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# Transportation Research Record

## Heterogeneous Effects of Pedestrian Environments on Walkability by Spatial Cognition Level and Travel Contexts: A Latent Class Ordered Choice Approach --Manuscript Draft--

<b>Full Title:</b>	Heterogeneous Effects of Pedestrian Environments on Walkability by Spatial Cognition Level and Travel Contexts: A Latent Class Ordered Choice Approach
<b>Abstract:</b>	<p>Measuring the pedestrian walkability is extensively emerged in urban planning and transportation area in the last decade, pointing to the walkable city. Previous studies have supposed to calculate the 'Walkability index', in terms of universal walking accessibility based on the walking facilities and street design indices. This approach implied the homogeneous perception of pedestrians, in order to assess the universal walking accessibility upon the road facilities. However, recent studies giving rise to claim that conventional approach is disesteem to the substances of personal contexts, which gained from the preceding experiences or memories of prior walking trips. The present study is position to the progress for heterogeneous perception of pedestrian satisfactions in relation to the contextual factors, such as sociodemographic attributes, spatial cognition level as the location familiarly, and the trip information. This study employing an advanced discrete choice modelling approach to develop the pedestrian walkability Latent Class Ordered Probit Choice model, it facilitates the influence of all the factors and their effects on the heterogeneity in walkability. The analysis results initialise the distinctive responses of walkability in relation to pedestrian latent groups, which can lead the attention to the variability of individual contextual variables. To this end, the paper presents more profound insights of pedestrian walkability, not only indicate the response of pedestrian environments, but also reveal the distinctive preferences on walkability between individuals.</p>
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1 **Heterogeneous Effects of Pedestrian Environments on Walkability by Spatial Cognition**  
2 **Level and Travel Contexts: A Latent Class Ordered Choice Approach**

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29

1 **ABSTRACT**

2 Measuring the pedestrian walkability is extensively emerged in urban planning and transportation  
3 area in the last decade, pointing to the walkable city. Previous studies have supposed to calculate  
4 the ‘Walkability index’, in terms of universal walking accessibility based on the walking facilities  
5 and street design indices. This approach implied the homogeneous perception of pedestrians, in  
6 order to assess the universal walking accessibility upon the road facilities. However, recent studies  
7 giving rise to claim that conventional approach is disesteem to the substances of personal contexts,  
8 which gained from the preceding experiences or memories of prior walking trips. The present  
9 study is position to the progress for heterogeneous perception of pedestrian satisfactions in relation  
10 to the contextual factors, such as sociodemographic attributes, spatial cognition level as the  
11 location familiarly, and the trip information. This study employing an advanced discrete choice  
12 modelling approach to develop the pedestrian walkability Latent Class Ordered Probit Choice  
13 model, it facilitates the influence of all the factors and their effects on the heterogeneity in  
14 walkability. The analysis results initialise the distinctive responses of walkability in relation to  
15 pedestrian latent groups, which can lead the attention to the variability of individual contextual  
16 variables. To this end, the paper presents more profound insights of pedestrian walkability, not  
17 only indicate the response of pedestrian environments, but also reveal the distinctive preferences  
18 on walkability between individuals.

19  
20 **Keywords:** Walkability, Pedestrian Contexts, Pedestrian Contextual Groups, Latent Class  
21 Ordered Probit Model, Spatial Cognition Level

## 1 INTRODUCTION

2 In recent years, it has increasingly been recognised the ‘walkability’ research, in terms of  
3 measuring the level of pedestrian satisfaction. It becoming a very popular topic for both urban  
4 planners and researchers have interested in the sustainable city (1). Previous studies have  
5 revealed the significant indicators based upon the (pedestrian) environments factors to measure  
6 the walking accessibility (2). Furthermore, empirical studies in US (1, 3–9) mainly complement  
7 to calculate the walkability based on the street design (8, 10) and environmental facilities.  
8 Overall, these studies paid attention to develop the universal measure of walking accessibility,  
9 and also investigate the (dominant) effects of environmental (pedestrian) facilities.

10  
11 However, previous studies have intend the inherent perceptions, i.e., fixed effects, of  
12 pedestrians; (surrounded) environment facilities to account for their walking satisfaction (11,  
13 12). In critique of these conventional studies, this study aiming to audit the heterogenous  
14 perception of walkability response to pedestrian environment, in terms of the distinctive  
15 perceptions of walking facilities of individuals. In this context, walkability may vary with  
16 personal characteristics, such as socio-demographics and spatial cognition levels, and also take  
17 into account for the retained perceptions (i.e. random effects), such as prior walking journeys. In  
18 addition to their retained perceptions, pedestrians recognised their surroundings while walking,  
19 so personal walking experiences, from their routine and habitual daily travels, providing the  
20 distinctive perception of walking environments. It supports to investigate a comprehensive set of  
21 psychological and environmental factors (2, 13).

22  
23 To support this extensive perspective, the latest studies (2, 12, 14–17) claiming the  
24 shortcoming of conventional researches, in which were supposed to be “one size fits all” (6).  
25 These studies also demonstrate the impact of environmental factors on the walkability might be  
26 overwhelmed, yet the substances of individual characteristics and contextual factors are  
27 disesteemed. For instance, Lee et al. (2) estimated the walkability indicators using discrete  
28 choice modelling approach, to exhibit the personal experience and contexts are dominant factor  
29 to influence on walkability than walking facilities and land-use indices. Building on this  
30 foundation, Erath et al. (17) developed a software tool that integrates the results from behaviour  
31 surveys with network-based walking facility datasets, to compute the walkability index  
32 incorporate with user preferences. Additionally, these studies also addressed the heterogeneous  
33 effects on walkability, which generated by the difference in individual trip characteristics:  
34 utilitarian and recreation trips (2, 13, 19) and fixed and flexible trip purpose (12), respectively.

35  
36 In overview, this study supports the alternative approach, not only demonstrate the level  
37 of pedestrian satisfaction, but also examine the heterogenous perception to walking facilities by  
38 pedestrian contextual variables. To achieve this task that the research examines the distinctive  
39 influence of pedestrian environments on walkability coupling with the pedestrian contexts, in  
40 regard to their previous travel experiences; for instance, sociodemographic attributes, spatial  
41 cognition level as the location familiarly, and the trip characteristics. Additionally, external  
42 conditional attributes such as weather conditions and crowdedness, in terms of relative excessive  
43 population volumes in relative to average pedestrian volumes at each survey location by the  
44 course of the day, are also considered as kind of contextual factors. To simultaneously consider  
45 all the factors and their effects on the heterogeneity in individual walkability, an advanced  
46 discrete choice modelling approach (i.e., Latent Class Ordered Probit Choice model) is

1 employed. This approach is allowing to provide more profound insights into the relationship  
2 between pedestrian environments, not only identifying the heterogeneous responses on  
3 pedestrian walkability, which simultaneously estimating why people feel differently (i.e. the  
4 contextual effects), but also interrogating what extent they feel differently with respect to the  
5 pedestrian walkability. Therefore, empirical results using this approach could be deliver the  
6 viable insight to urban planners, who aim to establish walkable cities with enhanced urban streets  
7 for everyone with resulting from a better understanding of walkable pedestrian environment  
8 design criteria. In the present study, the model was estimated based on 83,291 responses of  
9 ‘Seoul Pedestrian Survey’ administered in November 2009 in Seoul, Korea.

10  
11 This paper is structured in the following manner. In the next section, the data deployed in  
12 this research is described, it provides the details of data resources; the survey information and  
13 explanatory variables. In section three, the analytical strategy of the research is present the  
14 deployed approach, i.e., the Latent Class-Ordered Probit Model. The model estimation results are  
15 presented and described at the section four. Finally, we discuss the empirical summarises of the  
16 research findings.

## 17 18 **DATA USED**

19 The ‘Seoul Pedestrian Survey’ was conducted in the whole Seoul area from September to  
20 November in 2009. There were two stages in the survey data collection; on one hand an interview  
21 was taken from respondents, on the other hand various pedestrian environemtal charcteristics were  
22 investigated and recorded, such as pedestrian flow (in volumes per hour), walking road width,  
23 obstacles, pedestrian-support facilities (e.g., fence), and so forth. At each survey location,  
24 surveyors collected responses from 24 pedestrians on Tuesdays, Wednesdays, and Fridays,  
25 yielding 72 responses per location. Up to 2,400 individuals per day were interviewed (Seoul Metro  
26 Government, 2010). The questionnaire collected the pedestrian walkability, the pedestrian’s  
27 activity–travel, and personal information. There were over 1,170 survey locations, totalling 83,291  
28 responses (Lee et al., 2018, page 1).

29  
30 Based on the extensive dataset, we selected important variables and manipulated some of  
31 them to represent pedestrian environment and pedestrian contexts. **Table 1** presents the variables  
32 used for this study. First, the pedestrian walkability is considered which was measured by 3 ordinal  
33 scales: uncomfortable, neutral, and comfortable. The survey asked the respondents to indicate how  
34 much they feel comfortable to walk through the survey area. In addition, the respondents were  
35 asked to provide some information regarding pedestrian contexts such as socio-demographics,  
36 familiarity of the area, and trip characteristics. There are two variables to represent the familiarity  
37 of the area: residential location (whether or not the location is in the same administrative district  
38 with the survey location), and vsiting frequency of the area. The familiarity of the area is employed  
39 to measure the spatial cognitive level of each respondent. We assume that people who live close  
40 to the survey location and/or visit the location very frequently tend to be familiar with the area,  
41 and thus that they have a higher level of spatial cognition with respect to the area. Two variables  
42 were selected to represent the trip characterics of each respondent: trip purpose and travel  
43 companion (whether to walk with others or not).

44  
45  
46

1 **TABLE 1 Variable Lists with Descriptive Information of the Sample**

Variable			Data Type	Attributes	Sample Distribution
Pedestrian walkability			Ordinal	Uncomfortable Neutral Comforable	40.1% 41.9% 18.0%
Pedestrian contexts	Socio-demographics	Gender	Nominal	Male Female	44.6% 55.4%
		Age	Ordinal	Teenager (Ages 15-19) Adults (Ages 20-59) Elderly (Ages over 60)	7.0% 79.3% 13.7%
	Familiarity	Residence in the same administrative district with the survey location	Nominal	Yes (Dweller) No (Stranger)	54.3% 45.7%
		Visiting frequency	Ordinal	Every day 3–5 days/weak 1-2 days/week Rarely (1-3 days for 6 months and first visit)	40.4% 31.2% 14.2% 14.2%
		Trip characteristics	Trip purpose	Nominal	Utilitarian Others
	Travel companion		Nominal	With others Alone	73.0% 27.0%
	Pedestrian environments	Facility characteristics	Walking road width	Ratio	Meter
Presence of central lines			Nominal	Yes No	61.5% 38.5%
Number of total road lanes			Ratio	Lanes	mean: 3.47 std.: 2.64
Presence of walking obstacles			Norminal	Yes No	94.2% 5.8%
Mixed traffic street			Nominal	Yes No (Pedestrian only or with bike)	29.8% 70.2%
Presence of sidewalk fence			Nominal	Yes No	19.2% 80.8%
Presence of crossings			Nominal	Yes No	57.1% 42.9%
Presence of public transit stations within 50m			Nominal	Bus and subway Bus or subway Nothing	7.8% 31.6% 60.6%
Survey ponts characteristics		Crowdedness (Time-based)	Continuous	1 = average	mean: 0.993 std.: 0.374
		Type of spot	Nominal	Main road Residential area Others	14.1% 17.9% 68.0%
		Land-use mix	Continuous	Continuous	mean: 0.93 std.: 0.03

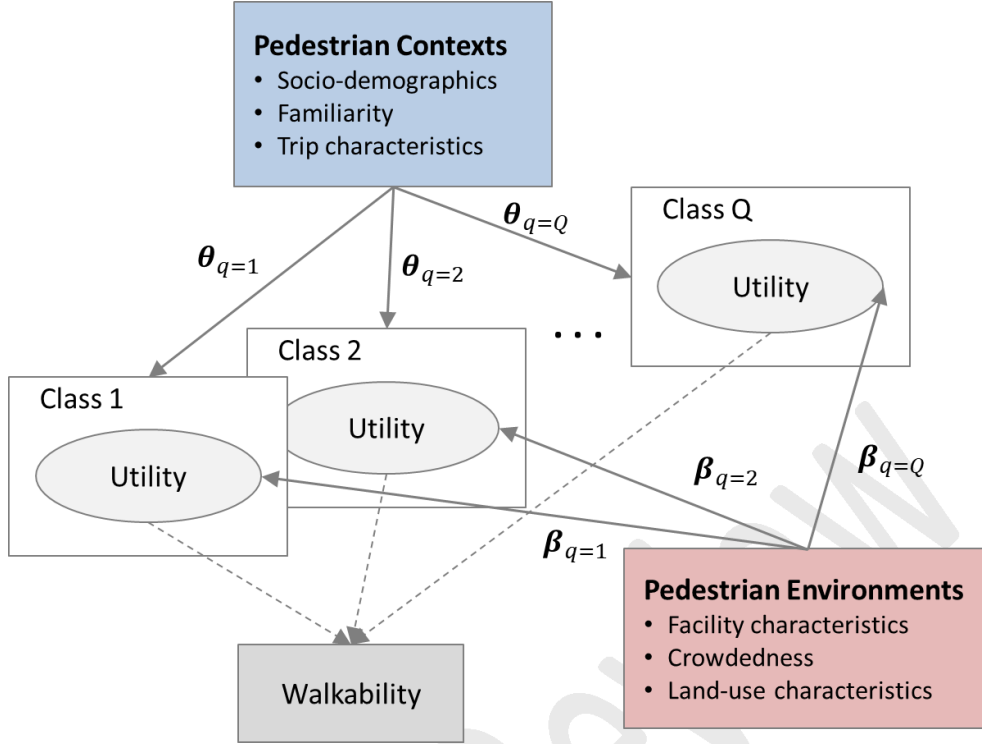
1 Variables representing the conditions of pedestrian environments can be classified into the  
2 facility characteristics and the survey point's characteristics. The pedestrian facilities have  
3 gathered during the survey periods to report the capability of pedestrian facilities, which  
4 encompassing the surrounding area (within 50m ranges) of each survey points. In other words, the  
5 pedestrian facilities depict the multitude typology of survey spots incorporate with the adjacent  
6 arterial road attributes. There are three kinds of environmental facilities, which describing the  
7 amenity level of pedestrian facilities as follows: 1) The constitutional survey spot facilities, such  
8 as the walking road width, and mixed traffic street. 2) The characteristics of adjacency roads, it  
9 comprising the number of total road lanes, and the installation of safeguard amenities for  
10 pedestrians: sidewalk fence, crossings, and walking obstacles. Additionally, in terms of walking  
11 obstacle that retraining the inconvenience of walking experiences to pedestrians, such as street  
12 trees, distribution transformer, street lights, and so forth. Finally, 3) convenience to the public  
13 transits (i.e. accessibility within 50m).  
14

15 Subsequently, the land use information of each survey points are also take into account for  
16 illustrating the typology of survey spots and adjacent arterial road areas. Crowdedness was  
17 measured by the periodic levels of pedestrian volumes in contrast to the average daily pedestrian  
18 volumes at each survey spots. As described in Lee et al (2), the pedestrian volumes may lead to  
19 establish the impression of the area, and further pedestrians may be obtained the walkability scores  
20 whereas retained by their walking experience. To account for this, the time-sensitive crowdedness  
21 index was employed to take account for the variation of pedestrian volumes by four periods per  
22 day at each survey points, if the value is equal to 1, it represents the sporadic pedestrian volume is  
23 equal to the average volume. A value lower or higher than one indicates, reprehensively, a lower  
24 or higher than usual volume.  
25

## 26 **METHODOLOGY**

27 This study employs a latent class choice approach. The underlying theory of the latent class  
28 choice approach assumes that a number of classes exists within the population, and that preferences  
29 of the individuals involved in a class tend to be homogeneous, but each of classes has different  
30 preferences. However, which class contains any particular individual is unknown to the analyst.  
31 The approach estimates simultaneously the probability that an individual is included in each latent  
32 class and the probability that the individual chooses a certain alternative across different classes  
33 (24). This approach is generally accepted in many transportation studies in order to investigate  
34 heterogeneity in travel behaviour. Application domains include, for instance, household location  
35 decisions (25), route choice behaviour of truck drivers (26), bicycle parking preferences of train  
36 travellers (27), preference of alternative-fuel vehicles (28), and departure time choice of train  
37 commuters (29).  
38  
39





1  
2 **Figure 1. Latent Class Order Probit Model framework.**  
3

4 In this study, a latent class ordered probit model is developed to account for the ordinal  
5 nature of the dependent variable (i.e. pedestrian walkability) and the heterogeneous effects of  
6 attributes (i.e. pedestrian environments) on the dependent variable. **Figure 1** depicts the model  
7 framework of this study. We assume that the level of pedestrian walkability indicated by an  
8 individual is associated with the utility of the individual stemming from the pedestrian  
9 environments (i.e. facility characteristics, crowdedness, and land-use characteristics) and that the  
10 utility varies according to the pedestrian contexts with respect to the individual (i.e. socio-  
11 demographics, familiarity, and trip characteristics). Based on this, the probability that individual  
12  $n$  indicates  $j$ th ordinal value for the walkability of pedestrian passage  $l$  can be formulated as  
13 follows:  
14

15 
$$P_{nl}(j) = \sum_{q \in Q} H_n(q) P_{nl}(j|q) \quad (1)$$

16 where  $H_n(q)$  is the probability that individual  $n$  involves in pedestrian-context class  $q$ , which is  
17 called as the class membership probability. The latent class modelling approaches generally  
18 suggest estimating the probability based on the formulation of the multinomial logit model as  
19 follows:  
20

21 
$$H_n(q) = \frac{\exp(\theta_q \mathbf{Z}_n)}{\sum_{g \in Q} \exp(\theta_g \mathbf{Z}_n)} \quad (2)$$
  
22  
23

1 where  $\mathbf{Z}_n$  indicates a  $(M \times 1)$  vector of the pedestrian context variables with respect to individual  
 2  $n$ , and  $\boldsymbol{\theta}_q$  is a  $(1 \times M)$  vector of parameters corresponding to pedestrian-context class  $q$ .

3 In equation (1),  $P_{nl}(j|q)$  is the conditional probability that individual  $n$  indicates  $j$ th  
 4 ordinal value for the walkability of pedestrian passage  $l$  in the context of class  $q$ . We employed a  
 5 system of ordered probit model for representing individual's decision-making process regarding  
 6 ordinal scale of the walkability as follows:

$$7 \quad U_{lnq} = \boldsymbol{\beta}_q \mathbf{X}_{ln} + \varepsilon_{lnq}, \quad (3)$$

$$8 \quad y_{ln} = \begin{cases} 1, & \mu_{q,0} < U_{lnq} < \mu_{q,1} \\ 2, & \mu_{q,1} < U_{lnq} < \mu_{q,2} \\ \dots & \dots \\ j, & \mu_{q,j-1} < U_{lnq} < \mu_{q,j} \\ \dots & \dots \\ J, & \mu_{q,J-1} < U_{lnq} < \mu_{q,J} \end{cases} \quad (4)$$

11 where  $U_{lnq}$  is the utility of individual  $n$  with respect to pedestrian passage  $l$  in the context of class  
 12  $q$ , which consists of the systematic utility  $\boldsymbol{\beta}_q \mathbf{X}_{ln}$  and the random disturbance  $\varepsilon_{lnq}$ .  $\mathbf{X}_{ln}$  indicates a  
 13  $(K \times 1)$  vector of the pedestrian environment variables with respect to pedestrian passage  $l$ , and  
 14  $\boldsymbol{\beta}_q$  is a  $(1 \times K)$  vector of parameters representing the effects of the pedestrian environment  
 15 variables on the utility which varies according to pedestrian-context class  $q$ .  $y_{ln}$  means an ordinal  
 16 value for the walkability indicated by individual  $n$  with respect to pedestrian passage  $l$ . According  
 17 to equation (4), the model assumes that individual  $n$  indicates the  $j$ th ordinal value for the  
 18 walkability if the utility  $U_{lnq}$  lies within a particular range between  $\mu_{q,j-1}$  and  $\mu_{q,j}$ . The cut-off  
 19 parameters  $\boldsymbol{\mu}_q = (\mu_{q,0}, \dots, \mu_{q,J})$  for each pedestrian-context class should be estimated. By  
 20 assuming the probability density of the random disturbance  $\varepsilon_{lnq}$  as the normal distribution, the  
 21 choice probability  $P_{nl}(j|q)$  can be expressed as follows:

$$22 \quad P_{nl}(j|q) = \Phi(\mu_{q,j} - \boldsymbol{\beta}_q \mathbf{X}_{nl}) - \Phi(\mu_{q,j-1} - \boldsymbol{\beta}_q \mathbf{X}_{nl}) \quad (5)$$

23 where  $\Phi(\cdot)$  represents the cumulative density function of the standard normal distribution.

24 To estimate the unknown parameters, the maximum likelihood method is employed in the  
 25 present study. All the parameters associated with the class membership and the choice probability  
 26 functions are estimated simultaneously. On the other hand, the optimal number of latent classes  
 27 ( $Q$ ) cannot be estimated simultaneously. Typically, the optimal number of classes is chosen based  
 28 on a comparison of the model fit of a series of models with different number of classes. To measure  
 29 and compare model fit, the present study employs Akaike information criterion (AIC) and  
 30 Bayesian information criterion (BIC). In the both criteria, a lower value indicates a better  
 31 goodness-of-fit.

## 32 ANALYSIS RESULTS

33 To identify the optimal number of latent classes, a series of models were estimated. Error!  
 34 Reference source not found. presents the model fit of the one-, two-, three-, and four-latent class  
 35 models. The value of both AIC and BIC continues to decrease as the number of class increase  
 36 indicating that more latent classes result in a better fit. Considering the model fit and the

1 interpretability of classes, we have selected the four-class model as the optimal model for  
 2 investigating the heterogeneous effects of pedestrian environments on walkability. Error!  
 3 Reference source not found. provides the estimation results of the one-class model (Model 1) and  
 4 the four-class model (Model 2).

5  
 6

**TABLE 2. Model Fit for the 1 to 4 Latent Class Membership Model.**

<b>Number of classes</b>	1	2	3	4
<b>Number of parameters</b>	15	40	65	90
<b>Log-likelihood value</b>	-84461.85	-84087.20	-83750.57	-83125.56
<b>AIC</b>	168953.69	168254.40	167631.15	166431.12
<b>BIC</b>	169093.45	168627.08	168236.75	167269.65

7  
 8

9 Model 1 presents the effects of the pedestrian environments on the walkability that are  
 10 assumed to be homogeneous across different pedestrians. According to the estimation results,  
 11 people tend to feel more walkable for a wider passage, and most of walking facilities tend to  
 12 represent a positive response. In contrast to common safety considerations, the existence of central-  
 13 line and mixed streets types exhibited a negative influence on the pedestrian satisfaction.  
 14 Subsequently, pedestrians perceived a place more walkable when the area was characterised by a  
 15 diversity of zoning types, yet too much crowding is experienced negatively as shown by the  
 16 estimated coefficients. Furthermore, the residence indicator was also selected as a significant  
 17 variable to influence the pedestrian walkability. In particular, the Seoul inhabitants, who hold an  
 18 extensive spatial cognition of their routinely visited areas, “perceived distress”. The visiting  
 19 frequency are used to unique define a routine visitor (i.e. every day and 3-5 days/week visitors).

20 Before moving on to the Latent Class Ordered Probit model (Model 2) estimation results,  
 21 the methodological approach of this study is as follows: the Latent class framework was applied  
 22 in this study. The Latent class framework indicated that relatively few classes are sufficiently  
 23 flexible to capture quite complex patterns of heterogeneity (30). In this context, Latent class choice  
 24 models are appropriate to distinguishing particular classes under the assumption that individuals’  
 25 preferences in the same classes (i.e. membership groups) are homogeneous (25, 26).

26

27 From the estimation results of both Model 1 and 2, the distinctive correlation patterns of  
 28 pedestrian environments on walkability were estimated. As stated before, the walking facilities,  
 29 particularly the presence of fences, crossings, and trees as obstructions, were found having a  
 30 positive influence. However, vehicle-oriented facilities were discovered to have negative effects  
 31 on walkability.

32 Furthermore, the indicator of crowdedness of the street was estimated to have a negative  
 33 influence. A mixed land-use index represents positive effects on walkability. The estimated results  
 34 show that accessibility, which comprises the accessible distance for walking, generally positively  
 35 contributes to pedestrian satisfaction levels (10) However, the crowdedness, the density of  
 36 pedestrians, may lead to a negative influence on walkability (31). In this context, the pedestrian  
 37 profiles may become the decisive factors to assess the pedestrian satisfaction level. Individual  
 38 contextual attributes thus lead to distinctive responses in the choice of walkability as demonstrated  
 39 by the Model 2 estimation results. In particular, it was discovered that dwellers (both in Seoul and  
 40 Metropolitan area) tend to be more sensitive for responding the negative walkability than visitors.  
 41 This may be explained by the spatial cognition and the density of residence in particular.

1 Model 2, latent class ordered probit model, first the four classes (the optimal number of  
2 latent membership groups according to the model estimation results using NLogit 4 software) were  
3 estimated. The profiles of latent classes were investigated, as it is the evidence shows that the  
4 characteristics of respondents are not only influenced by the choice of walkability, but also by the  
5 preferences of each class (being characterized by distinctive conditions and contexts). In other  
6 words, this study contribute to the development of latent class choice approach in walkability  
7 model, not only to distinguish the significant factors, but also explore the psychological indicators  
8 and preserved contexts for testing the stated preference experiments. The latent class modelling is  
9 deployed as an appropriate technique to detect the heterogeneous preferences in the contextual  
10 variables, according to the theoretical review. In the next section, each latent classes is explored,  
11 and distinct responses to their choices of (pedestrian) satisfaction are discovered in relation to  
12 pedestrian context membership groups.

13 Class 1: Male ‘Walk-holic’ adults, who lived in the metropolitan area, outside of Seoul.

14 Class 2: ‘Female Seoul inhabitants’, who generally walking together with companion

15 Class 3: ‘Female Seoul commuters’, who usually walking together with companion

16 Class 4: Male ‘Saunters’, who dwelled in metropolitan area

17  
18 This section describes the relationship between pedestrian environments and classes; that  
19 is the heterogeneous responses (coefficients) of each pedestrian context group to pedestrian  
20 environments from the estimation results (see Error! Reference source not found.).

21 Group 1 around a quarter (23.9%) of pedestrians belongs to this class. It is termed as ‘walk-  
22 holic’ male adults. This group has a very strong base dependence on land-use variables, such as  
23 the mixed land-use index and road types, especially in main-road survey spots. It can be interpreted  
24 as the local facilities may cause to the significant indicator to determine the preference of  
25 pedestrian walkability of Group 1.

26 Contradicting previous claim of Group 2 and group 3 are seems to be very similar classes,  
27 which being predominantly populated with Seoul female dwellers. Yet their trip purpose is  
28 distinctive, i.e. either walking commuters (i.e. utilitarian walking) or walkers for their discretionary  
29 travel purpose. Group 2 seems to be unique latent class in pedestrians because their response to  
30 pedestrian environments dramatically differs from overall tendency (i.e. the estimation results of  
31 Model 2). The female Seoul dwellers seem to be much stressed from crowdedness during walking,  
32 nor mixed land-use index which to be taken place the much volume of pedestrians in the walking  
33 street. In this sense, each latent membership group has a distinctive spatial cognition level  
34 depending on the locations. In particular, the locals who retained the knowledge of spatial  
35 recognition stated a positive satisfaction level for their walking experiences. Furthermore, they  
36 reacted also positively to accessible public transportation facilities, for which they have stronger  
37 preferences than any other classes.

38 Moreover, Group 3 mainly comprises the ‘Seoul female commuters’, who prefer walking  
39 together with someone for their commuting. As we discussed before, groups 2 and 3 have similar  
40 profiles of each group members, however their response for pedestrian environments are not  
41 identical. In other words, utilitarian pedestrian groups have a more negative influence than other  
42 travel purposes. Group 3 shows this difference more clearly. The membership groups have a  
43 negative effect from accessible public transportation facilities, yet they are positive at main roads  
44 survey spots and crowded walking streets. This preference can be interpreted in relation to their  
45 habitual and routine walking experiences which were accumulated during their previous  
46 commuting walking trips. Both subway gates and bus stops in Seoul are crowded during the peak

1 hours, as well as the main-streets, which are consistently a high dense area for moving pedestrians  
2 during the daytime.

3 Finally, Group 4, which is the largest class (28.7%), represents a large number of local  
4 metropolitan males who prefer to walk alone. It has a marginally different response compared to  
5 the overall tendency, especially for the width of walking street, mixed road types, and crossings.  
6 However, it may lead to investigate the importance of trip purpose, which caused the distinctive  
7 pedestrian walkability choice.

8  
9 In summary, we identified the effects of pedestrian contexts on walkability in the  
10 estimation results of Model 2, and compared it with Model 1. The most important indicators were  
11 identified from the estimation results: trip purpose and spatial cognition, and gender. In addition,  
12 the estimation results for Model 2 revealed the coefficients among the pedestrian environmental  
13 factors on walkability, which influence the pedestrian walkability between discrete membership  
14 groups by pedestrian contexts.

15 To be specific, 'Female Seoul dwellers', as membership group 2 in Model 2, yields the  
16 most heterogeneous response on the pedestrian environments compared to other groups, which is  
17 in contrast to the Model 1 estimation results. Additionally, the gender information is a substantial  
18 indicator for the discrete latent classes: Males (Group 1 and 4) and Females (Group 2 and 3). Each  
19 gender group shares some common characteristics regarding their pedestrian contexts; however,  
20 they have distinctive choice preferences regarding the pedestrian environments, in relation to  
21 walkability.

22 As indicated previously, the analysis results stem from the participants' historical  
23 experiences of walking and their spatial knowledge. As far as walkability is concerned, the  
24 psychological conditions and acknowledged experiences of pedestrians in respect to their routine  
25 walking behaviour may produce distinctive (individual) stated preferences for their walkability  
26 choices. From the extensive findings of this analysis, it was discovered that the walking-trip  
27 purposes and survey spots associated with the accumulated spatial cognition are substantial  
28 indicators to establish the perceived preference of each membership group.

29

1 **TABLE 3. Model Estimation Results**

Model Component	Attributes		Model 1 (The one-class model)			Model 2 (The four-class model)											
						Class 1 (23.9%)			Class 2 (20.3%)			Class 3 (27.1%)			Class 4 (28.7%)		
			B	Std.	Z.	B	Std.	Z.	B	Std.	Z.	B	Std.	Z.	B	Std.	Z.
Pedestrian walkability choice part	Constants		<b>0.5132</b>	0.0215	23.92	<b>1.0480</b>	0.1901	5.51	<b>-0.9515</b>	0.2005	-4.75	<b>0.8316</b>	0.1495	5.56	<b>1.4007</b>	0.1638	8.55
	Walking road width		<b>0.0214</b>	0.0020	10.97	<b>0.0500</b>	0.0172	2.90	<b>0.3694</b>	0.0259	14.25	<b>-0.0408</b>	0.0121	-3.37	<b>-0.0467</b>	0.0128	-3.67
	Presence of central lines		<b>-0.1168</b>	0.0068	-17.24	<b>-0.4493</b>	0.0679	-6.62	<b>0.6340</b>	0.0863	7.35	<b>-0.1487</b>	0.0439	-3.38	<b>-0.4222</b>	0.0562	-7.51
	Number of total road lands		<b>0.0557</b>	0.0023	24.76	<b>-0.0807</b>	0.0256	-3.16	<b>0.2708</b>	0.0248	10.94	<b>-0.1074</b>	0.0201	-5.33	<b>0.2396</b>	0.0270	8.88
	Presence of walking obstacles		<b>0.0375</b>	0.0085	4.44	<b>0.4340</b>	0.0655	6.63	<b>-0.6596</b>	0.0887	-7.44	<b>0.2167</b>	0.0508	4.26	<b>-0.0819</b>	0.0517	-1.59
	Mixed traffic street		<b>-0.0695</b>	0.0067	-10.32	<b>-0.7072</b>	0.0797	-8.87	<b>-0.0095</b>	0.0671	-0.14	<b>0.0421</b>	0.0396	1.06	<b>0.0944</b>	0.0482	1.96
	Presence of sidewalk fence		<b>0.0147</b>	0.0054	2.73	0.0553	0.0507	1.09	<b>-0.3339</b>	0.0473	-7.05	<b>0.1384</b>	0.0301	4.59	<b>0.0766</b>	0.0339	2.26
	Presence of crossings		<b>0.0249</b>	0.0050	5.01	<b>0.2826</b>	0.0508	5.57	<b>0.5210</b>	0.0530	9.84	<b>-0.0924</b>	0.0321	-2.88	<b>-0.2147</b>	0.0412	-5.21
	Presence of stations in 50m	Bus and Subway	<b>-0.0410</b>	0.0104	-3.95	0.0161	0.0865	0.19	<b>0.5539</b>	0.1057	5.24	<b>-0.3016</b>	0.0481	-6.27	<b>-0.2408</b>	0.0632	-3.81
		Bus or Subway	<b>0.0282</b>	0.0071	4.00	<b>-0.2511</b>	0.0665	-3.78	<b>-0.2464</b>	0.0668	-3.69	<b>0.2974</b>	0.0414	7.19	<b>0.1080</b>	0.0453	2.38
		Nothing	0.0128	-	-	0.2350	-	-	-0.3075	-	-	0.0042	-	-	0.1328	-	-
	Crowdedness (Time-based)		<b>-0.1149</b>	0.0106	-10.86	<b>-0.4778</b>	0.0915	-5.22	<b>-0.7274</b>	0.1014	-7.18	<b>0.3057</b>	0.0656	4.66	<b>-0.1569</b>	0.0669	-2.35
	Land-use mix		<b>0.3001</b>	0.0164	18.30	<b>1.1412</b>	0.1521	7.50	<b>-0.2289</b>	0.1465	-1.56	<b>0.3525</b>	0.0984	3.58	0.0414	0.0995	0.42
	Type of spot	Main road	-0.0046	0.0083	-0.55	<b>0.3715</b>	0.0787	4.72	<b>-1.4633</b>	0.1106	-13.23	<b>0.6160</b>	0.0711	8.66	-0.0654	0.0656	-1.00
Residential area		<b>-0.0395</b>	0.0082	-4.84	<b>-0.8551</b>	0.0856	-9.99	<b>2.3654</b>	0.1577	15.00	<b>-1.1001</b>	0.1034	-10.63	<b>0.1934</b>	0.0886	2.18	
Others		0.0441	-	-	0.4836	-	-	-0.9022	-	-	0.4841	-	-	-0.1279	-	-	
Pedestrian – context class membership part	Constants		-	-	-	<b>-0.3087</b>	0.1466	-2.11	<b>-0.3938</b>	0.1213	-3.25	-0.1690	0.2202	-0.77	-	-	-
	Gender	Male	-	-	-	<b>-0.0950</b>	0.0265	-3.58	<b>-0.1412</b>	0.0276	-5.11	<b>-0.1851</b>	0.0398	-4.65	-	-	-
		Female	-	-	-	0.0950	-	-	0.1412	-	-	0.1851	-	-	-	-	-
	Age	Teenager	-	-	-	<b>-0.6353</b>	0.0742	-8.56	<b>-0.5595</b>	0.0733	-7.63	<b>-0.7312</b>	0.1110	-6.59	-	-	-
		Adults	-	-	-	<b>0.1108</b>	0.0412	2.69	0.0210	0.0422	0.50	<b>0.2237</b>	0.0696	3.22	-	-	-
		Elderly	-	-	-	0.5245	-	-	0.5385	-	-	0.5076	-	-	-	-	-
	Region	Dweller	-	-	-	<b>-0.2525</b>	0.0327	-7.72	<b>0.1339</b>	0.0283	4.74	<b>0.1483</b>	0.0436	3.40	-	-	-
		Stranger	-	-	-	0.2525	-	-	-0.1339	-	-	-0.1483	-	-	-	-	-
	Visiting frequency	Every day	-	-	-	<b>0.1698</b>	0.0462	3.67	<b>0.3779</b>	0.0480	7.88	<b>0.3077</b>	0.0662	4.65	-	-	-
		3-5 days/week	-	-	-	0.0701	0.0454	1.54	<b>-0.1132</b>	0.0453	-2.50	<b>-0.3393</b>	0.0692	-4.90	-	-	-
		1-2 days/week	-	-	-	<b>-0.1162</b>	0.0511	-2.27	<b>-0.1804</b>	0.0533	-3.38	<b>-0.1368</b>	0.0829	-1.65	-	-	-
	Rarely		-	-	-	-0.1238	-	-	-0.0843	-	-	0.1684	-	-	-	-	-
			-	-	-	-	-	-	-	-	-	-	-	-	-	-	-
	Trip purpose	Utilitarian	-	-	-	<b>0.0444</b>	0.0270	1.65	<b>0.0606</b>	0.0279	2.17	<b>0.1199</b>	0.0427	2.81	-	-	-
Others		-	-	-	-0.0444	-	-	-0.0606	-	-	-0.1199	-	-	-	-	-	
Travel companion	Alone	-	-	-	<b>-0.0644</b>	0.0354	-1.82	<b>-0.2203</b>	0.0353	-6.23	<b>-0.2698</b>	0.0465	-5.81	-	-	-	
	With someone	-	-	-	0.0644	-	-	0.2203	-	-	0.2698	-	-	-	-	-	

Note: Estimates whose p-values are less than 0.05 are marked in bold.

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1  
2 **CONCLUSION**

3 With respect to the comparison between discrete choice model estimation results, a  
4 substantial influence of pedestrian contexts on walkability was observed, which not only accounts  
5 for the perceived preferences, but also produces the discrete membership groups in terms of latent  
6 classes.

7 The estimation results of Latent Class Ordered Probit model, provided evidence for the  
8 distinctive responses (i.e. diverse coefficients values) of walkability choices, in relation to  
9 individuals' contextual characteristics. These findings using a Latent class ordered-probit model  
10 indicated the significant influence of pedestrian contexts, which can lead to the perceived  
11 pedestrian walkability. Furthermore, the analysis results of this study support that perceived  
12 experiences and memories of individual groups may influence the walkability choices.

13 Moving on to the individual variability, i.e., pedestrian contextual attributes, such as spatial  
14 cognition level and trip characteristics; these were shown to behave as the most significant  
15 indicators, more than the pedestrian environments (especially facilities on the walking road),  
16 which is the fruitful information for urban planners and decision makers.

17 Finally, this study also initialises and updates the knowledge of the relationship between  
18 pedestrian environments and their perceived walkability with respect to personal contextual  
19 variables. It thus provides valuable information to urban planners and decision makers to update  
20 their best knowledge so they could design better walkable pedestrian environments, not only  
21 concerning the residential inhabitants, but also respect the time-specific visitors, i.e. visiting  
22 populations.

23  
24 Although the contributions of this study, i.e. the influence of environmental factors on  
25 walkability, and the significantly distinctive response in relation to the pedestrian (behavioural)  
26 contexts, are considerable, this study is also limited to appraise how much the heterogeneous  
27 individual contexts affects pedestrian walkability. This challenge may be addressed in futures  
28 studies, which could employ advanced discrete choice modelling approaches.

29  
30 In summary, future studies on the current topic, that is to investigate the external contextual  
31 attributes associated with daily activity-travel patterns, are recommended. For instance, weather  
32 conditions and the crowdedness by periods of day in a given area may generate distinctive choices  
33 of walkability according to the individuals' routine activities.

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