

# Economics of Public Wi-Fi Monetization and Advertising

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# Public Wi-Fi is everywhere



# Background

Venues: largest public Wi-Fi providers

- Top 3: Retailers > Cafes & Restaurants > Hotels

Venues	Retails	Cafes & Restaurants	Hotels	Others	Total
Wi-Fi hotspots	5,763,907	4,259,351	397,905	1,808,234	12,229,397

Predicted Wi-Fi Ownership by Venue Type, 2018 ©WBA

- Reasons to provide Wi-Fi
  - ▶ Enhance customers' experiences
  - ▶ Provide location-based services (e.g., navigation, billing, social interaction)
- Question: It is costly to deploy and operate the public Wi-Fi networks. How do venues generate revenue?

# First Approach: Ad Sponsored Wi-Fi Access

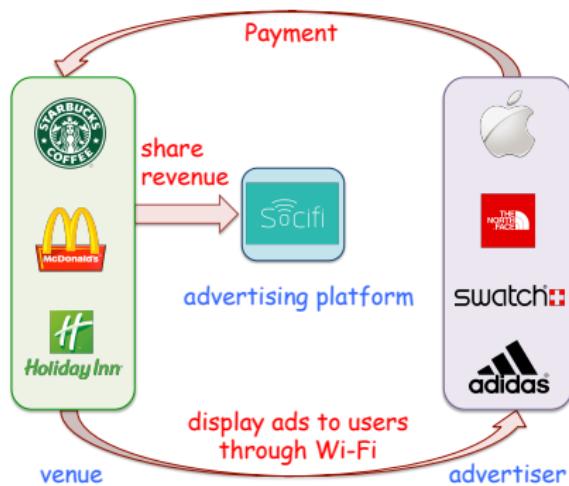
Users watches an ad (e.g., 30sec) and then connect Wi-Fi for a certain period (e.g., 30min)



play sponsored video (*advertisement*) and connect

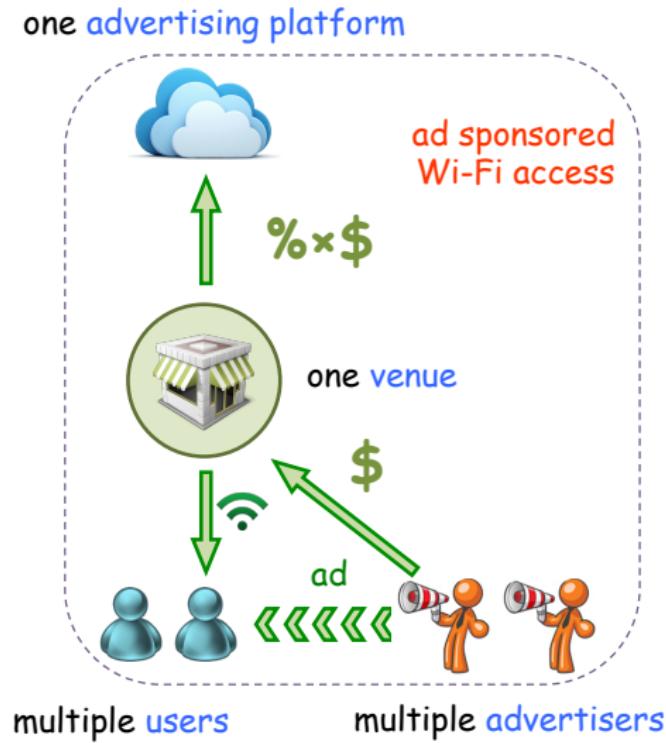
# First Approach: Ad Sponsored Wi-Fi Access

- Advertising platform (e.g., SOCIFI)  
organizes a two-sided market between venues and advertisers
- Example:
  - ▶ Starbucks (venue) displays Apple's (advertiser) ad to users in Wi-Fi;
    - ★ User watches the ad, and uses Wi-Fi for 30min for free
  - ▶ Apple pays Starbucks based on the ad display times;
  - ▶ Starbucks shares 30% revenue with SOCIFI (advertising platform)



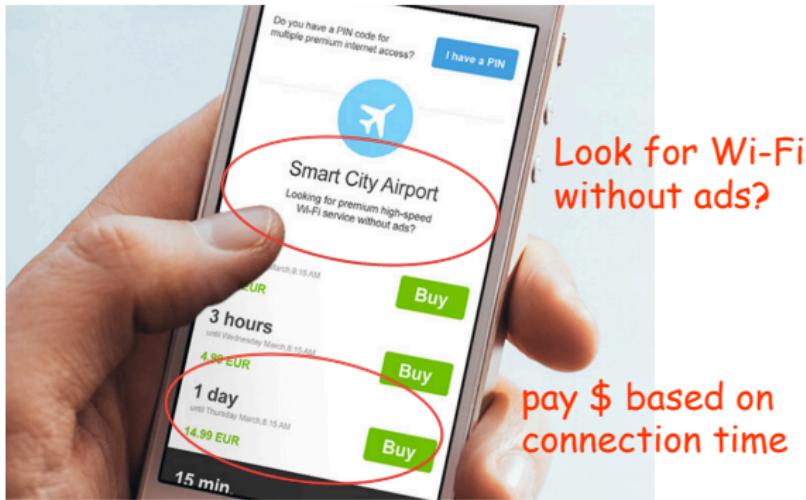
# First Approach: Ad Sponsored Wi-Fi Access

Illustration of ad sponsored Wi-Fi access

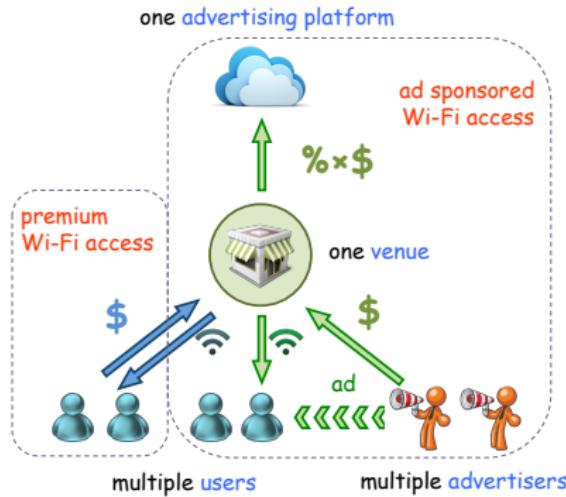


## Second Approach: Premium Wi-Fi Access

Users directly pay the [venue](#) to use Wi-Fi



# Public Wi-Fi Monetization Ecosystem



Understand each decision maker's **optimal** behavior

- **Advertising platform:** What is the ad revenue sharing proportion?
- **Venue:**
  - ▶ How much to charge **advertisers** for displaying ads?
  - ▶ How much to charge **users** for premium access?
- **Each advertiser:** How many ads to display at the **venue**?
- **Each user:** Which Wi-Fi access type to choose?

## System Model: Advertising Platform

Ad revenue sharing ratio  $\delta \in [0, 1]$ : the fraction of the ad revenue the venue needs to transfer to advertising platform.

# System Model: Venue

Two decision variables:

- Wi-Fi price  $p_f$  (premium access): if a user chooses premium access, venue charges the user  $p_f$  per session;
- Advertising price  $p_a$  (ad sponsored access): if a user chooses ad sponsored access, venue charges the corresponding advertisers  $p_a$  per displayed ad.

For example,

1 session=30 min

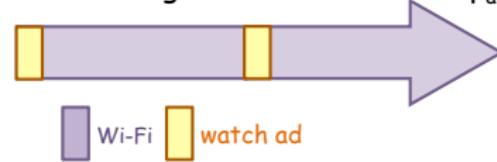
this user demands 1h (=2 sessions)



(a) if premium access  
venue charges this user  $2p_f$



(b) if ad sponsored access  
venue charges the advertisers  $2p_a$



## System Model: Users

- Consider  $N$  users, and each user's type  $\theta \sim U[0, \theta_{\max}]$  describes its valuation for Wi-Fi access
- A user's access choice  $d \in \{0, 1\}$ :
  - ▶  $d = 0$  denotes the ad sponsored access;
  - ▶  $d = 1$  denotes the premium access.
- A type- $\theta$  user's payoff in one session is:

$$\Pi^{\text{user}}(\theta, d) = \begin{cases} \theta(1 - \beta), & \text{if } d = 0 \text{ (ad sponsored access)}, \\ \theta - p_f, & \text{if } d = 1 \text{ (premium access)}, \end{cases} \quad (1)$$

where  $\beta < 1$  captures the inconvenience of watching ads.

- The number of sessions that a user demands within the considered time period (e.g., one week) follows the Poisson distribution with parameter  $\lambda > 0$ . Parameter  $\lambda$  describes users' visiting frequency at the venue.

## System Model: Advertisers

- Consider  $M$  advertisers, and each advertiser's type  $\sigma \sim U[0, \sigma_{\max}]$  describes its popularity (the popularity **decreases** with  $\sigma$ ).
- An advertiser's strategy  $m \geq 0$ : number of ads to display at the venue
- A type- $\sigma$  advertiser's payoff

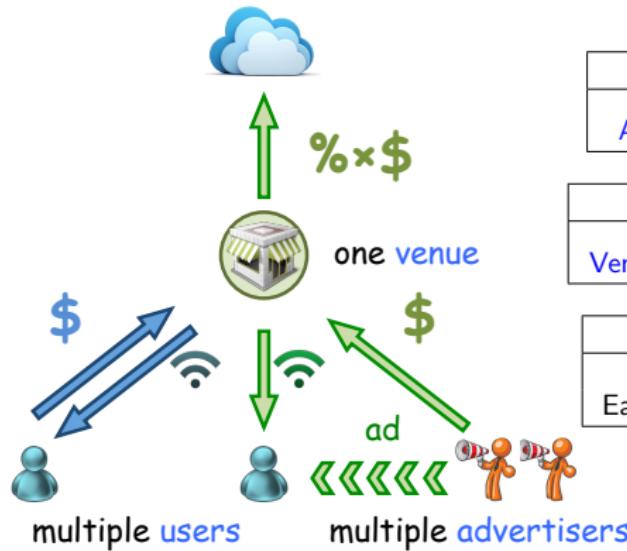
$$\Pi^{\text{advertiser}}(\sigma, m) = \underbrace{as(\sigma)N\varphi_a(p_f)\left(1 - e^{-\frac{m}{N\varphi_a(p_f)}}\right)}_{\text{Utility}} - \underbrace{p_a m}_{\text{Payment}} \quad (2)$$

- ▶  $a$ : unit profit of showing the ad to a targeted user
- ▶  $s(\sigma)$ : popularity of the advertiser (decrease with  $\sigma$ )
- ▶  $N\varphi_a(p_f)$ : number of users choosing the ad sponsored access
- ▶  $\left(1 - e^{-\frac{m}{N\varphi_a(p_f)}}\right)$ : probability for a user to see the advertiser's ad (obtained via computation), and is **concavely increasing** in  $m$
- ▶  $p_a$ : advertising price (set by the venue)

# Three-Stage Stackelberg Game

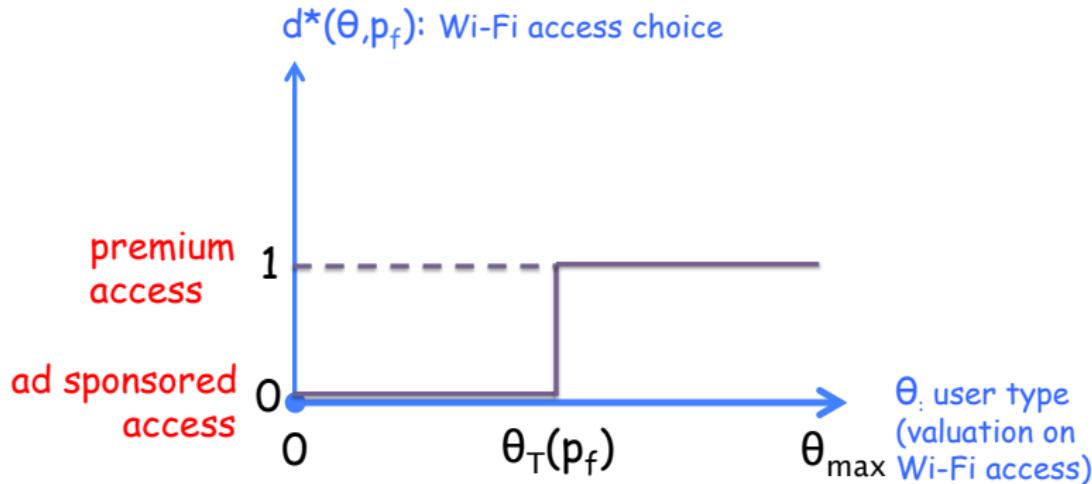
Solution: backward induction (Stage III→Stage II→Stage I)

one advertising platform



## Stage III: Users' Optimal Access Choices

- Users' threshold policy:  
If  $\theta < \theta_T(p_f)$ , use ad sponsored access;  
If  $\theta \geq \theta_T(p_f)$ , use premium access.
- Threshold  $\theta_T(p_f)$  is non-decreasing in  $p_f$

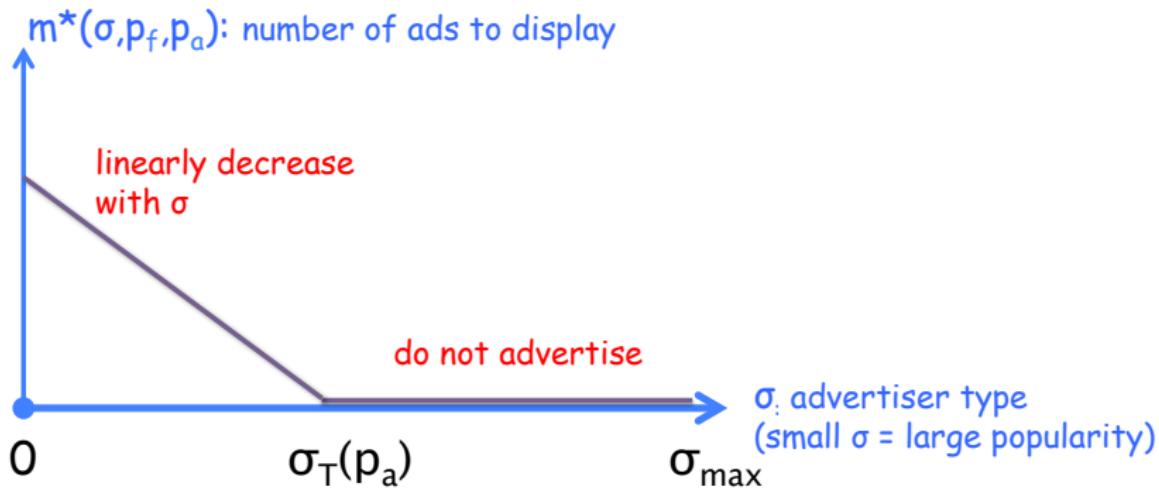


## Stage III: Advertisers' Optimal Advertising

- Advertisers' threshold policy:

If  $0 \leq \sigma \leq \sigma_T(p_a)$ , advertise, and number of ads decreases with  $\sigma$ ;

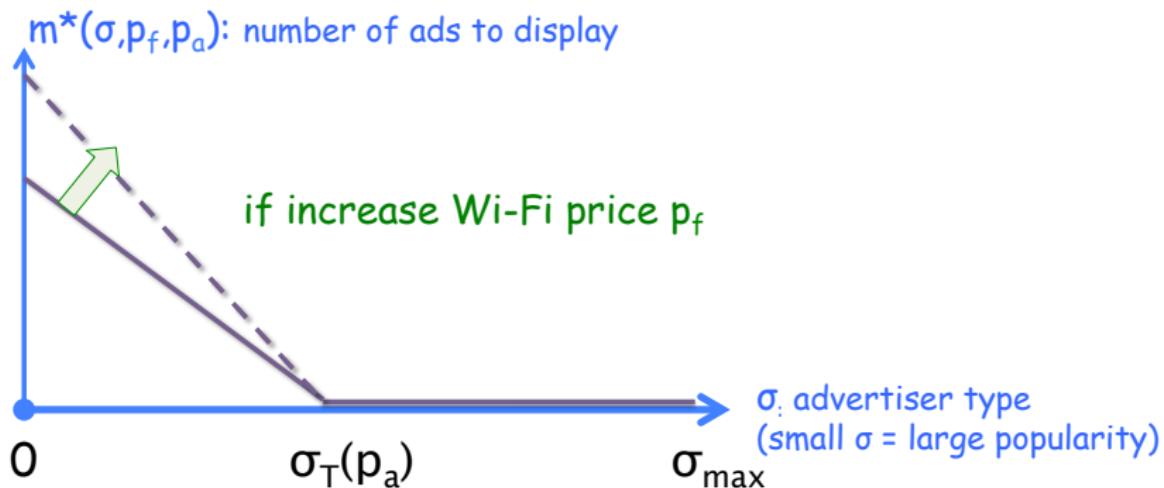
If  $\sigma_T(p_a) < \sigma \leq \sigma_{\max}$ , do not advertise



## Stage III: Advertisers' Optimal Advertising

If Wi-Fi price  $p_f$  increases (more users choose the sponsored access),

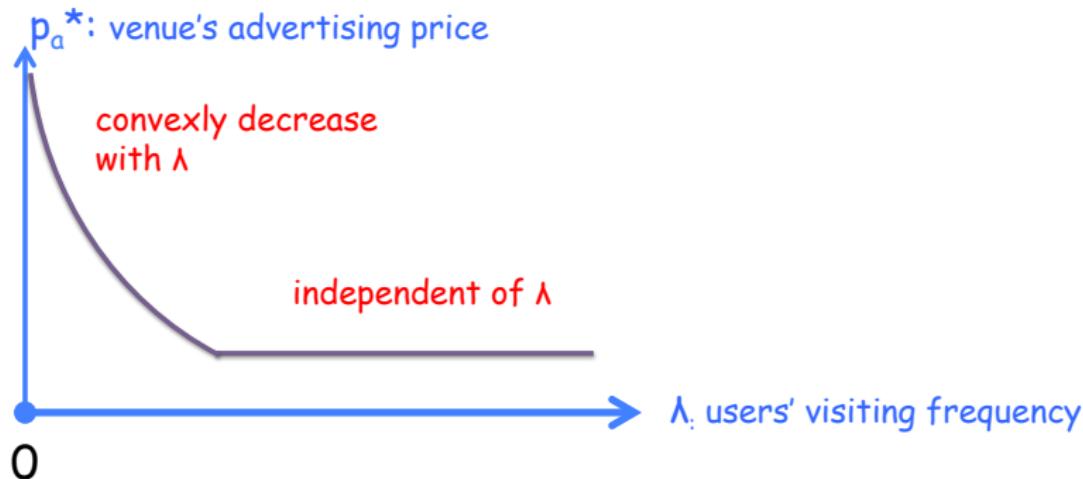
- threshold  $\sigma_T(p_a)$  does not change: the number of advertiser types that need to advertise does not change
- slope increases: the advertisers who originally advertise should display more ads



## Stage II: Venue's Optimal Advertising Price

Venue's optimal advertising price  $p_a^*$  (limit case  $M \rightarrow \infty$  and  $\sigma_{\max} \rightarrow \infty$ )

- (1)  $p_a^*$  is independent of the advertising platform's sharing ratio  $\delta$ ;
- (2)  $p_a^*$  decreases with  $\lambda$  for small  $\lambda$  region (reason: limited ad spaces);
- (3)  $p_a^*$  is independent of  $\lambda$  for large  $\lambda$  region (reason: enough ad spaces).



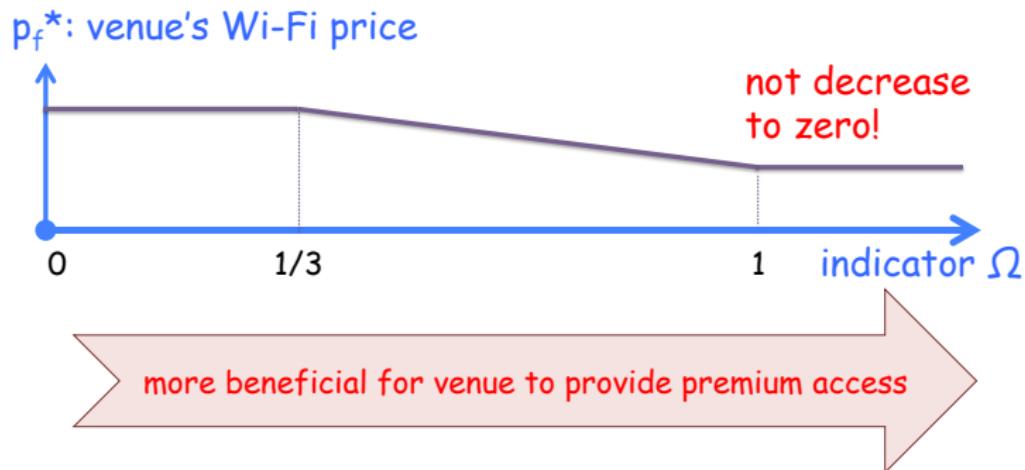
## Stage II: Venue's Optimal Wi-Fi Price

Define indicator  $\Omega \triangleq \frac{\lambda\beta\theta_{\max}}{ag(\lambda,\gamma,\eta)}$

- Parameters' meanings
  - ▶  $\lambda$ : users' visiting frequency
  - ▶  $\beta$ : users' payoff reduction due to watching ads
  - ▶  $\theta_{\max}$ : users' maximum valuation on Wi-Fi access
  - ▶  $a$ : unit profit for an advertiser of showing the ad to a targeted user
  - ▶  $\gamma$ : the venue's advertising concentration level
  - ▶  $\eta$ : the expected number of advertisers that a user likes
- Intuition: a large  $\Omega$  implies that the venue can earn more revenue by providing the premium access comparing to the ad sponsored access.

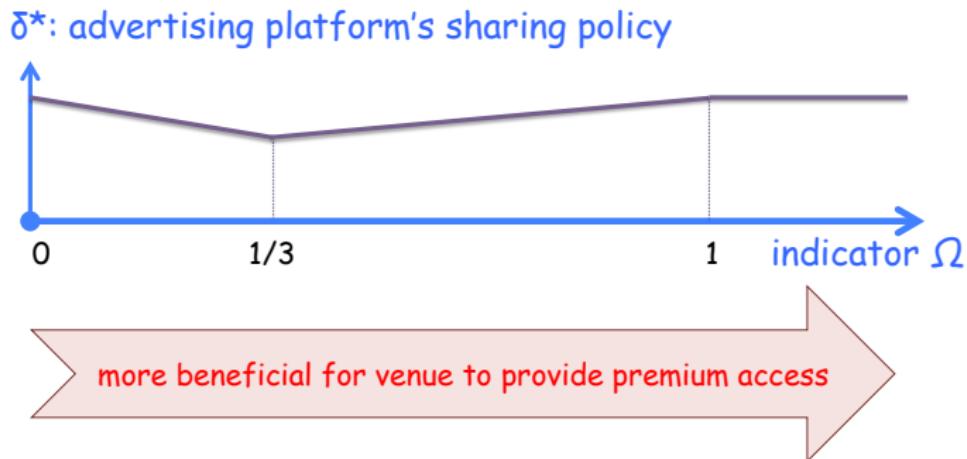
## Stage II: Venue's Optimal Wi-Fi Price

- **Indicator**  $\Omega$ : a large  $\Omega$  implies that the venue can earn **more** revenue by providing premium access comparing to ad sponsored access
- Wi-Fi price  $p_f^*$  is non-increasing in  $\Omega$



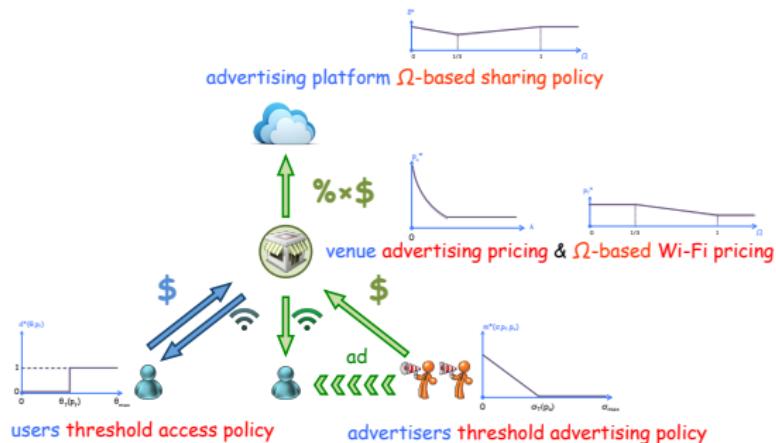
## Stage I: Advertising Platform's Optimal Sharing Policy

- **Indicator  $\Omega$ :** a large  $\Omega$  implies that the **venue** can earn **more** revenue by providing premium access comparing to ad sponsored access
- **Sharing ratio  $\delta^*$ :**
  - ▶ first decreases with  $\Omega$ : attract **venue** to provide ad sponsored access;
  - ▶ second increases with  $\Omega$ : directly extract more ad revenue from **venue**



# Conclusion and Future Work

- Public Wi-Fi monetization problem
  - ▶ Five threshold strategies for decision makers



- Future work
  - ▶ QoS differentiation (e.g., premium access with QoS guarantee)
  - ▶ Influence of Wi-Fi capacity

# THANK YOU



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## Stage III: ADs' Advertising

### AD's optimal advertising strategy

A type- $\sigma$  AD's optimal advertising strategy is

$$m^*(\sigma, p_f, p_a) = \begin{cases} N\varphi_a(p_f)\left(\ln\left(\frac{a\gamma}{p_a}\right) - \gamma\sigma\right), & \text{if } 0 \leq \sigma \leq \sigma_T(p_a), \\ 0, & \text{if } \sigma_T(p_a) < \sigma \leq \sigma_{\max}, \end{cases} \quad (3)$$

where  $\sigma_T(p_a) \triangleq \min\left\{\frac{1}{\gamma} \ln\left(\frac{a\gamma}{p_a}\right), \sigma_{\max}\right\}$  is the *threshold AD type*.

- (1)  $\sigma_T(p_a)$  and  $m^*(\sigma, p_f, p_a)$  decrease with  $p_a$ ;
- (2)  $m^*(\sigma, p_f, p_a)$  decreases with type  $\sigma$ .

## Stage III: Advertisers' Optimal Advertising

Threshold  $\sigma_T(p_a)$  is non-increasing in advertising price  $p_a$

