

Octopus: Recent advances in GPU development

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Why GPUs?

- Slower increase in CPU efficiency in last years
- Higher power efficiency of GPUs (FLOPS/Watt)
- All 3 US exascale machines will have GPUs

\rightarrow prepare now!

Octopus

- Density functional theory code with pseudopotentials
- Real-space grid + finite differences
- Real-time time-dependent calculations
- Mainly Fortran, plus some C
- Open source: octopus-code.org

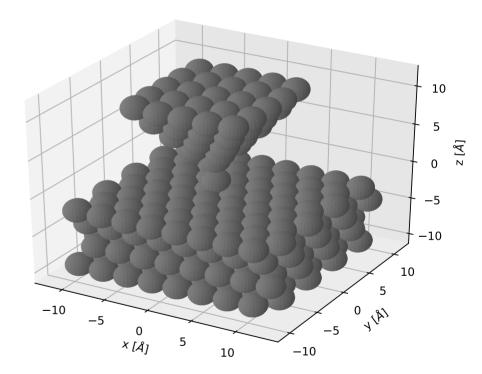


Octopus: GPU version

- Developed ~ 6 yr ago
- Written in OpenCL + wrapper for CUDA
- Interface Fortran $\leftrightarrow C$
- Kernels compiled during runtime

Example system

- Silver tip over crystal
- Periodic in x and y
- 312 Ag atoms
- 3200 orbitals
- 2.4 M grid points
- Compare TD runs



Benchmark clusters

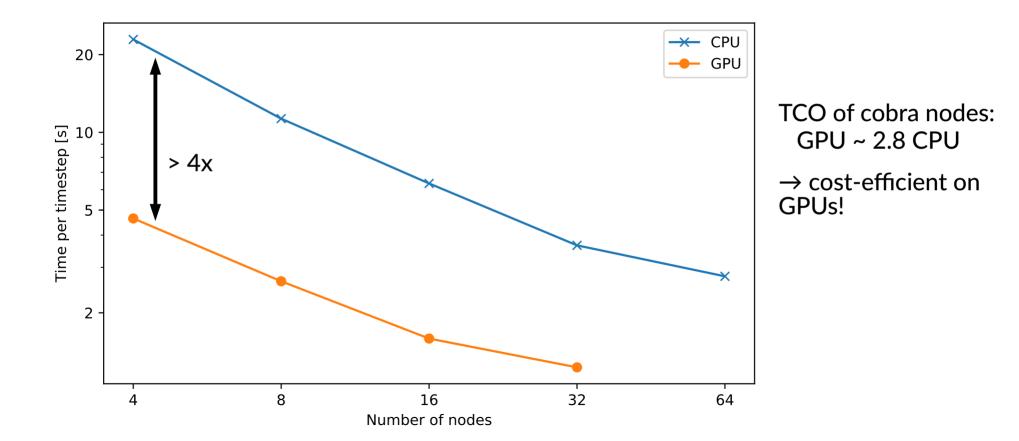
Cobra @ MPCDF



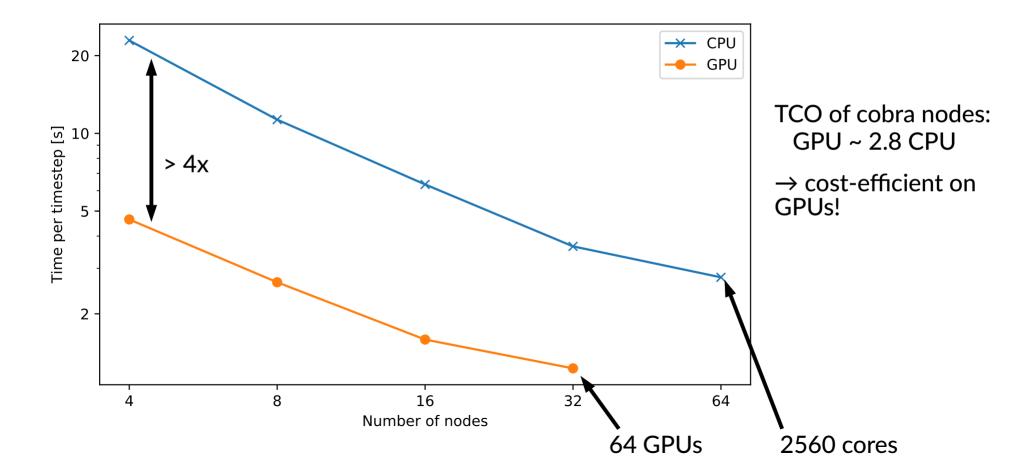
- CPU nodes: 40 cores/node (2x 20-core sockets)
- GPU nodes: CPU nodes + 2 Nvidia V100 GPUs
- Interconnect: Omnipath (100 Gbit/s)
- GPU machines @ MPSD
 - 2x 8-core sockets
 - 8 Nvidia V100 GPUs + NVLink



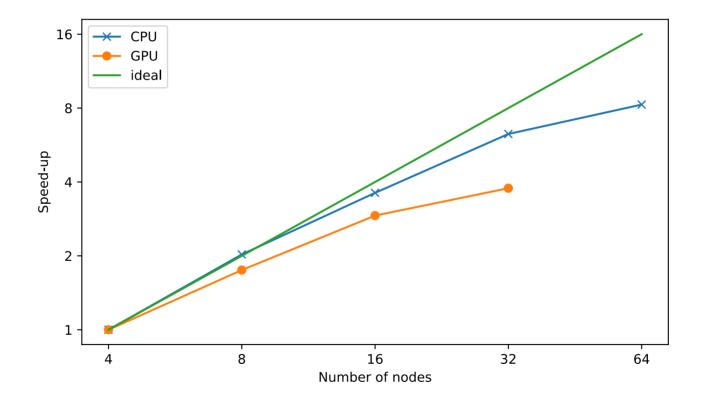
Comparison on cobra: CPU vs. GPU



Comparison on cobra: CPU vs. GPU



Scaling on cobra

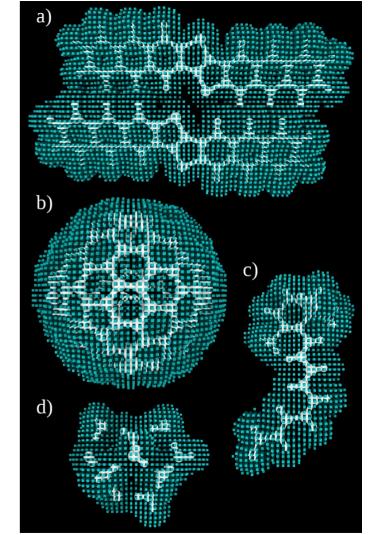


Implementation

- Pinned memory \rightarrow faster transfer speed
- Streams \rightarrow asynchronous operations
- CUDA-aware MPI \rightarrow GPU-GPU communication
- Prefetching \rightarrow overlap communication & computation

Data layout

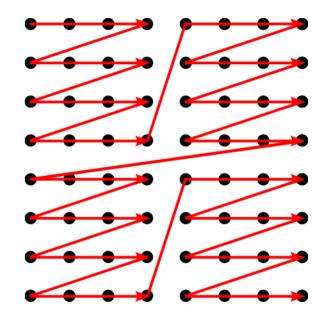
- Real-space grid for FD
- Complicated shape possible, e.g. molecules



X. Andrade & A. Aspuru-Guzik, J. Chem. Theory Comput. (2013), 9, 10, 4360-4373

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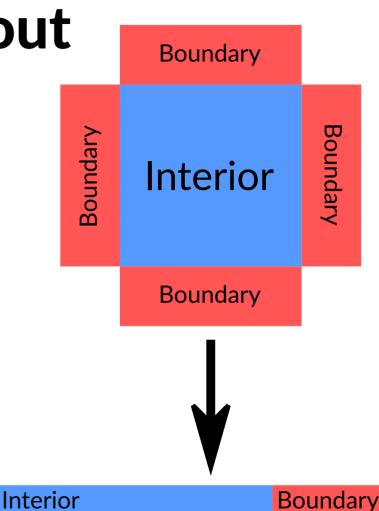
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Data layout

- Real-space grid for FD
- Complicated shape possible, e.g. molecules
- Cache-aware mapping to 1D array
- 1D data layout: 2 blocks
 - Interior points
 - Boundary/ghost points



• Timeline before (distributed mesh):



• Timeline before (distributed mesh):

Copy to CPU Operation: Inner Communication Copy to GPU Operation: Outer Gather

 With CUDA-aware MPI: communication between $GPUs \rightarrow no \text{ copies to/from } GPU$

Gather

Operation: Inner Communication (GPU-GPU) Operation: Outer

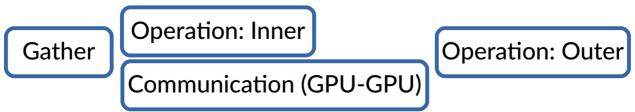
• Timeline before (distributed mesh):

Gather Copy to CPU Operation: Inner Communication Copy to GPU Operation: Outer

 With CUDA-aware MPI: communication between GPUs → no copies to/from GPU

Gather Operation: Inner Communication (GPU-GPU) Operation: Outer

CUDA-aware MPI + streams: overlap communication & computation



- Implementation:
 - Get pointers to GPU memory from C
 - Use c_f_pointer in Fortran to get a Fortran pointer to this memory
 - Use this Fortran pointer in the MPI calls
- On 8 GPUs with NVLink (machine @ MPSD)
 - Peer-to-peer transfer speed: ~24 GB/s
 - Speed-up of ~ 2.4x vs. normal MPI

Prefetching batches

• Timeline without prefetching:

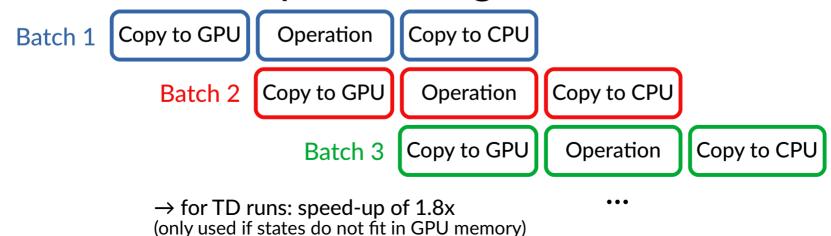


Prefetching batches

• Timeline without prefetching:



• Timeline with prefetching:



Summary



- Octopus more efficient on GPUs
- Improvements based on
 - Pinned memory
 - Streams
 - CUDA-aware MPI
 - Prefetching batches
- Outlook:
 - Port more parts of the code
 - Improve scaling

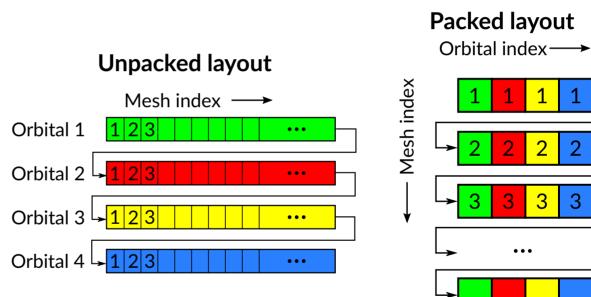


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Backup slides

Data layout II: batches

- Aggregate several orbitals into one batch
- Operations done over batches
- 2 layouts:
 - Unpacked
 - Packed \rightarrow vectorization, GPUs



3

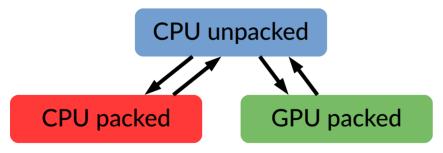
Batch handling

GPU packed

• Batch can have 3 states:



• Transitions before:



 \rightarrow always involves transposition

Batch handling

- Batch can have 3 states:
- CPU unpacked • Transitions before: CPU unpacked CPU packed CPU packed

 \rightarrow always involves transposition

 \rightarrow simple copy to GPU

Pinned memory

- Normal allocations: pageable memory
- Transfers to GPU: pinned memory needed
 → faster transfer
- Solution:
 - Allocate pinned memory in C (CUDA call)
 - Use c_f_pointer in Fortran to use this memory
- Transfer speed on PCIe 3: ~12 GB/s vs. ~5 GB/s

Streams

- Default: CUDA operations are blocking
- Streams needed to overlap operations
- Also needed for CUDA-aware MPI
- 32 Streams are initialized in the C layer
- Selection from Fortran layer
- Usage example: asynchronously launch norm kernels with strides

CUDA-aware MPI

- Extension of MPI, available for some flavours (OpenMPI, MPICH, MVAPICH, ...)
- Requires compatible low-level drivers
- Usage:
 - Pass GPU pointers to MPI calls
 - MPI library can directly access the GPU memory
- Advantages:
 - Peer-to-peer copies on the same node (even better with NVLink)
 - Less latency for inter-node communication

Overlap communication & computation

- 2 ways of running octopus on GPUs:
 - If enough GPU memory \rightarrow store all batches on GPU
 - Otherwise \rightarrow copy batch to GPU, operate, copy back
- For second way:
 - Overlap of communication & computation possible
 - Use asynchronous prefetching on different stream

Prefetching batches

- Advantage:
 - Hide copy latency, except for first & last copy
- Disadvantages:
 - Needs memory for 3 batches
 - Does not overlap completely if operation involves copies to/from the GPU
- For TD runs: speed-up of 1.8x

Timing data (Ag tip on cobra)

Nodes		4	8	16	32	64
CPU	Time [s]	22.9	11.3	6.34	3.65	2.77
	Speedup	1	2.0	3.6	6.3	8.3
GPU	Time [s]	4.64	2.65	1.59	1.23	
	Speedup	1	1.8	2.9	3.8	
GPU ParDomain=2	Time [s]	12.76	6.55	3.45	1.86	
	Speedup	1	2.0	3.7	6.9	
GPU StatesPack=no	Time [s]	5.26				

Timing data (Ag tip on MPSD machine)

Number of GPUs		1	2	4	8
GPU	Time [s]				4.44
GPU ParDomain=2	Time [s]				7.42
GPU ParDomain=2 normal MPI	Time [s]				17.9
GPU StatesPack=no	Time [s]	32.3	27.9	14.7	7.95
	Speedup	1	1.16	2.2	4.1
GPU StatesPack=no ParDomain=2	Time [s]		28	24	14
	Speedup		1	1.16	2