

# Optimized Travel Menus with a Flexible Mobility on Demand System

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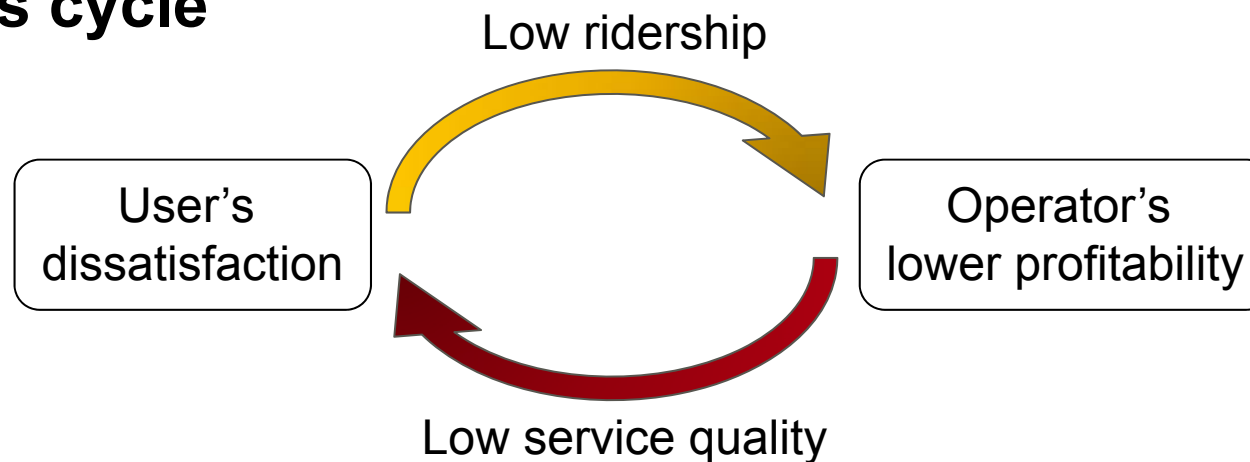
CASPT

- Motivation and background
  
- Flexible Mobility on Demand (FMOD) system
  - Concepts
  - Modeling Framework
  
- Simulation experiments
  - Myopic model v.s. Look-ahead model
  
- Summary and future directions

- **Conventional public transportation services** are not personalized.
  - Fixed route, Fixed schedule, Low frequency etc.
- Most people cannot afford to use **taxi service** on a daily basis.



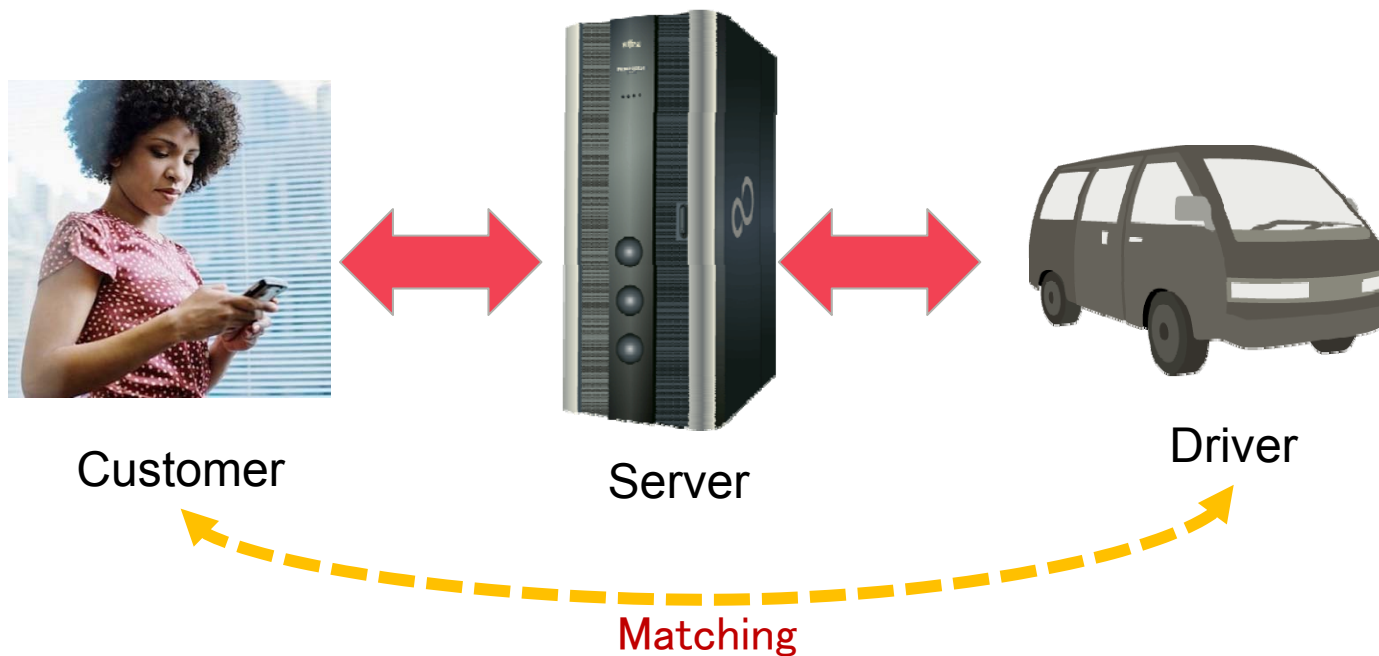
## ■ **Vicious cycle**



- ICT has a potential to break the vicious cycle?

# Motivation and background

- Personalized transportation services using mobile apps are emerging
  - Uber, Lyft, GrabTaxi, etc.



- Why not apply similar technologies to **public transportation services**?
  - DRT, fixed route bus etc.

# Problem

- “How to increase operator profit and user satisfaction? ”
- **Flexibility** to demand fluctuations is necessary.
- Currently, due to lack of the flexibility:
  - **Off-peak:**  
=> **Drivers** cannot find passengers
  - **On-peak:**  
=> **Passengers** cannot find drivers.



Some passengers may give up taking public transportation.  
=> Operator lose revenue opportunity.



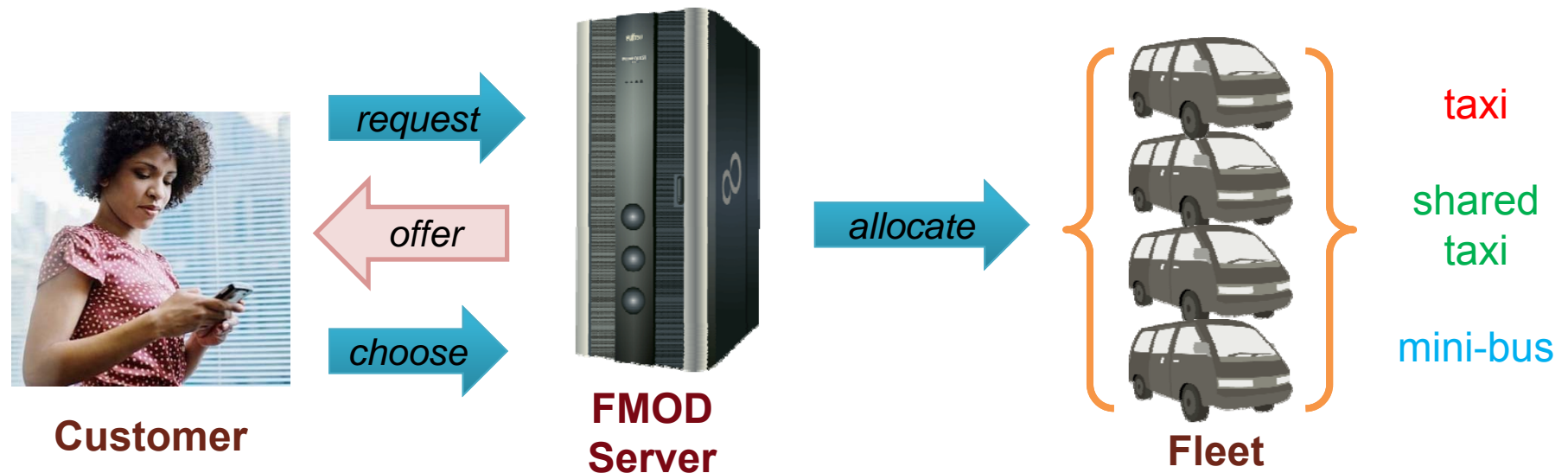
# What is FMOD?

## ■ Flexible Mobility on Demand

- Real-time system
- Flexibility to demand fluctuations

## ■ Concepts

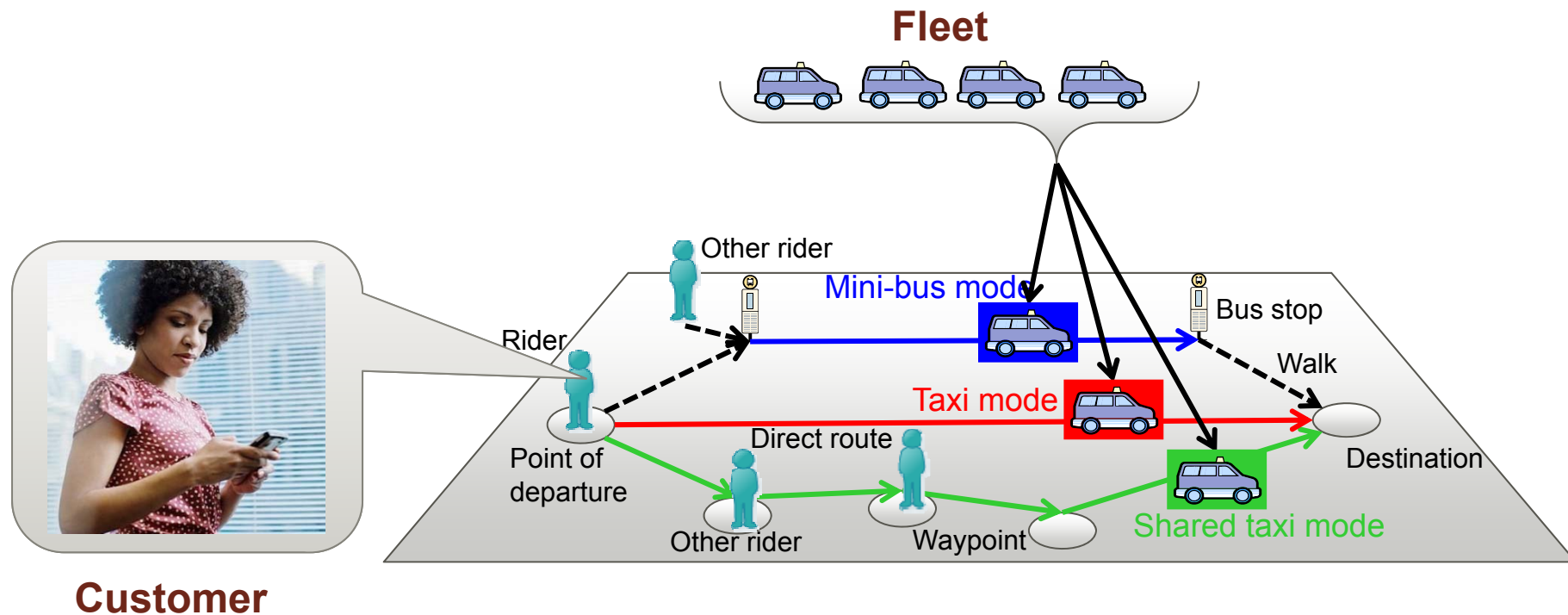
- **Dynamic allocation of vehicle** to service modes
- **Optimized travel menus** are offered to the customer



# Concept of FMOD (1/2)

## ■ Dynamic allocation of vehicle to service modes

- Same vehicle is dynamically reassigned to different service modes according to the evolving demand.



- **Taxi:** Flexible route, flexible schedule, private



- **Shared-taxi:** Flexible route, flexible schedule, **shared**



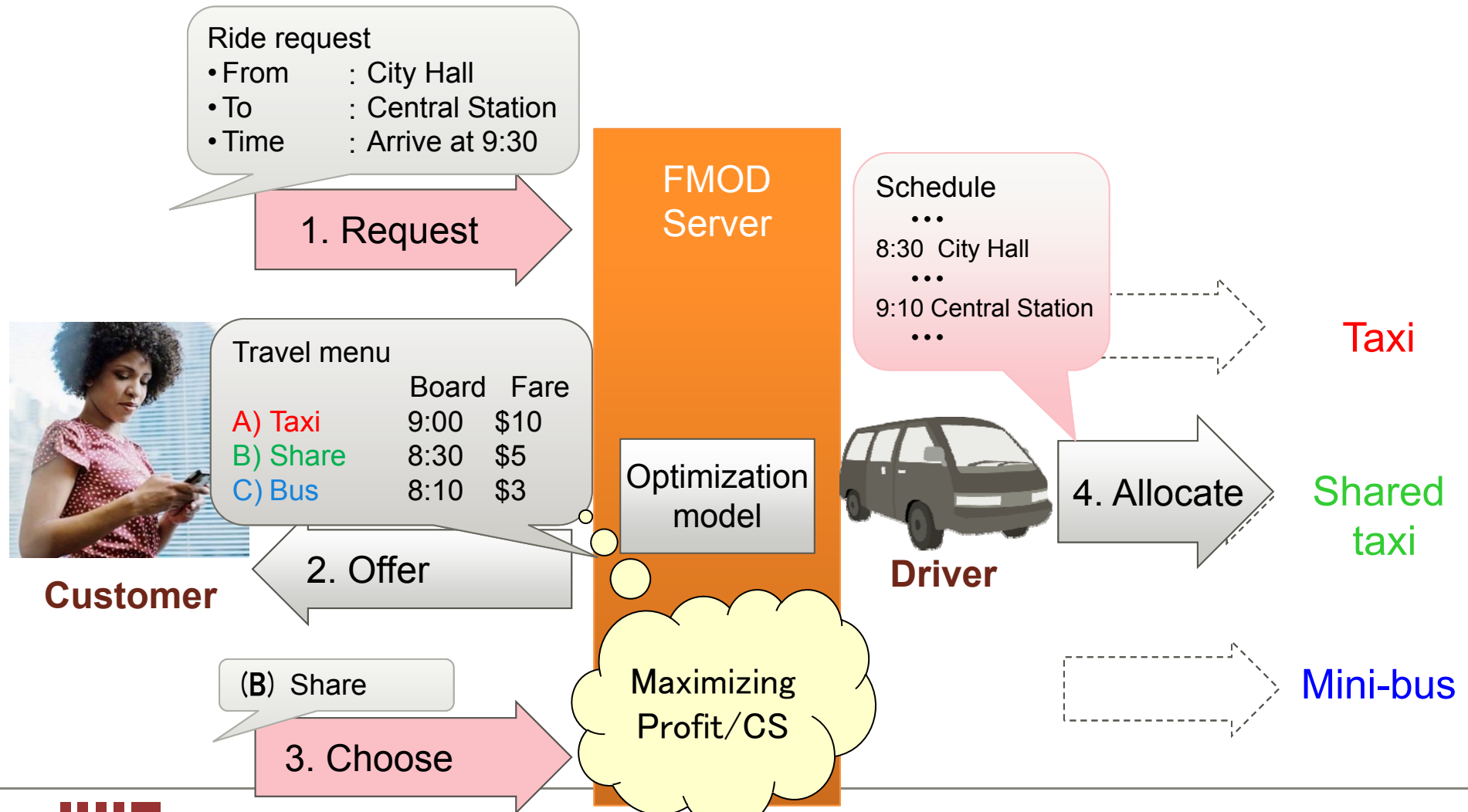
- **Mini-bus:** **Fixed route**, flexible schedule, shared





# Concept of FMOD (2/2)

- **Travel menu** is optimized in order to maximize operator profit / customer surplus



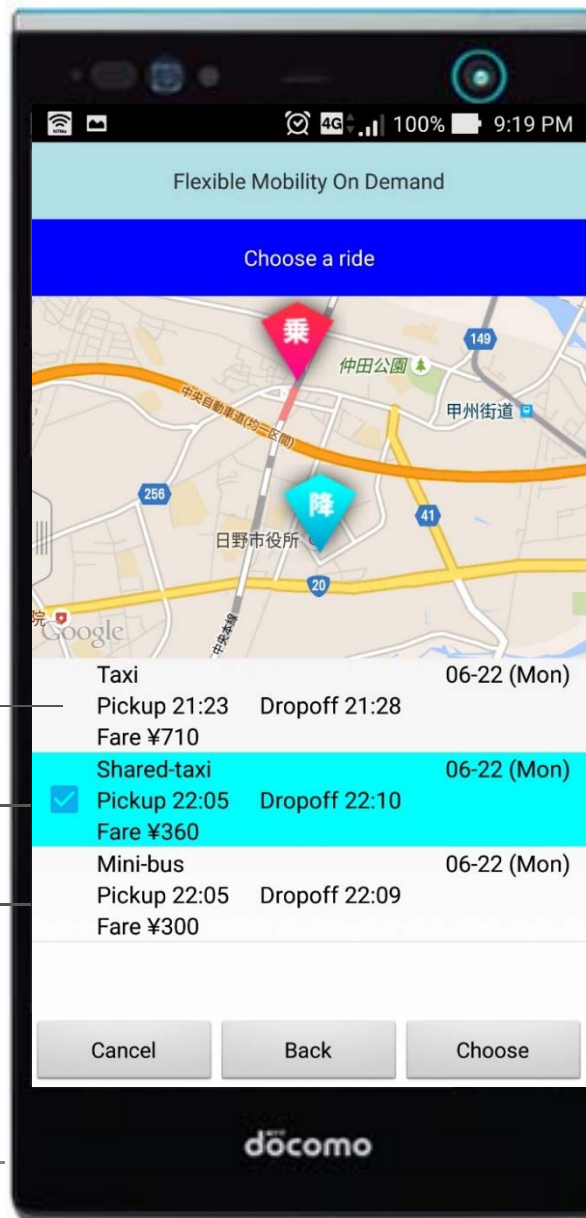
# Example of travel menu (Mobile app)

Options with different  
scheduled time and fare

taxi

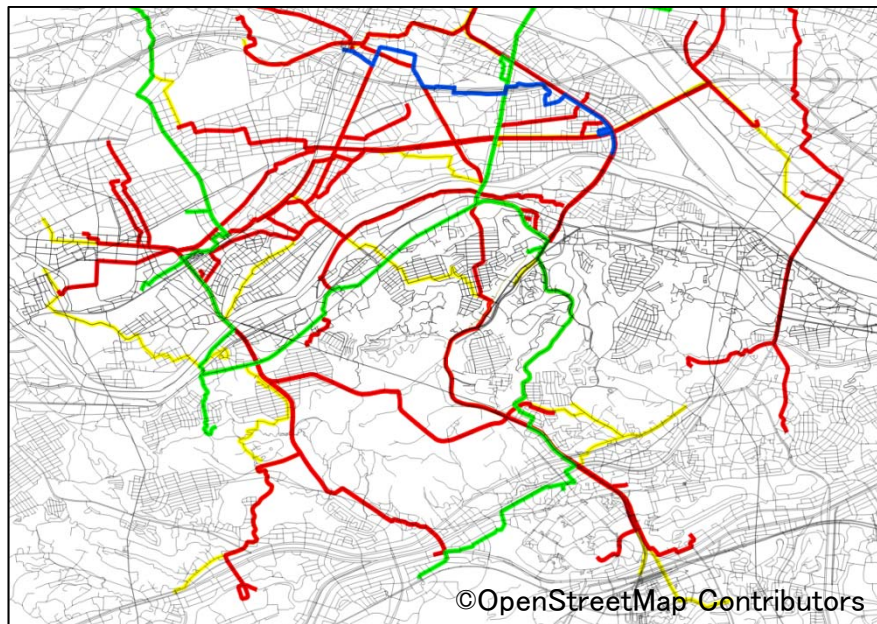
shared-taxi

mini-bus

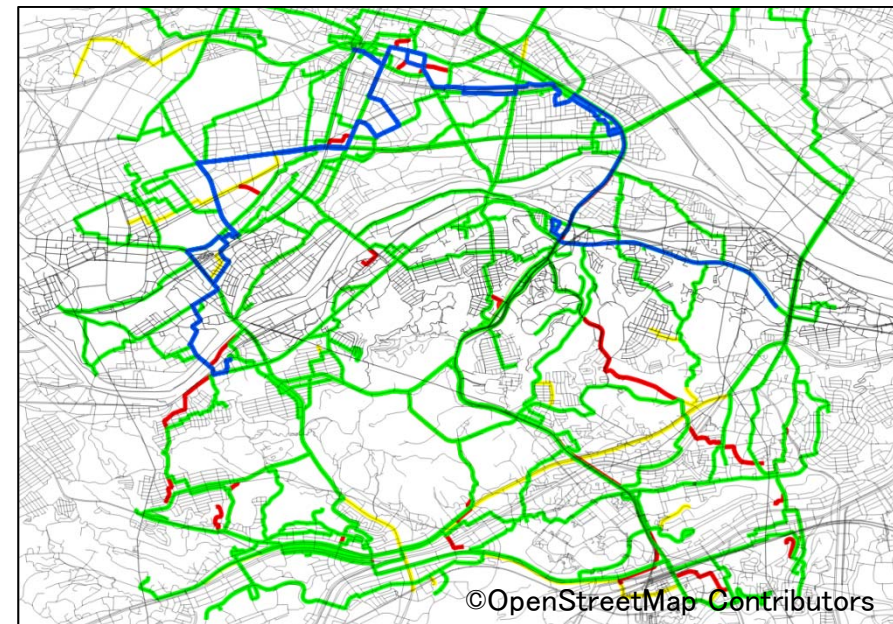


# Dynamic allocation (Simulation)

Red: Taxi, Green: Shared taxi, Blue: Mini-bus, Yellow: empty



Off-peak (AM 6:00)  
Taxi mode is dominant



On-peak (AM 8:00)  
Shared taxi / Mini-bus mode  
is dominant

## ■ Product $p_{n,m,l}$

- A service ( $m \in M$ ) on a vehicle ( $n \in N$ ) departing at a certain time period ( $l \in L$ )

$N$ : set of vehicles, $M$ : set of service modes $L$ : set of time periods
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## ■ Feasible product $p_{n,m,l} \in F$

- A product that satisfies the capacity and scheduling constraints
  - Vehicle capacity
  - No conflict with existing schedules
  - Deviation from preferred time window

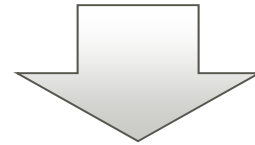
## ■ Assortment

- A list of feasible products on the travel menu

## Phase 1. Feasible product set generation

Feasible products set are generated taking into account:

- Capacity constraints
- Scheduling constraints



## Phase 2. Assortment optimization

Assortment to be offered to the customer is optimized

- Maximize operator's profit and/or consumer surplus based on a choice model

- **Logit model** is used for estimating the choice probabilities for each product and the reject option
- **Utility functions** are defined by:
  - Price
  - In-vehicle travel time
  - Out-vehicle travel time (for mini-bus)
  - Schedule delay

$$\text{Prob}_{n,m,l}(x) = \frac{x_{n,m,l} \exp(\mu V_{n,m,l})}{\exp(\mu V_{\text{reject}}) + \sum_{n' \in N} \sum_{m' \in M} \sum_{l' \in L} x_{n',m',l'} \exp(\mu V_{n',m',l'})}$$

# Assortment optimization model

## ■ Myopic model

- Consider the current request only

$$\max R_{\text{current}}(X)$$

## ■ Look-ahead model

- Take into account future demand

$$\max R_{\text{current}}(X) + R_{\text{future}}(X)$$

$$X = \{x_{n,m,l} \mid x_{n,m,l} \in \{0,1\}\}$$

*Decide which feasible products are included in the assortment*

$$x_{n,m,l} = 0 \quad \forall p_{n,m,l} \notin F$$

*Only feasible products are included*

# Myopic model

- Maximize expected profit from current customer

$$\max R_{\text{current}}(X) = \sum_{n \in N} \sum_{m \in M} \sum_{l \in L} r_{n,m,l} \text{Prob}_{n,m,l}(X)$$

Logit model

Expected profit from each product

$$\text{s.t. } \sum_{n \in N} \sum_{l \in L} x_{n,m,l} = 1 \quad \forall m \in M$$

One product is offered for each services

$r_{n,m,l}$

Profit associated with  $p_{n,m,l}$

$\text{Prob}_{n,m,l}$

Choice probability  $p_{n,m,l}$



# Look-ahead model

- Maximize expected profit from current customer and **expected future profit**

$$\max R_{\text{current}}(X) + \sum_{l \in L} \tilde{r}_l \tilde{z}_l$$

$$\text{s.t. } \sum_{n \in N} \sum_{l \in L} x_{n,m,l} = 1$$

$$z_{n,m,l} \leq \text{Cap}_{n,m,l} - x_{n,m,l}$$

$$\tilde{z}_l \leq \sum_{n \in N} \sum_{m \in M} z_{n,m,l}$$

$$\tilde{z}_l \leq \Phi^{-1}_{\text{Dem}_l}(\tau)$$

*Expected future profit  
as a function of the reserved capacity*

*Reserved capacity is limited by percentile  
of the future demand distribution*

$\tilde{r}_l$	Average future profit in time period $l$
$\tilde{z}_l$	Total reserved capacity in time period $l$
$z_{n,m,l}$	Reserved capacity of $p_{n,m,l}$
$\text{Cap}_{n,m,l}$	Capacity of $p_{n,m,l}$
$\text{Dem}_l$	Future demand in time period $l$

# Simulation Experiment - Conditions

- Network
  - Hino city in Tokyo (approx. 9km × 8km)
- Simulation horizon: 4 hours
- Supply
  - Fleet size: 12 (8 seats)
  - Bus line: actual route
- Demand
  - OD: From Hino station to arbitrary location (based on population density)
- Fare
  - Taxi: \$5 (base) + \$0.5 (per 320m)
  - Shared-taxi: 65% of taxi fare
  - Bus: \$4 (flat rate)
- Operation Cost
  - Variable cost \$0.2 / km
  - Fixed cost \$200 / day / vehicle



(Yellow: Bus line)

- Optimization models
  - Myopic
  - Look-ahead
  
- Objective function
  - Profit maximization
  
- Demand
  - 200, 400, 800 requests

# Results

- In all cases, look-ahead model improves the profit compared to the myopic model.
- As demand increase, improvement in profit increase,

		% change in profit	% change in cons. surplus	# of served pax.	# of no-offers
200 requests	myopic	reference		170	0
	look-ahead	+2.92%	-0.55%	200(+30)	0
400 requests	myopic	reference		269	20
	look-ahead	+30.8%	+16.4%	304(+35)	7(-13)
800 requests	myopic	reference		335	145
	look-ahead	+85.8%	+2.00%	356(+21)	112(-33)

# Results

- In all cases, look-ahead model accommodates more passengers compared to the myopic model

		% change in profit	% change in cons. surplus	# of served pax.	# of no-offers
200 requests	myopic	reference		170	0
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# Results

- Look-ahead model decrease the number of no-offers.

		% change in profit		% change in cons. surplus	# of served pax.	# of no-offers
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- Flexible Mobility on Demand (FMOD)
  - **Dynamic allocation of vehicle** to service modes
  - **Optimized travel menus** are offered to the customer
  
- We have developed and compared two optimization models.
  - Myopic model, Look-ahead model
  
- **Look-ahead model** improves the profit and accommodates more passengers compared to the **myopic model**
  - Especially in high demand scenarios.

- Test with different scenarios
  - Robustness for poor demand estimation
- Field test (Singapore, Japan etc.)
  - Dedicated + non-dedicated fleet
- Real world conditions
  - Traffic congestion, Cancelation / No show, Behind schedule
- Learning the behavior of customer through repeated usage
  - Online calibration of demand model parameters



Thank you for your attention!

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