




Sustainable Logistics

June 9, 2016
MIT Center for Transportation & Logistics
Cambridge, Massachusetts




Dr. Josué C. Velázquez Martínez
Director, SCALE Latin America



The SCALE LatAm - Mission

To lead impactful & applied research in logistics and supply chain management relevant to Latin America

- Lead & support academic partners
- “Push” an innovative research agenda
- “Push” an innovative education agenda

Research Agenda

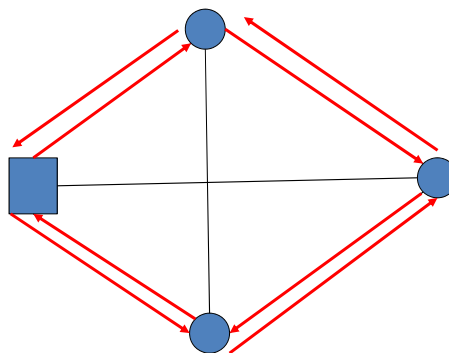
- Sustainability
- International Trade & Logistics
- Urban Logistics
- Applied Operations Research
- Innovative Case Studies in Logistics & SCM
- Logistics Efficiency in SMEs *



~80 Researchers from all over
Latin America
~20 Practitioners
~20 Students

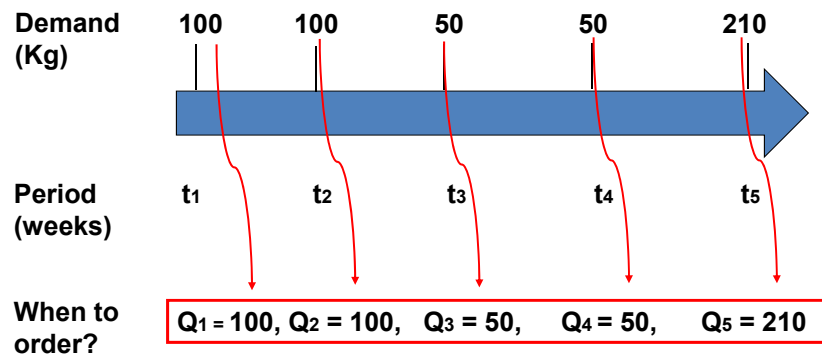


Decide the best route for this vehicle

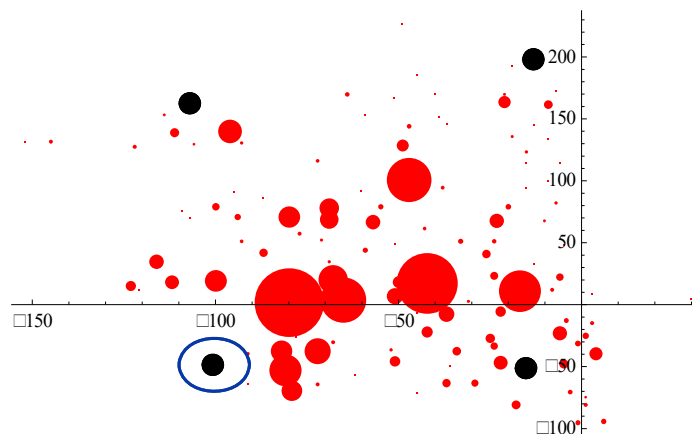


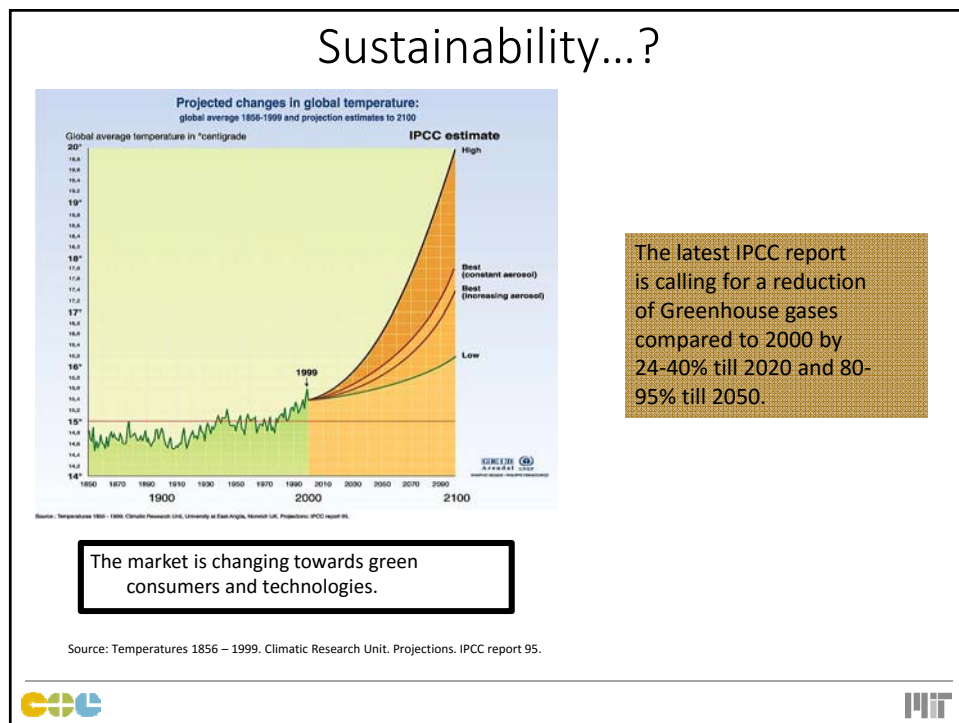
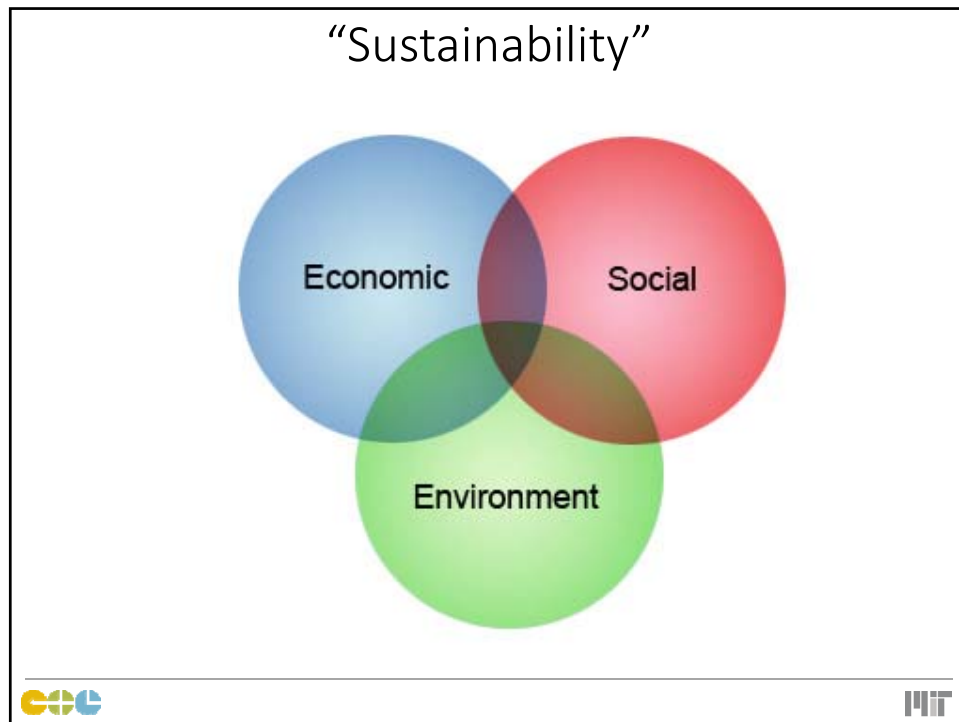
Decide how much and when to order?

Setup cost = \$5 Inventory cost = \$0.50



Decide the location of the DC?



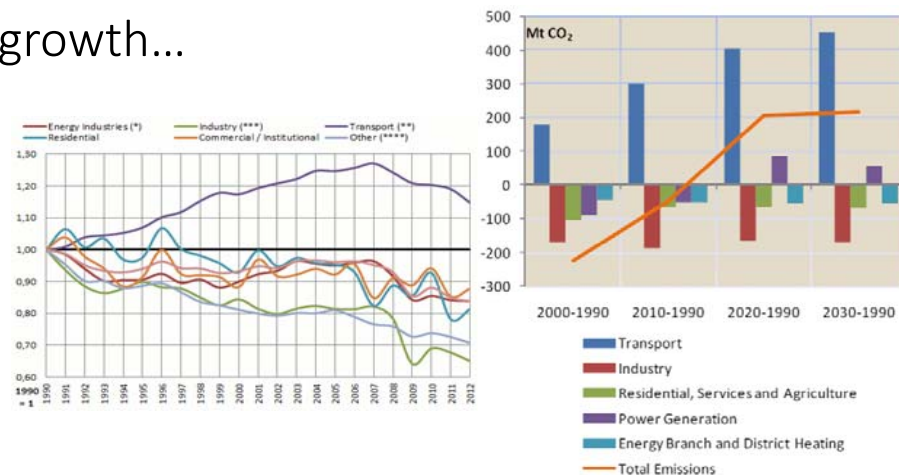


Market shift towards disclosing carbon emissions

1525 companies in 2010 & ~1000 commit to a self-imposed carbon target (CDP, 2011).



Transport emissions are by far the main contributor to emissions growth...



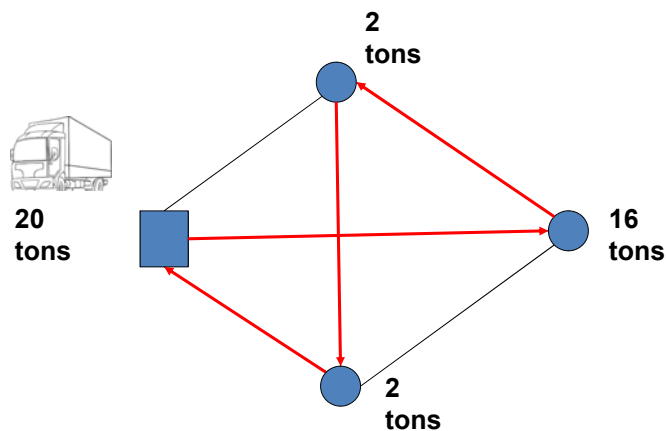
Source: European Commission (2007)



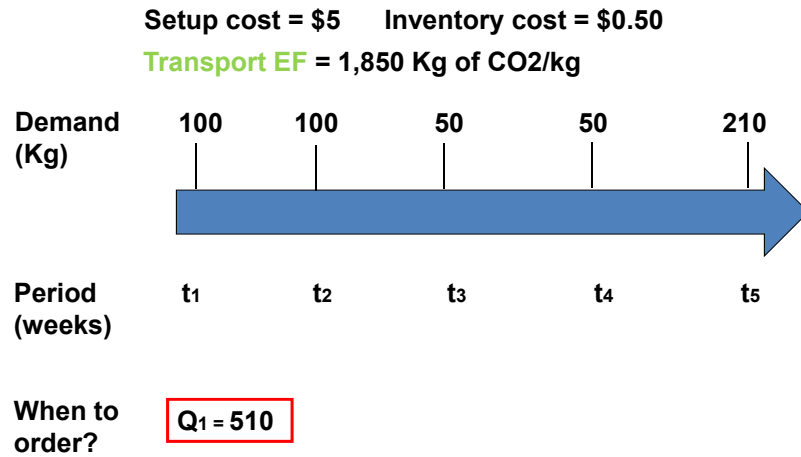
What could explain it? ... Urbanization



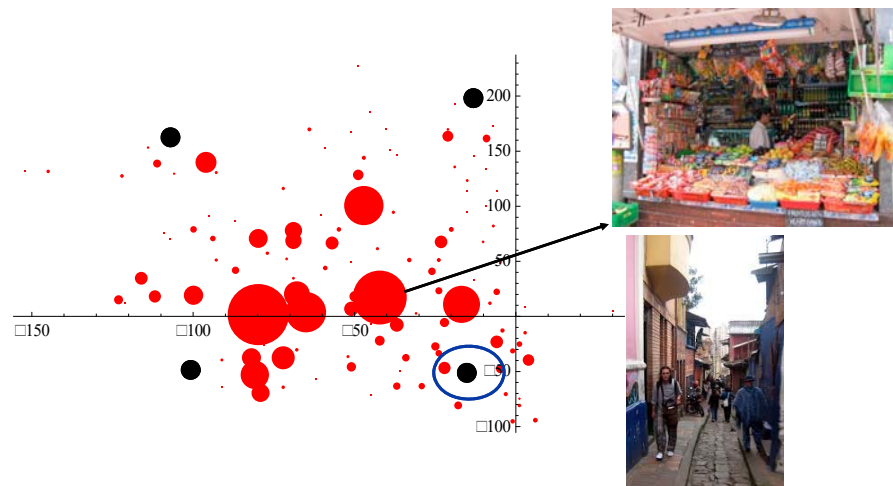
Decide the best route for this vehicle



Decide how much and when to order?



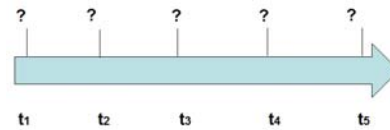
Decide the location of the DC



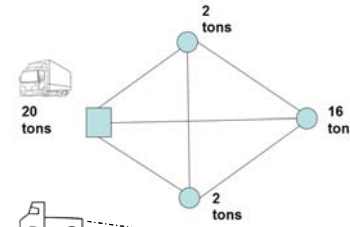
The Sustainable Logistics Initiative

How do companies can look for alternatives to reduce CO2 emissions by making smarter logistics decisions? Examples:

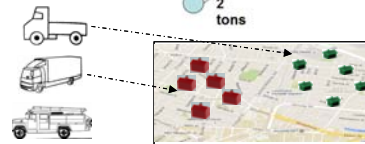
Decide how much and when to order?



Decide the best route for this vehicle



Decide the best vehicle for a region



Carbon Footprinting Aggregation?



- Support Measurement of CO2 emissions
- Identify potential savings during the logistics operations
- Impact of carbon footprinting aggregation– Published on 2014

Article
Flexible Services and Manufacturing Journal
June 2014, Volume 26, Issue 1, pp 196-220
First online: 05 January 2013

The impact of carbon footprinting aggregation on realizing emission reduction targets

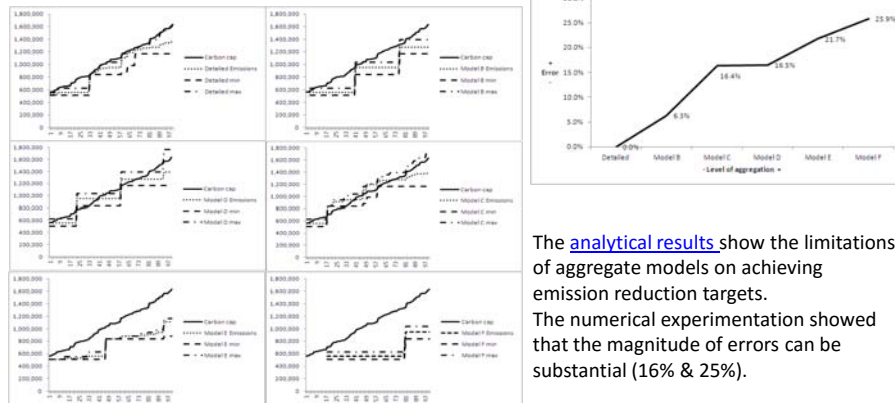
Josué C. Velázquez-Martínez, Jan C. Fransoo, Edgar E. Blanco, Jaime Mora-Vargas



Impact of aggregation

Dynamic Lot-Sizing Problem
(Wagner and Whitin, 1958)
with backorder assumption.

	Lot size per shipment	Average demand	FTL
EF per truck	Model A	Model C	Model E
Standard EF	Model B	Model D	Model F



The [analytical results](#) show the limitations of aggregate models on achieving emission reduction targets. The numerical experimentation showed that the magnitude of errors can be substantial (16% & 25%).

Source: Velázquez-Martínez et al. (2014). "The impact of Carbon Footprinting Aggregation on Realizing Emission Reduction Targets". Flex Serv Manuf J



Truck assignment for fuel savings



- Overall assignment of trucks for a DC in Mexico City
- Recommendations for fuel savings of ~3% & 26%
- 1st part of the Methodology –Published on 2016



Transportation Research Part D: Transport and Environment

Volume 43, March 2016, Pages 133–144

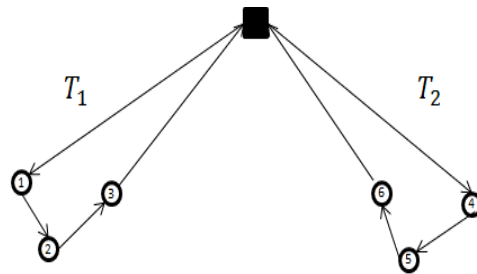


A new statistical method of assigning vehicles to delivery areas for CO₂ emissions reduction

Josué C. Velázquez-Martínez^{a,*}, Jan C. Fransoo^{b,1}, Edgar E. Blanco^{a,2}, Karla B. Valenzuela-Ocaña^{a,3}



Vehicle Assignment for CO2 emissions Reduction



Example:

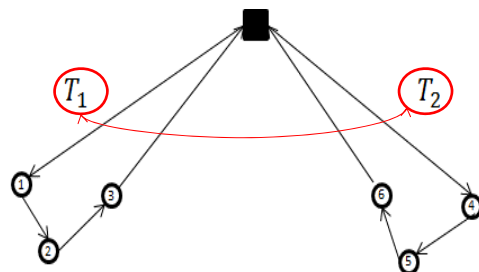
- Two trucks (T_1 & T_2) with same capacity
- Both travel same distance
- T_1 delivers in a flat area & regular (Av Speed 25 m/s)
- T_2 delivers in a pronounced area, congested, etc. (Av Speed 5.5 m/s)

Source: Velázquez-Martínez et al. (2016). "A new statistical method of assigning vehicles to delivery areas for CO2 emissions reduction". Transportation Research Part D



Vehicle Assignment for CO2 emissions Reduction

Suppose that by studying the historical performance of the trucks over time, we observe statistical evidence that a type of truck T_1 performs 20% better in regions with pronounced slopes and T_2 performs 12% better in plane roads.



Savings of 18% in carbon emissions

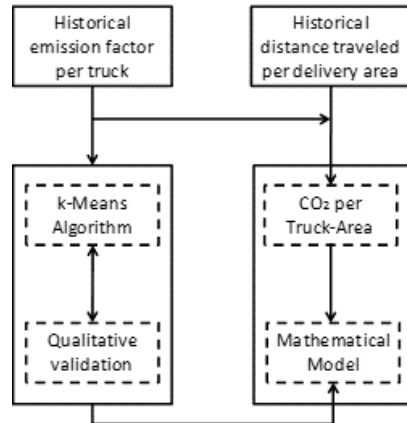
Note that this result is independent on the routing

Source: Velázquez-Martínez et al. (2016). "A new statistical method of assigning vehicles to delivery areas for CO2 emissions reduction". Transportation Research Part D

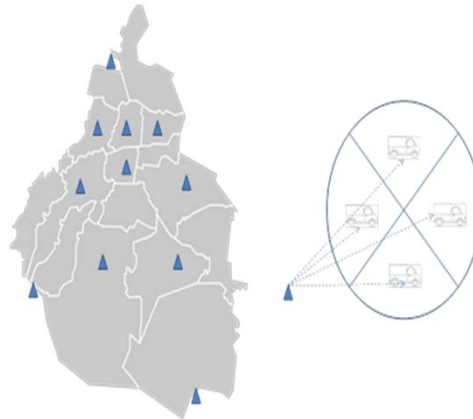


Vehicle Assignment for CO₂ emissions Reduction

Conceptual Model



Case Study: Parcel Company in Mexico City



Source: Velázquez-Martínez et al. (2016). "A new statistical method of assigning vehicles to delivery areas for CO₂ emissions reduction". Transportation Research Part D



Av emission factor analysis

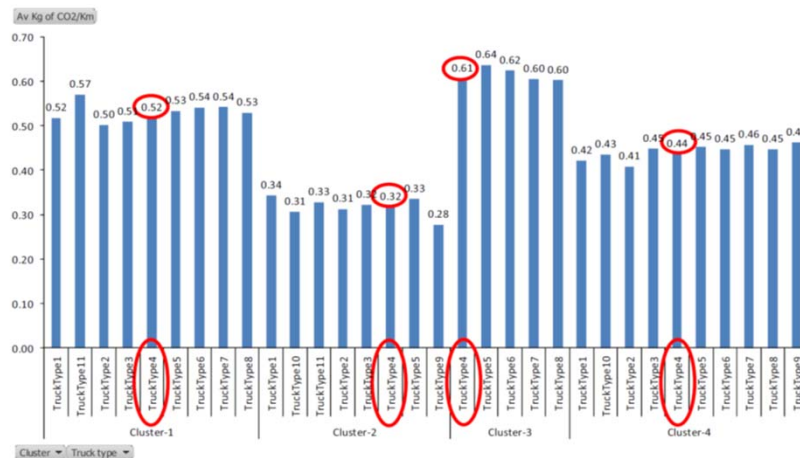


Fig. 7. Average emission factors per type of truck per cluster.

Source: Velázquez-Martínez et al. (2016). "A new statistical method of assigning vehicles to delivery areas for CO₂ emissions reduction". Transportation Research Part D



CO2 vehicle assignment comparison (-3% = 40 tons of CO2 = 12 vehicles)

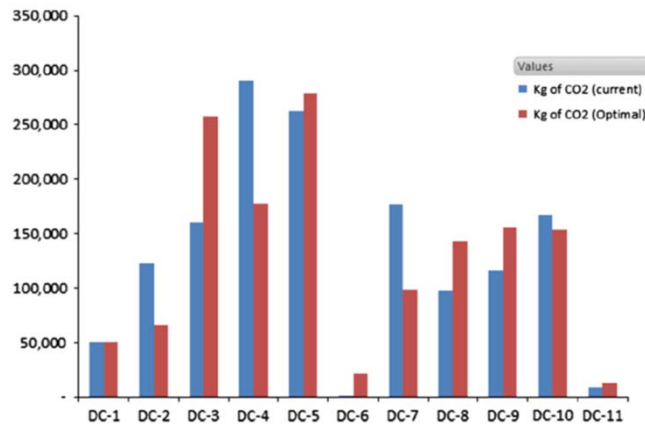


Fig. 8. Comparison between optimal and current CO₂ emissions per distribution center.

Source: Velázquez-Martínez et al. (2016). "A new statistical method of assigning vehicles to delivery areas for CO₂ emissions reduction". Transportation Research Part D



Practical application

Heineken study: Two trucks exchange pilot (Feb 2014 vs Feb 2015)

Route	Demand increase	Fuel efficiency (Km/l)		CO ₂ emissions (Tons)		CO ₂ Savings
		Feb 2014	Feb 2015	Feb 2014	Feb 2015	
RF1404	44%	3.25	3.26	1.47	1.12	25%
RF1407	42%	3.1	3.23	1.45	1.05	27%

With the truck exchange, Heineken distributes **more cargo** with **less fuel** consumption (about 292 liters less ~26% of savings)



Green Facility Location



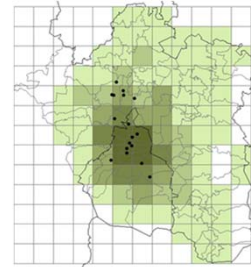
- Provide alternatives for the location of DCs in Mexico City
- Recommendations for savings of ~40% of CO2 emissions
- New Methodology – To be Published on 2016

Springer Series in Supply Chain Management
© 2016



Sustainable Supply Chains

Editors: Bouchery, Y., Corbett, C.J., Fransoo, J.C., Tan, T. (Eds.)



Green Facility Location

The Multiobjective Sustainable Facility Location Model

$$\text{Min} \rightarrow OF1 = \sum_{j \in J} \sum_{i \in I} \left(A_{ij} \frac{h_i}{W_i} + v_{ij} d_{ij} \left[\frac{h_i}{W_i} \right] \right) Y_{ij}$$



$$\text{Min} \rightarrow OF2 = \sum_{j \in J} \sum_{i \in I} d_{ij} \left[\frac{h_i}{W_i} \right] \left[f_i^e + (f_i^f - f_i^e) \frac{h_i}{W_i} \right] Y_{ij}$$



Subject to

$$\sum_{j \in J} Y_{ij} = 1 \quad \forall i \in I$$

$$\sum_{j \in J} X_j = p$$

$$Y_{ij} - X_j \leq 0 \quad \forall i \in I \quad \forall j \in J$$

$$X_j \in \{0,1\} \quad \forall j \in J$$

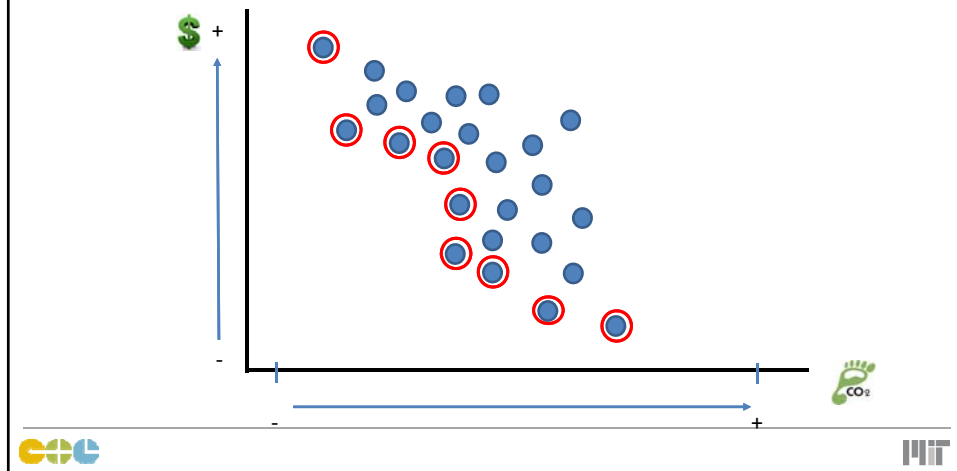
$$Y_{ij} \geq 0 \quad \forall i \in I, \forall j \in J$$



Solution Method

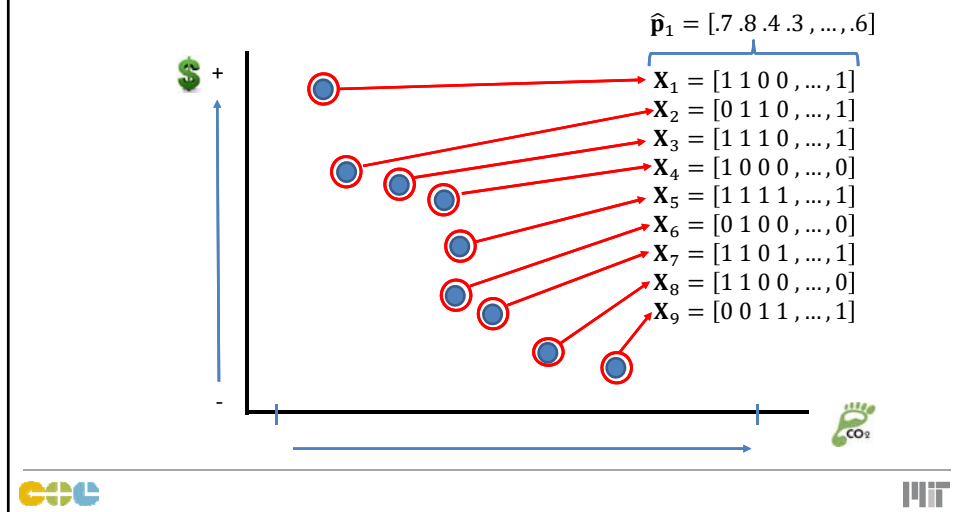
The Multiobjective Combinatorial Optimization CE Method (MOCO CEM)

$t = 0$, Draw a sample $\mathbf{X}_1, \dots, \mathbf{X}_N$ of Bernoulli vectors with success probability vector $\hat{\mathbf{p}}_{t-1}$.



The MOCO CEM - Algorithm

Pick the best solutions and estimate parameters

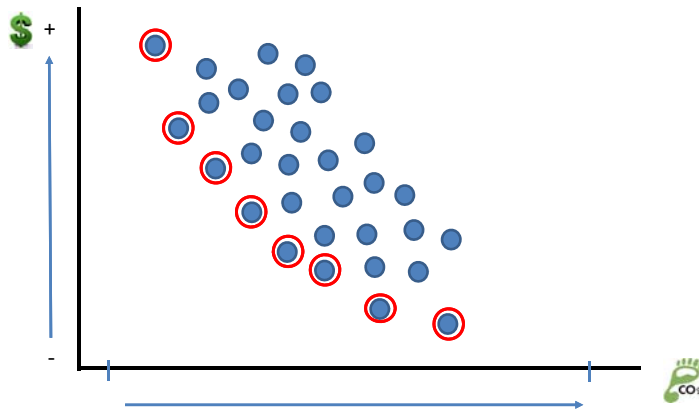


The MOCO CEM - Algorithm

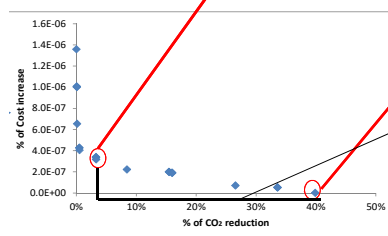
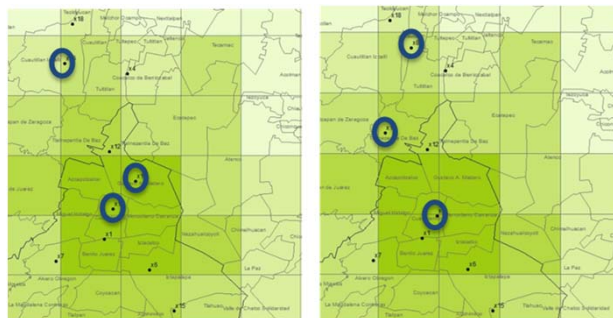
$t = 1$, Draw a sample $\mathbf{X}_1, \dots, \mathbf{X}_N$ of Bernoulli vectors with success probability vector

$\hat{\mathbf{p}}_{t-1}$

Pick the best solutions and estimate parameters...



Some Results...



40% of CO2 reduction -> No cost increase



