

# Estimation of Truck Counts with Multiple Truckload Categories: A Data-Fusion Approach and a Case Study in Florida

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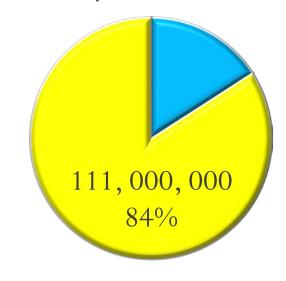
#### Truck trips in Florida in 2011



**Empty Trucks** 



Waste fuel







**Pollution** 



Pavement damage

## Works on estimating truck counts

- A two-stage approach by Jansuwan. et al.
- Network flow methodology by Mesa-Arango et al.
- Dynamic and stochastic models by Crainic et al.







Nonempty

We successfully estimated truck counts in different load categories

Three categories:





Fully loaded

Load categories:

**Empty** 

Half loaded

Five categories:









Load categories:

0-20%

20-40%

40-60%

60-80%

80-100%

## Data

• 1: Truck Weights  $(n_{la})$  and truck counts  $(n_a)$  on link a

Link ID	Vehicle weight (pounds)	Truck counts	
7_9918	26060	11755	

• 2: Commodity flows  $(m_w)$  and truck counts  $(n_w)$  between OD pair w

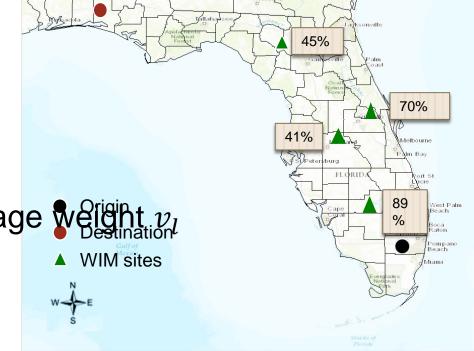
Origin- Destination	Commodity flows	Truck counts
319-323	36.68	1497

• 3: Path flows for freight trips  $p_{wa}$ 

OD pair	Link ID	Percentage		
319_359	18_9920	45%		

• 4. Truck weight categories l and average  $\mathbf{v} \in \mathbf{v}$ 

Category	Weight range	<b>Every weight</b>		
no.	(kips)	(kips)		
1	≤ 35	28.4		
2	35 - 60	42.3		
3	> 60	65.1		



#### Variables to estimate

- $y_{lw}$ : flow of category l trucks in between an OD pair w,  $y_{lw} \ge 0, l \in L, w \in W$
- $x_{la}$ : number of category l trucks passing through link a,  $x_{la} \ge 0, l \in L, a \in A$
- $\varepsilon_{la}$ : estimation error term

**Objective function:** minimize the sum of squared errors with  $C_1$  -  $C_5$  being weight

$$\min_{x,y} \left[ \sum_{l \in \mathcal{L}} \sum_{a \in A^{WS}} C_1 (n_{la} - x_{la})^2 \right] + \left[ \sum_{a \in A^T} C_2 \left( n_a - \sum_{l \in \mathcal{L}} x_{la} \right)^2 \right] + \left[ \sum_{w \in W^c} C_3 \left( m_w - \sum_{l \in \mathcal{L}} y_{lw} v_l \right)^2 \right]$$

$$+ \left[ \sum_{w \in W^T} C_4 \left( n_w - \sum_{l \in \mathcal{L}} y_{lw} \right)^2 \right] + \left[ \sum_{a \in A^T} C_2 \left( n_a - \sum_{l \in \mathcal{L}} x_{la} \right)^2 \right] + \left[ \sum_{w \in W^c} C_3 \left( m_w - \sum_{l \in \mathcal{L}} y_{lw} v_l \right)^2 \right]$$

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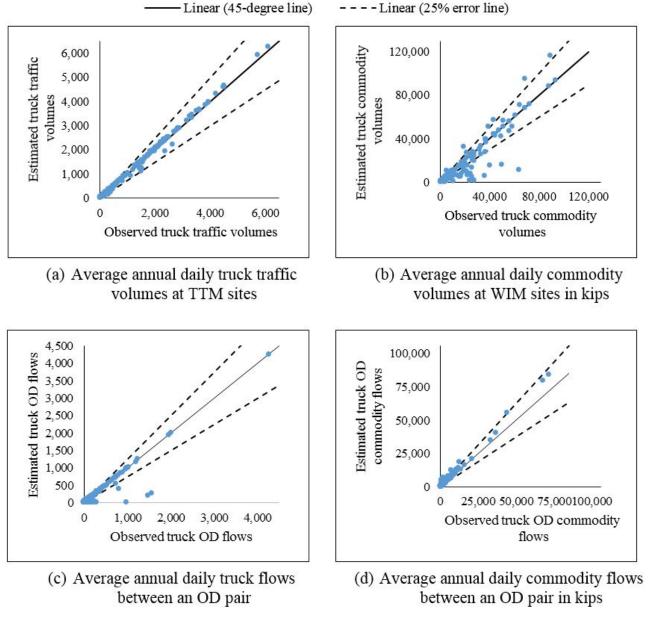
$$+ \left[ \sum_{a \in A^{WS}, l \in \mathcal{L}} C_5 \varepsilon_{la}^2 \right]$$

$$+$$

#### **Constraints**

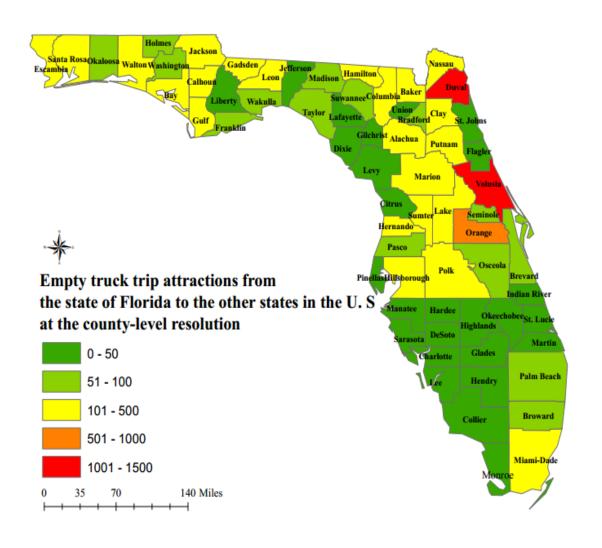
Subject to OD flow - link flow conversion:

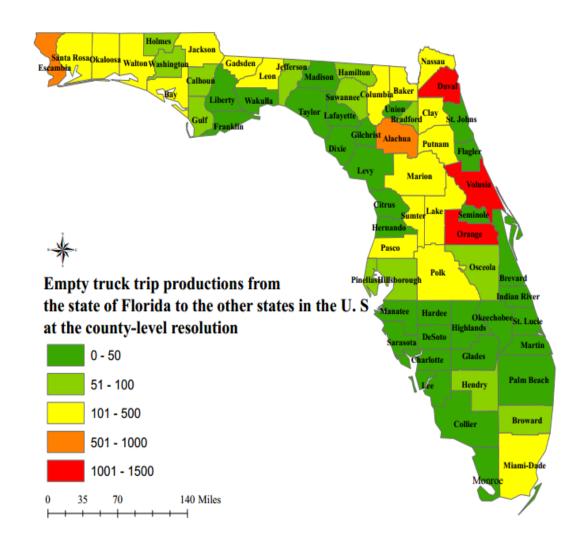
$$\sum_{w \in W} y_{lw} p_{wa} = x_{la} + \varepsilon_{la}, l \in L, a \in A$$



Observed versus estimated truck traffic volumes on links, truck OD flows, and commodity OD flows per day.

#### Empty truck flows between Florida and the other states in the US





- Conclusion:
- Dividing truck weights into several categories
- > An nonlinear optimization model
- Florida case study
- Acknowledgement

Thanks for the support of the Florida Department of Transportation

# Thank you!

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• Solver: Gurobi.



#### • Different sets for C's

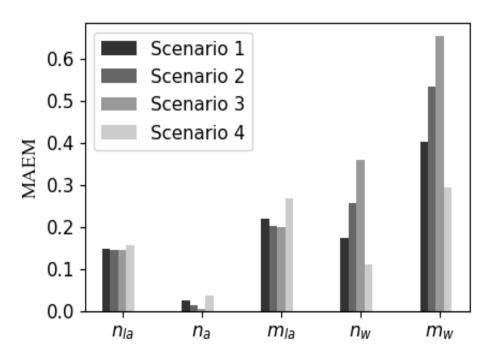
Weight	Scenario 1	Scenario 2 Scenario 3		Scenario 4
C1	1	10	100	1
C2	1	10	100	1
C3	1	1	100	1
C4	1	1	1	10
C5	1	1	1	10
C6	1	10	100	1

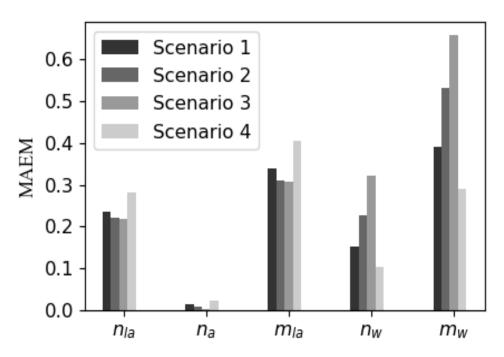
## Mean of Absolute Error to Mean (MAEM)

Mean of Absolute Error to Mean:

$$MAEM(\hat{\theta}) = \frac{E[|\hat{\theta} - \theta|]}{\bar{\theta}}$$

where  $E[|\hat{\theta} - \theta|]$  is the expected value of  $|\hat{\theta} - \theta|$ ,  $\hat{\theta}$  is the estimated value,  $\theta$  is the observed value, and  $\bar{\theta}$  is the mean of observed values.





(a) Three Truck-weight Categories Model

(b) Seven Truck-Weight Categories Model

 $n_{la}$  - total number of category l trucks passing through link a

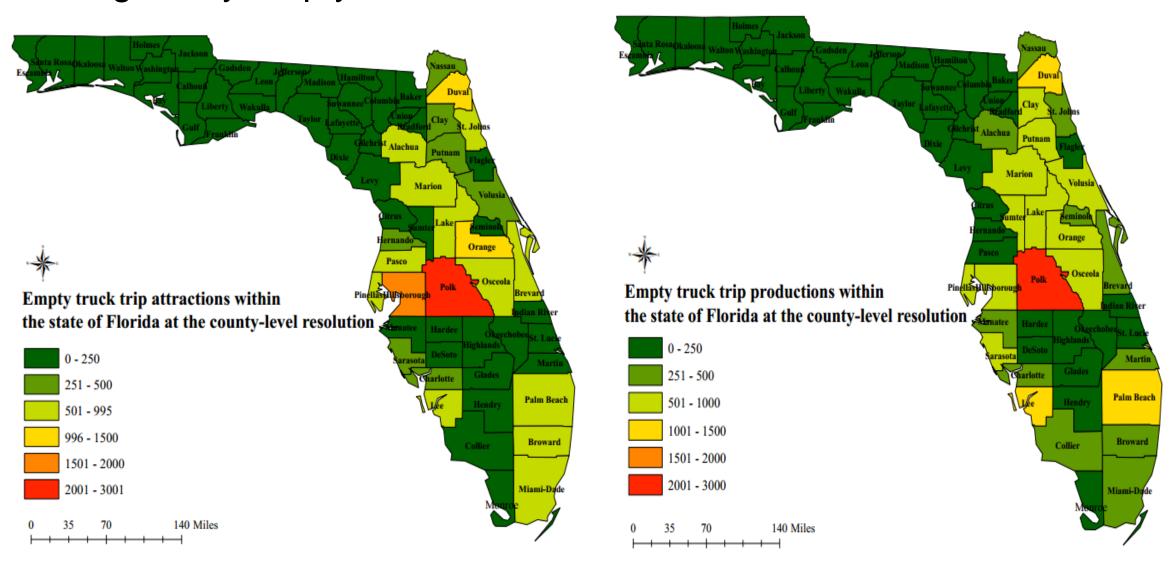
 $n_a$  - total truck flows on link a

 $m_{la}$  - total gross weight of category l trucks passing through link a

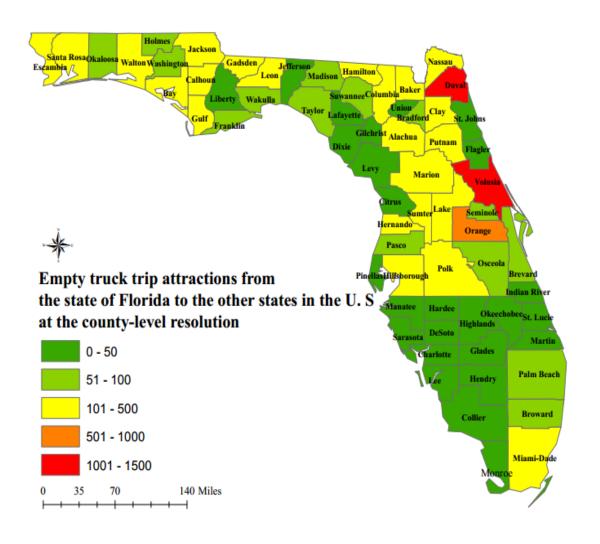
 $n_w$  - total truck flow between OD pair w

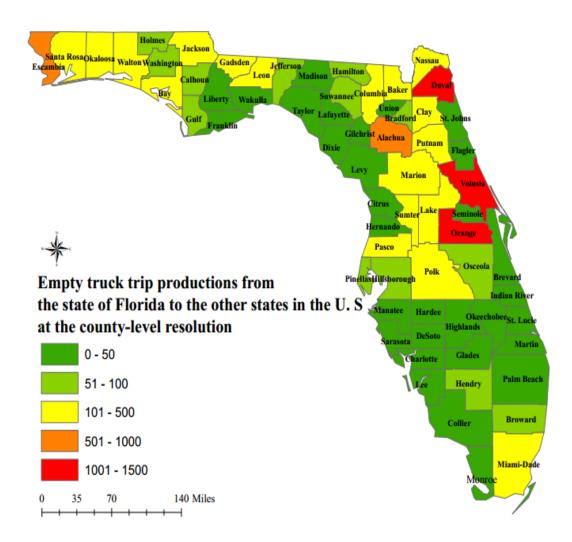
 $m_w$  - total commodity flow between OD pair w

### Average daily empty truck flows within Florida

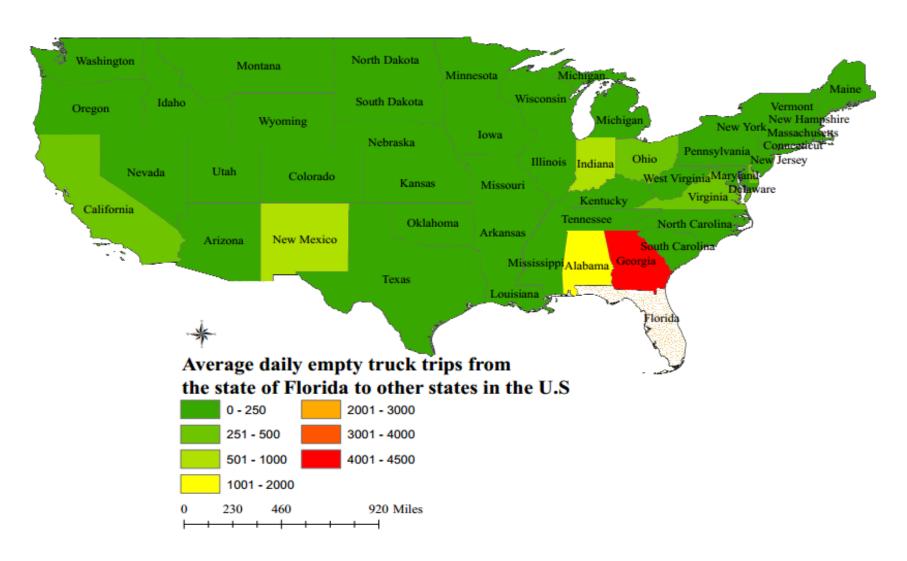


### Empty truck flows between other states and Florida

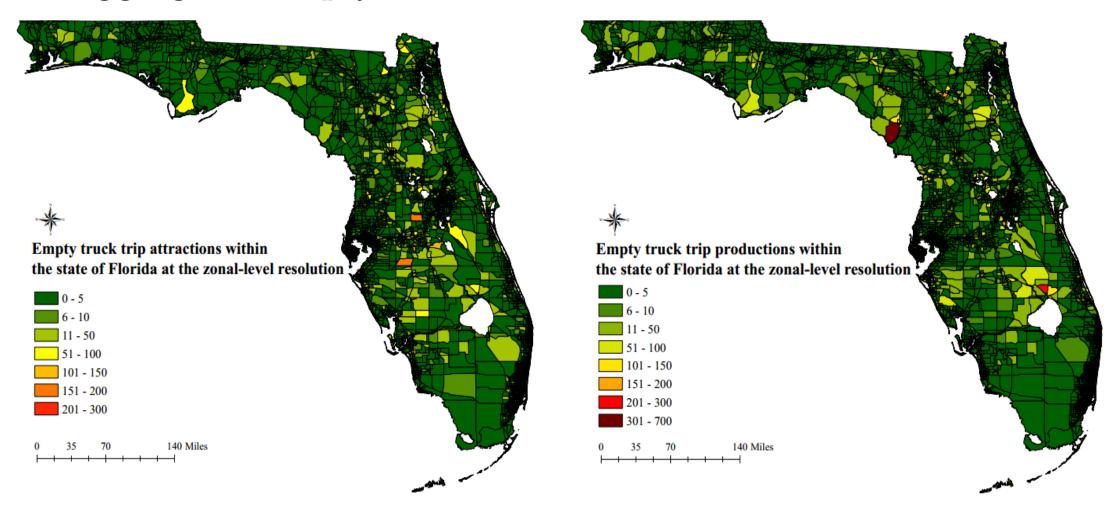




# **Estimated** empty truck flows from Florida to other states in the US



### Disaggregated empty truck flows within Florida



Estimated county level average daily empty truck trip attractions within Florida at SWTAZ level resolution

Estimated county level average daily empty truck trip productions within Florida at SWTAZ level resolution

# Methodology-Disaggregation of truck flows to newest FLSWM TAZ level (Holguín-Veras and Patil, 2008)

Holguín-Veras, J., and G. R. Patil. A multicommodity integrated freight origin—destination synthesis model. *Networks and Spatial Economics*, Vol. 8, No. 2, 2008, pp. 309-326.

# Estimated empty truck flows in TAZ level

#### **Parameters:**

- $k, \bar{k}$ : OD pairs in TAZ level,  $k, \bar{k}$  are in reverse direction
- $a_k$ : observed truck flows
- $w, \overline{w}$ : OD pairs in county level,  $w, \overline{w}$  are in reverse direction
- $p_w$ : in each OD pair w in county level, estimate empty truck flows of OD pair k as a proportion p of the corresponding truck flows of OD pair  $\bar{k}$
- $y_{0w}$ : empty truck flow in OD pair w

#### Variable:

•  $\bar{b}_k$ : estimated empty truck flows between OD pair k

• empty truck flows of OD pair k as a proportion p of the corresponding truck flows of OD pair  $\bar{k}$ .

$$\bar{b}_{\bar{k}} = p_w * a_k, \forall k \in w, \forall \bar{k} \in \overline{w}, \forall w, \overline{w} \in W$$

• sum of estimated empty truck flows  $\sum_{\bar{k}\in\bar{w}} \bar{b}_{\bar{k}}$  should be equal to the estimated empty truck flows between OD pair

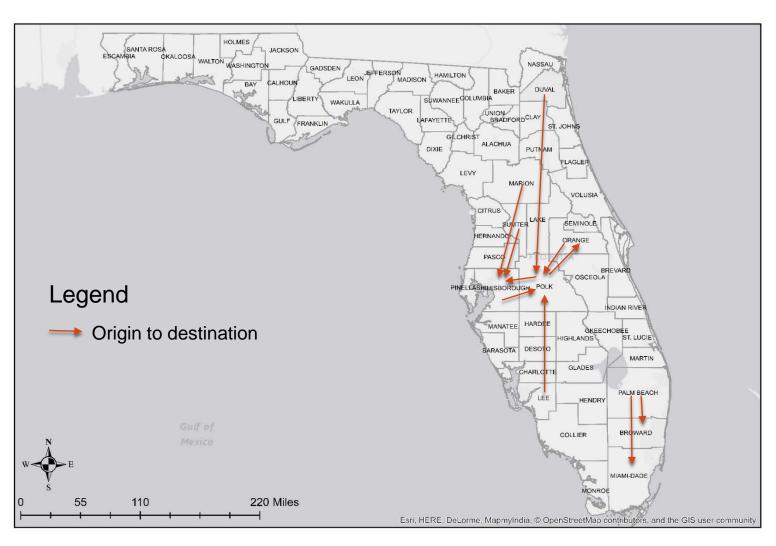
$$\overline{w}\sum_{ar{k}\in\overline{w}} \overline{b}_{ar{k}} = y_{0\overline{w}}, \forall \overline{w}\in W$$

• Union above equations and eliminate  $p_w$ :

$$\bar{b}_{\bar{k}} = y_{0\bar{w}} * a_k / \sum_{k \in w} a_k$$

## Top 10 OD pairs within Florida with highest number of empty flows

Origin	Destination	Average daily empty truck flows
Polk	Orange	447.42
Hillsborough	Polk	311.69
Sumter	Hillsborough	311.27
Palmbeach	Broward	299.00
Palmbeach	Miami Dade	266.32
Orange	Polk	261.06
Lee	Polk	242.67
Duval	Polk	234.11
Marion	Hillsborough	232.22
Polk	Hillsborough	228.84



## OD matrix of empty truck flows (deliverable format)

Destination zone Origin zone	1	2	3	4	5	6
1	175.37	0.98	0.00	1414.76	0.00	16.24
2	24.86	11015.72	3409.42	23643.44	13107.06	17295.67
3	0.00	7711.57	3730.46	10705.06	869.52	10351.20
4	514.71	14833.50	15086.26	51464.87	505.68	18441.74
5	0.00	12297.87	1055.78	4905.25	550.83	1144.43
6	0.00	18746.29	5225.65	13853.13	4673.33	22603.82
7	2239.04	51737.01	27705.42	111767.81	12362.30	65175.47
8	115.84	10912.60	794.89	13343.40	73.40	6368.06
9	289.42	2868.80	536.16	2346.72	98.89	4733.15

Top 1% OD pairs are highlighted in light red color