THE SELECTIVE TRAVELING SALESMAN PROBLEM WITH RELEASE DATES AND DRONE RESUPPLY

New and innovative use of drones to assist in parcel delivery

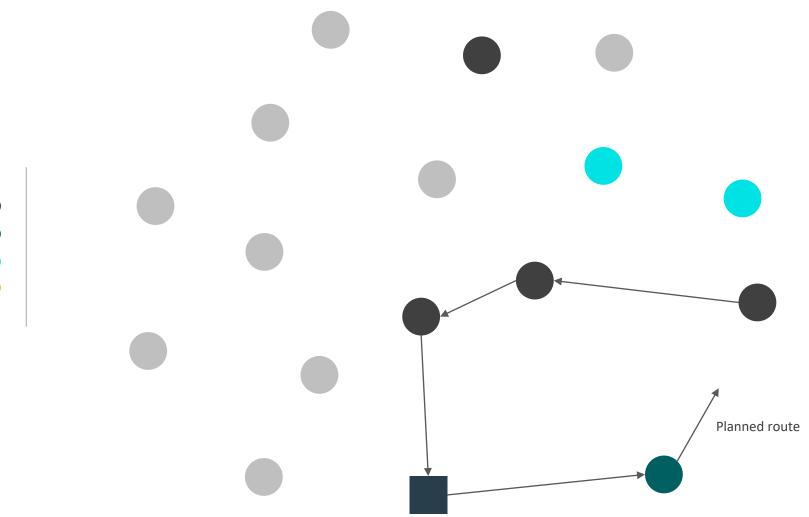
JUAN C. PINA-PARDO¹, DANIEL F. SILVA², ALICE E. SMITH²

¹ School of Industrial Engineering, Pontificia Universidad Catolica de Valparaiso, Chile ² Department of Industrial and Systems Engineering, Auburn University, USA

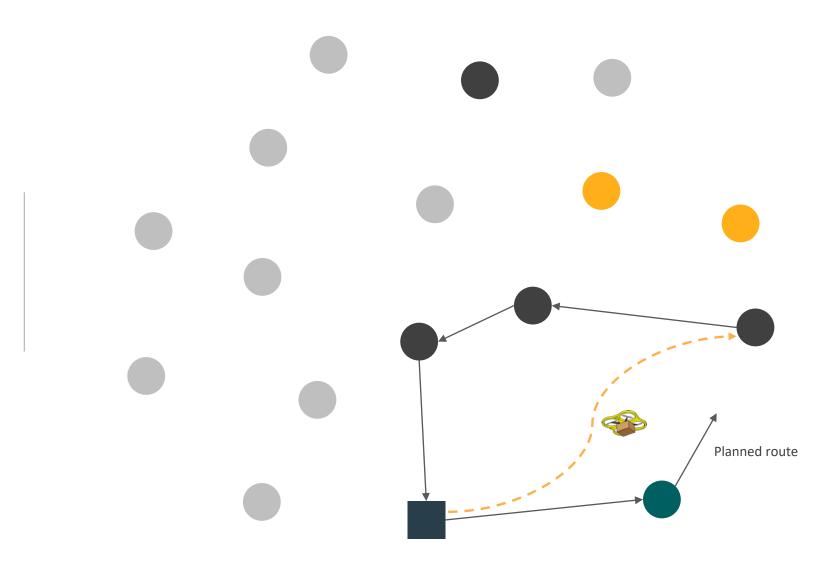


MOTIVATION

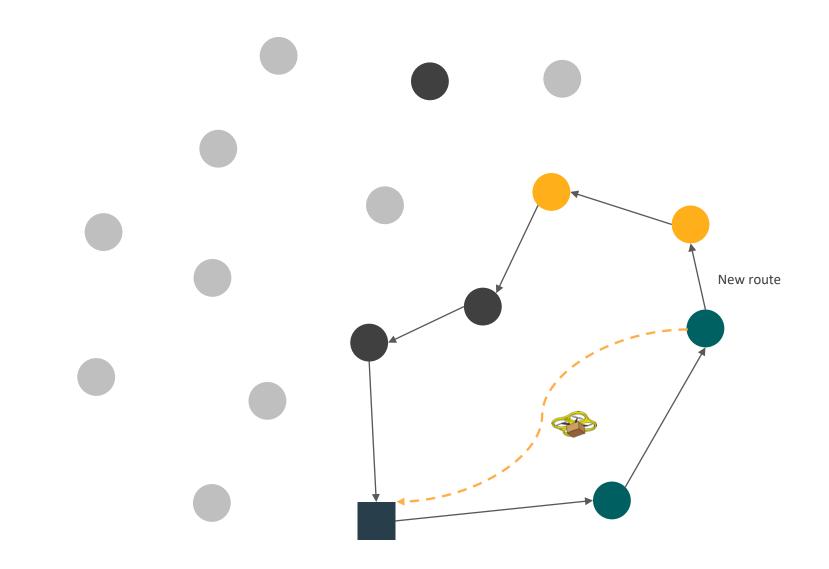
- Delivering a set of orders within a limited time frame
- Orders become available for dispatch at different times
- Vehicles can only deliver parcels loaded onto them before leaving the depot
- Using drones to resupply dispatch vehicles while en route
- No need for trucks to return to pick up new available orders



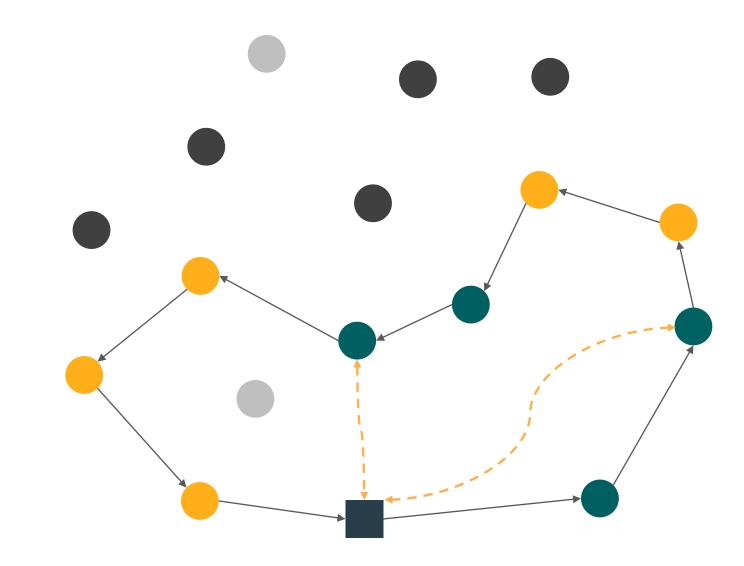
- Customers whose order arrived at the depot
 - Customers served so far
 - New orders available at the depot
 - Orders sent by the drone 🥚



- Customers whose order arrived at the depot
 - Customers served so far
 - New orders available at the depot
 - Orders sent by the drone 🥚



- Customers whose order arrived at the depot
 - Customers served so far
 - New orders available at the depot
 - Orders sent by the drone 🥚



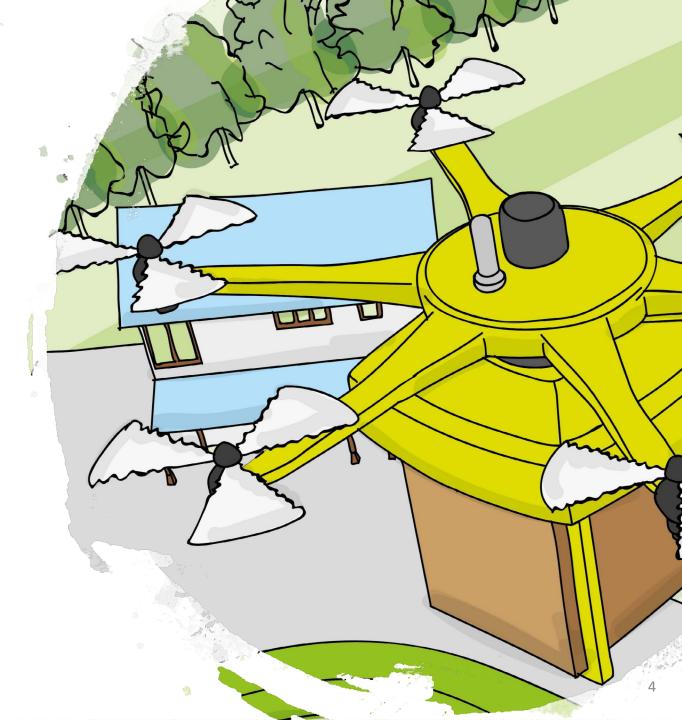
- Customers whose order arrived at the depot
 - Customers served so far
 - New orders available at the depot
 - Orders sent by the drone 🥚

PROBLEM STUDIED

Find a dispatch route for a single truck, synchronized with drone resupply, to minimize distribution costs and penalties incurred for not delivering.

ASSUMPTIONS AND CONSIDERATIONS:

- Single truck and single drone
- Deterministic case: Release dates are known in advance
- The drone has a given load capacity and flight endurance
- The truck is uncapacitated





MIP MODEL

PARAMETERS:

Release dates

Truck and drone travel times

Load capacity and flight endurance of the drone

Cost for wait times of the driver for the drone

Time for unloading orders from the drone

Transportation and penalty costs

Cost per use of the drone

Length of the delivery day

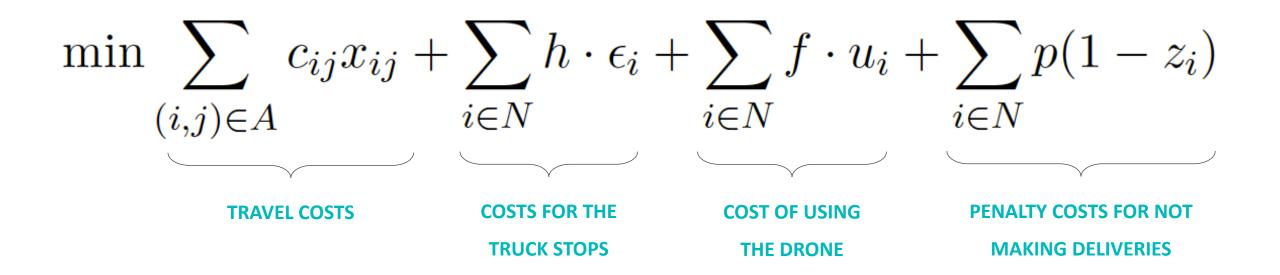


MIP MODEL

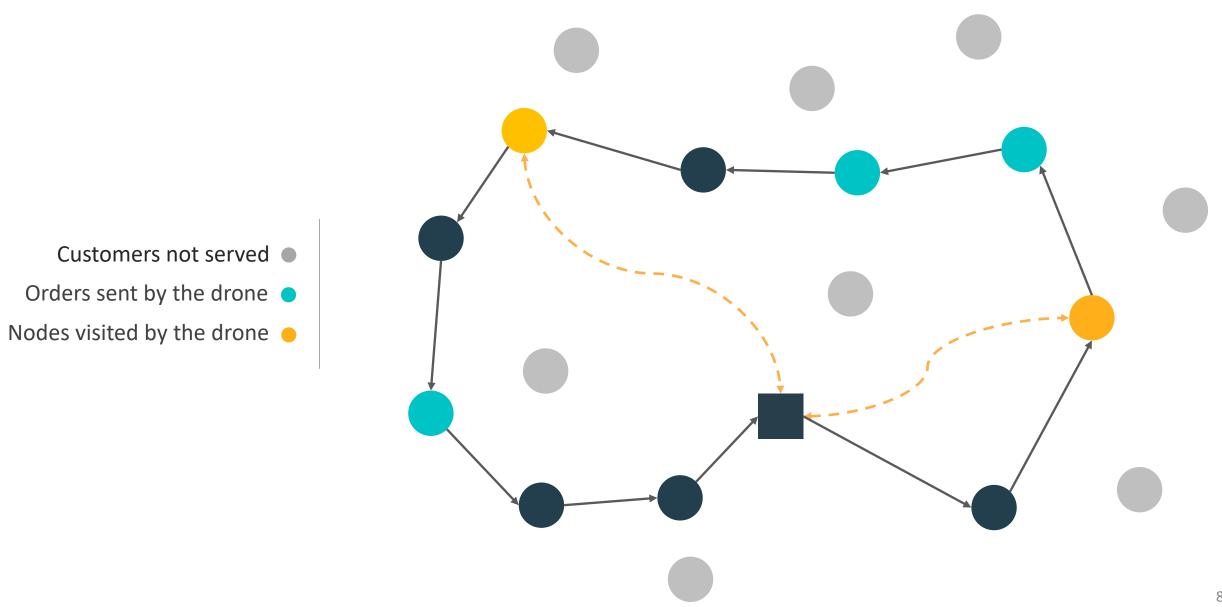
VARIABLES:

- x_{ij} : 1, if node j is visited immediately after node i by the truck. 0, otherwise.
- z_i : 1, if node *i* is visited by the truck. 0, otherwise.
- r_{ij} : 1, if node j is visited after node i by the drone. 0, otherwise.
- u_i : 1, if the drone flies to node *i* for resupplying the truck with new orders. 0, otherwise.
- y_{ij} : 1, if the order of customer j is loaded onto the truck at node i. 0, otherwise.
- T_i : Time when the truck departs node i.
- s_i : Time when the drone is launched for node *i*.
- ϵ_i : Elapsing time between arrival and departure of the truck at node *i*.

OBJECTIVE FUNCTION



CONSTRAINTS



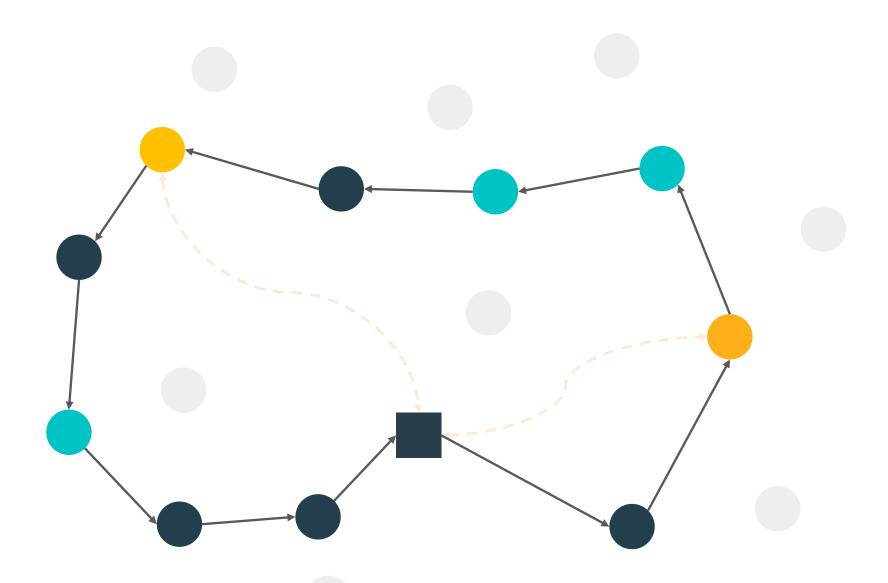
CONSTRAINTS



Which customers?

Which sequence?

Deadline





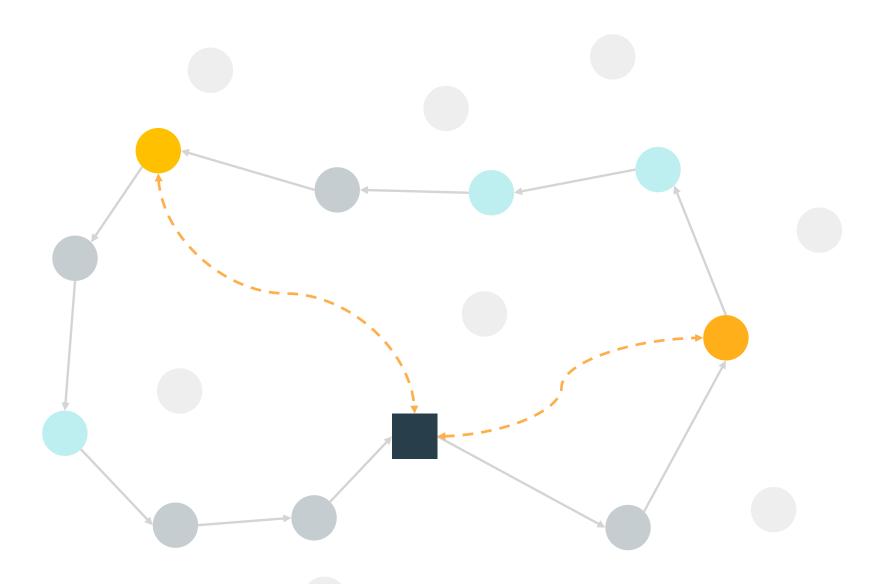


DRONE FLIGHTS:

Which nodes?

Which sequence?

Drone endurance



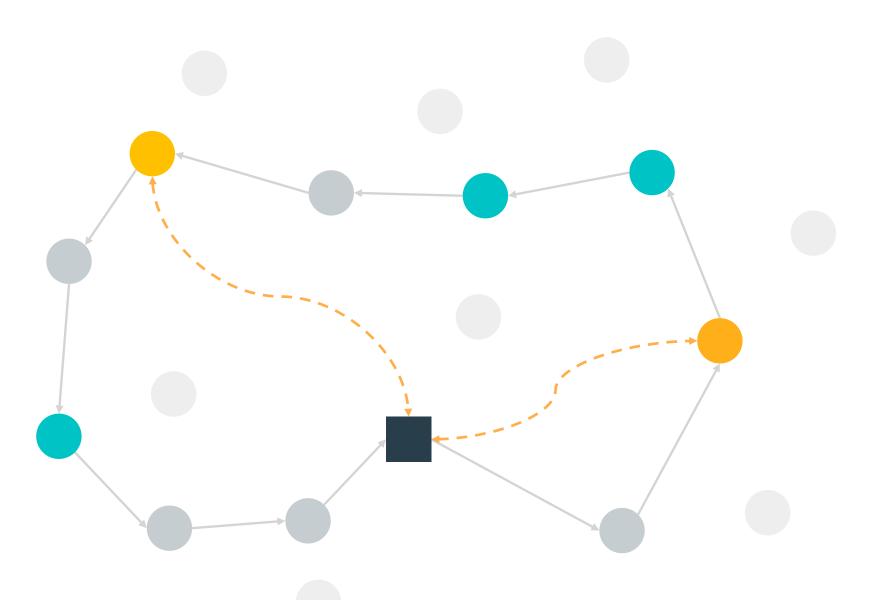




DRONE RESUPPLY:

Which orders?

Load capacity of drone



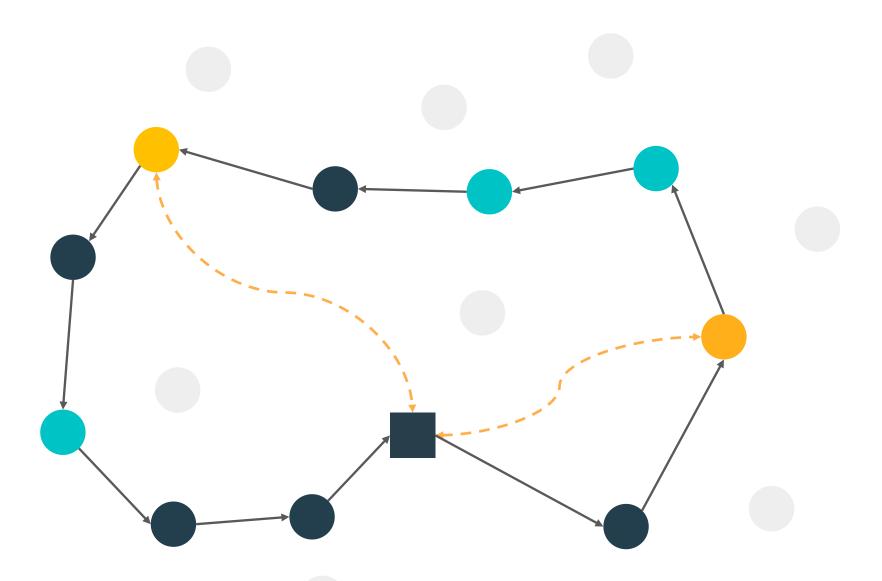
CONSTRAINTS



TIMING:

MTZ timing constraints

Synchronization of the routes

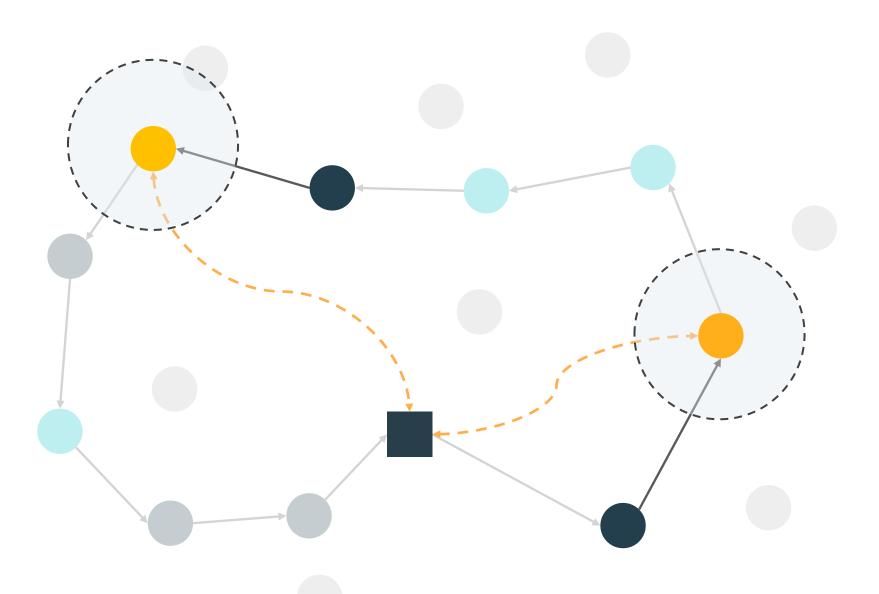


CONSTRAINTS



WAIT TIMES:

Elapsed time between arrival and departure of the truck at each node

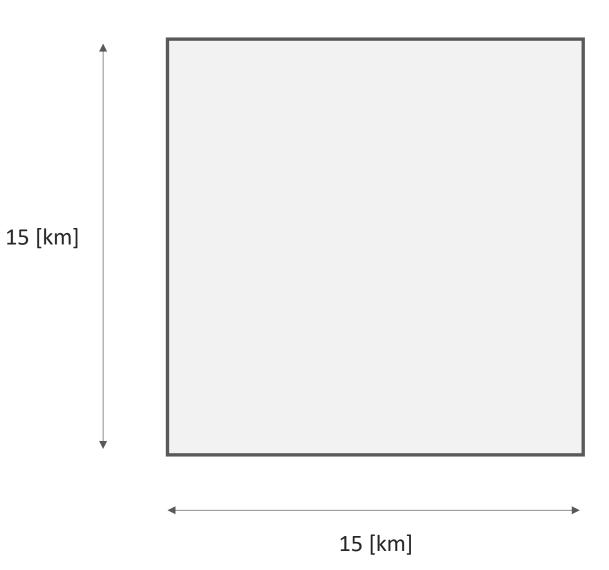




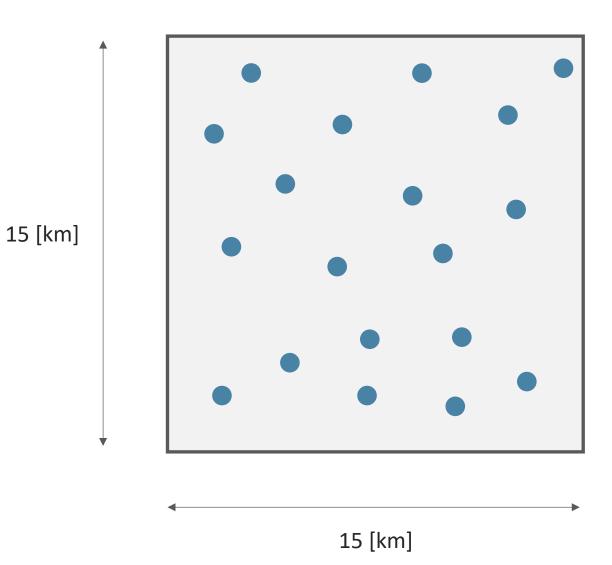
COMPUTATIONAL EXPERIMENTS

- Programed in Java and use of CPLEX 12.8 for solving MIP models
- Intel(R) Xeon(R) Gold 5118 CPU @2.30 GHz (12 cores) with 64 GB RAM

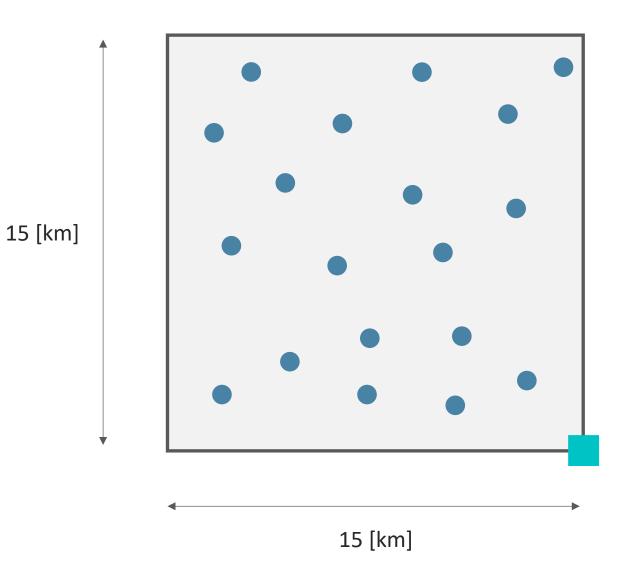




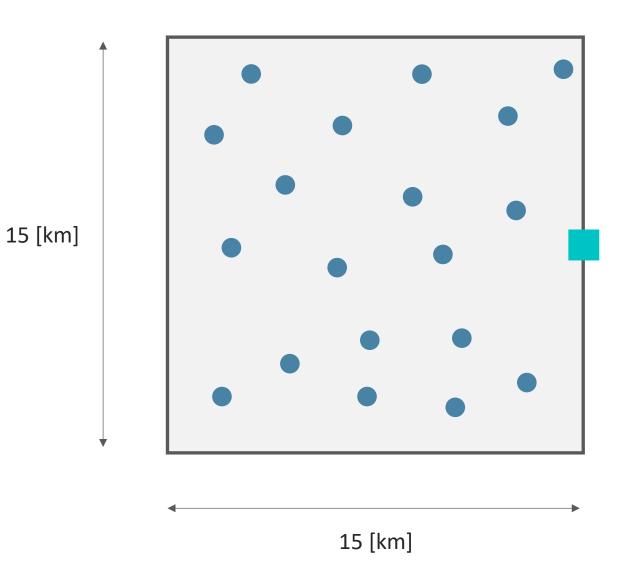














- 90 randomly generated instances of 10, 20, and 30 customers (30 for each size)
- Deliveries can be made between 8 AM and 6 PM
- Orders are only accepted until 4 PM
- All orders are the same size
- **Truck:** Speed of $30 \ km/hr$ and Manhattan distances
- **Drone:** Speed of 60 km/hr and Euclidean distances
- **Drone:** Flight endurance of 30 minutes and load capacity of 4 orders

		1004	100 0101 011	0 000 01 11	bouriees,	eemstaering	a ran on	
	Layout	n	CPU	Cost	GAP	Fill-rate	Drone	Orders
·		10	0.15			100.00%	0.03	0.03
	Right	20	99.82	82.65	0.01%	100.00%	0.70	1.53
	1018110	30	904.44	151.79	2.24%	98.22%	1.23	3.63
		10	0.17	64.13	0.00%	100.00%	0.07	0.10
	Corner	20	115.54	88.04	0.00%	100.00%	0.83	1.87
		30	1226.00	241.26	4.53%	96.33%	1.13	3.23

Table 1: Average results over the set of instances, considering a run-time of 30 minutes.

Table 1: Average results over the se	et of instances,	considering a run	time of 30 minutes.
--------------------------------------	------------------	-------------------	---------------------

Layout	n	CPU	Cost	GAP	Fill-rate	Drone	Orders
Right	$ \begin{array}{r} 10 \\ 20 \\ 30 \end{array} $	0.15 99.82 904.44	82.65	0.01%	100.00% 100.00% 98.22%	0.70	0.03 1.53 3.63
Corner	10 20 30	$0.17 \\ 115.54$	64.13 88.04	0.00% 0.00%	98.2270 100.00% 100.00% 96.33%	0.07 0.83	0.10 1.87 3.23

Table 1: Average results	over the set of instances,	considering a run-tin	ne of 30 minutes.
--------------------------	----------------------------	-----------------------	-------------------

Layout	n	CPU	Cost	GAP	Fill-rate	Drone	Orders
	10	0.15			100.00%		0.03
Right	20	99.82			100.00%		1.53
	30	904.44	151.79	2.24%	98.22%	1.23	3.63
	10	0.17			100.00%		0.10
Corner	20	115.54			100.00%		1.87
	30	1226.00	241.26	4.53%	96.33%	1.13	3.23

Table 1: Average results over the	e set of instances, o	considering a run-time	of 30 minutes.
-----------------------------------	-----------------------	------------------------	----------------

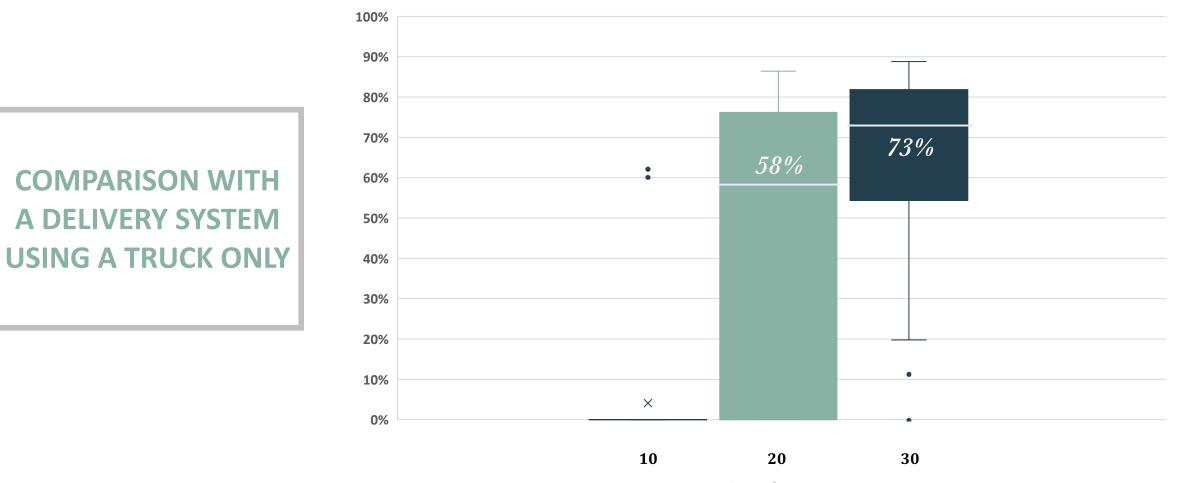
Layout	n	CPU	Cost	GAP	Fill-rate	Drone	Orders
	10	0.15	60.58	0.00%	100.00%	0.03	0.03
Right	20	99.82	82.65	0.01%	100.00%	0.70	1.53
	30	904.44	151.79	2.24%	98.22%	1.23	3.63
	10	0.17	64.13	0.00%	100.00%	0.07	0.10
Corner	20	115.54	88.04	0.00%	100.00%	0.83	1.87
	30	1226.00	241.26	4.53%	96.33%	1.13	3.23

Table 1: Average results over the	e set of instances, o	considering a run-time	of 30 minutes.
-----------------------------------	-----------------------	------------------------	----------------

Layout	n	CPU	Cost	GAP	Fill-rate	Drone	Orders
	10	0.15	60.58	0.00%	100.00%	0.03	0.03
Right	20	99.82	82.65	0.01%	100.00%	0.70	1.53
	30	904.44	151.79	2.24%	98.22%	1.23	3.63
	10	0.17	64.13	0.00%	100.00%	0.07	0.10
Corner	20	115.54	88.04	0.00%	100.00%	0.83	1.87
	30	1226.00	241.26	4.53%	96.33%	1.13	3.23

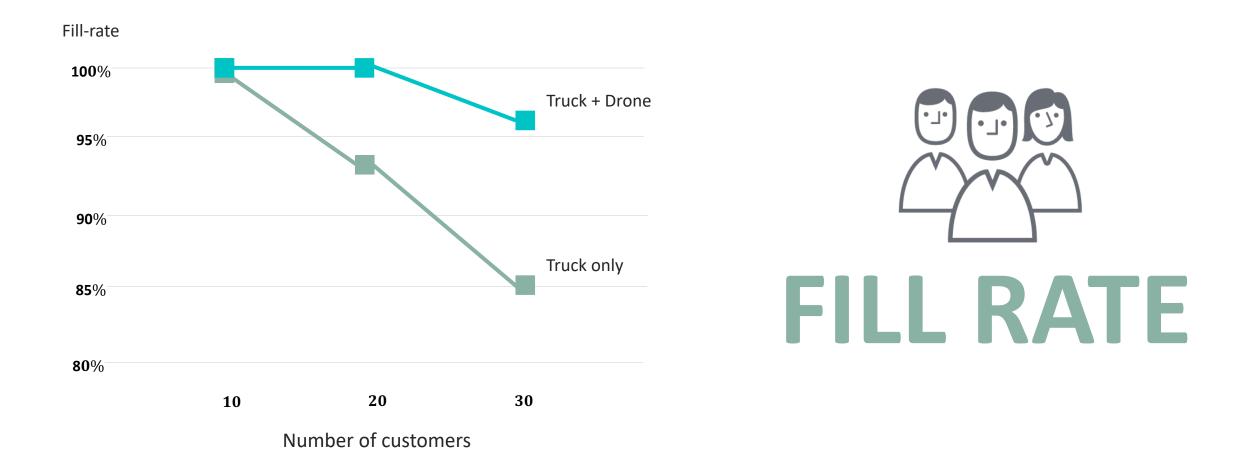
BENEFITS OF USING DRONES TO RESUPPLY VEHICLES

Cost savings



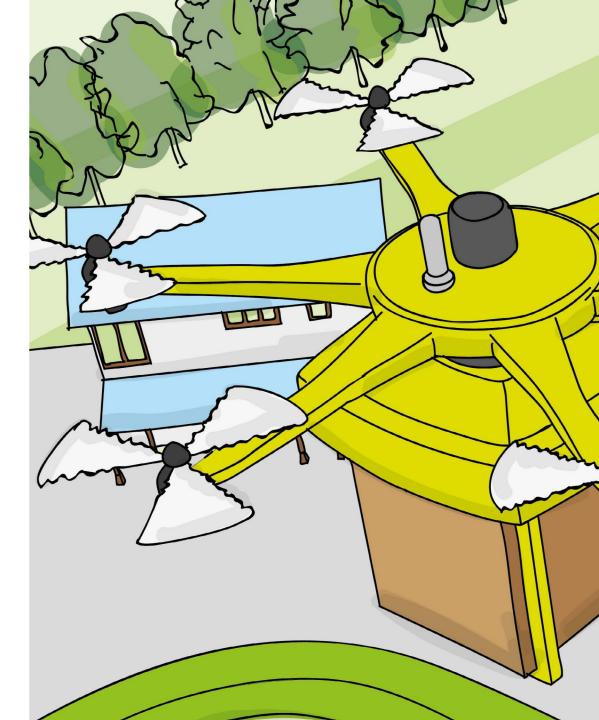
Number of customers

BENEFITS OF USING DRONES TO RESUPPLY VEHICLES



CONCLUSIONS

- Using drones to resupply vehicles can significantly reduce distribution costs and increase the number of orders delivered
- The MIP model is able to provide good gaps after a few minutes
- Extensions of the problem include considering a fleet of vehicles and drones, multiple depots and different load capacities
- Heuristic approaches could be developed to solve large size instances



THE SELECTIVE TRAVELING SALESMAN PROBLEM WITH RELEASE DATES AND DRONE RESUPPLY

This research was partially funded by **Toyota** Material Handling North America through their University Research Program

JUAN C. PINA-PARDO¹, DANIEL F. SILVA², ALICE E. SMITH²

¹ School of Industrial Engineering, Pontificia Universidad Catolica de Valparaiso, Chile ² Department of Industrial and Systems Engineering, Auburn University, USA

All images were retrieved from:

Joers, M., Schroder, J., Neuhauss, F., Klink, C. & Mann, F. (2016). Parcel delivery: The future of last mile. Travel, Transport and logistic: Mckinsey & Company. Published online on URL: https://mck.co/2n4sABU

KEY CONCEPTS ADDRESSED

RELEASE DATES

Moment when orders become ready to dispatch Cattaruzza et al. (2016) Archetti, Feillet, and Speranza (2015)

DEADLINE

Selection of customers to be visited Feillet, Dejax, and Gendreau (2005) Laporte and Martello (1990) Balas (1989)



DELIVERY USING DRONES

Use drones to reduce last-mile delivery costs Murray and Chu (2015) Otto et al. (2018)

DRONE RESUPPLY

Using drones to resupply dispatch vehicles Dayarian et al. (2018)

McCunney and Van Cauwenberghe (2019)

BENEFITS OF USING DRONES TO RESUPPLY VEHICLES

Table 2: Comparison between the new delivery system and a delivery system using a truck only.

	Full MIP	Model			Truck Or	Truck Only Model			
n	CPU	Cost	GAP	Fill-rate	CPU	Cost	GAP	Fill-rate	
10	0.17	64.13	0.00%	100.00%	8.70	72.07	0.00%	99.33%	
20	115.54	88.04	0.00%	100.00%	1269.53	260.62	7.90%	93.00%	
30	1226.00	241.26	4.53%	96.33%	1747.81	657.70	18.96%	85.33%	