

cuTensor-HT: High Performance Third-order Hierarchical Tucker Tensor Decomposition on GPUs

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Motivations



- Real data are often modeled as low-rank tensors.
- Tensor operations are compute-intensive.
- The running time and complexity grow rapidly with tensor order and size.

Contributions



We implement the HT decomposition on the GPU. We propose optimization strategies for memory access,reducing the amount of calculation and improving resource utilization We conduct experiment to evaluate the performance of HT decomposition algorithm on the GPU and we achieved 4.67 times speedups over the unoptimized GPU baseline.

Parallel HT Decomposition on the GPU







Input: Tensor $\mathcal{X} \in \mathbb{R}^{n_1 \times n_2 \times n_3}$, rank r_1, r_2, r_3, r_4 . 1: for i = 1 to 3 do 2: $H_i = X_{(i)} X_{(i)}^T$ 3: end for 4: for j = 1 to 3 in parallel do 5: $U_i \leftarrow (r_i \text{ leading eigen vectors}) \text{ of } H_i$ 6: end for 7: $U_4 \leftarrow (r_4 \text{ leading left singular vectors}) \text{ of } X_{(3)}^T$ 8: $\mathcal{B}_2 = \mathcal{U}_4 \times_1 U_1^T \times_2 U_2^T$, 9: $\mathcal{B}_1 = \mathcal{X} \times_1 U_4^T \times_2 U_3^T$, **Output:** $\mathcal{B}_1, \mathcal{B}_2, U_3, U_2, U_1$.

Design and Implementation





- Take the batch solution to calculate intermediate variables in parallel.
- Use shared memory to store data, which has low latency and high bandwidth relative to global memory.



Experiment Results



