

POP assessments with BSC tools

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Insights on a MPI study



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MPI study: background & FoA

- Name of the code: PORTA
- Programming: C; MPI (+POSIX threads)
- Scale (#cores): 141, 281, 561, 1121, 2241 (weak scaling)
- Platform: BSC MareNostrum4 (node=2 Intel Xeon Platinum 8160 24C @2.1 GHz)





Duration of the computing regions

With the small input (W1) the granularity of the main computations are in the range of 70-90 milliseconds (large computation 8 sec.)

Outside NPI
MPI_Send
MPI_Recv
MPI_Isend
MPI_Vait
MPI_Allreduce
MPI_Ssend
Communications using both
synchronous and asynchronous calls
and ranks are globally synchronized
with an MPI_Allreduce 3

Code scaling







All the timelines (right) have the scale of W5

- Up to W3 the scaling is good.
- W4 and W5 show an increase of the unbalance and a bad scaling of the large computing phase.
- Large computation (orange regions) have severe scaling problems



Efficiency model analysis



Global efficiency

- -- Parallel efficiency
 - -- Load balance
 - -- Communication efficiency
 - -- Serialization efficiency
 - -- Transfer efficiency
- -- Computation scalability
 - -- IPC scalability
 - -- Instruction scalability
 - -- Frequency scalability

_	141	281	561	1121	2241	100%
	91.42	83.74	77.62	66.02	53.60	100%
	91.42	88.35	85.46	81.02	72.48	90%
	92.47	90.34	89.81	91.28	86.62	
у	98.87	97.79	95.16	88.76	83.67	70%
ncy	99.23	98.04	95.44	89.40	84.54	
	99.64	99.75	99.70	99.29	98.97	< 50%
	100.00	94.78	90.82	81.48	73.96	
	100.00	102.30	100.99	100.22	103.10	
	100.00	92.64	88.15	81.27	71.76	
	100.00	100.02	102.03	100.04	99.96	
L						

- The scaling of the code is limited by:
 - An increase in the total number of instructions (code replication).
 - Serialization and dependencies.
- The main degradation for W1 is due to global load balance despite the efficiency is still very good.
- Efficiencies lower than 80% indicate space for improvement. Lower than 60% there is a clear need for improvement.
- The average IPC is 2.03 that is a good value for MN4 where frequently is limited to 1.2-1.5 $_{5}$



Computations scaling





- Tracking the evolution of the clusters we can see that:
 - Clusters 1 and 2 only increase the number of instructions (per instance) for the configurations that increase number of frequencies per rank.
 - The number of grid points increases the number of invocations of the clusters
 - The variability on instructions for cluster 1 increases drastically in W5.



MPI waiting time



 Most of the MPI waiting time for MPI_Recv and MPI_Allreduce is concentrated in 3 regions





Improved version (few months later)



FOA first Audit

144,867,61



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The improved code reduces 7% the iteration (smallest scale).

- Refactoring large computation
- Reduce code replication

Comparing the scaling efficiency and elapsed time with respect to the previous version

- The scaling (left) has been improved in average an 8% that goes up to 16% with W5
- The iteration time (right) has been reduced in average a 12% (close to 20% with W5)



Efficiency model analysis



Global efficiency

- -- Parallel efficiency
 - -- Load balance
 - -- Communication efficiency
- -- Computation scalability
 - -- IPC scalability
 - -- Instruction scalability
 - -- Frequency scalability

	141	281	561	1121	2241	100%
	92.04	89.81	81.69	71.01	62.61	90%
	92.04	89.23	86.07	83.35	73.38	80%
	92.84	91.03	89.91	90.19	82.45	70%
cy	99.14	98.03	95.73	92.42	88.99	60%
	100.00	100.64	94.91	85.19	85.33	< 50%
	100.00	100.46	100.19	98.68	98.85	
	100.00	100.18	94.70	86.30	86.39	
	100.00	100.01	100.03	100.03	99.92	

- The scaling of the code has been improved significantly by **reducing the code replication** (W5: instruction scaling from 71.76 to 86.39)
- The parallel efficiency reports very similar values (W5: 72.48 vs. 73.38)
 - Comm. eff. improves (83.67 → 88.99)
 - LB eff degrades (86.62 → 82.45)
- Efficiencies lower than 80% indicate space for improvement. Lower than 60% there is a clear need for improvement.
- The average IPC is 1.96 (previously 2.03)





Insights on an OpenMP study



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OpenMP study: background and FoA

- Name of the code: Explorer
- Programming: C++; MPI, OpenMP
- Scale (#cores): 1 MPI rank scaling OpenMP (4, 24, 48)
- Platform: BSC MareNostrum4 (node=2 Intel Xeon Platinum 8160 24C @2.1 GHz)

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Code scaling





The OpenMP scaling is poor. With 24 processes the efficiency is 49% (2.95x with 6x more resources). Doubling to 48 there is a very small time reduction (efficiency is only 28% of an ideal scaling).



Efficiency model analysis



Global efficiency

- -- Parallel efficiency
 - -- Load balance
 - -- Communication efficiency
- -- Computation scalability
 - -- IPC scalability
 - -- Instruction scalability
 - -- Frequency scalability

	4(1x4)[1]	24(1x24)[2]	48(1x48)[3]		- 100
-	72.82	35.82	20.47		100
-	72.82	66.93	63.56		- 80
-	96.34	85.55	81.95		
-	75.59	78.23	77.56		- 60
-	100.00	53.52	32.21		- 40
-	100.00	53.71	32.65		
-	100.00	99.88	99.65		- 20
-	100.00	99.76	99.00		0
				_	-0

- The scaling of the code is limited mainly by:
 - A drastic reduction of the IPC
 - An increase of the load unbalance
- The main degradation with 4 threads is due to time inside the **OpenMP runtime** (Comm. Eff.)
- Efficiencies lower than 80% indicate space for improvement. Lower than 60% there is a clear need for improvement.

Percentage(%)

• The average IPC is 1.14 with 4 threads and it significantly decreases. That seems to indicate the problem can be related with the accesses to the shared memory and/or with the frequent calls to the OpenMP runtime.



IPC degradation analysis





(left) All timelines are in the same color scale. The timelines show that **the IPC reduction is on the computation of almost all the phases**.

(right) Only the pink and red parallel functions maintain an acceptable IPC. The brown one has a very low IPC for all the runs but also suffers the biggest reduction with the scale.

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	Interact:compute	Interact:compute	Interact:compute	CachedPauteCache	CachedPauteCache	Interact:compute
Total	5.29	4.96	4.91	4.47	3.75	3.50
Average	1.32	1.24	1.23	1.12	0.94	0.88
Maximum	1.36	1.27	1.25	1.13	0.95	0.90
Minimum	1.29	1.22	1.21	1.10	0.93	0.86
StDev	0.03	0.02	0.02	0.01	0.01	0.01
Avg/Max	0.97	0.98	0.98	0.99	0.99	0.98

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Interact:compute	CachedPauteCache	Interact:compute	Interact:compute	CachedPauteCache	Interact:compute
26.31	23.66	22.89	22.86	19.85	5.47
1.10	0.99	0.95	0.95	0.83	0.23
1.15	1.01	1.00	1.01	0.84	0.29
1.05	0.85	0.90	0.90	0.80	0.21
0.03	0.03	0.03	0.03	0.01	0.02
0.95	0.98	0.95	0.94	0.98	0.78
	Interact:compute 26.31 1.10 1.15 1.05 0.03 0.95	Interactcompute CachedPauteCache 26.31 23.66 1.10 0.99 1.15 1.01 1.05 0.85 1.04 0.03 1.05 0.03 1.05 0.99	Interactcompute CachedPauteCache Interactcompute 26.31 23.66 22.89 1.10 20.09 0.05 1.115 1.01 1.01 1.105 0.085 0.090 1.105 0.085 0.090 1.105 0.085 0.090 1.105 0.085 0.090	Interactcompute CachedPauteCache Interactcompute Interactcompute ACAD ACAD ACAD ACAD ACAD ACAD ACAD ACAD	InteractcomputeCachedPauteCacheInteractcomputeCachedPauteCache100023.6022.8020.8019.85101010.00920.00520.00519.80101010.00920.00520.00520.008101010.01010.01010.01010.010101010.01020.01020.01020.010101010.01020.01020.01010.010101010.01010.01010.01010.010

CachedPa..uteCache

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	Interact:compute	CachedPauteCache	Interact:compute	Interact:compute	CachedPauteCache	Interact:compute
Total	37.94	36.74	31.66	31.56	28.14	5.80
Average	0.79	0.77	0.66	0.66	0.59	0.12
Maximum	0.90	0.82	0.70	0.82	0.62	0.21
Minimum	0.74	0.71	0.59	0.58	0.56	0.11
StDev	0.03	0.02	0.02	0.03	0.02	0.01
Avg/Max	0.88	0.93	0.95	0.80	0.95	0.56



Parallel functions computations



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	Interact:compute	Interact:compute	Interact:compute	CachedPauteCache	Interact:compute	CachedPauteCache
Total	2.82	2.80	3.28	3.00	2.78	2.90
Average	0.71	0.70	0.82	0.75	0.70	0.72
Maximum	0.71	0.71	0.84	0.77	0.74	0.80
Minimum	0.70	0.69	0.81	0.74	0.67	0.70
StDev	0.00	0.01	0.01	0.01	0.03	0.04
Avg/Max	1.00	0.99	0.98	0.98	0.93	0.91

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	Interact:compute	Interact:compute	Interact:compute	CachedPauteCache	Interact:compute	CachedPauteCache
Total	14.48	21.74	13.91	15.00	11.38	11.35
Average	0.60	0.91	0.58	0.62	0.47	0.47
Maximum	0.61	0.95	0.63	0.72	0.72	0.81
Minimum	0.59	0.90	0.57	0.62	0.46	0.46
StDev	0.01	0.01	0.01	0.02	0.05	0.07
Avg/Max	0.98	0.95	0.92	0.87	0.66	0.59

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Interact:compute	Interact:compute	Interact:compute	CachedPauteCache	Interact:compute	CachedPauteCache
26.29	43.74	24.76	25.28	18.77	17.49
0.55	0.91	0.52	0.53	0.39	0.36
0.56	0.97	0.60	0.64	0.69	0.78
0.54	0.91	0.50	0.51	0.38	0.35
0.00	0.01	0.01	0.02	0.04	0.06
0.97	0.93	0.86	0.82	0.56	0.47
	Interact:compute 26.29 0.55 0.56 0.54 0.00 0.07	Interact:compute Interact:compute 1000000000000000000000000000000000000	Interact:compute Interact:compute Interact:compute Interact:compute 2629 4374 0.055 0.091 0.056 0.097 0.057 0.091 0.058 0.091 0.059 0.091 0.050 0.091 0.051 0.001 0.052 0.011	Interact:compute Interact:compute CachedPauteCache 1000000000000000000000000000000000000	InteractcomputeInteractcomputeCachedPauteCacheInteractcomputeInteractcom

The average metric is pointing that the application has **a very fine granularity**.

The average row reflects the %time outside the OpenMP runtime

- →Values lower than 0.7 indicate a high OpenMP overhead.
- →With 24 and 48 threads all functions have a high OpenMP overhead (the brown one has a high value because computations do not scale).

The Avg/Max reflects the load balance

 \rightarrow Values lower than 0.8 indicate the balance need to be improved.

→ The unbalance with 24 and 48 is concentrated the pink and red parallel functions.



Efficiency model – OpenMP sched.

static

- 100

- 80

- 60

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ercentage(%)

 The application was using the default scheduling (dynamic with chunk size = 1)

dynamic 100 dynamic 1000 guided 1000

Global efficiency

- -- Parallel efficiency
- -- Load balance
- -- Communication efficiency
- -- Computation scalability
- -- IPC scalability
- -- Instruction scalability
- -- Frequency scalability

-	79.26	81.50	80.02	77.19					
-	79.26	78.06	77.28	74.06					
-	79.73	78.74	78.59	78.88					
-	99.41	99.14	98.34	93.89					
-	100.00	104.41	103.54	104.23					
-	100.00	104.18	103.18	103.98					
-	100.00	100.32	100.41	100.31					
-	100.00	99.90	99.94	99.93					

All configurations (24 threads) are very similar and improve the default (parallel efficiency = 66.9%)

• The average IPC is in the range 0.69 - 0.72. Being still low, but an improvement w.r.t. the default scheduling (0.61)











Insights on a MPI + CUDA study



MPI + CUDA: background and FoA



- Name of the code: Tsunami-HySEA
- Programming: C++; MPI, CUDA
- Scale: from 1 MPI rank + 1 GPU to 64 MPI ranks + 64 GPUs
- Platform: BSC CTE Power (first 2 studies) and Leonardo (last study)



300 iterations

Zoom in few iterations

Scalability



• Time scalability of the FOA with respect to linear scaling



The execution with 1 MPI + 1 GPU was discarded for the analysis because a significant different behaviour of the MPI task.

With 16 MPI + 16 CUDA the speed-up is 64% of lineal scaling.

With 64 MPI + 64 CUDA the speed-up goes down to 35%.





	4	8	16	24	32	48	64
Global efficiency	65.23	55.25	41.90	36.41	32.05	26.52	23.40
Parallel efficiency	65.23	64.32	48.9	44.79	41.7	36.53	33.94
Load Balance	72.57	77.79	67.41	67.86	68.15	64.92	64.67
Communication eff.	89.88	82.68	72.54	66.01	61.19	56.27	52.47
Computation scalability	100.00	85.89	85.69	81.30	76.87	72.60	68.94

- Mainly communications but also computations limit the code scalability
- The reference run with 4 MPI + 4 GPUs already reports poor efficiency
- Load balance is poor in all the configurations but there is limited degradation with the scale
- No counters available at GPUs to compute IPC and Instructions scaling factors



• Big impact from MPI and CUDA components.

• MPI flat behaviour except with 8 ranks



Efficiency model



Load balance and data transfer







Parallel efficiency 65%, Load Balance 76%, Comm. 85%

With instantaneous communications most of the point to point calls disappear and the allreduce increases a bit \rightarrow data transferred in point to point is limited by network resources





Code scaling (v1 vs. v2)







CUDA kernel

- The scaling is improved an average of 30% (40% for the runs of at least 4 nodes).
- Main improvements (previous audit suggestions):
 - Overlap MPI point to point calls with the kernels execution.
 - Improve balance between MPI ranks





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Efficiency model (all resources vs gpus)

	24	48	96	144	192	288	384	100
Global efficiency	44.72	41.83	37.86	34.28	30.85	27.14	22.62	- 100
Computation scalability	100.00	89.85	95.37	92.46	83.47	84.57	79.80	- 80
Hybrid Parallel efficiency	44.72	46.56	39.70	37.07	36.96	32.10	28.35	
Load balance	48.68	52.89	48.39	48.73	52.89	49.97	50.84	- 60
MPI contribution LB eff.	96.41	95.78	95.23	95.08	94.34	94.22	89.63	
CUDA contribution LB eff.	50.50	55.22	50.82	51.25	56.06	53.04	56.72	- 40
Communication efficiency	91.86	88.02	82.03	76.07	69.88	64.23	55.76	
MPI contribution comm. eff.	99.09	96.12	92.13	88.05	83.93	80.92	77.56	- 20
CUDA contribution comm. eff.	92.71	91.58	89.03	86.39	83.26	79.38	71.89	
	24	48	96	144	192	288	384	- 100
GPUs only Global efficiency	89.07	83.30	75.40	68.27	61.45	54.06	45.05	100
GPUs only Computation scalability	100.00	98.77	96.56	94.37	92.22	88.15	84.21	- 80
GPUs only Parallel efficiency -	89.07	84.34	78.09	72.34	66.63	61.33	53.50	- 60
GPUs only Load Balance	96.96	95.83	95.19	95.09	95.34	95.48	95.94	- 40
GPUs only Communication eff.	91.86	88.02	82.03	76.07	69.88	64.23	55.76	- 20



Programing model contribution





- **Bigger impact from CUDA component** with a similar contribution in all the scales
- MPI contribution increases with the scale, but still lower than CUDA



Quick validation of a bottleneck



• Discussing the analysis with the code developers allowed to identify a **performance problem with a CUDA reduction phase**.

<u>X</u> ×	CUDA call @ TsunamiHySEA.16N.chop1.clustered.prv	✓ ^ ⑧ X	CUDA call @ TsunamiHySEA.mod.16N.prv.gz	~ ^ 🕲
THREAD 1.1.1		THREAD 1.1.1		
CUDA-D3.Sd-p9r2nD9		CUDA-D3.54-p9+2n01		
CUDA-D2.52-p9r2n12		CUDA-D2.52-p9r2n88		
THREAD 1.33.1		THREAD 1.55.1		
CUDA-D3.S4-p9r2n18		CUDA-D3.S4-p9r2n13		- <u>- a.</u> 5 <u>a.</u> 5.
CUDA-D2.S2-p9r3n18		CUDA-D2.S2-psr2n1s		
CUDA-D4.S5-p9r3n13 176,899,0	016 mc	LEL 425,557 ns CUDA-D4.55-p9r2n18 38,245,449,34	4 #5	38,250,776,485 as
X ×	CUDA kernel @ TsunamiHySEA.16N.chop1.clustered.prv	✓ ^ ⑧ X	CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz	~ ^ 🔕
N X	CUDA kernel @ TsunamiHySEA.16N.chop1.clustered.prv	· · · · · · · · · · · · · · · · · · ·	CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz	~ ^ &
X // THREAD 1.1.1 CUDA-D3.54-p9r2n09	CUDA kernel @ TsunamiHySEA.16N.chop1.clustered.prv	V A S X X THREAD 1.1.1 CUDA-D3.54-p9r2n01	CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz	
X X THREAD 1.1.1 CUDA-D3.54-p9r2n09 CUDA-D3.52-p9r2n19	CUDA kernel @ TsunamiHySEA.16N.chop1.clustered.prv		CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz	
N X THREAD 1.1.1 CUUDA-D3.54-p9r2n09 CUUDA-D3.52-p9r2n13 THREAD 1.33.1	CUDA kernel @ TsunamiHySEA.16N.chop1.clustered.prv	X X THREAD 1.1.1 CUDA-D3.54-p9r2n91 CUDA-D2.53-p9r2n91 THREAD 1.3.1	CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz	
N N THREAD 1.3.1 0 CUDA-D3.54-p9r2n09 0 CUDA-D2.52-p9r2n12 0 THREAD 1.33.1 0 CUDA-D3.54-p9r2n13 0	CUDA kernel @ TsunamiHySEA-16N.chop1.clustered.prv	THREAD 1.1.1 CUDA-D3.54-p9r2n01 CUDA-D2.53-p9r2n01 THREAD 1.33.1 CUDA-D5.54-p9r2n08	CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz	
X X THREAD 1.1.1 CUDA-D3.54-ppr2h09 CUDA-D2.52-ppr2h12 THREAD 1.33.1 CUDA-D3.54-ppr2h18 CUDA-D3.54-ppr2h18	CUDA kernel @ TsunamiHySEA.16N.chop1.clustered.prv	Image: Second state Image: Second state Image: Second state Image: Second state <th>CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz</th> <th></th>	CUDA kernel @ TsunamiHySEA.mod.16N.prv.gz	

Audited version

Modified version

FOA of 2023 version

Useful Duration @ trace_Tsunami-HySEA_04.chop1.prv



Computations

MPI calls





CUDA kernel

CUDA	kernel	l@trace_Tsunami-HySEA_04.chop1.prv
CUDA	-D1.S2·	
CUDA	-D2.S2	
CUDA	-D3.S2·	
CUDA	-D4.S2	
		0 ns 1,610,220,703 ns

Focusing on 30 iterations and 4 GPUs we can already see:

- the weight of MPI is quite small (less than 1%)
- CPUs spend most of the time in the cudaMemcpy call (over 99%)
- GPUs are executing kernels most of the time (94%)





Scalability



• Time scalability of the FOA with respect to linear scaling



The application reports a good scaling up to 64 GPUs. In fact without instrumentation the user reports a little bit better scaling that is included in the plot and goes to 57.4 with 64 GPUs. With 16 GPUs the speed-up is 96% of lineal

scale of 64 GPUs the efficiency is 83%.

As reference of the code improvement, in the previous POP assessment in Nov'20 (for a different input case and configuration) the scaling efficiency with 64 GPUs was just over 50%.



Efficiency model



	1	4	8	16	24	32	40	48	56	64	100
Global Parallel efficiency	- 47.43	47.19	46.80	46.27	45.73	44.48	43.61	43.30	42.34	42.04	- 100
Global Load balance	- 50.00	49.91	49.97	49.91	49.92	49.83	49.59	49.64	49.59	49.40	- 80
GPUs Global efficiency	- 94.85	94.08	92.94	91.03	89.33	85.99	83.83	82.77	80.17	79.00	- 60 e
GPUs Parallel efficiency	- 94.85	94.33	93.53	92.39	91.23	88.67	86.87	86.18	84.19	83.52	ntag
GPUs Load Balance	- 100.00	99.79	99.86	99.65	99.59	99.33	98.77	98.80	98.62	98.15	- 40 - 40
GPUs Communication efficiency	- 94.85	94.53	93.66	92.71	91.61	89.27	87.95	87.23	85.37	85.09	- 20
GPUs Computation scalability	- 100.00	99.73	99.37	98.53	97.92	96.98	96.50	96.05	95.22	94.59	0

- Based on the low contribution of the CPUs, the efficiency analysis is focused on the GPUs including the global parallel efficiency and load balance as reference.
- Considering all the resources allocated, the efficiencies are around 40-50% because the work is concentrated in the GPUs (reported as unbalance). The small degradation in the parallel efficiency is also observed in the GPUs metrics.
- The GPUs global efficiency reports a degradation with the scale that is related mainly with the communication efficiency but also with the computation scalability. The load balance between GPUs is ok for all the scales.
- No counters available at GPUs to compute IPC and Instructions scaling factors.



GPU idling time vs. CPU activity



• Comparing 4 GPUs (up) and 64 GPUs (down)

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	Memory transfer	Configuring accelerator	Wait/WaitAll	Group Communication	Running	Immediate Send	Immediate Receive
Total	20.8905 %	0.6158 %	0.5416 %	0.4034 %	0.1522 %	0.0483 %	0.0122 %
Average	5.2226 %	0.1540 %	0.1354 %	0.1008 %	0.0380 %	0.0121 %	0.0031 %
Maximum	5.7037 %	0.1547 %	0.1422 %	0.1608 %	0.0414 %	0.0154 %	0.0037 %
Minimum	4.9679 %	0.1532 %	0.1284 %	0.0182 %	0.0345 %	0.0089 %	0.0024 %
StDev	0.2860 %	0.0007 %	0.0066 %	0.0561 %	0.0024 %	0.0032 %	0.0006 %
Avg/Max	0.9156	0.9952	0.9520	0.6271	0.9189	0.7828	0.8175

CUDA kerr	nel@tr	ace_Tsu	nami-H	ySEA_04	t.chop	1.prv	#1						
TASK 1.1													
TASK 1.2													
TASK 1.3													
TASK 1.4													
	321.881.9	983 ns									429.	495.8	29 ns
by state	@ trace	Tsunam	i-HySE/	A_04.ch	op1.pr	'V #1							
TASK 1.1													
TASK 1.2													

IC ID 30 🔍 🗮 H H H II 💥 Σ ½ ⊾ Average ∨ 💈 🛠

	Memory transfer	Group Communication	Wait/WaitAll	Configuring accelerator	Running	Immediate Send	Immediate Receive
Total	566.3759 %	264.3879 %	136.0887 %	55.1044 %	17.5300 %	9.5718 %	4.4942 %
Average	8.8496 %	4.1311 %	2.1264 %	0.8610 %	0.2739 %	0.1496 %	0.0702 %
Maximum	9.3694 %	5.1927 %	3.2838 %	1.8026 %	0.4432 %	0.2971 %	0.1390 %
Minimum	8.6255 %	2.0512 %	0.9941 %	0.5623 %	0.1805 %	0.0775 %	0.0299 %
StDev	0.1311 %	0.6532 %	0.7086 %	0.2559 %	0.0676 %	0.0519 %	0.0269 %
Avg/Max	0.9445	0.7956	0.6475	0.4777	0.6180	0.5034	0.5050
Memory trar	nsfer						

CUDA kernel	@ trace_	Tsunami-	HySEA_64	.chop1.	prv								
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TASK 1.49				1.1	<u> </u>	11		11	8			ž.	11
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When scaling the biggest increases are related with cudaMemcpy and MPI_Allreduce. The MPI_Allreduce has a better balance with 64 MPI ranks \rightarrow increase due to date transfer



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CUDA Kernels – scaling efficiency



• Top 11 kernels (>=4% execution time with 64 GPUs)



Eliminating the 4 kernels mentioned in the previous slide we can see that 10 kernels have a very good scaling efficiency between 0.94 and 0.98

Only the last kernel in the list that represents a 4.89% has a bigger degradation but still is 0.87 with 64 GPUs.



CUDA Kernels – load balance

0,8



• 64 GPUs



A load balance bigger than 0.8 can be considered good. There are 3 kernels with a load balance lower than 0.8.

The lowest load balance is 0.7 for *obtenerEstadoYDeltaTVolumenesNivelGPU* that represents 9% of the execution time. The other 2 kernels with low value represent less than 5% each.





Performance Optimisation and Productivity 3

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