Excercise 4 Implementing a centralized agent

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1 Solution Representation

We formalize the pickup and delivery problem as a constraint optimization problem. The COP is formally a tuple $\langle X, D, C, f \rangle$ where X is a set of variables describing a plan, D the domain of the variables $x_i \in X$, C a set of constraints that a plan needs to fulfill and f a cost function that we are trying to minimize. A plan P is given by an assignment of the variables in X (where the constraints in C are satisfied). We call N_v the number of vehicles and N_t the number of tasks.

1.1 Variables

Any plan can be described by a given assignment of the 4 following variables (in out implementation, a plan is represented by the plan class):

1. **nextTask_v**: the $nextTask_v$ variable is an array of size N_v where each element $nextTask_v[i]$ represents the first task that the i^{th} vehicle will perform.

if $nextTask_{-}v[i] = j$ it means that the i^{th} vehicle will start it's route by delivering the j^{th} task if $nextTask_{-}v[i] = NULL$ it means that the i^{th} vehicle has no task to deliver in the given plan.

2. $\mathbf{nextTask_t}$: the $nextTask_t$ variable is an array of size $2N_t$ where each element $nextTask_t[i]$ represents the next task that the vehicle that the vehicle will pickup (in this array the pickup and delivery are represented as two distinct array element to allow for multiple tasks to be picked up by the agent)

if $nextTask_t[i] = j$ it means that the vehicle that delivered the i^{th} task will deliver the j^{th} task next.

if $nextTask_t[i] = NULL$ it means that the vehicle that delivered the i^{th} has no task to deliver next.

- 3. **time**: the *time* variable is an array of size $2N_t$ where each element time[i] represents the position of the i^{th} task in the plan of the vehicle delivering it in the plan (so if it is the first task delivered by some vehicle we would have time[i] = 1, again it is of size $2N_t$ to represent pickup an delivery as distinct actions)
- 4. **vehicle**: the *vehicle* variable is an array of size $2N_t$ where each elements vehicle[i] describes which vehicle will delivered the i^{th} task

1.2 Cost function

Since we are looking for an optimal solution of our Constrained Optimization Problem (in the sense that it minimizes a cost function) we need to define our cost function.

Given:

Distance between two cities
$$c_1$$
 and c_2 :

Cost per kilometer for a given vehicle:

 $C_{km}(v)$

We define the cost function cost(P) as:

$$cost(P) = \sum_{v \in [1...N_v]} \left[\mathcal{C}_{km}(v) \cdot \sum_{path} dist(c_{i1}, c_{i2}) \right]$$

1.3 Constraints

Not all possible variable assignments correspond to a valid plan, a valid plan is a plan that satisfies all constraints $c \in C$, the constraints are the following:

The next task after a given task t cannot be itself $nextTask_t[t] \neq t$ The time variable must be coherent $nextTask_v[i] = j \Rightarrow time(j) = 1$ $nextTask_t[i] = j \Rightarrow time(j) = time(i) + 1$ # NULL values in nextTask_t All tasks must be delivered = # non-NULL values in nextTask_v $nextTask_v(k) = j \Rightarrow vehicle(j) = k$ From the definition of vehicle $nextTask_t[i] = j \Rightarrow vehicle[j] = vehicle[i]$ From the definitions of nextTask_t and vehicle $nextTask_v[i] = j \Rightarrow vehicle[j] = vehicle[i]$ From the definitions of nextTask_v and vehicle A vehicle cannot carry more than it's capacity $load(i) > capacity(k) \Rightarrow vehicle(i) \neq k$

2 Stochastic optimization

Because of the high computational complexity of the problem we search for a solution using a *Stochastic Local Search method (SLS)* that can be described by the following algorithm:

2.1 Stochastic optimization algorithm

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Algorithm 1: Stochastic Local Search for Constrained Optimization Problem S \leftarrow \text{GenerateInitialSolution}(< X, D, C, f >)

repeat
\begin{vmatrix} A_{old} \leftarrow A \\ N \leftarrow Succ(A_{old}, < X, D, C, f >) \\ A \leftarrow Choice(N, f) \\ \text{until } stable\_solution \lor timeout;

return A
```

This algorithm is implemented in the plan function of the CentralizedAssignement class.

2.2 Initial solution

We initialize our COP with a naïve solution to the pickup and delivery problem where we a

2.3 Generating neighbors

The set succ of neighbors is generated (implementation in the neighbours function of class Solution) by generating every possible exchange of two tasks in the order of pickup/delivery of tasks and every possible change of assignment of vehicle per task. So each succ is a set of size $N_t! + N_v \cdot N_t$.

2.4 Choosing neighbors

In order to minimize to cost we choose the neighbor that will replace the current plan in the next iteration according to the following law (let 0):

Algorithm 2: Choice(N,f)

With probability p:

return $plan \in N$ such that f is minimal

else

return $plan \in N$ where plan is randomly selected

3 Results

- 3.1 Experiment 1: Model parameters
- 3.1.1 Setting
- 3.1.2 Observations
- 3.2 Experiment 2: Different configurations
- 3.2.1 Setting
- 3.2.2 Observations