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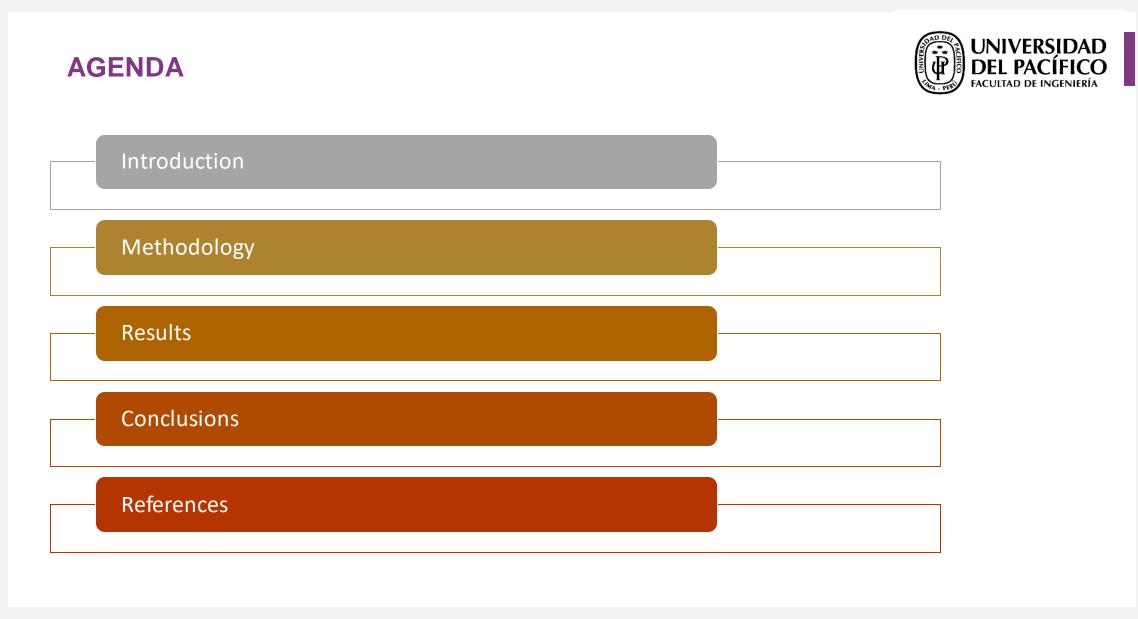
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Distribution of non-alcoholic beverages in the city of Lima





Introduction

The purpose



The purpose of this paper is to analyze the transportation configuration characteristics of non-alcoholic beverages in a Latin-American city

The city



Which are the more congested cities in the world?

The city



Ranking of the most congested cities <u>Globally</u> (Overall daily congestion level – extra travel time – population over 800,000):

1	Mumbai, India	65%	6	Istanbul, Turkey	53%
2	Bogota, Colombia	63%	7	Jakarta, Indonesia	53%
3	Lima, Peru	58%	8	Bangkok, Thailand	53%
4	New Delhi, India	58%	9	Mexico City, Mexico	52%
5	Moscow, Russia	56%	10	Recife, Brazil	49%

Source: TomTom. (2018). Tom Tom Traffic Index: Mumbai takes Crown of 'Most Traffic Congested City' in World

The city





The model



The model proposes routes that improve customer satisfaction due to on-time deliveries through the administration of inventories and the paths of the trucks filling pending orders



- Limited number of trucks that also have a limited capacity
- Loading requirements that are established in the contracts
- Inventory limitations
- Product stacking limitations



Methodology

The model



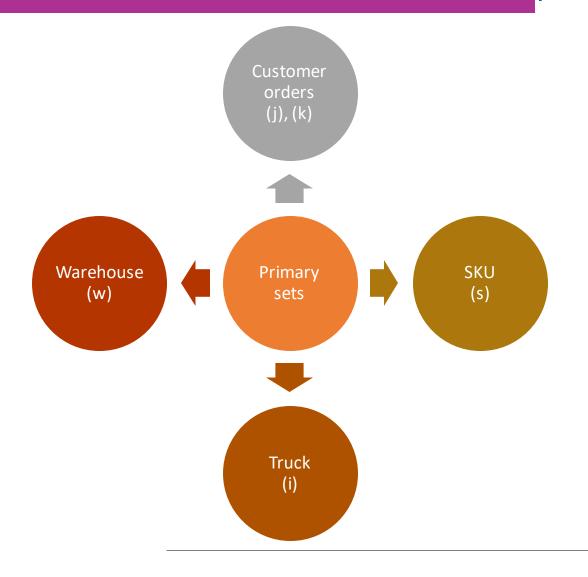
A goal programming model is proposed to schedule the transportation of non-alcoholic beverages from the distribution centers to the points of sale through fleets of trucks with different characteristics The goals

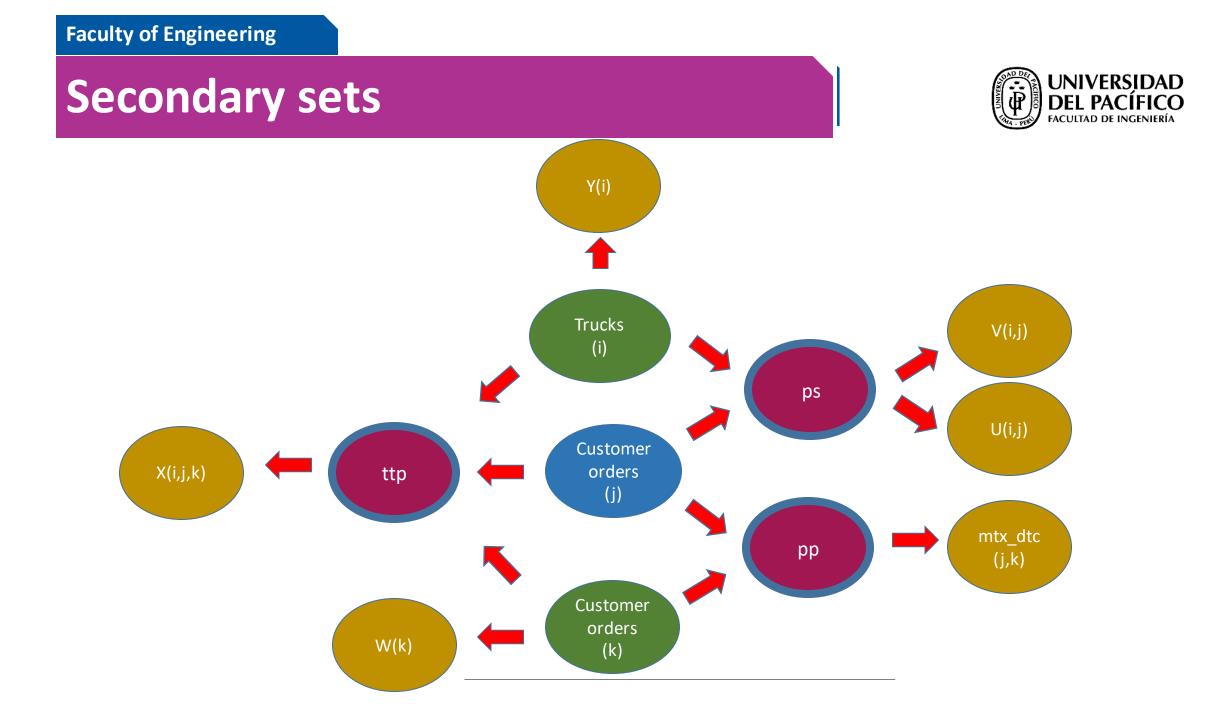


The goals include the fulfillment of the number of pending orders, the number of available trucks, the contractually bound capacity and the distance that trucks will travel to fill the orders

Primary sets

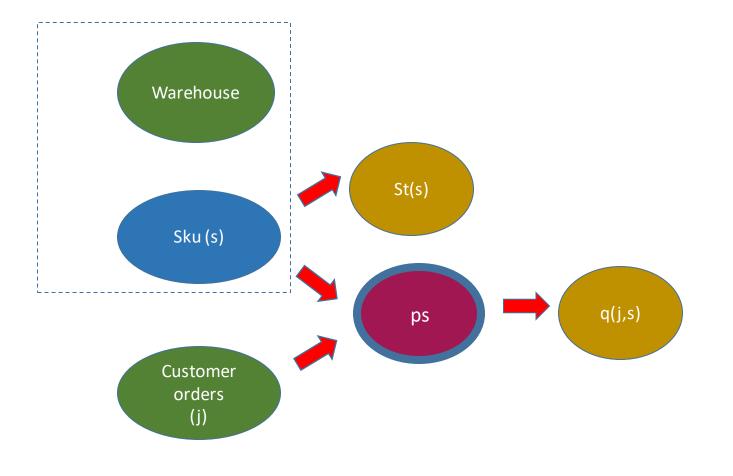






Secondary sets





The objective





$$Min \ z = not_attended + \frac{etrucks}{MinTrucks} + \sum_{i=1}^{m} e2cap_i + \frac{edst}{Max_Dst}$$

.
$$\sum_{j=1}^{n} \frac{W_j}{n} + not_attended = 1$$

Fullfilment

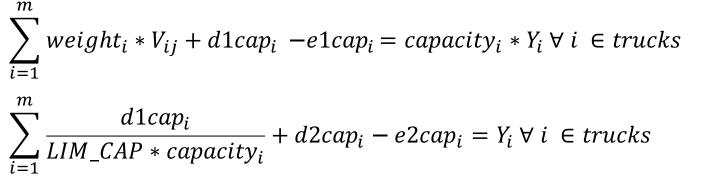
$$\sum_{i=1}^{m} Y_i + dtrucks - etrucks = MinTrucks$$



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The objective

Goals



 $\sum_{i=1}^{m} \sum_{j=1}^{n} \sum_{k=1/j \neq k}^{n} mtxdst_k * X_{ijk} + ddst - edst = Max_Dst$







$$\sum_{\substack{k=1\\n}}^{n} X_{i0k} = Y_i \forall i \in trucks$$
$$\sum_{\substack{k=1\\k=1}}^{n} X_{ik(n+1)} = Y_i \forall i \in trucks$$
$$\sum_{\substack{k=1\\k=1}}^{n} X_{i(n+1)k} = 0 \forall i \in trucks$$
$$X_{i0(n+1)} = 0 \forall i \in trucks$$

Trucks constraints: start point, routes, points



$$\sum_{k=1}^{n} \sum_{\substack{i=1 \\ m}}^{m} X_{ijk} \le 1 \forall j \in point_of_distribution \setminus \{0, n+1\}$$

$$\sum_{j=1}^{n} \sum_{\substack{i=1 \\ m}}^{m} X_{ijk} \le 1 \forall k \in point_of_distribution \setminus \{0, n+1\}$$
Route constraints - points

$$\sum_{j=1/j\neq k}^{n} X_{ijk} = \sum_{j=1/j\neq k}^{n} X_{ikj} \, \underline{\forall}_i \in trucks, k \in point_of_distribution \setminus \{0, n+1\}$$



 $U_{ik} \le U_{ij} + 1 - (n+2)(X_{ijk} - 1) \forall i \in trucks, (j,k) \in point_of_distribution \times point_of_distribution / j \neq k$

 $U_{ik} \ge U_{ij} + 1 + (n+2)(X_{ijk} - 1) \forall i \in trucks, (j,k) \in point_of_distribution \times point_of_distribution / j \neq k$

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\sum_{j=1}^{n} weight_{j}V_{ij} \leq capacity_{i}Y_{i} \forall i \in trucks
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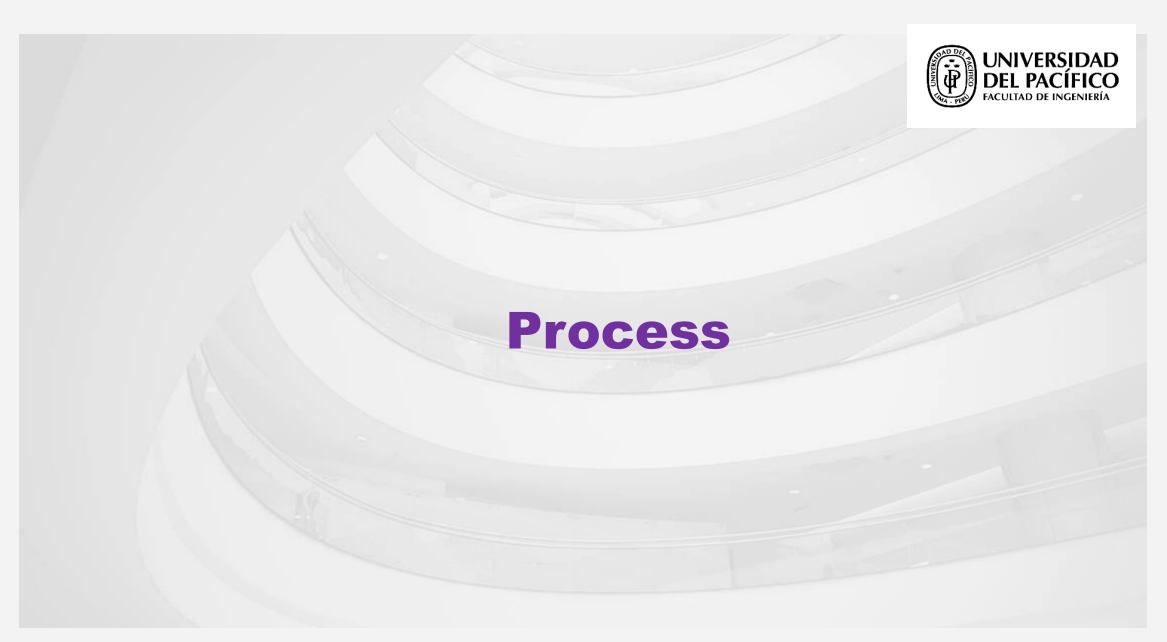
Balance constraints



$$\sum_{j=1/(j,s) \in ps}^{n} q_{js} W_j \le st_s \,\forall \, s \in sku$$

Inventory constraints

$$W_{k} = \sum_{i=1}^{m} \sum_{j=1}^{n} X_{ijk} \forall k \in point_of_distribution \setminus \{0, n+1\}$$



Data



The input was the structure of the trucks, orders, products and coordinates of the point of sales in a Latin American city

Data



Table 1Types of container used in the SKU of the simulated orders

	J	
	Quantity	Percentage
Container	SKU	SKU
Packaging board	7	15.9%
Glass	10	22.7%
PET	23	52.3%
Metal	4	9.1%
Total	44	100.0%

Source: author's own elaboration.

Process



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3 P1 4 P1638	0 569.23		301874.9 290936.9	from SolverStudio import *		
5 P1639	188.3		290930.9	m = Model("Camiones")		
6 P1655	46.6		292738.2			
7 P1658	66.3		290408.9	## Define variables		
8 P1659	104.8		290067.3	X=() for c in CAMION:		
9 P1660	4.1	0	290564.3	for d in DESTINO:		
10 P1661	15	2	290777.3.	for e in DESTINO:		
11 P1662	125.8	8	291094.4	<pre>if d!=e: X[c,d,e] = m.addVar(vtype=GRB.BINARY, name='X %s %s %s</pre>		
12 P1663	154.3	8	291538.8	<		
13 P1664	77.5		296415.3.	Model Output		
14 P1665	23.2		285327.8	## Executing C:\Users\Juan\Downloads\SolverStudio 00 09 03 00		
15 P1667	81.7		294124.4	20160520\SolverStudio_00_09_03_00 20160520\SolverStudio\SolverStudio\GurobiPython		
16 P1668	29.3		290373.6	\RunGurobiPython.py ## Building Gurobi input file, module 'SolverStudio.py', for 16 data items		
17 P1669	128.1		301851.	## Building Gurobi Input file, module 'Solverstudio.py', for 16 data items ## Writing data items		
18 P1670	120.93	3	284370.2	Academic license - for non-commercial use only		
19 P1671	Pedidos Detalle Distancias		290697.4	Changed value of parameter MIPGap to 0.05 Row 188, Column 15		

Process



NOATENDIDOS= m.addVar(obj=1000, name='NOATENDIDOS') ECAMIONES= m.addVar(obj=1, name='ECAMIONES') EDST= m.addVar(obj=1, name='EDST')

m.update()
m.setObjective(ECAMIONES+NOATENDIDOS*1000+EDST+quicksum(D2CAP[i] for i in CAMION));

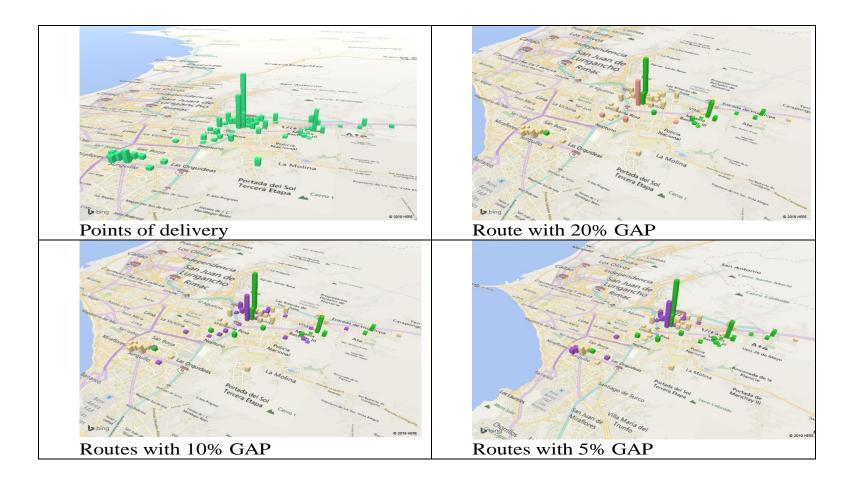
m.addConstr(quicksum(X[i,k,DESTINO[len(DESTINO)-1]] for k in DESTINO if 0!=DESTINO.index(k) and len(DESTINO)-1!=DESTINO.index(k))==Y[i]);



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Results – Order point of sales





Source: author's own elaboration.



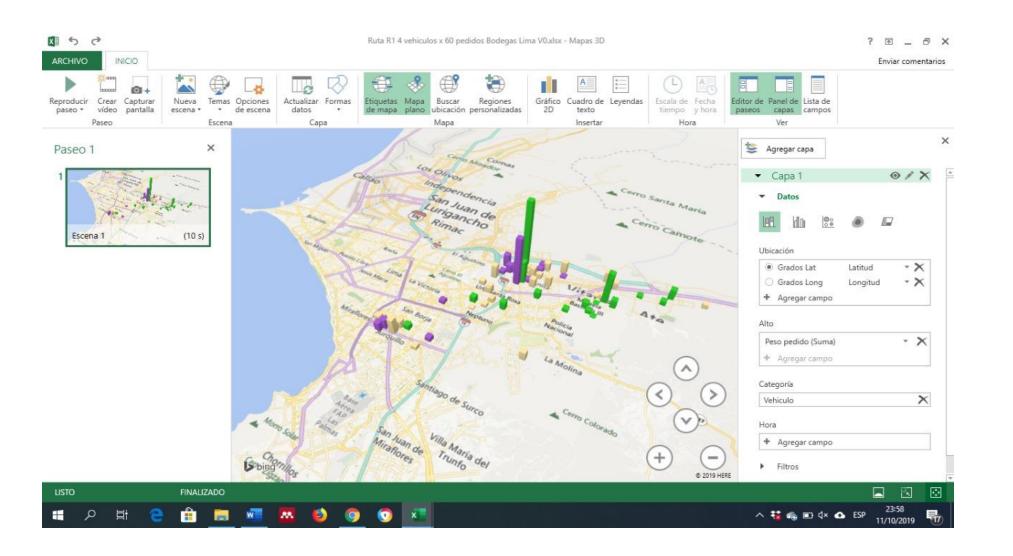




Table 2

The meeting of goals with sufficient inventory

Points of	Vehicles	% of	% of	Route
delivery	assigned	orders filled	capacity used	Distance (km)
10	1	100	23	15.9
20	1	100	56	23.6
30	1	100	50	24.6
40	1	100	66	26.6
50	1	100	88	29.4
60	2	100	78	33.2
70	2	100	79	36.6

Source: author's own elaboration.



The solutions allowed the scheduling of orders that had specific transportation units assigned with a graphical route



The percentage of filled order will improve the decision making process when the time for a route design is limited





We propose a model to design the shortest routes that can be used in an heterogeneous fleet with known capacities

The results show that it is possible to find a solution when the process time depends on the number of delivery points



One of the main contributions of the goal programming model proposal is that allows to manage the priorities in a distribution planning

As future research, we propose other constraints such as route options, time windows, among others



Managerial implications The results facilitate coordination within the The points of distribution can be supplied by located warehouses stakeholders Research implications We can optimize considering the preventive and The model combines multiple operational goals response stage





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