





Applications of Multi-Agent Systems in Logistics

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Applications of Multi-Agent Systems in Logistics

Overview



- **Transportation Domain**
- □ Agent-Based Optimization
 - Agent-Design
 - Allocation
 - Reallocation
 - User-Interaction
- Optimization Results
- \Box Conclusion

Freight Logistics



- □ Task: Create a plan to transport all orders
- □ Challenges
 - Complexity
 - Mergers
 - Higher volumes
 - Constraints
 - Dynamics
 - Ever shorter notice in advance
 - Unforeseen events
 - User interaction



Transportation Domain

Freight Logistics



- □ Today's Solution
 - Division into (regional) businesses
- □ Problems
 - Visibility
 - Communication
 - Missed synergetic effects



Spot-Market



- □ Trucks are ordered on the Spot-Market just in time
 - Assumption: truck is always available everywhere
 - We do not have to care for successor routes
- □ Dispatchers have to find a truck for each route (by phone)
- □ System has to suggest the routes
- □ Objective
 - Minimum cost
- □ Pickup and delivery has to obey
 - Load constraints
 - Time constraints

Transportation Domain

Cost Calculation



□ Spot-Market

- Variable costs depending on
 - Distance
 - Load
 - Relation (Start Destination)
 - Discount
- Minimum costs



Load Constraints



EXPLOSIVE

OXIDIZER

FLAMMABLE

RADIOACTIVEII

FLAMMAB

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- □ Precedence (pickup before delivery)
- □ Pairing (pickup and delivery by the same truck)
- □ LIFO (last in first out loading of orders)
- □ Capacity limitation (dependent on truck type)
- Weight limitation (dependent on truck type)
- □ Order truck compatibility (Type, Equipment)
- □ Order order compatibility (Dangerous goods)

Time Constraints



- Order dependent load and unload durations
- Earliest and latest pickup (Soft constraints)
- □ Earliest and latest delivery (Soft constraints)
- Opening hours for pickup and delivery
- Legal drive time restrictions
- Lead time for ordering spot market trucks
- Maximum Tour duration



Traditional Optimization (top-down)



- Batch optimizers using Linear Programming
- □ Complexity
 - Not able to deal with number of orders and trucks of big logistics companies
 - Harder to model and change constraints
- Dynamics
 - Need to know all orders in advance
 - Batch optimization, no real-time capabilities, no user interaction
- □ Advantages
 - Closer to optimum (for those domains they are able to solve and if time allows)

Agent Design



Each truck is represented by an autonomous agent

- Maximal scalability
- Overhead (messaging, memory, other resources)



Agent Design



- □ Regional cluster of trucks represented by an agent
 - Agents use almost same solution generation and optimization
- □ Solution generation
 - Start in the most promising cluster
- Optimization
 - Exchange request for tours running across multiple clusters



Allocation

Contract Net

- Coordinator agent asks each truck agent for a quote
- Each truck able to transport the order sends back a bid representing the cost increase when transporting the order
- Coordinator awards order to lowest bidding agent
- □ Example
 - Order1 is assigned to truck1, it is closer to the pickup location





Allocation

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□ Example

- Order2 is also assigned to truck1
- Additional distance for truck1 is less than additional distance for truck2
- truck1 is now fully loaded when order2 and order1 are picked up



Allocation

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Example

- Order3 is assigned to truck2
- truck1 does not bid, because
 - it is overloaded when picking it up together with order1 and order2
 - it is late at pickup location of order2 when picking it up before picking up order1 and order2
 - it is late at pickup location of order3 when picking it up after delivering order1 and order2



Allocation

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□ Example

- Order4 is also assigned to truck2
- Optimality
 - Contract net protocol produces sub-optimal results
 - Quality depends on the order in which transport orders arrive



Reallocation

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Mutual negotiation Lieferung alle Aufträge Agent with changed route sends request for checking order exchanges to other agents Abholung Abholung Other agent checks potential and Auftrag 1 . Auftrag 4 sends back request for order exchange if exchange creates a better solution Abholung Agent accepts if exchange creates Abholung Auftrag 3 Auftrag 2 a better solution Better solution win-win for inter-company , <u>- - -</u> - - - 0 LKW 1 Start LKW 2 Start systems (self-interested agents)

 win-loose for intra-company systems if win > loose

Reallocation



Example

- agent2 sends agent1 request for checking order exchanges
- agent1 sends back request for exchanging order2 and order3 to save costs
- agent2 accepts exchange, because it also saves costs
- □ b-cyclic k-transfer
 - b: number of involved trucks
 - k: number of involved orders
- Distributed Hill-climbing



User Interaction



- □ Agents and Users have to work together
- □ Performance
 - Users expect feedback within seconds (real-time optimization)
- Control
 - Full control to agents
 - Not feasible (can not know all constraints, spot market, ..)
 - Full control to dispatchers (all the time)
 - Slows down the system
 - Solution: different phases with handover of control

User Interaction



Planning phase

- Until n hours before route starts
- Trucks under full control of the agents
- Allocation and Reallocation without restrictions
- Dispatcher observes



User Interaction



Todo phase

- Trucks under control of agents, but dispatcher can take over
- Find alternatives to transport an order
- Add (1), Remove (2) or exchange (3) orders
- Change schedule (4)
- Agents may not change manually edited routes
- Dispatchers may violate some constraints



User Interaction



□ Tracking phase

- Trucks under full control of dispatcher
- Agents can make suggestions to add orders
- Telematics system keeps schedule updated
- Reactive alerting
- Proactive alerting

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Optimization Results

Overview



- □ Case study with 3500 real orders
- □ Objectives
 - Reduce cost
 - Reduce distance

□ Measures to ensure comparability with manual solution

- Same geo-coding
- Same cost parameterization
- No soft time constraints
- Result checked by dispatchers

Solution Quality



Comparison to manual dispatching plan

Objective	Reduce Cost	Reduce Distance
Cost	11.7%	not comparable
Kilometer	4.2%	13.4%
Vehicles	25.5%	40.4%

□ Cost savings

- 4.2% by reducing driven kilometers
- 2.2% by increasing cost-saving tramp routes (reuse of trucks)
- 5.3% by hiring trucks in cheaper regions

Solution Quality



- □ w/o ATN: Dispatchers (manual) solution
- □ ATN 1: Significant reduction of driven kilometers at same service level
- □ ATN 2: Reduced kilometers with significant higher service level
- □ ATN 3: Minimal additional kilometers while meeting all restrictions



Soft Time Windows



Optimization results with increased soft time windows



Optimization Results

k-transfers



 Optimization results with increasing k (number of orders exchanged)



Optimization Results

Clustering



- □ Cluster sizes can be varied
- □ Compared 4 clusters against 1 on a 4 CPU machine

	4 Clusters compared to 1	
Cost	+1.2%	
Runtime	11%	

□ Cost

- Higher costs due to distribution of trucks into clusters

Runtime

- Factor 3 by having computation distributed to 4 CPUs
- Factor 3 by having less trucks in one cluster

Tracking Feedback



- □ It is common to have feedback at pickup/delivery locations
- Truck positions are not known in between
- Knowing the position more precisely has advantages
 - Can suggest to add very short time notice orders
 - May detect late trucks earlier and replan
 - May make use of truck ahead of schedule

	Real time tracking
Cost	-3%



Software Agents



- □ Help our customers to reduce their costs
- □ Support the dispatchers in working in a
 - Complex,
 - Distributed and
 - Dynamic domain
- □ Help us to sell ATN
- □ Accounted for roughly 10% of the effort to create the system (assuming having an agent runtime available)



Thank you for your attention!

Heterogeneous Vehicles (FTL)



- Decision for which truck to choose matters
- □ Example
 - Order5 (24 tons) can not be transported by truck 17t
 - Routes in a have too many empty km
 - Simple Reallocation does not work



Network Optimization

- Transport orders between network of hubs
- □ Orders may be relayed at hubs
- Create a transport plan for all relations between hubs





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Network Optimization

- Distance optimal solution
 - 352 km
 - 3 trucks
- How does a cost optimal solution look like?
 - Distance costs
 - Fix costs for trucks
 - Costs for drivers not at home



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Network Optimization



□ Setup

- 175 orders
- 33 hubs
- 16 paths on average per relation for each order
- 61233 order kilometers

Approach	Result (avg) (km)	Result (best) (km)	Runtime (avg) (sec)	Result (best known)
Hill Climbing	33288	33288	9	30657
Tabu Search	32647	32647	60	32051
Simulated Annealing	28502	28325	65	28325
Genetic Algorithm	30684	30114	61	28180