Savings, Savings, Savings

Gabriel C-Parent

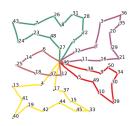
Overview

- cvrp problem
- implementation details
- improvement procedures
- construction procedure
- genetic algorithm
- tabu search
- QA

The Problem

The Problem 3/33

Capacitated Vehicle Routing



The Problem 4/33

Implementation

Implementation 5/33

Implementation Details







Implementation 6/33

Implementation Details

- reuse basic operators
- modularity
- concise

Implementation 7/33

Improvement

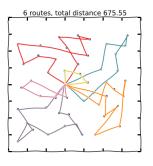
Improvement 8/3

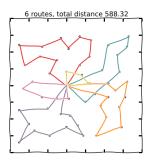
2-opt descent

- uses common 2-opt operator
- calculates all possible 2-opt for each iteration
- chooses the best available

Improvement 9/33

2-opt example





Improvement 10/33

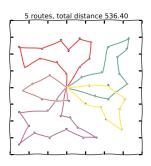
λ_1 -interchange definition

- λ -interchange, Osman, 1991
- exchange of customers between routes
- only feasible exchanges (capacity constraint)
- insertion (1, 0) and (0, 1) or interchange (1, 1)
- chooses the best option at each iteration
- apply 2-opt descent on routes implicated

mprovement 11/3:

λ_1 -interchange example





Improvement 12/33

Construction

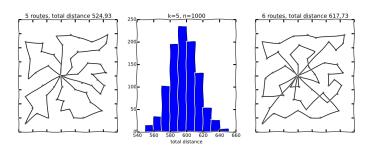
Construction 13/33

Random Savings Definition

- iterated local search
- variant of parallel savings
- at each iteration select randomly from top k best savings
- $k=1
 ightarrow ext{normal parallel savings}$
- once finished, apply improvement method

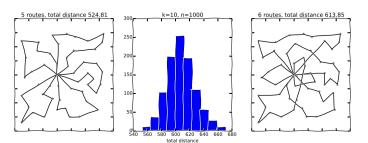
Construction 14/33

Random Savings, k=5



Construction 15/33

Random Savings, k=10



Construction 16/33

Best result in 60 secs



Construction 17/33

Genetic Algorithm

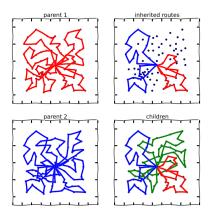
Genetic Algorithm 18/33

Crossover Description

- 1 arrange routes by angle of centroid to depot
- 2 choose $\frac{n}{4}$ contiguous routes from parent 1
- 3 choose at most $\frac{n}{4}$ contiguous non-intersecting routes from parent 2
- 4 assign routes to the rest of clients using parallel savings

Genetic Algorithm 19/33

Crossover example



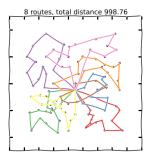
Genetic Algorithm 20/33

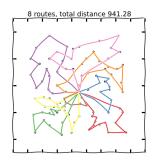
Mutation Description

- reuse the λ_1 -interchange operator
- use a fixed (in this case 5) number of iterations
- serves as local exploration and changes clusters of clients

Genetic Algorithm 21/3

Mutation example





Genetic Algorithm 22/33

Genetic Algorithm Parameters

- number of generations
- population size
- elitism (percentage of transfer)
- recombination probability
- mutation probability
- k for random savings

Genetic Algorithm 23/33

Results



Genetic Algorithm 24/33

Tabu Search

Tabu Search 25/33

Tabu Search Description

- Neighbourhood Structure
- Tabu List
- Diversification

Tabu Search 26/33

Neighbourhood Structure

- λ_1 -interchange
- only feasible solutions

Tabu Search 27/33

Tabu List

- avoid reversing a move
- remember pairs (client, route)
- $max{7, -40 + 9.6 \times ln(n \times v)}$

Tabu Search 28/33

Diversification by Multi-Start

- takes a parameter called patience
- patience replenish after a new best is found
- patience runs out → random savings

Tabu Search 29/33

Best Results in 60 sec



Tabu Search 30/33

Overall Performance

Overall Performance 31/33

Overall Performance

- complexity: [RS, GA, Tabu]
- speed: [Tabu, RS, GA]
- quality: [GA/Tabu, RS]

Overall Performance 32/33

QA

QA 33/33