

# Large-Scale Allocation of Personalized **Incentives** CY Transport & Urban Seminar

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- In France, the "eco-friendly" cars are subsidized and the most polluting cars are taxed.
- **Non-personalized policy:** everyone is facing the same subsidy or tax for the same vehicle model.
- Example: For electric cars, there is a subsidy of 6000 euros.
- **Inefficient policy:** individuals receive the same subsidy or pay the same tax, independently of their preferences or car usage.



Araldo, A., Gao, S., Seshadri, R., Azevedo, C.L., Ghafourian, H., Sui, Y., Ayaz, S., Sukhin, D. and Ben-Akiva, M., 2019. System-Level Optimization of Multi-Modal Transportation Networks for Energy Efficiency using Personalized Incentives: Formulation, Implementation, and Performance. Transportation Research Record, 2673(12), pp.425-438.

- Smartphone app proposing different alternatives (mode, departure time and route) to perform a trip.
- **Incentives** are proposed for energy-efficient alternatives.
- Incentive amount is the product of the energy savings with a **universal Token Energy Efficiency (TEE)**.
- Optimization problem to compute the best TEE, given a **limited budget**.
- Account for exact trip characteristics but **ignore individuals preferences**.









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- Tripod with a TEE of  $1/2$ : expenses of 3.5 euros,  $CO<sub>2</sub>$  reduced by 7.
- $\bullet$  Incentive of 1 euro to Alice and 2 euros to Bob for PT: expenses of 3 euros,  $CO<sub>2</sub>$  reduced by 7.



- **How to construct an efficient personalized policy, accounting for individual preferences?**
- **How much money is saved compared to non-personalized policies?**
- We focus on **incentive policies**, in a **discrete-choice framework**.
- A regulator uses his **limited budget** to improve a **social indicator** (e.g., CO<sub>2</sub> emissions reduction).



- The problem of finding an **optimal personalized-incentive policy** is a **Multiple-Choice Knapsack Problem**.
- We propose an algorithm to find a near-optimal policy with a large number of individuals and alternatives.
- **Numerical application** to mode choice in Lyon (France).
- Comparison with other policies.
- Discussion of **Perfect information** assumption.

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Model and Notations

- $\bullet$  Population of  $m$  individuals.
- Individual-specific choice-sets.
- $V_{i,j}$  is the **intrinsic utility** of alternative  $j$  of individual  $i$ (expressed in monetary units).
- bi*,*<sup>j</sup> is the **social indicator** of alternative j of individual i.



Incentives and Behavior

- $y_{i,j} \geq 0$  is the **monetary incentive** given to individual *i* if she accepts to choose alternative j
- Total **utility** of *i* when choosing *j* is thus

$$
U_{i,j}=V_{i,j}+y_{i,j}.
$$

 $\bullet$  The alternative chosen by individual *i* is such that

$$
j_i^* \in \arg \max_j U_{i,j}.
$$



**Regulator** 

- A regulator is endowed with a **budget** Q.
- His budget constraint is

$$
Y=\sum_i y_{i,j_i^*}\leq Q.
$$

His goal is to maximize the **global social indicator**:

$$
B=\sum_i b_{i,j_i^*}.
$$

• To do so, the regulator can choose the set of incentives  ${y_{i,j}}_{i,j}$ .



**Assumptions** 

- $\boldsymbol{\mathsf{Ind}}$ ependent intrinsic utilities:  $\boldsymbol{V}_{i,j}$  is independent of the choice of any individual  $i'\neq i.$
- **Independent social indicators:**  $b_{i,j}$  is independent of the choice of any individual  $i' \neq i$ .
- **Perfect information:** the regulator knows perfectly  $V_{i,j}$ and bi*,*<sup>j</sup> , ∀i*,* j.

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- $\bullet$  With incentives  $y_{\text{Alice, PT}} = 1 + \varepsilon$  (and nothing else), Alice will switch to public transit and it will cost  $1 + \varepsilon$  to the regulator.
- With incentives  $y_{\text{Alice}, \text{Walk}} = 2 + \varepsilon$  (and nothing else), Alice will switch to walking and it will cost  $2 + \varepsilon$  to the regulator.
- **•** Similarly, it will cost  $2 + \varepsilon$  to make Bob switch to public transit, and  $3 + \varepsilon$  to make him switch to walking.









• With budget  $Q = 1$ , the optimal policy is to induce Alice to switch to PT  $(B = 3)$ .





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- With budget  $Q = 2$ , the optimal policy is to induce Bob to switch to PT  $(B = 4)$ .





- With budget  $Q = 1$ , the optimal policy is to induce Alice to switch to PT  $(B = 3)$ .
- With budget  $Q = 2$ , the optimal policy is to induce Bob to switch to PT  $(B = 4)$ .
- With budget  $Q = 3$ , the optimal policy is to induce both Alice and Bob to switch to PT  $(B = 7)$ .



Expenses  $Y$  and social welfare  $B$  of all possible states.





**•** The **smallest incentive needed** to induce *i* to switch to alternative  *is* 

$$
w_{i,j}=V_{i,j_i^0}-V_{i,j},
$$

where  $\,V_{i,j^0_i}$  is the intrinsic utility of the alternative chosen i in absence of policy.

- One optimal policy is to set either  $y_{i,j}=0$  or  $y_{i,j}=w_{i,j}$ , ∀*i*, *j*, with at most one alternative *j* such that  $y_{i,j} > 0$ ,  $\forall i$ .
- Instead of choosing the incentive policy  $\{y_{i,j}\}_{i,j}$ , we can choose directly the set of alternatives chosen.
- **No individual gains or loses utility** with this policy.



• The optimization problem of the regulator can be written as

$$
\begin{cases}\n\max_{\{x_{i,j}\}_{i,j}} & \sum_{i} \sum_{j} b_{i,j} x_{i,j} \\
\text{s.t.} & \sum_{i} \sum_{j} w_{i,j} x_{i,j} \leq Q \\
& \sum_{j} x_{i,j} = 1, \qquad \forall i \\
& x_{i,j} \in \{0, 1\}, \qquad \forall i, \forall j\n\end{cases}
$$

- This is a **Multiple-Choice Knapsack-Problem**.
- An approximate solution can be found in polynomial time.

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### Optimal Policy

Multiple-Choice Knapsack-Problem



Multiple-choice knapsack problem

Source: Xu, S., Chen, X., Pi, X., Joe-Wong, C., Zhang, P. and Noh, H.Y., 2019, March. Incentivizing vehicular crowdsensing system for large scale smart city applications. In Sensors and Smart Structures Technologies for Civil, Mechanical, and Aerospace Systems 2019 (Vol. 10970, p. 109701C). International Society for Optics and Photonics.

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**•** The **efficiency** of switching individual *i* from alternative *i* to alternative  $j^\prime$  is

$$
\tilde{e}_{i,j\rightarrow j'}=\frac{b_{i,j'}-b_{i,j}}{w_{i,j'}-w_{i,j}}.
$$

**Iterative procedure:** we "switch" individuals one by one to their subsequent alternative, according to decreasing efficiency, until the budget is depleted.



Budget:  $Q = 3$ 







• Iteration 1: Best switch is Alice from Car to PT (efficiency 3, remaining budget 2).





- Iteration 1: Best switch is Alice from Car to PT (efficiency 3, remaining budget 2).
- Iteration 2: Best switch is Bob from Car to PT (efficiency 2, remaining budget 0).



- Solution is boundedly close to the optimum.
- Solution is optimal for the budget spent at any iteration.
- **Diminishing returns:** social welfare is concave with the expenses  $\bullet$ of the regulator.



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#### <span id="page-27-0"></span>Application to Mode Choice Data

- Census data for **220k individuals** in Lyon's area (France): home, workplace, mode of transportation for commuting, socio-demographic variables.
- **•** Travel times computed from OpenStreetMap and HERE data.
- **5 modes of transportation:** car, public transit, walking, cycling and motorcycle.
- For 16k individuals, public transit is not an alternative (no route).

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# Application to Mode Choice

Intrinsic Utilities and Social Indicators

- Intrinsic utilities are estimated from a Multinomial Logit model.
- Social indicators are the **CO<sup>2</sup> emissions**, computed with data from ADEME.



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#### Application to Mode Choice Social Welfare Curve



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### Application to Mode Choice **Results**

- Budget is set to  $Q = 1800$  euros (per day).
- After 3500 iterations, 1798.59 euros are spent.
- Only **1.57 % of individuals receive incentives**.
- Reduction of  $CO<sub>2</sub>$  of 18 tons per day (3 % of total emissions).
- Average cost is 100 euros per ton of  $CO<sub>2</sub>$ .

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#### Application to Mode Choice **Results**



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### Application to Mode Choice **Results**



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## Comparison with Other Policies

Application Results

- Enforcement: Individuals are forced to choose an alternative.
- Proportional tax: Alternatives are taxed proportionally to their  $CO<sub>2</sub>$  emissions.
- Tripod incentives: Incentives proportional to the  $CO<sub>2</sub>$ emissions of the alternatives.



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- Personalized-incentive policy boundedly close to optimum can be computed with **MCKP algorithms from Operations research**.
- The policy shows **diminishing returns** behavior.
- The allocation can be reached with **proportional taxation** or with **Tripod**.
- Numerical results show that the personalized policy can reach the same social target than non-personalized policies with **half the budget**.



- With **imperfect information**, we can use the same algorithm and set the incentive amounts to the expected utility difference. Results could be improved by learning from the responses of the individuals to the incentives.
- **Congestion** can be accounted for with an iterative procedure.