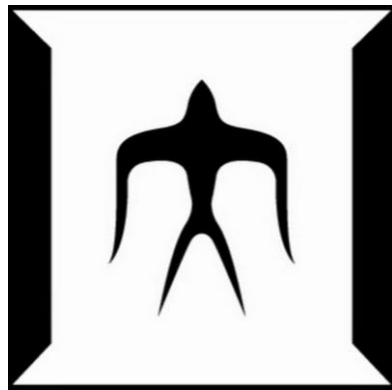


# A Recursive Logit Route Choice Model Considering Link Perceptions with An Application to Tokyo



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# CONTENTS

1. Background
2. Purpose
3. Methodology
4. Data
5. Target Area
6. Result
7. Conclusion
8. Future Works

# 1.BACKGROUND

- The construction of three ring roads
- Seamless tolls in highway with new toll collection
- Dynamic traffic information provided by ETC2.0

→The possibility of drastic change of traffic flow pattern



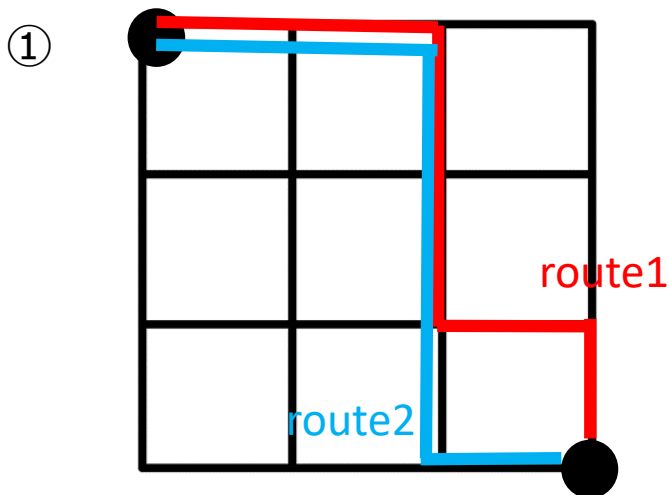
The necessity to conduct detailed **route choice model** to evaluate traffic policy

# 1. BACKGROUND

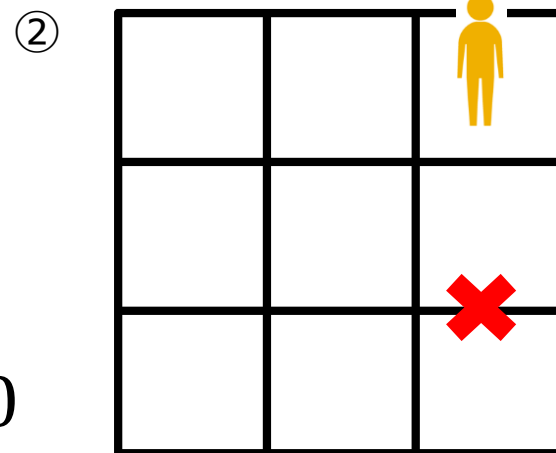
## Route Choice Model

The **path-based model** is typically used, but

- ① The difficulty to enumerate path alternatives
- ② The validity of assumption that traveler determine all of path at the origin



$${}_6C_3 = 20$$



# 1. BACKGROUND

## Route Choice Model

The **path-based model** is usually used

- ① The difficulty to enumerate path alternatives
- ② The validity of assumption that traveler determine all of path at the origin

Conventional Path-based model has limitations!



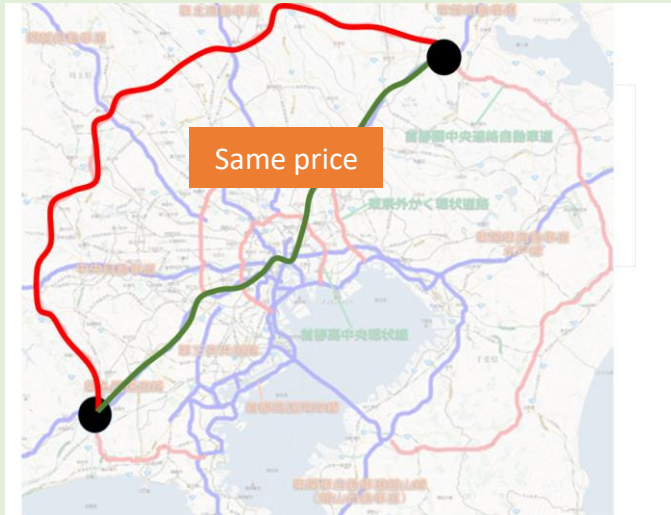
**Link-based Recursive Logit Model** Fosgerau et al. (2013)

the model thinks that traveler choose next link at each node recursively 5

# 1.BACKGROUND

	<b>Path Based Model</b>	<b>Link Based Model</b>
<b>Advantage</b>	<ul style="list-style-type: none"><li>• All of the attribute can be contained in</li></ul>	<ul style="list-style-type: none"><li>• Needless to enumerate path alternatives</li></ul>
<b>Dis-Advantage</b>	<ul style="list-style-type: none"><li>• The difficulty to enumerate path alternatives</li><li>• The validity of assumption that traveler determine all of path at the origin</li></ul>	<ul style="list-style-type: none"><li>• the difficulty to contain non link-additive attribute(Ex Price)</li><li>• The heavy computation load</li></ul>

# 1.BACKGROUND



The toll in highway depends on only origin and destination (non link-additive)

Advantage

Dis-Advantage

- The difficulty to enumerate path alternatives
- The validity of assumption that traveler determine all of path at the origin

- the difficulty to contain non link-additive attribute(Ex Price)
- The heavy computation load

# 1.BACKGROUND

$$\begin{array}{c} \text{Link number} \\ \left. \begin{array}{c} a_{11} \quad \cdots \quad a_{1n} \\ a_{12} \quad \cdots \quad a_{2n} \\ \vdots \quad \ddots \quad \vdots \\ a_{n1} \quad \cdots \quad a_{nn} \end{array} \right\} \begin{array}{c} \text{Link} \\ \text{number} \end{array} \end{array}$$

The necessity to calculate huge inverse matrix  
(It must be satisfied Hawkins-Simon's condition)

(Ex.)  $200000 \times 200000$   
(Tokyo Metropolitan Area)

**Advantage**

**Dis-Advantage**

- The difficulty to enumerate path alternatives
- The validity of assumption that traveler determine all of path at the origin

- the difficulty to contain non link additive attribute(Ex Price)
- **The heavy computation load**



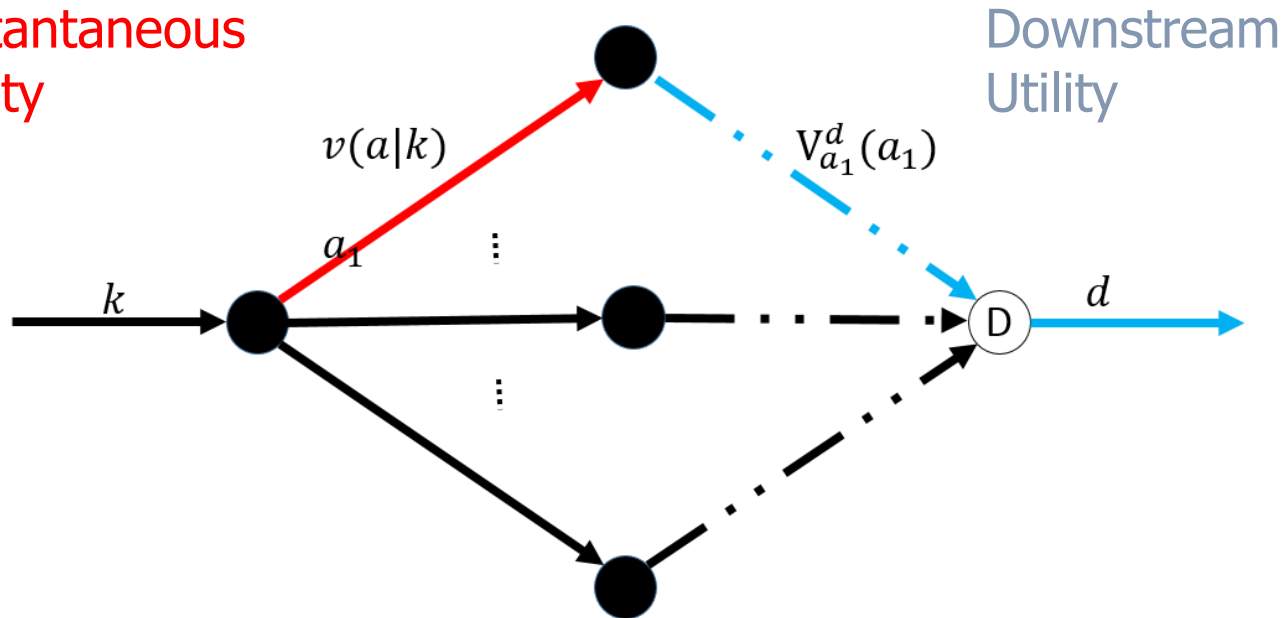
# 2. PURPOSE

Developing a recursive logit (RL) type route choice model in Tokyo Metropolitan Area. Introducing a link perception structure by which computational challenges in RL models would be relaxed.



# 3. METHODOLOGY

Instantaneous  
Utility



$G = (A, v)$ : network     $v$ : the set of node     $A$ : the set of link ( $k, a \in A$ )  
 $k$ : state link     $a$ : potential action link at  $k$      $A(k)$ : the link outgoing from  $k$   
 $d$ : dummy link from destination     $\tilde{A}$ : the set of all link  $\tilde{A} = A \cup d$

## ➤ The basic concept of RL

A traveler  $n$  choose the next link that maximize the sum of  
instantaneous utility  $v(a|k)$  and Downstream utility  $V_{a_1}^d(a_1)$

# 3. METHODOLOGY

## ➤ The basic concept of RL

A traveler  $n$  choose the next link that maximize the sum of **Instantaneous utility**  $v(a|k)$  and **Downstream utility**  $V_{a_1}^d(a_1)$

### ◆ Instantaneous utility $v(a|k)$

the sum of deterministic utility and stochastic utility

$$u_n(a|k) = v_n(a|k) + \tau \varepsilon_n(a) \quad (\tau: \text{scale parameter})$$

(ex)

$$v(a|k) = \beta_{time} Time_a + \beta_{cost} Cost_a + \beta_{turn} UturnDummy$$

### ◆ Downstream utility $V_{a_1}^d(a_1)$

expected utility to the destination when traveler choose link  $a$

$$V_n^d = E[\max(v_n(a|k) + \mu \varepsilon_n(a) + V_n^d(a) )]$$

### ◆ Link choice probability

i.i.d. extreme value type I error terms

$$P_n^d(a|k) = \frac{\exp\{v_n(a|k) + V_n^d(a)\}}{\sum \exp\{v_n(a|k) + V_n^d(a)\}}$$

The same shape  
as Multinomial  
Logit Model

# 3. METHODOLOGY

➤ To find the value of downstream utility  $V^d(a)$ ,

$$V_n^d = \begin{cases} \mu \ln \sum \delta(a|k) \exp \frac{1}{\mu} \{v_n(a|k) + V_n^d(a)\} \\ 0 \end{cases}$$

$$\Leftrightarrow M_{ka} = \begin{cases} \delta(a|k) \exp \left( \frac{1}{\mu} v_n(a|k) + V_n^d(a) \right) \\ 0 \end{cases}$$

$$\Leftrightarrow \mathbf{z} = \mathbf{Mz} + \mathbf{b} \quad \text{Bellman equation}$$

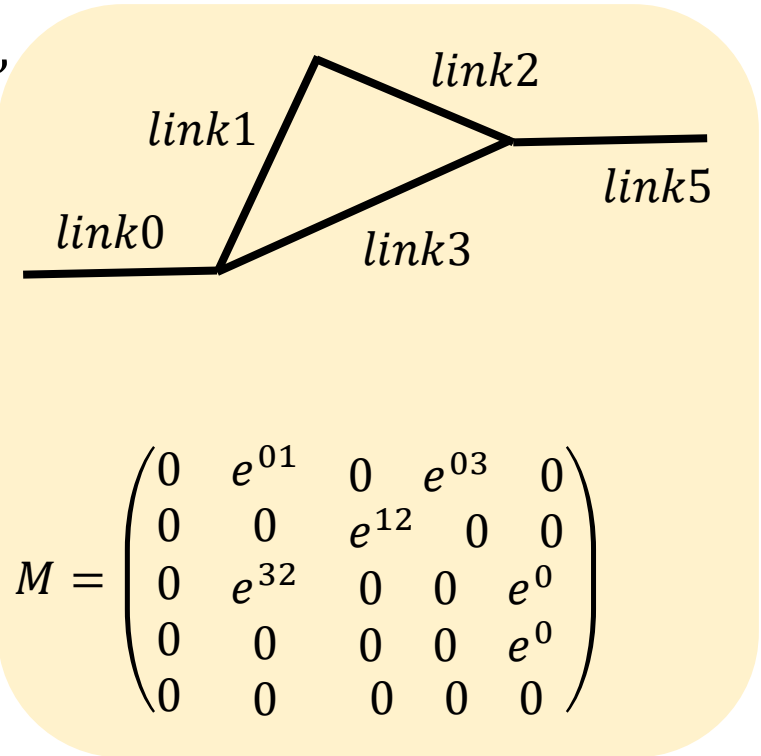
$$\Leftrightarrow \mathbf{z} = (\mathbf{I} - \mathbf{M})^{-1} \mathbf{b}$$

$I : (|\tilde{A}| \times |\tilde{A}|)$  the unit matrix

$M : (|\tilde{A}| \times |\tilde{A}|)$  the incidence matrix defining instantaneous utilities

$z : (|\tilde{A}| \times 1)$  a vector with  $z_k = \exp(1/\mu)V(k)$

$b : (|\tilde{A}| \times 1)$  a vector with  $b_k = 0(k \neq d), b_k = 1(k = d)$



# 3. METHODOLOGY

By the Markov property , path choice probability is

$$P(\sigma) = \prod_{i=0}^{I-1} P(k_{i+1}|k_i)$$

Rearranging this formula,

$$P(\sigma^n) = \frac{\exp(v(\sigma^n/\mu))}{\sum_{v(\sigma^{n'}) \in \Omega} \exp(v(\sigma^{n'}/\mu))}$$

This is the equivalent of MNL with infinite path alternatives!!

Maximum problem of log likelihood is written as bellow

$$\max LL_n(\boldsymbol{\beta}) = \frac{1}{N} \sum_{n=1}^N \ln P(\sigma_n; \boldsymbol{\beta}) = \frac{1}{\mu} \sum_{n=1}^N \left\{ \left[ \sum_{i=0}^{I^n-1} v(k_{i+1}^n | k_i^n) \right] - V(k_0^n) \right\}$$

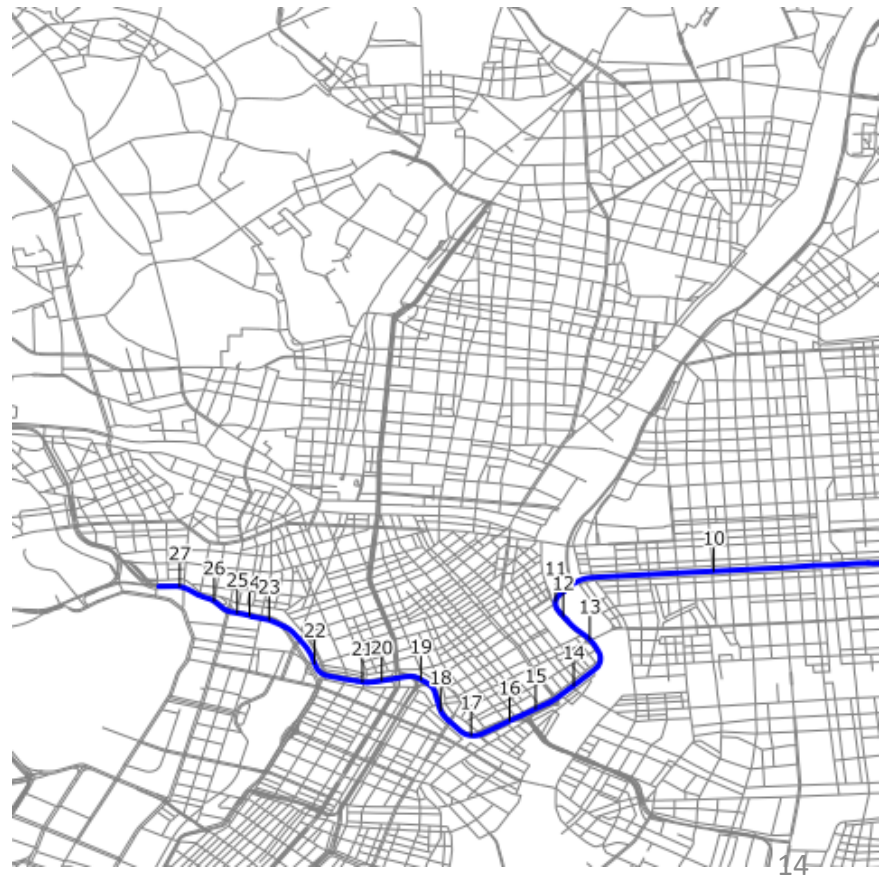
The formula is complicated!

The issue of Recursive logit model is the difficulty of calculations such as bellman equation, Log likelihood

# 3. DIFFICULTY

1. A huge inverse matrix ( Link num  $\times$  Link num ) must be calculated
2. An appearance of cyclic route

$$M = \begin{pmatrix} 0 & e^{-v(1|0)} & \dots & 0 \\ \vdots & 0 & 0 & e^0 \\ \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & 0 \end{pmatrix}$$

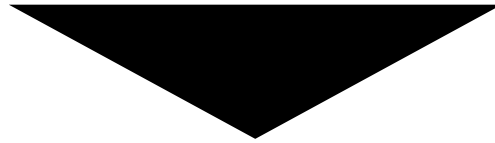




# 4. NEWPOINT

## Recursive Logit Model

The infinite number of route alternatives is assumed  
( including cyclic route )



## **Restrict to more feasible link**

Make the former step of choosing choice set  
Cascetta et al. (2001)

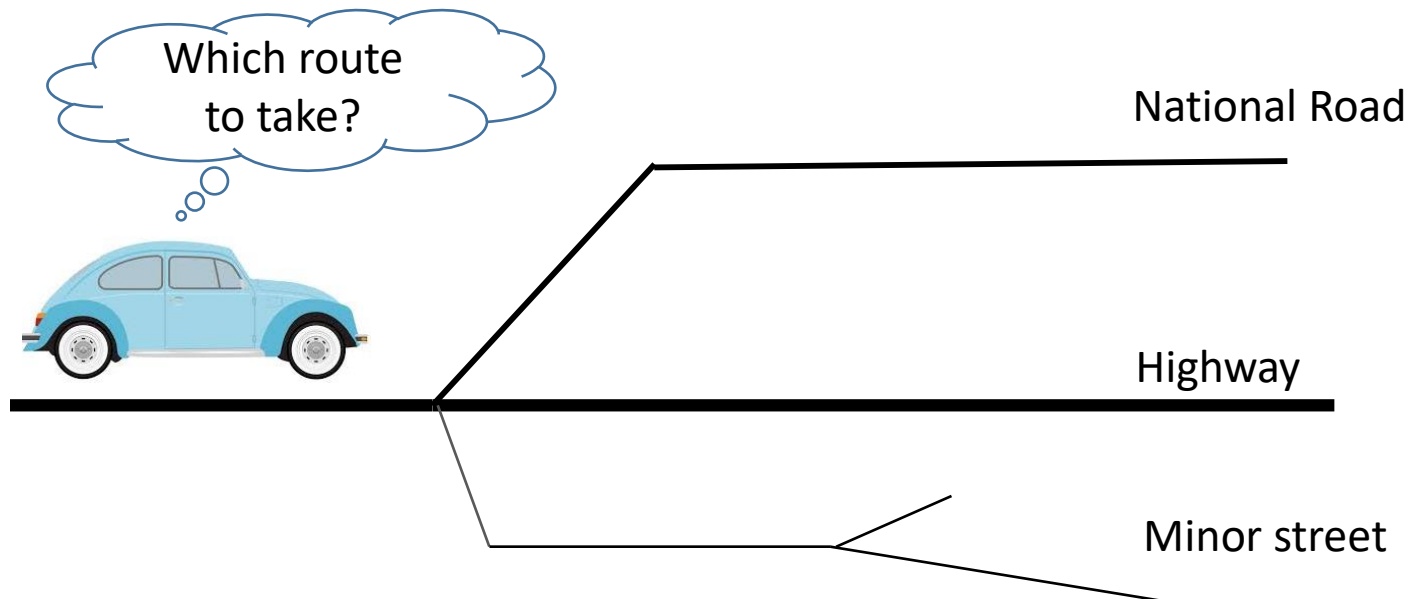


# 4. NEWPOINT

## Restrict to more feasible route

Make the former step of choosing choice set  
(Cascetta et al. 2001)

Using Link Perception



# 4. NEWPOINT

**Restrict to more feasible route in path based model**

**IAP** (Cascetta et al. 2001)

(the **P**erception/**A**vailability of alternative **I**mplicitly in the choice model)

$$U_j^n = V_j^n + \ln \mu_C^n(j) + \varepsilon_j^n$$

$U_j^n$  : Total utility of alternative j for decision maker n

$V_j^n$  : systematic utility of alternative j for decision maker n

$\mu_C^n(j)$  : the degree of membership of alternative j to the fuzzy choice set C for decision-maker n ( $0 \leq \mu_C^n \leq 1$ )

$\varepsilon_j^n$  : the random residual of alternative j for decision maker n

# 4. NEWPOINT

## Restrict to more feasible route in link based model

**IAP** (Cascetta et al. 2001)

(the perception/availability of alternative implicitly in the choice model)

◆ **Instantaneous utility  $v(a|k)$**

the sum of deterministic utility and stochastic utility

$$u_n(a|k) = v_n(a|k) + \ln \mu_C^n(k) + \varepsilon_n(a)$$

$u_n$  : Total instantaneous utility for decision maker  $n$

$v_n(a|k)$  : deterministic link utility of alternative  $k$  for decision maker  $n$

$\mu_C^n(k)$ : the degree of membership of alternative link  $k$  to the fuzzy choice set  $C$  for decision-maker  $n$  ( $0 \leq \mu_C^n(k) \leq 1$ )

$\varepsilon_n$  : the random residual of alternative  $j$  for decision maker  $i$

# 4. NEWPOINT

## Restrict to more feasible route in link based model

**IAP** (Cascetta et al. 2001)

(the perception/availability of alternative implicitly in the choice model)

◆ **Instantaneous utility  $v(a|k)$**

the sum of deterministic utility and stochastic utility

$$u_n(a|k) = v_n(a|k) + \ln \mu_C^n(k) + \varepsilon_n(a)$$

The factor of perception  
Ex.) Road type, the number of lane , ect

# 5. DATA

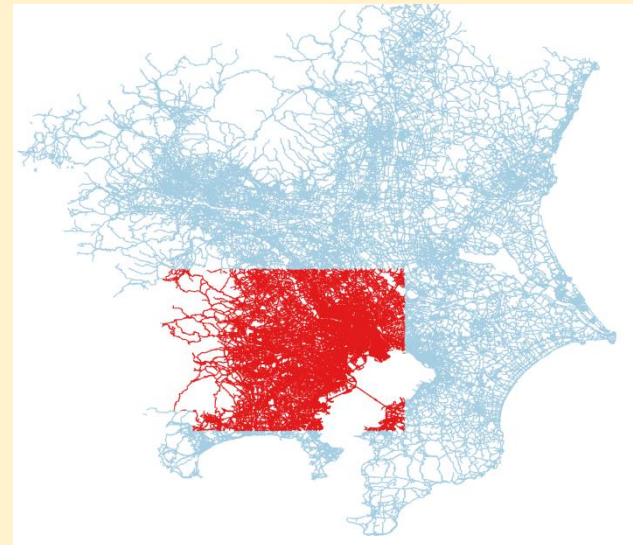
## Trip Data

ETC 2.0(trajjectory data)  
Electronic Toll Collection System



## Network Data

DRM  
Digital Road Map

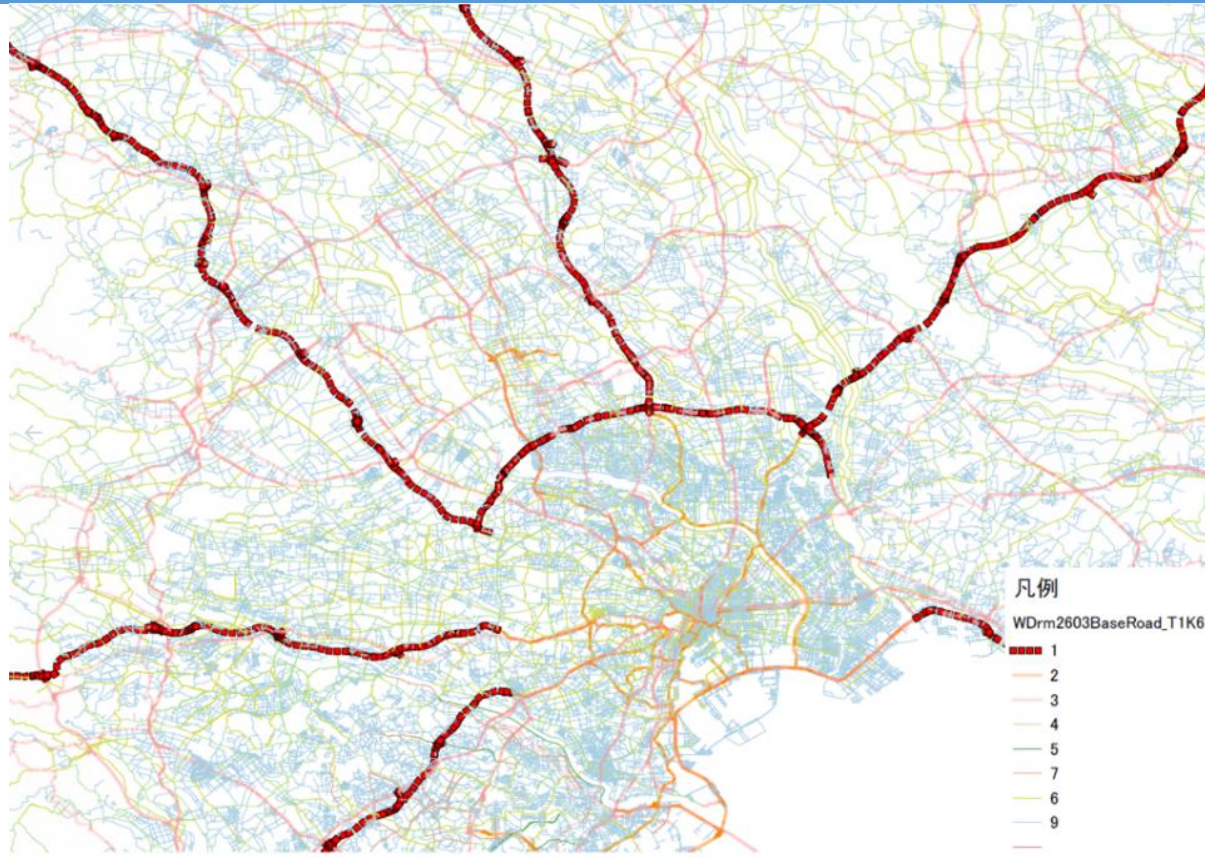


# 5. TRIP DATA



- Conducted by Ministry of Land Infrastructure and Transport
- Probe data
- All of the private vehicle route with ETC2.0
- The data of 24 hours a day, 365 days a year
- The data include
  - Car ID ,Car kind , time , GPS information ...

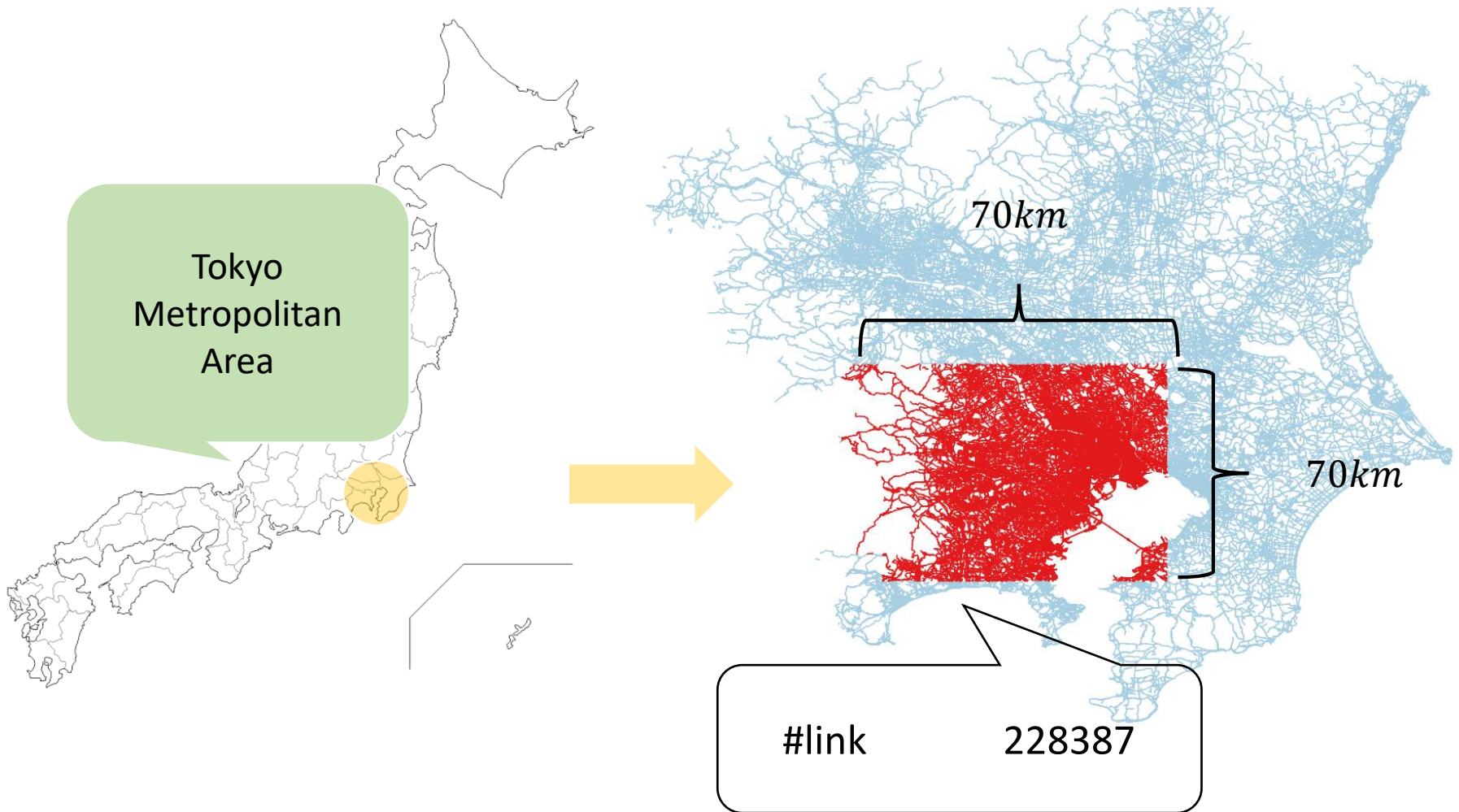
# 5.NETWORK DATA



- Made by The Japan Digital Road Map Association
- It includes road information  
Ex) road type, Road Length, GIS Information...
- Additional Information by IBS(transport consultant)  
Ex) Road cost, Time, traffic light...

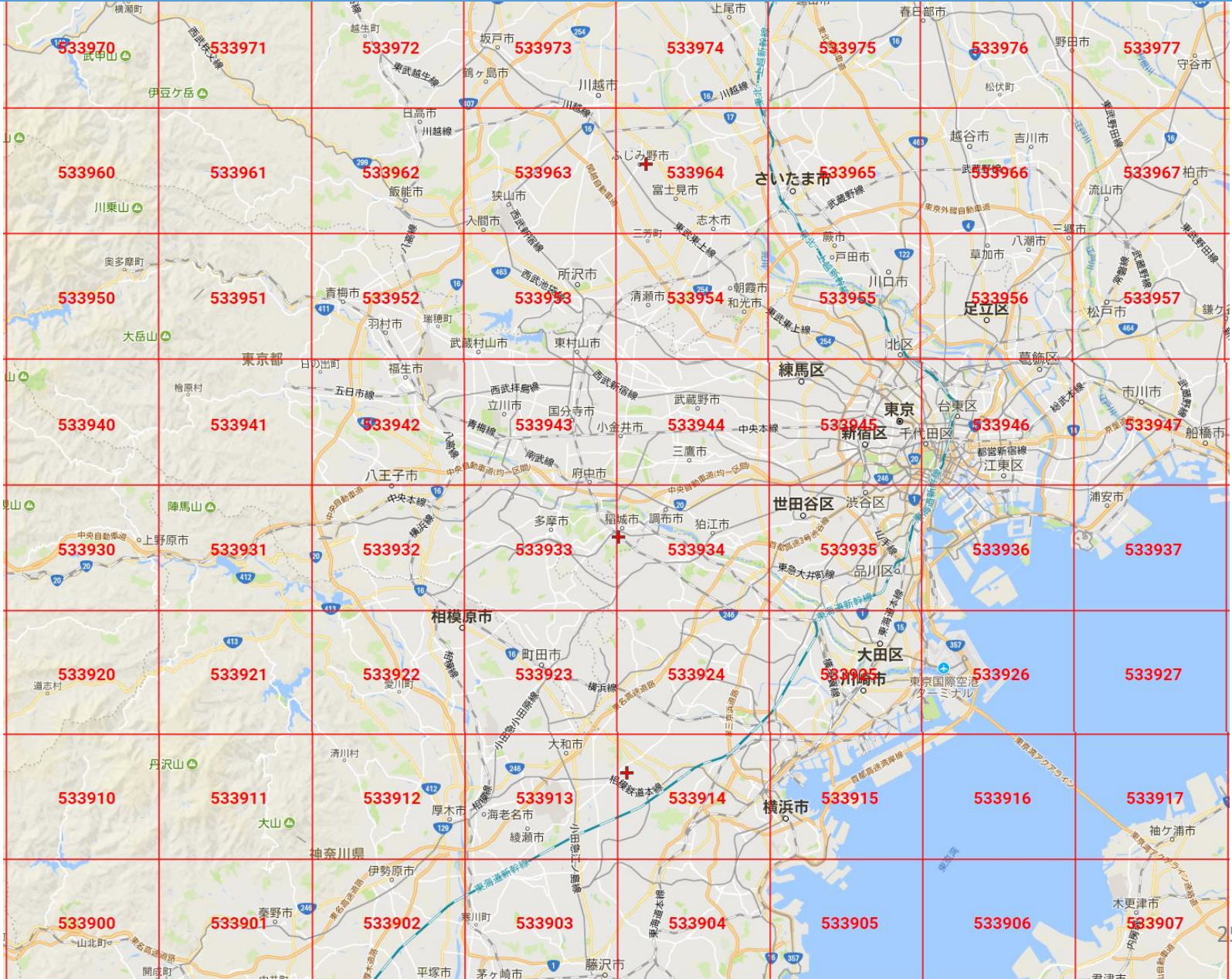


# 5.TARGET AREA



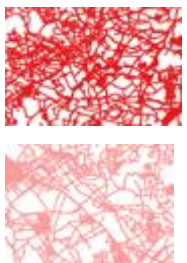
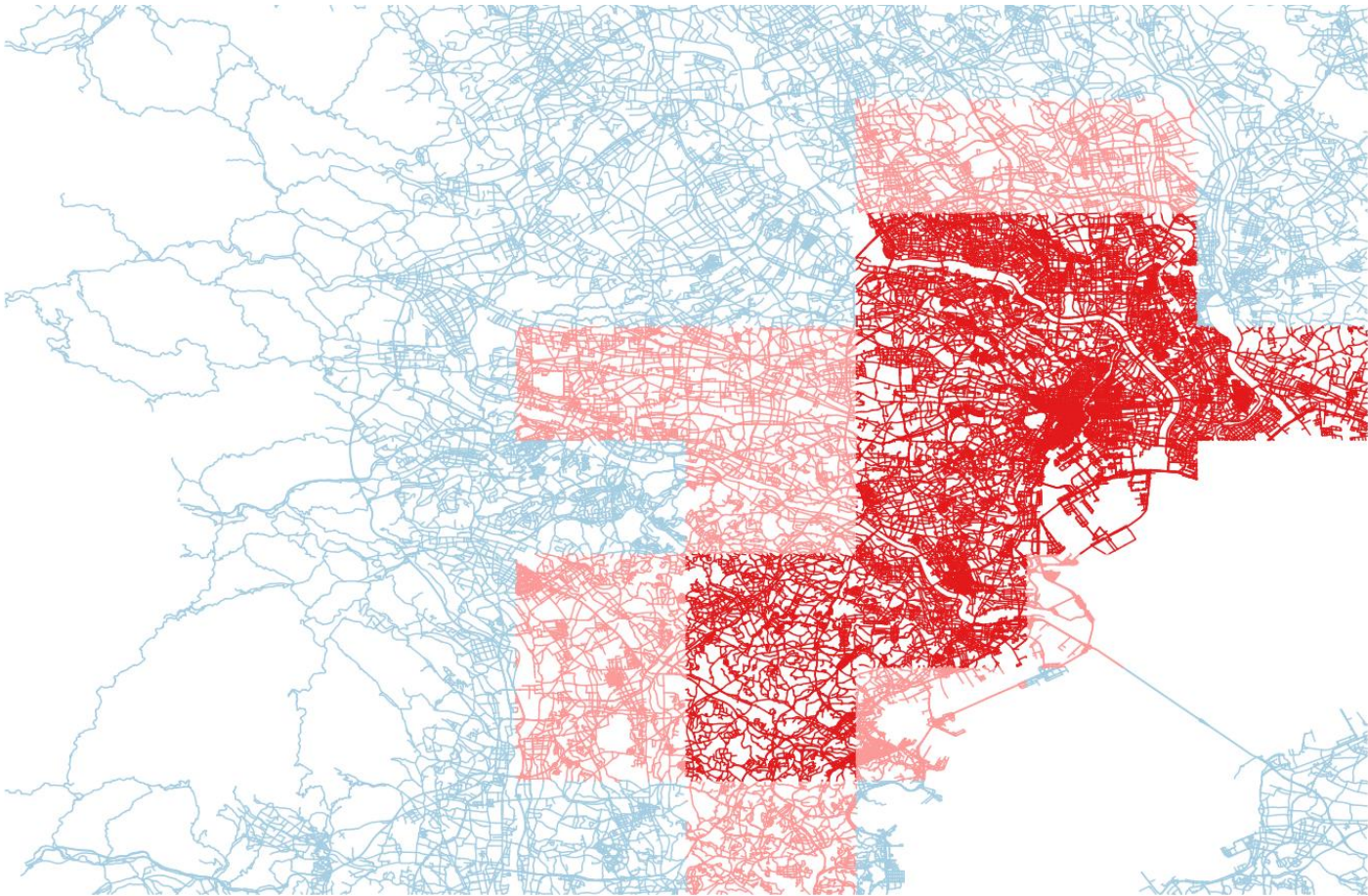


# 5.TARGET AREA





# 5.SETTING DESTINATION



... Congested Area Top 10

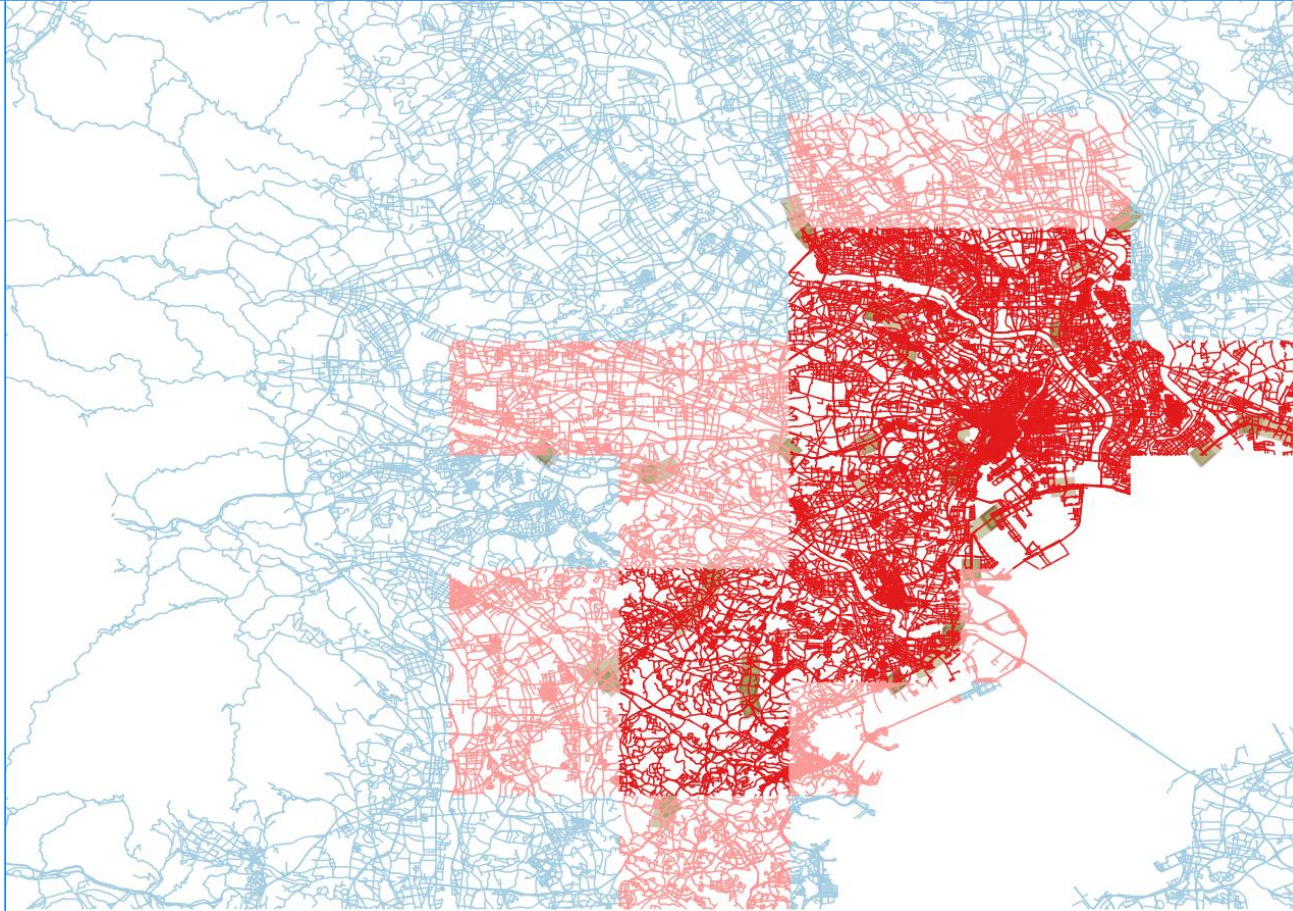
... congested Area Top 20



... each 5 destination (total 50)

... each 2 destination (total 20)

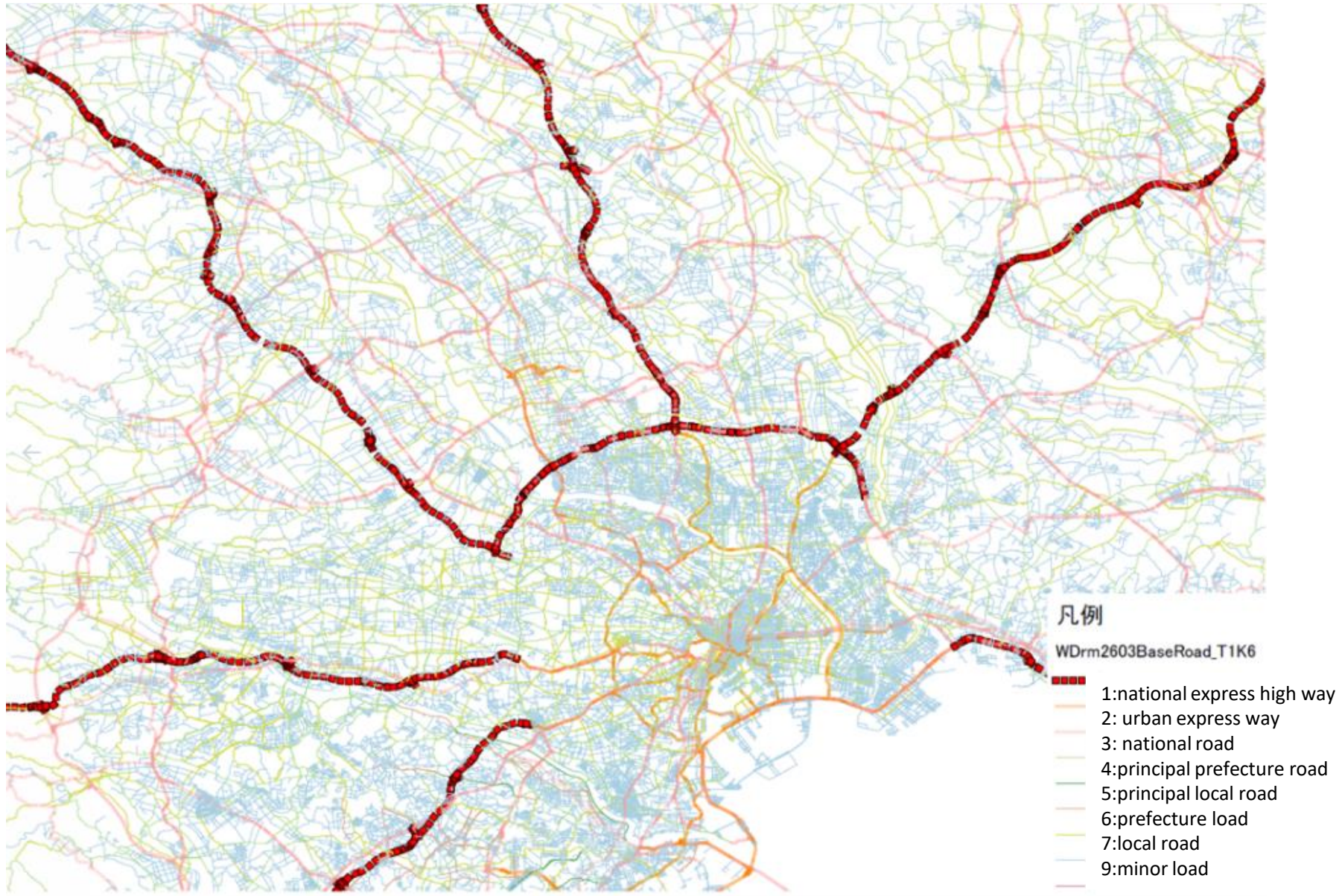
# 5.NETWORK DATA



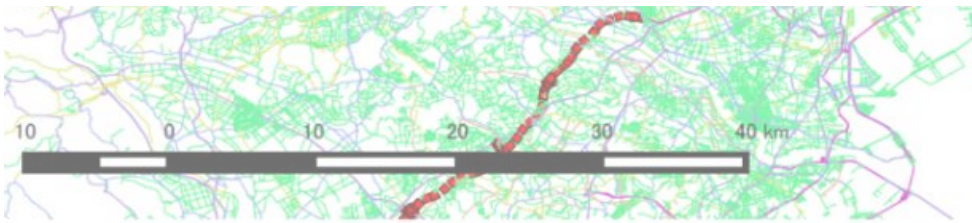
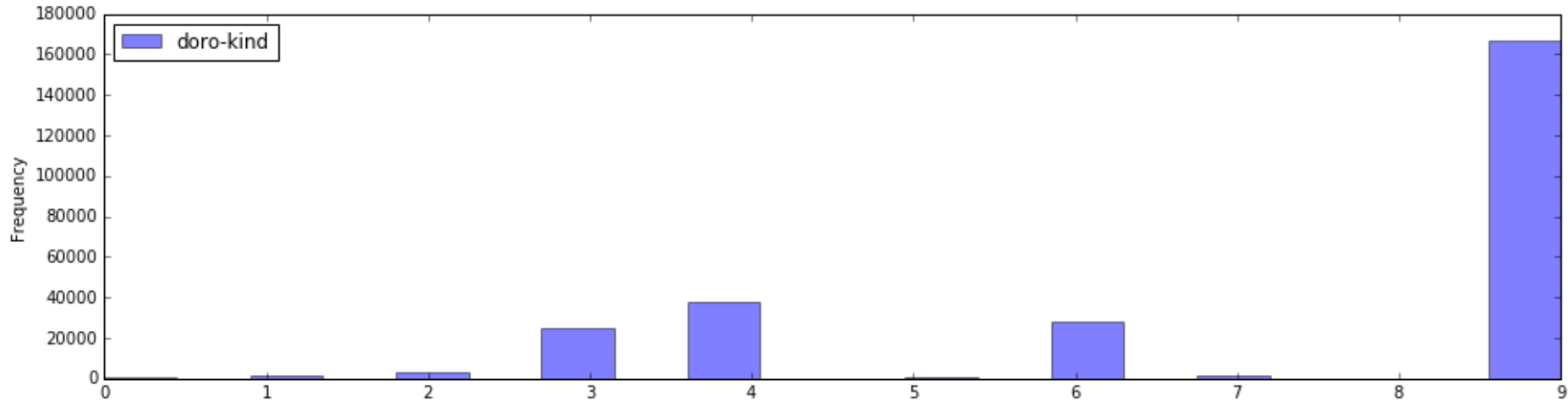
... destination node ( congested area in the target mesh )  
total : 70 destination



# 5.NETWORK DATA

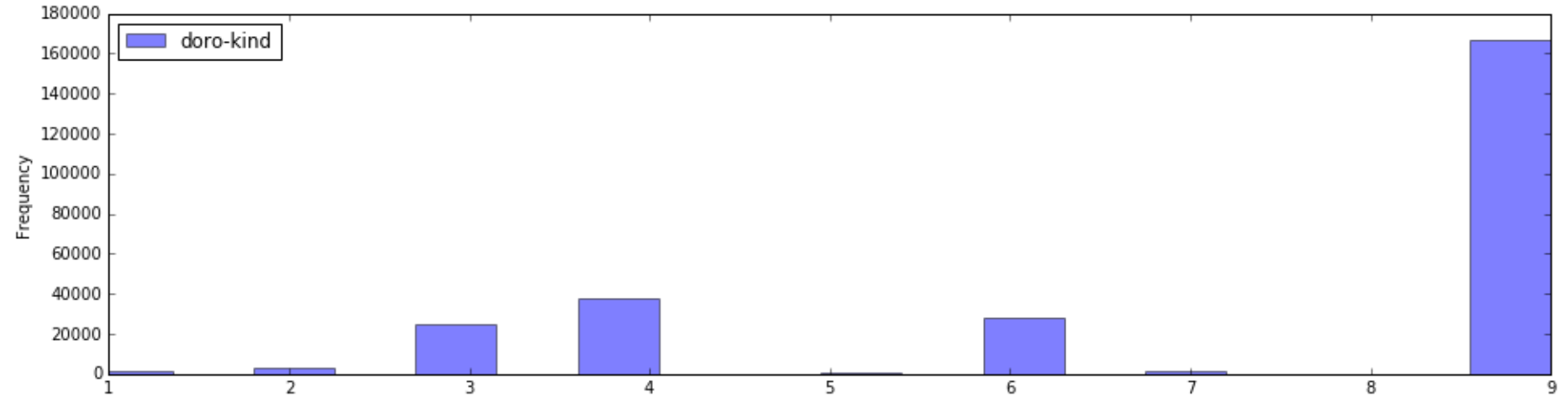


# 5.NETWORK DATA

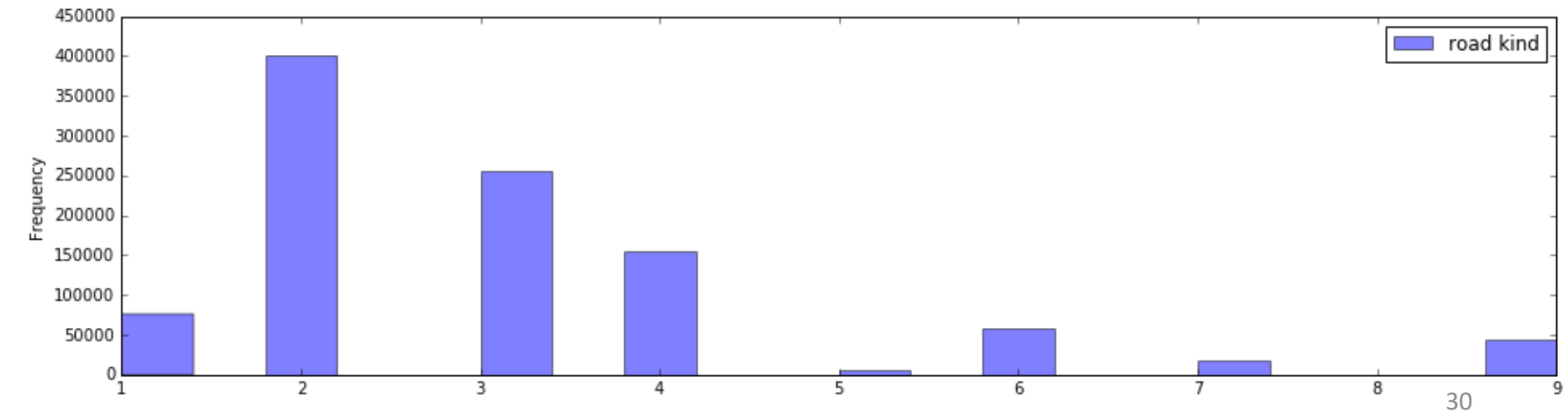


# 5.NETWORK DATA

- The road type distribution of all of the network

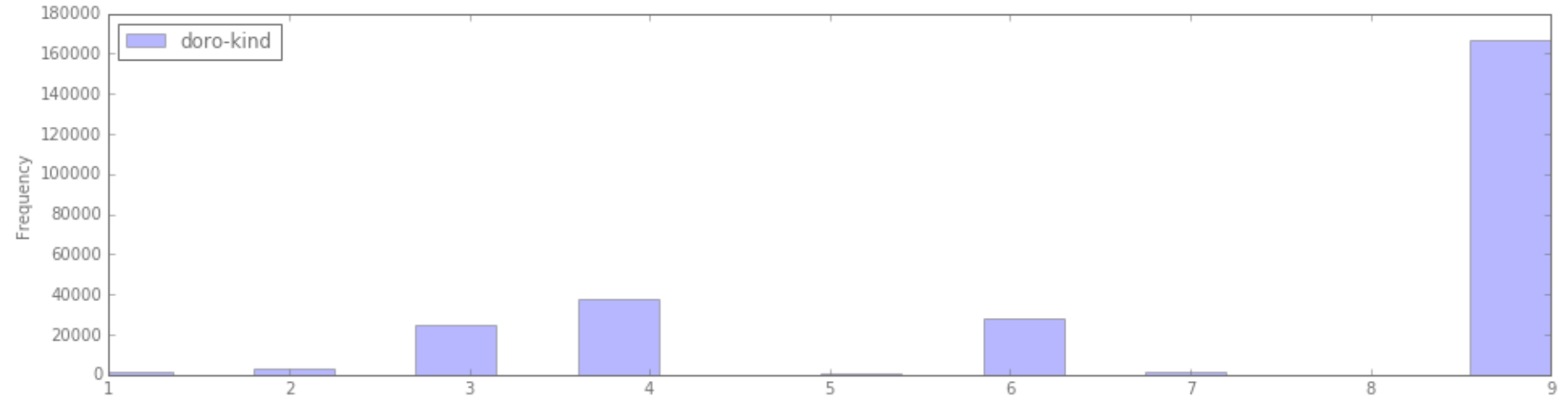


- The road type distribution of only used network

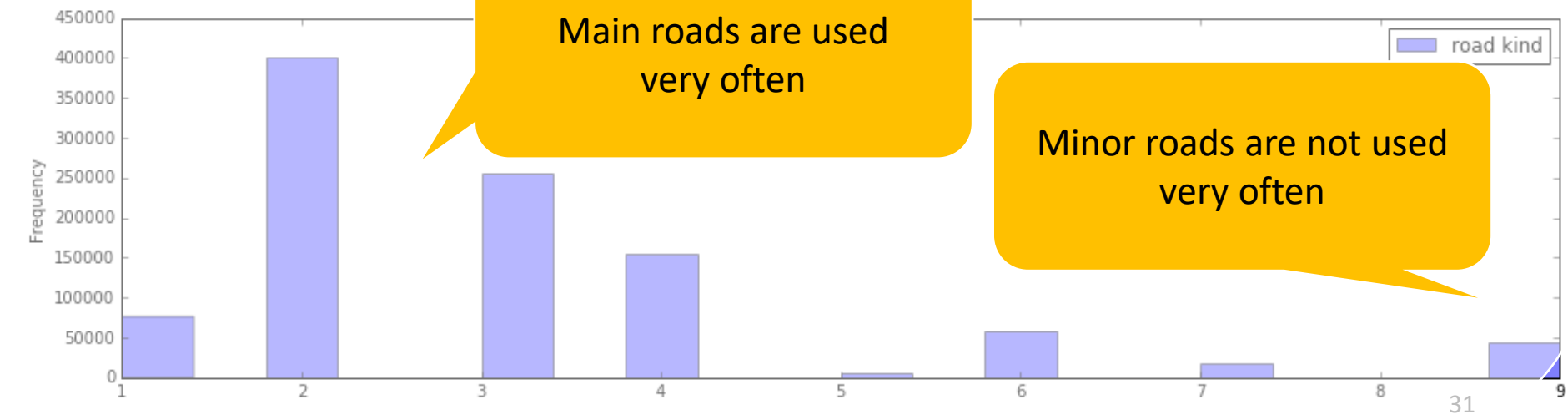


# 5.NETWORK DATA

## ■ Distribution of road types (all of the network)



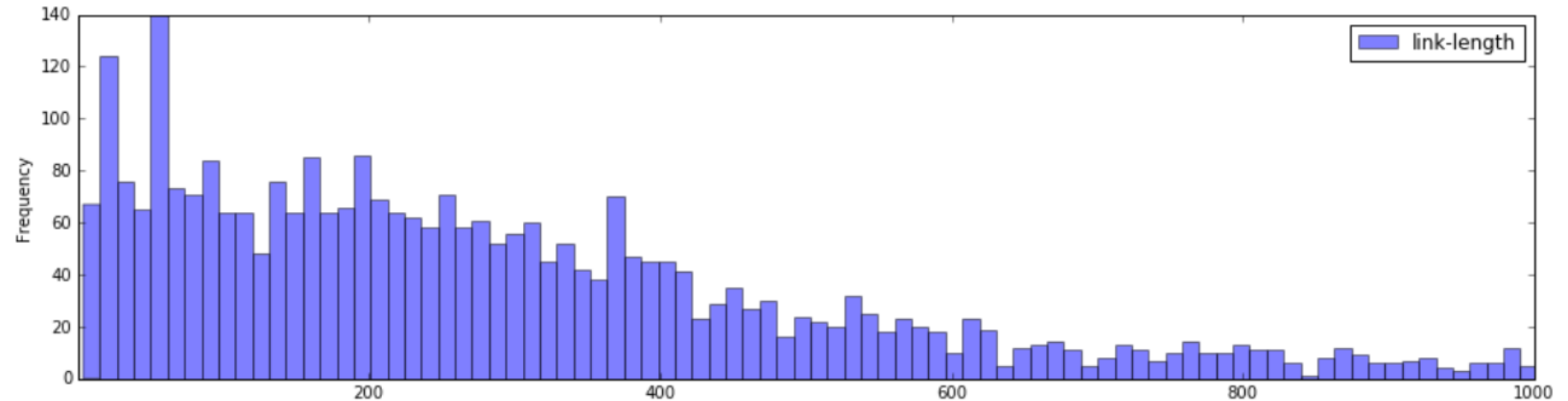
## ■ Distribution of road types (actually used links)





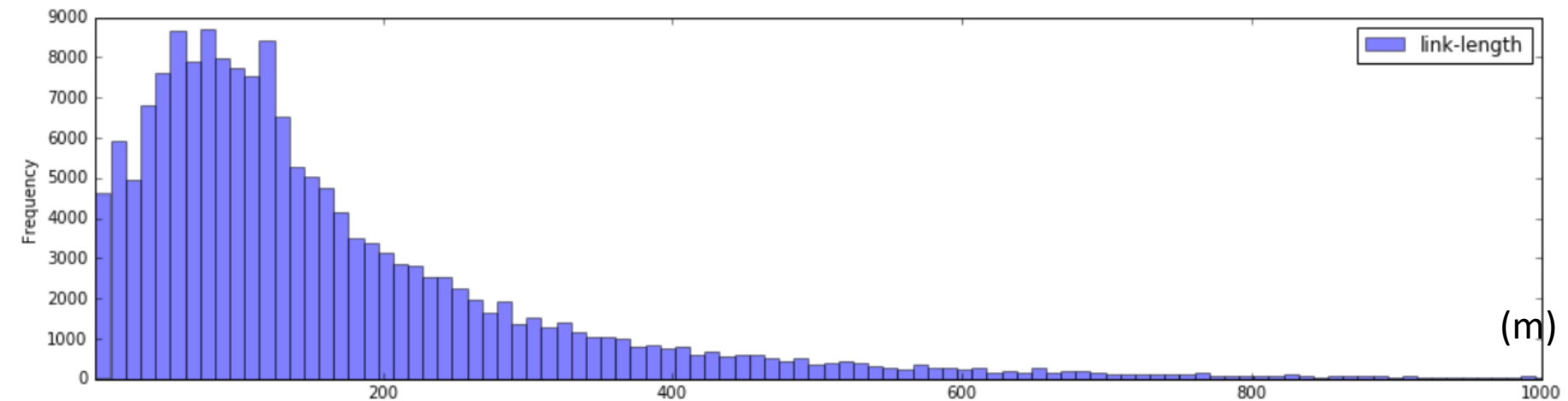
# 5. NETWORK DATA

## ■ Distribution of link length in road type 2 (main road)



## ■ Distribution of link length in road type 9 (minor road)

(m)

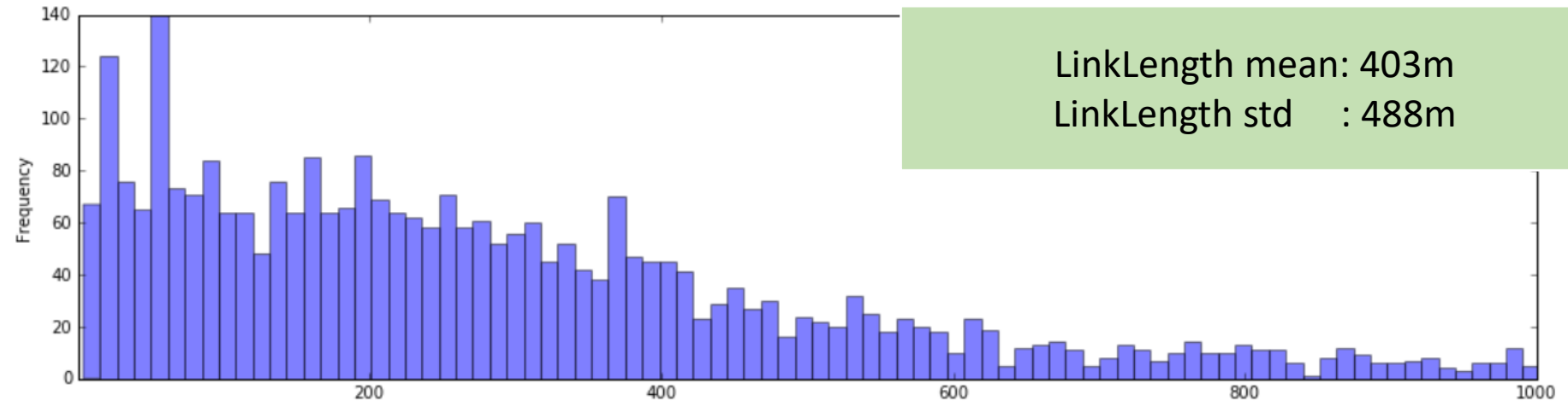


(m)

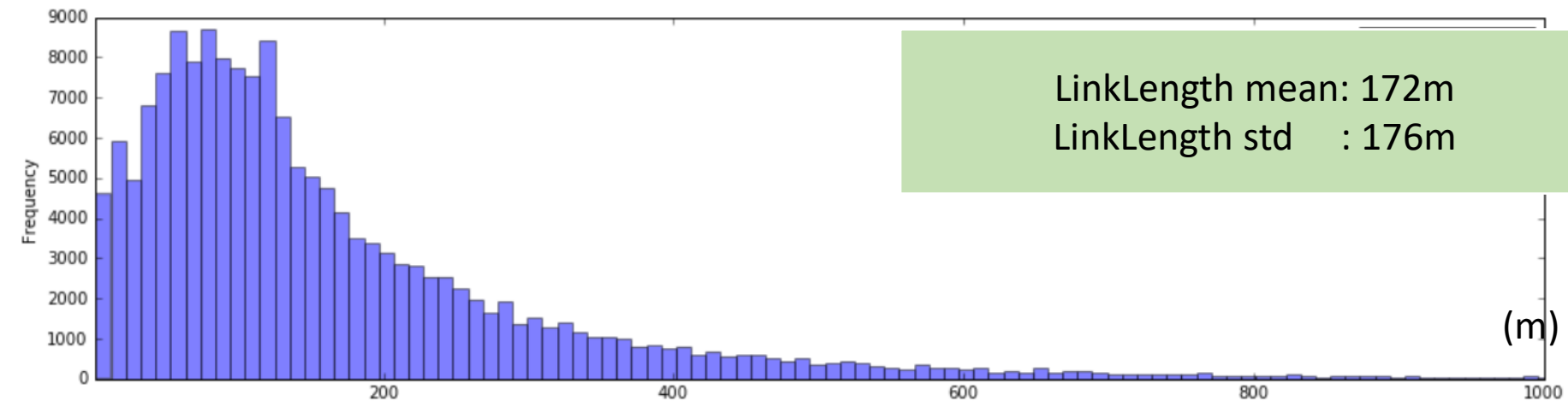


# 5. NETWORK DATA

## ■ Distribution of link length in road type 2 (main road)



## ■ Distribution of link length in road type 9 (minor road)

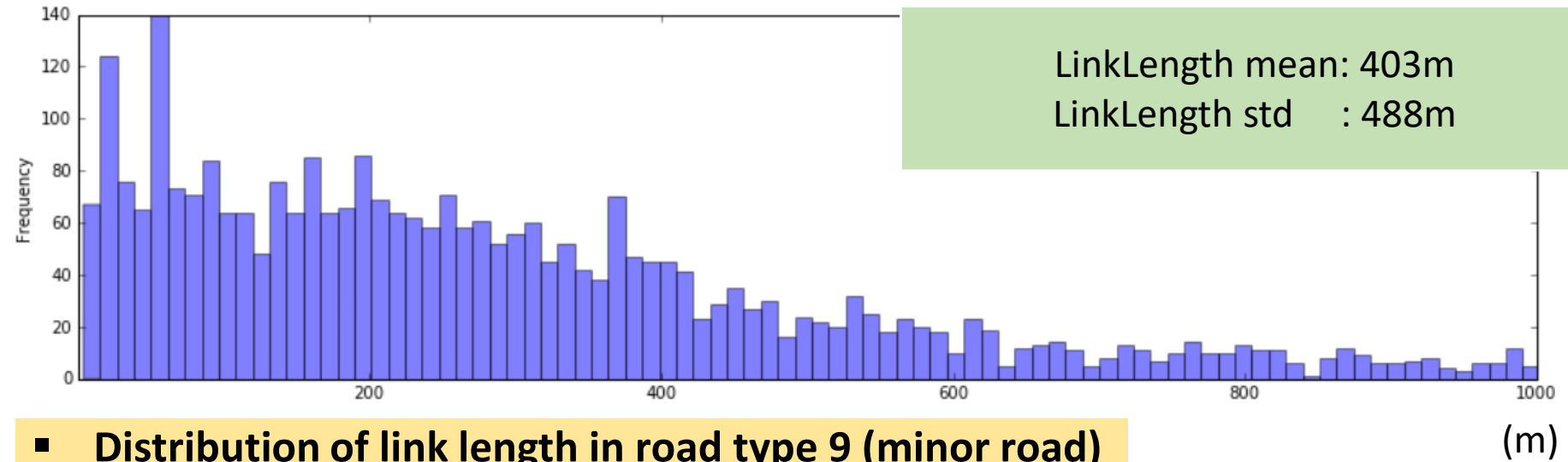


(m)

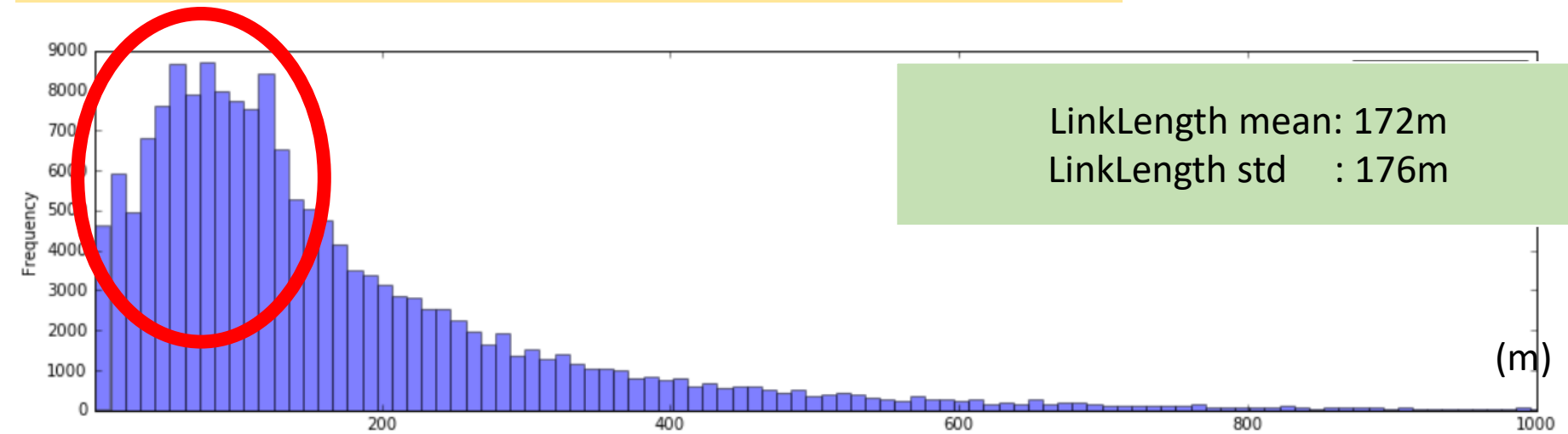
(m)

# 5. NETWORK DATA

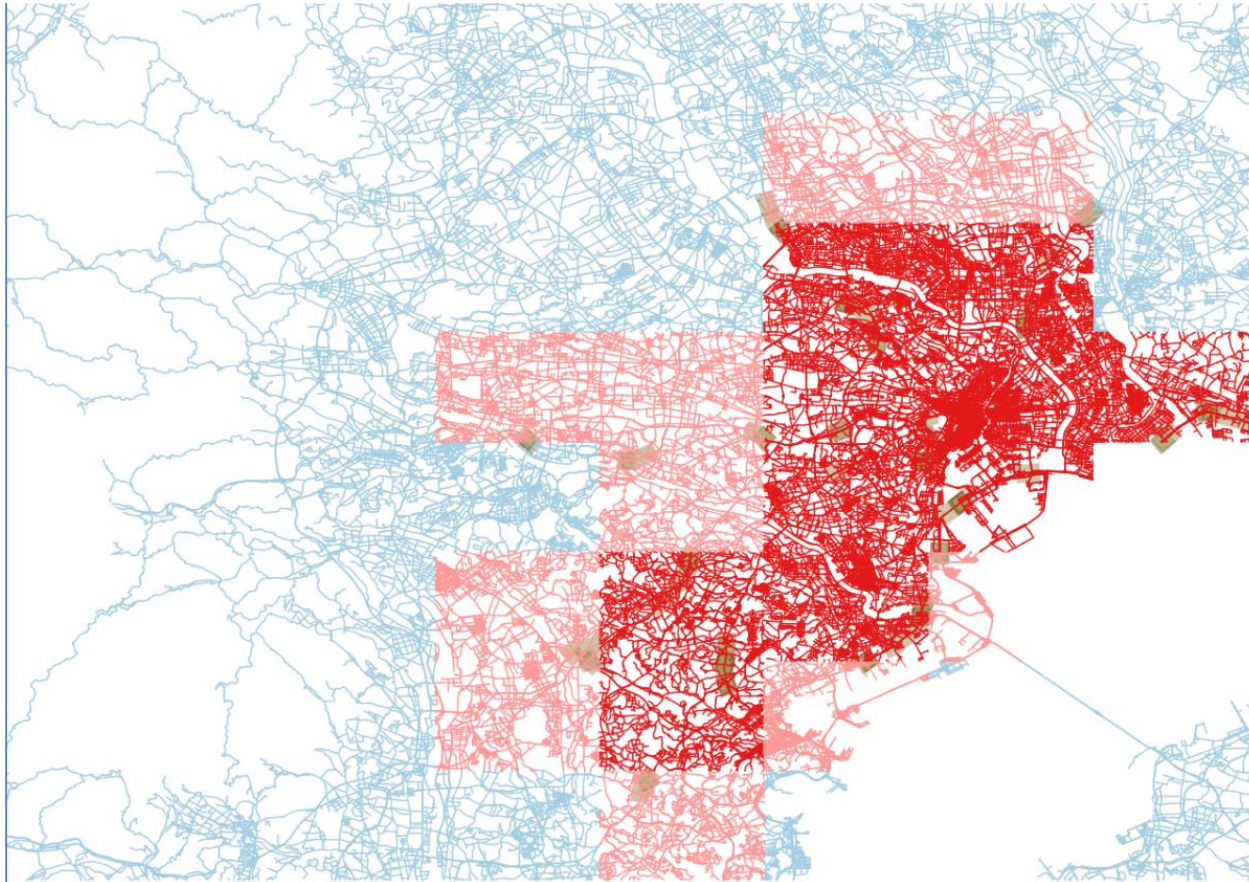
## ■ Distribution of link length in road type 2 (main road)



## ■ Distribution of link length in road type 9 (minor road)



# 5. TRIP DATA



Date : 2016.10.16 (sun)  
Destination # : 70  
link # : 264685  
Trip # : 13449  
Choice average # : 75.2

# 6.RESULT SO FAR ...apply to large area

- **How to define perception term**

$$v_n(a|k) = \beta_{time}x_{time} + \beta_{width}x_{width} +$$

$$\beta_{Uturn}Uturn + \beta_{Rturn}Rturn + \ln\mu(k) + ASC$$

The number of lane

	0	0	0.5	1	1.5	2	2.5	3	4	5	6
0	0	1	1	1	1	2	2	2	2	3	3
2	5	5	5	5	5	6	6	6	6	7	7
3	4	4	4	4	4	5	5	5	5	6	6
4	3	3	3	3	3	4	4	4	4	5	5
6	2	2	2	2	2	3	3	3	3	4	4
9	1	1	1	1	1	2	2	2	2	3	3

*perception term  $\mu = X/7$*

$$1/7 \leq \mu \leq 1$$

# 6. RESULT SO FAR ...apply to large ara

**Model1** (Without perception)

$$v_n(a|k) = \beta_{time}x_{time} + \beta_{width}x_{width} + \beta_{Uturn}Uturn + \beta_{Rturn}Rturn + ASC$$

**Model1** (With perception)

$$v_n(a|k) = \beta_{time}x_{time} + \beta_{width}x_{width} + \beta_{Uturn}Uturn + \beta_{Rturn}Rturn + \ln\mu(k) + ASC$$

# 6.RESULT SO FAR ...apply to large area

**Model1 (without perception)**

**Model2 (with perception)**

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-1.00	-7.23
$\beta_{width}$	-0.61	-0.81
$\beta_{Uturn}$	-1789	-0.08
$\beta_{Rturn}$	-2.24	-0.60
$\beta_{perception}$	1( <i>fixed</i> )	
<i>sample size</i>	13449	
<i>Max LL</i>	-1422210	
<i>estimation time (s)</i>	23600	

# 6.RESULT SO FAR ...apply to large area

**Model1 (without perception)**

**Model2 (with perception)**

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	0.2	-0.16
$\beta_{Rturn}$		
<i>sample size</i>		
<i>Max LL</i>	755171	
<i>estimation time (s)</i>	26404	

positive

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-1.00	-7.23
$\beta_{width}$	-0.61	-0.81
$\beta_{Uturn}$	0.789	-0.08
$\beta_{Rturn}$		
$\beta_{perception}$		
<i>sample size</i>	15449	
<i>Max LL</i>	-1422210	
<i>estimation time (s)</i>	23600	

negative

# 6.RESULT SO FAR ...apply to large area

**Model1 (without perception)**

**Model2 (with perception)**

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	24	-0.50
<i>sample size</i>	<b>negative</b>	
<i>Max LL</i>		
<i>estimation time (s)</i>		

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-1.00	-7.23
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$\beta_{Rturn}$	24	-0.60
$\beta_{perception}$	<b>negative</b>	
<i>sample size</i>		
<i>Max LL</i>		
<i>estimation time (s)</i>	23600	



# 6.RESULT SO FAR ...apply to large area

## Model1 (without perception)

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

## Model2 (with perception)

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-1.00	-7.23
$\beta_{width}$	-0.61	-0.81
$\beta_{Uturn}$	-1789	-0.08
$\beta_{Rturn}$	-2.24	-0.60
$\beta_{perception}$	1( <i>fixed</i> )	
<i>sample size</i>	13449	
<i>Max LL</i>	-1422210	
<i>estimation time (s)</i>	23600	

Estimation time is improved

# 6.RESULT SO FAR ...apply to large area

## Model1 (without perception)

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<b><i>Max LL</i></b>	<b>-753471</b>	
<i>estimation time (s)</i>	26404	

## Model2 (with perception)

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-1.00	-7.23
$\beta_{width}$	-0.61	-0.81
$\beta_{Uturn}$	-1789	-0.08
$\beta_{Rturn}$	-2.24	-0.60
$\beta_{perception}$	1( <i>fixed</i> )	
<i>sample size</i>	13449	
<b><i>Max LL</i></b>	<b>-1422210</b>	
<i>estimation time (s)</i>	23600	

Max LL is improved

# 6.RESULT SO FAR ...apply to large area

**Model1 (without perception)**

**Model2 (with perception)**

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
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$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-1.00	-7.23
$\beta_{width}$	-0.61	-0.81
$\beta_{Uturn}$	-1789	-0.08
$\beta_{Rturn}$	-2.24	-0.60
$\beta_{perception}$	1( <i>fixed</i> )	
<i>sample size</i>	13449	
<i>Max LL</i>	-1422210	
<i>estimation time (s)</i>	23600	

# 6.RESULT SO FAR ...apply to large area

## Model1 (without perception)

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

## Model2 (with perception)

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-1.00	-7.23
$\beta_{width}$	-0.61	-0.81
$\beta_{Uturn}$	-1789	-0.08
$\beta_{Rturn}$	-2.24	-0.60
$\beta_{perception}$	1( <i>fixed</i> )	
<i>sample size</i>	13449	
<i>Max LL</i>	-1422210	
<i>estimation time (s)</i>	23600	

Estimation time is improved

# 6.RESULT SO FAR ...apply to small area

*20 km*

*20 km*

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

Day 2016.10.16 (Sun)

Trip # 2120

Link # 56897

Destination # 10

# 6.RESULT SO FAR ...apply to small area

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

## Model1 (without perception)

$$v_n(a|k) = \beta_{time}x_{time} + \beta_{length}x_{length} + \beta_{cost}x_{cost} + \beta_{width}x_{width} + \beta_{Uturn}Uturn + \beta_{Rturn}Rturn + constant$$

## Model2 (with perception)

$$v_n(a|k) = \beta_{time}x_{time} + \beta_{length}x_{length} + \beta_{cost}x_{cost} + \beta_{width}x_{width} + \beta_{Uturn}Uturn + \beta_{Rturn}Rturn + \ln\mu(k) + constant$$

# 6.RESULT SO FAR ...apply to small area

## model1

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-3.74	-13.86
$\beta_{length}$	-1.12	-33.79
$\beta_{cost}$	-3.65	-50.25
$\beta_{width}$	2.09	10.82
$\beta_{Uturn}$	-623	-0.24
$\beta_{Rturn}$	-38.7	-136
<i>sample size</i>	1039	
<i>Max LL</i>	-1520211	
<i>estimation time (s)</i>	1850	

## model2

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-4.21	-11.3
$\beta_{length}$	-1.14	-2.67
$\beta_{cost}$	-3.20	-3470
$\beta_{width}$	2.09	16.7
$\beta_{Uturn}$	-563	-1.01
$\beta_{Rturn}$	-39.5	38.3
$\beta_{perception}$	1( <i>fixed</i> )	
<i>sample size</i>	1039	
<i>Max LL</i>	-1082644	
<i>estimation time (s)</i>	1368	

# 6.RESULT SO FAR ...apply to small area

## model1

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-2.187	-13.86
$\beta_{length}$	-3.607	-33.79
$\beta_{cost}$	-2.169	-50.25
$\beta_{width}$	1.693	10.82
$\beta_{Uturn}$	-563	-0.24
$\beta_{Rturn}$	-49.54	-136
<i>sample size</i>	1039	
<i>Max LL</i>	-1439386	
<i>estimation time</i>	8497	

## model2

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-3.068	-29.3
$\beta_{length}$	-4.740	-7.91
$\beta_{cost}$	-2.061	-34.2
$\beta_{width}$	1.433	11.8
$\beta_{Uturn}$	-756.8	-287
$\beta_{Rturn}$	-48.44	-174
$\beta_{perception}$	1(fixed)	
<i>sample size</i>	1039	
<i>Max LL</i>	-1330211	
<i>estimation time</i>	1368	



# 6.RESULT SO FAR ...apply to small area

## model1

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-2.187	-13.86
$\beta_{length}$	-3.607	-33.79
$\beta_{cost}$	-2.169	-50.25
$\beta_{width}$	1.693	10.82
$\beta_{Uturn}$	-563	-0.24
$\beta_{Rturn}$	-49.54	-136
<i>sample size</i>	1039	
<i>Max LL</i>	-1439386	
<i>estimation time</i>	8497	

## model2

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-3.068	-29.3
$\beta_{length}$	-4.740	-7.91
$\beta_{cost}$	-2.061	-34.2
$\beta_{width}$	1.433	11.8
$\beta_{Uturn}$	-756.8	-287
$\beta_{Rturn}$	-48.44	-174
$\beta_{perception}$	1.29	85.8
<i>sample size</i>	1039	
<i>Max LL</i>	-1330211	
<i>estimation time</i>	1368	

# 7.CONCLUSION and FUTURE WORKS

## Conclusion

1. Route choice model in large area are created using link based model
2. Adding the step of choosing choice set, the estimation of RL can be more easily and stability and increase accuracy of estimation

## Future works

1. Including non link-additive term like cost as explanatory variables
2. Including perception term more precisely
3. Including explanatory variables more in large network





# RESULT SO FAR ...apply to small area

*20 km*

*20 km*

<i>parameter</i>	<i>Value</i>	<i>T - value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

Day 2016.10.16 (Sun)

Trip # 2120

Link # 56897

Destination # 10

# RESULT SO FAR ...apply to small area

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	1.19	11.0
$\beta_{width}$	-1.30	-2.29
$\beta_{Uturn}$	-28.2	-0.16
$\beta_{Rturn}$	-10.04	-0.50
<i>sample size</i>	13449	
<i>Max LL</i>	-753471	
<i>estimation time (s)</i>	26404	

## Model1 (without perception)

$$v_n(a|k) = \beta_{time}x_{time} + \beta_{length}x_{length} + \beta_{cost}x_{cost} + \beta_{width}x_{width} + \beta_{Uturn}Uturn + \beta_{Rturn}Rturn + constant$$

## Model2 (with perception)

$$v_n(a|k) = \beta_{time}x_{time} + \beta_{length}x_{length} + \beta_{cost}x_{cost} + \beta_{width}x_{width} + \beta_{Uturn}Uturn + \beta_{Rturn}Rturn + \ln\mu(k) + constant$$

# RESULT SO FAR ...apply to small area

## model1

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-3.74	-13.86
$\beta_{length}$	-1.12	-33.79
$\beta_{cost}$	-3.65	-50.25
$\beta_{width}$	2.09	10.82
$\beta_{Uturn}$	-623	-0.24
$\beta_{Rturn}$	-38.7	-136
<i>sample size</i>	1039	
<i>Max LL</i>	-1520211	
<i>estimation time (s)</i>	1850	

## model2

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-4.21	-11.3
$\beta_{length}$	-1.14	-2.67
$\beta_{cost}$	-3.20	-3470
$\beta_{width}$	2.09	16.7
$\beta_{Uturn}$	-563	-1.01
$\beta_{Rturn}$	-39.5	38.3
$\beta_{perception}$	1( <i>fixed</i> )	
<i>sample size</i>	1039	
<i>Max LL</i>	-1082644	
<i>estimation time (s)</i>	1368	

# RESULT SO FAR ...apply to small area

## model1

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-2.187	-13.86
$\beta_{length}$	-3.607	-33.79
$\beta_{cost}$	-2.169	-50.25
$\beta_{width}$	1.693	10.82
$\beta_{Uturn}$	-563	-0.24
$\beta_{Rturn}$	-49.54	-136
<i>sample size</i>	1039	
<i>Max LL</i>	-1439386	
<i>estimation time</i>	8497	

## model2

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-3.068	-29.3
$\beta_{length}$	-4.740	-7.91
$\beta_{cost}$	-2.061	-34.2
$\beta_{width}$	1.433	11.8
$\beta_{Uturn}$	-756.8	-287
$\beta_{Rturn}$	-48.44	-174
$\beta_{perception}$	1.29	85.8
<i>sample size</i>	1039	
<i>Max LL</i>	-1330211	
<i>estimation time</i>	1368	



# RESULT SO FAR ...apply to small area

## model1

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-2.187	-13.86
$\beta_{length}$	-3.607	-33.79
$\beta_{cost}$	-2.169	-50.25
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<i>sample size</i>	1039	
<i>Max LL</i>	-1439386	
<i>estimation time</i>	8497	

## model2

<i>parameter</i>	<i>Value</i>	<i>T – value</i>
$\beta_{time}$	-3.068	-29.3
$\beta_{length}$	-4.740	-7.91
$\beta_{cost}$	-2.061	-34.2
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$\beta_{perception}$	1.29	85.8
<i>sample size</i>	1039	
<i>Max LL</i>	-1330211	
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