads around the platters). With Solid State Disks (SSDs), no such electro-mechanical exist, so sequential versus random IO on SSDs is not a consideration in terms of seek times and latency.

However, in terms of setting expectations for the results, it is still important to factor in IO sizes. For IO loads that tend to be small (< 500kB), the performance metric of relevance is IO operations per second (IOPS) and latency, versus throughput, because loads with smaller IO sizes will not necessarily maximize available throughput. If assessing maximum throughput is the goal, larger IO sizes should be used when benchmarking.
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