



Barcelona Supercomputing Center Centro Nacional de Supercomputación

CAMP First GPU Solver: A Solution to Accelerate Chemistry in Atmospheric Models

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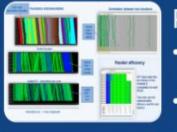
Barcelona SuperComputing Center *National Center for Atmospheric Research (NCAR)

7th ENES workshop

BSC Departments



To develop and implement global and regional state-of-the-art models for shortterm air quality forecast and long-term climate applications



Performance Team

- Provide HPC Services (profiling, code audit, ...) to find main bottlenecks of our operational models
- Research and apply new computational methods for current and new platforms

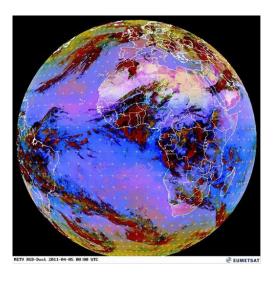


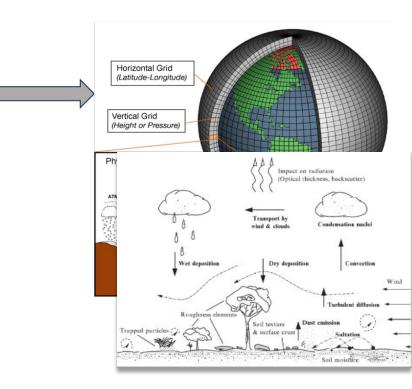
Background



Atmospheric models

Atmospheric models are a **mathematical representation** of atmospheric water, gas, and aerosol cycles.

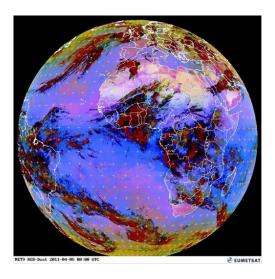




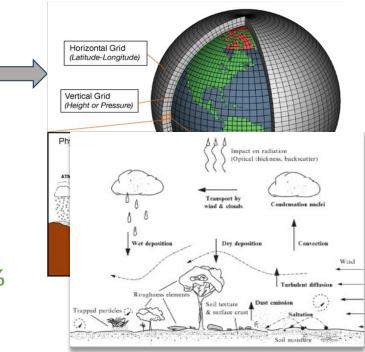


Atmospheric models

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Resolution of chemical processes can take up to 80% of the time execution!





State of the art - KPP GPU

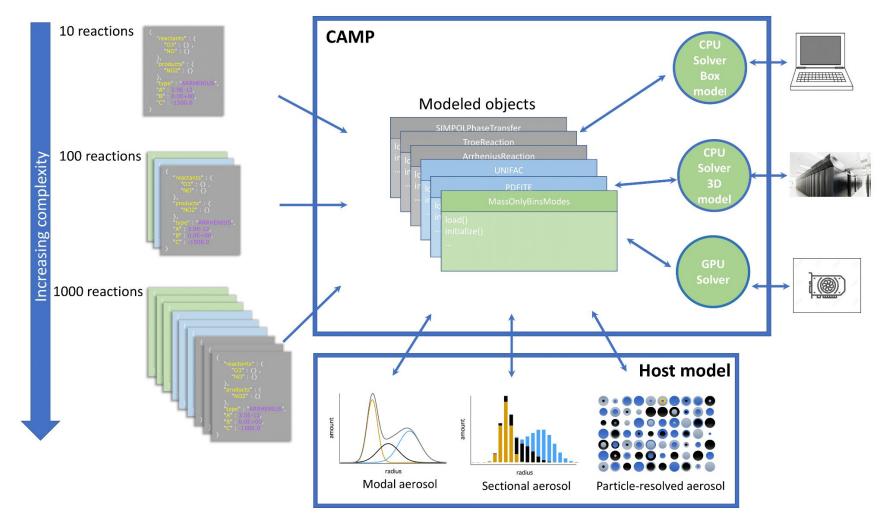
- Kinetic PreProcessor (KPP) is a analysis tool to solve chemical mechanisms using Rosenbrock methods
- KPP is widely used in the atmospheric community
- The GPU version for the EMAC climate model achieves up to 20x speedup against CPU single-core and 1.86x against 2 CPUs

Configuration		Median CPI exec tim		an acce- ted exec time (s)	Perform	over CPU
Intel Xeon X5650 + M20	4.50	2	0.999		4.50×	
Intel Xeon E5-2680 v3 +	-K80	1.47	6	0.283	-	5.21×
IBM POWER8 + P100		3.04	0	0.149	20	0.40×
Configuration	MPI I	Processes	CPU exec time (s)	Accelera exec time		rformance over CPU
2 × 6-core Intel Xeon X5650 +	2 MPI	processes	5199	2	358	2.27 ×
2 × NVIDIA M2070	12 MI	PI processes	1388	1	368	$1.01 \times$
2 × 12-core Intel E5-2680 v3 +	4 MP	processes	7362	3	384	2.17 ×
$2 \times NVIDIA K80$	24 MI	PI processes	1756	1.	473	1.19 ×
2 × 10-core IBM POWER8 +	4 MP	processes	2294		918	$2.50 \times$
4 × NVIDIA P100		PI Processes	814		437	$1.86 \times$

Michail Alvanos and Theodoros Christoudia, GPU-accelerated atmospheric chemical kinetics in the ECHAM/MESSy (EMAC) Earth system model , 2017



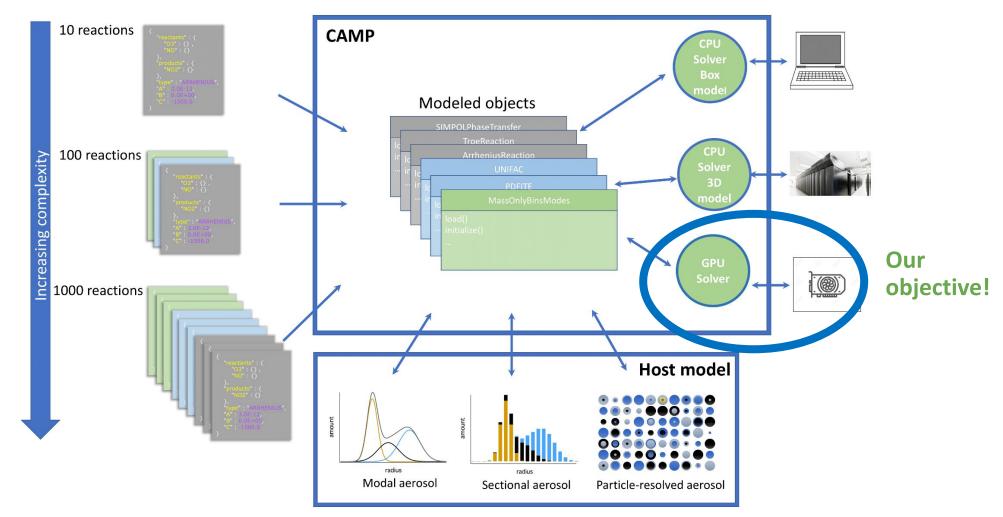
CAMP: Chemistry Across Multiple Phases



Dawson, Guzman, Curtis, Acosta, et. al., Chemistry Across Multiple Phases (CAMP) version 1.0, GMD 2022



CAMP: Chemistry Across Multiple Phases

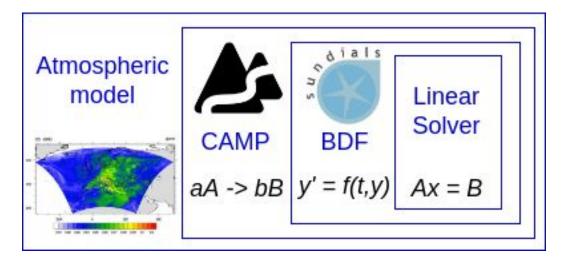


Dawson, Guzman, Curtis, Acosta, et. al., Chemistry Across Multiple Phases (CAMP) version 1.0, GMD 2022



CAMP CPU Solver

- CAMP uses the Backward Differentiation Formula (BDF) from CVODE, which is a solver for ordinary differential equation (ODE) systems.
- BDF requires a linear solver package. The default option is the KLU algorithm for the CPU execution, while it also has a CUDA version of the Biconjugate Gradient (BCG) algorithm.



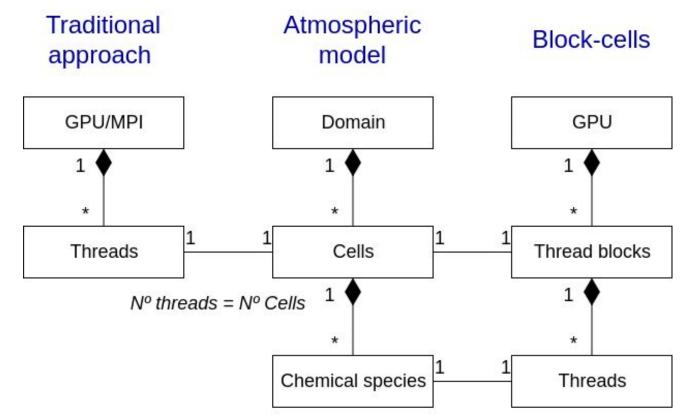


Implementation



Block-cells (GPU parallelization strategy)

- Block-cells assigns each atmospheric cell to a GPU thread block
- Uses as many threads as chemical species



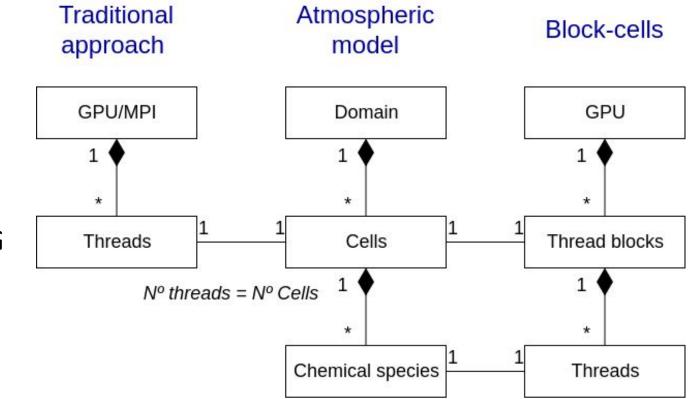
Nº threads = Nº Species

Guzman et. al. Studying a new GPU treatment for chemical modules inside CAMP, 19th ECMWF Workshop



Introduction | Background | Implementation | Test environment | Results | Conclusions Block-cells (GPU parallelization strategy)

- Higher occupancy than traditional approaches (more threads computing data)
- 34x speedup against CPU single-thread for the CAMP BCG linear solver



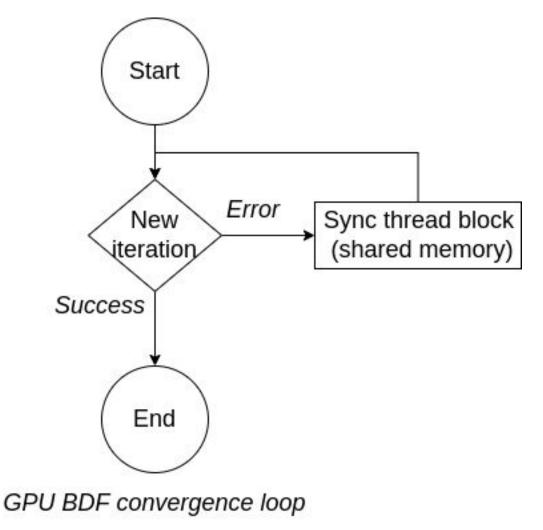
 N^{o} threads = N^{o} Species

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Communicating data between threads

- All communications are performed at thread block level
- An example: A thread triggers an error due to a negative concentration
- The error is shared between the other threads in the block by using shared memory





Test environment



Hardware

- CTE-POWER cluster:
 - 2 x IBM Power9 8335-GTH @ 2.4GHz (3.0GHz on turbo, 20 cores and 4 threads/core, total 40 cores per node)
 - 4 x GPU NVIDIA V100 (Volta) with 16GB HBM2.
 - Compilers: GCC version 6.4.0 and NVCC version 10.2



Software configuration

Architecture	Parallel resources	Parallelization language				
CPU	1, 40	MPI				
GPU	Nº of different chemical concentrations (species x cells)	CUDA				

- The evaluation is performed over the code included in the most external loop in BDF. The code related to previous initializations is excluded.
- Chemical mechanism: Gas phase chemistry from Carbon bond 2005 (CB05) | Chemical species: 156 | Cells (ODE systems): 100-10,000 | GPU Shared memory size per block: 256 | CVODE absolute tolerance: 0.01% | BCG tolerance: 1.0e-30

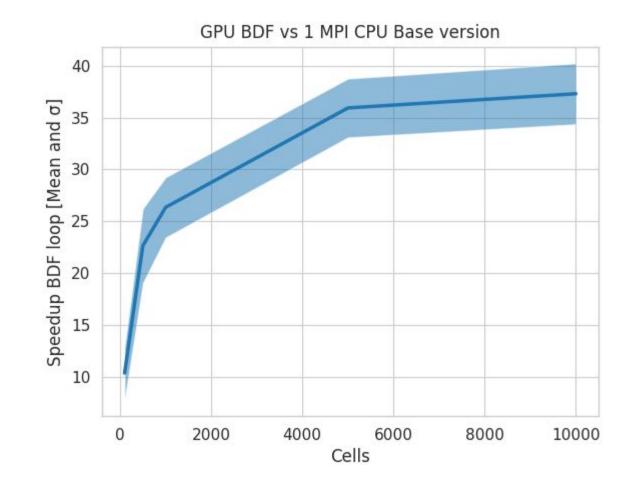


Results



Speed-up

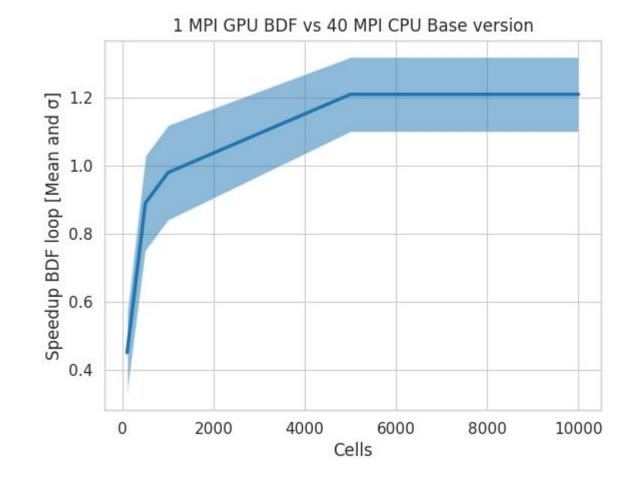
- Up to 35x speedup in average vs single-thread
- Standard deviation around 2





Speed-up against 40 processes

- 1.2x speed-up against a fully CPU node (40 MPI processes)
- Since there's no communication
 between threads, we estimate 4.8x
 speed-up using the full GPU resources
 in a node (4 GPUs) Ongoing work





Kernel profiling

- Some optimizations already performed (like adjusting the number of registers per thread)
- The register usage is likely preventing the kernel from fully utilizing the GPU
- This usage is mostly produced by the algorithm definition, which computes big data like the Jacobian matrix

Variable	Achieved	Theoretical	Device Limit	Grid S	ize: [10000),1,1]	(10000	block	s)Blo	ock Size	e: [73,1	,1](7	3 th	reads
Occupancy Per SM															
Active Blocks		10	32	0	3	6	9	12	15	18	21	24	27	30	32
Active Warps	29,67	30	64	0	6	12	18	24	30	36	42	48	54	60	64
Active Threads		960	2048	0	256	6	512	768	102	4	1280	1536	179	2	204
Occupancy	46,4%	46,9%	100%	0%			25% 50%				75% 1			100	
Warps				1											
Threads/Block		73	1024	0	128	8	256	384	51	2	640	768	896	5	102
Warps/Block		3	32	0	3	6	9	12	15	18	21	24	27	30	32
Block Limit		21	32	0	3	6	9	12	15	18	21	24	27	30	32
Registers															
Registers/Thread		64	255	0	32		64	96	12	8	160	192	224	4	255
Registers/Block		6144	65536	0			16k		32	.k		48k			64
Block Limit		10	32	0	3	6	9	12	15	18	21	24	27	30	32
Shared Memory		-10													
Shared Memory/Block		1024	98304	0		16k		32k	48	lk.	64	k	80k	_	96
Block Limit		76	32	0	2	6	9	12	15	18	21	24	27	20	32



Conclusions



Conclusions

- Our Block-cells strategy increase the GPU parallel threads against traditional implementations (Nº Threads = Nº Cells)
- The CUDA BDF loop performs up to 35x times faster than CPU single-thread
 - 1.2x speed-up against CPU using the fully resources node.
 - Since the load is independent between threads, we estimate up to 4.8x speedup using 4 GPUs per node.
- The kernel profiling suggests a limitation on the performance by memory
- Our approach can be used in more chemical applications thanks to the versatility of the CAMP module.



- Add multi-device functionality to compute up to 4 GPus per node.
- Balance load between CPU and GPU architectures.
- Integrate our implementation inside an atmospheric model.





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Thank you

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