

Center for Information Services and High Performance Computing (ZIH)

### Scalable High-Quality 1D Partitioning

### 2014 International Conference on High Performance Computing & Simulation (HPCS 2014)

21 – 25 July 2014, Bologna, Italy

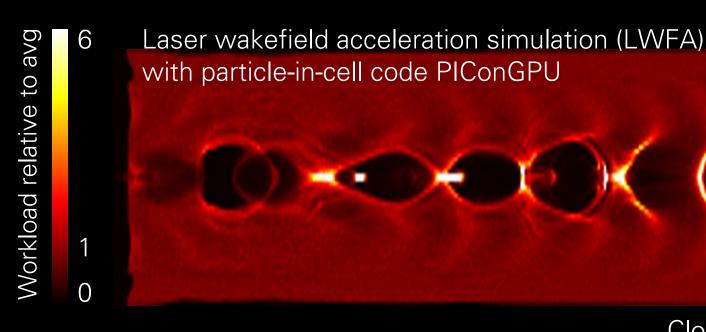
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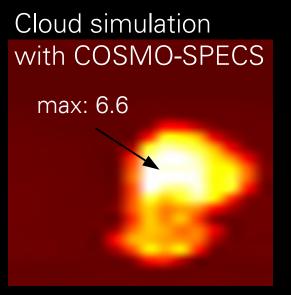
# Load balance is a major challenge for scalable HPC applications, especially if the workload changes dynamically.



Sources of workload variations:

- Adaptive spacial grids (e.g. AMR)
- Adaptive time stepping techniques
- Complex models for physical phenomena

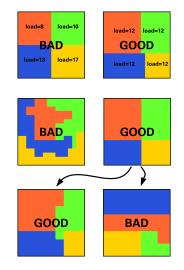
Common solution: dynamic load balancing - Requires (frequent) repartitioning of domain



max: 91.7

#### Repartitioning

- Four objectives of repartitioning algorithm
  - Balance workload
  - Reduce communication between partitions (due to data dependencies)
  - Reduce migration, i.e. communication when changing the partitioning
  - Compute partitioning as fast as possible
- Contradictory goals
- Existing methods (heuristics) provide different trade-offs between the four objectives
  - Bisection methods, space-filling curves, graph methods, diffusion methods, ...



Teresco, Devine, Flaherty, *Partitioning* and Dynamic Load Balancing for the Numerical Solution of Partial Differential Equations, LNCSE, vol. 51, pp. 55-88, 2006.

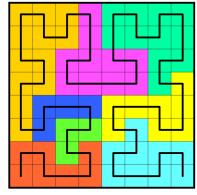




#### From SFC Partitioning to 1D Partitioning

- Space-filling curve (SFC) partitioning widely used
  - nD space is mapped to 1D by SFC
  - Mapping is fast and has high locality
  - Migration typically between neighbor ranks
- ID partitioning is core problem of SFC partitioning
  - Decomposes task chain into consecutive parts
- Two classes of existing 1D partitioning algorithms:
  - Heuristics: fast, parallel, no optimal solution
  - Exact methods: slow, serial, but optimal

Dynamic applications need fast, parallel, and high-quality methods to master Exascale!



Hilbert SFC

Pilkington, Baden, Dynamic partitioning of non-uniform structured workloads with spacefilling curves, IEEE T. Parall. Distr., vol. 7, no. 3, pp. 288-300, 1996.

Pinar, Aykanat, *Fast* optimal load balancing algorithms for 1D partitioning, J. Parallel Distr. Com., vol. 64, no. 8, pp. 974-996, 2004.





#### **Overview**

- ID Partitioning Problem Definition
- State of the Art 1D Partitioning
- Improving an Exact Method
- Scalable High-Quality 1D Partitioning
- Conclusion





#### **1D Partitioning Problem Definition**

Decompose a vector wi of N positive task weights into P consecutive partitions while minimizing the maximum load (i.e. bottleneck B) among the partitions.

Example for N=16 tasks and P=4 partitions:

Task weight vector wi (i=1,2, ..., N):

1 1 1 1 1 1 1 1 1 1 1 1 <mark>5</mark> 1 <mark>3</mark>

Optimal solution:

Load Lp of the partitions:

- Maximum load / Bottleneck B = max(Lp) = 6
- Load balance  $\Lambda_{opt} = (\sum w_i / P) / B_{opt} = 0.92$  ( $\Lambda = 1$  would be perfect)



#### State of the Art 1D Partitioning: Heuristics

- Recursive bisection commonly used
- Heuristic H2 by Miguet and Pierson better to parallelize
- Example for H2 with N=16, P=4, task weights wi:

1 1 1 1 1 1 1 1 1 1 1 1 1 <mark>5</mark> 1 <mark>3</mark>

(1) Prefix sum  $W_j = \sum_{i=1}^{j} w_i$ 

1 2 3 4 5 6 7 8 9 10 11 12 13 <mark>18</mark> 19 <mark>22</mark>

(2) First task of partition p: min s with  $W_s > p$  (WN / P)

(3) Increment s if Ws is closer to p (WN / P) than W(s-1)

- Bottleneck B = max(Lp) = 7
- Load balance  $\Lambda$  = (WN / P) / B = 5.5 / 7 = 0.79





Recursive Bisection Heuristic: Oden, Patra, Feng, *Domain Decomposition for Adaptive hp Finite Element Methods*, Contemp. Math., vol. 180, 1994.

Method described here: Miguet, Pierson, *Heuristics for 1D rectilinear partitioning as a low cost and high quality answer to dynamic load balancing*, LNCS, vol. 1225, 1997, pp. 550-564.

#### State of the Art 1D Partitioning: Exact Methods

- Various exact methods have been proposed
- Fastest method by Pinar and Aykanat: ExactBS
- Based on binary search for the optimal bottleneck Bopt
- Search Interval:  $W_N / P \le B_{opt} \le B_{Heuristic}$
- Requires probing whether a partition exists for given B
  - Binary search on Wj for each P: O(P log N)
  - Han et al.: O(P log (N/P))
  - Pinar and Aykanat: O(P log P + P log(wmax/wavg))
- Problems
  - Slower than heuristics
  - More fatal: serial only, one process needs to collect task weights from all other processes



Extensive Overview: Pinar, Aykanat, *Fast* optimal load balancing algorithms for 1D partitioning, J. Parallel Distr. Com., vol. 64, no. 8, pp. 974-996, 2004.

Han, Narahari, Choi, *Mapping a chain task to chained processors*, Inform. Process. Lett., vol. 44, no. 3, pp. 141-148, 1992.

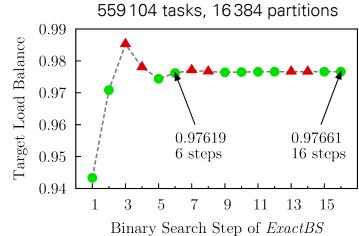


#### Improving an Exact Method: ExactBS $\rightarrow$ QBS $\rightarrow$ QBS\*

(1) New probe algorithm, tuned for small N / P

- Guess that consecutive partitions have same number of tasks, if not, do linear (!) search on W<sub>j</sub> starting from guessed position
- (2) QBS: Quality-Assuring Bisection Algorithm
  - Stop binary search at given quality q
  - Guarantees load balance  $\Lambda = q \Lambda_{opt}$
  - Reduces number of search steps
  - Exact method for q = 1
- (3) QBS\*: Parallelization of binary search
  - Search interval for Bopt is split between processes
  - Reduces number of search steps
  - Downside: all processes need task weights of all processes



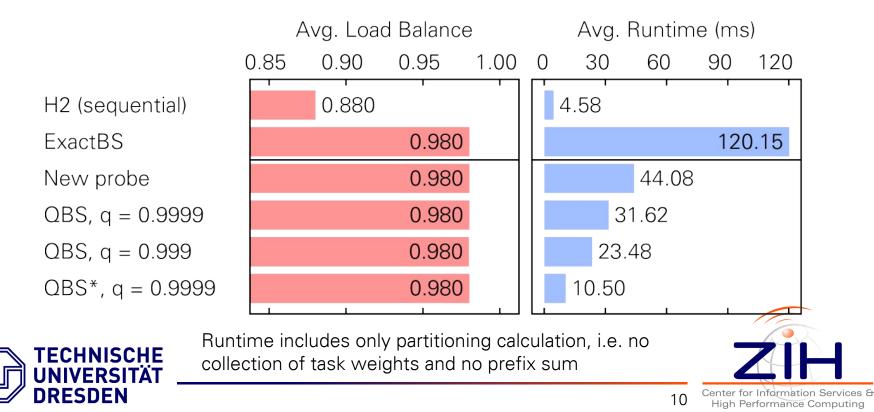




#### Improving an Exact Method: Comparison Benchmark

- Comparison to existing methods H2 and ExactBS
- Averages over 100 successive task weight vectors from the cloud simulation
- System: JUQUEEN, IBM Blue Gene/Q
- 559 104 tasks, 16 384 partitions / MPI ranks

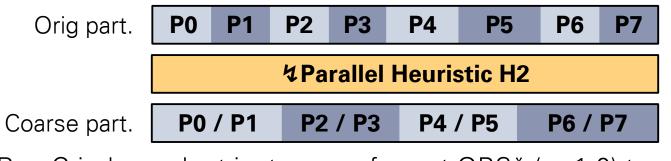




#### Scalable High-Quality 1D Partitioning: Algorithm HIER\*

## Large scale applications require a fully parallel method, i.e. without gathering all task weights

Run parallel H2 to create G < P coarse partitions:</p>



Run G independent instances of exact QBS\* (q=1.0) to create final partitions within each group:

|             | <b>ቱOBS</b> * |    | なQBS*      |           | 40BS*     |    | なQBS* |    |
|-------------|---------------|----|------------|-----------|-----------|----|-------|----|
| Final part. | <b>P0</b>     | P1 | <b>P</b> 2 | <b>P3</b> | <b>P4</b> | P5 | P6    | P7 |

 Parameter G allows trade-off between scalability (high G → heuristic dominates) and load balance (small G → exact method dominates)





H2 nearly optimal if wmax << WN / P: Miguet, Pierson, Heuristics for 1D rectilinear partitioning as a low cost and high quality answer to dynamic load balancing, LNCS, vol. 1225, 1997, pp. 550-564.

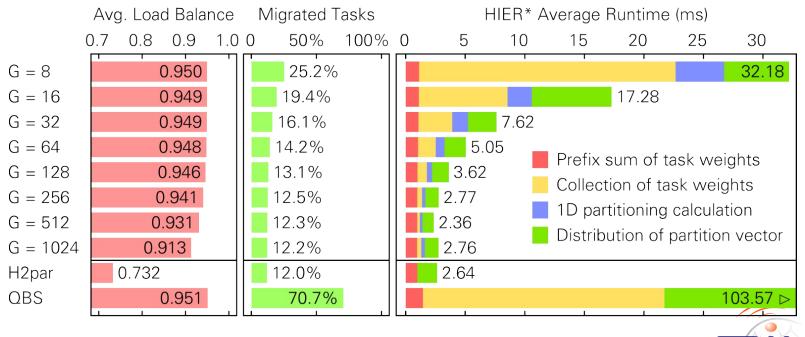
#### Scalable High-Quality 1D Partitioning: Group Count G

- Comparison of HIER\* with parallel heuristic H2 and QBS
- Averages over 2000 successive task weight vectors of the LWFA simulation
- System: JUQUEEN, IBM Blue Gene/Q



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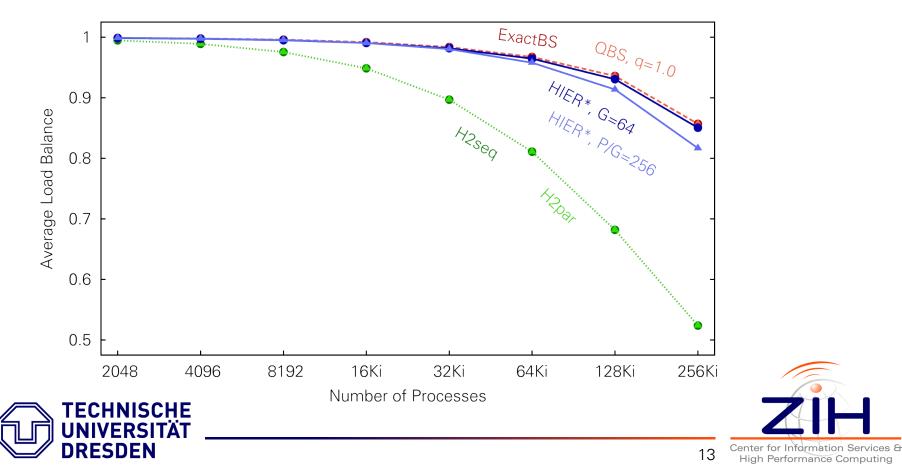
1048576 tasks, 16384 partitions / MPI ranks





#### Scalable High-Quality 1D Partitioning: Load Balance

- Cloud simulation, 1357824 tasks
- System: JUQUEEN, IBM Blue Gene/Q
- HIER\*, G=64 achieves 99.2% of the optimal load balance at 262 144 processes

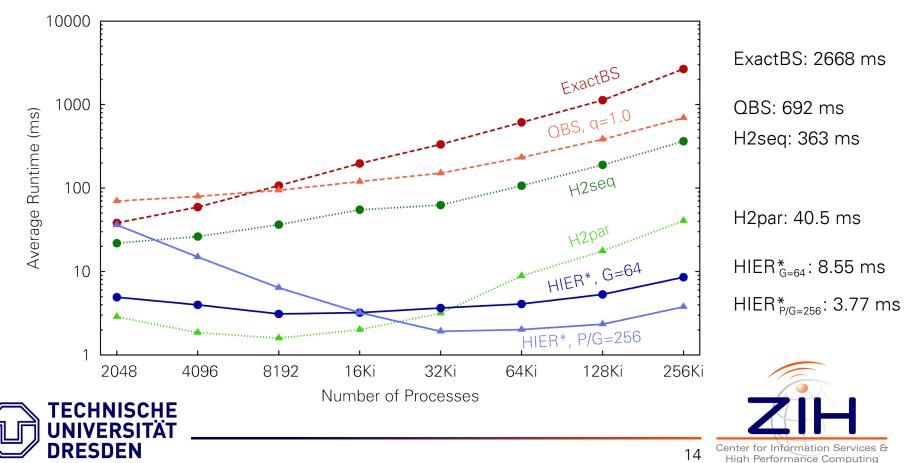




#### Scalable High-Quality 1D Partitioning: BG/Q Scalability

- Cloud simulation, 1357824 tasks
- System: JUQUEEN, IBM Blue Gene/Q
- HIER\*, G=64 runs at 262 144 processes ~300x faster than ExactBS

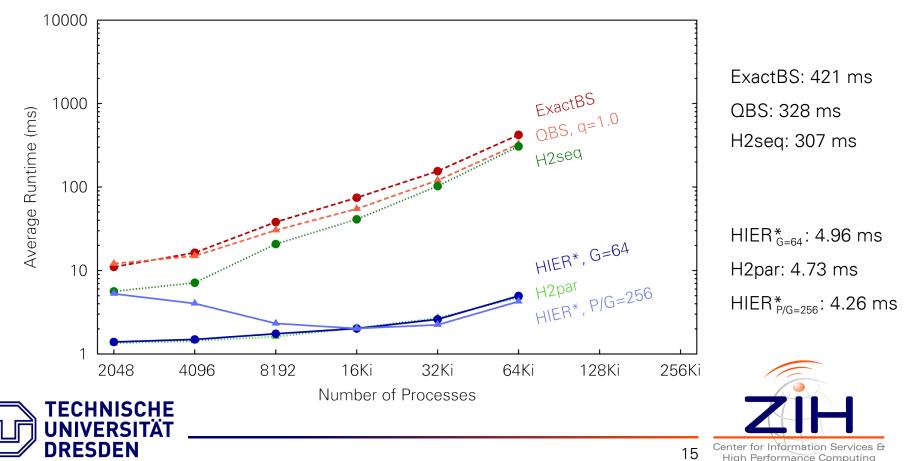




#### Scalable High-Quality 1D Partitioning: iDataPlex Scalability

- Cloud simulation, 1357824 tasks
- System: SuperMUC, IBM iDataPlex (E5-2680, IB FDR10)
- HIER\*, G=64 runs at 65 536 processes ~85x faster than ExactBS





#### Conclusions

- Optimized published exact 1D partitioning algorithm
- Developed scalable, hierarchical algorithm
- Implemented in the open source dynamic load balancing and model coupling framework FD4
  - Benchmark program, and cloud dataset available to reproduce results
  - Enables dynamic load balancing up to 262 144 processes for COSMO-SPECS+FD4

Future work

- Comparison using other applications workload data
- 1D partitioning algorithms tuned for low migration
- Avoid replication of full partition vector on all ranks

FD4 website and benchmark download: http://wwwpub.zih.tudresden.de/~mlieber/fd4

Lieber, Nagel, Mix, Scalability Tuning of the Load Balancing and Coupling Framework FD4, NIC Symposium 2014, pp. 363-370.





#### Thank you very much for your attention!



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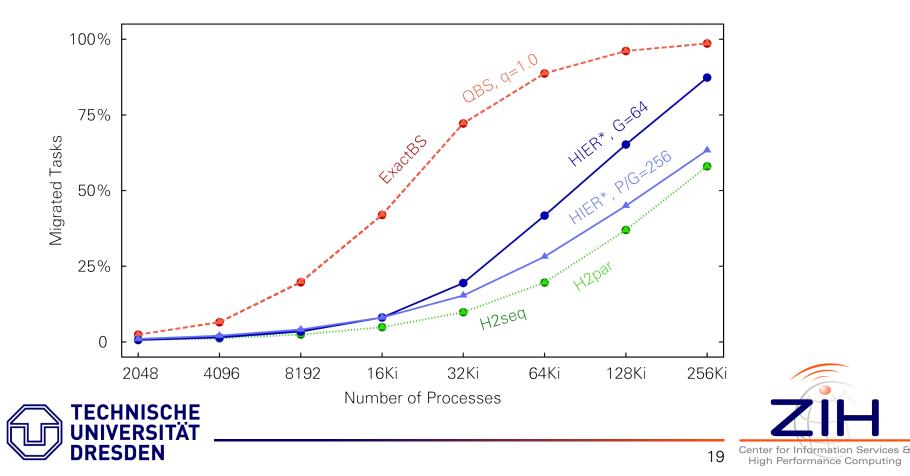




#### Scalable High-Quality 1D Partitioning: Migration

- Cloud simulation, 1357824 tasks
- System: JUQUEEN, IBM Blue Gene/Q
- Hierarchical method lies between exact and heuristic





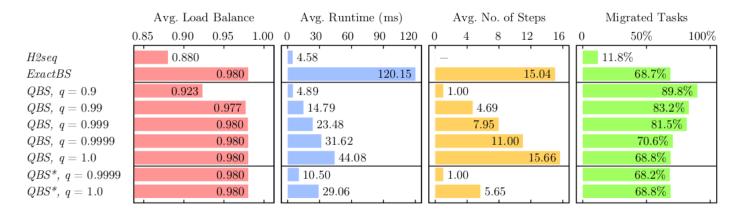


Fig. 5. 1D partitioning benchmark results for the sequential algorithms *H2seq*, *ExactBS*, and *QBS* with the CLOUD dataset (559 104 tasks) for 16 384 processes on JUQUEEN. *QBS\** is a version with parallelized search for the optimal bottleneck. The runtime includes the 1D partitioning calculation only.





#### **HIER\* Algorithm as seen from MPI**

- Prefix sum of task weights + broadcast of total load
  - *MPI\_Exscan* (parallel prefix sum)
  - MPI\_Bsend + MPI\_Recv (consistency at partition borders)
  - MPI\_Bcast (last rank broadcasts total load to all)
- Compute coarse partitioning with parallel H2
  - MPI\_Isend + MPI\_Recv + MPI\_Waitall (send partition borders to group master)
  - MPI\_Bcast (group master broadcasts borders within group)
- Collection of task weights within groups
  - MPI\_Isend + MPI\_Irevc + MPI\_Waitall (send weights from other groups ranks to the nearest ranks in the group these tasks belong to in coarse partitioning)
  - MPI\_Allgather (exchange task weights within group)
- Exact partitioning within group with QBS\*, q=1
  - MPI\_Allreduce
- Distribition of global partition vector
  - MPI\_Allgather (between all group masters)
  - MPI\_Bcast (within groups)





#### HIER and HIER\* Performance (16 384 processes)

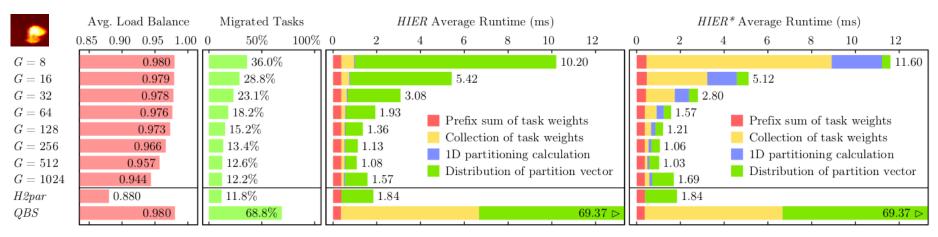


Fig. 6. 1D partitioning benchmark results of the hierarchical methods *HIER* and *HIER*<sup>\*</sup> with the CLOUD dataset (559 104 tasks) for 16 384 processes on JUQUEEN. For comparison, the results of the parallel heuristic *H2par* and the sequential exact algorithm *QBS* (q = 1) are shown.

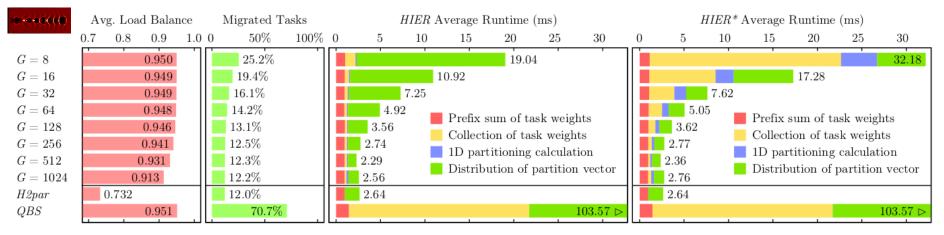
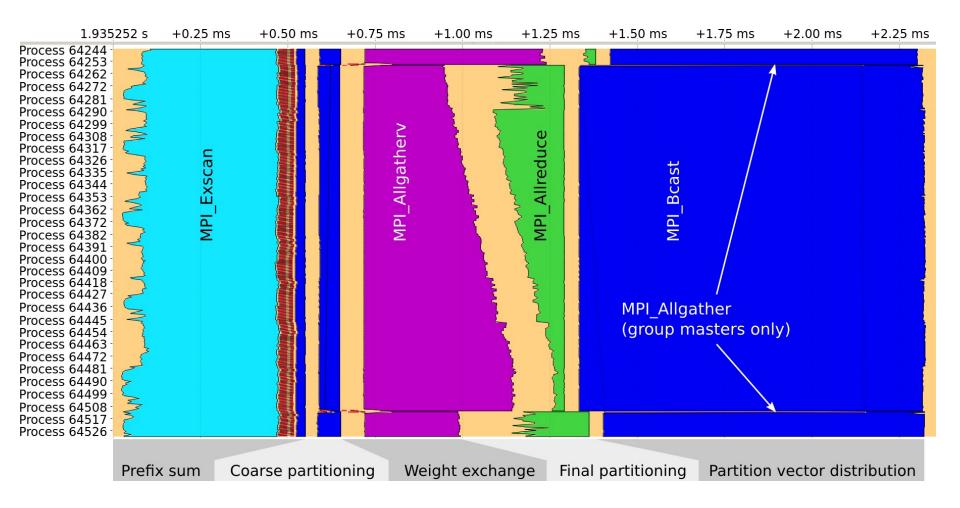


Fig. 7. 1D partitioning benchmark results of the hierarchical methods *HIER* and *HIER*<sup>\*</sup> with the LWFA dataset (1048576 tasks) for 16384 processes on JUQUEEN. For comparison, the results of the parallel heuristic *H2par* and the sequential exact algorithm *QBS* (q = 1) are shown.





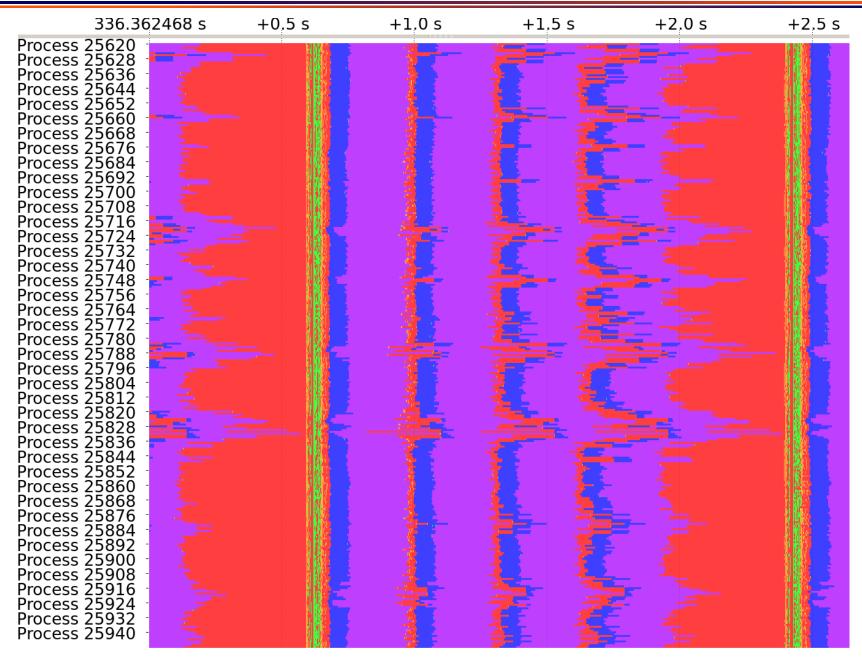
#### HIER\* seen in Vampir (one Group of 256 out of 64Ki)



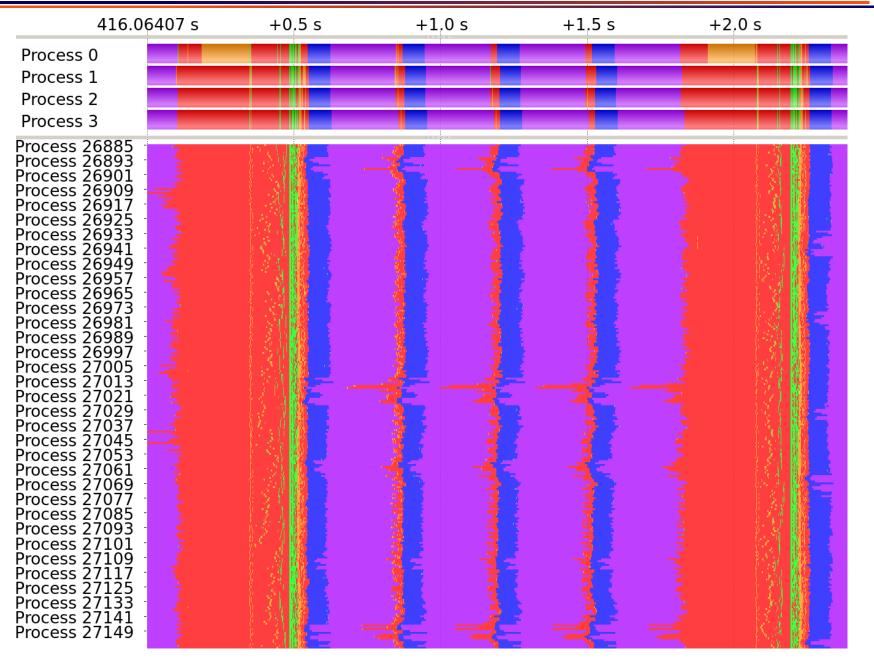




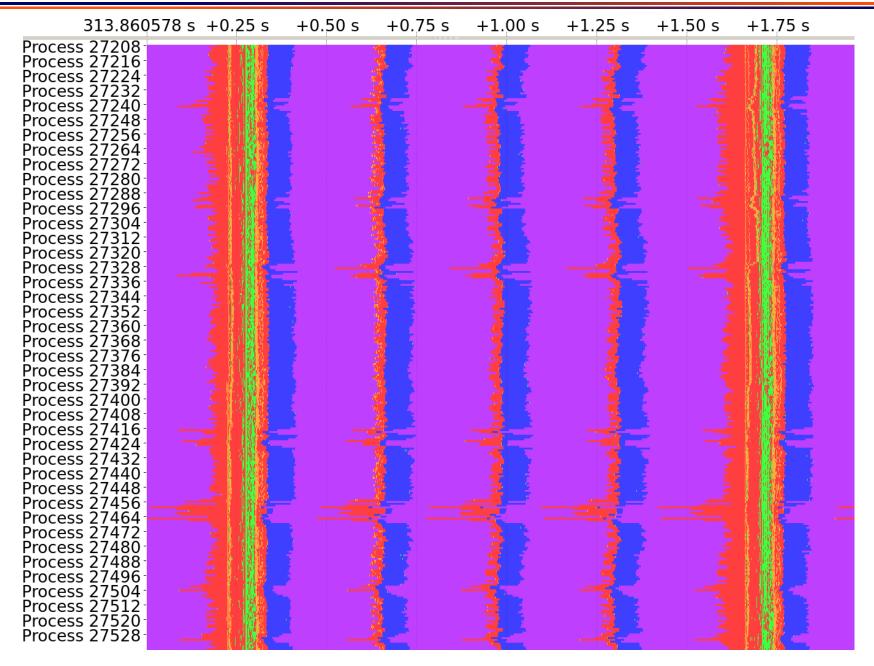
#### Heuristic H2 in Action (COSMO-SPECS+FD4)



#### ExactBS in Action (COSMO-SPECS+FD4)



#### **HIER\* in Action (COSMO-SPECS+FD4)**



#### **COSMO-SPECS+FD4: Comparison of Methods**

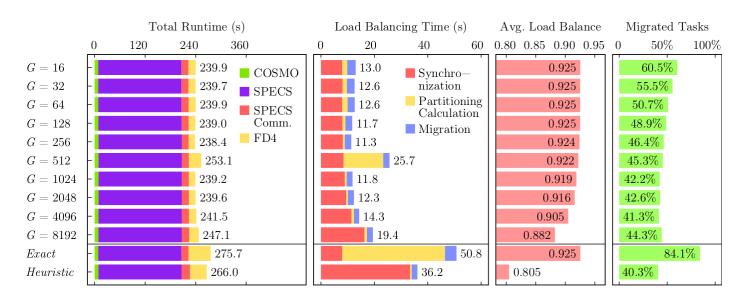


Figure 3. Influence of the group count G on the hierarchical 1D partitioning algorithm in COSMO-SPECS+FD4 with 65 536 processes on BlueGene/Q. The exact method and the heuristic are included as reference.



Lieber, Nagel, Mix,

Load Balancing and

Coupling Framework

FD4, NIC Symposium

2014, pp. 363-370.

Scalability Tuning of the

