

# Market Your Venue with Mobile Applications

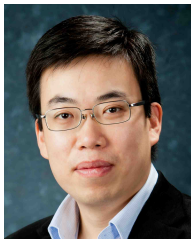
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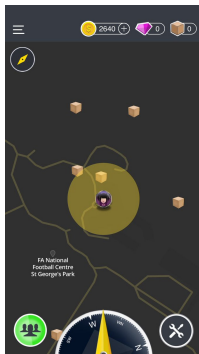
# Mobile Apps with Augmented Reality Features



Ingress



Pokemon Go



Snatch

Other examples include Red Envelope Game, Snapchat's Geo-filter, Yinyangshi, etc.

- (i) Apps label real-world locations as **places of interest (POIs)**.
- (ii) After physically visiting the locations, users can win items (e.g., Pokemons and coupons) in the apps.

# Apps Benefit Venues

If venues (e.g., restaurants) are labeled as *POIs*, they can attract more visitors, which potentially increases the venues' sales.

## Pokemon Go Lures Customers to Small Businesses

July 2



Pokémon Go and the Future of Marketing with Augmented Reality

Mo

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## Pokémon Go Is Good Business for Small Businesses

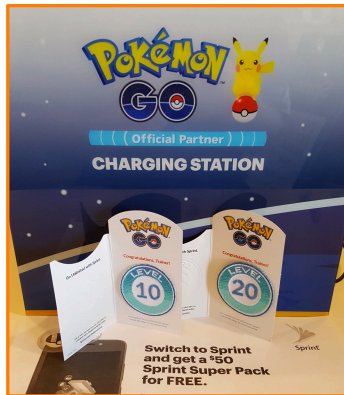
Kerry Close  
Jul 12, 2016



Pokémon Go users gotta catch 'em all—and small businesses are taking advantage of their obsession with the viral smartphone game.

## Venues Benefit Apps

If users play the apps at venues with good **infrastructure** (e.g., charging stations and Wi-Fi networks), they will have enhanced app experience.



free smartphone charging stations  
@Sprint



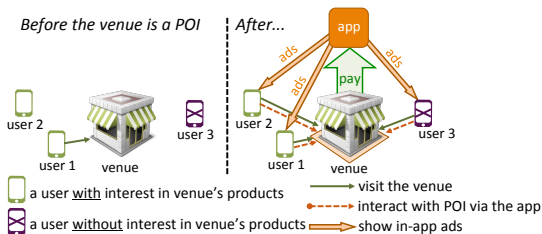
free Wi-Fi networks  
@Starbucks

# POI-Based Collaborations of Apps and Venues

- Practical examples
  - ▶ **Pokemon Go** collaborated with 3,000 **McDonald's** restaurants in Japan, and 10,500 **Sprint** stores and 12,800 **Starbucks** locations in the U.S.;
  - ▶ **Yinyangshi** collaborated with 5,000 **KFC** and 1,700 **Pizza Hut** in China;
  - ▶ **Snapchat** offered specialized “geo-filters” for **Wendy's** in the U.S.;
  - ▶ **Snatch** labeled locations of **Mitchells & Butlers** and **Topshop** in the U.K. as “safehouses”.
- AR/VR market's worldwide revenue might exceed \$162 billion in 2020, so POI-based collaboration could create substantial revenues.

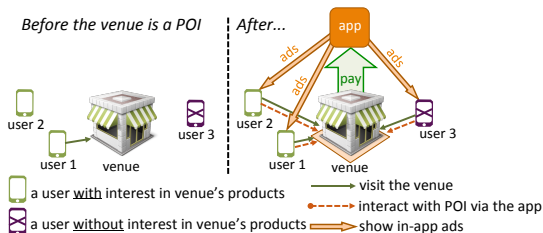
# Illustration of POI-Based Collaboration

- Consider an app and a store/restaurant chain's representative venue.



- ▶ After becoming a POI, venue's investment in the app-related infrastructure affects the number of visitors.
- **Misaligned interests**
  - ▶ App: benefits from both **green** & **purple** users (as they interact with POI).
  - ▶ Venue: gains profits only from **green** users (with interests in its products).
  - ▶ Restrict venue's willingness to invest and also the created revenue.

# Problem Description



- Current practices of app's tariffs
  - ▶ **Lump-sum-only tariff:** based on a lump-sum fee
    - ★ Example: Snapchat.
  - ▶ **Per-player-only tariff:** based on number of users interacting with POIs
    - ★ Example: Pokemon Go charges a venue \$0.5 per player.
- **Question:** Can these tariffs solve the “misaligned interests” problem?  
**Our answer:** No. We design an optimal **two-part tariff**, which incentivizes the highest venue's investment and creates the highest revenue compared with the two state-of-the-art tariffs.

# Related Work

- Cooperation between online and offline businesses
  - ▶ There are very few papers in this area.
  - ▶ POI-based collaboration: there are only empirical studies ([V. Pamuru *et al.* 2017], [A. Colley *et al.* 2017]), and we provide **first analytical study**.
- Competition between online and offline businesses
  - ▶ Empirical study on users' choices between online & offline businesses
    - ★ [A. Goolsbee 2000], [JT. Prince 2007], [C. Forman *et al.* 2009]
  - ▶ Analytical study on price competition between online & offline businesses
    - ★ [S. Balasubramanian 1998], [X. Pan *et al.* 2002], [S. Viswanathan 2005]
  - ▶ The studied online & offline businesses sell the **same** type of products.



# App's Two-Part Tariff $(l, p)$

- We assume that the app is free to all users.
- The app announces a two-part tariff  $(l, p)$  to venue:
  - ▶  $l \in \mathbb{R}$ : lump-sum fee,  $p \in \mathbb{R}$ : per-player charge.
  - ▶ When the venue becomes a POI, it pays:  
 $l + p \times \text{number of users interacting with the POI}$ .
  - ▶ Note that the app can incentivize the venue with negative  $l, p$ .

# Venue's POI Decision $r$ and Investment Decision $I$

- Venue's choices
  - ▶ POI decision  $r \in \{0, 1\}$ :  $r = 1$  if and only if it becomes a POI;
  - ▶ Investment level  $I \geq 0$  on the app-related infrastructure.
- We use parameter  $I_0$  to denote the **initial investment level**, and call  $I + I_0$  as the **total investment level**.

## A Type- $(\omega, c)$ User's Decision $d$

We consider a continuum of users who use the app, and denote the population size by  $N$ . Each user is described by two attributes:

- $\omega \in \{0, 1\}$  captures a user's intrinsic interest in venue's products.
  - ▶ We assume that  $\eta N$  users have  $\omega = 1$ , and  $(1 - \eta) N$  users have  $\omega = 0$ .
- $c \in [0, c_{\max}]$  captures a user's transportation cost to visit the venue.
  - ▶ We assume that  $c$  uniformly takes a value from  $[0, c_{\max}]$ .

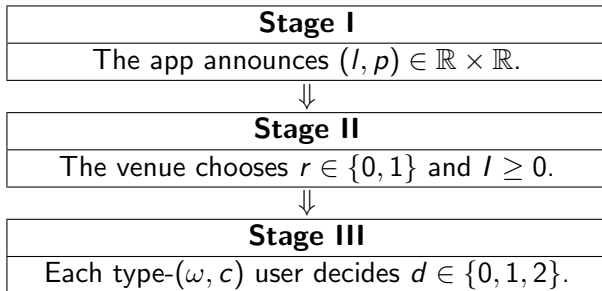
## A Type- $(\omega, c)$ User's Decision $d$

- Denote a user's decision by  $d \in \{0, 1, 2\}$ :
  - ▶  $d = 0$ : do not visit the venue;
  - ▶  $d = 1$ : visit the venue but do not interact with the POI;
  - ▶  $d = 2$ : visit the venue and interact with the POI.
- A type- $(\omega, c)$  user's payoff under the venue's choices  $r$  and  $l$  is

$$\Pi^{\text{user}}(\omega, c, d, r, l) = \begin{cases} 0, & \text{if } d = 0, \\ U\omega - c, & \text{if } d = 1, \\ U\omega - c + V \underbrace{+ \theta \bar{y}(r, l) N}_{\text{network effect}} - \underbrace{\frac{\delta}{l + l_0} \bar{y}(r, l) N}_{\text{congestion effect}}, & \text{if } d = 2. \end{cases}$$

- ▶  $U > 0$ : utility of a user with  $\omega = 1$  when it consumes venue's products;
- ▶  $V > 0$ : a user's base utility of interacting with the POI;
- ▶  $\theta \geq 0$ : network effect factor;
- ▶  $\bar{y}(r, l) \in [0, 1]$ : the fraction of users interacting with the POI, given the venue's choices  $r$  and  $l$  (depend on all users' equilibrium decisions);
- ▶  $\delta > 0$ : congestion effect factor.

# Three-Stage Game



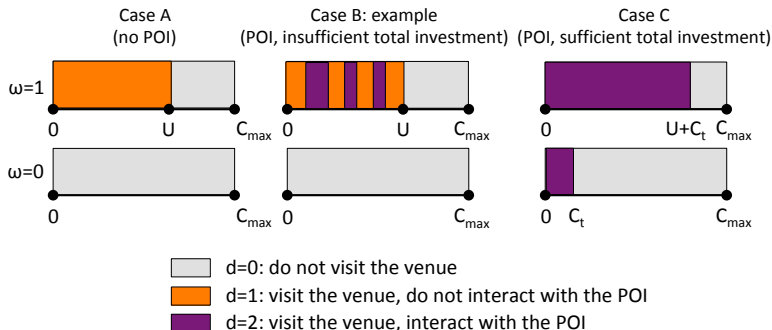
# Users' Equilibrium at Stage III

A type- $(\omega, c)$  user decides  $d^*$  by solving

$$\begin{aligned} & \max \Pi^{\text{user}}(\omega, c, d, r, l) \\ & \text{var. } d \in \begin{cases} \{0, 1\}, & \text{if } r = 0, \\ \{0, 1, 2\}, & \text{if } r = 1. \end{cases} \end{aligned}$$

# Users' Equilibrium at Stage III

Under venue's POI and investment decisions, we have three possible cases.



## Venue's Equilibrium Decisions at Stage II

The venue makes the POI choice  $r^*$  and investment choice  $l^*$  by solving

$$\max \Pi^{\text{venue}}(r, l, l, p) \triangleq \underbrace{bN\bar{x}(r, l)}_{\text{profit from sales}} - \underbrace{kl}_{\text{investment cost}} - \underbrace{r(l + pN\bar{y}(r, l))}_{\text{payment}}$$

var.  $r \in \{0, 1\}$ ,  $l \geq 0$ .

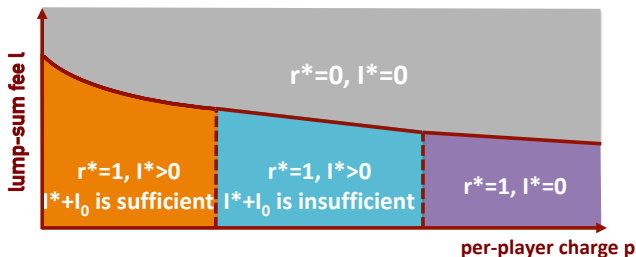
- $b > 0$ : the venue's profit due to one user's consumption of products;
- $\bar{x}(r, l) \in [0, 1]$ : fraction of users that have  $\omega = 1$  and visit the venue under  $r$  and  $l$  (depend on users' equilibrium decisions at Stage III);
- $k > 0$ : unit investment cost.



## Venue's Equilibrium Decisions at Stage II

Based on initial investment level  $I_0$  and congestion effect factor  $\delta$ , we have three situations:

- Small  $I_0$  and large  $\delta$  (only illustrate this situation below);
- Small  $I_0$  and small  $\delta$ ;
- Large  $\delta$ .



(We analytically characterize all boundaries in the paper.)

# App's Optimal Tariff at Stage I

The app determines  $(l^*, p^*)$  by solving

$$\max R^{\text{app}}(l, p) \triangleq \underbrace{r^*(l, p) \left( l + pN\bar{y}(r^*(l, p), l^*(l, p)) \right)}_{\text{venue's payment}} + \underbrace{\phi N\bar{y}(r^*(l, p), l^*(l, p))}_{\text{advertising revenue}}$$

$$\text{var. } l, p \in \mathbb{R}.$$

Here,  $\phi \geq 0$  is the **unit advertising revenue**, representing the app's advertising revenue because of a user's interaction with the POI.

# App's Optimal Tariff at Stage I

## App's optimal two-part tariff

- (i) Per-player charge  $p^* = -\phi \leq 0$  ( $\phi$  is app's unit ad revenue);
- (ii) Lump-sum fee  $l^* \geq 0$  is the maximum lump-sum fee under which venue becomes a POI, given  $p^* = -\phi$  (*concrete expression is given in our paper*).

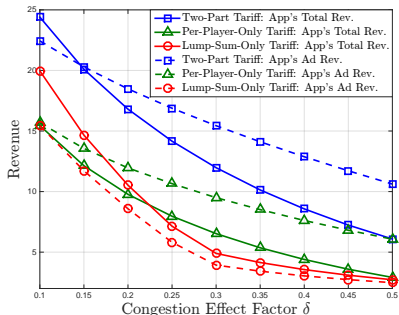
- Reason

- ▶ When  $p^* = -\phi$ , the venue's investment level in response to  $p^*$  will maximize the summation of the app's revenue and venue's payoff;
- ▶ App chooses  $l^*$  to extract all the venue's surplus.

- Practical insight: charge-with-subsidy scheme

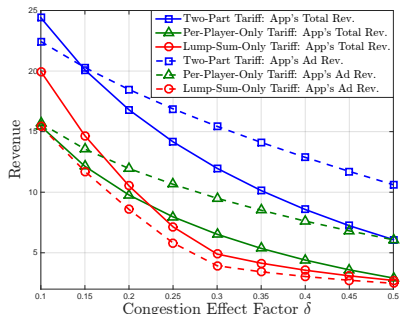
- ▶ In order to be a POI, the venue needs to first pay  $l^*$ ;
- ▶ Every time a user interacts with the POI, the app pays the venue  $\phi$ .

# Comparison with State-of-The-Art Schemes



- Our tariff always achieves the highest app's total revenue.
  - ▶ Can prove it is true even compared with a general class of tariffs.
- Our tariff always achieves the highest app's ad revenue.
  - ▶ This implies highest investment and highest number of interactions.

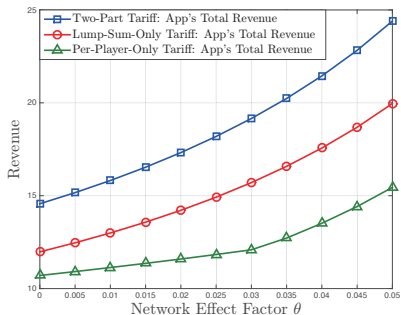
# Comparison with State-of-The-Art Schemes



## Comparison with lump-sum-only tariff

- Our tariff's performance improvement is most obvious at **medium  $\delta$** .
- Our tariff's strength: subsidize the venue to incentivize investment, which relieves the congestion.

# Comparison with State-of-The-Art Schemes



## Comparison with per-player-only tariff

- Our tariff's performance improvement is most obvious at **large  $\theta$** .
- Our tariff's strength: extract high venue's payment via lump-sum fee.

# Conclusion and Future Direction

## ● Conclusion

- ▶ Model the emerging POI-based collaboration by a three-stage game.
- ▶ Design an optimal two-part tariff to realize its full business potential.

## ● Our other results

- ▶ Survey venues' influences on 103 Pokemon Go players' experience.
- ▶ Study implementation of optimal two-part tariff under uncertainty.
- ▶ Analyze which type of venues is the best choice to collaborate.
  - ★ Counter-intuitive insights, e.g., a **bandwidth-consuming** app should collaborate with a **low-quality** venue, rather than a **high-quality** venue.

## ● Future direction

- ▶ Consider heterogeneity of users' sensitivities to the network effect and congestion effect.
- ▶ Investigate the competition among multiple venues in becoming POIs.

# *THANK YOU*



Network Communications and Economics Lab

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