Modelling competition in ride hailing platforms

The limits of network effects replicated using a biological model

Remi K. Uzel

June 23 2020





The growth of ride-hailing platforms

Ride Hailing Platforms are everywhere

 After 11 years, Uber serves 785 metropolitan areas (63 countries)



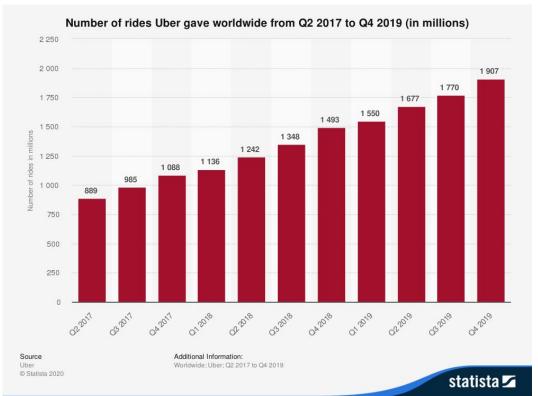
- 14 million rides a day
 - o 18k per city
- 26% of 18-24 yo in London use Uber





Their growth is impressive

- Number of rides given by Uber nearly doubles in 2 years
- Averages a quarterly growth of 8%



They are raising questions about competition

- Will Uber bring healthy competition, or crush the market into a monopoly?
- These platforms are thought to benefit from a "rich get richer" effect¹. Is this growth uncontrolled?



Uber versus black cabs: Battle lines



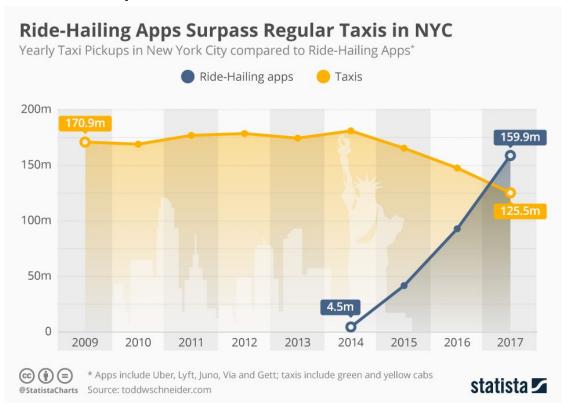
A clash is about to take place on the capital's streets that sees the old pitted against the new. The livelihoods of London's iconic cabbies are at stake. But could they be stifling opportunities for a new breed of driver - that could even include you?

London's black-cab drivers are preparing to cause gridlock in the capital on Wednesday, in protest at Transport for London's stance towards a mobile phone app that allows users to book private vehicles, using GPS technology to set the fare.

It's the hottest topic of conversation amongst cab drivers since Olympic traffic lanes. What should be done about Uber?

Black-cab driver Lloyd Baldwin is in no doubt. "Our beef with Uber is that these drivers have come straight into London, and have been licensed straight away by

Their growth is impressive

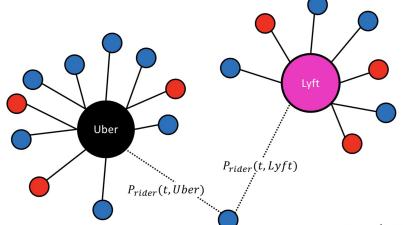


What are the dynamics of this growth?

Network Effects: a key contributor

Rider agent Driver agent

- More platform riders -> more drivers
- More drivers -> overall service is more appealing to riders
- Ride hailing platforms are known to have this dynamic 1



$$P_{rider}(t, Uber) \sim \frac{k_{Uber}}{\sum_{i} k_{i}}$$
 $P_{rider}(t, Lyft) \sim \frac{k_{Lyft}}{\sum_{i} k_{i}}$

$$P_{rider}(t, Lyft) \sim \frac{k_{Ly}}{\sum_{i} k_{i}}$$

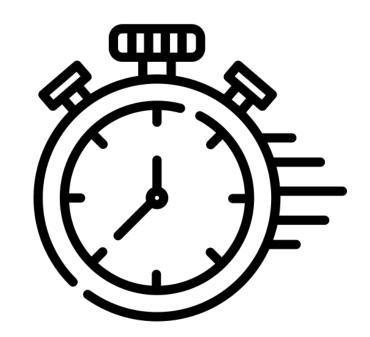
There seems to be an inherent limit...

- Unmitigated network effects lead to explosive growth
- Competition would be improbable everywhere

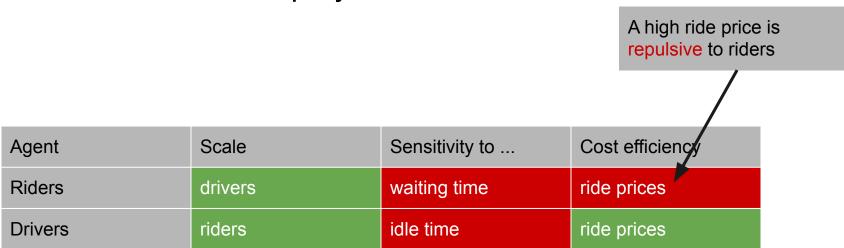
Because of their growth, it is assumed that a strong first-mover advantage prevails in the market. This isn't always the case and we study what could limit this.

Could the limit be the waiting time?

- Platforms have a key attractor: waiting/idle time
- Once it is low enough, is it still attractive?
 - o 3 min vs 6 min
 - o 1 min vs 2 min?
 - o 30s vs 1 min??



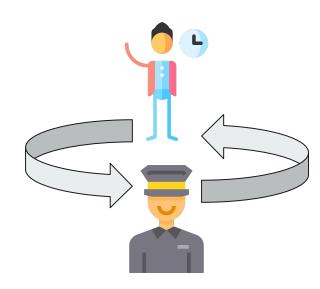
What are the forces at play?



green = positive factor | red = negative factor

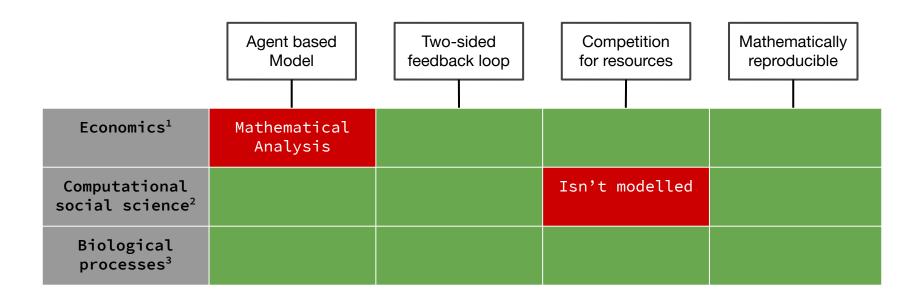
A two sided feedback loop: agent based competition

- As a rider, drivers are a vital resource
- As a driver, riders become the resource
- This results in a two sided competitive feedback loop
 - competition among drivers (and riders)
 - competition for riders (and drivers)



Summary of applicable models

Overview of modelling literature



¹ Sun, M. and Tse, E., 2009. The resource-based view of competitive advantage in two-sided markets. Journal of Management Studies, 46(1), pp.45-64.

² Barabási, A.L. and Albert, R., 1999. Emergence of scaling in random networks. *science*, 286(5439), pp.509-512.

³ Delitala, M., Lorenzi, T. and Melensi, M., 2014. A structured population model of competition between cancer cells and T cells under immunotherapy. In *Mathematical Models of Tumor-Immune System Dynamics* (pp. 47-58). Springer, New York, NY.

Ride-hailing platforms in immunology?

Platform agents:

agents compete amongst themselves for service (either rides or riders)

As the population of agents increase, the effect of their respective waiting/idle times is magnified

Human cells:

 Systematically compete amongst themselves for the body's ressources

Overcrowding naturally hinders cell-growth

From immunology to ride-hailing platforms

- Directly models the population of two cell types (agents)
- Rigorous mathematical model
- Parameters can be mapped naturally to our situation
- Overcrowding and competition for resources are taken into account

A Structured Population Model of Competition Between Cancer Cells and T Cells Under Immunotherapy

Marcello Delitala, Tommaso Lorenzi, and Matteo Melensi

The model's growth rate equations

$$R_{rider}(t,u_i) := \kappa_{driver}(t,u_i) - \mu_{waiting} \rho_{rider}(t,u_i) - \eta \rho_{rider}(t,u_i)$$

$$R_{driver}(t,u_i) := \kappa_{rider}(t,u_i) - \mu_{idle} \rho_{driver}(t,u_i) + \eta \rho_{rider}(t,u_i)$$
Price factors

with
$$\rho_A(t,u) = \frac{f_A(t,u)}{f_{rider}(t,u) + f_{driver}(t,u)}$$
 Overpopulation factor

with
$$\rho_A(t,u) = \frac{f_A(t,u)}{f_{rider}(t,u) + f_{driver}(t,u)} \qquad \kappa_A(t,u_i) = \frac{f_A(t,u_i)}{\sum_{j\neq i}^U f_A(t,u_j)} \qquad A \in \{rider, driver\}$$

Results

The data: NYC TLC*

- Monthly ride data from 2015 to 2019
- About 800 million individual rides
 - Spread out through 4 platforms: Uber,
 Lyft, Via and Juno
- Pickup/dropoff datetime (to the second)
- Location IDs:
 - LON/LAT for yellow/green cab
 - Split into neighbourhoods for our platforms



The Simulation

At each timestep X agents are generated and become riders with probability p

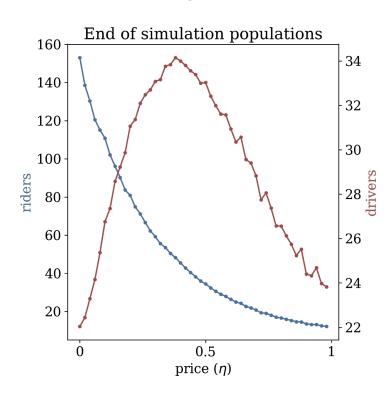
• Agents choose which platform to join w.r.t. their type's rate equations

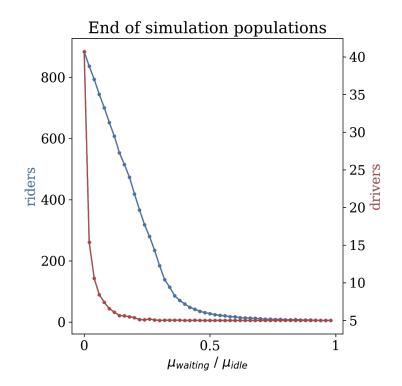
The three* free parameters are:

- $m{\mu}_{ ext{waiting}}$ the sensitivity of riders on waiting time
- $m{\mu}_{ ext{idle}}$ the sensitivity of drivers on idle time
- p the proportion of agents that are riders / drivers

$$P_A(t, u_i) = \frac{R_A(t, u_i)}{\sum_j R_A(t, u_j)}$$

Reproducing intuitive results

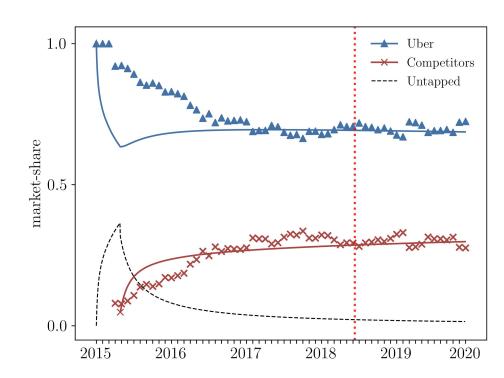




Market-share simulation results

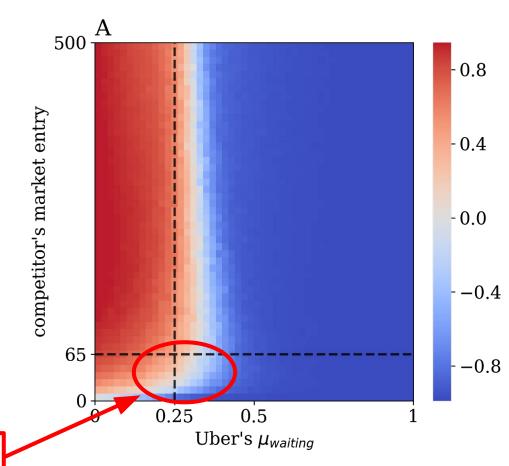
Solid lines are the simulation, markers are the real-world data

- Competitors: Lyft, Via and Juno
- The untapped market: unassigned agents
- The model was fit with the RMSE values to the left of the vertical red line



Alternative worlds

- sensitivity to waiting time is different (X)
- the competitors enter the market at different times (Y)
- Red: Uber has more market share
- Dotted lines: the real world

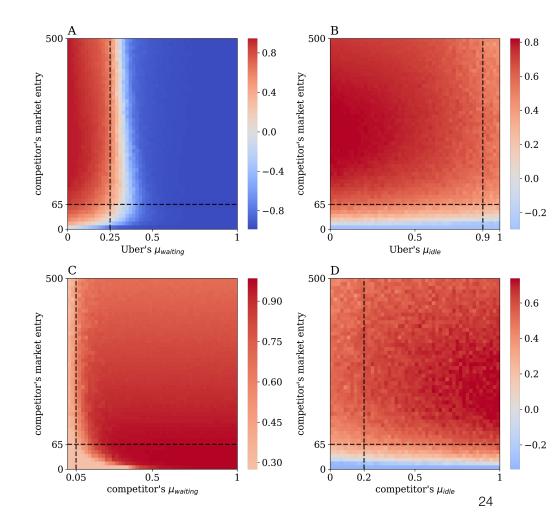


Waiting time can control first movers advantage

Sensitivity analysis

Extending this analysis we can see:

- C: a competitive boundary still exists
- **B, D:** the weakness of μ_{idle}



Future work

Improving rider/driver realism

- ullet Increasing the complexity of the $oldsymbol{\mu}_{ ext{waiting}}$ and $oldsymbol{\mu}_{ ext{idle}}$ terms
 - Suggestion: model them as a distribution that changes over time
- Accounting for agent-level variance
 - \circ Suggestion: sampling from a distributions of $\pmb{\mu}_{ ext{waiting}}$ and $\pmb{\mu}_{ ext{idle}}$



Improving platform realism

• Considering the geography of cities

- Suggestion: model market-shares using an ensemble graph model (e.g. voronoi tessellation)
- Modelling the untapped market more accurately
 - Suggestion: Decrease the number of new agents over time



Studying the impact of price

• Study the impact of price

 Suggestion: add η as a parameter in our model search / sensitivity analysis

• Increase the complexity of the η term

 Suggestion: Sampling from a distribution for each agent, and have that distribution change over time



Key achievements

- Developed a simple graph-based model which captures micro-level intuitions and is capable of reproducing macroscopic behaviour
- ✓ Reproduced market-share growth from Uber and its competitors in NYC
- ✓ Sensitivity analysis based on the parameters that reflect the real-world situation
- ✓ Due to the model's origin, this work may shine a light on universal competitive dynamics (found both in ride-hailing platforms and in the human body)

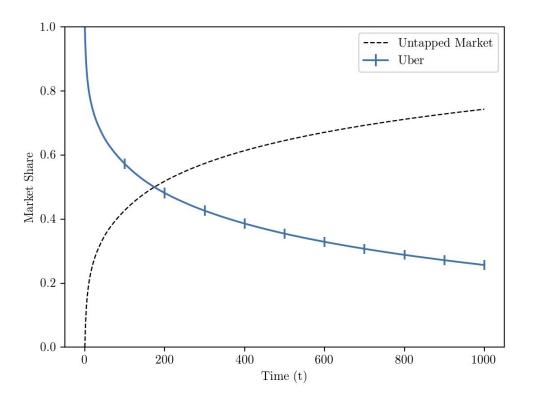
Questions?

Backup

Limits of the model

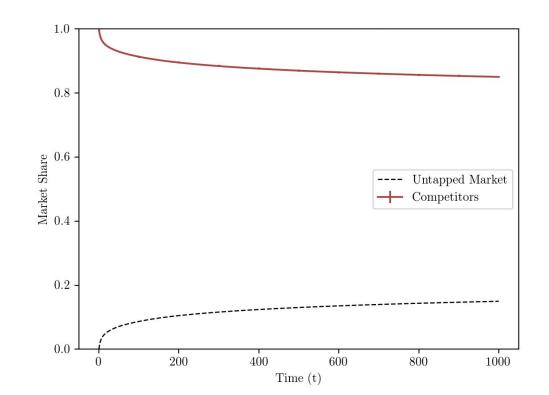
Limits of the simulation in a monopolistic setting

Uber in a monopoly (with NYC parameters)



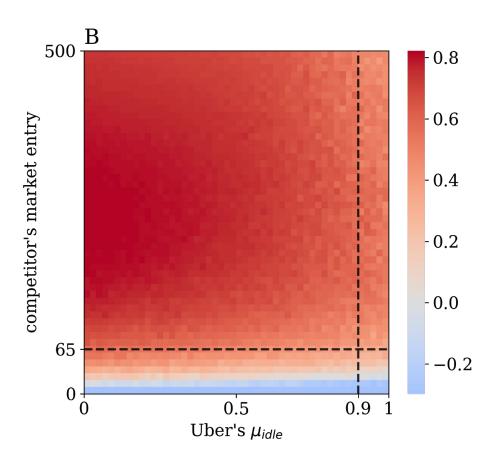
Limits of the simulation in a monopolistic setting

Competitors in a monopoly (with NYC parameters)

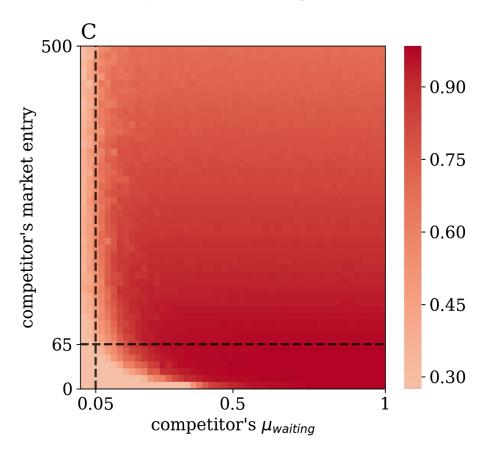


Individual sensitivity analyses

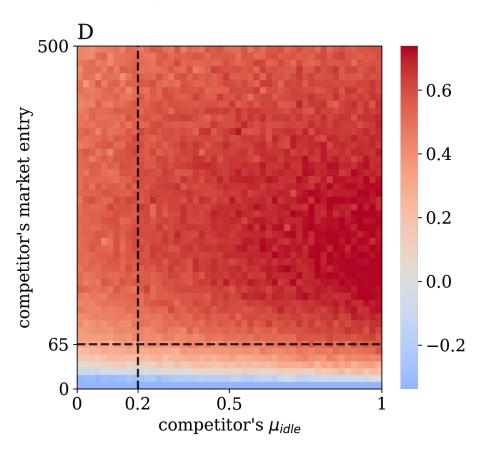
Uber's sensitivity to idle time



Competitor's sensitivity to waiting time



Competitor's sensitivity to idle time



The original cell population model

Population Model of Competition Between Cancer Cells and T Cells

$$R_C(t,u) := \underbrace{(\kappa_C - \mu_C \rho_C(t))}_{\text{cancer cell proliferation and competition for resources}} - \underbrace{\int_V \eta_{\theta_I}(|u-v|) f_I(t,v) dv}_{\text{action of T cells against cancer cells}}$$

$$R_I(t,v) := \underbrace{\left[\int_U \eta_{\theta_E}(|u-v|) f_C(t,u) du + \kappa_P c_P(t)\right]}_{} - \underbrace{\frac{\mu_I}{1 + \mu_M c_M(t)} \rho_I(t)}_{}.$$

clonal expansion and boosting of T cell proliferation

homeostatic regulation and boosting of immune memory (5)