

12th Behavior Model Summer School

Combined estimation of activity generation models incorporating unobserved small trips using probe person data

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- ▶ Research background
- ▶ Comparison between PT and PP data
- ▶ Combined estimation model
- ▶ Correcting sampling bias
- ▶ Conclusion

1960s **Person Trip survey** (Paper-based)

(1955 CATS, 1967 Hiroshima)

1980s **Activity based model** – disaggregate data

2000s **Probe Person survey** (GPS-based)

(Zitto and D'este, 1995; Murakami and Wagner, 1999;
Asakura and Hato, 2004; Hato et al., 2006; Stopher et al., 2011)

Non-response activities

**Short trips and activities
are often underreported**

(Wolf et al., 2001; Bricka and Bhat, 2006;
Itsubo and Hato, 2006)



Non-response bias

Changes of activity patterns

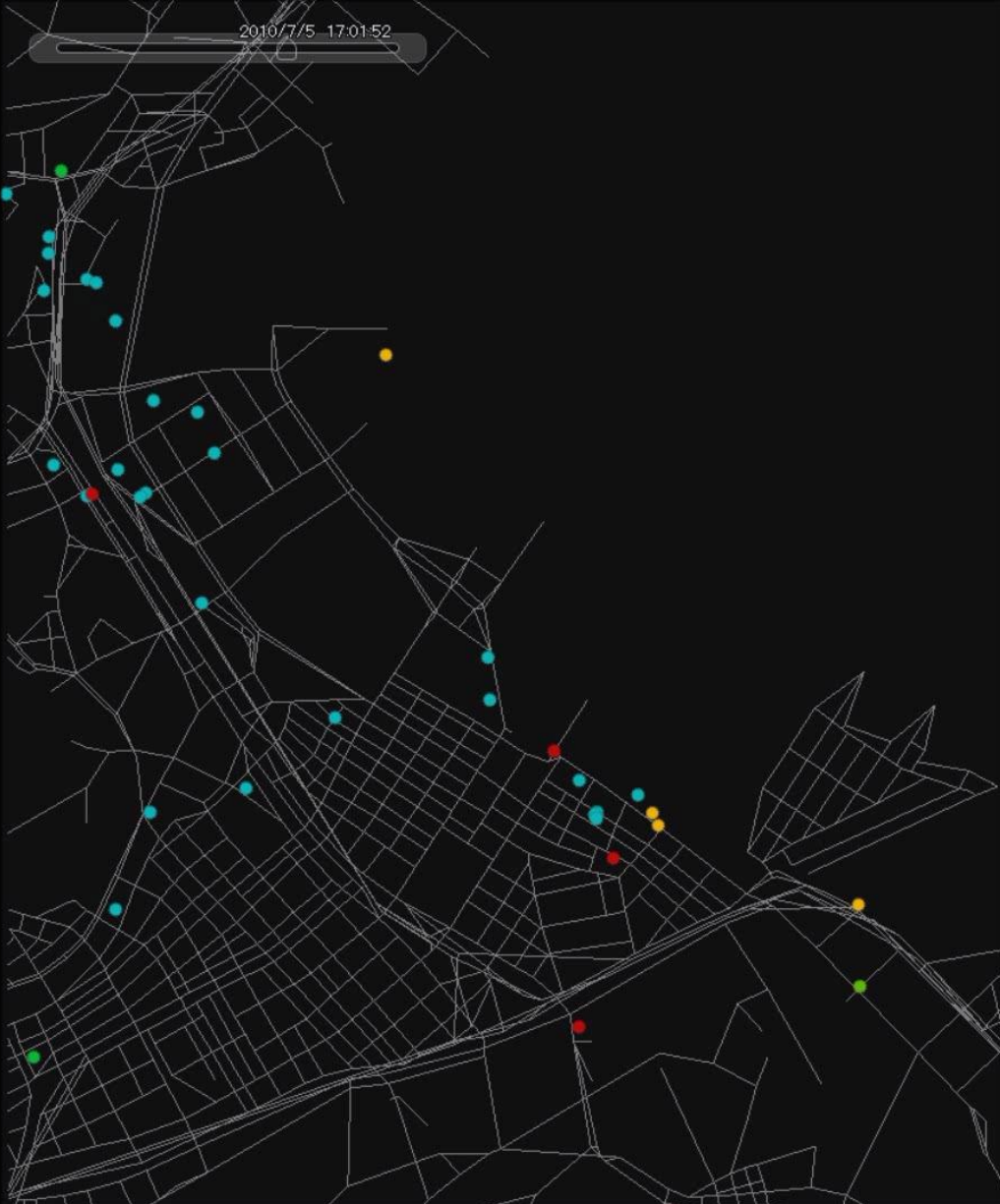
- **Aging society**
- **Inner-city problems**



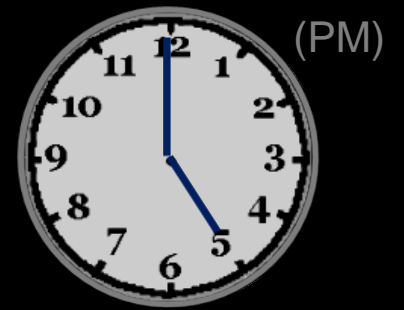
Short activities becomes important

▶ PP survey data

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- Walk
- Car
- Bike
- Motorcycle
- Bus
- Train



Comparison between survey data

PT survey data

Massive location data

PP survey data

Paper-based
(Rely on respondents' memories)

GPS (Automatical but
fragmentary)

GPS (Automatical)
+ Web diary

Zone-based

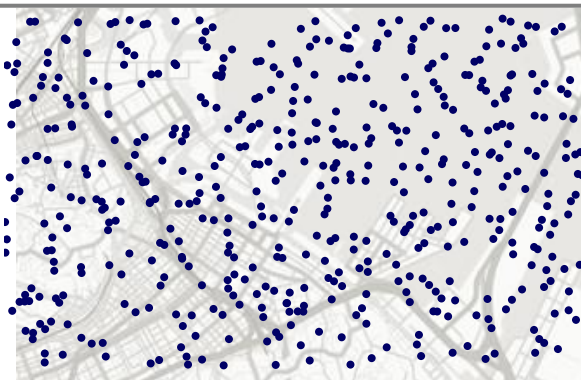
Dot data
(High-resolution)

Dot-based
(High-resolution)

Large sample

Large sample

Small sample



Combined Estimation using both PT and PP data

- Both data are obtained in Yokohama, Japan
- Respondents are resided in Yokohama

■ PT survey

Surveillance period	2008/10 - 2008/11 (each respondent answers his/her travel behavior of 1 day in surveillance period)
Method	Paper questionnaire
The number of all trips	1,906,032 trips
The number of trips in Yokohama	253,737 trips

■ PP survey

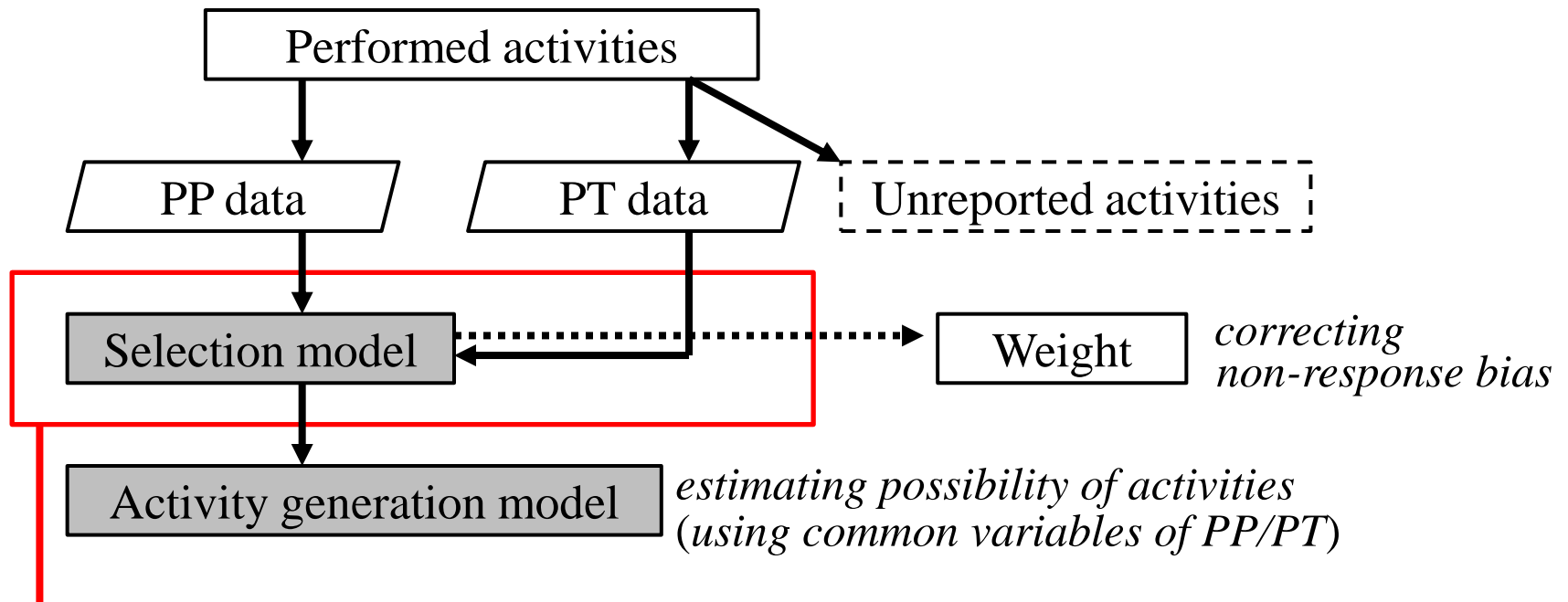
Surveillance period	35 days (2010/07/05 - 2010/08/08)
Survey methods	Probe Person survey with GPS cell phone + Web diary
The number of samples	40 people
The number of Trips	3,617 trips
The number of location data	789,074 points

- In almost all of categories, the number of activities of PT data is smaller than that of PP data

	The number of activities			The sum of activity duration		
	mean		t-statistics	mean (min.)		t-statistics
	PT	PP		PT	PP	
age 20s	1.26	1.39	2.62*	457.0	544.0	5.95*
age 30s	1.40	1.60	3.12*	426.9	389.0	1.84
age 40s	1.53	1.74	2.63*	445.0	288.5	8.60*
age 50s	1.55	1.80	1.98*	412.2	325.9	3.73*
age 60s+	1.56	1.58	0.19	233.3	298.2	1.63
male	1.49	1.78	4.86*	459.7	497.9	2.90*
female	1.43	1.43	0.00	309.1	281.7	2.14*
total	1.46	1.60	5.39*	383.0	389.5	0.65

* : reject the null hypothesis of no difference between the mean of PT data and that of PP data at 5% significant level

- It is assumed that PP data does not have unreported activities.
- If missing activities have some characteristics in common, sampling bias affects the estimation result



Detecting the factors influencing the propensities to record activities

Apply Tobit selection model to activity generation and its observation

● Activity generation model

$$y_{in1}^* = \beta_1 x_{in1} + \varepsilon_{in1}$$

Latent variable **about activity generation** of individual i and zone n

$$\begin{cases} y_{in1} = 1 & \text{if } y_{in1}^* > 0 & \text{generate} \\ y_{in1} = 0 & \text{if } y_{in1}^* \leq 0 & \text{not generate} \end{cases}$$

x_{in1} : explanatory variables of individual i and zone n

ε_{in1} : error term of individual i and zone n

● Selection model

$$y_{in2} = \beta_2 x_{in2} + \varepsilon_{in2}$$

Latent variable **about observation** of individual i and zone n

$$\begin{cases} \text{if } y_{in2} > 0 & y_{in1} \text{ is observed} \\ \text{if } y_{in2} \leq 0 & y_{in1} \text{ is not observed} \end{cases}$$

x_{in2} : explanatory variables of individual i and zone n

ε_{in2} : error term of individual i and zone n

y_{in2} : unobserved variable of individual i and zone n

$$\begin{pmatrix} \varepsilon_{in1} \\ \varepsilon_{in2} \end{pmatrix} \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_1^2 & \rho\sigma_1 \\ \rho\sigma_1 & 1 \end{pmatrix} \right)$$

● Activity generation model

$$y_{in1}^* = \beta_1 x_{in1} + \varepsilon_{in1}$$

● Selection model

$$y_{in2} = \beta_2 x_{in2} + \varepsilon_{in2}$$

Expected value of latent variable y_{in1} after considering selection bias

$$E(y_{in1} | y_{in2} > 0) = \beta_1 x_{in1} + E(\varepsilon_{in1} | \varepsilon_{in2} > -\beta_2 x_{in2})$$

$$= \beta_1 x_{in1} + \rho \sigma_1 \frac{\phi(\beta_2 x_{in2})}{\Phi(\beta_2 x_{in2})}$$

Correction term
(apply only for PT data)

Φ : cumulative distribution function of the standard normal distribution
 ϕ : probability density function of the standard normal distribution

Independent variables	The normal activity generation model		The sample selection model	
	Parameter	t score	Parameter	t score
For activity generation model				
Constant	-1.902	-76.64 *	-1.808	-79.24 *
Male	0.091	12.59 *	0.069	7.51 *
Age \geq 60	-0.116	-15.37 *	-0.106	-10.89 *
Single-member household	0.090	8.79 *	0.100	7.73 *
Car ownership	-0.003	-0.42	-0.002	-0.17
Distance from home (km)	-0.108	-98.83 *	-0.117	-58.97 *
Distance from workplace (km)	-0.025	-43.52 *	-0.028	-35.70 *
Store space (ha) ¹⁾	0.043	71.31 *	0.035	39.55 *
γ	0.125	5.09 *	-	-
ρ	-	-	0.435	16.94 *
For selection model				
Male	-	-	0.466	14.18 *
Age 20-39 years	-	-	-0.545	-7.07 *
Age \geq 60	-	-	0.355	4.20 *
Distance from home (km)	-	-	0.071	0.66
Distance from workplace (km)	-	-	0.020	0.23
Stay Duration (min.)	-	-	0.044	4.99 *
μ	-	-	3.557	17.67 *
Observations (PT)	1,780,164		1,780,164	
Observations (PP)	23,000		23,000	
Initial log-likelihood	-1,249,858		-1,249,858	
Final log-likelihood	-65,013		-64,272	
Rho-squared $\bar{\rho}^2$	0.948		0.949	

Following attributes associate with activity under-reporting at the significant level

- male
- stay duration
- age 20-39 years
- age 60+

- Not relevant; * Significant at 5% level.

1) : The sum of space about retail stores in the zone

To correct the bias, the inverse of observation probability is considered **the weight** as:

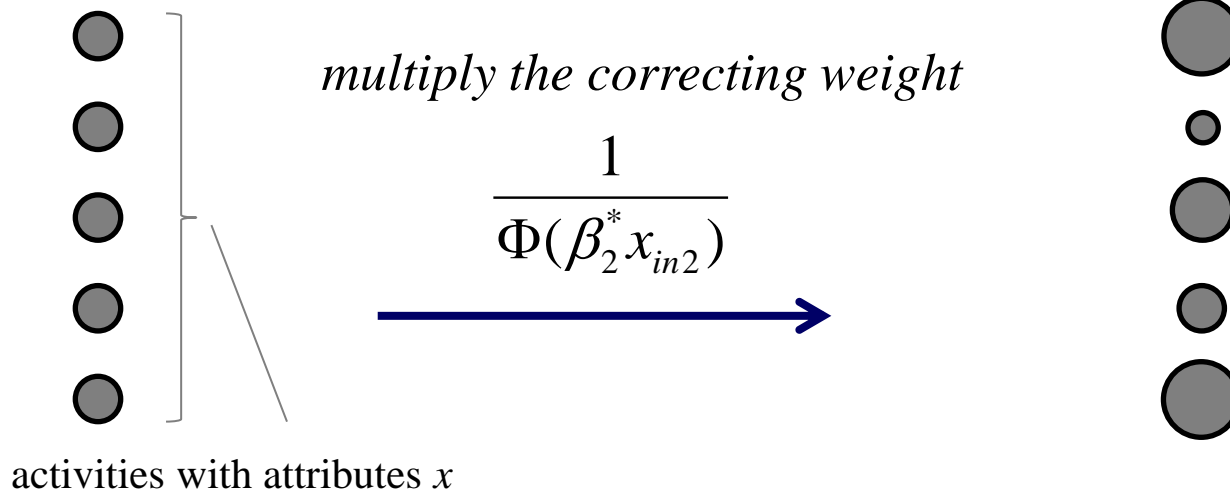
$$w_{in} = \frac{1}{p(y_{in2} > 0 | x_{in2})} = \frac{1}{\Phi(\beta_2^* x_{in2})}$$

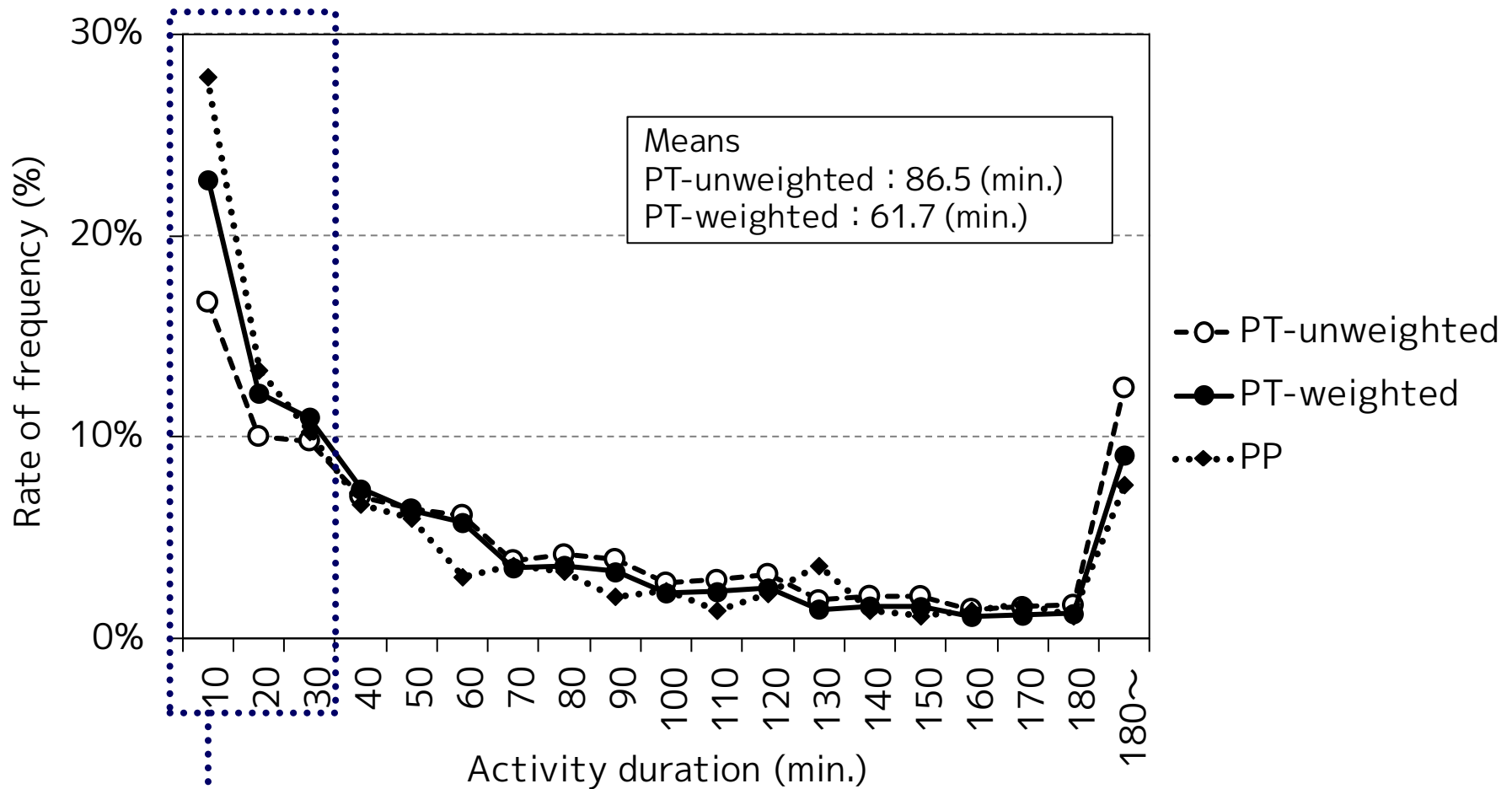
β^* : the parameter estimated in the model

comes from the estimation results

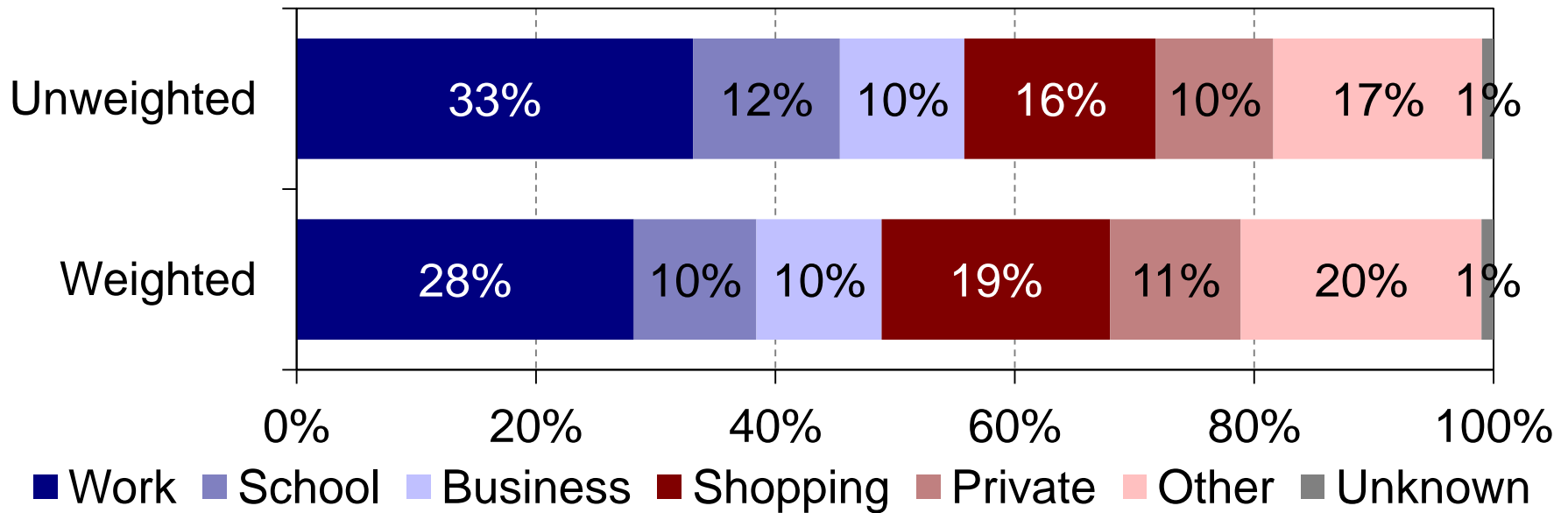
Observation activity data (disaggregate)

Corrected results

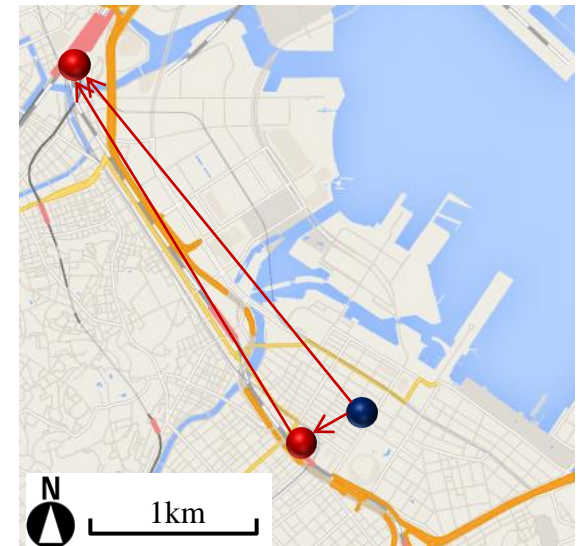
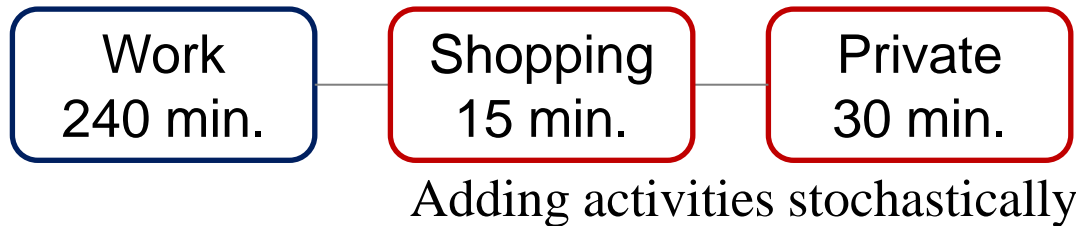




The rate of frequency of weighted PT is similar to PP, which represents the bias of short activities is corrected



The rate of discretionary activities is expanded by weighting.



Comparison between PT and PP

We have discussed the advantages of both new GPS-based PP surveys and conventional PT surveys

Combined estimation using PT and PP data

Introducing the selection model, we show several demographic attributes and activity characteristics associate if activities are missed or not and consider the selection bias

Correcting the sampling bias

By multiplying the inverse of probabilities of observation obtained from the selection model, the bias is appropriately assessed and corrected



Thank you for your attention!



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