# Vehicle Routing Problem with a Helper

## EKESH KUMAR

## Bruce L. Golden

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## §1 Introduction

### §1.1 Preliminaries

National postal services seek to minimize the makespan of their deliveries by designing optimal routes for their vehicles. The *vehicle routing problem* (VRP) is an optimization problem in operations research in which one seeks to find an optimal set of routes for a fleet of vehicles. Vehicles spend time travelling to predefined service stops and spend a predefined amount of time at the service stop to satisfy a customers' demands. Variants of the vehicle routing problem seek to minimize differing objective functions such as the sum of the route lengths or the makespan of the routes.

In this paper, we explore a variant of the vehicle routing problem — the vehicle routing problem with a helper (VRPWH). Instead of having a fleet of vehicles, this problem studies the case in which a single truck driver and a helping agent work together to provide service at a set of service stops. More precisely, the VRPWH problem has a single truck driver that drives a delivery truck. The truck driver works with a helping agent that can either ride in the delivery truck with the truck driver or travel by foot by themselves. The driver of the delivery truck and the helping agent seek to visit and service a set of predefined service stops S along a long path. We impose the additional the constraint that the helping agent walks at a constant factor  $\alpha$  slower than the delivery truck. That is, if the delivery truck travels at a speed of v, then the helping agent walks at a speed of  $\frac{v}{\alpha}$ . Furthermore, we impose the additional constraint that the helping agent is only permitted to service a fixed subset  $S' \subseteq S$  of service stops that is known ahead of time.

We explore a few well-performing heuristics and metaheuristics for the problem at hand. Although this paper focuses on the case in which there is exactly one helping agent, the heuristics discussed in this report can easily be generalized to two or more helping agents.

### §1.2 Past Work

In 2008, [1] proposed an idea about having a helper dispatch tool for UPS's delivery system. The authors delineate how the decision tool could have a direct impact on savings while also providing a more efficient delivery solution. However, the authors do not mention any routing decisions that can be applied in practice.

In 2017, [2] introduced the dependent driver helper dispatching problem (DHDP), a related problem that "had never been studied before." The DHDP problem is similar to the vehicle routing problem problem: it seeks to minimize the cost required to service a set of customers with the assistance of a helper. With the assistance of a helper, the driver can share the delivery workload. Furthermore, the DHDP problem is similar to the VRPWH problem as both problems impose the constraint that the helping agent cannot move on its own. However, the DHDP problem differs from the VRPWH problem in that the helping agent is less efficient at providing service to service stops than the truck driver. This is not the case in the VRPWH problem: the truck driver and the helping agent require the same amount of time to provide service at a fixed service stop. The DHDP problem also differs in that the demands of a service stop can be split among both the primary driver and the helper. This is not permitted in the VRPWH problem — the truck driver or the helping agent must spend an uninterrupted amount of time to finish providing service at a service stop. Finally, the author of [2] only presents a single heuristic for the DHDP problem in general case. In this report, we provide several heuristics specialized for the case in which the underlying graph resembles a circular path.

## §2 Problem Definition

## §2.1 Assumptions and Constraints

Consider a mail route with a vehicle that makes numerous stops along a path. At each stop, the mail carrier exits the vehicle and delivers mail to a set of customers clustered along a closed walk on foot. After delivering the mail, the mail carrier returns to their vehicle and continues along the path. Once the truck driver and the helping agent have finished servicing all of the customers, they must return to their initial starting point.

In this problem, we seek to find the minimum time required to satisfy a set of customers' demands and return back to the starting point with the assistance of a helping agent.

We can model the problem by representing the n service stops as points  $p_1, p_2, \ldots, p_n$  on a circular path in the real plane. With this circular representation, an instance of the VRPWH problem is fully defined by the following parameters:

- 1. An radius parameter r, which is used to describe the circumference of the underlying circle.
- 2. A set of service stops  $S = \{p_1, \dots, p_n\}$ , where  $p_i$  denotes the location of the  $i^{\text{th}}$  service stop.
- 3. A time function  $t: S \to \mathbb{R}$ , which maps each service stop to the number of time units required to service the customer at that stop.
- 4. A parameter  $\alpha$  used to describe the factor by which the helping agent's walking speed is slower than the truck driver's driving speed. In other words, if the truck drives at a speed v, then the helping agent's walking speed is  $v/\alpha$ .

5. A subset  $S' \subseteq S$  of service stops describing where the helping agent can be utilized. If  $S' = \emptyset$ , then the helping agent cannot be used whatsoever. Conversely, if S' = S, we can employ the helping agent at any service stop.

We impose the following additional constraints:

- 1. The locations of the service stops are known ahead of time.
- 2. The time required to service each customer is known ahead of time.
- 3. The truck in which the truck driver and the helping agent begin in starts at the point  $p_0 = (-r, 0)$ .
- 4. The process of serving a customer cannot be split. At every service stop  $u \in S$ , either the truck driver or the helping agent must spend exactly t(u) uninterrupted time units and service the customer to completion.
- 5. The truck that the truck driver drives can only move in one of two directions (clockwise or counterclockwise) but not both. On the other hand, the helping agent is able to walk in either direction.

An example of an instance of the VRPWH problem with parameter r=1 is depicted below:

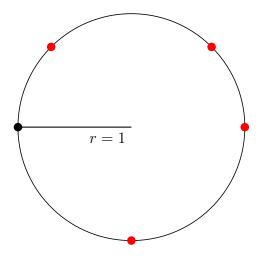


Figure 1: VRPWH Problem Instance

In this figure, the starting location of the truck is marked by a solid black dot. There are four service stops whose locations are marked with solid red dots.

For notational convenience, we define the following functions:

- 1.  $CW: S \mapsto S$  maps a service stop to the next service stop in the clockwise direction.
- 2.  $CCW: S \mapsto S$  maps a service stop to the next service stop in the counter-clockwise direction.

## §2.2 Hardness

We now present a proof demonstrating that the VRPWH problem is at least NP-Hard:

#### **Proposition 2.1**

The VRPWH Problem is at least NP-hard.

*Proof.* Consider an instance of the PARTITION problem in which we are given a multiset of positive integers  $S = \{a_1, \ldots, a_n\}$ , and we are tasked with deciding whether or not S can be partitioned into two subsets  $S_1$  and  $S_2$  such that the sum of the numbers in  $S_1$  equals that of  $S_2$ . It is well-known that PARTITION is an NP-Hard problem.

We will demonstrate that the VRPWH Problem on a Circle is also NP-hard by reducing an arbitrary instance of PARTITION to an instance of the VRPWH problem on a circle.

Consider an instance of PARTITION in which we are provided with a multiset of positive integers  $S = \{a_1, \ldots, a_n\}$ . We wish to determine whether S can be decomposed into the disjoint union of two multisubsets  $S_1$  and  $S_2$  such that  $\sum_{a_i \in S_1} a_i = \sum_{a_j \in S_2} a_j$ . We construct an instance of the VRPWH on a Circle problem whose optimal solution

will yield such a partitioning.

Construct an instance of the VRPWH on a Circle problem in which the underlying circle has radius r, and we have n service stops with service times  $a_1, \ldots, a_n$ . In the limit as  $r \to 0$ , the time required to go around the circle approaches zero. Furthermore, as  $\alpha \to 0$ , the helping agent's walking speed approaches that of the truck driver's driving speed. In other words, there is no advantage in having the helping agent dropped off at the service stops it visits over having the helping agent walking to that service stop on its own.

Thus, an optimal solution to such an instance of the VRPWH on a Circle problem becomes one in which the customers' demands are divided evenly. In other words, if  $T_1$  denotes the time that the truck driver spends servicing customers and  $T_2$  denotes the time that the helping agent spends servicing customers, then an optimal solution to the VRPWH on a Circle instance minimizes the quantity  $|T_1 - T_2|$ . By definition, the two multisubsets of S that attain the sums  $T_1$  and  $T_2$  form an optimal solution to the corresponding PARTITION instance.

## §3 Properties of Efficient Solutions

When designing efficient heuristics for the VRPWH problem, we take several observations into account. First, we present a few loose upper bounds for the problem at hand. Next, we several strategies that are weakly dominated (i.e., there is always another course of action that is at least as good as that strategy).

By taking these bounds and dominated strategies into consideration, we can significantly reduce our search space for an efficient solution.

### §3.1 Bounds

#### **Proposition 3.1**

Consider an instance of the VRPWH Problem on a Circle in which the underlying circle has radius r. Let the n service stops be  $a_1, \ldots, a_n$  with respective service times  $t(a_1), t(a_2), \ldots, t(a_n)$ . No solution to the VRPWH problem will take less than  $2\pi r + \frac{1}{2} \sum_i t(a_i)$  time.

Proof. If the truck driver were to service every customer by themselves, they would spend a total of  $\sum_i t(a_i)$  time. However, if this process were fully parallelized with the helping agent (i.e., for every time unit spent by the truck driver servicing a customer, the helping agent also spends a time unit servicing a customer), they would spend a total of  $\frac{1}{2}\sum_i t(a_i)$  time units. This does not any travel costs, which can only make this number increase. In particular, one can note that the travel costs will always be at least  $2\pi r$  as the truck driver can only move in one direction meaning that it must go around the circle at least once. Thus, we conclude that no solution to an instance of the VRPWH problem will ever take less than  $2\pi r + \frac{1}{2}\sum_i t(a_i)$  time.

With this result in mind, we seek to find efficient heuristics and results that will allow us to approximate an optimal solution in a computationally feasible manner.

## §3.2 Dominated Strategies

## Proposition 3.2 (Dominated Strategies: Helping Agent Idleness)

It is never advantageous for the helping agent to remain idle. For each solution in which the helping agent remains idle, there is a corresponding solution that can only improve upon the initial solution in which the helping agent always spends its time moving (possibly in the truck) or servicing a customer. *Proof.* Consider an instance of the VRPWH problem. Let IDLE denote a feasible solution in which the helping agent remains idle for a non-zero period of time. We will construct a new solution SOL that performs at least as well as IDLE in which the helping agent remains idle for exactly zero time units.

Consider an arbitrary point in time in IDLE in which the helping agent remains idle. Let  $p_k$  be the next point (either a service stop or the starting point) that the helping agent will visit. We consider two collectively exhaustive cases:

- 1. The helping agent will walk to  $p_k$  on its own. In this case, it is clearly disadvantageous to remain idle. By assumption, the helping agent is not already servicing a customer. Moreover,  $p_k$  is the next service stop (or initial point) that the helping agent will visit. By removing the idle time and instead starting to walk to  $p_k$  immediately, we obtain SOL, which improves upon IDLE by exactly the time that the helping agent remained idle before starting to walk to  $p_k$ .
- 2. The helping agent will be dropped off at  $p_k$  by the truck. In this case, we can construct SOL by removing the idle time of the helping agent and instead have the helping agent walk towards the truck. This is always possible because the helping agent is permitted to walk in both the clockwise and counterclockwise direction. Furthermore, this solution can only improve upon IDLE due to the inequality  $\alpha \leq 1$ .

There is one last special case that needs to be considered. Consider the case in which there is no "next point"  $p_k$  that the helping agent needs to visit. This can happen when the helping agent has finished servicing the subset of customers that it was assigned by a feasible solution, and it has returned to the initial starting point on its own. In this case, we can still construct SOL by having the helping agent walk in the direction that is opposite of the direction that the truck is moving in. Since the

truck cannot move in both directions, we are guaranteed that the helping agent will encounter the truck. Once the helping agent encounters the truck, it should ride in the truck with the truck driver for the remainder of the solution.

This course of action cannot make the preceding feasible solution any worse since the makespan of the solution is already being limited by the time that the truck driver requires to return back to the starting point.

#### **Proposition 3.3**

It is never advantageous for the helping agent to walk in the direction opposite of which that the truck is moving in.

*Proof.* TODO.  $\Box$ 

The next proposition illustrates that it is only optimal for the helping agent to re-enter the truck at a service stop.

#### **Proposition 3.4**

Suppose the helping agent is no longer in the truck with the truck driver. There is an optimal solution in which the helping agent only re-enters the truck at a service stop.

*Proof.* Suppose, for the sake of contradiction, that the proposition is not true. In other words, this means that there exists an instance of the VRPWH problem in which it is strictly advantageous for the helping agent to re-enter the truck driver's truck at a point that is not a service stop.

Let  $a_1, \ldots, a_n$  denote the n service stops, labeled in a clockwise manner around the circle. Suppose the helping agent re-enters the truck between service stops  $a_i$  and  $a_{i+1}$ , where  $a_{i+1}$  is taken to be the starting point when i = n. Since the truck can

only move in one direction along the circle (either clockwise or counterclockwise), there are exactly two cases to consider.

In the first case, we consider the scenario in which the truck is driving towards the helping agent. This means that the helping agent must have walked ahead and ended up in front of the truck at some previous point in time. An optimal solution would only have a helping agent walk ahead if it the helping agent were to service a service stop beyond the truck's current position. However, by Proposition 3.3, it is never strictly advantageous for the helping agent to walk backwards after walking ahead. If the helping agent stays where it is after completing the demands at a service stop, the result follows. On the other hand, consider the case in which the helping agent walks forward after completing the demands at a service stop. This can never improve the final solution if the helping agent is being picked up by the truck because the helping agent will always arrive at its subsequent destination at the same time as the truck. Thus, having the helping agent wait at the last service stop it completes results in a corresponding solution that is at least as good as one in which the helping agent walks forward after servicing its last service stop.

Next, we consider the case in which the helping agent walks towards the truck. This can only be the case if the helping agent were left behind at some point. An optimal solution would only leave the helping agent behind if it the helping agent was being used to satisfy the demands at some service stop that the truck driver drove past by. By extension, the truck driver must have gone on to service some other customer. If this were not the case, the truck driver would have just serviced the customer that the helping agent was left behind to service. Consider the last customer that the truck driver services prior to the helping agent catching up to the truck. In an optimal solution, it only makes sense for the truck to wait at that last service stop since moving ahead can only increase the gap between the helping agent and the

truck (the helping agent's walking speed is less than or equal to the driving speed of the truck). By assumption, we are considering the last service stop prior to the helping agent catching up, which means that we can only decrease the overall time of our solution by having the truck wait at the last service stop that the truck driver processed.

#### 

## §3.3 Undominated Strategies

#### **Proposition 3.5**

Having the truck driver wait is *not* a dominated strategy (i.e., it may be optimal for the truck driver to remain idle).

*Proof.* It suffices to construct an instance of the VRPWH problem in which having the truck driver wait is undominated. Consider the following instance in which  $\alpha = 1 - \epsilon$  for some suitable choice of  $\epsilon > 0$  (i.e., the walking speed of the helping agent is just under that of the driving speed of the truck):

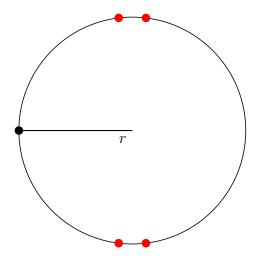


Figure 2: Primary Deliverer Waiting is Undominated

We can label the four service stops in clockwise order as  $p_1, p_2, p_3$ , and  $p_4$ . We assume that the helping agent can be utilized at each of the four service stops (that is,

S'=S). Furthermore, we assume  $t(p_i) \ll 2\pi r$ , which means that the time required to service the customers is inconsequential when compared to the time required to travel around the circle.

With these parameters in mind, consider the following construction of a solution, denoted IDL, in which it makes sense for the truck driver to remain idle:

- The truck driver and the helping agent begin at the point (-1,0), denoted by the solid black circle. They begin to travel in the clockwise direction.
- Upon encountering the first service stop  $p_1$ , the helping agent is dropped off to service the customers at that station. The truck driver continues along its route in the clockwise direction until it reaches the second service stop,  $p_2$ . The truck driver services the customer at this second station.
- By construction, we can come up with an instance of the VRPWH problem in which the truck driver and the helping agent finish servicing  $p_1$  and  $p_2$  at exactly the same time.
- If  $r \gg 0$ , it would take a notable amount of more time for the helping agent to travel to the next service stop on its own than if it were dropped off by the truck. Furthermore, it would be too costly for the truck driver to make a full lap around the circle to pick up the helping agent. Since the distance between  $p_1$  and  $p_2$  is relatively small, the merits of having the truck driver wait for the helping agent to arrive at  $p_2$  are clear.

## §4 Heuristics

### §4.1 Baseline Heuristic

To begin our discussion, we introduce a baseline heuristic that does not utilize the helper vehicle at all. The purpose of the baseline heuristic is to simply provide a baseline so that we can easily compare the performance of other heuristics relative to this one. This heuristic is summarized below:

- 1. The truck begins at the point (-r,0) with the truck driver and the helping agent in the truck.
- 2. While at least one unvisited service stop exists, the two vehicles travel clockwise around the circle to the next unserviced service stop. The truck driver satisfies the demand at the service stop while the helping agent remains idle.

In a real-world setting, the baseline heuristic produces the cost of servicing all of the customers when we do not have the option to deploy the helper vehicle at all. It acts as a "control" in our subsequent comparisons.

The method by which we can improve upon the baseline heuristic is by dispatching the helper vehicle at some service stop at which it is permitted and subsequently servicing some number of customers at service stops before returning to the helper vehicle.

## §4.2 Look-ahead Heuristic

#### §4.2.1 Heuristic

The next heuristic we will discuss is called the **look-ahead** heuristic. The look-ahead heuristic decreases the total time spent by attempting to serve future service stops with the primary vehicle while the helper vehicle stays behind to serve a service stop with a large demand. The heuristic is summarized below:

- 1. The principal and helper vehicles both start at 0 on the underlying number line.
- 2. While at least one unvisited service stop exists, the two vehicles travel together along the one-dimensional number line to the next unserved service stop  $a_i$ . There are now two cases to consider:
  - a) If  $a_i \notin S'$ , then we service the customers at the service stop immediately using the primary vehicle. In the meantime, the helper vehicle remains idle for  $t(a_i)$  minutes as it cannot move on its own.
  - b) If  $a_i \in S'$ , then we greedily exercise our option to dispatch the helper vehicle. In the meantime, we "look ahead" with our primary vehicle, and we service as many service stops ahead of the current one as possible such that the helper vehicle's idle time does not increase at all.

#### §4.2.2 Demonstration

Consider the following instance of the VRPWH Problem on a Circle with r=1 and  $\alpha=0$  in which the locations of the service stops are marked with solid red dots. The starting location of the truck is the solid black dot. For the sake of simplicity, we assume the helping agent can be dispatched at any one of the four service stops.

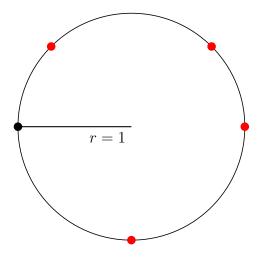


Figure 3: Look-ahead Demonstration

There are three service stops depicted in the instance above:  $a_1 = (0,1), a_2 = (-\sqrt{2}/2, \sqrt{2}/2), a_3 = (\sqrt{2}/2, \sqrt{2}/2),$  and  $a_4 = (0, -1)$ . Suppose the service times are given by  $t(a_1) = 5$ ,  $t(a_2) = 2$ ,  $t(a_3) = 2$ , and  $t(a_4) = 20$ . Furthermore, suppose that the time required to move a single unit on the Euclidean plane is equal to one time unit. Thus, we spend a total of  $2\pi$  time units travelling around the entire circle once.

In this instance of the VRPWH on a Circle problem, the look-ahead heuristic would proceed as follows:

- 1. The truck begins at the point (-1,0) with both the truck driver and the helping agent.
- 2. The truck travels to the point  $a_1$  at a cost of  $\pi/3$  time units. Since  $a_1$  can be serviced by the helping agent, we greedily exercise our option to dispatch the helping agent.
- 3. While the helping agent spends  $t(a_1) = 5$  uninterrupted time units servicing  $a_1$ , the truck driver travels  $\pi/2$  units to  $a_2$  in the truck, services it in two time units, travels  $\pi/2$  units back to  $a_1$ , and waits for the helping agent to finish servicing  $a_1$ . That is, the truck driver "looks ahead" to see if it can service future service stops while the helping agent stays behind servicing a different service stop.
- 4. The helping agent and truck driver travel together in the truck to  $a_3$  in a total of  $5\pi/4$  time units. The helping agent services  $a_3$  while the truck driver remains idle for  $t(a_3) = 2$  time units.
- 5. The helping agent and the truck driver travel together in the truck to  $a_4$  in a total of  $\pi/2$  time units. The helping agent services  $a_4$  in  $t(a_4) = 20$  time units while the truck driver remains idle.

6. Finally, the helping agent and the truck driver travel together to their starting point in  $\pi/2$  units. All customers' demands have been satisfied, so the heuristic terminates here.

The total cost of this solution is

$$\frac{\pi}{4} + 5 + \frac{3\pi}{4} + 2 + \frac{\pi}{2} + 20 + \frac{\pi}{2} = 27 + 2\pi.$$

Despite the straightforward manner of this heuristic, one can show that it performs well under certain circumstances. For example, it is easy to see that in an instance of the VRPWH on a Circle problem in which the circle has radius r in which the time required to move one unit of distance is equal to one unit of time, the total time spent travelling will never exceed  $2\pi r$ .

The lookahead heuristic improves upon the baseline heuristic by exercising the option to service customers ahead of the one that the helping agent is servicing in such a way that the helping agents' idle time does not increase. Computational results demonstrate that the lookahead heuristic actually manages to exercise this "lookahead" more than one might expect. For example, the table below demonstrates some of these results for varying radii when the number of service stops is fixed at 100 and the distribution of demands follows a standard normal distribution. Note, however, that we do not permit negative service times. Thus, if the generated value is below 0 or above  $2\mu$ , it is discarded and generated again. This procedure preserves symmetry and ensures demands are non-negative.

The table below depicts the mean number of service stops skipped across  $10^6$  random instances of the VRPWH on a Circle with a varying radii. In this instance, we assume that the helping agent can service any one service stop with probability p = 0.50.

# of Service Stations	Radius	Mean # of Skipped Stations
100	1	17.620
100	2	16.897
100	5	14.516
100	10	11.324
100	25	6.508
100	50	3.772

Figure 4: Average Number of Skipped Stations in Randomly Generated VRPWH Instances with p=0.50

As demonstrated in the table above, even when the ratio of the number of service stops to the circumference of the circle in the VRPWH instance is small, we can still improve upon the baseline heuristic more than one may expect.

## §4.3 Knapsack Look-ahead Heuristic

#### §4.3.1 Motivation

The next heuristic we present is the **knapsack look-ahead heuristic**, which we motivate through an example.

Recall that the look-ahead heuristic greedily exercises the option to dispatch the helper vehicle whenever possible. The primary vehicle subsequently serves as many service stops beyond the current one by considering the service stops in order of their distance to the current service stop. However, this approach may have drawbacks. Consider an instance of the VRPWH on a Circle problem, and suppose the following represents a small segment of the circle (curvature is not depicted here):

Figure 5: Knapsack Look-ahead Example

Once again, for simplicity, we assume that the helper vehicle is allowed to be dispatched at any of the three service stops. Suppose the time required to satisfy each service stop's demands is given by t(0) = 10, t(1) = 1, and t(2) = 7.

The look-ahead heuristic would dispatch the helper vehicle at 0. While the helper vehicle serves this service stop, the primary vehicle would travel to 1, service it, and return. The primary and helper vehicles would then travel together to 2, and the primary vehicle would service it while the helper vehicle remains idle. The total cost of the look-ahead heuristic in this instance is t(0)+2+t(2)=19 time units.

A better solution would be to dispatch the helper vehicle at 0 while the primary vehicle travels and services the station located at 2. The primary vehicle could then return to pick up the helper vehicle at 0, they could both travel to 1, service it, and finally travel to 2 to finish. The total cost of this solution would be t(0)+1+t(1)+1=13 time units.

In summary, we see that the lookahead heuristic can be improved by not just servicing subsequent service stops in order by their distance to the current service stop but instead servicing subsequent service stops so that the time spent by the primary vehicle is approximately equal to that of the helper vehicle. Fundamentally, the problem can be reduced to finding a subset of numbers that sum up closest to a target number.

#### §4.3.2 Heuristic

The knapsack look-ahead heuristic operates as follows:

- 1. The truck driver and helping agent begin in the truck located at the point (-r,0).
- 2. The truck driver and helping agent travel together in the clockwise direction. While at least one unvisited service stop exists, the two vehicles travel together along the one-dimensional number line to the next unserved service stop  $a_i$ . There are now two cases to consider:
  - a) If  $a_i \notin S'$ , then we service the service stop immediately using the primary vehicle. In the meantime, the helper vehicle remains idle for  $t(a_i)$  minutes as it cannot move on its own.
  - b) If  $a_i \in S'$ , then we greedily exercise our option to dispatch the helper vehicle. In the meantime, we "look ahead" with our primary vehicle, and we service a subset of service stops ahead of the current one such that the time spent (round-trip time plus service time) by the primary vehicle is as close as possible (but not more than)  $t(a_i)$ . That is, the helper vehicle's idle time does not increase at all.

## §4.4 Neighbor-Expansion Heuristic

#### §4.4.1 Motivation

The next heuristic we present is the **neighbor-expansion heuristic**, which is a greedy algorithm that uses some elements from both the look-ahead and knapsack look-ahead heuristics.

The preceding two heuristics we discussed both sought to decrease the total time spent by servicing some set of customers by "looking ahead" with the primary vehicle while the helper vehicle stayed behind at a particular service stop.

Unfortunately, not every optimal solution takes this form. It can be disadvantageous

for us to immediately dispatch the primary vehicle at service stops where the helper vehicle cannot be dispatched. Consider the following example:



Figure 6: Neighborhood-Expansion Example

Suppose we have the option to dispatch the helper vehicle at both 0 and 1. If t(0) = 5 and t(1) = 15, then dropping the helper vehicle off at 1, returning to service 0, and returning to 1 would produce a solution that takes 16 time units. On the other hand, both the look-ahead heuristic and the knapsack look-ahead heuristic would take 21 time units. This demonstrates that it may not always be optimal to "look ahead" with the primary vehicle while the helper vehicle stays behind.

In order to combat this issue, we propose a heuristic that considers both "looking ahead" and "looking behind" as possible choices.

#### §4.4.2 Heuristic

Before introducing the neighbor-expansion heuristic, we introduce some terminology.

**Definition 4.1.** For each service stop  $a_i \in S'$  at which the helper vehicle can be dispatched at, we define a set of **neighbors**  $N_{a_i}$ . The set  $N_{a_i}$  contains the set of service stops for which it is possible for the primary vehicle to visit, service, and return to  $a_i$  without exceeding a total time of  $t(a_i)$ .

Essentially, the neighbors of a service stop  $a_i \in S'$  represent the set of service stops that the primary vehicle can service while the helper vehicle services  $a_i$  without leaving the helper vehicle idle at  $a_i$ .

The neighbor-expansion heuristic works as follows:

1. We begin with a set  $A = \{a_1, a_2, \dots, a_n\}$  containing every service stop.

- 2. While there are service stops in A that can be serviced by the helper vehicle (i.e., while  $A \cap S' \neq \emptyset$ ), we perform the following:
  - a) For every service stop  $a_i \in A \cap S'$ , we compute the set of neighbors  $N_{a_i}$ .
  - b) Next, for each set of neighbors  $N_{a_i}$  we compute a subset  $N'_{a_i} \subseteq N_{a_i}$  of neighbors such that servicing all of the customers in  $N'_{a_i}$  (with travel costs included) is as large as possible but does not exceed  $t(a_i)$ .
  - c) Among all of the computed values, we take the  $a_i$  for which the difference between  $t(a_i)$  and the time needed to service all of the customers in  $N'_{a_i}$  is minimal. Ties are broken arbitrarily.
  - d) We remove both  $a_i$  and all of the service stops in  $N'_{a_i}$  from A, and we repeat this process until  $A \cap S' = \emptyset$ .

Upon termination, the set A represents the service stops that should be served immediately when the primary and helper vehicles visit them for the first time. On the other hand, when we visit some  $a_i \notin A$  for the first time, we should not immediately process it. Instead, when we visit the service stop  $a_k$  for which  $a_i \in N_{a_k}$  holds, then we should dispatch the helper vehicle at  $a_k$ , and the primary vehicle should service every service stop in  $N_{a_k}$ , which would include servicing  $a_i$ .

#### §4.4.3 Demonstration

As a brief demonstration, consider an instance of the VRPWH Problem on a Circle (curvature not shown) in which the time required to travel one unit of distance equals one time unit:

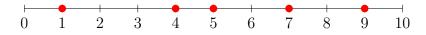


Figure 7: Neighborhood-Expansion Demonstration

For simplicity, we assume the helper vehicle can be dispatched at any one of the four service stops. Furthermore, suppose we have t(1) = 10, t(4) = 7, t(5) = 12, t(7) = 1, t(9) = 1. The neighbor-expansion heuristic proceeds as follows:

- 1. First, we initialize a set  $A = \{1, 4, 5, 7, 9\}$  containing the locations of our service stops.
- 2. Next, we compute the set of neighbors for each of the four service stops. We have N<sub>1</sub> = Ø (we cannot travel, service, and return from some other service stop while the helper vehicle remains at 1), N<sub>4</sub> = {7}, N<sub>5</sub> = {4, 7, 9}, N<sub>7</sub> = Ø, and N<sub>9</sub> = Ø.
- 3. Subsequently, for each  $N_{a_i}$ , we compute a subset  $N'_{a_i}$  of service stops for which the time needed to service all of the customers in  $N'_{a_i}$  (with travel costs included) is as large as possible but does not exceed  $t(a_i)$ . One can find  $N'_1 = \emptyset$ ,  $N'_4 = \{7\}$ ,  $N'_5 = \{4\}$ ,  $N_7 = \emptyset$ , and  $N_9 = \emptyset$ . Among all of the computed values, we select service stop 4 since |t(4) 3 t(7) 3| = 0. From here, we remove 4 and 7 from A, and we iterate again.
- 4. In the second iteration, we follow the same procedure, and we end up removing 5 and 9 (the other neighbor sets are empty).
- 5. The final solution operates generated by this heuristic operates as follows:
  - a) The primary and helper vehicles begin at 0.
  - b) The primary and helper vehicles move together to 1 at which the primary vehicle immediately spends t(1) = 10 uninterrupted minutes servicing the required demand.
  - c) The primary and helper vehicle move together to the service stop at 4.

    The helper vehicle is dispatched here while the primary vehicle travels to 7, services 7, and returns to 4.

- d) The primary and helper vehicle move together to the service stop at 5.

  The helper vehicle is dispatched here while the primary vehicle travels to
  - 9, services 9, and returns to 5.
- e) The primary and helper vehicle navigate to 10. All of the demands have been satisfied.

## §4.5 Randomized-Expansion Heuristic

#### §4.5.1 Motivation

The computational results in the subsequent section demonstrate that the proposed heuristics already appear to perform much better than the baseline solution. Notably, the knapsack look-ahead and the neighbor-expansion heuristics appear to perform better than the standard look-ahead heuristic.

However, this improvement comes at a price: both the knapsack look-ahead and neighbor-expansion heuristics are more involved, and they perform constant-time operations on the same service stop several times. Although the two heuristics appear to perform very quickly when the service stops and service times are generated according to naturally arising probability distributions, step 2 of either algorithm runs in worst-case exponential time. When working with several service stops, these approaches may be computationally infeasible.

This motivates us to seek different approaches. In this section, we present a randomized metaheuristic that is structurally similar to the neighbor-expansion heuristic. However, it speeds up and potentially improves upon the neighbor-expansion heuristic by utilizing stochastic decision making. We refer to this approximation as the randomized-expansion heuristic.

#### §4.5.2 Heuristic

In the knapsack look-ahead and neighbor-expansion heuristics, the process of finding the neighbors of some service stop  $a_i \in S'$  is fast (i.e., they can be computed in polynomial time), but computing the desired optimal subset of S'. The randomizedexpansion heuristic improves upon the latter step by making randomized decisions.

The randomized-expansion heuristic works as follows:

- 1. We begin with a set  $A = \{a_1, a_2, \dots, a_n\}$  containing every service stop.
- 2. While there are service stops in A that can be serviced by the helper vehicle (i.e., while  $A \subseteq S' \neq \emptyset$ ), we perform the following:
  - a) For each service stop  $a_i \in A \cap S'$ , we compute the set of neighbors  $N_{a_i}$ .
  - b) Next, for each set of neighbors  $N_{a_i}$ , we generate a random subset  $N'_{a_i}$  of  $N_{a_i}$ . If the time required to service all of the service stops (with travel costs included) exceeds  $t(a_i)$ , we discard the generated subset. This step is guaranteed to terminate because any singleton would satisfy the desired properties.<sup>1</sup>.
  - c) For each random subset  $N'_{a_i}$ , we compute the absolute difference between  $t(a_i)$  and the time required to service (with travel costs) each of the service stops in  $N'_{a_i}$  and return to  $a_i$ .
  - d) Steps a, b, c are repeated a fixed number of times. Among all of the computed absolute differences, we take the  $a_i$  and  $N'_{a_i}$  for which the absolute difference was minimal.
  - e) We remove both  $a_i$  and all of the service stops in  $N'_{a_i}$ , and we repeat Step 2 until  $A \cap S' = \emptyset$ .

<sup>&</sup>lt;sup>1</sup>The expected number of steps needed to terminate is polynomial in n.

- 3. At this point, we can construct a solution just as we would in the neighborexpansion heuristic.
- 4. We can repeat this entire algorithm several times, and we can take the best objective value found from all repetitions.

Introducing randomness into the expansion heuristic serves two purposes:

- 1. First, we are able to significantly improve the running time of the previous algorithms that we proposed.
- 2. Next, there is a high chance that some of the found solutions will differ from what the other proposed algorithms found. These differences are analogous to the concept of mutations in genetic algorithms.

## §5 Computational Results

In order to benchmark the proposed algorithms, we generated random instances of the VRPWH on a Circle. For a fixed radius r > 0 and a fixed number of points N, the locations of the N service stops were generated randomly on the circle  $x^2 + y^2 = r^2$ . More precisely, this is done by generating a random angle  $\theta$  by sampling a uniform random variable  $Z \sim U(0,1)$  and setting  $\theta = 2\pi Z$ . From here, we can easily find our random point (x,y) on the circumference of the circle by setting  $x = r\cos(\theta)$  and  $y = r\sin(\theta)$ .

On the other hand, the service times of the N service stops were generated according to a "restricted"  $N(\mu, \sigma^2)$  distribution in which service times above  $2\mu$  or below 0 were regenerated in order to preserve symmetry and avoid negative service times. In each instance, a parameter p was also specified, which was used to denote the probability of any one service stop belonging to the subset of service stops that the helper vehicle can be dispatched at. Each computed cost is calculated by averaging

over  $10^6$  simulations each.

N	r	p	$\mu$	$\sigma^2$	$\alpha$	Baseline	Lookahead	Knapsack	NE	RE
10	1	0	1	1	1	2.351	1.641	1.476	1.752	1.380
10	1	0.25	1	1	1	2.359	1.624	1.479	1.877	1.291
10	1	0.50	1	1	1	2.416	1.681	1.424	1.829	1.400
10	1	0.75	1	1	1	2.473	1.760	1.592	1.843	1.452
10	1	1	1	1	1	2.465	1.737	1.539	1.821	1.475
25	1	0	2	1	4	10.173	7.745	7.126	8.076	5.885
25	1	0.25	2	1	4	10.6513	8.271	8.425	8.439	6.146
25	1	0.50	2	1	4	10.295	7.881	6.095	7.909	5.896
25	1	0.75	2	1	4	10.582	8.191	6.794	8.104	6.054
25	1	1	2	1	4	10.298	7.871	8.221	7.812	5.945
50	1	0	3	1	2	24.523	17.612	16.783	18.724	14.045
50	1	0.25	3	1	2	24.345	17.535	15.504	18.625	13.895
50	1	0.50	3	1	2	24.659	17.791	15.071	18.925	14.406
50	1	0.75	3	1	2	24.485	17.713	15.022	18.723	13.997
50	1	1	3	1	2	24.713	17.822	15.191	18.957	14.081
75	1	0	5	1	$\frac{4}{3}$	43.829	26.052	57.642	33.586	24.999
75	1	0.25	5	1	$\frac{4}{3}$	43.905	26.111	24.342	23.610	21.067
75	1	0.50	5	1	$\frac{4}{3}$	43.905	26.261	25.342	23.614	21.069
75	1	0.75	5	1	$\frac{4}{3}$	43.904	26.923	25.276	23.567	21.109
75	1	1	5	1	$\frac{4}{3}$	43.855	26.051	24.012	23.535	21.079
100	1	0	10	1	$\infty$	83.617	36.012	34.924	32.691	34.950
100	1	0.25	10	1	$\infty$	83.783	35.241	32.792	34.155	33.826
100	1	0.50	10	1	$\infty$	83.698	35.012	31.461	33.950	32.891
100	1	0.75	10	1	$\infty$	83.698	35.212	31.421	32.945	27.121
100	1	1	10	1	$\infty$	78.746	29.521	28.192	43.421	30.991

Figure 8: Performance of Heuristics on Random Instances of the VRPWH Problem. N denotes the number of service stations; r is the radius of the underlying circle; p denotes the probability of a fixed service stop belonging to the subset of service stops that the helping agent can service;  $\mu$  and  $\sigma^2$  denote the mean and variance of the underlying normal distribution used to generate the service times of each stop;  $\alpha$  denotes the constant factor by which the helping agent walks slower than the truck.

From the computational results, it is immediately clear that all of the proposed heuristics outperform the baseline heuristic. This is expected since the baseline heuristic does not utilize the helping agent at all; it simply provides us with an initial value that we can use to compare to the other heuristics.

We see that the simple lookahead heuristic always outperformed the baseline heuristic but it never outperformed any of the other heuristic. Despite its poor performance relative to the other heuristics, the lookahead heuristic is not computationally intensive compared to the other heuristics. On the other hand, the knapsack heuristic typically outperforms the lookahead heuristic but it is extremely computationally intensive. This stems from the fact that the knapsack heuristic is a pseudopolynomial dynamic programming algorithm, and its asymptotic runtime grows according to the sum of the service demands.

The neighbor expansion heuristic was also computationally intensive; however, it was not nearly as intensive as the knapsack heuristic. On most instances, it seemed to perform worse than both the lookahead and knapsack heuristics. Finally, the randomized expansion heuristic took a Monte Carlo approach, and it seems to perform the best out of all of the proposed heuristics. The running time of the randomized expansion heuristic typically fell between that of the neighbor expansion heuristic and the knapsack heuristic.

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