

# An agent-based model of platform competition in the Food Delivery Industry

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# Outline

- 1 Motivation
- 2 Theoretical background
- 3 Model description
- 4 First results and outlook

# Motivation: The rise of the “platform economy”



| Value Rank ▼ | Company              | Market Value | Sector     |
|--------------|----------------------|--------------|------------|
| 1            | Apple                | \$921 bil.   | Technology |
| 2            | Amazon.com           | \$765 bil.   | Retailing  |
| 3            | Alphabet             | \$750 bil.   | Technology |
| 4            | Microsoft            | \$746 bil.   | Technology |
| 5            | Facebook             | \$531 bil.   | Technology |
| 6            | Berkshire Hathaway   | \$492 bil.   | Financials |
| 7            | JPMorgan Chase & Co. | \$388 bil.   | Financials |

As of 5/21/2018

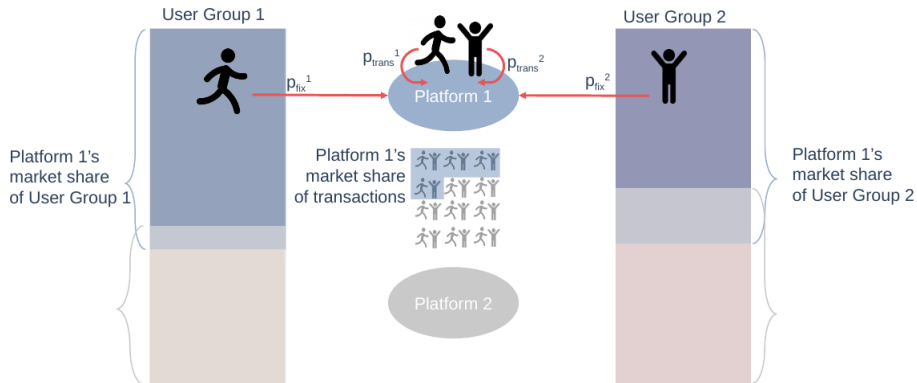
SOURCE: [Fortune 500](#)

## Controversies around “platformization”

- massive data collection
- negative ecological impact (energy, hardware)
- precarious working conditions
- concentration of power in the hands of few platforms

▷ How exactly do these mechanisms of power concentration work?

# Two-sided industries – an economical framework



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► **network effects** as characteristic feature of two-sided industries

# History of theoretical approaches at a glance

**Arthur (1989):** Technology competition under increasing returns

- including network size into utility calculation and adoption decisions with regard to technologies
- different return regimes change adoption processes fundamentally (e.g. with regard to path-dependency)

**Rochet and Tirole (2003):** conceptualization of “two-sidedness”

- platforms as intermediaries between a supply- and a demand-side
- equilibrium model of optimal prices under competition and monopoly
- starting point for further equilibrium models, e.g. Evans and Schmalensee (2008), Armstrong (2010) ...

**Heinrich and Gräbner (2017):** agent-based model of two-sided industries

- allows for studying different decision making rules and more than two platforms

# What determines concentration in digital two-sided industries?

Theoretically known factors which decrease concentration in multi-sided industries:

- weak or asymmetric indirect network effects
- strength of direct network effects
- multi-homing
- market-size
- compatibility of different platforms
- strong innovation dynamics

(Budzinski and Kuchinke (2018), Haucap and Stühmeier (2015))

# Research goal

## Overarching research goal

Development of an empirically validated agent-based model to understand drivers of concentration and diffusion of digital platforms in multisided industries

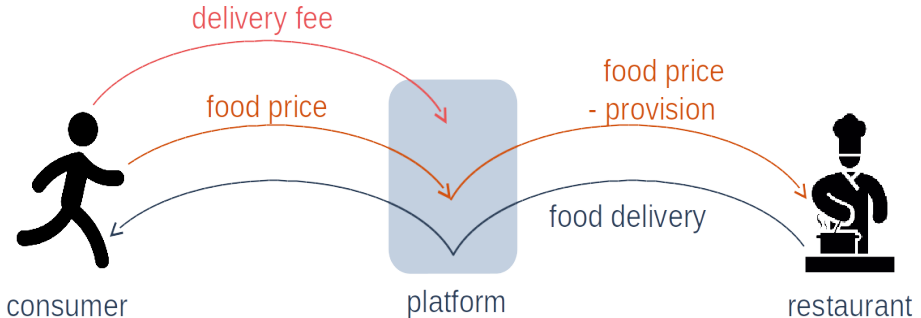
## Subgoal

Development of a simple model reproducing basic features of concentration

# The case of the Food Delivery Industry

## Definition: Food Delivery Platform

- provides consumers with meals from partner restaurants
- offers to handle the delivery process





# Agent types



## Consumers:

- may carry out transactions (order food)
- in case of transaction: decide for a platform to use



## Restaurants:

- decide which platform to sign up with



## Platforms:

- will set restaurants commission and consumers delivery fee
- current state of the model: constant prices

# Parameters and variables

Exogenous simulation parameters:

- $N$  number of time steps
- $n_c$  total number of consumers
- $n_r$  total number of restaurants
- $n_{pf}$  total number of platforms
- $n_{trans}$  number of transactions carried out per time step

Observed variables:

- $n_r^{pf}(t)$  number of restaurants at each platform
- $n_{trans}^{pf}(t)$  number of transactions carried out via each platform
- $n_p^{pf}(t)$  number of platforms

# Simulation steps

In each time step  $t = 1, \dots, N$  following steps are taking place:

- 1 Pick  $n_{trans}$  random consumers
- 2 Iterate over chosen consumers:
  - ▷ each chosen consumer places an order
- 3 Iterate over all restaurants
  - ▷ each restaurants updates its platform memberships
- 4 Update observed variables at time  $t$  for all platforms:
  - $sh_{trans}^{pf}(t) \in [0, 1]$ : share of transactions carried out via  $pf$  at  $t$
  - $n_r^{pf}(t)$ : number of member restaurants of  $pf$
  - $n_{mult}^{pf}(t)$ : number of multihoming restaurants  $pf$

# Consumer iterations

## Parameters of a consumer $c$

$w_c$ : weight of network utility component

- Interpretation: How important is a big choice of restaurants to her?

## Steps of a single transaction (placing an order)

- 1 evaluate expected returns of all platforms
- 2 choose best platform for the transaction (in case of equality: randomly choose one of the best platforms)

## Expected returns of choosing platform $pf$ at time $t$

$$ret_{c,t}(pf) = \underbrace{a_c^{pf} - p_{c,t}^{pf}}_{\text{platform-intrinsic net returns}} + \underbrace{w^c \cdot n_r^{pf}(t)}_{\text{network-based returns}}$$

# Restaurant iterations

## Parameters of a restaurant $r$

- $w_r$ : weight of network utility component
- $c_f^r$ : fixed costs of using a platform
- $ret_{min}^r$ : expected returns threshold to join a platform

## Steps of an updating process

- 1 Evaluate expected returns of membership of all available platforms
- 2 leave all platforms with negative expected returns
- 3 join all platforms with exp. returns exceeding  $ret_{min}^r$

## Expected returns of membership with platform $pl$ at time $t$

$$ret_{r,t}(pf) = \underbrace{a_r^{pf} - c_f^r}_{\text{platform-intrinsic net returns}} + \underbrace{w^r \cdot sh_{trans}^{pf}(t-1) \cdot (1 - p_{t,r}^r)}_{\text{network-based returns}}$$

# Platform iterations

## Parameters of a platform $p_l$

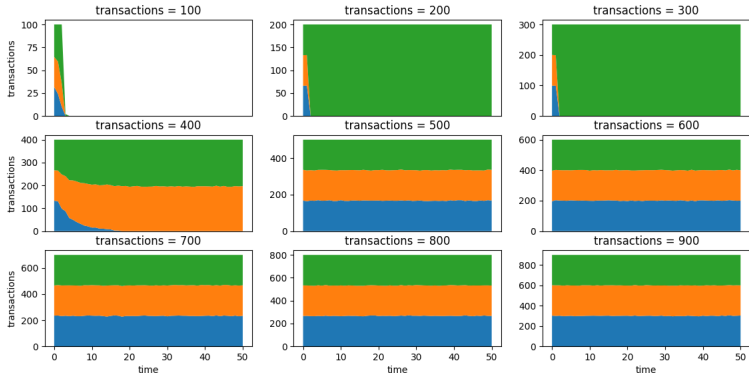
|                           |  |
|---------------------------|--|
| $p_{c,t}^{pf}$            | consumers' transaction fee   |
| $p_{r,t}^{pf} \in [0, 1]$ | restaurants' commission per transaction                                |
| $a_c^{pf}$                | consumers' platform-intrinsic returns<br>(e.g. usability...)           |
| $a_r^{pf}$                | restaurants' platform-intrinsic returns<br>(e.g. increased popularity) |

- current model: no iteration of platforms yet
- intended: price adaptations via learning algorithm

# Preliminary results

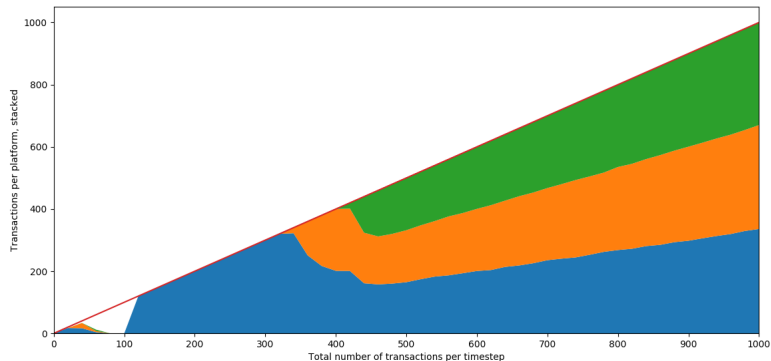
- ▷ number of existing platforms increases with number of transactions

Varying number of transactions, average transaction distribution out of 50 runs, timesteps = 50



# Preliminary results

Varying transaction number, average transaction distribution after 50 timesteps out of 50 runs





# Outlook

## Next steps:

- variable prices, i.e. platforms apply learning strategies to adapt prices
- endogenous network effects, i.e. consumers and restaurants evaluate past transactions via a platform instead of a fixed network returns component in their returns function
- empirical validation: ongoing data collection of Food Delivery platforms in Berlin and Madrid via webscraping