Merchandiser: Data Placement on Heterogeneous Memory for Task-Parallel HPC Applications with Load-Balance Awareness

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Motivation and Introduction

➢ HM raises a data placement problem
➢ Because of small capacity of fast memory and relatively worse performance of slow memory.
➢ Memory pages must be allocated and migrated between fast and slow memories.
➢ To make sure that most of memory accesses can happen in fast memory for high performance.
➢ Many solutions to address the data placement problem on HM uses a profiling-guided optimization (PGO) approach
➢ These solutions identify frequently accessed memory pages ("hot pages") by periodically sampling memory pages and tracking memory accesses to them.
➢ Hot pages are then migrated to fast memory for better performance.

Research questions:
➢ The PGO on HM cannot work well for task-parallel applications
➢ Because they lack a view of “finishing all tasks fast” for high performance.
➢ They migrate and place hot pages into fast memory, but do not consider which task accesses those memory pages.

Methodology

➢ Introduce a load balance-aware data placement system for HM, named Merchandiser, to address the problem
➢ Merchandiser introduces task semantics during memory profiling. This means Merchandiser associates memory accesses with tasks during profiling, instead of being application-agnostic.
➢ Using limited task semantics, Merchandiser effectively sets up coordination among tasks on the usage of HM.
➢ Merchandiser uses historical, fine-grained profiling results of the task to guide data placement for the subsequent executions of the same task with new inputs.
➢ Performance modeling to predict execution time of the task:
➢ The novelty of our performance modeling lies in the modeling of performance correlation between different data placements of the task.
➢ Performance modeling takes the performance of a data placement as input, and then predicts the performance of another data placement.
➢ Greedy heuristic algorithm:
➢ Decide how to allocate the fast memory space among tasks to maximize performance benefit of all tasks.

Evaluation Results

Observation 1:
Figure 1 shows overall performance normalized to PM-only (Optane-only). Merchandiser introduces 23.6%, 17.1%, and 15.4% performance improvement on average (up to 37.8%, 26.0%, and 23.2%) over PM-only, Memory Mode, and MemoryOptimizer respectively.

Observation 2:
For applications with large load imbalance, such as BFS and NWChem-TC, the number of pages being migrated among tasks can vary by up to 21.4 times.

Observation 3:
Compared with Memory Mode and MemoryOptimizer, Merchandiser reduces A.C.V (average coefficient of variation of execution time across threads/processes) by 51.6% and 42.7% on average respectively.

Observation 4:
Compared with Memory Mode, Merchandiser increases average DRAM bandwidth usage from 5.98 GB/s to 24.31 GB/s, indicating the usage of fast memory is improved.

Meanwhile, the average PM bandwidth usage is reduced from 13.74 GB/s to 9.97 GB/s, indicating the effectiveness of page migration in Merchandiser.

Conclusion

➢ The traditional wisdom “migrating frequently accessed pages to fast memory leads to better performance” is not necessarily correct.
➢ We introducing task semantics during memory profiling and migration to address the limitation of traditional wisdom.
➢ We introduce a load balance-aware data placement system for HM.

References