A short discussion about travel demand models

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Travel demand

Most people don't travel for the sake of it Travel demand = derived demand **Results of many choices:** Choice of activity Choice of destination Choice of departure time Choice of transportation mode Choice of access point (parking, bus stop) Choice of itinerary Etc...

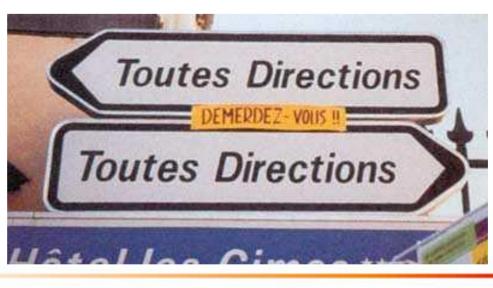




Route choice for car drivers









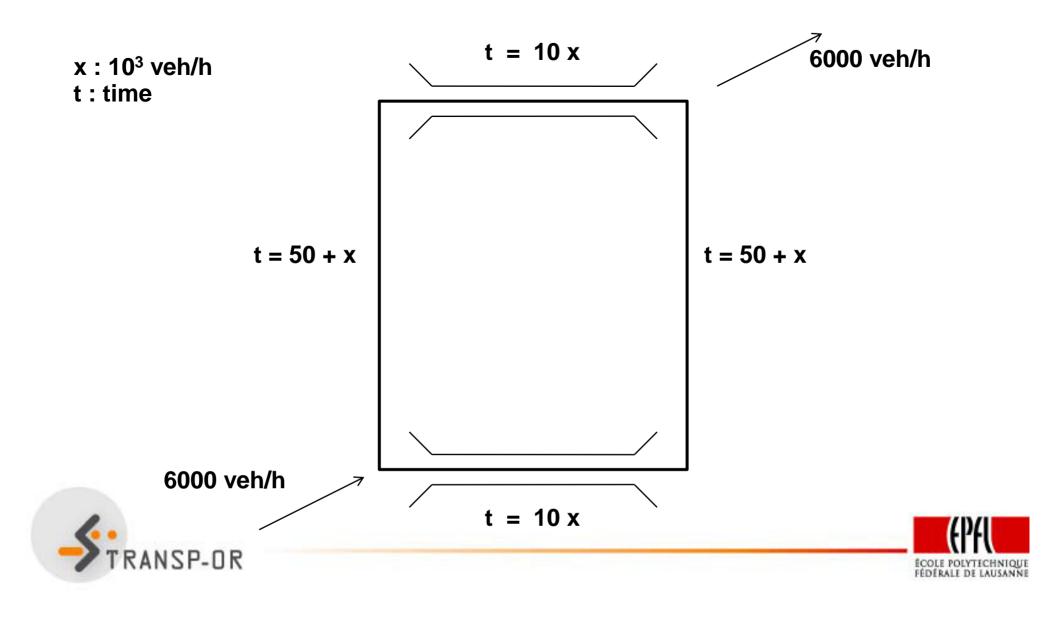


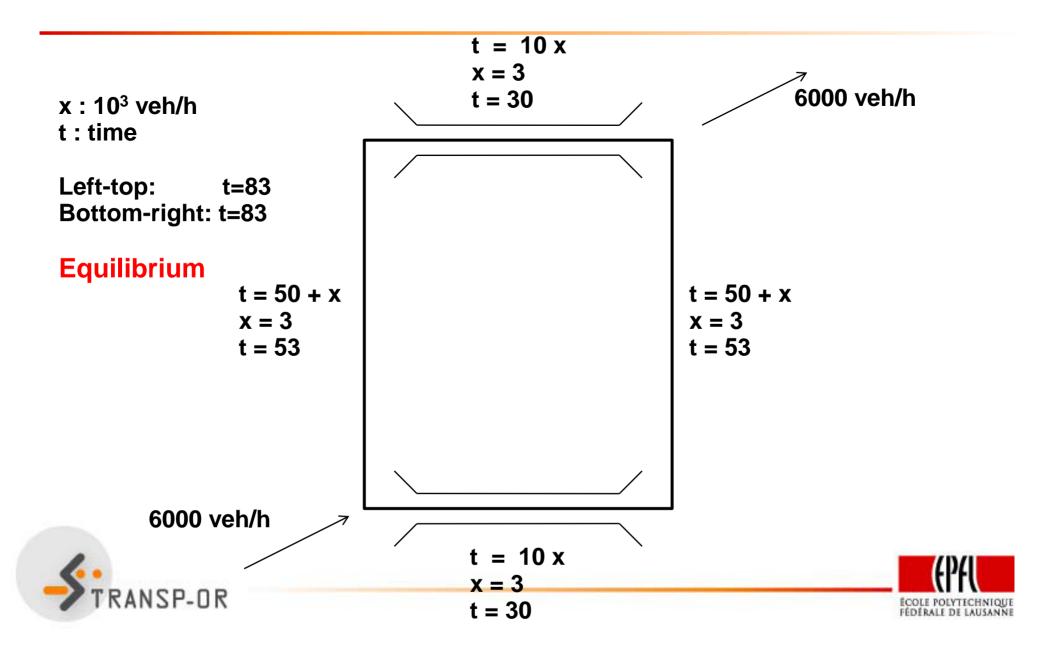
Route choice for car drivers

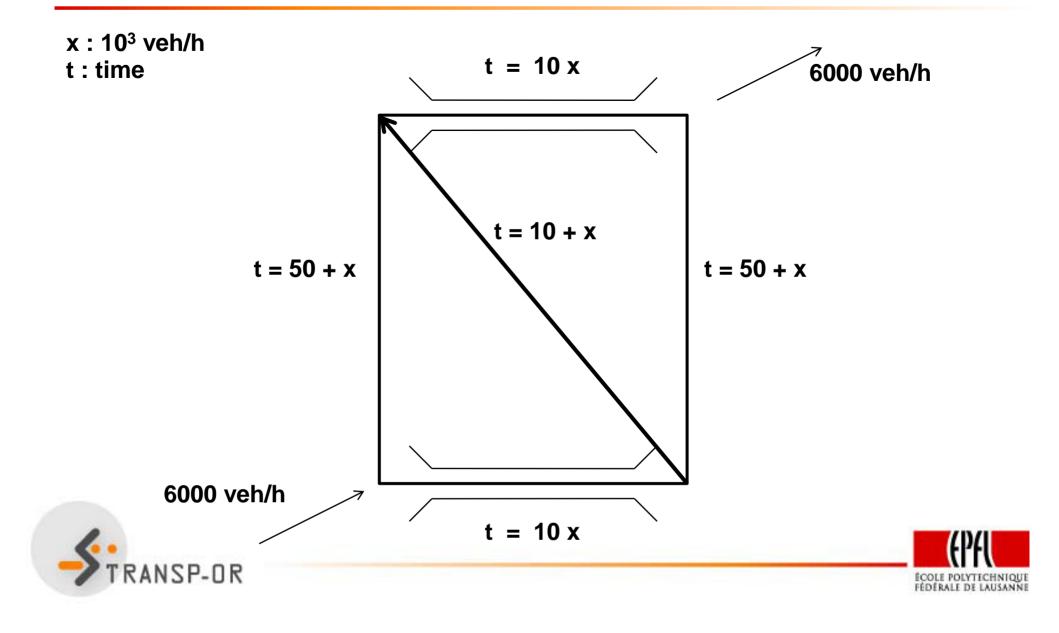
- Assumption #1: drivers prefer the fastest route
- Warning:
 - Their presence affects the other drivers
 - More cars = increased travel time
- So...
 - Travel time influences route choice
 - Route choice influences travel time

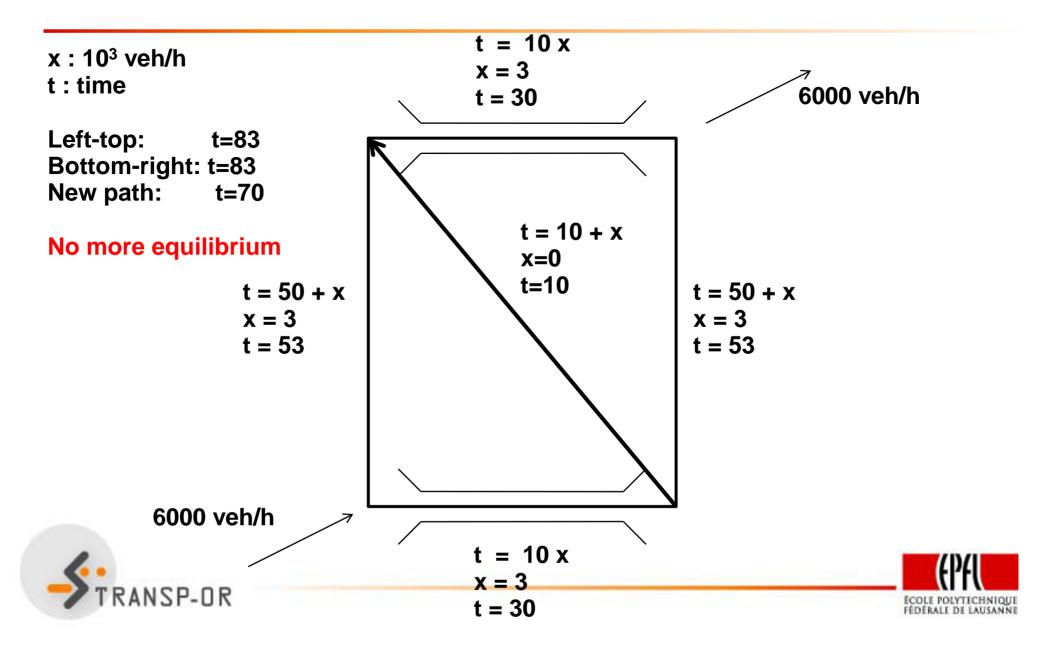


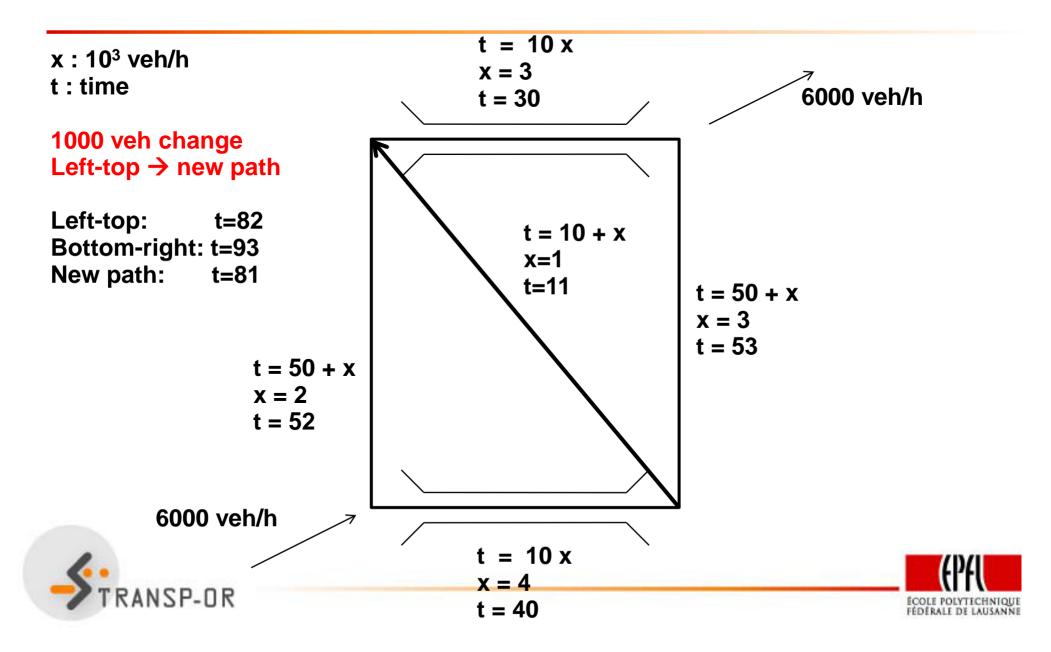


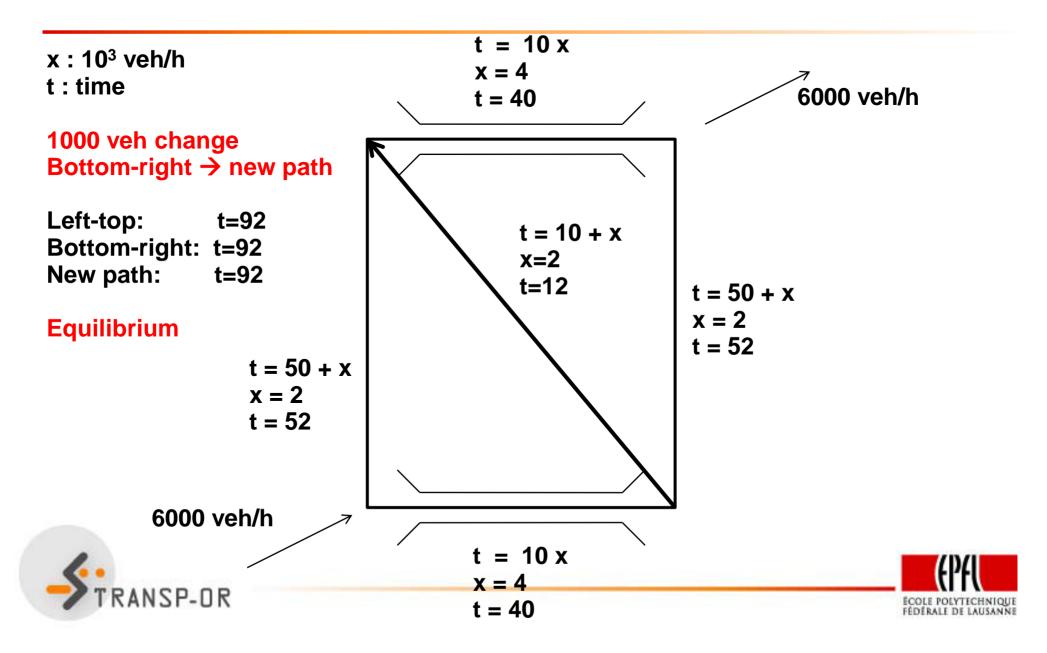












- A new infrastructure is built
- Before, travel time = 83 minutes
- After, travel time = 92 minutes

Increasing the physical capacity of the network does not necessarily increase the mobility

Braess' paradox





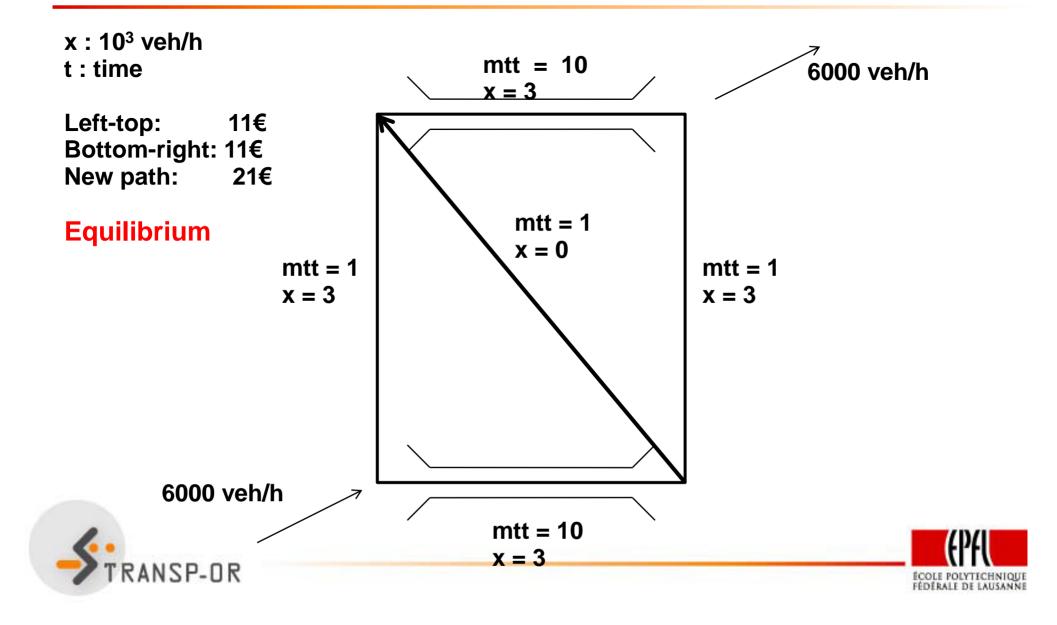
Polluters pay principle

- Concept of marginal travel time
 - t = 50 + x Marginal ttime = 1
 - t = 10 + x Marginal ttime = 1 t = 10 x Marginal ttime = 10
 - Drivers are tolled proportionally to the nuisance they produce
- 1 min marginal travel time = 1€
- Assumption #2: drivers prefer the cheapest route





Back to the simple example



Behavioral assumption?

- Do people minimize time?
- Do people minimize cost?
- Each assumption gives different results
- Behavior is more complex...





Time is money

Path 1: 11€ - 83 minutes
Path 2: 11€ - 83 minutes
Path 3: 21€ - 70 minutes



- Would you be willing to pay 10€ to save 13 minutes ?
- Assumption #3: drivers consider both time and cost
- But how do we identify the best path then?





- Idea : drivers combine cost and time into a number called "utility"
- The selected route is the one with the largest utility.
- Example with two routes:

$$U_1 = -\beta t_1 - \gamma c_1$$
$$U_2 = -\beta t_2 - \gamma c_2$$

β, γ > 0





$$U_1 = -\beta t_1 - \gamma c_1$$
$$U_2 = -\beta t_2 - \gamma c_2$$
$$U_1 > U_2 \text{ if } -\beta t_1 - \gamma c_1 \ge -\beta t_2 - \gamma c_2$$

$$-\frac{\beta}{\gamma}t_1 - c_1 \ge -\frac{\beta}{\gamma}t_2 - c_2$$

$$c_1 - c_2 \le -\frac{\beta}{\gamma}(t_1 - t_2)$$

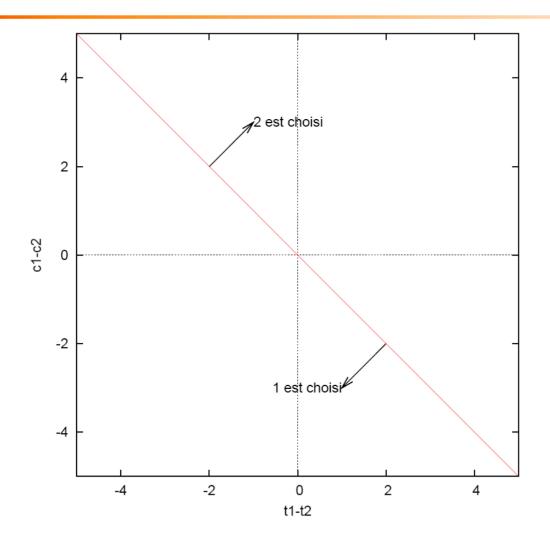




- Dominated cases:
- $c_1 > c_2$ and $t_1 > t_2$: 2 is dominating 1
- c₂ > c₁ and t₂ > t₁: 1 is dominating 2
 What about the trade-offs for non-dominated cases?

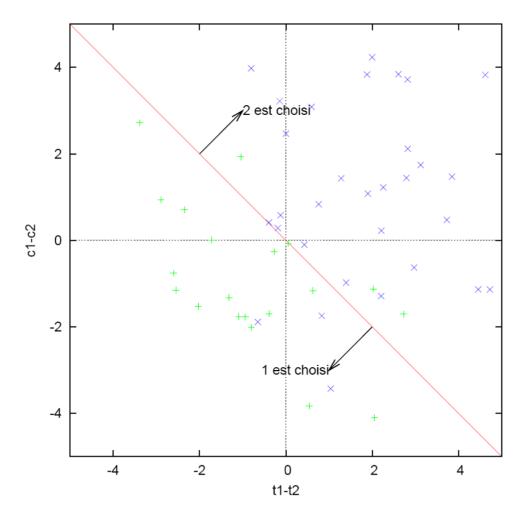
















Need for a random termNow, probability must be used

$$U_1 = -\beta t_1 - \gamma c_1 + \varepsilon_1$$
$$U_2 = -\beta t_2 - \gamma c_2 + \varepsilon_2$$

P(1) = P(U₁>U₂)
Most famous model : the multinomial logit model





Multinomial logit model

$$U_{in} = V_{in} + \varepsilon_{in} = \beta_1 x_{in1} + \beta_2 x_{in2} + \ldots + \varepsilon_{in}$$

 where x include time, cost, number of speed bumps, number of left turns, type of routes, etc.

$$P_n(i|\mathcal{C}_n) = \Pr(U_{in} \ge U_{jn} \; \forall j \in \mathcal{C}_n)$$

$$P_n(i|\mathcal{C}_n) = \frac{e^{V_{in}}}{\sum_{j \in \mathcal{C}_n} e^{V_{jn}}}$$





Value of time in Switzerland

• We can measure the willingness to pay for travel time savings

Axhausen, K., Hess, S., Koenig, A., Abay, G., Bates, J., and Bierlaire, M. (to appear). Income and distance elasticities of values of travel time savings: new Swiss results, *Transport Policy*

	Trip purpose					
WTP at sample mean	Business	Commuting	Leisure	Shopping		
PT travel time (CHF/hour)	49.57	27.81	21.84	17.73		
Car travel time (CHF/hour)	50.23	30.64	29.2	24.32		
Headway red.(CHF/hour)	14.88	11.18	13.38	8.48		
Interchange red. (CHF/change)	7.85	4.89	7.32	3.52		





Value of time in Switzerland

WTP at sample mean	Business	Commuting	Leisure	Shopping
PT Travel time (€/h)	30.2	17.0	13.3	10.8
Car travel time (€/h)	30.6	18.7	17.8	14.8
Headway red. (€/h)	9.1	6.8	8.2	5.2
Interchange red. (€/change)	4.8	3.0	4.5	2.1





Optimal pricing

- Price = z, Population = N
- Choice model:
- P(choosing the train | z)
 Number of people choosing the train: N P(choosing the train | z)
- Revenues:

R(z) = N P(choosing the train ¦ z) z Optimal pricing:

 $Max_{z} R(z)$





Recent developments in route choice

Route choice modeling difficult because

- Large number of alternatives
- High structural correlation due to the physical overlap of paths
- Difficulty to collect data (reports, GPS)
 Solutions we have proposed
- Sampling of alternatives
- Concept of subnetworks
- Measurement equations





Summary

- Travel demand is complex
- Simple assumptions are useful but not sufficient
- Need to analyze the situation as a whole (beware of the Braess paradox)
- Observing and measuring behavior is critical (ex: willingness to pay)
- Random utility models are at the core of disaggregate demand modeling
 Hot topic: route choice models



