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Synchromodality in the Physical Internet Real-time Switching in a Multimodal Network with Stochastic Transit Times

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A deep decarbonization of the logistics industry is needed.

- Enormous amount of CO₂ emissions from the freight transportation sector
- Modal shift towards less carbon-intensive transportation modes



The Physical Internet

Connect logistics networks into an integrated network



Figure 1. Adapted from Roadmap: Corridors, Hubs and Synchromodality, by ETP-Alice. Retrieved from http:// etp- logistics.eu.

Synchromodality offers a solution to reduce emissions.

- Enormous amount of CO₂ emissions from the freight transportation sector
- Modal shift towards less carbon-intensive transportation modes



The Physical Internet > Synchromodality
 Connect logistics networks into an integrated network



Synchromodality - Towards the Physical Internet.



- Select the best transportation mode at all times
- Use modalities more efficiently and exploit all advantages

Synchromodality efficiently copes with uncertainty in transit times.

Synchromodality in a network with stochastic transit times

- Unreliability in transportation system
- Optimal transportation decision given the transit time outcome





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Synchromodality in a network with stochastic transit times

- Unreliability in transportation system
- Optimal transportation decision given the transit time outcome



Research objective

- Develop synchromodal planning model to construct optimal transportation routes
- Insights in the advantages of synchromodality







Cost

Service quality

Sustainability



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Synchromodality in a network with stochastic transit times

- Unreliability in transportation system
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Research objective

- Develop synchromodal planning **model** to construct optimal transportation routes
- **Insights** in the advantages of synchromodality









Service quality

Sustainability





Cost

Our model constructs optimal transportation routes given the stochastic transit times.

Input	Optimization model	Output
	- Operative to active al	- Decision avrida

- Multimodal network
- Set of orders

- Construct optimal transportation routes
- MILP

 Decision guide conditional on transit time outcome



Input of the model

Input

Multimodal network

2

- Set of orders
- Input parameters



− − ► Rail

Truck

8

6

·····▶ Barge



8

Destination terminals

Input of the model: stochasticity is modeled through scenarios.

Input

- Multimodal network
- Set of orders

Input parameters



Transit time stochasticity

- Transit time scenarios
- Probability distribution

The optimization model is formulated as a mixed-integer linear programming problem.



Optimization model

- Construct optimal transportation routes
- MILP

- Construct optimal transportation routes
 - Routing decisions
 - Stochasticity: adapt routes to real-time information

- Mixed-integer linear programming problem
 Minimize Leg transportation cost
 - Terminal transshipment cost
 - Overdue penalty cost

- Low cost

 On-time delivery



Output of the model



Output

 Decision guide conditional on the transit time outcome Which decision to take in a terminal, given the time period in which the decision is to be made







Output of the model

Terminal

Numerical study





Numerical study - Performance analysis.

Common practice	One mode	Multiple modes	Real-time information
 Unimodal road transportation 	X		
 Unimodal rail transportation 	X		
 Unimodal barge transportation 	x		
 Multimodal transportation 		X	
 Synchromodal transportation 		X	Х

Value of real-time planning



Synchromodality performs well in terms of cost, service quality and environmental impact.



Synchromodality offers a combination of advantages that allows to achieve sustainable freight transportation services at a favorable price and service quality.

Numerical study – Sensitivity analyses.



Numerical study – Sensitivity analyses.





Numerical study – Impact of changing cost parameters.



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Synchromodality > Multimodality

- Cost
- Service quality

regardless of the parameter value

Numerical study – Impact of changing cost parameters.



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Synchromodality > Multimodality

- Cost
- Service quality

regardless of the parameter value

Numerical study – Impact of changing cost parameters.



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Synchromodality > Multimodality > Value of real-time information

- Cost
- Service quality

regardless of the parameter value

Numerical study – Impact of carbon tax.





Numerical study – Impact of carbon tax.







Synchromodal

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Conclusion

We developed a synchromodal planning model to construct optimal transportation routes in a multimodal network with stochastic transit times.

Synchromodality offers a combination of advantages that allows to achieve sustainable freight transportation services at a favorable price and service quality.









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Additional information





Objective function



Optimization model

- Construct optimal transportation routes
- MILP

Objective function

Minimize

 $\sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\mathcal{C}^V]_{i,j,m,0}^n + \sum_{n \in \mathcal{O}} \sum_{g,m \in \mathcal{M}} E[\mathcal{C}^T]_{i,g,m,0}^n + \sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\rho]_{i,j,m,0}^n$





Optimization model

- Construct optimal transportation routes
- MILP

Objective function

Minimize

 $\sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\mathcal{C}^{V}]_{i,j,m,0}^{n} + \sum_{n \in \mathcal{O}} \sum_{g,m \in \mathcal{M}} E[\mathcal{C}^{T}]_{i,g,m,0}^{n} + \sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\rho]_{i,j,m,0}^{n}$

variable leg transportation cost





Optimization model

- Construct optimal transportation routes
- MILP

Objective function

Minimize

 $\sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\mathcal{C}^{V}]_{i,j,m,0}^{n} + \sum_{n \in \mathcal{O}} \sum_{g,m \in \mathcal{M}} E[\mathcal{C}^{T}]_{i,g,m,0}^{n} + \sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\rho]_{i,j,m,0}^{n}$ $\mathsf{terminal}$

transshipment cost



Optimization model

- Construct optimal transportation routes
- MILP

Objective function

Minimize

 $\sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[C^V]_{i,j,m,0}^n + \sum_{n \in \mathcal{O}} \sum_{g,m \in \mathcal{M}} E[C^T]_{i,g,m,0}^n + \sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\rho]_{i,j,m,0}^n$

overdue penalty cost





Optimization model

- Construct optimal transportation routes
- MILP

Objective function







Optimization model

 Construct optimal transportation routes

MILP

Objective function





Optimization model

- Construct optimal transportation routes
- MILP

Objective function

Minimize

 $\sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\mathcal{C}^V]_{i,j,m,0}^n + \sum_{n \in \mathcal{O}} \sum_{g,m \in \mathcal{M}} E[\mathcal{C}^T]_{i,g,m,0}^n + \sum_{n \in \mathcal{O}} \sum_{(i,j,m) \in \mathcal{A}} E[\rho]_{i,j,m,0}^n$

variable leg transportation cost terminal transshipment cost overdue penalty cost

on-time delivery



Optimization model

 Construct optimal transportation routes

MILP

Objective function





Constraints



Subjected to three constraint sets.



Constraints

- Network flow constraints
- Expected cost constraints

Optimization model

- Construct optimal transportation routes
- MILP

Expected penalty constraints





Optimization model

Constraints

Network flow constraints

 $\sum_{(o_n, j, m) \in \mathcal{A}} x_{o_n, j, m, 0}^n = 1$ $\forall n \in \mathcal{O}$ $\sum_{(i,j,m)\in\mathcal{A}} x_{i,j,m,0}^n = 1$ $\forall n \in \mathcal{O}$ $x_{i,j,m,t}^n \leq \sum_{(j,k,h)\in\mathcal{A}} x_{j,k,h,t+l_{i,j,m}}^n$ $\forall n \in \mathcal{O}, \forall (i, j, m) \in \mathcal{A}, \forall t \in \mathcal{T}, \forall s \in \mathcal{S}$ $\sum_{(i,j,m)\in\mathcal{A}} x_{i,j,m,t}^n \leq 1$ $\forall n \in \mathcal{O}, \forall i \in \mathcal{N}, \forall t \in \mathcal{T}$ $x_{d_n,d_n,storage,T}^n = 1$ $\forall n \in \mathcal{O}$ $\sum_{(i,j,m)\in\mathcal{A}} x_{i,j,m,T}^n = 1$ $\forall n \in \mathcal{O}$ $1 + w_{i,g,m,t+l_{k,i,g}^{s}}^{n} \ge x_{k,i,g,t}^{n} + x_{i,j,m,t+l_{k,i,g}}^{n} \quad \forall (k,i,g) \in \mathcal{A}, \forall (i,j,m) \ \mathcal{A}, \forall t \in \mathcal{T}, \forall s \in \mathcal{S}$ $w_{i,storage,m,0}^n \ge x_{i,j,m,0}^n$ $\forall (i, j, m) \in \mathcal{A}, i \in \mathcal{N}_o$ $x_{i,j,m,t}^n \in \{0,1\}$ $\forall n \in \mathcal{O}, \forall (i, j, m) \in \mathcal{A}, \forall t \in \mathcal{T}$ $w_{i,g,m,t}^n \in \{0,1\}$ $\forall n \in \mathcal{O}, \forall i \in \mathcal{N}, \forall g, m \in \mathcal{M}, \forall t \in \mathcal{T}$



Optimization model

Constraints

Expected cost constraints

$$\begin{split} E[C^V]_{i,j,m,t}^n &\geq c_{i,j,m}^V f_n + \sum_s p_{i,j,m}^s \sum_{(j,k,h) \in \mathcal{A}} E[C^V]_{i,j,m,t+l_{i,j,m}}^n - Z(1 - x_{i,j,m,t}^n) \\ &\forall n \in \mathcal{O}, \forall (i,j,m) \in \mathcal{A}, \forall t \in \mathcal{T} \\ E[C^V]_{i,j,m,t}^n &\geq 0 \qquad \qquad \forall n \in \mathcal{O}, \forall (i,j,m) \in \mathcal{A}, \forall t \in \mathcal{T} \end{split}$$

$$\begin{split} E[C^{T}]_{i,g,m,t}^{n} &\geq c_{g,m}^{T} f_{n} + \sum_{s} p_{i,j,m}^{s} \sum_{(j,k,h) \in \mathcal{A}} E[C^{T}]_{i,m,h,t+l_{i,j,m}}^{n} - Z\left(1 - w_{i,g,m,t}^{n}\right) - Z\left(1 - x_{i,j,m,t}^{n}\right) \\ &\quad \forall n \in \mathcal{O}, \forall (i,j,m) \in \mathcal{A}, i \in \mathcal{N}_{t}, \forall g \in \mathcal{M}, \forall t \in \mathcal{T} \\ E[C^{T}]_{i,g,m,t}^{n} &\geq \sum_{s} p_{i,j,m}^{s} \sum_{(j,k,h) \in \mathcal{A}} E[C^{T}]_{i,m,h,t+l_{i,j,m}}^{n} - Z\left(1 - w_{i,g,m,t}^{n}\right) - Z\left(1 - x_{i,j,m,t}^{n}\right) \\ &\quad \forall n \in \mathcal{O}, \forall (i,j,m) \in \mathcal{A}, i \in \mathcal{N}_{o}, \forall g \in \mathcal{M}, \forall t \in \mathcal{T} \\ E[C^{T}]_{i,g,m,t}^{n} &\geq 0 \qquad \forall n \in \mathcal{O}, \forall i \in \mathcal{N}, \forall g, m \in \mathcal{M}, \forall t \in \mathcal{T} \end{split}$$



Optimization model

Constraints

Expected penalty constraints

$$\begin{split} E[\rho]_{i,j,m,t}^{n} &\geq \sum_{s} p_{i,j,m}^{s} \sum_{(j,k,h) \in \mathcal{A}} E[\rho]_{j,k,h,t+l_{i,j,m}}^{n} - Z(1-x_{i,j,m,t}^{n}) \\ &\quad \forall n \in \mathcal{O}, \forall (i,j,m) \in \mathcal{A}, i \in \mathcal{N}_{0} \cup \mathcal{N}_{t}, \forall t \in \mathcal{T} \\ E[\rho]_{i,j,m,t}^{n} &\geq (t-k_{n}) f_{n} \rho - Z(1-x_{i,j,m,t}^{n}) \\ &\quad \forall n \in \mathcal{O}, \forall (i,j,m) \in \mathcal{A}, i \in \mathcal{N}_{d}, \forall t \in \mathcal{T} \\ E[\rho]_{i,j,m,t}^{n} &\geq 0 \\ &\quad \forall n \in \mathcal{O}, \forall (i,j,m) \in \mathcal{A}, \forall t \in \mathcal{T} \end{split}$$





Numerical study

Sensitivity analyses



Numerical study – Impact of penalty cost





Numerical study – Impact of penalty cost



Numerical study – Impact of transshipment cost





Numerical study – Impact of transshipment cost



Numerical study – Sensitivity analyses



 \rightarrow

Synchromodality > Multimodality

- Cost
- Service quality
 regardless of the parameter value

- Relative cost reduction \uparrow :
 - Penalty ↑
 - Transshipment cost ↓

Numerical study – Impact of carbon tax





Numerical study – Impact of carbon tax



Numerical study – Impact of carbon tax









Synchromodal

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