

# A new mathematical formulation to integrate supply and demand within a choice-based optimization framework

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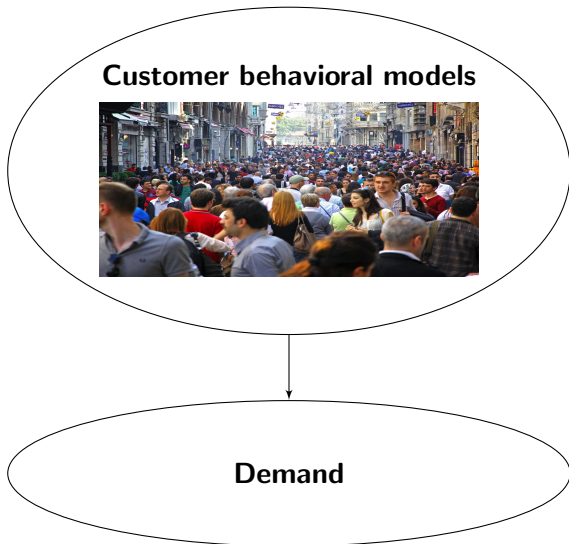
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# Outline

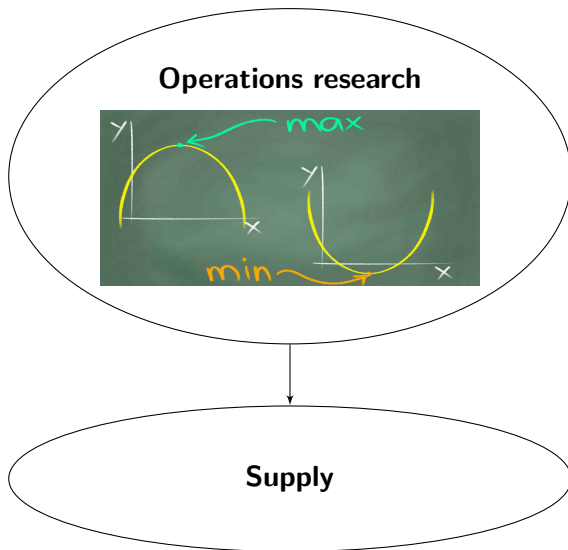
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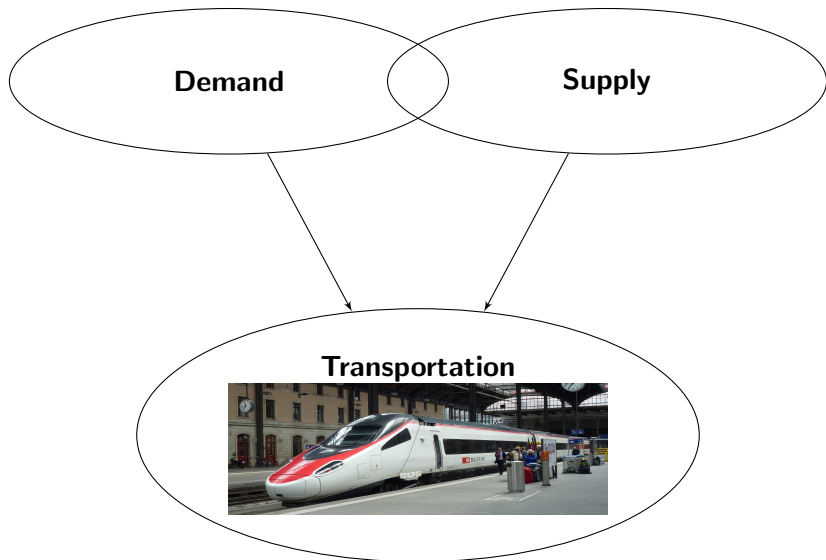
# Motivation



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# Demand vs. supply

## Customer behavioral models

- Given the configuration of the system  $\Rightarrow$  predict the demand
- Maximize satisfaction
- **Here:** discrete choice models

## Operations Research

- Given the demand  $\Rightarrow$  configure the system
- Minimize costs
- **Here:** MILP

## Discrete choice models in optimization problems

- Integrated choice model  $\Rightarrow$  source of nonconvexity
- Many techniques to convexify and linearize. **Here:** different approach
  - Nonconvex representation of choice probabilities
  - Include a wide class of discrete choice models

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# Utilities



## Demand and supply

- Population of  $N$  individuals
- Set of products  $\mathcal{C}$  in the market
  - artificial "opt-out" product
- $\mathcal{C}_n \subseteq \mathcal{C}$  subset of available products to individual  $n$

## Utility

$U_{in}$  associated score with alternative  $i$  by individual  $n$ :  $U_{in} = V_{in} + \varepsilon_{in}$

- $V_{in}$ : deterministic part
- $\varepsilon_{in}$ : error term

**Behavioral assumption:**  $n$  chooses  $i$  if  $U_{in}$  is the highest in  $\mathcal{C}_n$

# Probabilistic model

## Choice variable

$$w_{in} = \begin{cases} 1 & \text{if } n \text{ chooses } i \\ 0 & \text{otherwise} \end{cases} \quad \forall n, \forall i \in \mathcal{C}.$$

## Probabilistic model

$$\Pr(w_{in} = 1) = \Pr(U_{in} \geq U_{jn}, \forall j \in \mathcal{C}_n)$$

$$D_i = \sum_{n=1}^N \Pr(w_{in} = 1)$$

# Simulation

## Non linearity

$D_i$  is in general non linear

### Example:

$$\Pr(w_{in} = 1) = \frac{y_{in}e^{V_{in}}}{\sum_{j \in C} y_{jn}e^{V_{jn}}} \quad (\text{logit model})$$



## Simulation

- Assume a distribution for  $\varepsilon_{in}$
- Generate  $R$  draws  $\xi_{in1} \dots \xi_{inR}$
- $r$  behavioral scenario
- The choice problem becomes **deterministic**

# Demand model

$$U_{inr} = V_{in} + \xi_{inr} = \sum_k \beta_k x_{ink} + f(z_{in}) + \xi_{inr}$$

$\Rightarrow U_{inr}$  is not a random variable

## Endogenous part of $V_{in}$

- Linear in the variables  $x_{ink}$
- Decision variables (involved in the optimization problem)
- Assumption for the integration in a MILP

## Exogenous part of $V_{in}$

- Depends on other variables  $z_{in}$
- $f$  not necessarily linear

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# Mixed Integer Linear Problem

## Availability of alternatives

- Availability at operator level
- Availability at scenario level (e.g. demand exceeding capacity)

## Preference of alternatives

- Take into account only the available alternatives
- Choose the alternative with highest utility

## Choice

Choice at scenario level:  $w_{inr}$

$$D_i = \frac{1}{R} \sum_{n=1}^N \sum_{r=1}^R w_{inr}$$

# Demand based revenues maximization (I)

## Application

- Operator selling services to a market, each offered service:
  - Price
  - Capacity (number of customers)
- Demand is price elastic and heterogenous
- **Goal:** best strategy in terms of capacity allocation and pricing

## Maximization of revenues

- $p_{in}$  price that individual  $n$  has to pay to access service  $i$

$$\max R_i = \max \frac{1}{R} \sum_{n=1}^N p_{in} \sum_{r=1}^R w_{inr}$$

- $p_{in}$  endogenous variable  $\Rightarrow R_i$  non linear

# Demand based revenues maximization (II)

## Pricing

- Linearization of the price
- Discretization  $\Rightarrow$  price levels

## Capacity allocation

- Capacity for each alternative  $i$ 
  - We assume it given
  - It could be a decision variable
- Who has access?
- Provide a priority list to the model



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# Perspective

## Conclusions

- High dimensionality of the problem ( $INR$ )
- Any assumption can be made for the  $\varepsilon_{in}$

## Ongoing research

- **Proof of concept:** case study from the literature (mixed logit)
- Define different scenarios to test the formulation
- More accurate values (e.g. price levels)

## Future work

- Decomposition techniques to speed up the computational results
  - By **customer**: capacity!
  - By **scenario**: only considered together in the objective function
- Introduce new features (e.g.  $N$  as a group of individuals), capacity?

# Questions?

