

# Trends in Transportation and Logistics and the Role of Optimization

M. Grazia Speranza University of Brescia

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# **Optimization in transportation and logistics**

1960's and 70's

"Transportation science" emerged

"Transportation" meant traffic and public transportation

"Logistics" referred to physical distribution and inventory management





### The contributions



# The contributions: efficiency

#### Efficient use of scarce resources

Scarcity is a fact of life. While people's desire for goods is unlimited, the resources to produce them are limited





# The contributions: systemic approach





# The contributions: looking ahead

#### Anticipation of consequences





### The contributions: in one word





# Technology







# Trends in transportation













### Moreover...



Meeting the needs of the present without compromising the ability of future generations to meet their needs



# Technology and optimization





# Transportation (freight and people)

Freight-empty returns:31% for companies23% for logistic companies

**People**-use of personal cars: 70-75%

Waste of capacity

**Unnecessary emissions** 



### Directions in research

**Systemic** 

**Data-driven** 

#### **Technology-driven**

**Collaborative** 



Dynamic

# Freight







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Pick-up

Delivery

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Archetti, Christiansen, Speranza, EJOR, 2018

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- Sequential policy: each customer independent
- Coordinated policy

	% total cost (average)	% total cost (max)
T=3	40.47	63.19
T=6	27.66	40.28
Low inventory cost	36.36	52.97
High inventory cost	34.67	63.19
All	35.54	63.19



#### **Reduction of emissions: 35.54%**







Archetti, Feillet, Mor, Speranza, EJOR, 2020 Archetti, Feillet, Mor, Speranza, COR, 2018













	Avg.	Avg.	Avg.	Avg.	Avg.	Avg.		
	Deterministic	Stochastic	R	RD	RDW	"myopic"		
$\delta = 0$	2.69	0.72	1.77	1.68	1.67		11.96	
$\delta=$ 0.5	2.96	1.40	2.35	2.17	2.02		14.74	
$\delta = 1$	3.10	1.34	3.03	2.39	2.25		10.85	
$\beta = 0.5$	1.91	1.28	1.67	1.61	1.51		24.22	
eta = 1	3.79	0.90	2.37	2.56	2.11		11.10	
$\beta = 1.5$	3.05	1.29	2.40	2.09	2.02		2.23	
Avg.	2.92	1.16	2.14	2.09	1.88		12.52	

#### Iterated local search

**Reduction of emissions: 13%** 















Causes of double parking:

- Scarcity of L/U areas
- Need to deliver
- Individual routes and schedules

Consequences of double parking:

- One less lane available
- Queues of vehicles
- Emissions









A vehicle makes a booking of the L/U areas

#### Windows of (un)availability for the following vehicles



Mor, Speranza, Viegas, TR E, 2020

OF BRESC





### Fixed and flexible starting time

**Fixed starting time of the route**:  $T_0 = q$ 

(the vehicle decides the starting time of the route)

#### **Routing only**

**Flexible starting time of the route**:  $T_0 \ge q$  and  $T_0 \le q+f$ 

(the starting time of the vehicle is between q and q+f)

#### **Routing + Scheduling**



### **Computational experiments**





Lisbon: 560m x 260m

### Tests

Independent vehicles: TSP for each vehicle (Lin-Kernighan heuristic)

Sequential booking with fixed starting time (formulation solved with CPLEX 12.6)

Sequential booking with flexible starting time (formulation solved with CPLEX 12.6)

Trade-off between routing time and number of double parked vehicles



# Results (up to 50 vehicles)





# People













### Emissions and occupation of space



Public transportation





Personal car



Shared taxi





#### A simulation/optimization model

#### Input:

- Origins
- Destinations
- Request time
- Desired departure time
- Flexibility factor





#### Archetti, Speranza, Weyland, ITOR, 2018





#### **Reduction of emissions: more than 50%**

# Fair collaboration

Shared taxi service (reservation-based)

Multiple companies

Collaboration initiative







# Fair collaboration



### UNIVERSITY OF BRESCIA

M.Grazia Speranza

#### Routes without collaboration

# Fair collaboration







Routes with full/ unconstrained collaboration Routes with fair/constrained collaboration

# The problem: basic formulation

$$\min z = \sum_{i \in V} \sum_{j \in V} \sum_{k \in K} c_{ij} x_{ij}^k$$

+ Routing constraints

+ Time windows and maximum time on board constraints

- + Capacity constraints
- $\begin{array}{lll} y_c^k \in \{0,1\} & k \in K \quad c \in C \\ x_{ij}^k \in \{0,1\} & k \in K \quad i \in V \quad j \in V \\ w_i^k \in \mathbb{N} & k \in K \quad i \in V \\ u_i^k \geq 0 & k \in K \quad i \in V \\ r_c \geq 0 & c \in C. \end{array}$

#### **Unconstrained optimization**



# The problem: time balance





# The problem: customer balance





# Adaptive large neighborhood search

while  $T_{\rm max} > 1$  do  $-q := resizeNeighborhood(E, w, q, q_{min}, q_{max}, p);$ Draw a destroy and a repair operator; - Destroy the current solution x; - Repair the destroyed solution and obtain x'; - if  $cost(x') \ge cost(x^*)$  then | -w = w + 1;-u randomly drawn in U(0,1); A worse solution may be if  $u < e^{\frac{cost(x^*) - cost(x')}{T_{\max}}}$  then accepted (with decreasing -x = x': probability) if  $cost(x') < cost(x^*)$  then  $-x^* = x'$ : - if r > R then -r = 0: – Set destroy and repair operators' scores to initial values; else Update destroy and repair operators' scores -r = r + 1; $-T_{\max} = T_{\max} * \gamma;$ return  $x^*$ :



# Instances

Map-based 142 instances (real travel time with Graphhopper)

112 instances (2 to 10 companies, 4 to 10 customers)
4 cities (Paris, Berlin, London, Rome)
4 sizes
7 demand scenarios per city

30 instances - Paris only – 5 different demand scenarios for:

- E1: 6 companies, 50 customers
- E2: 3 companies, 100 customers
- E3: 2 companies, 150 customers
- E4: 1 company with 150 customers, 3 with 50
- E5: 1 company with 150 customers, 1 with 100, 1 with 50
- E6: 1 company with 200 customers, 2 with 50



# Savings on large instances

Acceptable % of more or less time or customers							
Model	$\alpha$ (%)	Group	Group E				
UC	-	Avg. 18.05	Max. 25.30	Std. 3.07			
T-CDARP	10	17.08	24.56	3.55	14		
	$\frac{20}{30}$	$\frac{17.73}{18.1}$	24.79 25.14	$\frac{3.25}{2.98}$			
C-CDARP	10	15.19	23.14 22.32	3.76			
	20	16.6	23.16	3.93			
	30	17.03	25.21	3.32			
TC-CDARP	10	15.01	22.30	3.71			
	20	16.57	23.16	2.75			
	30	16.97	24.26	3.48			



#### **Reduction of emissions: 15%**













# Traffic assignment





 $t_{ij}^{FF} \left[1 + 0.15 \left(\frac{x_{ij}}{u_{ij}}\right)^4\right]$ 

Travel time on arc (i,j) with flow  $x_{ij}$ 

Min Total travel time on <u>paths of limited length</u>

Non linear optimization problem

on an exponential number of paths



Min Total travel time (piecewise approximation)

#### LP and MILP with exponential number of binary variables (FP-UC-SO and L-UC-SO)











#### **Reduction of emissions: 20%**



### The future





#### Ridesharing in the sky





# The future



