ROAR-NET Cost Action CA22137 First Training School



Integrating Passenger Transport and Parcel Deliveries:
An example of Share-a-Ride problem in Luxembourg

D. H. Stolfi, W. Mtalaa



AGENDA



- Research at LIST
- Last Mile Delivery overview
- Combining Passenger and Parcel Transport
- Description of the Problem
- Mathematical Model
- Algorithms Developed
- Numerical Experiments
- Conclusion and Perspectives

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HUMAN RESOURCES

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a

756 employees

34% women

61 nationalities

114
people recruited

researchers or Innovation experts

78%

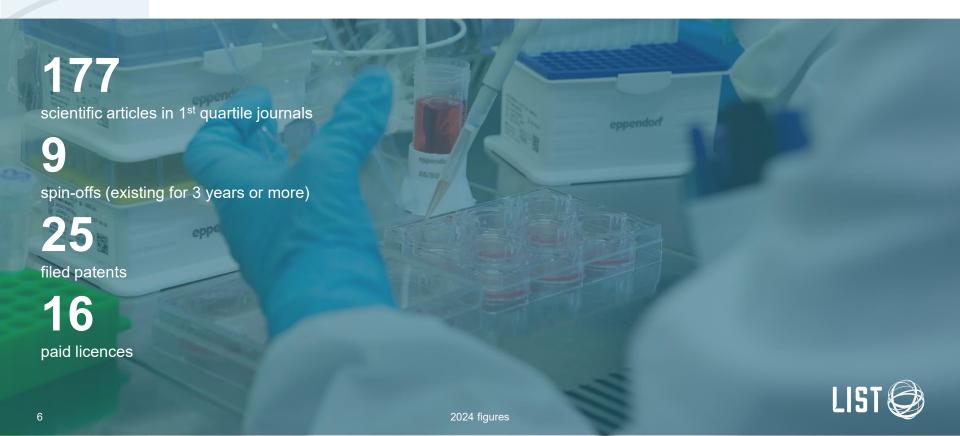
83

PhD students hosted



SCIENCE AND TRANSFER IN 2024





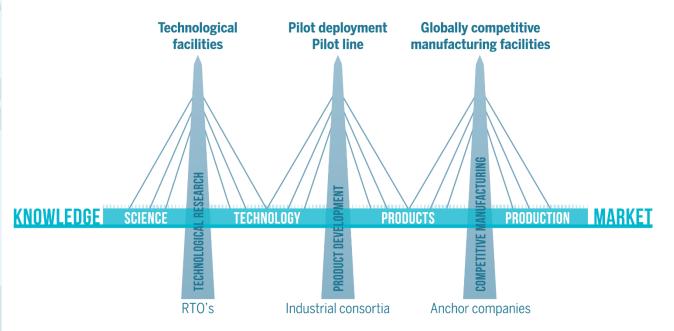


4 AREAS OF RESEARCH EXCELLENCE

Working across the entire innovation chain

- Fundamental & applied research
- Incubation & transfer of technologies
- Policy support
- Doctoral & post-doctoral training

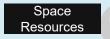
BUILDING A BRIDGE BETWEEN LUXEMBOURG RESEARCH AND SUIL SINE

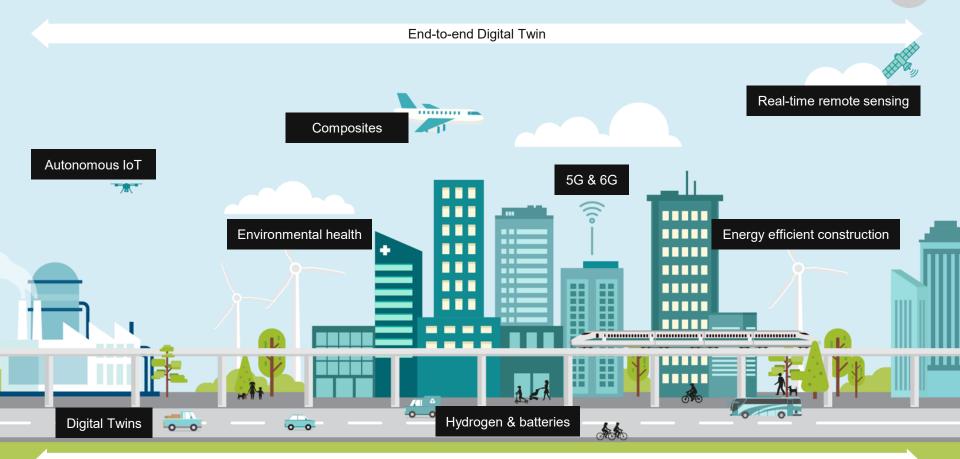


Technology Readiness Level (TRLs)



DIGITALISATION AND SUSTAINABILITY: 12 LIST CORE TECHNOLOGIES





LIST INFORMATICS



Responsible
Data Science
& Analytics



Reliable Distributed Systems



Software Engineering RDI



AIDA Lab









Al & Data Analytics

- Data science
- Al prototyping
- Software development
- Research data management
- Infrastructure provision





5 PARTNERSHIP MODELS

A solution for every problem





LIST VENTURES

LIST's Tech Venture Builder and Start-up / Industry Cooperation

LIST Entrepreneurs

Are you a researcher at LIST planning to set up a company?

LIST Researchers

Are you a researcher at LIST with an interest in entrepreneurship?

Startups & Entrepreneurs

Are you a Startup looking for a tecnological partner?

Entrepreneurial Corporations

Are you a company interested in launching new businesses?

Investors

Are you an investor looking for technological deep moat investments?

THANKS TO





LE GOUVERNEMENT DU GRAND-DUCHÉ DE LUXEMBOURG Ministère de l'Agriculture, de la Viticulture et du Développement rural







LE GOUVERNEMENT DU GRAND-DUCHÉ DE LUXEMBOURG Ministère des Affaires étrangères et européennes

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DU GRAND-DUCHÉ DE LUXEMBOURG Ministère de l'Environnement, du Climat et du Développement durable

Direction de la Défense

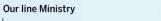
LE GOUVERNEMENT









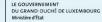




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Service des médias et des communications



LE GOUVERNEMENT DU GRAND-DUCHÉ DE LUXEMBOURG Ministère de la Mobilité et des Travaux publics

Administration des ponts et chaussées



LE GOUVERNEMENT DU GRAND-DUCHÉ DE LUXEMBOURG Ministère de la Santé









































THANKS TO

















































































































































LAST MILE DELIVERY OVERVIEW



Final leg of the logistics chain, from hub to consumer



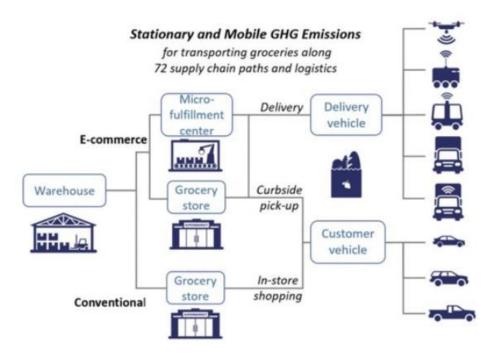
- •Key issues: High costs, traffic congestion, delivery delays
- •Growing demand from e-commerce and food delivery

LAST MILE DELIVERY OVERVIEW



The majority of households are unoccupied during the postal delivery times

- Failed deliveries increase the CO2 emissions
- Light duty vehicles
- Electric tricycles
- Drones
- Public transport



COMBINING PASSENGER AND PARCEL TRANSPORT



Challenges of Integration

- Operational complexity (routing, timing, space constraints)
- Safety and privacy concerns for passengers
- Legal and insurance hurdles
- Balancing delivery efficiency with passenger satisfaction

Opportunity

Shared infrastructure for transporting goods and people

COMBINING PASSENGER AND PARCEL TRANSPORT



- Routing problems
- Lockers
- Delivery by personal cars (light duty vehicles)
- Use of drones/ robots/ public transportation
- Multi-purpose vehicle design (modular seating/storage)
- Al for dynamic routing and load management
- Digital twins and simulation for urban mobility planning
- Policy frameworks supporting multi-use transport



Autonomous shuttles

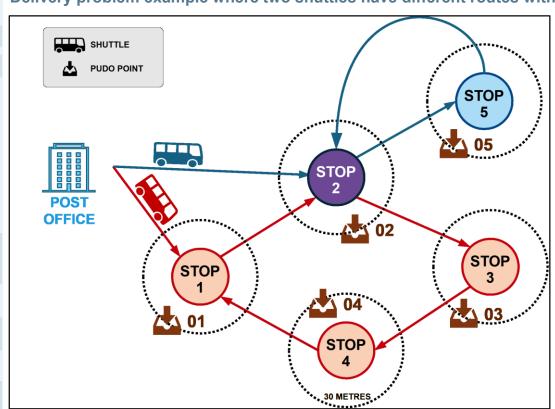


Pack up stations





Delivery problem example where two shuttles have different routes with one common stop.





Entities and Notations

```
 \begin{split} &\{SHUTTLE_1, \ldots, SHUTTLE_N\}, \\ &STOPS_i = \{st_{i,1}, st_{i,2}, \ldots, st_{i,s}\} \\ &B_i = \{b_{i,1}, b_{i,2}, \ldots, b_{i,M}\} \\ &Maximum \ capacity \ of \ a \ Basket \ (H_j) \\ &SPACE_i = \{sp_{i,1}, sp_{i,2}, \ldots, sp_{i,s}\}, \\ &shuttle's \ capacity \ S_i, sp_{i,s} \leq S_i \end{split}
```

x_{i,j,k} indicates if the parcel p_{i,j,k} is included into the delivery batch where i is the shuttle, j is the basket, and k is the parcel's position in the basket

MATHEMATICAL MODEL



Maximise:

$$f = \sum_{i=1}^{N} \sum_{j=1}^{M} \sum_{k=1}^{K_{i,j}} [p_{i,j,k} \times x_{i,j,k}]$$
 (1)

Subject to:

$$\sum_{k=1}^{K_{i,j}} h(p_{i,j,k}) \le H_j \tag{2}$$

$$Stop(p_{i,j,k}) \in STOPS_i$$
 (3)

 $|STOPS_i|$

$$\sum_{st-k} |b_{i,st}| \le sp_{i,k}, \ k = \{1, 2, \dots, |STOPS_i|\}$$
 (4)

Where:

$$p_{i,j,k} = \begin{cases} 1, & \text{if } p_{i,j,k} \in P_T \\ |P_T| + 1, & \text{if } p_{i,j,k} \in P_{SD} \end{cases}$$
 (5)

$$x_{i,j,k} \in \{0,1\}$$
 (6)



Maximize the number of parcels delivered

Problem

Number of stops

Number of parcels to be delivered the :

- Same day
- Tomorrow

Request

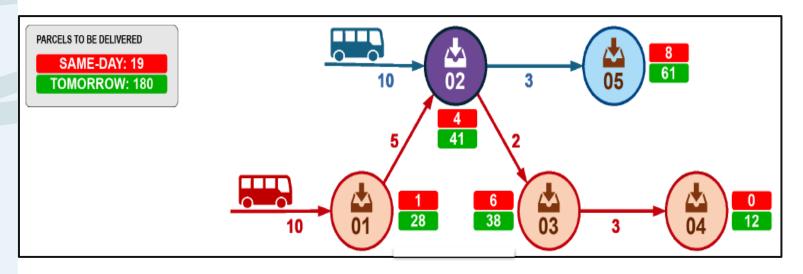
Delivery location = PUDO points Load

Constraints

Capacity
Priority to same day parcels

PROBLEM INSTANCE

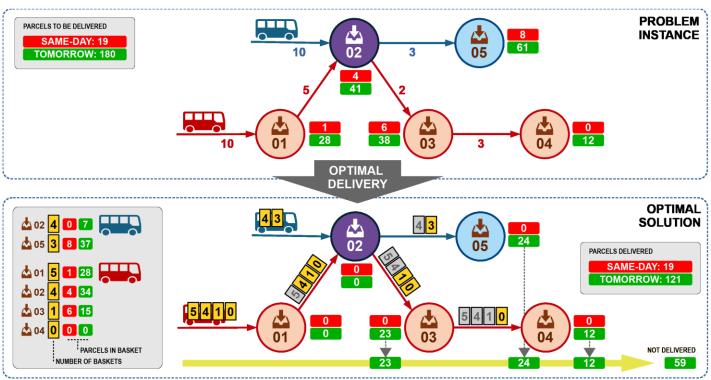




The objective is to maximise the number of parcels delivered the "same-day"

PROBLEM AND SOLUTION REPRESENTATION



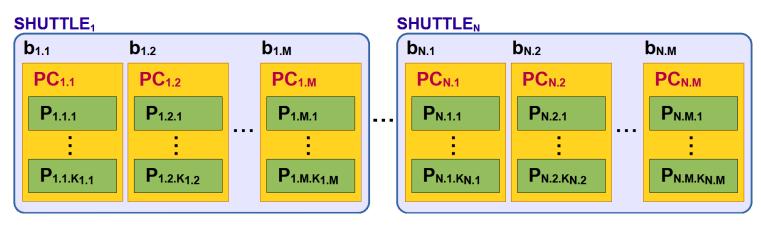


A possible optimal solution with detailed delivery plan and results.

SHARE-A-RIDE PROBLEM



Representation of a solution



- N shuttles having up to M baskets.
- Each basket has K parcels to be delivered at the stop having the postal code PC.

SHARE-A-RIDE PROBLEM



Representation of a solution

```
\begin{split} \vec{x} &= \{shuttle_1, \dots, shuttle_i, \dots, shuttle_N\} \\ shuttle_i &= \{basket_{i,1}, \dots, basket_{i,j}, \dots, basket_{i,M}\} \\ basket_{i,j} &= \{p_{i,j,1}, \dots, p_{i,j,k}, \dots, p_{i,j,K_{i,j}}\} \\ basket_{i_j} &\in STOPS_i \\ p_{i,j,k} &\in PARCELS \end{split}
```

MATHEMATICAL MODEL



Minimisation problem

$$F(\vec{x}) = \Phi R_{SD} + R_T + \xi \tag{12}$$

$$\Phi = |P_{SD}| + 1 \tag{13}$$

$$R_{SD} = |P_{SD}| - |P_{SD}^{\star}| \tag{14}$$

$$R_T = |P_T| - |P_T^*| \tag{15}$$

$$\xi = \begin{cases} \Phi \times |P_{SD}|, & \text{if } \vec{x} \text{ is not viable} \\ 0, & \text{otherwise} \end{cases}$$
 (16)

 $|P_{SD}^{\star}|$ is the number of parcels actually delivered by the proposed solution

ALGORITHMS DEVELOPED



- Greedy algorithm (GR)
- Genetic Algorithm (GA)
- Random Search as a way to assess the validity of the GA and GR

ALGORITHMS DEVELOPED

Pseudocode of the Greedy Algorithm (GR)

```
GR()
P_{SD} \leftarrow Sort(parcels, same-day)
P_T \leftarrow Sort(parcels, tomorrow)
sol \leftarrow EmptySolution()
{First: add "same-day" parcels}
for all shuttle in sol do
  for all stop in Stops(shuttle) do
     for all p in P_{SD} do
       if Stop(p) = stop then
          if Fits(baskets(shuttle), p) then
             basket \leftarrow basket \cup \{p\}  {add to basket}
          else if Viable(sol, new\_basket) then
             sol[shuttle][new\_basket] \leftarrow \{p\} \{new\}
 Second: complete with "tomorrow" parcels}
for all shuttle in sol do
   for all basket in shuttle do
     for all p in P_T do
       if Stop(p) = Stop(shuttle) then
          if Fits(basket, p) then
            basket \leftarrow basket \cup \{p\}
 Finally, add new baskets for "tomorrow" if possible}
for all shuttle in sol do
  for all stop in Stops(shuttle) do
     while Viable(sol, new basket) do
       for all p in P_T do
          if Stop(p) = stop then
             if Fits(new\_basket, p) then
               new\ basket \leftarrow new\ basket \cup \{p\}
return sol
```



ALGORITHMS DEVELOPED Genetic algorithm



Pseudocode of the Genetic Algorithm (GA)

```
GA(n_i, p_c, p_m)
t \leftarrow 0; n_{ev} \leftarrow 0
Q(0) \leftarrow \emptyset {Q=auxiliary population}
P(0) \leftarrow Initialisation(n_i) \{ \text{Pop=population} \}
while n_{ev} < MAX\_EVALUATIONS do
  Q(t) \leftarrow Selection(P(t))
  Q(t) \leftarrow Crossover(Q(t), p_c)
  Q(t) \leftarrow Mutation(Q(t), p_m)
  Evaluation(Q(t))
   P(t+1) \leftarrow Replacement(Q(t), P(t))
  t \leftarrow t + 1
return Best(P)
```

ALGORITHMS DEVELOPED Genetic algorithm_Selection operator

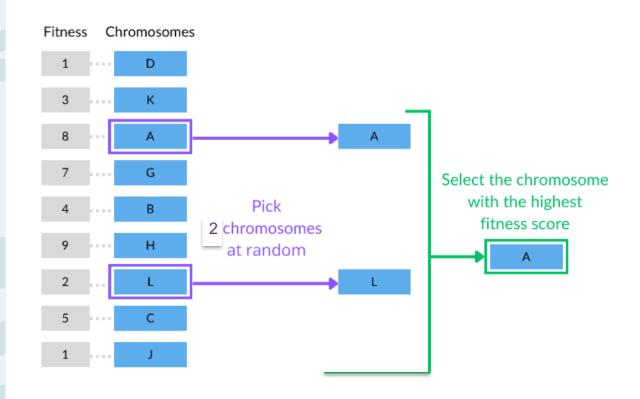


- In a **binary tournament**, two individuals are randomly selected, and the one with the best fitness is chosen.
- **Simplicity:** It's a simple and computationally inexpensive selection method.
- **Commonly Used:** Binary tournament selection is a frequently used technique in genetic algorithms due to its balance of efficiency and effectiveness.
- Larger tournaments tend to favor fitter individuals, leading to faster convergence but potentially reduced diversity.
- This process is repeated until a new population of the desired size is created

ALGORITHMS DEVELOPED



Genetic algorithm_Binary Tournament



ALGORITHMS DEVELOPED Genetic algorithm_Crossover operator



Pseudocode of the Crossover

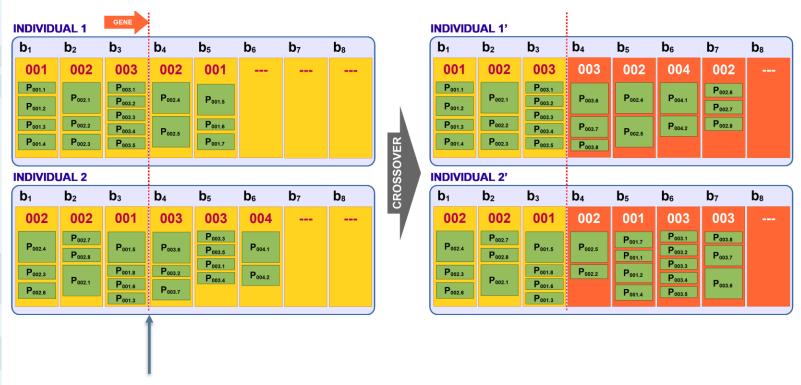
```
Crossover(ind_1, ind_2, cross\_point)
qene \leftarrow 0
for all sh_1, sh_2 in ind_1, ind_2 do
  for all basket_1, basket_2 in sh_1, sh_2 do
     if qene < cross point then
        of f_1[sh_1][basket_1] \leftarrow ind_1[sh_1][basket_1]
        of f_2[sh_2][basket_2] \leftarrow ind_2[sh_2][basket_2]
     else
        if basket_2 \neq \emptyset and Viable(off_1, basket_2) then
           of f_1[sh_1][basket_1] \leftarrow ind_2[sh_2][basket_2]
        else
           of f_1[sh_1][basket_1] \leftarrow \emptyset {if invalid}
        if basket_1 \neq \emptyset and Viable(off_2, basket_1) then
           of f_2[sh_2][basket_2] \leftarrow ind_1[sh_1][basket_1]
        else
           of f_2[sh_2][basket_2] \leftarrow \emptyset {if invalid}
     gene \leftarrow gene + 1
of f_1 \leftarrow RemoveDuplicatesAndComplete(of f_1)
```

 $off_1 \leftarrow RemoveDuplicatesAndComplete(off_1)$ $off_2 \leftarrow RemoveDuplicatesAndComplete(off_2)$ **return** off_1, off_2

ALGORITHMS DEVELOPED



Genetic algorithm_Crossover operator



Cut point located at the third gene

ALGORITHMS DEVELOPED Genetic algorithm_Mutation operator



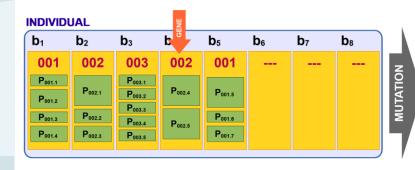
Pseudocode of the Mutation

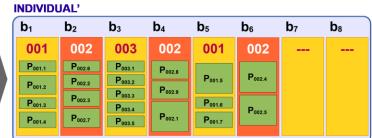
```
Mutation(ind, gene)
pc \leftarrow PostalCode(ind, gene)
for all shuttle in ind do
  for all basket in shuttle do
     if PostalCode(basket) = pc then
       basket \leftarrow \emptyset {All baskets for pc are emptied}
parcels \leftarrow GetParcels(pc) {All parcels for pc}
found \leftarrow \mathbf{false}
for all p in parcels do {Assign parcels (sorted)}
  shuttle \leftarrow RandShuttle(pc)
  for all basket in RandBasket(shuttle, pc) do
     if Fits(basket, p) then {there is space}
        basket \leftarrow basket \cup \{p\}
        found \leftarrow true
        break
  if not found then {parcel was not added}
     for all basket in RandBasket(shuttle, pc) do
        if basket = \emptyset then
          if Viable(ind, basket) then
             ind[shuttle][basket] \leftarrow \{p\}  {new basket}
             break
return ind
```

ALGORITHMS DEVELOPED



Genetic algorithm_Mutation operator





ALGORITHMS DEVELOPED



Pseudocode of the Random search

Pseudocode of Random Search(RS)

```
\begin{aligned} & \mathbf{RS}() \\ & best\_sol \leftarrow \emptyset; \quad min\_f \leftarrow 0; \quad n_{ev} \leftarrow 0 \\ & \mathbf{repeat} \\ & sol \leftarrow RandomSolution() \\ & f \leftarrow Evaluate(sol) \\ & \mathbf{if} \ f \leq min\_f \ \mathbf{then} \\ & best\_sol \leftarrow sol; \quad min\_f \leftarrow f \\ & n_{ev} \leftarrow n_{ev} + 1 \\ & \mathbf{until} \ n_{ev} = MAX\_EVALUATIONS \\ & \mathbf{return} \ best\_sol \end{aligned}
```

Problem instances



Characteristics of each case study

# shuttles	# instances	# shared	# parcels			
# shuttles	# Histalices	stops	Same-Day	Tomorrow		
1	30	0	18 - 40	146 – 193		
2	60	0,1	17 - 39	146 – 194		
3	90	0,1,2	18 - 47	149 – 189		
4	90	0,1,2	17 - 43	155 – 198		
5	90	0,1,2	16 – 45	152 – 190		

- 360 problem instances in total,30 for one-shuttle scenarios, 60 for two shuttles, and 90 for the rest.
- Each group of 30 instances corresponds to the number of shared stops, i.e. 0, 1, and 2,
- The range of the number parcels was calculated by using a uniform probability distribution
- The number of empty spaces (0 14) and parcels' height (1 50 cm) are common to all case studies.
- The rest of parameters are:
 - basket capacity (Hi = 100 cm, ∀i),
 - shuttle capacity (Si = 14, ∀i),
 - Number of stops $(5 \le |STOPSi| \le 20)$.
 - The maximum number of baskets per shuttle (M) corresponds to the shuttle capacity, i.e. M = 14.

Problem instances



Hyperparameters of the GA calculated by IRACE

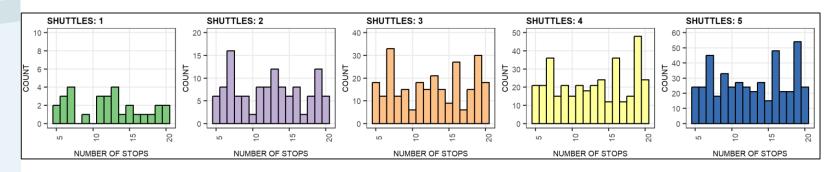
# shuttles	1	2	3	4	5
Pc	0.6000	0.9444	0.7442	0.6233	0.8388
P _m	0.1867	0.1010	0.1492	0.1615	0.1910

- Irace is an implementation of the iterated racing procedure which uses Friedman's nonparametric two-way analysis of variance by ranks
- maximum number of evaluations : 10,000
- Number of individuals: 400 according to the observed algorithm's convergence.
- crossover probability Pc and mutation probability Pm, calculated by performing 2000 experiments per case study with a confidence interval for the elimination test of 0.95.

Problem instances



Histograms with number of stops, randomly calculated, used in the scenarios of each case study





Results for the 360 scenarios, best fitness and run time values for 30 runs of RS and GA (GR is one run)

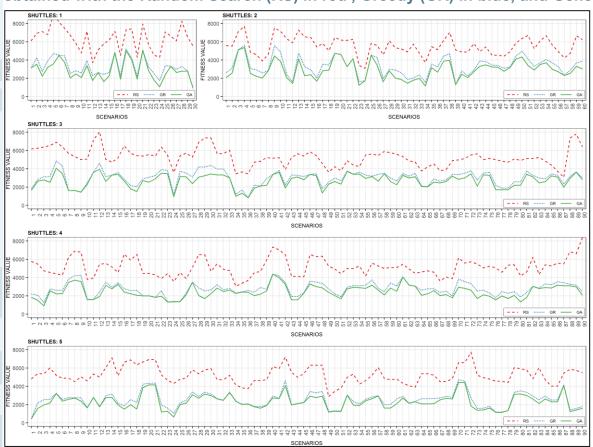
		RS		GR		GA		Improvements					
# shuttles	# scenarios	Avg.	Avg.	Avg.	Avg.	Avg.	Avg.		GA vs. RS		GA vs. GR		
		fitness	time (s)	fitness	time (s)	fitness	time (s)	Min.	Avg.	Max.	Min.	Avg.	Max.
1	30	6338.1	20.0	3262.6	0.001	2853.5	39.9	45.4%	146.4%	503.4%	0.2%	17.7%	62.4%
2	60	5938.5	22.8	3339.3	0.002	2927.6	68.4	35.9%	128.4%	340.0%	0.2%	17.7%	44.0%
3	90	5556.5	25.1	3017.6	0.002	2693.2	69.9	48.2%	146.9%	293.2%	0.2%	15.0%	36.4%
4	90	5305.3	26.7	2758.0	0.002	2429.6	70.4	45.5%	157.5%	426.3%	1.0%	15.1%	48.5%
5	90	5310.5	28.2	2541.7	0.002	1633.3	69.7	66.6%	197.4%	989.1%	0.8%	16.3%	61.7%

Total number of parcels to be delivered by case study and solutions obtained by each optimization algorithm

	# shuttles	# scenarios	Parcels			GR		GA			
					Delivered		Not delivered	Delivered		Not delivered	
			Same-Day	Tomorrow	Same-Day	Tomorrow	by shuttles	Same-Day	Tomorrow	by shuttles	
	1	30	862	5069	316	469	5146 (86.8%)	386	891	4654 (78.5%)	
	2	60	1749	10290	635	1230	10174 (84.5%)	773	2280	8986 (74.6%)	
	3	90	2682	15321	1169	2196	14638 (81.3%)	1330	4016	12657 (70.3%)	
	4	90	2653	15410	1283	2160	14620 (80.9%)	1443	4355	12265 (67.9%)	
	5	90	2746	15318	1478	2389	14197 (78.6%)	1589	4878	11597 (64.2%)	

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Comparison of the best fitness values for the 360 scenarios,
obtained with the Random Search (RS) in red , Greedy (GR) in blue, and Genetic Algorithm (GA) in green.



CONCLUSION



- Design of two algorithms for last mile parcel delivery using mobility services.
- Greedy algorithm rapidly finds good solutions (hundredths of second)
- Genetic algorithm improves the quality of the solutions but slower
- Adoption of the solution would:
 - reduce the delivery costs
 - improve the road traffic conditions (jams and greenhouse gas emissions)





- Real tests in Belval and other places
- Real passenger and logistics data
- Explore other operators/ local search algorithms that would be more efficient.
- Hybridise a local search with an evolutionary schema
- Increase convergence of mobility and logistics platforms
- Increase integrated collaboration between public and private agencies

REFERENCES



- A. Adibfar, S. Gulhare, S. Srinivasan, and A. Costin, "Analysis and modeling of changes in online shopping behavior due to covid-19 pandemic: A florida case study," Transport Policy, vol. 126, pp. 162–176, 2022.
- V. Gatta, E. Marcucci, M. Nigro, and S. Serafini, "Sustainable urban freight transport adopting public transport-based crowdshipping for b2c deliveries," European Transport Research Review, vol. 11, no. 1, p. 13, 2019
- I. Orenstein, T. Raviv, and E. Sadan, "Flexible parcel delivery to automated parcel lockers: models, solution methods and analysis," EURO Journal on Transportation and Logistics, vol. 8, no. 5, pp. 683–711, 2019.
- R. Masson, A. Trentini, F. Lehuédé, N. Malhéné, O. Péton, and H. Tlahig, "Optimization of a city logistics transportation system with mixed passengers and goods," EURO Journal on Transportation and Logistics, vol. 6, no. 1, pp. 81–109, 2017.
- Y. Vakulenko, D. Hellström, and K. Hjort, "What's in the parcel locker? exploring customer value in e-commerce last mile delivery," Journal of Business Research, vol. 88, pp. 421–427, 2018.
- E. Chaalal, C. Guerlain, E. Pardo, and S. Faye, "Integrating connected and automated shuttles with other mobility systems: Challenges and future directions," IEEE Access, vol. 11, pp. 83 081–83 106, 2023. F. Cavallaro and S. Nocera, "Integration of passenger and freight transport: A concept-centric literature review," Research in Transportation Business & Management, vol. 43, p. 100718, 2022.

QUESTIONS?





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