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1.1. AT A GLANCE

1.1.1. MPS

The Multi-Process Service (MPS) is an alternative, binary-compatible implementation of the CUDA Application Programming Interface (API). The MPS runtime architecture is designed to transparently enable co-operative multi-process CUDA applications, typically MPI jobs, to utilize Hyper-Q capabilities on the latest NVIDIA (Kepler-based) GPUs. Hyper-Q allows CUDA kernels to be processed concurrently on the same GPU; this can benefit performance when the GPU compute capacity is underutilized by a single application process.

1.1.2. Volta MPS

The Volta architecture introduced new MPS capabilities. Compared to MPS on pre-Volta GPUs, Volta MPS provides a few key improvements:

   Volta MPS clients submit work directly to the GPU without passing through the MPS server.
   Each Volta MPS client owns its own GPU address space instead of sharing GPU address space with all other MPS clients.
   Volta MPS supports limited execution resource provisioning for Quality of Service (QoS).

This document will introduce the new capabilities, and note the differences between Volta MPS and MPS on pre-Volta GPUs. Running MPS on Volta will automatically enable the new capabilities.
1.1.3. Intended Audience

This document is a comprehensive guide to MPS capabilities and usage. It is intended to be read by application developers & users who will be running GPU calculations and intend to achieve the greatest level of execution performance. It is also intended to be read by system administrators who will be enabling the MPS capability in a user-friendly way, typically on multi-node clusters.

1.1.4. Organization of This Document

The order of the presentation is as follows:

- Introduction and Concepts – describes why MPS is needed and how it enables Hyper-Q for multi-process applications.
- When to Use MPS – describes what factors to consider when choosing to run an application with or choosing to deploy MPS for your users.
- Architecture – describes the client-server architecture of MPS in detail and how it multiplexes clients onto the GPU.
- Appendices – Reference information for the tools and interfaces used by the MPS system and guidance for common use-cases.

1.2. Prerequisites

Portions of this document assume that you are already familiar with:

- the structure of CUDA applications and how they utilize the GPU via the CUDA Runtime and CUDA Driver software libraries.
- concepts of modern operating systems, such as how processes and threads are scheduled and how inter-process communication typically works.
- the Linux command-line shell environment.
configuring and running MPI programs via a command-line interface

1.3. Concepts

1.3.1. Why MPS is needed

To balance workloads between CPU and GPU tasks, MPI processes are often allocated individual CPU cores in a multi-core CPU machine to provide CPU-core parallelization of potential Amdahl bottlenecks. As a result, the amount of work each individual MPI process is assigned may underutilize the GPU when the MPI process is accelerated using CUDA kernels. While each MPI process may end up running faster, the GPU is being used inefficiently. The Multi-Process Service takes advantage of the inter-MPI rank parallelism, increasing the overall GPU utilization.

1.3.2. What MPS is

MPS is a binary-compatible client-server runtime implementation of the CUDA API which consists of several components.

- Control Daemon Process – The control daemon is responsible for starting and stopping the server, as well as coordinating connections between clients and servers.
- Client Runtime – The MPS client runtime is built into the CUDA Driver library and may be used transparently by any CUDA application.
- Server Process – The server is the clients’ shared connection to the GPU and provides concurrency between clients.

1.4. See Also

Manpage for nvidia-cuda-mps-control (1)
Manpage for nvidia-smi (1)
Blog “Unleash Legacy MPI Codes With Kepler’s Hyper-Q” by Peter Messmer (http://blogs.nvidia.com/2012/08/unleash-legacy-mpi-codes-with-keplers-hyper-q)
WHEN TO USE MPS

2.1. The Benefits of MPS

2.1.1. GPU utilization

A single process may not utilize all the compute and memory-bandwidth capacity available on the GPU. MPS allows kernel and memcopy operations from different processes to overlap on the GPU, achieving higher utilization and shorter running times.

2.1.2. Reduced on-GPU context storage

Without MPS each CUDA processes using a GPU allocates separate storage and scheduling resources on the GPU. In contrast, the MPS server allocates one copy of GPU storage and scheduling resources shared by all its clients. Volta MPS supports increased isolation between MPS clients, so the resource reduction is to a much lesser degree.

2.1.3. Reduced GPU context switching

Without MPS, when processes share the GPU their scheduling resources must be swapped on and off the GPU. The MPS server shares one set of scheduling resources between all of its clients, eliminating the overhead of swapping when the GPU is scheduling between those clients.

2.2. Identifying Candidate applications

MPS is useful when each application process does not generate enough work to saturate the GPU. Multiple processes can be run per node using MPS to enable more concurrency. Applications like this are identified by having a small number of blocks-per-grid.

Further, if the application shows a low GPU occupancy because of a small number of threads-per-grid, performance improvements may be achievable with MPS. Using fewer blocks-per-grid in the kernel invocation and more threads-per-block to increase
the occupancy per block is recommended. MPS allows the leftover GPU capacity to be occupied with CUDA kernels running from other processes.

These cases arise in strong-scaling situations, where the compute capacity (node, CPU core and/or GPU count) is increased while the problem size is held fixed. Though the total amount of computation work stays the same, the work per process decreases and may underutilize the available compute capacity while the application is running. With MPS, the GPU will allow kernel launches from different processes to run concurrently and remove an unnecessary point of serialization from the computation.

2.3. Considerations

2.3.1. System Considerations

2.3.1.1. Limitations

MPS is only supported on the Linux operating system. The MPS server will fail to start when launched on an operating system other than Linux.

MPS is not supported on Tegra platforms. The MPS server will fail to start when launched on Tegra platforms.

MPS requires a GPU with compute capability version 3.5 or higher. The MPS server will fail to start if one of the GPUs visible after applying CUDA_VISIBLE_DEVICES is not of compute capability 3.5 or higher.

The Unified Virtual Addressing (UVA) feature of CUDA must be available, which is the default for any 64-bit CUDA program running on a GPU with compute capability version 2.0 or higher. If UVA is unavailable, the MPS server will fail to start.

The amount of page-locked host memory that can be allocated by MPS clients is limited by the size of the tmpfs filesystem (/dev/shm).

Exclusive-mode restrictions are applied to the MPS server, not MPS clients.

Only one user on a system may have an active MPS server.

The MPS control daemon will queue MPS server activation requests from separate users, leading to serialized exclusive access of the GPU between users regardless of GPU exclusivity settings.

All MPS client behavior will be attributed to the MPS server process by system monitoring and accounting tools (e.g. nvidia-smi, NVML API)

2.3.1.2. GPU Compute Modes

Three Compute Modes are supported via settings accessible in nvidia-smi.

PROHIBITED – the GPU is not available for compute applications.

EXCLUSIVE_PROCESS – the GPU is assigned to only one process at a time, and individual process threads may submit work to the GPU concurrently.

DEFAULT – multiple processes can use the GPU simultaneously. Individual threads of each process may submit work to the GPU simultaneously.
Using MPS effectively causes EXCLUSIVE_PROCESS mode to behave like DEFAULT mode for all MPS clients. MPS will always allow multiple clients to use the GPU via the MPS server.

When using MPS it is recommended to use EXCLUSIVE_PROCESS mode to ensure that only a single MPS server is using the GPU, which provides additional insurance that the MPS server is the single point of arbitration between all CUDA processes for that GPU.

2.3.2. Application Considerations

The NVIDIA Codec SDK: https://developer.nvidia.com/nvidia-video-codec-sdk is not supported under MPS on pre-Volta MPS clients.

Only 64-bit applications are supported. The MPS server will fail to start if the CUDA application is not 64-bit. The MPS client will fail CUDA initialization.

If an application uses the CUDA driver API, then it must use headers from CUDA 4.0 or later (i.e. it must not have been built by setting CUDA_FORCE_API_VERSION to an earlier version). Context creation in the client will fail if the context version is older than 4.0.

Dynamic parallelism is not supported. CUDA module load will fail if the module uses dynamic parallelism features.

MPS server only supports clients running with the same UID as the server. The client application will fail to initialize if the server is not running with the same UID.

Stream callbacks are not supported on pre-Volta MPS clients. Calling any stream callback APIs will return an error.

CUDA graphs with host nodes are not supported under MPS on pre-Volta MPS clients.

The amount of page-locked host memory that pre-Volta MPS client applications can allocate is limited by the size of the tmpfs filesystem (/dev/shm). Attempting to allocate more page-locked memory than the allowed size using any of relevant CUDA APIs will fail.

Terminating an MPS client without synchronizing with all outstanding GPU work (via Ctrl-C / program exception such as segfault / signals, etc.) can leave the MPS server and other MPS clients in an undefined state, which may result in hangs, unexpected failures, or corruptions.

CUDA IPC between CUDA contexts which are created by processes running as MPS clients and CUDA contexts which are created by processes not running as MPS clients is supported under Volta MPS.

2.3.3. Memory Protection and Error Containment

MPS is only recommended for running cooperative processes effectively acting as a single application, such as multiple ranks of the same MPI job, such that the severity of the following memory protection and error containment limitations is acceptable.

2.3.3.1. Memory Protection

Volta MPS client processes have fully isolated GPU address spaces.
Pre-Volta MPS client processes allocate memory from different partitions of the same GPU virtual address space. As a result:

- An out-of-range write in a CUDA Kernel can modify the CUDA-accessible memory state of another process, and will not trigger an error.
- An out-of-range read in a CUDA Kernel can access CUDA-accessible memory modified by another process, and will not trigger an error, leading to undefined behavior.

This pre-Volta MPS behavior is constrained to memory accesses from pointers within CUDA Kernels. Any CUDA API restricts MPS clients from accessing any resources outside of that MPS Client’s memory partition. For example, it is not possible to overwrite another MPS client’s memory using the cudaMemcpy() API.

### 2.3.3.2. Error Containment

Volta MPS supports limited error containment:

- A fatal GPU exception generated by a Volta MPS client process will be contained within the subset of GPUs shared between all clients with the fatal exception-causing GPU.
- A fatal GPU exception generated by a Volta MPS client process will be reported to all the clients running on the subset of GPUs in which the fatal exception is contained, without indicating which client generated the error. Note that it is the responsibility of the affected clients to exit after being informed of the fatal GPU exception.
- Clients running on other GPUs remain unaffected by the fatal exception and will run as normal till completion.

Once a fatal exception is observed, the MPS server will wait for all the clients associated with the affected GPUs to exit, prohibiting new client connecting to those GPUs from joining. When all the existing clients associated with the affected GPUs have exited, the MPS server will recreate the GPU contexts on the affected GPUs and resume processing client requests to those GPUs.

For example, if your system has devices 0, 1, and 2, and if there are four clients client A, client B, client C, and client D connected to the MPS server: client A runs on device 0, client B runs on device 0 and 1, client C runs on device 1, client D runs on device 2. If client A triggers a fatal GPU exception:

- Since device 0 and device 1 share a comment client, client B, the fatal GPU exception is contained within device 0 and 1.
- The fatal GPU exception will be reported to all the clients running on device 0 and 1, i.e. client A, client B, and client C.
- Client D running on device 2 remain unaffected by the fatal exception and continue to run as normal.

The MPS server will wait for client A, client B, and client C to exit and reject any new client requests to device 0 and device 1 will be rejected with error CUDA_ERROR_MPS_SERVER_NOT_READY. After client A, client B, and client C have
exited, the server recreates the GPU contexts on device 0 and device 1 and then resumes accepting client requests to device 0 and device 1.

Pre-Volta MPS client processes share on-GPU scheduling and error reporting resources. As a result:

A GPU exception generated by any client will be reported to all clients, without indicating which client generated the error.
A fatal GPU exception triggered by one client will terminate the MPS server and the GPU activity of all clients.

CUDA API errors generated on the CPU in the CUDA Runtime or CUDA Driver are delivered only to the calling client.

2.3.4. MPS on Multi-GPU Systems

The MPS server supports using multiple GPUs. On systems with more than one GPU, you can use CUDA_VISIBLE_DEVICES to enumerate the GPUs you would like to use. See section 4.2 for more details.

On systems with a mix of Volta / pre-Volta GPUs, if the MPS server is set to enumerate any Volta GPU, it will discard all pre-Volta GPUs. In other words, the MPS server will either operate only on the Volta GPUs and expose Volta capabilities, or operate only on pre-Volta GPUs.

2.3.5. Performance

2.3.5.1. Client-Server Connection Limits

The pre-Volta MPS Server supports up to 16 client CUDA contexts per-device concurrently. Volta MPS server supports 48 client CUDA contexts per-device. These contexts may be distributed over multiple processes. If the connection limit is exceeded, the CUDA application will fail to create a CUDA Context and return an API error from cuCtxCreate() or the first CUDA Runtime API call that triggers context creation. Failed connection attempts will be logged by the MPS server.

2.3.5.2. Volta MPS Execution Resource Provisioning

Volta MPS supports limited execution resource provisioning. The client contexts can be set to only use a portion of the available threads. The provisioning capability is commonly used to achieve two goals:

Reduce client memory footprint: Since each MPS client process has fully isolated address space, each client context allocates independent context storage and scheduling resources. Those resources scale with the amount of threads available to the client. By default, each MPS client has all available threads useable. As MPS is usually used with multiple processes running simultaneously, making all threads accessible to every client is often unnecessary, and therefore wasteful to allocate full context storage. Reducing the number of threads available will effectively reduce the context storage allocation size.
Improve QoS: The provisioning mechanism can be used as a classic QoS mechanism to limit available compute bandwidth. Reducing the portion of available threads will also concentrate the work submitted by a client to a set of SMs, reducing destructive interference with other clients’ submitted work.

Setting the limit does not reserve dedicated resources for any MPS client context. It simply limits how much resources can be used by a client context. Kernels launched from different MPS client contexts may execute on the same SM, depending on load-balancing.

By default, each client is provisioned to have access to all available threads. This will allow the maximum degree of scheduling freedom, but at a cost of higher memory footprint due to wasted execution resource allocation. The memory usage of each client process can be queried through nvidia-smi.

The provisioning limit can be set via a few different mechanisms for different effects. These mechanisms are categorized into two mechanisms: active thread percentage and programmatic interface. In particular, partitioning via active thread percentage are categorized into two strategies: uniform partitioning and non-uniform partitioning.

The limit constrained by the uniform active thread percentage is configured for a client process when it starts, and cannot be changed for the client process afterwards. The executed limit is reflected through device attribute cudaDevAttrMultiProcessorCount whose value remains unchanged throughout the client process.

The MPS control utility provides 2 sets of commands to set/query the limit of all future MPS clients. See section 4.1.1 for more details.
Alternatively, the limit for all future MPS clients can be set by setting the environment variable CUDA_MPS_ACTIVE_THREAD_PERCENTAGE for the MPS control process. See section 4.2.5.1 for more details.

The limit can be further constrained for new clients by solely setting the environment variable CUDA_MPS_ACTIVE_THREAD_PERCENTAGE for a client process. See section 4.2.5.2 for more details.

The limit constrained by the non-uniform active thread percentage is configured for every client CUDA context and can be changed throughout the client process. The executed limit is reflected through device attribute cudaDevAttrMultiProcessorCount whose value returns the portion of available threads that can be used by the client CUDA context current to the calling thread.

The limit constrained by the uniform partitioning mechanisms can be further constrained for new client CUDA contexts by setting the environment variable CUDA_MPS_ACTIVE_THREAD_PERCENTAGE in conjunction with the environment variable CUDA_MPS_ENABLE_PER_CTX_DEVICE_MULTIPROCESSOR_PARTITIONING.
See sections 4.2.5.3 and 4.2.6 for more details.

The limit constrained by the programmatic partitioning is configured for a client CUDA context created via cuCtxCreate_v3() with the execution affinity CUexecAffinityParam which specifies the number of SMs that the context is limited to use. The executed limit of the context can be queried through cuCtxGetExecAffinity(). See section 5.2 for more details.
A common provisioning strategy is to uniformly partition the available threads equally to each MPS client processes (i.e. set active thread percentage to 100% / n, for n expected MPS client processes). This strategy will allocate close to the minimum amount of execution resources, but it could restrict performance for clients that could occasionally make use of idle resources.

A more optimal strategy is to uniformly partition the portion by half of the number of expected clients (i.e. set active thread percentage to 100% / 0.5n) to give the load balancer more freedom to overlap execution between clients when there are idle resources.

The near optimal provision strategy is to non-uniformly partition the available threads based on the workloads of each MPS clients (i.e. set active thread percentage to 30% for client 1 and set active thread percentage to 70 % client 2 if the ratio of the client1 workload and the client2 workload is 30%: 70%). This strategy will concentrate the work submitted by different clients to disjoint sets of the SMs and effectively minimize the interference between work submissions by different clients.

The most optimal provision strategy is to precisely limit the number of SMs to use for each MPS clients knowing the execution resource requirements of each client (i.e 24 SMs for client1 and 60 SMs for client 2 on a device with 84 SMs). This strategy provides finer grained and more flexible control over the set of SMs the work will be running on than the active thread percentage.

If the active thread percentage is used for partitioning, the limit will be internally rounded down to the nearest hardware supported thread count limit. If the programmatic interface is used for partitioning, the limit will be internally rounded up to the nearest hardware supported SM count limit.

2.3.5.3. Threads & Linux Scheduling

On pre-Volta GPUs, launching more MPS clients than there are available logical cores on your machine will incur increased launch latency and will generally slow down client-server communication due to how the threads get scheduled by the Linux CFS (Completely Fair Scheduler). For setups where multiple GPUs are used with an MPS control daemon and server started per GPU, we recommend pinning each MPS server to a distinct core. This can be accomplished by using the utility ‘taskset’, which allows binding a running program to multiple cores or launching a new one on them. To accomplish this with MPS, launch the control daemon bound to a specific core, e.g. ’taskset –c 0 nvidia-cuda-mps-control –d’. The process affinity will be inherited by the MPS server when it starts up.

2.3.6. Interaction with Tools

2.3.6.1. Debugging and cuda-gdb

Under certain conditions applications invoked from within cuda-gdb (or any CUDA-compatible debugger, such as Allinea DDT) may be automatically run without using MPS, even when MPS automatic provisioning is active. To take advantage of this automatic fallback, no other MPS client applications may be running at the time. This enables debugging of CUDA applications without modifying the MPS configuration for the system.
Here’s how it works:

cuda-gdb attempts to run an application and recognizes that it will become an MPS client.
The application running under cuda-gdb blocks in cuInit() and waits for all of the active MPS client processes to exit, if any are running.
Once all client processes have terminated, the MPS server will allow cuda-gdb and the application being debugged to continue.
If any new client processes attempt to connect to the MPS server while cuda-gdb is running, the new MPS client will block in cuInit() until the debugger has terminated.
The client applications will continue normally after the debugger has terminated.

2.3.6.2. cuda-memcheck

The cuda-memcheck tool is supported on MPS. See the cuda-memcheck documentation for usage instructions.

2.3.6.3. Profiling

CUDA profiling tools (such as nvprof and Nvidia Visual Profiler) and CUPTI based profilers are supported under MPS. See the profiler documentation for usage instructions.
3.1. Background

CUDA is a general purpose parallel computing platform and programming model that leverages the parallel compute engine in NVIDIA GPUs to solve many complex computational problems in a more efficient way than on a CPU.

A CUDA program starts by creating a CUDA context, either explicitly using the driver API or implicitly using the runtime API, for a specific GPU. The context encapsulates all the hardware resources necessary for the program to be able to manage memory and launch work on that GPU.

Launching work on the GPU typically involves copying data over to previously allocated regions in GPU memory, running a CUDA kernel that operates on that data, and then copying the results back from GPU memory into system memory. A CUDA kernel consists of a hierarchy of thread groups that execute in parallel on the GPUs compute engine.

All work on the GPU launched using CUDA is launched either explicitly into a CUDA stream, or implicitly using a default stream. A stream is a software abstraction that represents a sequence of commands, which may be a mix of kernels, copies, and other commands, that execute in order. Work launched in two different streams can execute simultaneously, allowing for coarse grained parallelism.

CUDA streams are aliased onto one or more ‘work queues’ on the GPU by the driver. Work queues are hardware resources that represent an in-order sequence of the subset of commands in a stream to be executed by a specific engine on the GPU, such as the kernel executions or memory copies. GPU’s with Hyper-Q have a concurrent scheduler to schedule work from work queues belonging to a single CUDA context. Work launched to the compute engine from work queues belonging to the same CUDA context can execute concurrently on the GPU.

The GPU also has a time sliced scheduler to schedule work from work queues belonging to different CUDA contexts. Work launched to the compute engine from work queues belonging to different CUDA contexts cannot execute concurrently. This can cause underutilization of the GPU’s compute resources if work launched from a single CUDA context is not sufficient to use up all resource available to it.
Additionally, within the software layer, to receive asynchronous notifications from the OS and perform asynchronous CPU work on behalf of the application the CUDA Driver may create internal threads: an upcall handler thread and potentially a user callback executor thread.

### 3.2. Client-server Architecture

This diagram shows a likely schedule of CUDA kernels when running an MPI application consisting of multiple OS processes without MPS. Note that while the CUDA kernels from within each MPI process may be scheduled concurrently, each MPI process is assigned a serially scheduled time-slice on the whole GPU.
When using pre-Volta MPS, the server manages the hardware resources associated with a single CUDA context. The CUDA contexts belonging to MPS clients funnel their work through the MPS server. This allows the client CUDA contexts to bypass the hardware limitations associated with time sliced scheduling, and permit their CUDA kernels execute simultaneously.

Volta provides new hardware capabilities to reduce the types of hardware resources the MPS server must managed. A client CUDA context manages most of the hardware resources on Volta, and submits work to the hardware directly. The Volta MPS server mediates the remaining shared resources required to ensure simultaneous scheduling of work submitted by individual clients, and stays out of the critical execution path.

The communication between the MPS client and the MPS server is entirely encapsulated within the CUDA driver behind the CUDA API. As a result, MPS is transparent to the MPI program.

MPS clients CUDA contexts retain their upcall handler thread and any asynchronous executor threads. The MPS server creates an additional upcall handler thread and creates a worker thread for each client.
3.3. Provisioning Sequence

System-wide provisioning with multiple users.

3.3.1. Server

The MPS control daemon is responsible for the startup and shutdown of MPS servers. The control daemon allows at most one MPS server to be active at a time. When an MPS client connects to the control daemon, the daemon launches an MPS server if there is no server active. The MPS server is launched with the same user id as that of the MPS client.

If there is an MPS server already active and the user id of the server and client match, then the control daemon allows the client to proceed to connect to the server. If there is an MPS server already active, but the server and client were launched with different user id’s, the control daemon requests the existing server to shutdown once all its clients have disconnected. Once the existing server has shutdown, the control daemon launches a new server with the same user id as that of the new user’s client process. This is shown in the figure above where user Bob starts client C’ before a server is available. Only once user Alice’s clients exit is a server created for user Bob and client C’.
The MPS control daemon does not shutdown the active server if there are no pending client requests. This means that the active MPS server process will persist even if all active clients exit. The active server is shutdown when either a new MPS client, launched with a different user id than the active MPS server, connects to the control daemon or when the work launched by the clients has caused an exception. This is shown in the example above, where the control daemon issues a server exit request to Alice’s server only once user Bob starts client C, even though all of Alice’s clients have exited.

The control daemon executable also supports an interactive mode where a user with sufficient permissions can issue commands, for example to see the current list of servers and clients or startup and shutdown servers manually.

### 3.3.2. Client Attach/Detach

When CUDA is first initialized in a program, the CUDA driver attempts to connect to the MPS control daemon. If the connection attempt fails, the program continues to run as it normally would without MPS. If however, the connection attempt succeeds, the MPS control daemon proceeds to ensure that an MPS server, launched with same user id as that of the connecting client, is active before returning to the client. The MPS client then proceeds to connect to the server.

All communication between the MPS client, the MPS control daemon, and the MPS server is done using named pipes and UNIX domain sockets. The MPS server launches a worker thread to receive commands from the client. Upon client process exit, the server destroys any resources not explicitly freed by the client process and terminates the worker thread.
The following utility programs and environment variables are used to manage the MPS execution environment. They are described below, along with other relevant pieces of the standard CUDA programming environment.

### 4.1. Utilities and Daemons

#### 4.1.1. nvidia-cuda-mps-control

Typically stored under /usr/bin on Linux systems and typically run with superuser privileges, this control daemon is used to manage the nvidia-cuda-mps-server described in the section following. These are the relevant use cases:

- `man nvidia-cuda-mps-control` # Describes usage of this utility.
- `nvidia-cuda-mps-control -d` # Start daemon in background process.
- `ps -ef | grep mps` # See if the MPS daemon is running.
- `echo quit | nvidia-cuda-mps-control` # Shut the daemon down.
- `nvidia-cuda-mps-control -f` # Start daemon in foreground

When used in interactive mode, the available commands are

- `get_server_list` – this will print out a list of all PIDs of server instances.
- `start_server -uid <user id>` - this will manually start a new instance of nvidia-cuda-mps-server with the given user ID.
- `get_client_list <PID>` - this lists the PIDs of client applications connected to a server instance assigned to the given PID
- `quit` – terminates the nvidia-cuda-mps-control daemon

Commands available to Volta MPS control:

- `get_device_client_list [<PID>]` - this lists the devices and PIDs of client applications that enumerated this device. It optionally takes the server instance PID.
- `set_default_active_thread_percentage <percentage>` - this overrides the default active thread percentage for MPS servers. If there is already a server spawned, this
command will only affect the next server. The set value is lost if a quit command is executed. The default is 100.

get_default_active_thread_percentage - queries the current default available thread percentage.

set_active_thread_percentage <PID> <percentage> - this overrides the active thread percentage for the MPS server instance of the given PID. All clients created with that server afterwards will observe the new limit. Existing clients are not affected.

g_get_active_thread_percentage <PID> - queries the current available thread percentage of the MPS server instance of the given PID.

Only one instance of the nvidia-cuda-mps-control daemon should be run per node.

4.1.2. nvidia-cuda-mps-server

Typically stored under /usr/bin on Linux systems, this daemon is run under the same $UID as the client application running on the node. The nvidia-cuda-mps-server instances are created on-demand when client applications connect to the control daemon. The server binary should not be invoked directly, and instead the control daemon should be used to manage the startup and shutdown of servers.

The nvidia-cuda-mps-server process owns the CUDA context on the GPU and uses it to execute GPU operations for its client application processes. Due to this, when querying active processes via nvidia-smi (or any NVML-based application) nvidia-cuda-mps-server will appear as the active CUDA process rather than any of the client processes.

4.1.3. nvidia-smi

Typically stored under /usr/bin on Linux systems, this is used to configure GPU’s on a node. The following use cases are relevant to managing MPS:

man nvidia-smi # Describes usage of this utility.
nvidia-smi -L # List the GPU’s on node.
nvidia-smi -q # List GPU state and configuration information.
nvidia-smi -q -d compute # Show the compute mode of each GPU.
nvidia-smi -i 0 -c EXCLUSIVE_PROCESS # Set GPU 0 to exclusive mode, run as root.
nvidia-smi -i 0 -c DEFAULT # Set GPU 0 to default mode, run as root. (SHARED_PROCESS)
nvidia-smi -i 0 -r # Reboot GPU 0 with the new setting.

4.2. Environment Variables
4.2.1. **CUDA_VISIBLE_DEVICES**

CUDA_VISIBLE_DEVICES is used to specify which GPU’s should be visible to a CUDA application. Only the devices whose index or UUID is present in the sequence are visible to CUDA applications and they are enumerated in the order of the sequence.

When CUDA_VISIBLE_DEVICES is set before launching the control daemon, the devices will be remapped by the MPS server. This means that if your system has devices 0, 1 and 2, and if CUDA_VISIBLE_DEVICES is set to “0,2”, then when a client connects to the server it will see the remapped devices - device 0 and a device 1. Therefore, keeping CUDA_VISIBLE_DEVICES set to “0,2” when launching the client would lead to an error.

The MPS control daemon will further filter-out any pre-Volta devices, if any visible device is Volta+.

To avoid this ambiguity, we recommend using UUIDs instead of indices. These can be viewed by launching nvidia-smi -q. When launching the server, or the application, you can set CUDA_VISIBLE_DEVICES to “UUID_1,UUID_2”, where UUID_1 and UUID_2 are the GPU UUIDs. It will also work when you specify the first few characters of the UUID (including “GPU-“) rather than the full UUID.

The MPS server will fail to start if incompatible devices are visible after the application of CUDA_VISIBLE_DEVICES.

4.2.2. **CUDA_MPS_PIPE_DIRECTORY**

The MPS control daemon, the MPS server, and the associated MPS clients communicate with each other via named pipes and UNIX domain sockets. The default directory for these pipes and sockets is /tmp/nvidia-mps. The environment variable, CUDA_MPS_PIPE_DIRECTORY, can be used to override the location of these pipes and sockets. The value of this environment variable should be consistent across all MPS clients sharing the same MPS server, and the MPS control daemon.

The recommended location for the directory containing these named pipes and domain sockets is local folders such as /tmp. If the specified location exists in a shared, multi-node filesystem, the path must be unique for each node to prevent multiple MPS servers or MPS control daemons from using the same pipes and sockets. When provisioning MPS on a per-user basis, the directory should be set to a location such that different users will not end up using the same directory.

4.2.3. **CUDA_MPS_LOG_DIRECTORY**

The MPS control daemon maintains a control.log file which contains the status of its MPS servers, user commands issued and their result, and startup and shutdown notices for the daemon. The MPS server maintains a server.log file containing its startup and shutdown information and the status of its clients.

By default these log files are stored in the directory /var/log/nvidia-mps. The CUDA_MPS_LOG_DIRECTORY environment variable can be used to override the default value. This environment variable should be set in the MPS control daemon’s
environment and is automatically inherited by any MPS servers launched by that control daemon.

### 4.2.4. CUDA_DEVICE_MAX_CONNECTIONS

When encountered in the MPS client’s environment, CUDA_DEVICE_MAX_CONNECTIONS sets the preferred number of compute and copy engine concurrent connections (work queues) from the host to the device for that client. The number actually allocated by the driver may differ from what is requested based on hardware resource limitations or other considerations. Under MPS, each server’s clients share one pool of connections, whereas without MPS each CUDA context would be allocated its own separate connection pool. Volta MPS clients exclusively own the connections set aside for the client in the shared pool, so setting this environment variable under Volta MPS may reduce the number of available clients.

### 4.2.5. CUDA_MPS_ACTIVE_THREAD_PERCENTAGE

On Volta GPUs, this environment variable sets the portion of the available threads that can be used by the client contexts. The limit can be configured at different levels.

#### 4.2.5.1. MPS Control Daemon Level

Setting this environment variable in an MPS control’s environment will configure the default active thread percentage when the MPS control daemon starts.

All the MPS servers spawned by the MPS control daemon will observe this limit. Once the MPS control daemon has started, changing this environment variable cannot affect the MPS servers.

#### 4.2.5.2. Client Process Level

Setting this environment variable in an MPS client’s environment will configure the active thread percentage when the client process starts. The new limit will only further constraint the limit set by the control daemon (via set_default_active_thread_percentage or set_active_thread_percentage control daemon commands or this environment variable at the MPS control daemon level). If the control daemon has a lower setting, the control daemon setting will be obeyed by the client process instead.

All the client CUDA contexts created within the client process will observe the new limit. Once the client process has started, changing the value of this environment variable cannot affect the client CUDA contexts.

#### 4.2.5.3. Client CUDA Context Level

By default, configuring the active thread percentage at the client CUDA context level is disabled. User must explicitly opt-in via environment variable CUDA_MPS_ENABLE_PER_CTX_DEVICE_MULTIPROCESSOR_PARTITIONING. See section 4.2.6 for more details.

Setting this environment variable within a client process will configure the active thread percentage when creating a new client CUDA context. The new limit will only further
constraint the limit set at the control daemon level and the client process level. If the control daemon or the client process has a lower setting, the lower setting will be obeyed by the client CUDA context instead. All the client CUDA contexts created afterwards will observe the new limit. Existing client CUDA contexts are not affected.

**4.2.6. CUDA_MPS_ENABLE_PER_CTX_DEVICE_MULTIPROCESSOR_PARTITIONING**

By default, users can only partition the available threads uniformly. An explicit opt-in via this environment variable is required to enable non-uniform partitioning capability. To enable non-uniform partitioning capability, this environment variable must be set before the client process starts.

When non-uniform partitioning capability is enabled in an MPS client's environment, client CUDA contexts can have different active thread percentages within the same client process via setting CUDA_MPS_ACTIVE_THREAD_PERCENTAGE before context creations. The device attribute cudaDevAttrMultiProcessorCount will reflect the active thread percentage and return the portion of available SMs that can be used by the client CUDA context current to the calling thread.

**4.3. MPS Logging Format**

**4.3.1. Control Log**

The control daemon's log file contains information about the following:

- Startup and shutdown of MPS servers identified by their process ids and the user id with which they are being launched.

  `[2013-08-05 12:50:23.347 Control 13894] Starting new server 13929 for user 500`
  `[2013-08-05 13:02:26.226 Control 13894] Server 13929 exited with status 0`

- New MPS client connections identified by the client process id and the user id of the user that launched the client process.

  `[2013-08-05 13:02:10.866 Control 13894] NEW CLIENT 19276 from user 500: Server already exists`
  `[2013-08-05 13:02:10.961 Control 13894] Accepting connection...`

- User commands issued to the control daemon and their result.

  `[2013-08-05 12:50:23.347 Control 13894] Starting new server 13929 for user 500`

- Error information such as failing to establish a connection with a client.

  `[2013-08-05 13:02:10.961 Control 13894] Accepting connection...`
  `[2013-08-05 13:02:10.961 Control 13894] Unable to read new connection type information`
4.3.2. Server Log

The server’s log file contains information about the following:

- New MPS client connections and disconnections identified by the client process id.

  [2013-08-05 13:00:09.269 Server 13929] New client 14781 connected

  [2013-08-05 13:00:09.270 Server 13929] Client 14777 disconnected

- Error information such as the MPS server failing to start due to system requirements not being met.

  [2013-08-06 10:51:31.706 Server 29489] MPS server failed to start

  [2013-08-06 10:51:31.706 Server 29489] MPS is only supported on 64-bit Linux platforms, with an SM 3.5 or higher GPU.

4.4. MPS KNOWN ISSUES

Clients may fail to start, returning ERROR_OUT_OF_MEMORY when the first CUDA context is created, even though there are fewer client contexts than the hard limit of 16.

Comments: When creating a context, the client tries to reserve virtual address space for the Unified Virtual Addressing memory range. On certain systems, this can clash with the system linker and the dynamic shared libraries loaded by it. Ensure that CUDA initialization (e.g., cuInit(), or any cuda*() Runtime API function) is one of the first functions called in your code. To provide a hint to the linker and to the Linux kernel that you want your dynamic shared libraries higher up in the VA space (where it won’t clash with CUDA’s UVA range), compile your code as PIC (Position Independent Code) and PIE (Position Independent Executable). Refer to your compiler manual for instructions on how to achieve this.

Memory allocation API calls (including context creation) may fail with the following message in the server log: MPS Server failed to create/open SHM segment.

Comments: This is most likely due to exhausting the file descriptor limit on your system. Check the maximum number of open file descriptors allowed on your system and increase if necessary. We recommend setting it to 16384 and higher. Typically this information can be checked via the command ‘ulimit –n’; refer to your operating system instructions on how to change the limit.
The convention for using MPS will vary between system environments. The Cray environment, for example, manages MPS in a way that is almost invisible to the user, whereas other Linux-based systems may require the user to manage activating the control daemon themselves. As a user you will need to understand which set of conventions is appropriate for the system you are running on. Some cases are described in this section.

5.1. Starting and Stopping MPS on LINUX

5.1.1. On a Multi-User System

To cause all users of the system to run CUDA applications via MPS you will need to set up the MPS control daemon to run when the system starts.

5.1.1.1. Starting MPS control daemon

As root, run the commands

- `export CUDA_VISIBLE_DEVICES=0` # Select GPU 0.
- `nvidia-smi -i 0 -c EXCLUSIVE_PROCESS` # Set GPU 0 to exclusive mode.
- `nvidia-cuda-mps-control -d` # Start the daemon.

This will start the MPS control daemon that will spawn a new MPS Server instance for any $UID starting an application and associate it with the GPU visible to the control daemon. Note that CUDA_VISIBLE_DEVICES should not be set in the client process’s environment.

5.1.1.2. Shutting Down MPS control daemon

To shut down the daemon, as root, run

```
echo quit | nvidia-cuda-mps-control
```

5.1.1.3. Log Files

You can view the status of the daemons by viewing the log files in
5.1.2. On a Single-User System

When running as a single user, the control daemon must be launched with the same user id as that of the client process

5.1.2.1. Starting MPS control daemon

As $UID, run the commands

```bash
export CUDA_VISIBLE_DEVICES=0 # Select GPU 0.
export CUDA_MPS_PIPE_DIRECTORY=/tmp/nvidia-mps # Select a location that’s accessible to the given $UID
export CUDA_MPS_LOG_DIRECTORY=/tmp/nvidia-log # Select a location that’s accessible to the given $UID
nvidia-cuda-mps-control -d # Start the daemon.
```

This will start the MPS control daemon that will spawn a new MPS Server instance for that $UID starting an application and associate it with GPU visible to the control daemon.

5.1.2.2. Starting MPS client application

Set the following variables in the client process’s environment. Note that CUDA_VISIBLE_DEVICES should not be set in the client’s environment.

```bash
export CUDA_MPS_PIPE_DIRECTORY=/tmp/nvidia-mps # Set to the same location as the MPS control daemon
export CUDA_MPS_LOG_DIRECTORY=/tmp/nvidia-log # Set to the same location as the MPS control daemon
```

5.1.2.3. Shutting Down MPS

To shut down the daemon, as $UID, run

```bash
echo quit | nvidia-cuda-mps-control
```

5.1.2.4. Log Files

You can view the status of the daemons by viewing the log files in

```bash
$CUDA_MPS_LOG_DIRECTORY/control.log
$CUDA_MPS_LOG_DIRECTORY/server.log
```

5.1.3. Scripting a Batch Queuing System
5.1.3.1. Basic Principles

Chapters 3-4 describe the MPS components, software utilities, and the environment variables that control them. However, using MPS at this level puts a burden on the user since

At the application level, the user only cares whether MPS is engaged or not, and should not have to understand the details of environment settings etc. when they are unlikely to deviate from a fixed configuration.

There may be consistency conditions that need to be enforced by the system itself, such as clearing CPU- and GPU- memory between application runs, or deleting zombie processes upon job completion.

Root-access (or equivalent) is required to change the mode of the GPU.

We recommend you manage these details by building some sort of automatic provisioning abstraction on top of the basic MPS components. This section discusses how to implement a batch-submission flag in the PBS/Torque queuing environment and discusses MPS integration into a batch queuing system in-general.

5.1.3.2. Per-Job MPS Control: A Torque/PBS Example

Note: Torque installations are highly customized. Conventions for specifying job resources vary from site to site and we expect that, analogously, the convention for enabling MPS could vary from site to site as well. Check with your system’s administrator to find out if they already have a means to provision MPS on your behalf.

Tinkering with nodes outside the queuing convention is generally discouraged since jobs are usually dispatched as nodes are released by completing jobs. It is possible to enable MPS on a per-job basis by using the Torque prologue and epilogue scripts to start and stop the nvidia-cuda-mps-control daemon. In this example, we re-use the “account” parameter to request MPS for a job, so that the following command.

```bash
qsub -A "MPS=true" ...
```

will result in the prologue script starting MPS as shown:

```bash
# Activate MPS if requested by user
USER=$2
ACCTSTR=$7
echo $ACCTSTR | grep -i "MPS=true"
if [ $? -eq 0 ]; then
    nvidia-smi -c 3
    USERID=`id -u $USER`
    export CUDA_VISIBLE_DEVICES=0
    nvidia-cuda-mps-control -d && echo "MPS control daemon started"
sleep 1
```
echo "start_server -uid $USERID" | nvidia-cuda-mps-control & & echo "MPS server started for $USER"
fi
and the epilogue script stopping MPS as shown:

# Reset compute mode to default
nvidia-smi -c 0
# Quit cuda MPS if it’s running
ps aux | grep nvidia-cuda-mps-control | grep -v grep > /dev/null
if [ $? -eq 0 ]; then
 echo quit | nvidia-cuda-mps-control
fi
# Test for presence of MPS zombie
ps aux | grep nvidia-cuda-mps | grep -v grep > /dev/null
if [ $? -eq 0 ]; then
 logger "'hostname' epilogue: MPS refused to quit! Marking offline"
 pbsnodes -o -N "Epilogue check: MPS did not quit" `hostname`
fi
# Check GPU sanity, simple check
nvidia-smi > /dev/null
if [ $? -ne 0 ]; then
 logger "'hostname' epilogue: GPUs not sane! Marking 'hostname' offline"
 pbsnodes -o -N "Epilogue check: nvidia-smi failed" `hostname`
fi

5.2. BEST PRACTICE FOR SM PARTITIONING

Creating a context is a costly operation in terms of time, memory, and the hardware resources.
If a context with execution affinity is created at kernel launch time, the user will observe a sudden increase in latency and memory footprint as a result of the context creation. To avoid paying the latency of context creation and the abrupt increase in memory usage at kernel launch time, it is recommended that users create a pool of contexts with different SM partitions upfront and select context with the suitable SM partition on kernel launch:

int device = 0;
cudaDeviceProp prop;
const Int CONTEXT_POOL_SIZE = 4;
CUcontext contextPool[CONTEXT_POOL_SIZE];
int smCounts[CONTEXT_POOL_SIZE];
cudaSetDevice(device);
cudaGetDeviceProperties(&prop, device);
smCounts[0] = 1; smCounts[1] = 2;
smCounts[3] = (prop.multiProcessorCount - 3) / 3;
smCounts[4] = (prop.multiProcessorCount - 3) / 3 * 2;
for (int i = 0; i < CONTEXT_POOL_SIZE; i++) {
    CUexecAffinityParam affinity;
    affinity.type = CU_EXEC_AFFINITY_TYPE_SM_COUNT;
    affinity.param.smCount.val = smCounts[i];
cuCtxCreate_v3(&contextPool[i], affinity, 1, 0, deviceOrdinal);
}
for (int i = 0; i < CONTEXT_POOL_SIZE; i++) {
    std::thread([i]() {
        int numSms = 0;
        int numBlocksPerSm = 0;
        int numThreads = 128;
        CUexecAffinityParam affinity;
        cuCtxSetCurrent(contextPool[i]);
        cuCtxGetExecAffinity(&affinity, CU_EXEC_AFFINITY_TYPE_SM_COUNT);
        numSms = affinity.param.smCount.val;
        cudaOccupancyMaxActiveBlocksPerMultiprocessor(
            &numBlocksPerSm, kernel, numThreads, 0);
        void *kernelArgs[] = { /* add kernel args */ };
        dim3 dimBlock(numThreads, 1, 1);
        dim3 dimGrid(numSms * numBlocksPerSm, 1, 1);
        cudaLaunchCooperativeKernel((void*)my_kernel, dimGrid, dimBlock, kernelArgs);
    });
}
The hardware resources needed for client CUDA contexts is limited and support up to 48 client CUDA contexts per-device on Volta MPS. The size of the context
pool per-device is limited by the number of CUDA client contexts supported per-device. The memory footprint of each client CUDA context and the value of CUDA_DEVICE_MAX_CONNECTIONS may further reduce the number of available clients. Therefore, CUDA client contexts with different SM partitions should be created judiciously.
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