



# An Artificial Intelligence-based software module for the optimization of collaborative delivery in last-mile logistics

Jenny Fajardo-Calderin<sup>1</sup>, Antonio D. Masegosa<sup>1,2</sup>, Maria Pilar Elejoste<sup>1</sup>, Asier Moreno-Emborujo<sup>1</sup>, Xabier Cantero-Lopez<sup>1</sup> and Ignacio Angulo<sup>1</sup>

- 1. Deusto Institute of Technology (DeustoTech), Faculty of Engineering, University of Deusto, 48007 Bilbao, Spain;
  - 2. IKERBASQUE, Basque Foundation for Science, 48013 Bilbao, Spain Corresponding author: <a href="mailto:fajardo.jenny@deusto.es">fajardo.jenny@deusto.es</a>

Abstract: This paper presents a route delivery planning and simulation module that forms a core part of the ICT Platform of the H2020 SENATOR project, which aims to enhance the sustainability of cities by developing a new urban logistic model. The module utilizes AI-based optimization algorithms to support the matching of supply and demand, identify the best fleet mix, and estimate the best delivery route based on real-time conditions. It also allows last-mile delivery planning using different transport modes, inter-modality, and driving restrictions, and simultaneously optimizes different performance indicators. The paper provides a detailed description of the AI-based optimization method and the architecture and components of the software module. Finally, the software module is validated in two scenarios (current operations and implementation of a Low Emission Zone) using real shipment data from a postal operator company in the living lab that the SENATOR project is implementing in Zaragoza.

**Keywords:** last-mile logistics, vehicle routing problem, optimization, dynamic planning, multi-modal, collaborative delivery.

Conference Topic(s): distributed intelligence last mile & city logistics; PI modelling and simulation; technologies for interconnected logistics (5G, 3D printing, Artificial Intelligence, IoT, machine learning, augmented reality, blockchain, cloud computing, digital twins, collaborative decision making)

**Physical Internet Roadmap** (Link): Select the most relevant area(s) for your paper:  $\square$  PI Nodes,  $\square$  PI Networks,  $\boxtimes$  System of Logistics Networks,  $\square$  Access and Adoption,  $\boxtimes$  Governance.

#### 1 Introduction

The growth of last-mile logistics is a continuous trend, driven mainly by urbanization and changes in consumer behaviour. The surge in online retailing, e-groceries, and e-commerce has contributed to this phenomenon (European Commission, 2023). This trend has resulted in an increase in freight traffic in urban areas, which negatively impacts the sustainability and livability of our cities. The additional traffic generated by vehicles for deliveries leads to congestion and emissions, with CO2 accounting for 25% and PM and NOx accounting for 30-50%. Furthermore, heavy vehicles also reduce road safety (European Commission, 2020). On top of that, the COVID-19 pandemic has further accelerated the growth of online purchasing and logistic innovations(DHL, 2023). However, despite efforts to improve logistics sustainability, the challenges of the entire process are still subject to debate. Nevertheless, new technologies, processes and distribution strategies offer significant potential for enhancing the impact of last-mile deliveries in urban areas.

With these challenges in mind, the SENATOR project aims to create a new urban logistic model for enhancing the sustainability of cities. For this purpose, the project will develop a smart network operator, as a control tower supported on an ICT Platform that will work as a support tool for decision-making, integration and planning of all logistics operations. In consequence, it will minimize the negative impacts that this distribution causes in the cities and will constitute an effective means of collaboration between agents (citizens, operators, carriers, and administrations).

The objective of this paper is to present one of the core parts of the mentioned ICT Platform. Concretely, the route delivery planning and simulation module whose aim is to support the matching of supply (vehicle, transport operators, etc.) with demand; to identify the best fleet mix (e.g. fuel, electric or zero-emission vehicles, cargo-bikes) to fulfil the customer demands

- **VRP Service from ArcGIS**<sup>5</sup> is a commercial service developed by ESRI to address different routing problems. It can be accessed in different ways such as JavaScript APIs and SDKs in different programming languages.
- **Circuit**<sup>6</sup> is a commercial tool available on web service, Android and iOS platforms, that allows the optimization of up to 1000 stops considering stop time windows, first and last stop, and priority levels. Its main use cases are Driver Tracking, Local Delivery, Route Planning, Proof of Delivery, and Courier Management.
- LOCUS<sup>7</sup> is a commercial tool that allows the planning and optimization of routes and vehicle assignation for orders considering problems such as Travelling Salesman, Vehicle Routing and Knapsacking. LOCUS implements exact, heuristic, and hybrid algorithms. Its main use cases are Last-Mile Delivery Routing, Field Service Dispatch Planning, Dynamic Route Planning and Optimization, Territory-Based Route Planning, and Reverse/Returns Logistics.
- **OptaPlanner**<sup>8</sup> is an open-source tool developed in Java that implements several optimization problems, including VRP, Capacitated VRP, and VRP with Time Windows. It also allows integration with Google Maps and OpenStreetMap.
- **HERE**<sup>9</sup> is a commercial tool that provides a route planning API to solve the VRP, implementing the Capacitated VRP, VRP with Time Windows, Multi-Depot VRP, Open Vehicle Routing, Heterogeneous or Mixed Fleet VRP, and Pickup and Delivery VRP. It allows the calculation of routes using real-time and historical traffic information and the re-planning of routes in real time if new orders appear.
- **GraphHopper**<sup>10</sup> is an open-source software tool developed in Java that uses JSprit as the route optimization engine. It provides an API to solve a variety of vehicle routing problems, including the Traveling Salesman optimization problem, and all the VRP variants implemented in JSprit. Its main advantages are the possibility of designing vehicle types and defining time windows and service times for drivers.

However, none of the tools discussed above offer at the same time the functionalities of dynamic route optimisation, multi-modal fleet optimisation, inter-modal and/or transfer route optimisation, multi-objective optimisation and consideration of driving constraints.

# 3 Artificial Intelligence based software module for collaborative last-mile logistics

The software module for collaborative last-mile logistics is based on the well-known Rich Vehicle Routing Problem (Lahyani, Khemakhem, & Semet, 2015) which is a class of optimization problems that represent some or all aspects of a real-world application of vehicle routing including optimization criteria, constraints, and preferences. These problems deal with more realistic optimization functions, uncertainty, and dynamism, along with a wide variety of real-life constraints related to time and distance factors, and the use of heterogeneous fleets. Specifically, this software module internally implements a more realistic and complex model of the Rich Vehicle Routing Problem than those currently available in the literature as it can simultaneously incorporate the following aspects:

<sup>8</sup> https://www.optaplanner.org/

 $<sup>^{5}\ \</sup>underline{\text{https://desktop.arcgis.com/es/arcmap/latest/extensions/network-analyst/vehicle-routing-problem.htm}$ 

<sup>&</sup>lt;sup>6</sup> <u>https://getcircuit.com/</u>

<sup>&</sup>lt;sup>7</sup> https://locus.sh/

<sup>9</sup> https://developer.here.com/products/tour-planning

<sup>10</sup> https://github.com/graphhopper/graphhopper

- **Dynamism:** given that some elements of the delivery planning may change over time, the software module allows the re-adjustment of the planning (e.g. last-minute orders that may appear, a vehicle breaks down and it is necessary to assign orders to other routes)
- **Multi-modality:** the module allows last-mile delivery planning using different transport modes (e.g. walking, bikes, motorbikes, vans, etc.).
- **Inter-modality:** the software allows that one shipment can be transported by different transport modes along its route between the depot/pick-up location and the destination.
- **Multi-objective:** the module deals with the simultaneous optimization of different performance indicators (e.g. distance, time, emissions, etc.)
- **Driving restrictions:** the component allows the modelling of areas with access restrictions to specific vehicles (e.g. pedestrian zones, low emission zones, etc.)

In order to solve this highly complex Rich Vehicle Routing Optimization model, a specific sort of Artificial Intelligence techniques has been used. Concretely, we have used metaheuristics (Potvin & Gendreau Jean-Yves, 2019) because of their high efficiency and efficacy for this type of problem (Goel & Bansal, 2019). In a more specific way, the optimization algorithm designed is a hybrid metaheuristic(Gu, Cattaruzza, Ogier, & Semet, 2019) based on Large Neighbourhood Search (Pisinger & Ropke, 2019).

In the following subsections, we will describe the Artificial Intelligence-based resolution algorithm and the architecture of the software module.

# 3.1 Artificial Intelligence-based resolution algorithm

The Large Neighborhood Search (LNS) algorithm was employed in the resolution of the presented module, utilizing a metaheuristic in which the neighbourhood of a solution is implicitly defined through the use of destroying and repairing operators. The destroy operator eradq0ETa octurr ci0f T/Fo2019(th)-11(e)4()20(c)u1(r)-4(re)7(0f 839(solut)-4(i,0f 831(w3(le)-7(2)40(the2229(

# 3.2 Architecture and Components

The architecture of the proposed module is based on the distinct responsibilities and functionalities of its various components, each with its own distinct behaviours. It depicts the different components and their relationships. Certain components serve a specific purpose and have been designed to consolidate the optimization model. Within the optimization 2 Tfn4.09()] TJET

# 4 Experiments and Results

In order to validate the results of the system presented in this article, we have designed two scenarios: **Scenario 1**) which reflects the current operations of a real postal operator in the city of Zaragoza, and **Scenario 2**) which corresponds to the implementation of a Low Emission Zone in the centre of the city of Zaragoza. In addition, for each of the scenarios considered, three different fleet compositions have been simulated in terms of vehicle electrification ratio, with the aim of showing the capabilities of the tool for measuring the impact that different levels of fleet electrification would have on the two scenarios defined.

Below we provide more details about the experimental framework designed and the results obtained for the two scenarios designed and the alternative fleet compositions

### 4.1 Experimental framework

This section aims to define the two scenarios considered for the validation of the route optimisation software module, as well as the different fleet compositions considered in the experimentation.

#### 4.1.1 Scenario 1: Baseline Scenario

Scenario 1 reflects the current situation of postal operations in the urban area of Zaragoza, where the following infrastructure is in place: A) Nine Delivery Units (DUs) distributed throughout the city and dedicated to the delivery of postal items and small parcels. Most of the routes are done by postmen/postwomen on foot; B) Two Special Service Units (SSU) that are located in the northern and southern areas of the city, respectively. They specialise in the delivery of larger parcels and therefore all routes are done with a motorised vehicle.

Figure 2 A) shows the distribution of the DUs and SSUs in the city of Zaragoza. As for the operation of the routes, the postal operator works in two shifts, one in the morning and one in the afternoon. The morning shift runs from 7:00 am to 3:00 pm, while the afternoon shift runs from 3:00 pm to 10:00 pm. Since postmen need time at the beginning of the shift to sort and prepare the items to be delivered and at the end of the shift to dispose of undelivered items, the time slots in which postmen run their delivery routes are from 8:00 am to 2:00 pm in the morning shift, and from 4:00 pm to 9:00 pm in the afternoon shift. As for the demand data, for the purpose of analysis, we have chosen the data on deliveries made by the postal operator on 13 and 14 September 2022, which we show in Table 1 as distributed by day and by shift.

#### 4.1.2 Scenario 2: Deployment of a Low-Emission Zone

The second scenario we have defined for this analysis considers the deployment of a Low Emission Zone in the historic centre of the city of Zaragoza, whose delimitation is shown in Figure 2 B). The deployment of this Low Emission Zone implies that polluting vehicles cannot enter the area between 7 am and 11 pm. This would affect postal operations in the area since combustion vehicles would not be able to access the zone for delivery. Only postmen/postwomen on foot or electric vehicles would be able to deliver items to the designated area. The percentage of the orders that would be affected by the Low Emission Zone is shown in Table 1. As can be seen, the percentage of orders falling within the Low Emission Zone ranges between 9% and 15% depending on the shift and the day.



Figure 2 – A) Image of the area of study for the Zaragoza Scenario 1 and location of the DUs and SSUs, and B) Delimitation of the area considered for the deployment of the Low Emission Zone in Zaragoza

Day	Working Shift	Number of Shipments	% of shipments in the Low Emission Zone
Sept 13, 22	Morning	11723	15%
Sept 13, 22	Afternoon	3794	9%
Sept 14, 22	Morning	11429	15%
Sept 14, 22	Afternoon	4296	13%

Table 1 Number of shipments per day and shift, and percentage of shipments in the Low-Emission Zone in each zone

#### 4.1.3 Alternative fleet compositions

As mentioned before, have defined three different fleet compositions in terms of fleet electrification in order to understand what impact it may have in environmental and operational terms. The three fleet compositions considered are as follows:

- **Current fleet composition:** in this alternative, the fleet has the same composition as the current postal operator fleet shown in Table 2.
- **Electrification 50%:** in this case, a fleet electrification of around 50% is considered, following the same composition in terms of vehicle typology.
- **Electrification 100%:** in this last alternative fleet composition, 100% electrification is considered, i.e., all vehicles in the fleet are electric. Similar to the previous case, the typology of the vehicles is maintained with respect to the current composition.

Vehicle Type	Technology	<b>Current composition</b>	<b>Electrification 50%</b>	<b>Electrification 100%</b>
Large Van	Combustion	9	9	-
Large Van	Electric	-	-	9
Motorcycle	Electric	12	27	27
Small Van	Combustion	37	15	-
Small Van	Electric	2	24	39

Table 2 - Number of vehicles per type and technology for each of the fleet composition alternatives

#### 4.2 Result analysis

In this section, we will analyse the results obtained by the software module presented in the two scenarios defined and for each of the fleet compositions. For this analysis of the results, we will consider the indicators shown in Table 3, which are provided by the route optimisation system presented in this article. In the two following subsections, we discuss the results obtained in each of the scenarios. More details about the calculation of the indicators can be found in (Vincenzo et al., 2022).

Impact area	Criteria	Indicator	Data/unit
<b>Environment &amp;</b>	Air quality	CO concentration	g/day
Society		SOx concentration	g/day
		NOx concentration	g/day
		NH3 concentration	g/day
		PM10 concentration	g/day
	GHG emissions	CO2	g/day
		CH4	g/day
		N2O	g/day
	Social costs	Social costs of air quality and GHG emissions	€/day
Transport & mobility	Accessibility	Number of shipments	n./day
		Number of routes	n./day
		Total km covered (including walking)	km/day
		Total km covered by green modes (including walking)	km/day
		Total veh-km covered by freight vehicles	Veh-km/day
		Total veh-km covered by green freight vehicles	Veh-km/day
	UFT vehicles	Vehicle utilisation factor	%/day
	Operative costs	Fixed costs	€/day
		Running costs	€/day
		Capital costs	€/day

Table 3 Indicators considered for the analysis of results

#### 4.2.1 Results for Scenario 1

Table 4 shows the results for scenario 1 with the different fleet compositions. As can be seen, with the current fleet having a high percentage of conventional vehicles, the environmental impact is high, reaching a social cost of more than €17,75 per day. However, by increasing the electrification of the fleet, emissions are reduced by around 85%, as is the social cost. This more than 50% increase is due to the fact that the postal operator fleet is oversized to cope with peak demand. Therefore, with 50% of the current fleet electrified, emissions would be reduced by 85% in periods of intermediate demand. With 100% electrification of the fleet, as expected, the environmental impact is reduced by 100%.

Furthermore, as we can see and as expected, the impact in terms of operations is nil as in all cases the same service levels are maintained. However, in terms of operational costs, the electrification of the fleet implies a slight increase in fixed costs and mainly in capital costs (60%), due to the higher price of electric vehicles. On the positive side, however, running costs would be reduced by 40%.

#### 4.2.2 Results for Scenario 2

The results of scenario 2 for the implementation of a Low Emission Zone are shown in the table below. If we look at the environmental impact of the different levels of electrification, we see that the results are very similar to those of the previous scenario, as is to be expected. Where we do see some differences is in the number of shipments delivered, which increases by around 1% with higher electrification of the fleet, which is the same as the decrease in the number of shipments delivered when compared to scenario 0 for the current postal fleet. This 1% increase is due to the fact that with increased electrification of the fleet, more vehicles can access the Low Emission Zone and therefore deliver more parcels.

Overall, the presented tool has shown that increasing the electrification of the fleet is an effective way to reduce the environmental impact of postal operations, with the added benefit of increased access to Low Emission Zones. The drawbacks found by the tool are that there may be some additional costs associated with electrification.

In short, these results have validated that the presented tool is novel and that it allows the optimisation and simulation of last-mile logistics considering different elements of high relevance nowadays such as fleet electrification or the implementation of low-emission zones. More details about the results of the algorithm in other scenarios and using real data from the city of Dublin can be found in (Vincenzo et al., 2022)

## Acknowledgements

This research has been supported by the Spanish Ministry of Science and Innovation through research project PID2019-109393RA-I00. This research has also been supported by European Union's Horizon 2020 research and innovation programme under grant agreement No. 861540 [project SENATOR (Smart Network Operator Platform enabling Shared, Integrated and more Sustainable Urban Freight Logistics)].

#### References

- DHL. (2023). Logistics Trend Radar, Delivering Insight Today, Creating Value Tomorrow DHL Global. Retrieved April 10, 2023, from https://www.dhl.com/global-en/home/insights-and-innovation/insights/logistics-trend-radar.html
- European Comission. (2020). A Policy Brief from the Policy Learning Platform on Low-carbon economy Sustainable Urban Logistics Policy Learning Platform on Low-carbon economy Policy brief: Sustainable urban logistics.
- European Comission. (2023). E-commerce statistics for individuals Statistics Explained. Retrieved April 10, 2023, from https://ec.europa.eu/eurostat/statistics-explained/index.php?title=E-commerce\_statistics\_for\_individuals
- Goel, R. K., & Bansal, S. R. (2019). Hybrid algorithms for rich vehicle routing problems: A survey. In *Smart Delivery Systems: Solving Complex Vehicle Routing Problems* (pp. 157–184). Elsevier. https://doi.org/10.1016/B978-0-12-815715-2.00011-7
- Gu, W., Cattaruzza, D., Ogier, M., & Semet, F. (2019). Adaptive large neighborhood search for the commodity constrained split delivery VRP. *Computers & Operations Research*, *112*, 104761. https://doi.org/10.1016/j.cor.2019.07.019
- Lahyani, R., Khemakhem, M., & Semet, F. (2015). Rich vehicle routing problems: From a taxonomy to a definition. *European Journal of Operational Research*, 241(1), 1–14. https://doi.org/10.1016/j.ejor.2014.07.048
- Pisinger, D., & Ropke, S. (2019). Large neighborhood search. *International Series in Operations Research and Management Science*, 272, 99–127. https://doi.org/10.1007/978-3-319-91086-4\_4/COVER
- Potvin, M. and, & Gendreau Jean-Yves. (2019). *Handbook of Metaheuristics*. (M. Gendreau & J.-Y. Potvin, Eds.) (Vol. 272). Cham: Springer International Publishing. https://doi.org/10.1007/978-3-319-91086-4
- Vincenzo, C., Muccio, E., Masegosa, A. D., & Fajardo, J. (2022). D2.5 Feedback from ZGZ and DUBLIN about first usage of the SENATOR dynamic planning.
- https://www.senatorproject.eu/wp-content/uploads/2023/01/D2.5-Feedback-from-ZGZ-and-DUBLIN-about-first-usage-of-the-SENATOR-dynamic-planning\_rev.pdf