Parallel Beam Search Algorithms for Domain-Independent Dynamic Programming

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Background

Domain-Independent Dynamic Programming (DIDP)

A user can solve a combinatorial optimization problem by formulating a Dynamic Programming (DP) model



Recently proposed by us [Kuroiwa and Beck 2023b]

Example: DP Model for TSP

- TSP: Minimize the total travel cost to visit all customers and return
- DP: State space representation of the problem Customers are visited one by one in TSP



State transition graph for the DP model of TSP

Example of DIDP with Python

import didppy as dp

```
model = dp.Model(maximize=False)
customer = model.add object type(number=4)
c = model.add int table([[0, 3, 4, 5], [3, 0, 5, 4], [4, 5, 0, 3], [5, 4, 3, 0]])
u = model.add set var(object type=customer, target=[1, 2, 3])
i = model.add element var(object type=customer, target=0)
for j in range(1, 4):
    visit = dp.Transition(
        name="visit {}".format(j),
        cost=c[i, j] + dp.IntExpr.state cost(),
        effects=[(u, u.remove(j)), (i, j)],
        preconditions=[u.contains(j)],
    model.add transition(visit)
model.add base case([u.is empty()], cost=c[i, 0])
model.add dual bound(0)
solver = dp.CABS(model, threads=32)
solution = solver.search()
```



Define constants and state variables

Define transitions between states

Define goal conditions

Call a solver

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                        Contribution of this paper
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Solving DP with Heuristic Search

Solving the DP model by finding a path in the state transition graph

f-value: priority to expand, g + h

g-value: actual path cost h-value: estimation by a heuristic function (given with a DP model in current DIDP)



- Beam search expands the *b* states minimizing *f*-values in each layer
- Complete Anytime Beam Search (CABS) repeats beam search with increasing *b* until finding an optimal solution [Zhang 1998]

b = 2

Initial state

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Parallel Beam Search Algorithms

Approach 1: Shared Beam Search (SBS)

- Expand the best *b* states (obtained by **parallel sort**) in parallel
- Use a **concurrent hash table** for duplicate detection Divided into multiple shards, and each shard has a lock



Similar to problem-specific parallel beam search by Frohner+ (2023)

 Send a state to a thread determined by its hash value using message passing (duplicate states sent to the same thread)

b = 2, #threads = 2

• Each thread **locally detects duplicates** and expands $\frac{b}{\# threads}$ states



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HDBS1: Immediate Layer Synchronization



HDBS2: Delayed Layer Synchronization



Experimental Evaluation

SBS vs. HDBS: Mean Speedup against Single Thread



Used with CABS and measure the time to solve optimally (limits: 5-min and 188GB)

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DIDP vs. Commercial Parallel Optimization Solvers

Problem	Description	Gurobi	СРО	DIDP (HDBS2)
TSPTW (340)	TSP with time	239/4.2	27/0.1	262 /13.3
CVRP (207)	vehicle routing	29 /5.3	0/ -	8/ 9.3
SALBP-1 (2100)	line balancing	1351/1.3	1581/1.4	1826 /18.8
Bin Packing (1615)	bin packing	1192/6.4	1251 /9.2	1239/39.6
MOSP (570)	manufacturing	238/3.1	397/0.3	531 / 9.0
Graph-Clear (135)	building security	16/2.0	4/3.2	113 /10.3

#optimally solved / mean speed up

- Resources: 32 threads, 5-min, and 188GB
- Gurobi: mixed-integer programming solver
- CPO: IBM ILOG CP Optimizer (constraint programming solver)

Conclusion

- A parallel beam search algorithm, HDBS2, shows good speedup and yields a high-performance parallel combinatorial optimization solver
- Start DIDP with Python: pip install didppy

Tutorials and API References

Project Page





GitHub Repo

