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PASSENGER REACTION IN **RIDE-ON-DEMAND SERVICES** WITH DYNAMIC PRICING

Suiming Guo, Chao Chen, Yaxiao Liu, Ke Xu, Dah Ming Chiu

Suiming Guo, Chao Chen, Yaxiao Liu, Ke Xu, Dah Ming Chiu, Modelling Passengers' Reaction to Dynamic Prices in Ride-on-demand Services: A Search for the Best Fare, Volume 1, Issue 4, 136:1-136:23, IMWUT



The Chinese University of Hong Kong





Outline

data analysis & modelling

Background

dynamic pricing + app-based usage --> new passenger reaction pattern

Data

spatio-temporal distribution of price multipliers

patterns of passenger reaction

Modelling

Numerical Results

What is RoD Service?

- Uber, Lyft,...
- Grab,...
- Didi, Shenzhou UCar, ...
- emerging these years...

What we do?

- real-data, new service, new idea;
- what influence passenger reaction?
- how to model it?

What data?

- Event-log data from Shenzhou UCar.
- Beijing
- Late 2015 to Early 2016.



BACKGROUND

RoD Service: Dynamic Pricing + App-based Usage

Dynamic Pricing Taxi: fixed price RoD: *dynamic* price multiplier

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The figures "showing price multiplier" and "showing street hailing" are from Google search, and we don't retain any copyright.

App-based

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Erom:		XXX Train Station
🔞 То:	х	XX Shopping Mall
When to ride?		Now
Coupon		N/A
Sedan ⊘	MPV	Luxury Car
Current price multiplier: 1.0x		
Est. Fare: ¥33 🗓	Othe	r available promotions
Dide a Carl		

instead of



New Passenger Reaction Patterns



Estimate the fare for multiple times before giving up or taking a ride

Motivation:

1. Quantify passenger perception & dissipate concerns from outside;

2. Make better and more responsive pricing algorithms.

DATA ANALYSIS

Spatial-temporal Distribution of Price Multipliers



low (1.0x), *avg* (1.2x) and *high* (1.4 & 1.6x) price multipliers

• clear time-of-day pattern:

- -- transportation area: more stable
- -- business/residential: going-to-work or back-home patterns.

The distribution varies according to the price multiplier, time, location, etc.

DATA ANALYSIS

Patterns of Passenger Reaction to Dynamic Prices

■ ⊖ E, 🔶 C, 🧎 ⊖ "nothing happens". (a) Freqs of (EC) and (E) in Business Area (b) Freqs of (EC) and (E) in Residential Area 50% 40% 30% 20% 10% 0% 0:00 , Jio Hio Cio Sio 10:00 13:00 14:00 18:00 Join Jio 2.00 8:00 10:00 12:00 14:00 16:00 20:00 D:00 6.00 ×.00 0.00 (d) Freqs of (EEC) and (EE) in Business Area (e) Freqs of (EEC) and (EE) in Residential Area 12% 9% 6% 3% 0% 6:00 e:00 10:00 14:00 0:00 6.00 12:00 14:00 16:00 18:00 ×:00 2:00 0.00 10:00 2:00 10:00 18:00 20:00 Dig 20:00 2:00 2:00 e.60 0:00 0.00

Most don't hesitate long (EC and E are the most common patterns)

business area:

Most eager to get a car, & highest demand transportation area: Not that eager, and can wait longer



business/residential area:

During rush hour, more are inclined to estimate once and give up -- too high price.



Distribution of Price Multipliers + Patterns of Passenger Reaction

Both the *dynamic prices* & *passenger demand elasticity* influence their reaction in RoD service.

This inspires our modelling.



Modelling

Technical details referred to the paper, not shown here.

Distribution of Price Multipliers Passenger Demand Elasticity / Search Cost Passenger Reaction

Any two decide the other one

a search = estimate fare for once search cost = passenger demand elasticity

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A SEARCH THEORY MODEL

- buying or giving up.

- Resulting stopping strategy:
- take the ride.
- multiplier threshold.
- What can the model be used for?
- the distribution of price multipliers.

• Just like buying things at stores: comparing prices, repeating many times,

• We adopt and extend a search theory model to model passenger reaction. • Basically, "every search incurs a search cost; and as long as the expected price decrease is smaller than the search cost, you should stop."

• (a) If the search cost is large enough, you should "only search once and

• (b) Otherwise, stop searching as long as you get a price multiplier <= a

• (a) *learning the search cost*: to understand passenger demand elasticity; • (b) predicting passenger reaction: based on the learned search cost and

SOME NUMERICAL RESULTS

based on our data and the model



Fig. 14. The search cost: leaving busi- Fig. 15. The search cost: leaving resi- Fig. 16. The search cost: leaving trans-ness area on weekdays.dential area on weekdays.portation area on weekdays.

Hours-of-the-day

Hours-of-the-day

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Hours-of-the-day

Some quick observations:

• peaks-and-troughs roughly agree that of high price multipliers, but not the same -- joint result of both price multiplier distribution and search cost.

• time-of-day variation of multiplier threshold can be a useful guidance for both passengers and the service provider.

• search cost v.s. multiplier threshold: more accurate, with finer granularity

information obtained is different. Example: when the multiplier threshold is the same (e.g., 1.4x), the search cost is significantly lower in residential area.

• can infer the relationship between the search cost and trip intention.

Thank you for your interest!

l'm from

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