

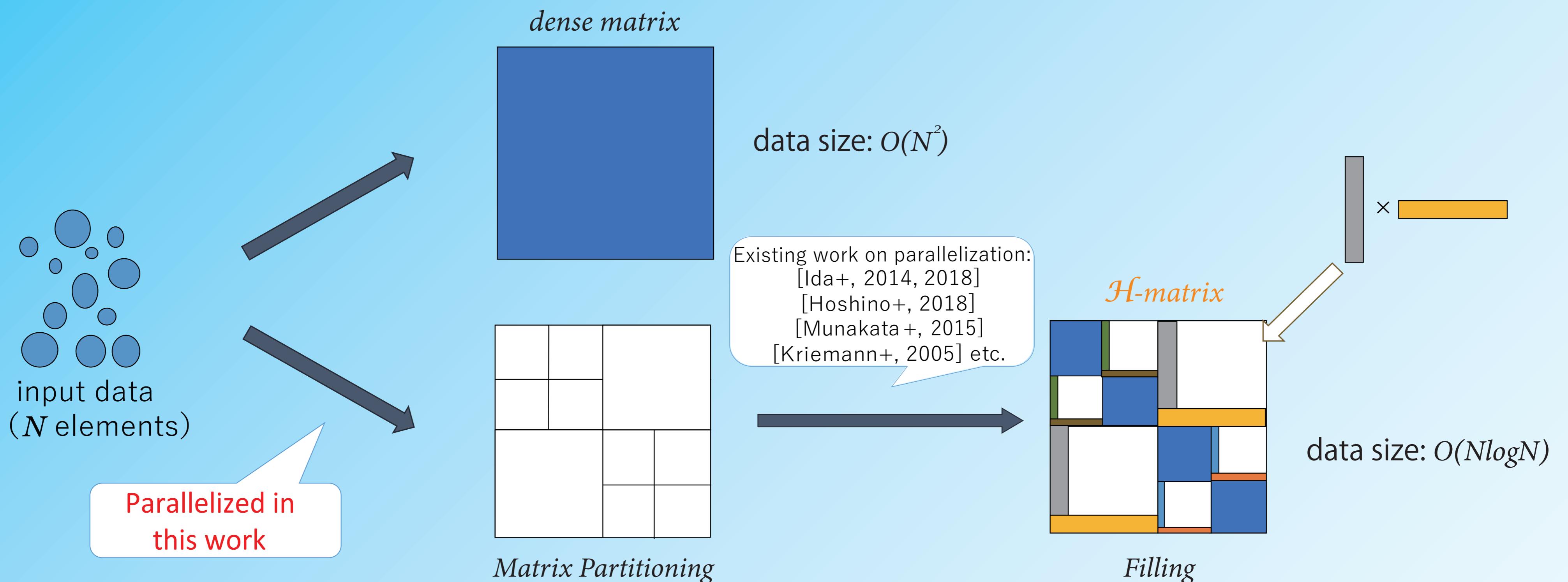
# Implementation of Partitioning of Hierarchical Matrices using Task Parallel Languages

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## Hierarchical Matrix ( $\mathcal{H}$ -matrix)

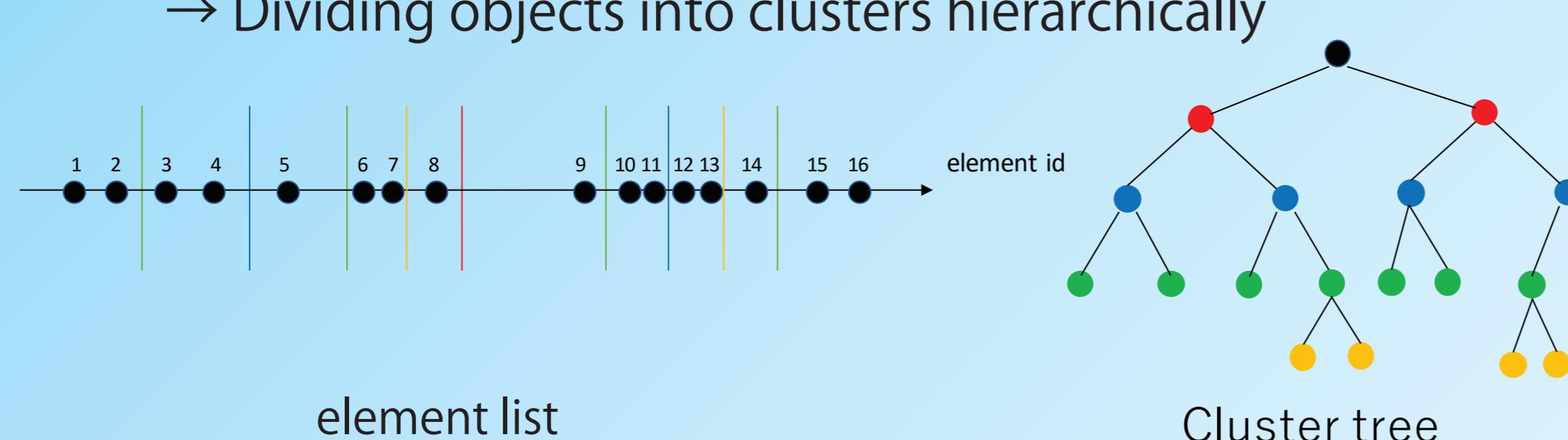
- An approximated form that represents  $N \times N$  correlations of  $N$  objects



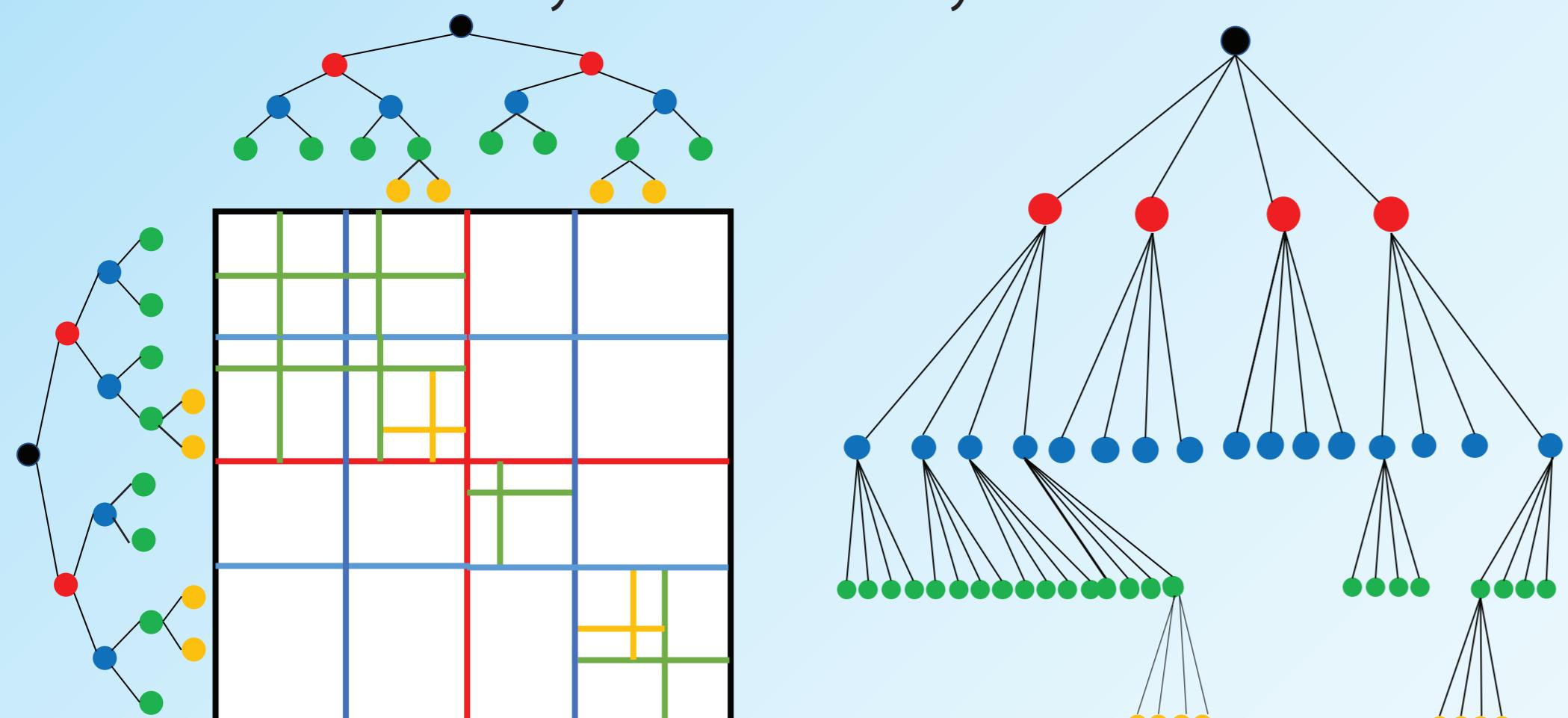
## Parallelization of Matrix Partitioning using Task Parallel Languages

- Matrix Partitioning

- Step 1: Cluster tree (CT) construction  
→ Dividing objects into clusters hierarchically



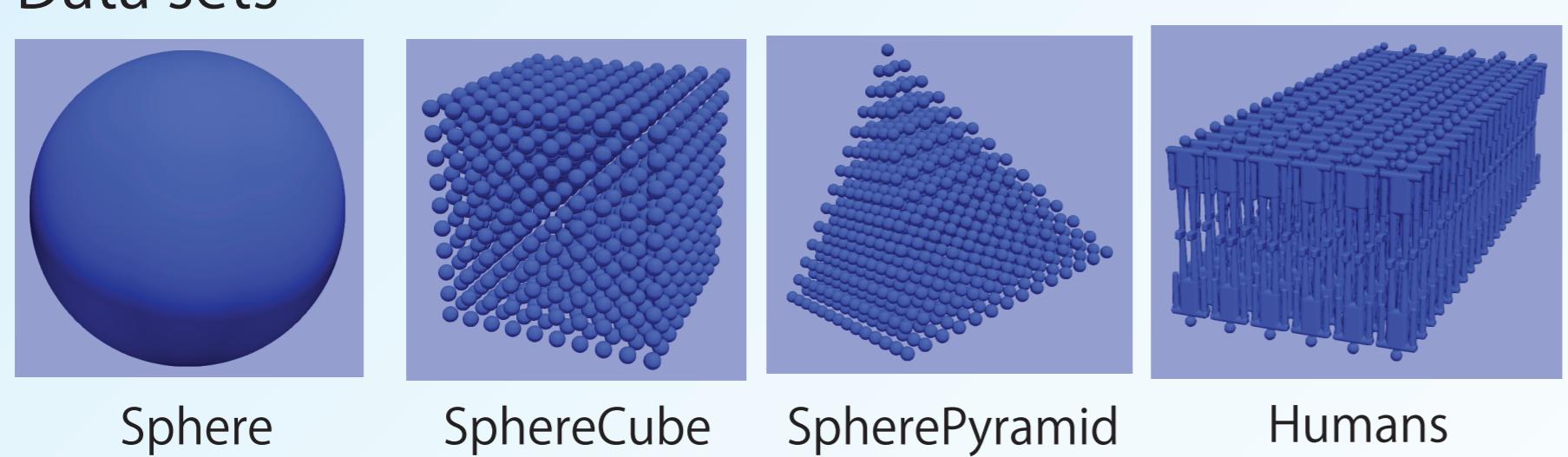
- Step 2: Block cluster tree (BCT) construction  
→ Observe all cluster pairs at the same level of the cluster tree that satisfy an admissibility condition



- CT and BCT are unpredictably unbalanced  
→ We used task parallel languages, Cilk Plus and Tascell, to get good load balance
  - Parallelized recursive calls using spawn/sync (Cilk Plus) or do\_two (Tascell)
  - In CT construction, we also parallelized the inside of each recursive step

## Evaluation

- Data sets



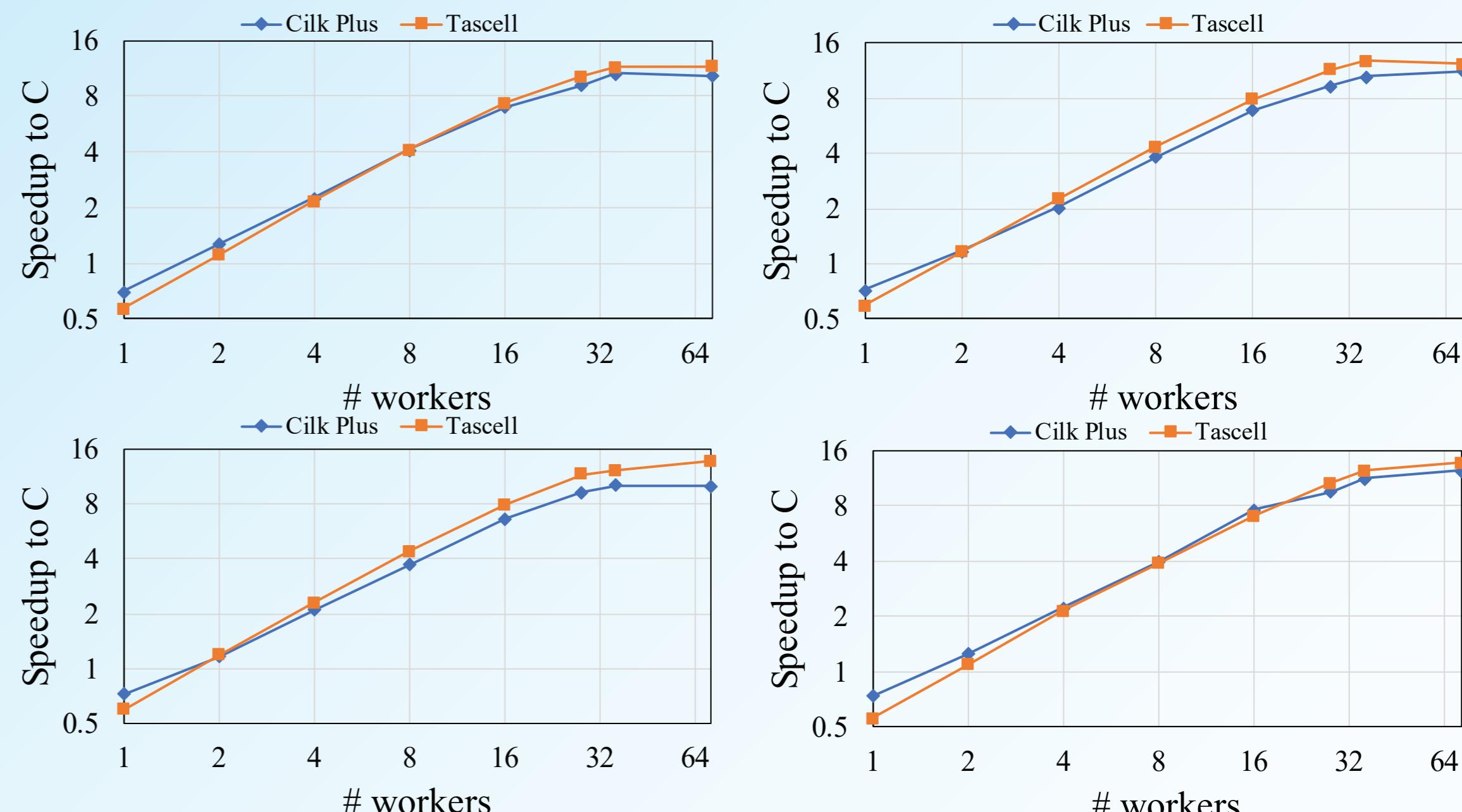
# elements: 50,000,000

SphereCube 101,250,000

SpherePyramid 102,768,750

Humans 98,320,000

- Speedup (Xeon Broadwell 18 cores × 2 socket, HT enabled)



- Performance Parameter Tuning (using 36 workers in Sphere)

$T_N$  : Execute recursive calls sequentially when # elements >  $T_N$   
 $T_S$  : Execute pivoting (recursive step) sequentially when # elements >  $T_S$   
 $C$  : Chunk size in the parallelization of pivoting

