Parallel Beam Search for Functionality Partial Matching

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What is the functionality of an object?



Chair Sitting + Leaning



Crib
Storage + Rocking

What is functionality partial matching?





Problem Statement

- Develop a parallel beam search algorithm
- Speed up the existing functionality partial matching method

Literature Review

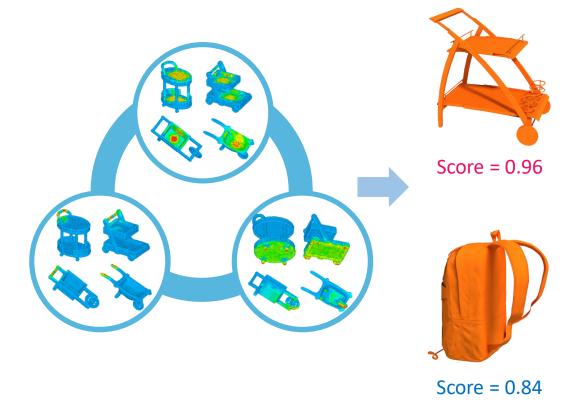
Functionality Analysis in 3D Shape Modeling



Interaction landscapes [8]

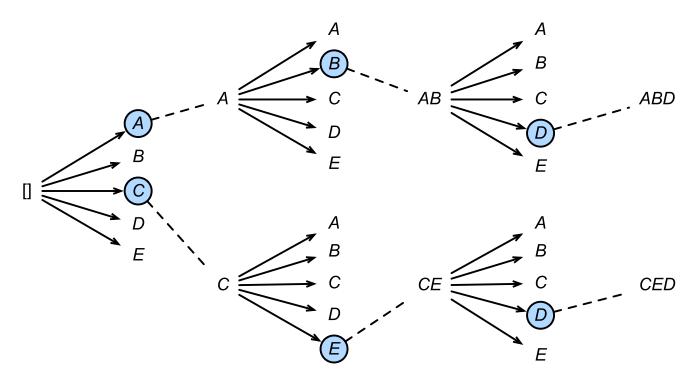


Interaction context (ICON) [5]



Category functionality model [4]

Beam Search and Its Applications

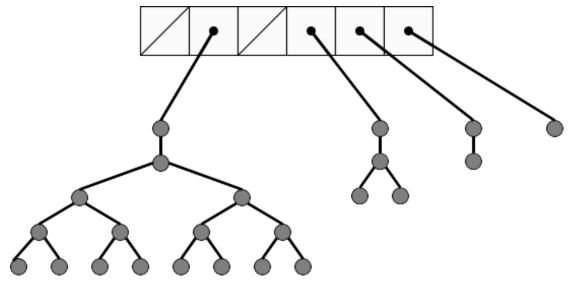


Applied in Speech recognition systems [7, 2], machine translation systems [10], sequence-to-sequence models [9], and so on...

Parallel Breadth-First Search (BFS) Algorithms

```
1 for all v \in V in parallel do
       d[v] \leftarrow -1;
 \mathbf{3} \ d[s] \leftarrow 0;
 4 Q \leftarrow \phi;
 5 Enqueue s \leftarrow Q;
 6 while Q \neq \phi do
        for all u \in Q in parallel do
             Delete u \leftarrow Q;
 8
             for each v adjacent to u in parallel do
 9
                  if d[v] = -1 then
10
                      d[v] \leftarrow d[u] + 1;
11
                      Enqueue v \leftarrow Q;
12
```

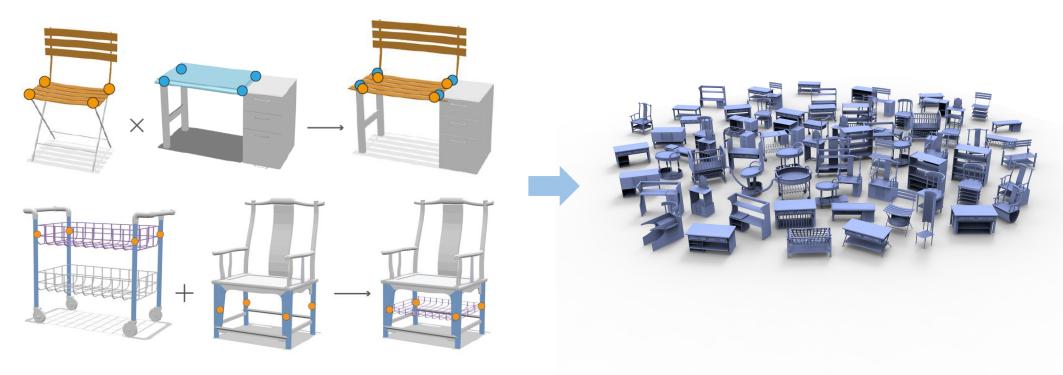
Level synchronization [1, 11]



Multi-set data structure [6]

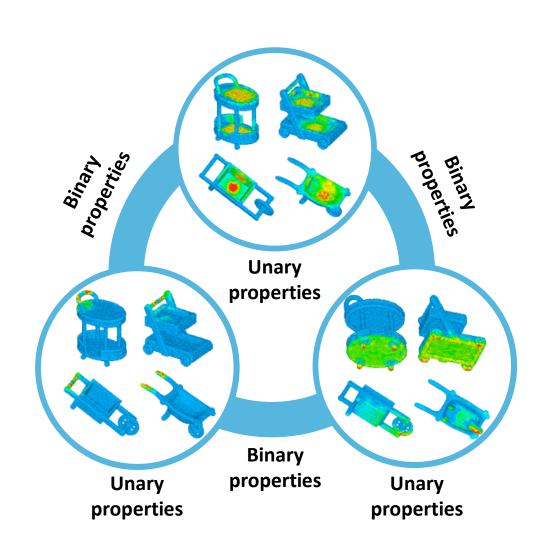
Our Method

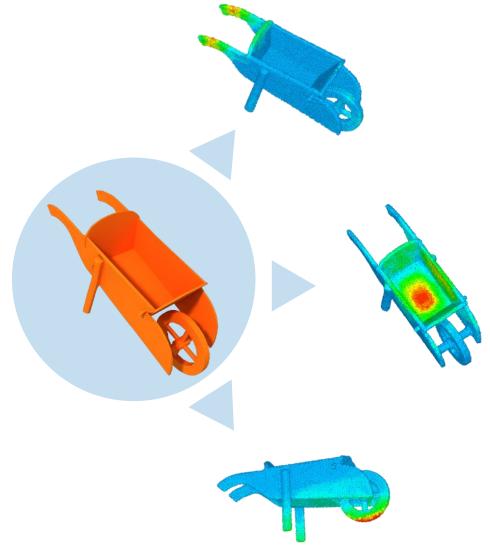
Input: Hybrid Shapes



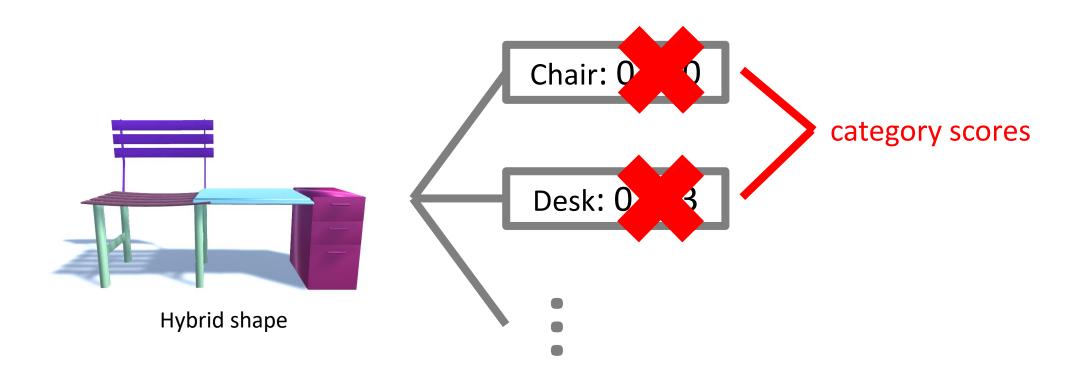
Shape hybridization via evolution [3]

Category Functionality Model





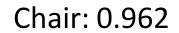
Category Functionality Model



Category models [4]

Functionality Partial Matching





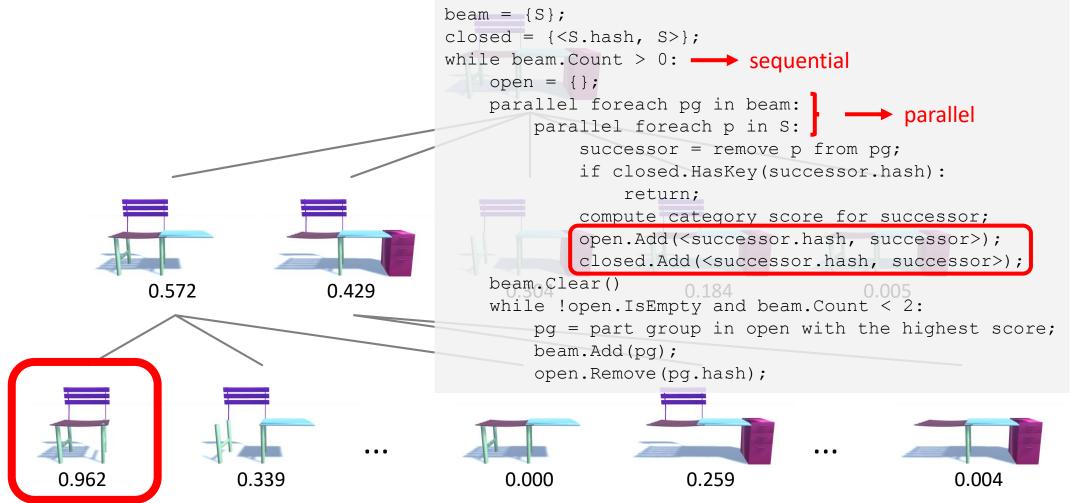


Desk: 0.988



Final score: 0.988

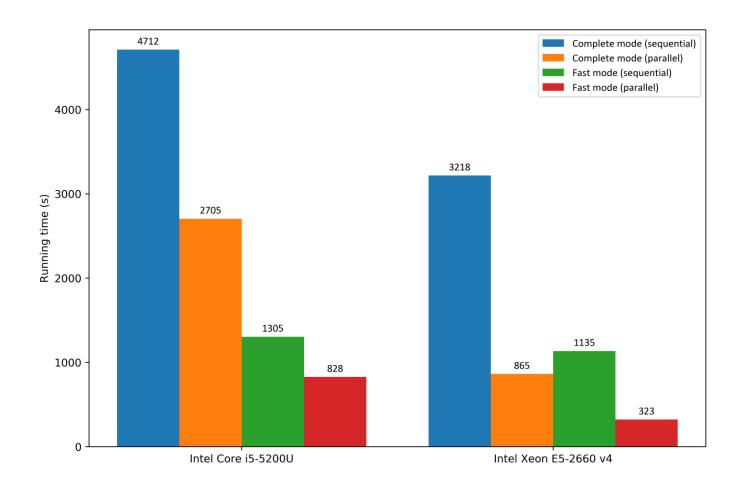
Parallel Beam Search



Results and Applications

Shape	Combinatorial search		Beam search		Time
	$s_{\text{chair}} = 0.96$	$s_{\text{desk}} = 0.98$	$s_{\text{chair}} = 0.96$	$s_{\text{desk}} = 0.98$	722s
	$s_{\text{chair}} = 0.93$	$s_{\text{handcart}} = 0.98$	$s_{\text{chair}} = 0.90$	$s_{\text{handcart}} = 0.98$	1161s
	$s_{\text{chair}} = 0.96$	$s_{\text{shelf}} = 0.94$	$s_{\text{chair}} = 0.96$	$s_{\text{shelf}} = 0.93$	232s
	$s_{\text{desk}} = 0.98$	$s_{\text{shelf}} = 0.96$	$s_{\text{desk}} = 0.98$	$s_{\text{shelf}} = 0.96$	1258s
	$s_{\text{handcart}} = 0.96$	$s_{\text{shelf}} = 0.97$	$s_{\text{handcart}} = 0.92$	$s_{\text{shelf}} = 0.96$	765s

Performance and Scalability



Applications



Functionally plausible

Applications



Initial population



Sitting + Leaning



Placement + Storage



Grasping + Rolling

Conclusion and Future Work

- A parallel beam search can largely speed up the process of functionality partial matching
- The implementation of the category functionality model can be further optimized

References

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Questions

- Why do we need functionality partial matching?
- What is the key idea of parallelizing a beam search?
- Why do we need concurrent open and closed lists in the parallel beam search?