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# Memory-Aware Scheduling of Tasks Sharing Data on Multiple GPUs

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# Memory-Aware Scheduling of Tasks Sharing Data on Multiple GPUs

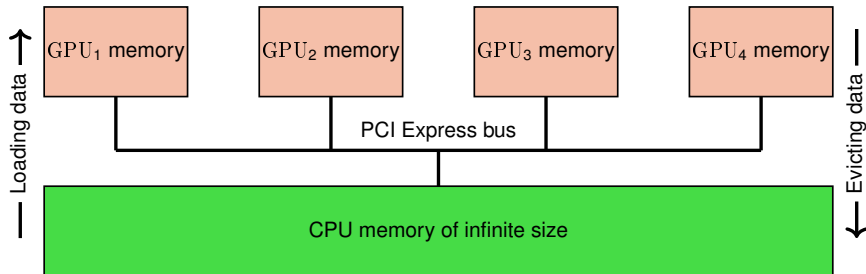
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**LIP - ROMA - LaBRI - STORM - Inria  
ENS de Lyon - Université de Bordeaux**

March 23, 2023

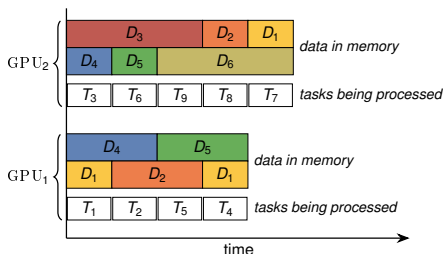
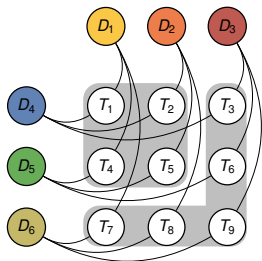
# Why focus on memory-aware scheduling ?



GPUs are quick but have a limited memory

They share the same PCI bus with limited bandwidth

# Example with 2D grid dependencies



## GOALS:

- **Balancing** tasks among GPUs → Reduce total execution time
- **Ordering** tasks inside each GPU → Reduce data transfers
- Having a **generic** scheduler → Efficient with any application or memory limitation

# Dynamic scheduler of STARPU: DMDAR

## Strategy

Schedule tasks so their completion time is minimal based on computation + communication times

## + **Ready** reordering heuristic on $\text{GPU}_k$

From the list  $L$  of tasks allocated on  $\text{GPU}_k$ , it will search the task  $T \in L$  requiring the fewest data transfers

# Data-Aware Reactive Task Scheduling: DARTS

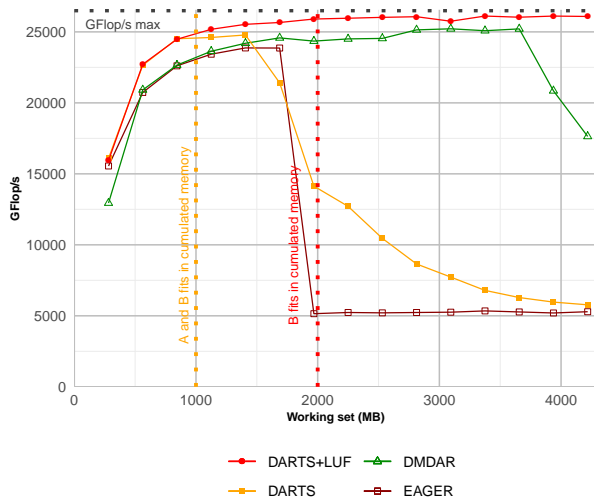
The scheduling is done by  $\text{GPU}_k$  when it needs new tasks.

- Find  $D_{opt} \in \text{dataNotInMem}_k$  such that the number of tasks depending on  $D_{opt}$  and on other data already in memory is maximum (if there is a tie: choose with transfer time, priority of the associated tasks and most total unprocessed task)
- $\text{plannedTasks}_k \leftarrow$  set of unprocessed tasks depending only on  $D_{opt}$  and on other data already in memory

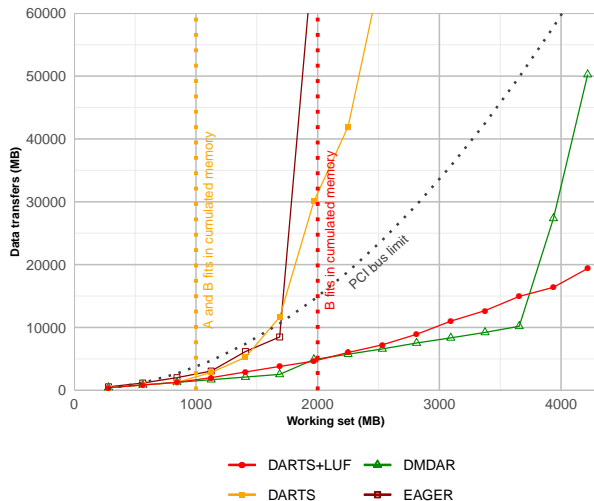
Our eviction policy: LUF (Least Used in the Future)

- Try to evict a data not useful for any task already in the GPU's buffer.
- Try to evict a data used by a minimal number of tasks in the set of task planned by the scheduler.

# 2D matrix multiplication with 2 Tesla V100 GPUs

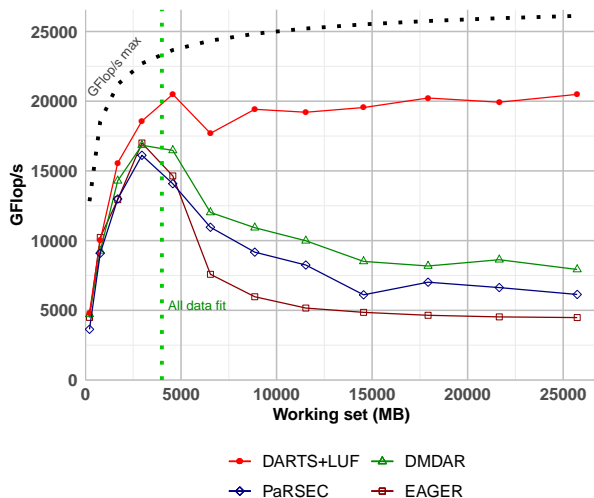


# Amount of data transfers on the 2D matrix multiplication with 2 Tesla V100 GPUs

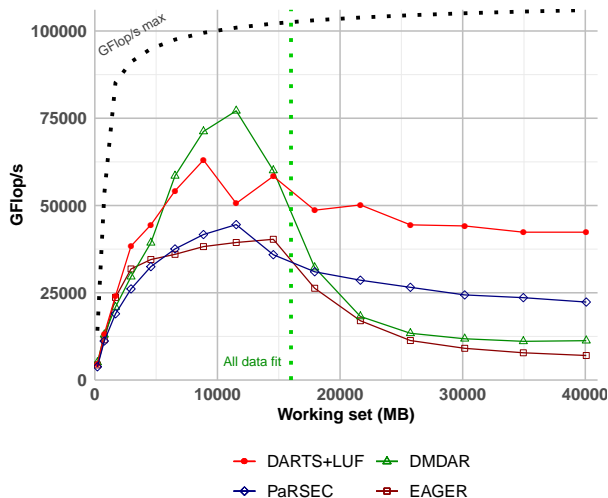




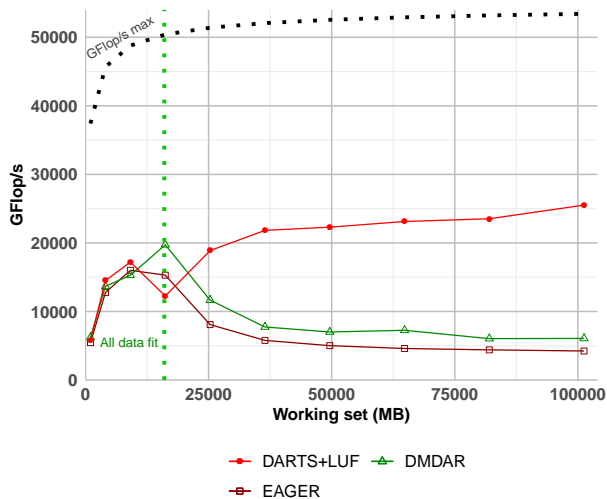
# Cholesky decomposition with 2 GPUs



# Cholesky decomposition with 8 GPUs



# LU with 8 GPUs



# Conclusion

Limiting data movements is crucial

→ **New strategy: DARTS, focused on data movement**

Tested on 2D matrix multiplication, Cholesky decomposition, LU, GEMM and sparse matrix multiplication → **DARTS can achieves good performance when the memory is not constrained and always get very good performance when it is a scarce resource**

Areas for improvement

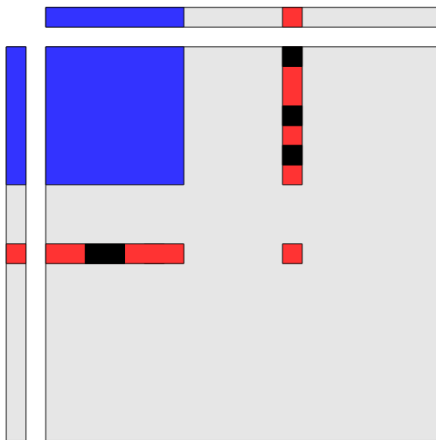
- Improve computational complexity. It's holding us back to get good performances on Gemms and large matrices of the outer product
- Manage multiple MPI nodes
- **Export it to other out-of-core applications. DARTS is not specific to GPUs!**

# If you have an open postdoc position contact me :)

## Contact:

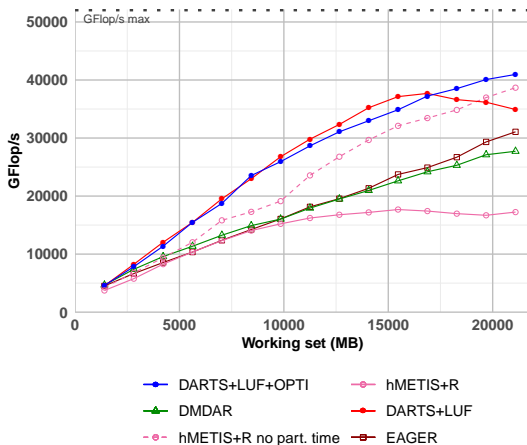
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# Joker slide



**Source:** Analysis of Dynamic Scheduling Strategies for Matrix Multiplication on Heterogeneous Platforms - Marchal - Beaumont

# Sparse 2D matrix multiplication without memory limitation (32GB by GPU) with 4 Tesla V100 GPUs



- No memory constraint
- DARTS produces a **processing order that best distributes transfers over time**