

Genetic Algorithms

The algorithm that we developed is based on Vidal 2021, Computers & Operations Research

Hybrid genetic search for the CVRP: Open-source implementation and SWAP* neighborhood

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ABSTRACT

The vehicle routing problem is one of the most studied combinatorial optimization topics, due to its practical importance and methodological interest. Yet, despite extensive methodological progress, many recent studies are hampered by the limited access to simple and efficient open-source solution methods. Given the sophistication of current algorithms, reimplementing is becoming a difficult and time-consuming exercise that requires extensive care for details to be truly successful. Against this background, we use the opportunity of this short paper to introduce a simple – open-source – implementation of the hybrid genetic search (HGS) specialized to the capacitated vehicle routing problem (CVRP). This state-of-the-art algorithm uses the same general methodology as Vidal et al. (2012) but also includes additional methodological improvements and lessons learned over the past decade of research. In particular, it includes an additional neighborhood called SWAP* which consists in exchanging two customers between different routes without an insertion in place. As highlighted in our study, an efficient exploration of SWAP* moves significantly contributes to the performance of local searches. Moreover, as observed in experimental comparisons with other recent approaches on the classical instances of Uchoa et al. (2017), HGS still stands as a leading metaheuristic regarding solution quality, convergence speed, and conceptual simplicity.

1. Introduction

A decade has passed since the introduction of the hybrid genetic search with advanced diversity control (HGS in short) in Vidal et al. (2012) and the generalization of this method into a unified algorithm

To facilitate future studies, we use the opportunity of this short paper to introduce an open-source HGS algorithm for the canonical capacitated vehicle routing problem (CVRP). We refer to this specialized implementation as HGS-CVRP. The C++ implementation of this algorithm has been designed to be transparent, specialized, and

Implementation

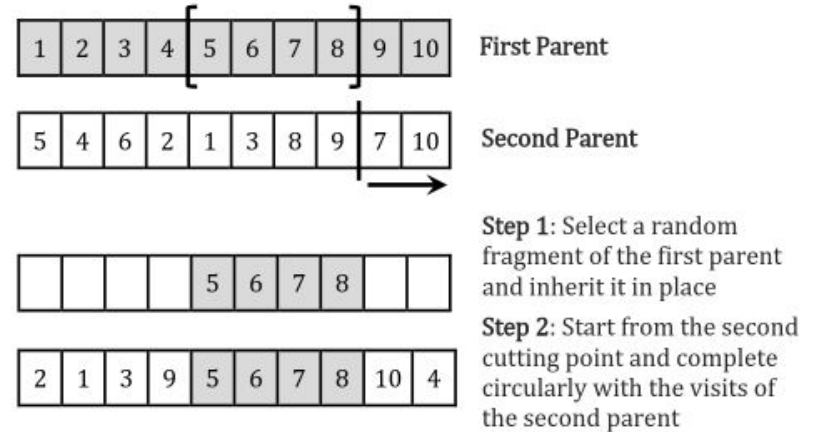
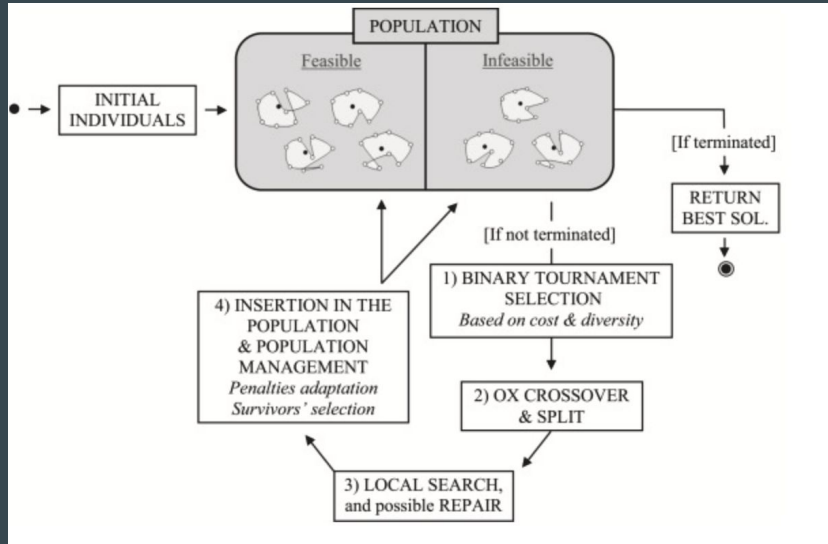


Figure 2: OX Crossover

Some Tuning

- The SWAP* technique mentioned in the paper did not actually perform well (maybe due to our smaller instances and Python overhead)
- Stop condition: 290 seconds passed OR 200 iterations without improvement
- Lots and lots of data structures

Results

- Reasonably competitive solutions for all instances
- Overhead of Python was a major bottleneck
- Move-testing (working with linked lists) in local search step is very slow

```
Collecting samples from 'python src/main.py input/386_47_1.vrp' (python v3.9.21)
Total Samples 28100
GIL: 100.00%, Active: 100.00%, Threads: 1
```

%Own	%Total	OwnTime	TotalTime	Function (filename)
14.00%	14.00%	34.16s	35.98s	move3 (hga_local.py)
8.00%	8.00%	32.95s	34.49s	move5 (hga_local.py)
5.00%	8.00%	29.37s	40.68s	move1 (hga_local.py)
13.00%	13.00%	28.66s	31.33s	move2 (hga_local.py)
6.00%	78.00%	21.59s	257.6s	run (hga_local.py)
5.00%	5.00%	18.68s	23.89s	updateRouteData (hga_local.py)
5.00%	5.00%	17.83s	18.49s	move6 (hga_local.py)
2.00%	2.00%	16.56s	18.80s	move4 (hga_local.py)
5.00%	7.00%	14.89s	19.46s	move8 (hga_local.py)
5.00%	5.00%	13.30s	13.30s	setLocalVariables (hga_local.py)
9.00%	9.00%	12.53s	15.87s	move9 (hga_local.py)
7.00%	7.00%	8.27s	8.27s	propagate (hga_split.py)
1.00%	1.00%	6.66s	6.66s	calc_penalty (hga_local.py)
2.00%	21.00%	5.85s	21.43s	split_1f (hga_split.py)
0.00%	0.00%	4.01s	5.90s	move7 (hga_local.py)
8.00%	8.00%	3.30s	3.30s	dominates (hga_split.py)
0.00%	0.00%	2.36s	3.04s	is_enclosed (hga_circle.py)
0.00%	0.00%	1.61s	4.92s	extend (hga_circle.py)
0.00%	0.00%	1.51s	1.51s	dominates_right (hga_split.py)
1.00%	1.00%	1.08s	1.08s	brokenPairsDistance (hga_population.py)
0.00%	0.00%	0.950s	0.950s	positive_mod (hga_circle.py)
2.00%	2.00%	0.870s	0.870s	size (hga_split.py)
0.00%	1.00%	0.840s	1.33s	loadIndividual (hga_local.py)
1.00%	1.00%	0.790s	0.790s	get_front (hga_split.py)
1.00%	1.00%	0.350s	0.350s	get_back (hga_split.py)
0.00%	0.00%	0.250s	0.250s	evaluateCompleteCost (hga_structures.py)

Thank you py-spy

Reflection

- Time Spent: Too long (90+ student hours)
- Many sunrises witnessed
- Slept for 1 hour night before a final
- Code that got nuked harder than our subpopulations