

The
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Choice, price, and traffic: an assignment model for MaaS

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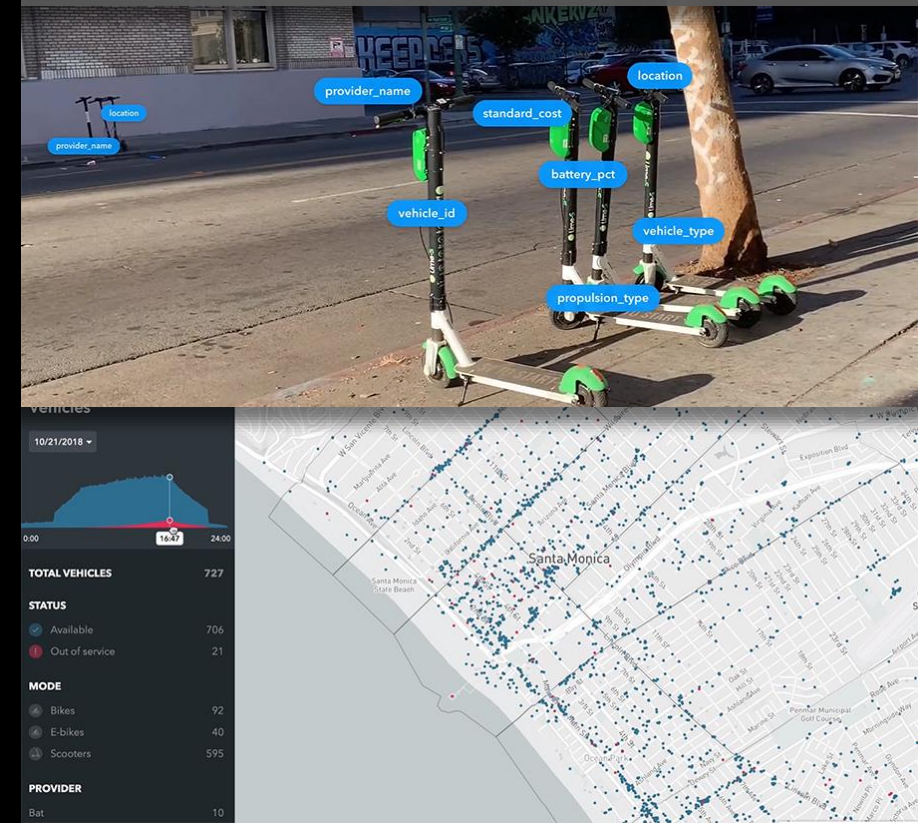


Cities need to be able to model MaaS, to support policy decisions.

- For simple policy decisions, the city needs to predict what will happen.
e.g. If we licensed Uber, how many people would use it, and on which routes?
- For advanced city management, the city needs to understand how MaaS users and operators react.
e.g. If we used micro-charging via an API, as Los Angeles is contemplating, how should we set charges?

Seleta Reynolds, general manager,
Los Angeles Department of Transport:

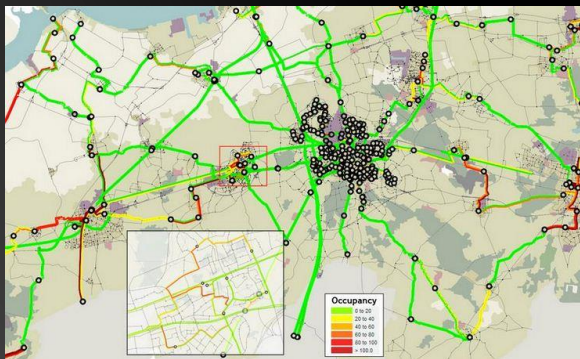
“For example, let’s say the Lyft app pings the city and says, ‘I want to pick up and drop off at this corner’. But [at that corner] there’s a bike lane where we don’t want pickups and drop-offs. The city’s operating system might say, ‘You need to go right around the corner. We have a clear space for you and it’ll cost you 50 cents a minute’.”



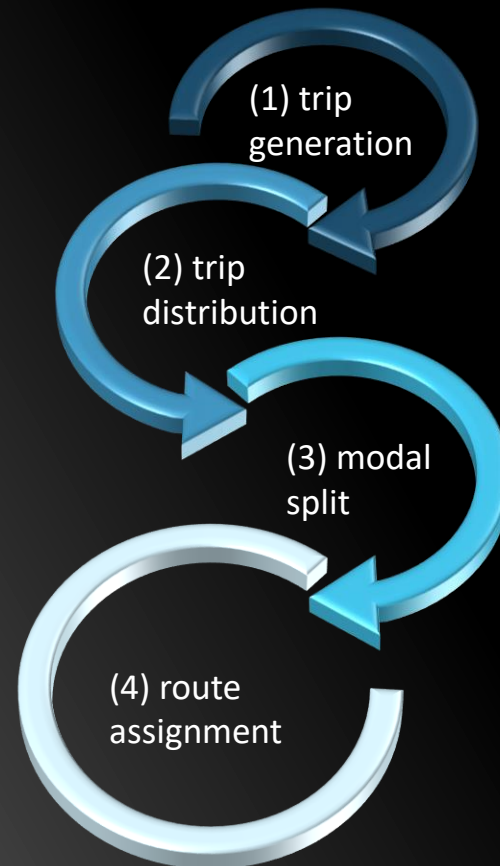
For conventional road planning, the four-step approach is in widespread use. But it doesn't work for MaaS.

The four-step model is in widespread use, for predicting traffic on roads.

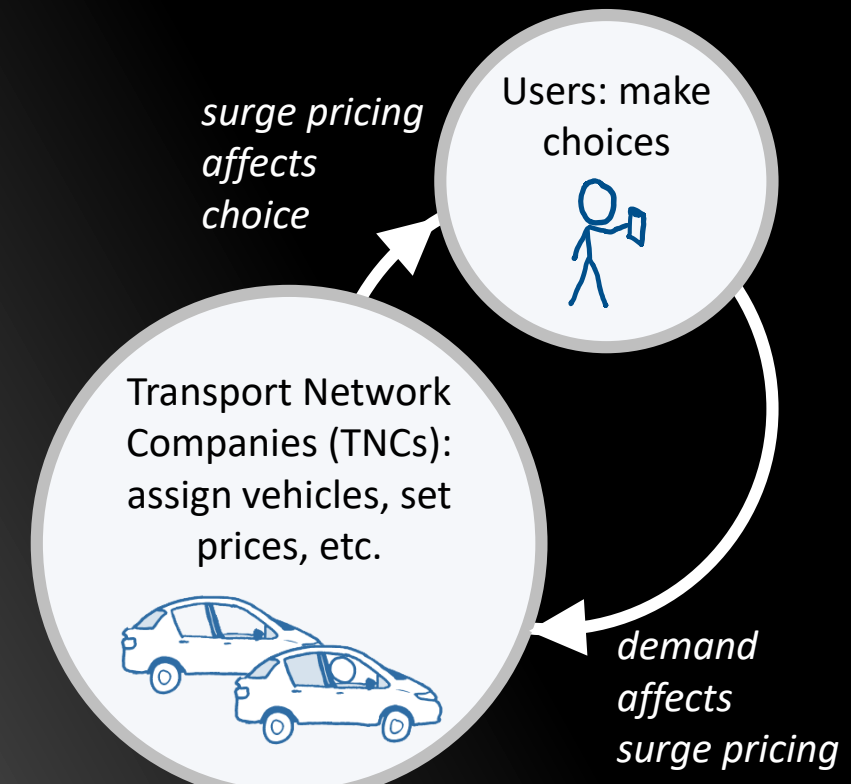
Software packages:
Aimsun, PTV VISUM, etc.



In the four-step model, we estimate the number of car trips (3), before working out the routes (4).



- In MaaS, there is a feedback loop between mode choice (3) and vehicle routing (4).
- TNCs control a whole fleet, so their route assignment is much more complex than in the four-step model.



We have developed new economics-based theory for assignment, which extends the four-step model.

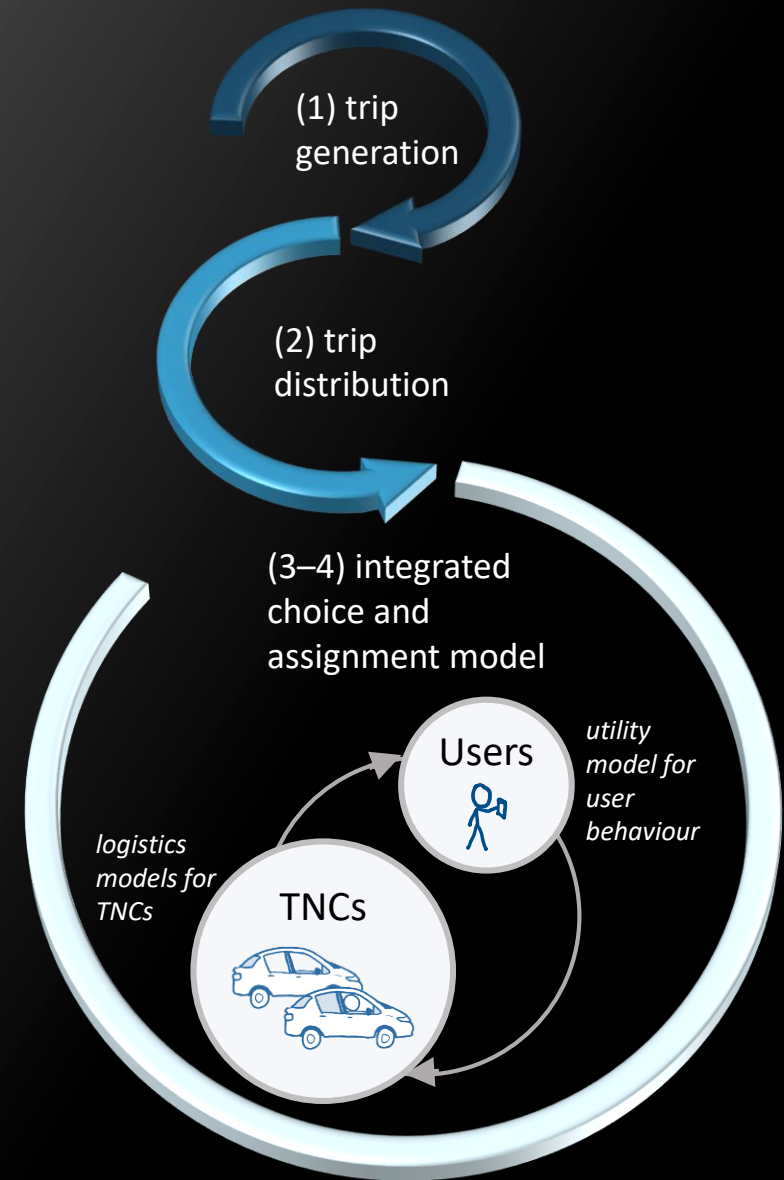
Model output

- rideshare prices
- driver assignments
- passenger choice of rideshare or other modes

Model input

- operational cost of running rideshare (travel time, fuel and wear and tear, surcharges)
- fares and travel time for other modes

We can explore policy variations by modifying the operational parameters for rideshare.



- *We model users using utility theory—this is an economics-based approach*
- *We treat TNCs as efficient fleet operators: we assume they've worked out how to manage their drivers*

Some MaaS scenarios, which we have considered in more or less detail ...

What is a fair cost for running a rideshare service?

To model this: compare a for-profit rideshare model, with a non-profit model.

Should the city levy a surcharge on empty rideshare vehicles?

To model this: increase the operational cost for empty vehicles.

Should rideshare vehicles be allowed to use bus lanes?

To model this: alter the travel time for rideshare.

What is the impact of rideshare on congestion and travel time?

To model this: include congestion delay in the assignment model.

Can we replace under-used bus routes with rideshare?

To model this: remove the public transport mode option from selected areas.

Where should the rideshare company be allowed to operate?

To model this: set the user choice to always prefer other modes, outside the permitted area.

Should the city levy a congestion charge on rideshare vehicles?

To model this: use two types of rideshare vehicle, one that has paid, one that hasn't.

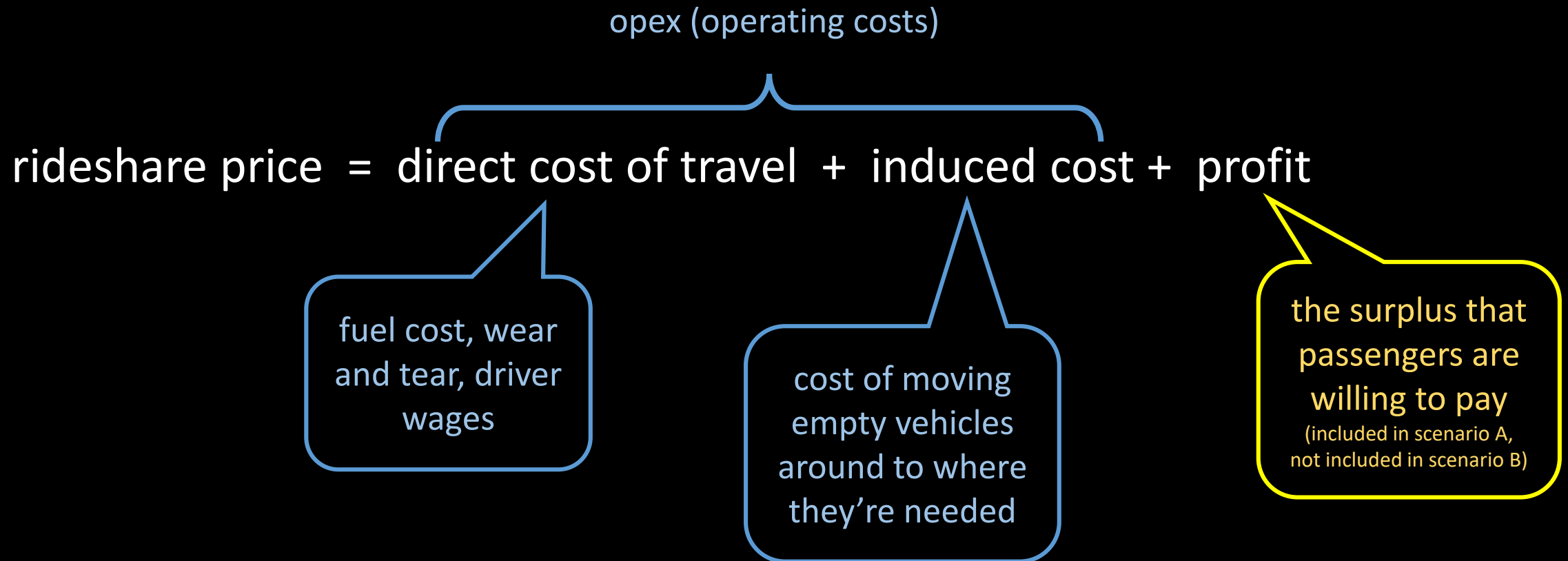
scenarios we've explored

*scenarios that are easy to explore:
all that's needed is the data*

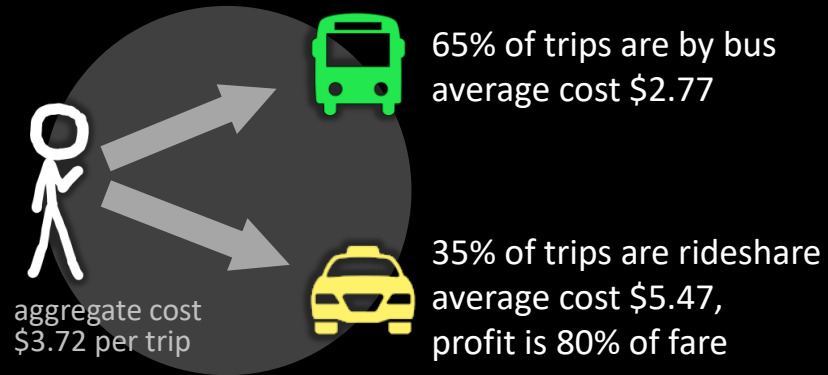
*scenarios that take a little more
model development work*

Case study: what is a fair cost for running a rideshare service?

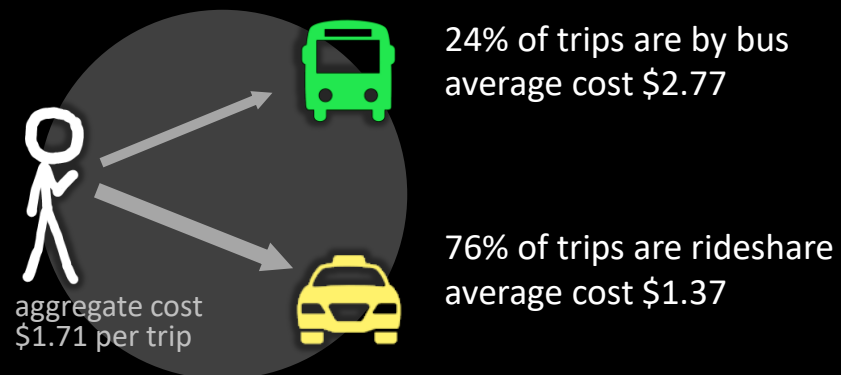
- We ran two scenarios:
(A) for-profit monopoly rideshare, competing with buses
(B) non-profit rideshare, competing with buses
- We modelled user preference and travel demand, from the Bay Area Travel Survey 2000, and we used data from bus trips and taxi trips (so this is just a demonstration of the technique, not a serious estimate!)
- We modelled just the rideshare operator's task of assigning vehicles and setting prices; we ignored congestion. We treated the rideshare operator as efficient: we didn't model the details of driver incentives.



SCENARIO A: buses, and for-profit monopoly rideshare



SCENARIO B: buses, and non- profit rideshare



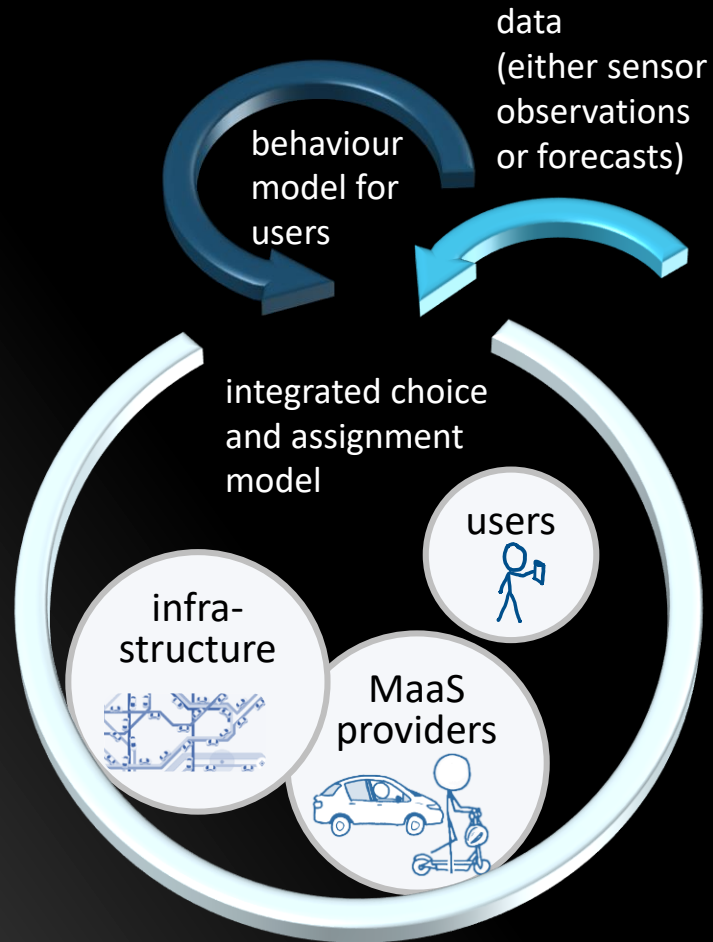
Sanity check:

For this exercise, we used demand data from 2000. We found that the profit is 80%, for a monopoly rideshare operator.

A recent survey found average profit 35–38%, worldwide, for competitive operators.

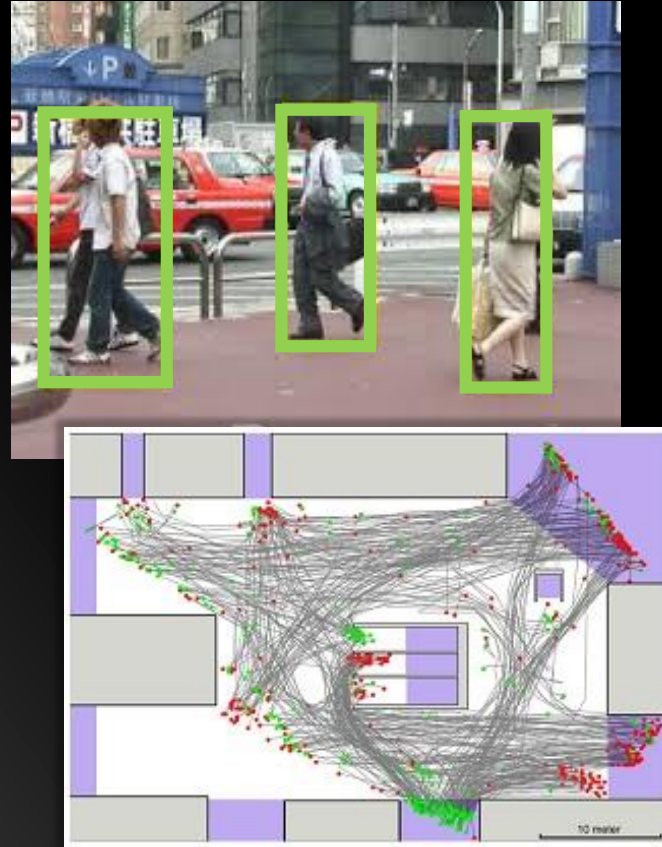
<https://jalopnik.com/uber-and-lyft-take-a-lot-more-from-drivers-than-they-sa-1837450373>

Our model extends further, to a wide class of problems based on prediction or inference about user choice.



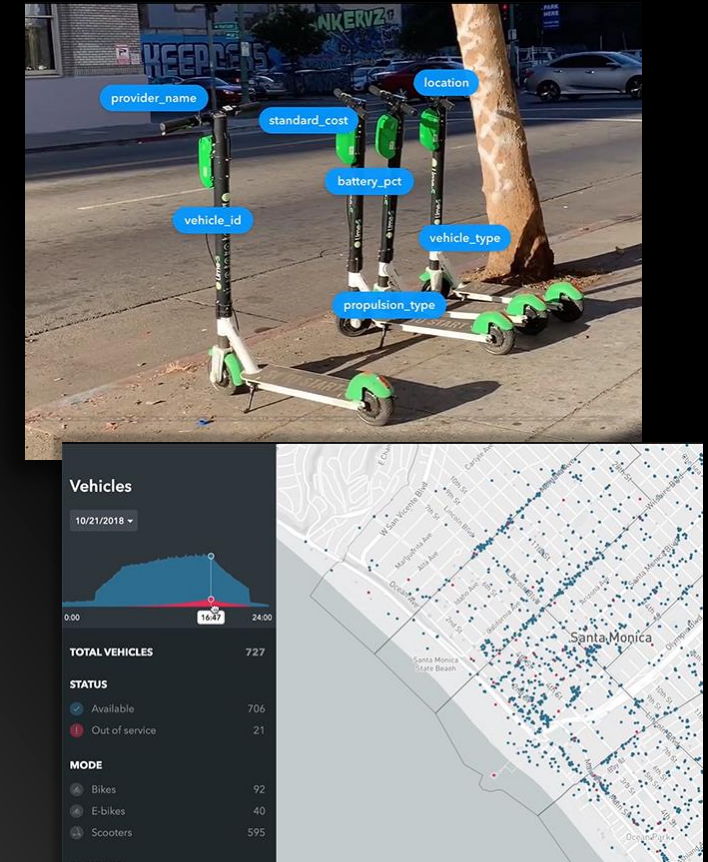
The full economic assignment model

We have developed a flexible economics-driven approach. It unifies steps 2–4 of the four step model. It permits machine-learnt user models.



How do people use the streets?

Using pedestrian counts from stationary cameras, we can infer the patterns of movement around a district.



Other MaaS options

The model can be extended to describe dockless scooters etc. This could be used to set tariffs relating to availability or clutter.

We are building a “Mobility Data Toolkit”, an interactive system for exploring MaaS scenarios (and more). It combines macrosimulation, microsimulation, and report generation. See the accompanying video for more.

