

Pathfinding and Routing

NAIL137

Vehicle routing problem

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Overview

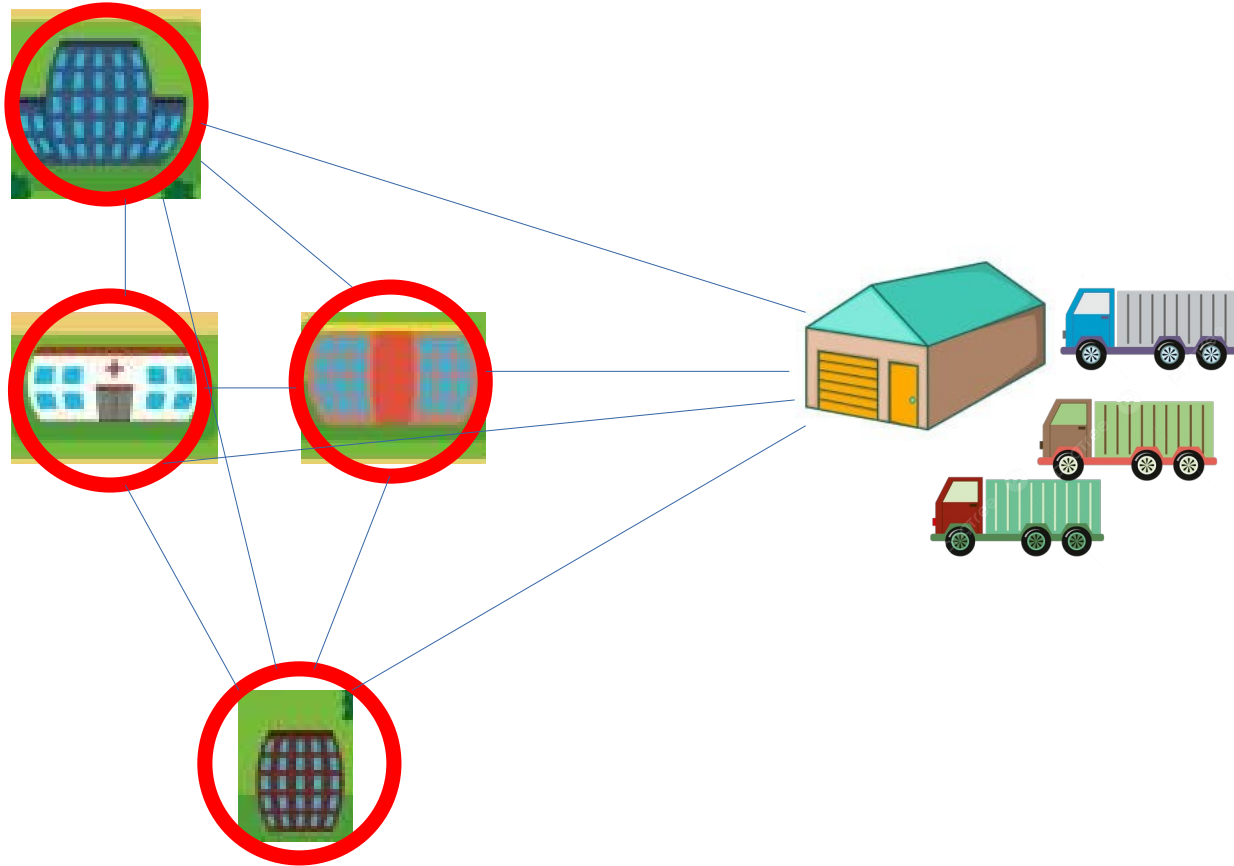
- Definitions
- Variations
- Optimal solvers
- Suboptimal solvers
- Use cases

Motivation



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Motivation - simplification



Definition

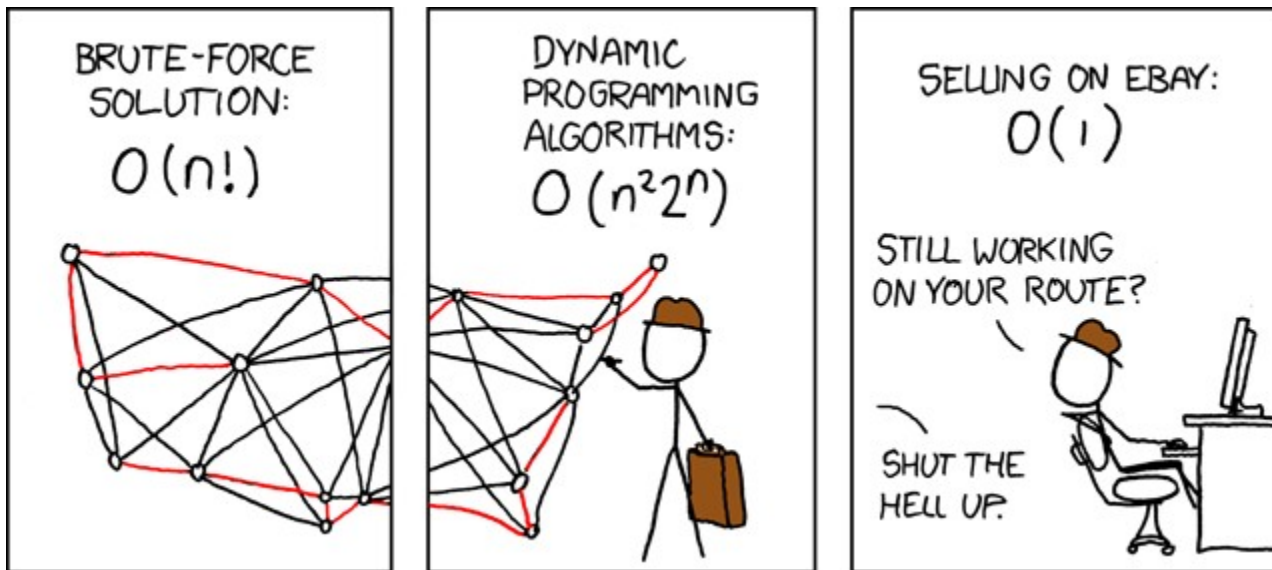
- A complete weighted graph $G=(V,E)$
- v_0 is a depot, v_1, \dots, v_n are customers
- A set of k identical vehicles

Find a set of paths, one for each vehicle, such that each customer is visited exactly once.

Introduced in 1959 as „Truck Dispatching Problem“

Special case - single vehicle

Traveling salesman problem



Complexity

- TSP is NP-Hard
- VRP is even harder, first decide partition, each vehicle is solving a separate TSP problem

(This is not a formal proof)

Variations

- Distance constrained VRP (DCVRP)
 - Each vehicle has a maximum distance it can travel (cost, safety, balance, ...)
- Capacity-constrained VRP (CVRP)
 - Each vehicle has a specific capacity, each customer demands specific amount
- VRP with Time Window (VRPTW)
 - Demands must be met in specific time-window

Variations cont.

- VRP with Pickup and Delivery (VRPPD)
 - Each delivery needs to be picked first
- Multiple depot VRP (MDVRP)
 - There are more depots
 - Vehicles can start anywhere
- VRP with stochastic (VRPS)
 - Some parts of the problem is probabilistic
 - Travel time
 - Demand
 - Customer presence
 - ...

Optimization functions

- Total distance traveled (=cost)
- Total number of vehicles used
- Number of unfulfilled demands
- Number of broken constraints on time

Exact approaches

- Bruteforce
- Dynamic programming
- Reduction to ILP, CP

Dynamic programming - TSP

- subset $S \subseteq \{1, \dots, n\}$
- $C(S, j)$ – cost of path starting at depot, ending at j
- $C(\{j\}, j) = c_{0j}$
- $C(S, j) = \min_{i \in S \setminus \{j\}} [C(S \setminus \{j\}, i) + c_{ij}]$
- **$T(S) = \min_{j \in S} [C(S, j) + c_{j0}]$**

Dynamic programming - VRP

- $F(S, q)$ – cost to serve customers from S via q vehicles
- $F(S, 1) = T(S)$
- $F(S, q) = \min_{A \subseteq S} [T(A) + F(S \setminus A, q - 1)]$
- **$F(\{1, \dots, n\}, m)$**

Naive ILP model

x_{ijk} - vehicle k is using edge (i,j)

subject to

$$\text{minimise } \sum_{i,j} c_{ij} \sum_k x_{ijk}$$
$$\sum_i \sum_k x_{ijk} = 1 \quad \forall j$$

One incoming visit

$$\sum_j \sum_k x_{ijk} = 1 \quad \forall i$$

One outgoing visit

$$\sum_i x_{ihk} - \sum_j x_{hjk} = 0 \quad \forall k, h$$

Flow conservation for each vehicle

$$\sum_{ijk} x_{ijk} = |S| - 1 \quad \forall S \subseteq P(N), 0 \notin S$$

Subtour elimination

$$x_{ijk} \in \{0, 1\}$$

ILP - column generation

$$\min \sum_k c_k x_k$$

subject to

$$\sum_k a_{ik} x_k = 1 \quad \forall i$$

$$x_k \in \{0,1\}$$

x_k - route of vehicle k

c_k - cost of route

a - 0/1 matrix of visited customers

All customers are covered by all routes exactly once

- 1) Start with a set of routes (columns)
- 2) Create a dual problem to generate new columns (pricing problem)
- 3) Repeat until no improving column (route) exists

ILP - branch and bound

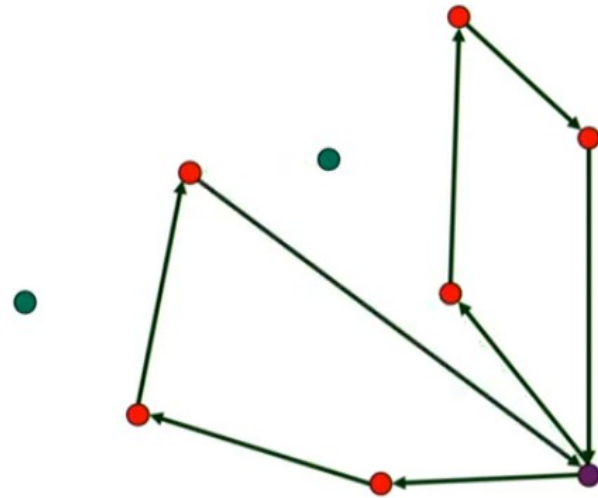
- The model may be solved as LP
 - This creates non-integer solution!
- Branch and bound to extract integer solution
- Combine with the pricing of column generation → branch and price algorithm

- Useful constraint types
 - AllDifferent
 - Circuit

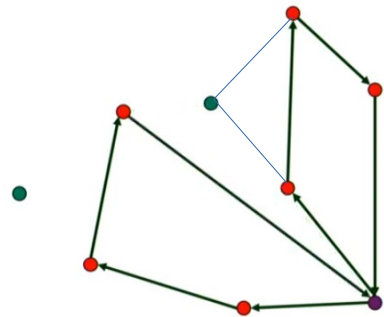
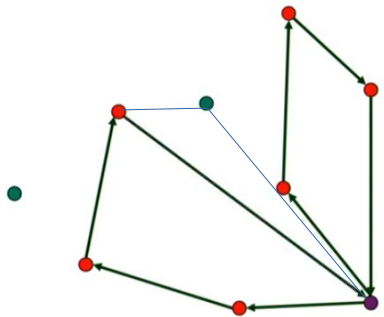
Suboptimal solvers

- Construct by insertion
 - 1) Start with empty solution
 - 2) Choose a customer to insert
 - 3) Choose a place to insert it
- Greedy - choose the cheapest customer to include and the cheapest place to insert
- Regret - insert the customer that would cause the largest regret if postponed

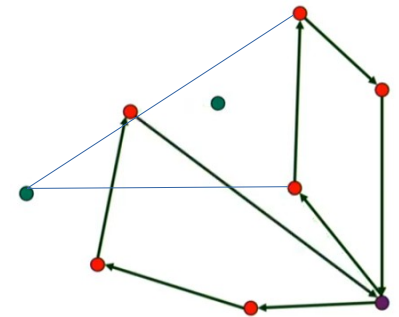
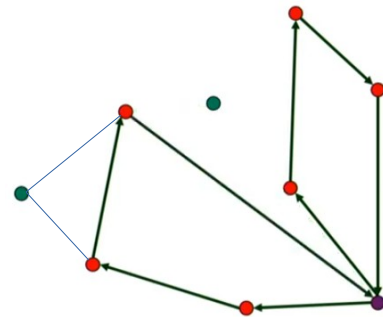
Regret example



Current routes



Second best option is not much worse
→ low regret



Second best option is much worse
→ high regret

Suboptimal solvers cont.

- Improvement methods
- Operators to change solution
 - Move one customer to different position in the vehicle route or different route
 - Swap one customer with another from different vehicle route
 - Swap tails of routes
 - Remove 2 arcs and replace them (reverse order of the customers in between)
 - Select a sequence of customers and try to place them everywhere

Suboptimal solvers cont.

- Improvement methods
 - Local search (such as hill climb, simulated annealing)
 - Genetic algorithms
 - Escape local optimum → take larger neighborhoods

Suboptimal solvers cont.

- Large neighborhood search
 - 1) Delete part of solution (unassign customers)
 - 2) Construct the solution back (any construction algorithm, even optimal or a portfolio of solvers)
 - 3) Keep if better

Suboptimal solvers cont.

- Large neighborhood search improvement
 - Keep worse solution with some prob.
 - Depends on temperature (like SA)
 - Adaptive LNS
 - Change the prob. of selecting an operator based on previous success (in this run)

Use cases

- Logistics
 - Deliveries (ordered packages, recurring deliveries)
 - Trash collection
 - Post delivery
 - Maintenance
- Robotics
 - Task assignment for agents
- Search and rescue

