

# Efficient Delivery with Mobile Agents



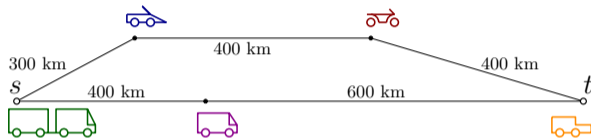
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CNLS Postdoc Seminar  
April 18, 2019

# Model of delivery

## Setting

- undirected graph  $G = (V, E)$   
with edges  $e \in E$  having lengths  $\ell_e$
- $m$  packages: source  $s_i$  and target  $t_i$
- $k$  agents, each starting at node  $p_i$ ,  
*budget*  $\beta_i$ , *weight*  $\omega_i$  & *velocity*  $v_i$ ,  
able to transport 1 package at a time.



## Assumptions

- agents cooperate by  
global, centralized coordination
- handovers possible at nodes  $V$   
as well as inside edges

**Task:** Find an *efficient delivery* in terms of

*energy* consumption  $\mathcal{E} \rightsquigarrow$  terms of form  $\omega_i \cdot \ell_e$

delivery *time*  $\mathcal{T} \rightsquigarrow$  terms of form  $\frac{1}{v_i} \cdot \ell_e$

constrained *resources*  $\beta_i \rightsquigarrow$  range of agents

# Outline

## 1 Examples and Results

- Resource-efficiency: Budgets only
- Energy-efficiency: Weights only
- Time-efficiency: Velocities only

## 2 Dynamic programming

- Technique
- Matrix chain multiplication

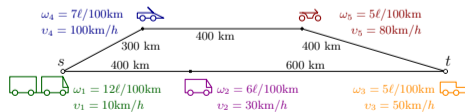
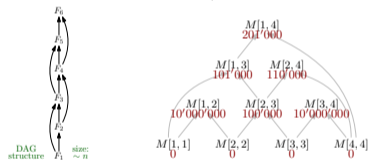
## 3 A detailed discussion

- Combining energy- and time-efficiency
- OPT characterization
- Polynomial-time algorithm

Resource-efficiency

Energy-efficiency

Time-efficiency

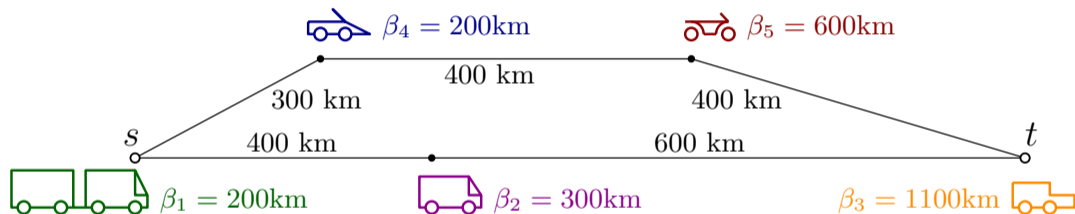


# Examples and Results

## Agents with budgets

The agents' resources constrain how far each agent can go (*budgets*  $\beta_i$ ).

Can we *decide* whether there is a package delivery which respects all battery constraints?



not on shortest path

in-edge-handovers

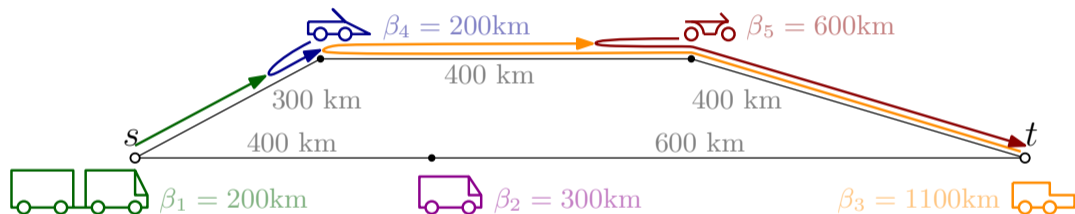
no clear characterization

[1] B., Chalopin, Das, Dissler, Geissmann, Graf, Labourel, Mihalák: Collaborative delivery with energy-constrained mobile robots.

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## Resource-efficiency

NP-hard, even for a single package ( $m = 1$ ) on simple graphs.

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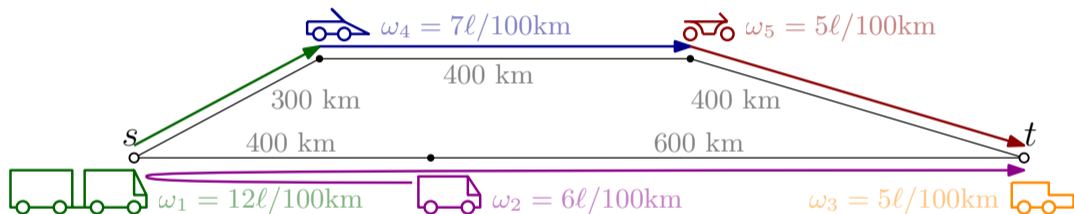
## Energy-efficiency

## Time-efficiency

# Agents with weights

Each agent has its individual energy consumption (*weight*)  $\omega_i$ .

Can we *optimize* the total *energy consumption*  $\mathcal{E}$  needed to deliver the package?



not on shortest path

vertex handovers

decreasing weights

[2] B., Chalopin, Das, Disse, Graf, Hackfeld, Penna: Energy-Efficient Delivery by Heterogeneous Mobile Agents.

[3] B., Graf, Penna: Truthful Mechanisms for Delivery with Agents.



## Resource-efficiency

NP-hard, even for a single package ( $m = 1$ ) on simple graphs.

**Energy-efficiency** Agents face 3 major challenges:

*Collaboration:* How to work together on a package?

– 2-approximation without collaboration.

*Planning:* Which route to take?

– NP-hard, polynomial-time 2-approximation.

*Coordination:* How to assign agents to packages?

– NP-hard, polynomial-time  $\max \frac{\omega_i}{\omega_j}$ -approximation.

⇒ Polynomial-time  $4 \max \frac{\omega_i}{\omega_j}$ -approximation.

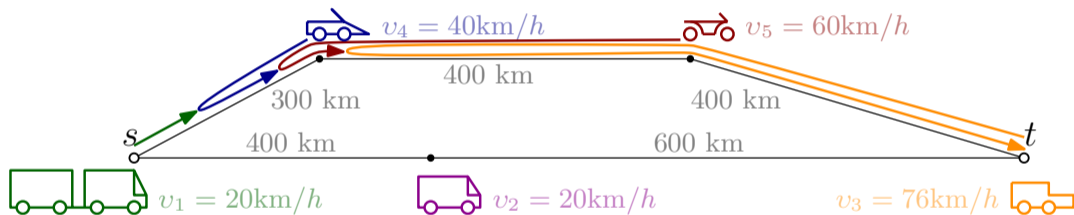
⇒ Polynomial-time algorithm for one package.

## Time-efficiency

# Agents with velocities

Each agent has its individual *velocity*  $v_i$ .

Can we *optimize* the total *time*  $\mathcal{T}$  needed to deliver the package?



not on shortest path

(multiple) in-edge-handovers

increasing velocities

[4] B., Graf, Mihalák: Collective fast delivery by energy-efficient agents.

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*Collaboration*: How to work together on a package?

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⇒ Polynomial-time algorithm for **one package**.

**Time-efficiency**

Existence of optima is unclear, maybe only infima exist.

→ NP-hard.

Poly-time algo. for **one package**.

**Can we combine energy- and time-efficiency for  $m = 1$ ?**

## Combining energy- and time-efficiency

Each agent has its individual *weight*  $\omega_i$  and *velocity*  $v_i$ .

We want a delivery that is both efficient in its energy consumption *and* its delivery time:

- Fastest delivery among all energy-optimum ones.  
Task: lexicographically minimize  $(\mathcal{E}, \mathcal{T})$ . ⇒ **Part 3.** [5]
- Energy-optimum delivery among all fastest ones.  
Task: lexicographically minimize  $(\mathcal{T}, \mathcal{E})$ . } ⇒ NP-hard. [4]
- Tradeoff between the two measures.  
Task: minimize  $\varepsilon \cdot \mathcal{T} + (1 - \varepsilon) \cdot \mathcal{E}$ ,  $\varepsilon \in (0, 1)$ .

[4] B., Graf, Mihalák: Collective fast delivery by energy-efficient agents.

[5] B., Tschager: Energy-Efficient Fast Delivery by Mobile Agents.

# Dynamic programming

# Dynamic programming: Technique

Programming technique which can be used if an optimal solution of a problem can be found by combining optimal solutions of subproblems: **optimal substructure**.

## Applications:

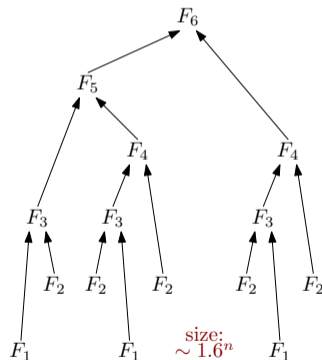
- Graph: Shortest Paths & Dijkstra & Floyd-Warshall
- Bioinformatics: De Novo Peptide Sequencing
- Economics: Optimal Saving
- Control Theory
- Matrix chain multiplication

## Toy Example:

### Fibonacci Sequence

- $F_n = F_{n-1} + F_{n-2}$
- $F_2 = 1$
- $F_1 = 1$

*Task:* Compute  $n$ -th Fibonacci number.



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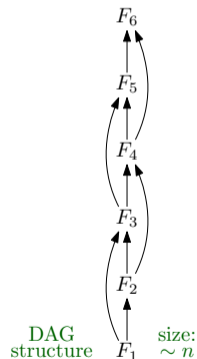
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## Toy Example:

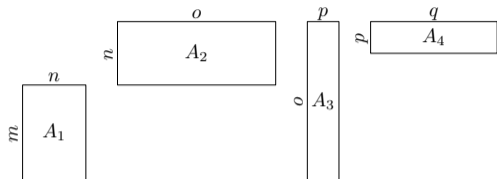
### Fibonacci Sequence

- $F_n = F_{n-1} + F_{n-2}$
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# Dynamic programming: Matrix chain multiplication



$m = 100, n = 10, o = 10'000, p = 1, q = 1'000.$

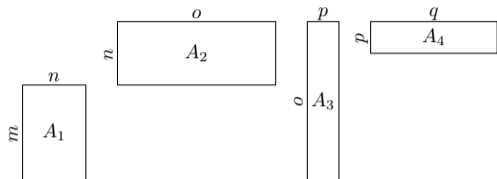
$$\underbrace{\underbrace{(A_1 \times A_2)}_{m \cdot n \cdot o = 10^7} \times \underbrace{(A_3 \times A_4)}_{o \cdot p \cdot q = 10^7}}_{m \cdot o \cdot q = 10^9} \quad \blacksquare \quad (A_1 \times A_2) \times (A_3 \times A_4) \Rightarrow 1'020'000'000$$

## Find best multiplication order by:

- Testing all ways to insert parentheses:  $\sim 4^{(\#\text{Matrices})}$  many!



# Dynamic programming: Matrix chain multiplication



$m = 100, n = 10, o = 10'000, p = 1, q = 1'000.$

$$\underbrace{\underbrace{\underbrace{((A_1 \times A_2) \times A_3) \times A_4}_{m \cdot n \cdot o = 10^7}}_{m \cdot o \cdot p = 10^6}}_{m \cdot p \cdot q = 10^5}$$

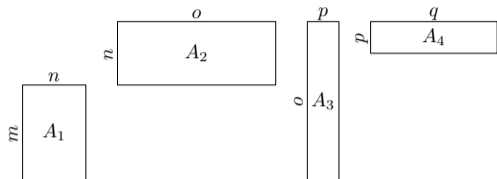
- $((A_1 \times A_2) \times A_3) \times A_4$   
 $\Rightarrow 11'100'000$

- $(A_1 \times A_2) \times (A_3 \times A_4)$   
 $\Rightarrow 1'020'000'000$

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# Dynamic programming: Matrix chain multiplication



$m = 100, n = 10, o = 10'000, p = 1, q = 1'000.$   $A_1 \times A_2 \times A_3 \times A_4 \Rightarrow 1'020'000'000$

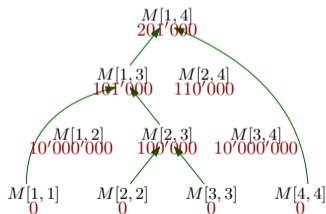
- $(A_1 \times (A_2 \times A_3)) \times A_4 \Rightarrow 201'000$
- $((A_1 \times A_2) \times A_3) \times A_4 \Rightarrow 11'100'000$
- $(A_1 \times A_2) \times (A_3 \times A_4) \Rightarrow 1'020'000'000$

## Find best multiplication order by:

- Testing all ways to insert parentheses:  $\sim 4^{(\#\text{Matrices})}$  many!
- **Dynamic Programming**:  $\sim (\#\text{Matrices})^2$  subproblems:

$M[x, y]$  = Cost of best parentheses for  $A_x \times \dots \times A_y$

$$= \min_{x \leq i < y} \left\{ \begin{array}{l} M[x, i] + M[i+1, y] \\ + \text{cost of multiplying two subproblems} \end{array} \right\}$$

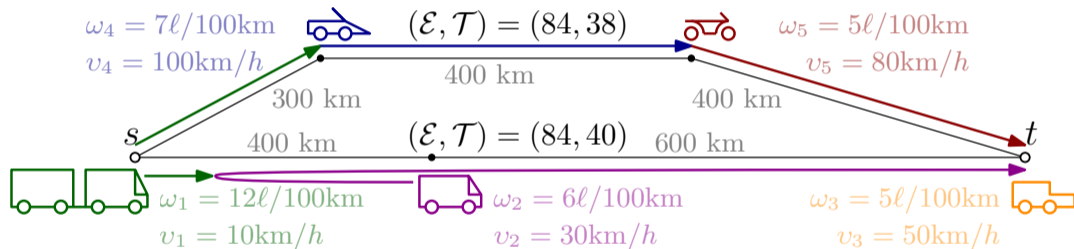


## A detailed discussion

# Agents with weights and velocities

Each agent has its individual *weight*  $\omega_i$  and *velocity*  $v_i$ .

Among all energy-optimum deliveries, can we find the fastest?



not on shortest path

0 or 1 in-edge-handovers

decreasing tuples  $(\omega_i, v_i^{-1})$

# Characterization of an optimum delivery

## Theorem (OPT characterization)

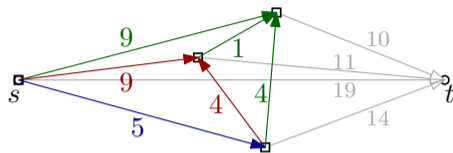
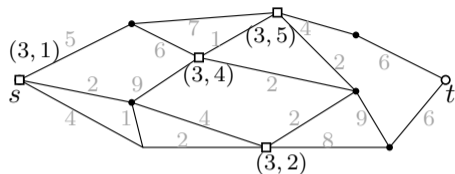
There is an optimum delivery, with the involved agents denoted by  $1, \dots, i, i+1, \dots, k$ , in which the following holds for each consecutive pair of agents:

- Decreasing weights:  $\omega_i \geq \omega_{i+1}$ .
- If  $\omega_i = \omega_{i+1}$ , then  $v_i < v_{i+1}$ .
- If  $\omega_i = \omega_{i+1}$ , then agent  $i+1$  does not move without the package.

$$\underbrace{\omega_1 = \omega_2 = \dots}_{W_1} >_x \underbrace{v_i < v_{i+1} < \dots \quad \omega_i = \omega_{i+1} = \dots}_{W_2} >_y \underbrace{\dots = \omega_k}_{W_3}$$

⇒ First look at the problem for each weight class  $W_j$  separately.

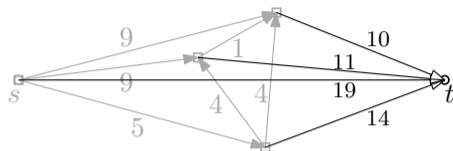
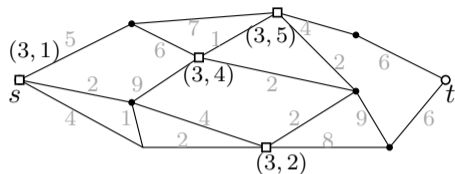
## Uniform energy consumption weights



Example: 4 agents, weight 3, velocities  $v_1 = 1, v_2 = 2, v_3 = 4, v_4 = 5$ . Approach:

- 1 Move closest agent to source  $s$ . Costs  $(\mathcal{E}, \mathcal{T})[p_1] = (3 \cdot 3, 3/1) = (9, 3)$ .
- 2 Order agents by increasing velocity. Transform graph to DAG. Compute  $(\mathcal{E}, \mathcal{T})[p_i] =$  the energy consumption  $\mathcal{E}[p_i]$  and delivery time  $\mathcal{T}[p_i]$  of an optimum delivery of the package from  $s$  to  $p_i$ , using only agents  $1, \dots, i - 1$ .

## Uniform energy consumption weights



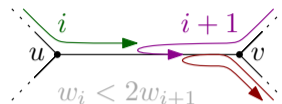
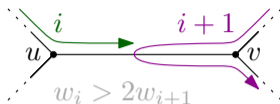
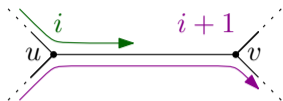
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- 2 Order agents by increasing velocity. Transform graph to DAG. Compute
 
$$(\mathcal{E}, \mathcal{T})[p_2] = (\mathcal{E}, \mathcal{T})[p_1] + (5 \cdot 3, 5/1) = (24, 8)$$

$$(\mathcal{E}, \mathcal{T})[p_3] = \min\{(\mathcal{E}, \mathcal{T})[p_1] + (9 \cdot 3, 9/1), (\mathcal{E}, \mathcal{T})[p_2] + (4 \cdot 3, 4/2)\} = (36, 10)$$

$$(\mathcal{E}, \mathcal{T})[p_4] = \min\{\dots, (\mathcal{E}, \mathcal{T})[p_2] + (4 \cdot 3, 4/2), \dots\} = (36, 10)$$
- 3 Compute optimum *delivery time* among energy-optimum deliveries:  $(\mathcal{E}, \mathcal{T}) = (66, 12)$ .

## Vertex handovers



If  $w_i \neq 2w_{i+1}$ , then agent  $i$  does not handover the package to  $i+1$  inside an edge.

Assume we have  $w_i \neq 2w_j \forall i, j$ .

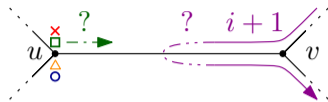
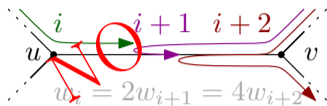
$\Rightarrow$  We can use the precomputed weight class solutions. Define the subproblems

$(\mathcal{E}, \mathcal{T})[j, y]$  = the energy consumption  $\mathcal{E}[j, y]$  and delivery time  $\mathcal{T}[j, y]$  of an optimum delivery of the package from  $s$  up to node  $y$ , using only agents from the first  $j$  weight classes  $W_1, \dots, W_j$ .

$\Rightarrow$  We can compute  $(\mathcal{E}, \mathcal{T})[j, y]$  from all smaller subproblems  $(\mathcal{E}, \mathcal{T})[j-1, x]!$



# In-edge handovers



We can have at most one in-edge-handover per edge.

But: **Which agents** are involved in an in-edge-handover?

Pareto frontier!

Adapt previous methods as follows:

- 1 Incorporate the Pareto frontier into the weight class computations.
- 2 Incorporate in-edge-handovers into the main dynamic program.

## Running time:

- Preprocessing  $poly(\text{APSP} + k + |V|)$
  - Per weight class:  $poly(|V| \cdot k^2 + |V|^2 \cdot k)$
  - Main dynamic program:  $poly(k \cdot |V| \cdot |V|)$
- }
- $poly(k + |V|^3)$

**Resource-efficiency,  $m = 1$**  NP-hard on simple graphs.

### Energy-efficiency

*Collaboration:*

2-apx without collaboration

*Planning:*

NP-hard, poly-time 2-apx

*Coordination:*

NP-hard, poly-time  $\max \frac{\omega_i}{\omega_j}$ -apx

Poly-time algorithm for  $m = 1$

### Time-efficiency

Existence of optima is unclear (maybe only infima)

NP-hard

Poly-time algo. for  $m = 1$

### Combining energy- and time-efficiency, $m = 1$

Some versions NP-hard, but lexicographically minimizing  $(\mathcal{E}, \mathcal{T})$  can be done in polynomial time.

#### UPCOMING AND RECENTLY-ACHIEVED SELF-DRIVING CAR MILESTONES

- AUTOMATIC EMERGENCY BRAKING
- HIGHWAY LANE-KEEPING
- SELF-PARKING
- FULL HIGHWAY AUTONOMY
- FIRST SEX IN A SELF-DRIVING CAR
- FULL TRIPS WITH NO INPUT FROM DRIVER
- FULL TRIPS BY EMPTY CARS
- SELF-REFUELING OF EMPTY CARS
- AN EMPTY CAR WANDERING THE HIGHWAYS FOR MONTHS OR YEARS UNTIL SOMEONE NOTICES THE CREDIT CARD FUEL CHARGES
- CARS THAT READ OTHER CARS' BUMPER STICKERS BEFORE DECIDING WHETHER TO CUT THEM OFF
- AUTONOMOUS ENGINE REVVING AT RED LIGHTS
- SELF-LOATHING CARS
- AUTONOMOUS CANYON JUMPING
- CARS CAPABLE OF ARGUING ABOUT THE TROLLEY PROBLEM ON FACEBOOK

<https://xkcd.com/1925/>