

Integration of the Load Matching and Routing Problem with Equipment Balancing for Small Package Carriers...

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Michigan**Engineering**

Industrial and Operations Engineering

...or,
Struggling to Apply OR
in the Real World

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Michigan**Engineering**
Industrial and Operations Engineering

Industry, OR, and Academia

- My background
- What do trucking, passenger aviation, and automotive manufacturing all have in common?
- Challenging scheduling problems
 - ◆ Complex networks
 - ◆ Tightly constrained resources
 - ◆ Large numbers of interconnected decisions

Industry, OR, and Academia, cont.

- What do trucking, passenger aviation, and automotive manufacturing all have in common?
- Challenges in applying academic OR solutions
 - ◆ Emphasis on optimality
 - ◆ Simplifying for tractability
 - ◆ Stochastic and dynamic environments
 - ◆ Desire for solution regularity

Industry, OR, and Academia, cont.

- Academic goals:
 - ◆ Clean, elegant, provably optimal
- Industry desires:
 - ◆ High quality solutions
 - ◆ Timely results
 - ◆ Regularity
 - ◆ Robustness
 - ◆ Real-world detail

Large-Scale Network Problems in Transportation and Logistics

Challenges of Complex Systems

- Large numbers of decisions
- Tightly inter-coupled because of network structures
- Very constrained resources (time, money, raw materials, labor...)
- Timeliness

Math Programming Approach

■ Pros:

- ◆ Global viewpoint
- ◆ General solution techniques
- ◆ Established commercial solvers

■ Cons:

- ◆ Large numbers of constraints, variables
- ◆ Non-linearities
- ◆ Weak LP relaxations
- ◆ No exploitation of problem-specific structures

Customized Algorithms

■ Pros:

- ◆ Exploit problem structure
- ◆ More realistic
- ◆ Fast run times, convergence

■ Cons:

- ◆ Feasibility?
- ◆ Solution quality?
- ◆ Adaptability?

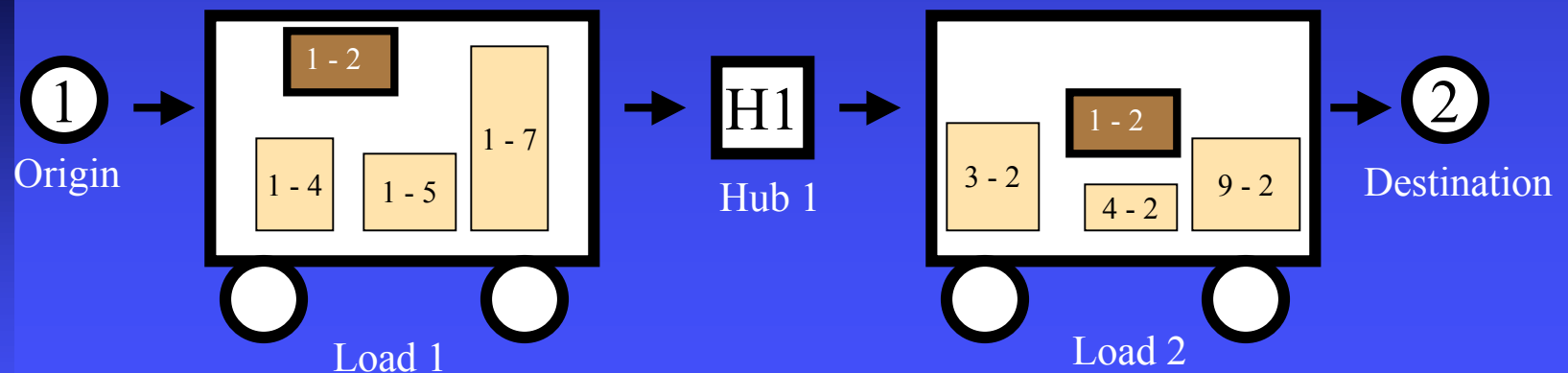
Our Hybrid Approach

- Seek to combine the strengths of both approaches
- MP-based *master problem*
- *Complex variable definitions*
 - ◆ Multiple decisions in a single variable
 - ◆ Capture problem complexity within variable definition
 - ◆ Eg. crew pairing
- Application-specific algorithms to generate columns
 - ◆ Wealth of resources for generating candidate columns
 - ◆ Leveraging user expertise
 - ◆ Trading solution time vs quality

Integrated Planning for Small Package Carriers

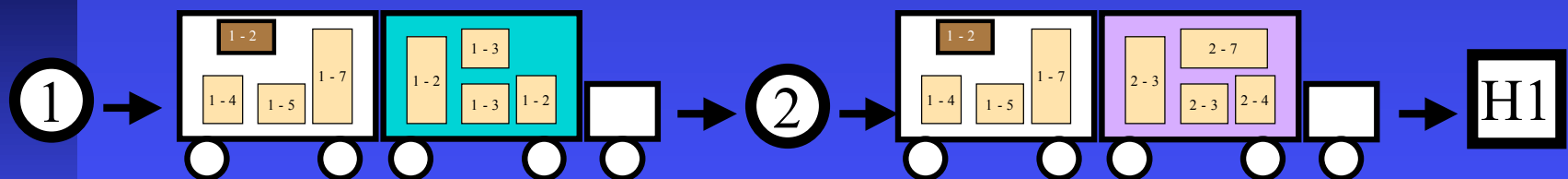
Express Package Networks

- Packages move from origin to destination via one or more hubs at which freight is unloaded, sorted, and reloaded
- Each trailer movement is called a *load*



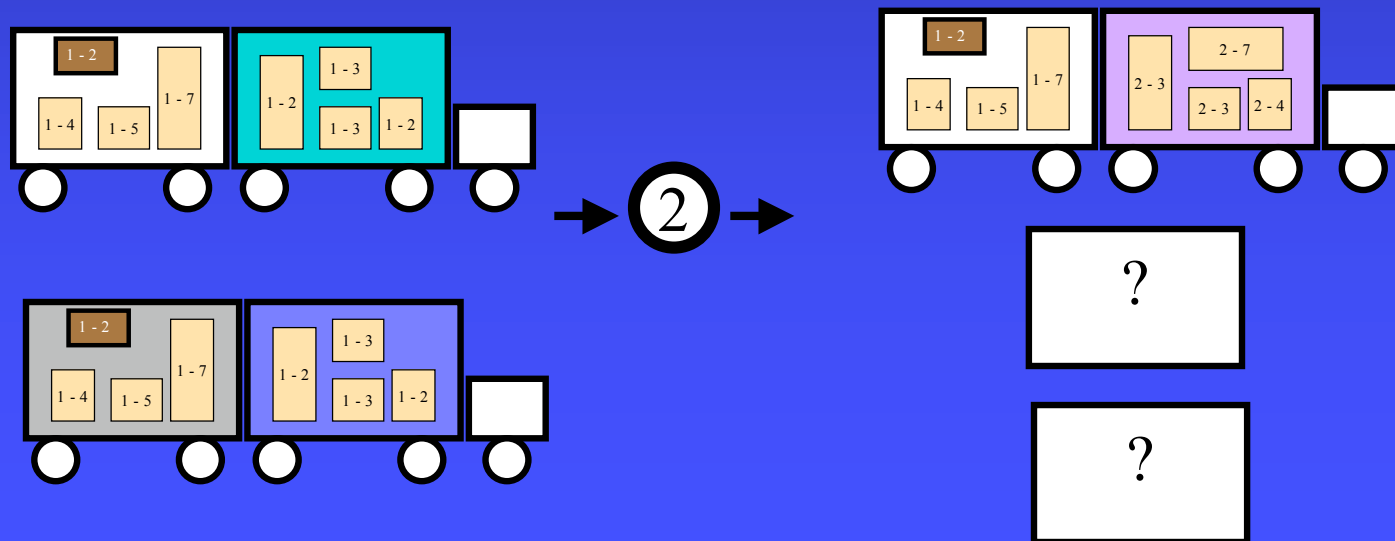
Load Matching

- Loads (trailers) are often coupled with other trailers for reduced transportation cost
 - ◆ The cost of pulling two trailers behind the same tractor is less than twice the cost of pulling a single trailer
- Loads may stop at intermediate points along the way to re-configure the tractor



Equipment Balancing

- Loaded trailer movements can result in system imbalances – some locations have more outbound loads than inbound, others more inbound than outbound
- We need to move empty trailers throughout the system to regain balance



Integrated Load Matching and Equipment Balancing Problem

- What loads should be matched?
 - ◆ When is it worth driving extra miles to combine two trailers?
- How should empty trailers be redistributed throughout the system?
 - ◆ At each location, total trailer flow out must equal total flow in
- Can we make these two decisions together for even greater savings?
 - ◆ Loaded and empty trailers can be paired together

Problem Description: Inputs

- Set of loads (each load has an origin, destination, trailer type, earliest available time, and latest delivery time)
- Permissible tractor configurations between each pair of locations
- Cost per mile of each configuration
- Mileage and travel time between each pair of locations

Problem Description: Constraints

- Loads must move contiguously in space and time from origin to destination
- Loads must satisfy time windows
- Each location must be balanced (i.e. total number of trailers in must equal total number out, including both loaded and empty trailers)

Problem Description: Objective

Minimize total transportation cost,
incorporating cost function for different
modes and configurations

MCF Model with Side Constraints

- This can be thought of as a multi-commodity flow problem with side constraints
- Each load is a commodity that must be moved from origin to destination
- Side constraints are needed to ensure timing restrictions, linearize the objective function, and balance the network

Multi-commodity flow based model

- Use time-space network to capture time constraints
- Decision variables
 - ◆ x_{ijk} – flow of commodity k on arc (i, j)
 - ◆ y_{ij} – flow of empty trailers on arc (i, j)
 - ◆ s_{ij} – flow of single trailer combinations on arc (i, j)
 - ◆ d_{ij} – flow of double trailer combinations on arc (i, j)

Multi-commodity flow based model

$$\min \sum_{(i,j) \in A} c_{ij}^s s_{ij} + \sum_{(i,j) \in A} c_{ij}^d d_{ij}$$

$$\text{s.t.} \quad \sum_{i:(j,i) \in A} x_{jik} - \sum_{i:(i,j) \in A} x_{ijk} = b_{jk} \quad \forall j \text{ in } V, k \text{ in } K$$

$$s_{ij} + 2d_{ij} = \sum_{k \in K} x_{ijk} + y_{ij} \quad \forall (i,j) \text{ in } A$$

$$\sum_{j \in V_f} \left(\sum_{k \in K} b_{jk} + \sum_{i:(j,i) \in A} y_{ji} - \sum_{i:(i,j) \in A} y_{ji} \right) = 0 \quad \forall f \text{ in } F$$

$$x_{ijk}, y_{ij}, s_{ij}, d_{ij} \text{ in } \mathbb{Z}^+$$

Challenges of MCF

- Very large number of nodes, constraints
 - ◆ Time/space network
 - ◆ If non-facility meet points are permitted, grows further
- Weakness of LP relaxation – half a double is always better than a whole single

Alternative Model: Load Clusters

- As an alternative, we define the notion of a *load cluster* – a set of interconnected loads, the path each load takes, and the tractor configurations used to combine these loads
 - ◆ All loads must move fully from origin to destination
 - ◆ All loads must satisfy their time windows
 - ◆ No outside loads can be partially included in the configurations

Example Cluster

A load from A to B moves directly, as a single trailer:



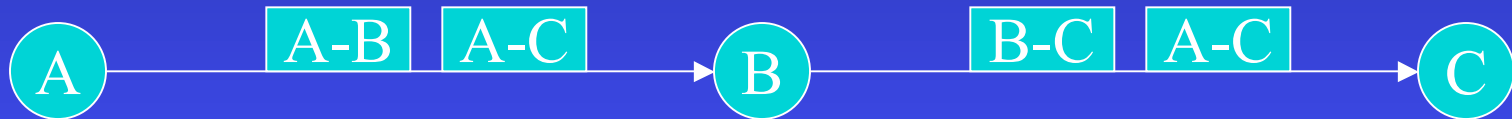
Example Cluster

A load from A to B moves directly, in combination with an empty trailer:



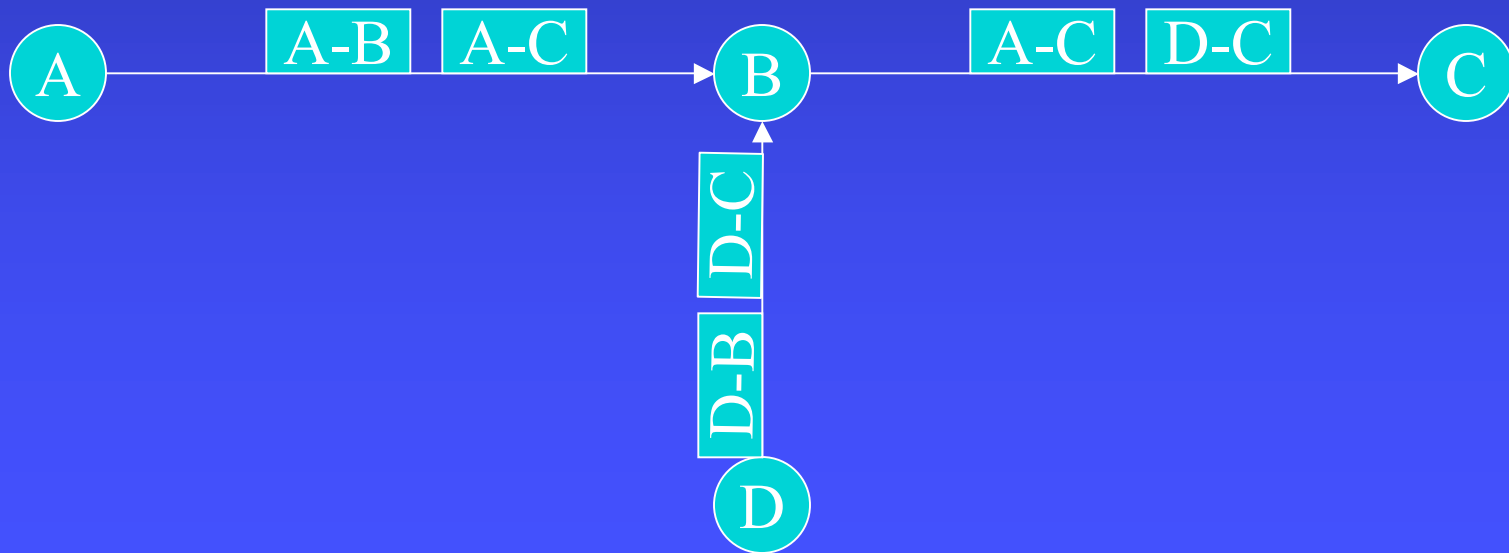
Example Cluster

Loads from A to B and A to C move via B, where the A to B load is dropped off and a load from B to C is picked up:



Example Cluster

Loads from A to B and A to C move via B,
as do loads from D to B and D to C:



The Cluster-Based Model

- L is the set of loads
- C is the set of load clusters
- c_i is the cost of load cluster i
- F is the set of facilities
- δ_{il} is defined to be 1 if cluster i contains load l and 0 otherwise
- λ_{if} is the net surplus of trailers at facility f relative to cluster i
- x_i is a binary variable that takes value 1 if cluster i is chosen and 0 otherwise

The Cluster-Based Model

$$\text{Min.} \quad \sum_{i \in C} c_i x_i$$

$$\text{St} \quad \sum_{i \in C} \delta_{il} x_i = 1 \quad \forall l \in L$$

$$\sum_{i \in C} \lambda_{if} x_i = 0 \quad \forall f \in F$$

$$x_i \in \{0, 1\} \quad \forall i \in C$$

Benefits

- The objective function is linear
- The LP relaxation is strengthened
- Timing issues are imbedded within the variable definition
- The number of constraints is greatly reduced
- Expanded solution space – the set of meet locations does not have to be limited to carrier facilities
- Multiple transportation modes, trailer configurations can easily be incorporated

Original Solution Plan

- Too many variables to explicitly enumerate in the model
- Use *branch-and-price*
 - ◆ Branch-and-bound to solve the IP
 - ◆ *Column generation* to solve the LP nodes
- What is the pricing problem???

Computational Experience

- Started off with “initial” set of cluster *templates* to find seed duals
- Found significant improvement in solution quality just from using the templates!
 - ◆ Approximately 10% on a small network, with expectations of better savings from larger networks
- Feasibility
- Regularity
 - ◆ Simplifies cluster-based model
 - ◆ Makes MCF much more complex

Future Research

- Symmetry
- Time window analysis
- Assigning volume to trailer types

Conclusions, Challenges, and Future Research

Conclusions

- Many applied problems are hard!
- Simplification can lead to tractability but real-world infeasibility
- Optimality is not all it's cracked up to be...
- ...but the quality of our solution really matters!
- Sometimes, a hybrid approach with a complex variable model can help

Challenges

- How to generate candidate columns
- Feasibility vs. optimality
- Lower bounds
- Degeneracy, convergence

Future Research

- Expanding problem scope
- Generalizing results
- Robustness