

# CIVIL-557

## Decision-Aid Methodologies in Transportation

### Applications

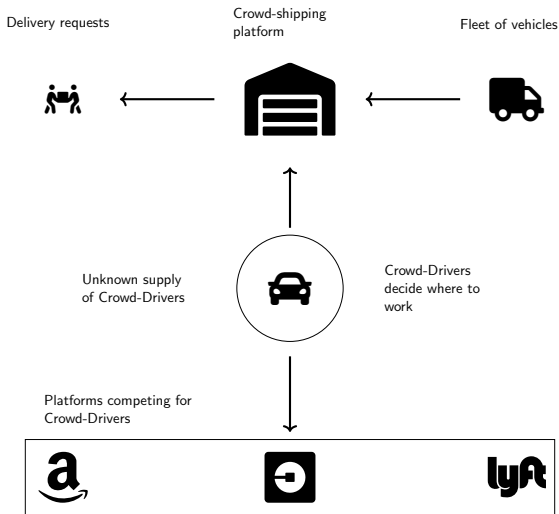
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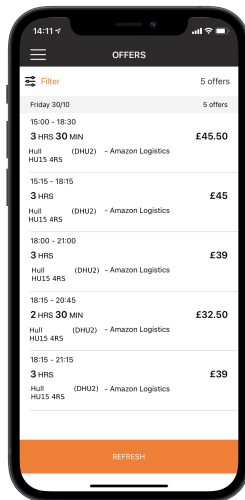
- 1 Crowd-shipping: Case study AmazonFlex
- 2 Tourist Itinerary planning: Cases study Unki
- 3 Elevator Dispatching: Case study Schindler

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# Crowd-shipping

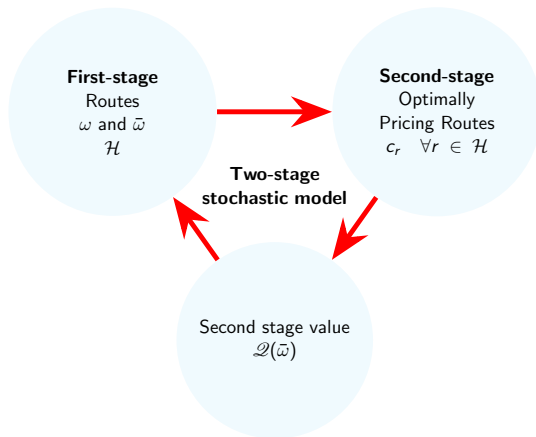


# AmazonFlex delivery



$$\begin{aligned} \min \quad & \sum_{r \in \Omega \cup \Omega'} c^r \omega^r + \mathcal{Q}(\bar{\omega}) \\ \text{s.t.} \quad & \sum_{r \in \Omega \cup \Omega'} a_i^r \omega^r = 1 & \forall i \in N \\ & \sum_{r \in \Omega'} \omega_r \leq M \\ & \omega_r \in \{0, 1\} & \forall r \in \Omega \cup \Omega' \end{aligned}$$

# Two-stage framework



# Demand model: logit model

We have a set of attributes of a route that influence the decision of Crowd-drivers. For example, the total distance, stops, load, and more importantly compensation (i.e.,  $c_r$ ).

$$u_r = \sum_{i \in \mathcal{A}} \beta_i x_i^r + \beta_c c_r + \varepsilon \quad \forall r \in \mathcal{H}$$

$$u_0 = 0$$



## Second stage problem: Notation

- Let  $\mathcal{H}$  be the set of routes created in the first stage for crowd drivers.
- $\aleph$  is the penalty if the route is rejected.
- $V_r$  is the deterministic part of the utility function, i.e., without the error term for route "r".
- In the logit model the probability of acceptance is equal to  $\frac{e^{V_r}}{e^{V_r} + 1}$

## Second stage problem

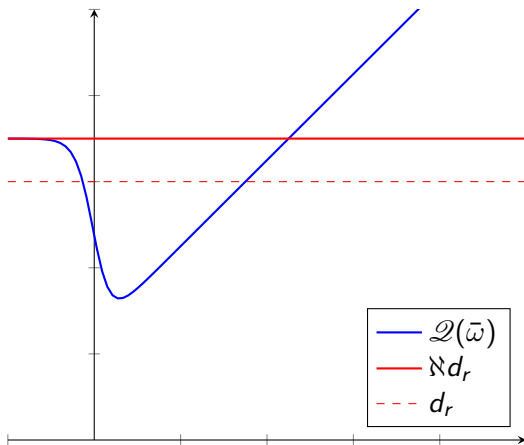
### Route Assignment and Pricing (RAP)

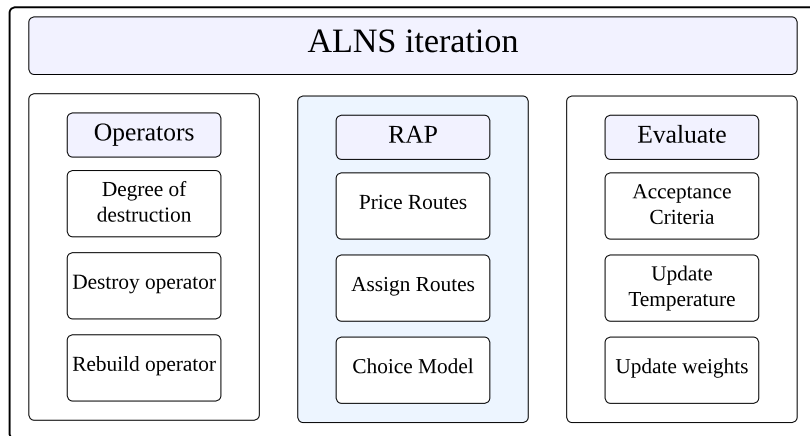
$$c_r = \frac{V_r - x_r}{\beta_c}$$

$$\mathcal{Q}(\bar{\omega}) = \min \sum_{r \in \mathcal{H}} \bar{\omega}_r \frac{e^{V_r}}{e^{V_r} + 1} \cdot \frac{V_r - x_r}{\beta_c} + \sum_{r \in \mathcal{H}} \bar{\omega}_r \left( 1 - \frac{e^{V_r}}{e^{V_r} + 1} \right) \cdot \mathbb{N} d_r$$

## Second stage problem

### Route Assignment and Pricing (RAP)





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```
Incumbent sol = Initial solution;
Sol = Incumbent sol;
Set k;
Set Temperature;
Set Cooling rate;
for  $it = 1$  to  $k$  do
     $Sol^* = Sol$ ;
    Select removal operator;
    RAP;
    Select insertion operator;
    RAP;
    if  $Sol^* \text{ value} < \text{Incumbent sol value}$  then
        | Incumbent sol =  $Sol^*$ ;
    end
    if  $Accept(Sol^*, Sol)$  then
        |  $Sol = Sol^*$ ;
    end
    Cool Temperature;
    if  $Temperature < \epsilon$  then
        | Stop;
    end
end
return Incumbent sol;
```

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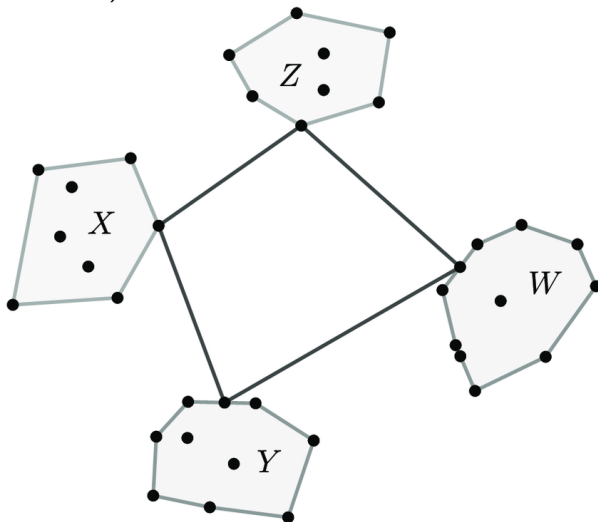
## Unki

Unki is an EPFL startup. The company wants to create itineraries for tourists that visit a new city.

- Based on the preferences of customers, a utility function is created for each possible activity.
- Tourist schedules are Hotel – Activity – Eat – Activity – Eat – Hotel.
- Realistic constraints that consider the cost and time to travel from one activity to another are necessary. For example, going from the hotel to an activity that is too far away takes a lot of time.

# Problem

The problem is an elementary shortest path problem with resource constraints (ESPPRC).





# Problem: Notation

- $U_j = \begin{cases} U_i + U_{a_j}^{participation} + U_{a_j}^{starttime} + U_{a_i}^{duration}, & \text{if } a_j \neq a_i \\ U_i, & \text{otherwise} \end{cases}$
- $t_j = \begin{cases} t_i + t_{(a_i, a_j)}, & \text{if } a_j \neq a_i \\ t_i + 1, & \text{otherwise} \end{cases}$
- $d_j = \begin{cases} 0, & \text{if } a_j \neq a_i \\ d_i + 1, & \text{otherwise} \end{cases}$
- $c_j = c_i + c_{(a_i, a_j)}$
- $b_j = b_i + b_{(a_i, a_j)}$
- $\mathcal{U}_j = \begin{cases} \mathcal{U}_i \cup a_i, & \text{if } a_j \neq a_i \\ \mathcal{U}_i, & \text{otherwise} \end{cases}$

# Dominance rules:

- ①  $a_1 = a_2$
- ②  $t_1 = t_2$
- ③  $d_1 = d_2$
- ④  $U_1 - c_1 > U_2 - c_2$
- ⑤  $b_1 \leq b_2$
- ⑥  $\mathcal{U}_1 \subseteq \mathcal{U}_2$

# Labeling algorithm

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**Algorithm 1:** Basic DP

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Initiate time,  $t \leftarrow 0$ ;

Initiate the list of labels,  $\text{Label}[1] = \{\mathcal{L}_0 = (\text{Hotel}, 0, 1, 1, 0, 0, \{\})\}$ ;

**while**  $t \neq \mathcal{T}$  **do**

$t \leftarrow t + 1$ ;

**for**  $\ell_1$  in  $\text{Label}[t]$  **do**

        Extend label to all feasible extensions;

$\mathcal{L}^* \leftarrow$  feasible extension;

$t^* \leftarrow \text{REF}(t)$ ;

**for**  $\ell_2$  in  $\text{Label}[t^*]$  **do**

            Apply dominance rules;

**if**  $\ell_2$  is dominated **then**

                delete  $\ell_2$ ;

                continue;

**if**  $\mathcal{L}^*$  is dominated **then**

                delete  $\mathcal{L}^*$ ;

                break;

$\text{Label}[t^*] \cup \mathcal{L}^*$

**return** Best label at time  $\mathcal{T}$

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# Elevator Dispatching Problem with Destination Control

Elevators are vertical transportation systems.

- Historically, elevators had a conductor inside that would control the elevator as passengers got on and off at each floor.
- In most modern elevators, the controls remain inside the elevator. Passengers enter the elevator and then choose the destination floor.
- In Destination Control, passengers choose the destination floor outside the elevator when making the call.



**Schindler**

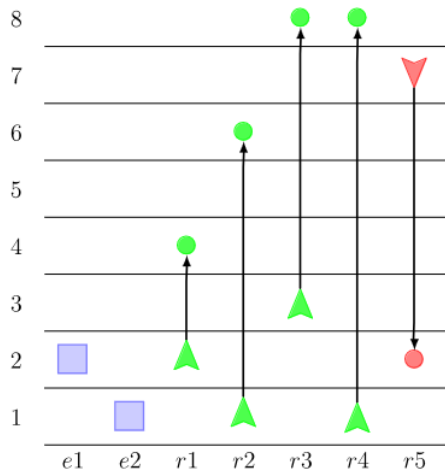
Schindler wants to evaluate the performance of their heuristic with an exact method.

We can assume that perfect information is known about future passenger arrivals.

# Destination Control



# Passenger Calls





Elevators must follow a set of rules:

- 1 Passengers should never travel in the direction opposite to their destination.
- 2 Passengers that travel in different directions cannot be in the same elevator at the same time.
- 3 Passengers should not pass their destination floor without getting out.
- 4 The elevators can only change direction when empty

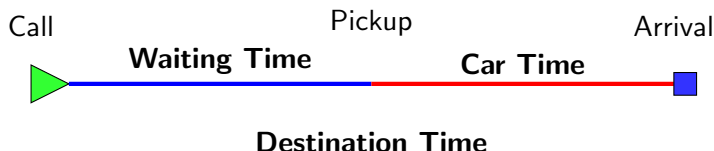
# Static EDPDC

In the static problem, assume we have perfect information about passenger arrivals. **We know the future.**



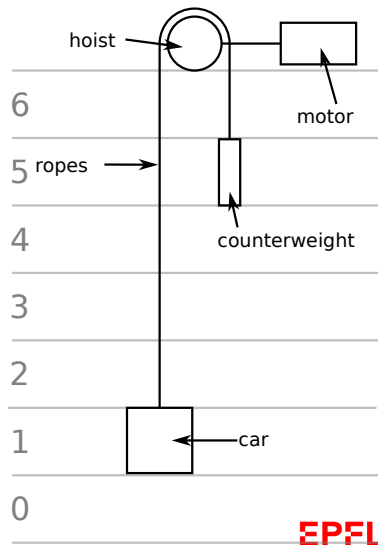
# Objective 1: minimize destination time

**Objective:** Minimize average destination time.



## Objective 2: minimize energy

$$\bar{\xi}_{ij}(q) = \left( q[kg] - \frac{Q}{2}[kg] \right) (j - i) \times$$
$$\text{distance}[m] \times g[m/s^2]$$
$$\xi_{ij}(q) = \begin{cases} \frac{1}{0.75} \times \bar{\xi}_{ij}(q) & \text{if } \bar{\xi}_{ij}(q) > 0 \\ 0.75 \times \bar{\xi}_{ij}(q) & \text{if } \bar{\xi}_{ij}(q) \leq 0. \end{cases}$$



# Calculating time

- **Doors open:** 3 seconds;
- **Doors close:** 4 seconds;
- **Elevator moves:** " $t_{i,j}$ " time to travel from floor  $i$  to floor  $j$
- **Waiting Time:** begins when a passenger makes a call and ends when the elevator arrives, before opening the doors.
- **Car Time:** begins when elevator arrives, before opening the doors and ends when the elevator arrives at the destination, before opening the doors.
- **Destination Time:** is the sum of waiting time and car time.
- **Passengers enter:** It takes 1 second for a passenger to enter the elevator once the doors have opened;
- **Passengers exit:** It takes 1 second for a passenger to exit the elevator once the doors have opened.

# Input

- **External Id** / id of each passenger;
- **Set of origins** =  $P$ ;
- **Set of destinations** =  $D$ ;
- **Time of calls** =  $a_i \quad \forall i \in P$ ;
- **Set of Elevators** =  $E$ ;
- **Set of Floors** =  $F$ ;
- **Elevator Capacity** =  $Q_e$ .

# Formulation

$$\begin{aligned} \min \quad & \sum_{r \in \Omega} c_r \lambda_r \\ \text{s.t.} \quad & \sum_{r \in \Omega} a_i^r \lambda^r = 1 && \forall i \in P \\ & \sum_{r \in \Omega} \lambda_r \leq |E| \\ & \lambda_r \in \{0, 1\} && \forall r \in \Omega \end{aligned}$$

## Part 1: Pricing problem

The reduced cost of a variable is the following:

$$\bar{c}_r = (1 - \omega)d_r + \sum_{i \in \bar{P}} \alpha_i a_i^r + \omega \times \xi - \mu \quad \forall r \in \Omega$$

- $\alpha$  is the dual variable from the constraint of each passengers in the master problem;
- $\mu$  is the dual variables for the constraint of the total number of elevators;
- $c$  is the total cost, e.g., average destination time;
- $\xi$  the energy used in the route;
- $\omega$  is a weight in the objective that multiplies the energy and the destination cost.



Let  $\mathcal{L} = (f, \ell, t, c, \delta, \bar{D}, \bar{P})$  be a label that represents a partial path (or state) of an elevator.

- $f$  current floor;
- $\ell$  current load, i.e., the total number of passengers currently in the elevator;
- $t$  the current time;
- $c$  the cost;
- $\delta$  the direction of travel, i.e., up or down;
- $D$  the set of deliveries;
- $P$  the set of all passengers picked up.

We start with a label  $\mathcal{L}_0$  and make an extension to all possible directions, i.e., pick up locations that do not violate constraints.

- **Resource extension functions (REF)** are used to keep track of resources as they pass through an arc.
- **Feasibility checks** make sure every extension is feasible, i.e., passengers do not travel in the wrong direction.
- **Dominance rules** eliminate non-promising labels

# Dominance Destination time

## Theorem 1:

Label  $\mathcal{L}_1$  dominates label  $\mathcal{L}_2$  if the following conditions are met:

- 1  $f_1 = f_2$
- 2  $\delta_1 = \delta_2$
- 3  $t_1 \leq t_2$
- 4  $\ell_1 \leq \ell_2$
- 5  $c_1 + \ell_1 * (t_2 - t_1) \leq c_2$
- 6  $D_1 \subseteq D_2$
- 7  $P_1 \subseteq P_2$

# Dominance DT and Energy $\xi$

## Theorem 2:

Label  $\mathcal{L}_1$  dominates label  $\mathcal{L}_2$  if the following conditions are met:

1  $f_1 = f_2$

2  $\delta_1 = \delta_2$

3  $t_1 \leq t_2$

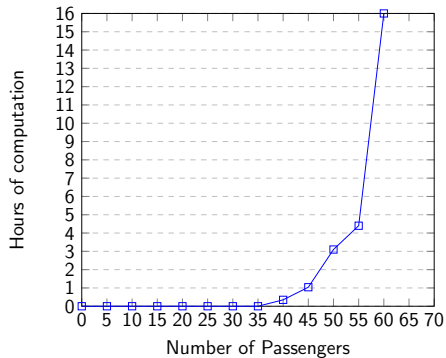
4 
$$\begin{cases} \ell_1 \leq \ell_2 & \text{if } \delta_1 = \text{up } \uparrow \\ \ell_1 = \ell_2 & \text{if } \delta_1 = \text{down } \downarrow \end{cases}$$

5  $c_1 + (1 - \omega)\ell_1 * (t_2 - t_1) \leq c_2.$

6 
$$\begin{cases} D_1 \subseteq D_2 & \text{if } \delta_1 = \text{up } \uparrow \\ D_1 \equiv D_2 & \text{if } \delta_1 = \text{down } \downarrow \end{cases}$$

7  $P_1 \subseteq P_2$

# Complexity



n	Av.DT	T[sec]
5	21.20	0.15
10	34.25	0.21
15	35.75	0.43
20	38.30	0.79
25	38.96	6.63
30	39.93	5.66
35	42.23	18.99
40	42.50	1266.17
45	42.75	3759.45
50	42.50	11157.50
55	42.44	15758.81
60	42.21	58239.26