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# ASSESSING THE PEDESTRIAN EXPERIENCE IN PUBLIC SPACES

#### **Ioannis Kaparias**

Centre for Transport Studies, Department of Civil and Environmental Engineering, Skempton Building, South Kensington Campus, Imperial College London, London SW7 2BU, UK Email: ik00@imperial.ac.uk

## Michael G. H. Bell

Centre for Transport Studies, Department of Civil and Environmental Engineering, Skempton Building, South Kensington Campus, Imperial College London, London SW7 2BU, UK Email: mghbell@imperial.ac.uk

#### **Edward Gosnall**

Centre for Transport Studies, Department of Civil and Environmental Engineering, Skempton Building, South Kensington Campus, Imperial College London, London SW7 2BU, UK Email: edward.gosnall07@imperial.ac.uk

#### **Daban Abdul-Hamid**

Centre for Transport Studies, Department of Civil and Environmental Engineering, Skempton Building, South Kensington Campus, Imperial College London, London SW7 2BU, UK Email: daban.abdul-hamid06@imperial.ac.uk

#### **Michael Dowling**

Centre for Transport Studies, Department of Civil and Environmental Engineering, Skempton Building, South Kensington Campus, Imperial College London, London SW7 2BU, UK Email: michael.dowling06@imperial.ac.uk

#### Ishaan Hemnani

Centre for Transport Studies, Department of Civil and Environmental Engineering, Skempton Building, South Kensington Campus, Imperial College London, London SW7 2BU, UK Email: ishaan.hemnani07@imperial.ac.uk

#### **Bill Mount**

Centre for Transport Studies, Department of Civil and Environmental Engineering, Skempton Building, South Kensington Campus, Imperial College London, London SW7 2BU, UK Email: bill.mount49@googlemail.com

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# ASSESSING THE PEDESTRIAN EXPERIENCE IN PUBLIC SPACES

# I. Kaparias, M.G.H. Bell, E. Gosnall, D. Abdul-Hamid, M. Dowling, I. Hemnani, B. Mount

Centre for Transport Studies, Imperial College London, UK

#### ABSTRACT

The assessment of the pedestrian experience in public spaces is increasingly becoming an essential constituent of urban street design. This paper first presents a new methodology for evaluating pedestrian environments through on-street surveys, building upon well-established comprehensive pedestrian audit tools, such as PERS and PEDS. The methodology is applied on the South Kensington area of London, in light of recent redevelopments seeing the conversion of the previous car-oriented layout to a more pedestrian-friendly one. The results suggest that the new design is generally perceived positively by pedestrians, but point out that there may be room for improvement in terms of pedestrian comfort. The results are then further analysed statistically in order to draw generic conclusions and investigate the effects of different aspects of the pedestrian environment on each other with respect to the pedestrian experience. By fitting a series of ordered logistic regression models, a number of interdependences are identified and interpreted.

## **INTRODUCTION**

Traditionally, street design and traffic engineering have been driven by the concept of segregation of pedestrians and vehicles. As such, the priority has been to allow for quicker access and movement of the vehicular traffic by limiting conflicts and human-dependent decisions and by designing streets according to the desire lines of vehicle drivers. As opposed to that, the desire paths of pedestrians have often been neglected, with structures such as pedestrian subways, bridges, guardrails and walls restricting their movement with the objective of protecting them and preventing their involvement in road accidents. The concept of segregation is set out most lucidly in Buchanan's "Traffic in Towns" report (1) which served as a street design manual in the UK for many decades.

In recent years, however, there has been a trend away from traffic segregation, driven by developments in architecture and urban planning. Segregation has been deemed by some detrimental for urban environments due to its perception as resulting in "the domination of vehicular traffic and associated noise and air pollution alongside street clutter and ugly surroundings" (2). Instead, street design and traffic engineering have seen a shift in focus from vehicles to pedestrians as a means of creating a better environment, mainly through the introduction of single surfaces and the removal of features such as street furniture, signage, delineation and kerbs. Notable examples include the Complete Streets initiative in the USA (3) and the shared space concept in the UK (4).

In light of the increasing importance of catering for pedestrians in urban street design and traffic engineering, the review and assessment of the pedestrian environment is becoming an essential constituent of any new scheme or redevelopment, alongside the conventional assessment of the traffic efficiency, before and after implementation. This is generally carried out using either well-established comprehensive pedestrian audit tools (such as PERS in the UK or PEDS in North America) addressed to small numbers of experts and investigating individual aspects of a design (e.g. the provision of adequate pedestrian crossing facilities), or through shorter purpose-developed general surveys targeting larger numbers of respondents from the general public and thus aiming at broader but statistically significant results.

The present study has two goals: 1) to develop a pedestrian evaluation methodology of the latter category for use in on-street surveys and to apply it to a specific area in order to draw conclusions on the pedestrian environment; and 2) to investigate the effects of different aspects of the

pedestrian environment on each other with respect to the pedestrian experience. In particular the second goal aims at investigating the previously unexplored topic of the interrelationships between the factors constituting the pedestrian experience (perception), with the prospect of being used in predictive models in future, in cases with limited data availability or with constraints in the data collection (e.g. survey time).

The study focuses on London's South Kensington area, which has recently been redeveloped as part of the Exhibition Road project. The project comprises, among others, the implementation of a single shared surface as a replacement of a conventional dual carriageway, the unravelling of a pedestrian-unfriendly one-way system, and the provision of more pedestrian facilities (Figure 1). The work described has been carried out as part of a traffic monitoring programme of the Exhibition Road project. Other activities conducted within the framework of the monitoring programme include traffic conflicts analyses (5), behavioural studies (6), and generic shared space perception surveys (7,8).



FIGURE 1 South Kensington before and after the redevelopment

The paper is structured as follows: Section 2 sets out the background of the study, which includes a review of existing pedestrian evaluation methods. Section 3 describes the survey design methodology and reports on the results obtained. Section 4 then goes on to investigate the interrelationships between the individual aspects of the pedestrian environment and to present the corresponding statistical analysis through a series of regression models. Finally, Section 5 concludes the paper and identifies areas of future research.

#### 2 BACKGROUND

A variety of methods have been developed with the aim of assessing the street environment from the pedestrians' point of view in order to be able to address their needs. These methods usually involve the conduct of surveys with pedestrians, covering aspects such as the accessibility of the area, the availability of the required street furniture, the availability of the required services, the aesthetic appearance and the cleanliness of the space. Such an approach is adopted, for example, in a study by Jones et al (2), who analyse a street space in London by asking users about their levels of satisfaction when using the space. However, since external environments are subject to users who have different tastes and needs, what is important for one user may be less important for another, thus making it a necessity to find a common standard for the comparison.

Previous research has attempted to address the topic of pedestrian environment assessment by identifying the key influencing factors and developing appropriate tools. While attempts have been made to evaluate pedestrian facilities objectively, it has been highlighted in the literature that many of the factors involved are subjective in nature, and as such studies have adopted various approaches (9).

Examples include: a method presented by Sarkar (10), which defines a level of service measure for pedestrian movement based on safety, security, comfort and convenience, continuity, system coherence, and attractiveness; a study by Pikora et al (11), evaluating the quality of walking as a function of four criteria (functional, safety, aesthetic and destinations); and the works of Emery et al (12) and Day et al (13), which assess features such as the condition of pavements and sidewalks, safety, lighting, aesthetics, and public transport facilities. Further methods and tools are presented and appraised in a study by Vernez-Moudon and Lee (14).

An important contribution to the field has been made by Clifton et al (15) through the development of the "Pedestrian Environment Data Scan" (PEDS) tool, which is currently one of the most widely-used tools in the USA. Focussing on individual streets rather than districts or areas, PEDS consists of a single review sheet containing three sections. The first section aims at defining the street to be reviewed, which is down to the discretion of the reviewer but is typically the pavement segment between two junctions; the second section allows for a small number of subjective scores to be given (assessing attractiveness and perceived safety); and the third section, which is the main part of the form, requires quantitative scores input to four sets of questions covering the categories of environment, pedestrian facility, road attributes and walking/cycling environment. In total 40 questions are included, and to ensure the integrity of the respondents, a two-day training is required prior to using the tool.

Equivalently to PEDS, the "Pedestrian Environment Review System" (PERS) has been developed in the UK by the Transport Research Laboratory (TRL) (16). Implementing a three-tier system, where each level offers a more detailed assessment of the previous level (Tier 1: Public space -> Tier 2: Route -> Tier 3: Link, Crossing), PERS is based on the completion of a series of separate independent review forms. A wide range of parameters are covered by the review framework, relating to different aspects of urban design features, such as safety, legibility of space, user conflicts, and walking surface quality. Respondents are required to assign scores between -3 and +3 to each of the parameters, which can be subsequently weighted according to their importance (17) in the pedestrians' perceived utility. Overlaying the results from PERS on a map enables their visualisation, which offers a quick insight into the street space under analysis.

While PEDS and PERS are fairly comprehensive analysis tools for pedestrian environments and offer standardised methods of measuring pedestrian views, they are mainly intended for comprehensive interviews with experts. In fact, the completion of the numerous review forms in the case of PERS and the previous training required for the respondents of PEDS are both timeconsuming procedures and are not suited for the conduct of on-street surveys with pedestrians. What is more, both tools require very detailed information from the respondents about all parameters, which is mostly not noticed by "ordinary" pedestrians. Conversely, a more general assessment of the pedestrian perceptions of a space through a shorter purpose-developed on-street questionnaire is often the preferred approach, as it enables interviewing larger numbers of respondents and thus collecting less-detailed but statistically-significant data.

An example of a study of that category is a 2007 assessment of the pedestrian perceptions at London's Oxford Circus site. For the purposes of the study an on-site questionnaire was designed and implemented with the prospect of identifying the issues relating to movement within and around the space. Surveys were held on two different study days: one during a specially organised pedestrianised day (called Very Important Pedestrians (VIP) day, held on 1 December 2007), on which vehicles were severely limited from passing through the area; and one on a "normal" day, held a week later, so as to compare the results with everyday conditions. The study was commissioned by Transport for London and was carried out by Atkins Intelligent Space (18). The surveys enquired visitors to the area about their perception of public transport services, way-finding into and around the area, the ease of moving around the space and their perceived safety. A demographic dataset describing the respondents was also collated in the process. The study eventually helped clarify important problems relating to the pedestrian environment, in light of the subsequent redevelopment of the space in 2010, and was also

coupled with an after-study using a similar questionnaire, which demonstrated improvements in the aspects identified (19).

A similar evaluation is described in the next section, where a new questionnaire has been developed and tested on the South Kensington site.

## **3 SURVEY DESIGN AND RESULTS**

In order to evaluate the pedestrian experience in the South Kensington area following the redevelopments within the framework of the Exhibition Road project, a purpose-developed short onstreet questionnaire was developed and was applied on a large sample of respondents. The study area is briefly described first, followed by the questionnaire development and the discussion of the results.

#### **3.1** The study area

South Kensington station is one of the busiest underground stations in London with around 30 million users every year (20), as it is the gateway to Exhibition Road, which is home to a number of London's museums, academic institutions and events venues. Before the redevelopment of the immediate surrounding area, traffic congestion, overcrowding and high numbers of pedestrian-vehicle conflicts were significant problems affecting the pedestrian environment. As part of the Exhibition Road project, the road layout was radically altered, with many of the previous one-way roads becoming two-way and with additional pedestrian crossings being introduced. The phasing of traffic signals has also been improved, and large open pedestrian spaces have been provided to the North and South of the station. These spaces have been designed to provide a more attractive and comfortable environment for pedestrians to interact and relax, rather than a busy traffic intersection where pedestrians would rush through.

The study area analysed here is shown in Figure 2, and consists of the immediately adjacent space of South Kensington station, including five pedestrian crossings and two pedestrian spaces surrounding the station entrance.



#### **3.2** Questionnaire development

The design of the questionnaire used in the on-street survey was an important task, as careful thought needed to go into the selection of the attributes examined and the formulation of the questions. As the number of questions directly influences the amount of information gained but also the time taken to complete the questionnaire, a trade-off between brevity and level of detail was required. Naturally, respondents may become restless and frustrated with an overlong questionnaire, while too few

questions would limit the amount of information that could be obtained. Looking at previous on-street questionnaires, it was decided in this study that 10 questions were required.

	Imperial College London			Study of South Kensington Station area improvements					
1.	Age: Under 21	21-	-30 🔀	31-40	41-5	0 🛛 51	- 60 🔀	60+ 🔀	
2.	Gender: Male	F	emale 🔀	)					
3.	How often do y	<b>you visit ti</b> infrequ <mark>en</mark> t	ne South I	Kensingto at least or	n Station	area?			
	at least once a w	reek 🔀	at least 5	times a we	ek 🗡				
4.	How do you ra	te the visi	ual enviro	nment of	the area	(street fu	rniture, pa	ving, <mark>cleanlin</mark>	ess, et
		-3	-2	-1	0	+1	+2	+3	
	Comments:								1078
5.	How easy do y	ou find me	oving arou	un <mark>d in t</mark> he	area? (o	bstructions	, congestio	on, signage, e	etc)
		-3	-2	-1	0	+1	+2	+3	
	Comments:	32 KG	48 (K. 1968)	6985	0.05	(a) (a			
6.	How do you ra	te the con	nfort of th	e area? (I	noise, sea	ting, shelte	er, etc)		
		-3	-2	-1	0	+1	+2	+3	
	Comments:		-						
7.	How do you ra	te the safe	ety of the	area?					
		-3	-2	-1	0	+1	+2	+3	
	Comments:								
8.	How do you ra you rather cros	te the pos ss elsewhe	itioning o ere?)	f the ped	estrian cr	ossings? (a	re they ou	t of your way	? woul
		-3	-2	-1	o	+1	+2	+3	
	Comments:								
9.	How do you ra	te the wai	ting time:	s of the cr	ossings?				
		-3	-2	-1	o	+1	+2	+3	
	Comments:								
10	How do you ra	te the cap	acity of th	ne crossin	gs? (widt	h, congesti	on, obstru	ctions, etc)	
		-3	-2	-1	0	+1	+2	+3	
	Comments:								
			FIC	TIRE 3	Questio	nnaire			

The first three questions reviewed the respondent's age, gender and frequency of visit to the area, in order to collect some demographical data of the sample so as to allow for the examination of any possible correlation between demographics and pedestrian perceptions. The remaining questions

were then formulated in a way that would allow for the most useful conclusions to be drawn, and given the nature of the area, it was decided that focus would be given to the assessment of the pedestrian crossings and the public spaces, following the structure of the PERS software. An important feature that was also added was the provision of space for the recording of additional comments that respondents would state. This would enable subsequent qualitative analysis so as to investigate the causes behind specific ratings.

The concept of the PERS software was also followed in the rating scale implemented, as respondents were asked to give a rating between -3 and +3 to each of the attributes examined, with -3 implying "very bad", 0 implying "neutral" and +3 implying "excellent". Indeed, it was deemed useful that respondents were able to give negative/neutral/positive ratings to aspects towards which they felt negative/indifferent/positive, as such a rating scale resulted in better clarity of the questions and hence more confident and honest answers.

The complete questionnaire is shown in Figure 3. Prior to the conduct of the actual survey the questionnaire was piloted in the field to identify potential shortcomings and need for improvements, particularly with respect to the interview length and the formulation of the questions (e.g. the existence of ambiguous statements, leading to the confusion of the respondents). 30 responses were collected in the pilot test and it was found that the time to complete the questionnaire was less than two minutes, and no signs of restlessness or frustration were shown by any of the respondents. Also, the respondents seemed to understand the questions fairly well and gave confident responses, in line with what was expected. It was hence decided that no modifications to the survey were required.

#### 3.3 Survey results

Surveys were carried out in several two-hour periods between January and June 2011, and the total number of responses obtained was 202. Looking at the demographics, there were 97 male respondents (48%) and 105 female ones (52%), resulting in a fairly well-balanced sample in terms of gender. Considering age, the following distribution was obtained: 4 respondents were under 21 years old (2%); 55 were 21-30 (27%); 46 were 31-40 (23%); 42 were 41-50 (21%); 35 were 51-60 (17%); and 20 were over 60 (10%). As such, the sample was well-balanced in terms of age, with the exception of the under 21 category, whose small number of responses could give statistically insignificant results in the analysis if considered on its own. With respect to the frequency of visit to the area, the following distribution was obtained: 25 respondents stated that it was their first visit to the area (12%); 48 said that they were infrequent visitors, i.e. less than once a month (24%); 51 visited the area at least once a month (25%); 50 visited the area at least once a week (25%); and 28 were daily visitors to the site, visiting it at least 5 times per week (14%). Consequently, a well-balanced sample was achieved.

The average score for each of the questions of the survey across the whole sample of respondents is shown in Figure 4a. As an overall conclusion about the area, the respondents rated most examined pedestrian experience aspects positively, with the exception of comfort. In particular the visual environment and the perceived safety were rated strongly positive, with the average ratings being above 1; comfort, on the other hand, scored clearly lower than the other aspects, with a rating of zero, indicating that the respondents were indifferent towards it. When looking at some of the respondents' comments, it seems that the lack of seating and the high levels of noise were the explanations behind that score.

Considering the results of the survey in conjunction with the demographics of the sample, i.e. the demographical categories' distributions of the responses to each question, interesting inferences can be drawn. Namely, it can be seen in Figure 4b that responses from males and females were similar, suggesting that gender did not affect the response to each of the questions, and that men and women perceived the public space in a similar way. On the other hand, looking at the responses of the different age groups in Figure 4c, it can be seen that ratings seem to decrease with increasing age in most questions, such that older pedestrians appeared to be more critical towards public space perception parameters. Finally, as concerns the frequency of visit, results suggest that this did not affect public space perception in general (Figure 4d). However, a notable finding is that people who

visited the area at least five times per week seemed to have rated safety slightly higher than others, whilst rating ease of movement, positioning of crossings and waiting times of crossings slightly lower. This might suggest that increased familiarity with the area resulted in noticing more negative aspects.



FIGURE 4 Average response ratings to each question

The effects of the demographics on the perceptions of pedestrians, as well as the

interrelationships of the parameters making up pedestrian perceptions, are investigated in more detail by means of statistical analysis in the next section.

# 4 ANALYSIS AND MODELLING

The statistical analysis of the results is carried out by fitting a series of regression models to the data, each time taking a different attribute as the dependent variable.

#### 4.1 Analysis methodology

As most variables/attributes had categorical outcomes ordered from low to high (-3 to +3), ordered logistic regression is used. Ordered logit models take the form  $Y^* = X \cdot \beta + \varepsilon$ , where  $Y^*$  is the unobserved dependent variable, X is the vector of independent variables,  $\beta$  is the vector of regression coefficients to be estimated, and  $\varepsilon$  is a random disturbance term following a logistic distribution. The observed ordinal variable Y is a function of  $Y^*$ , which has various category thresholds. For example,

$$\begin{aligned} \mathbf{Y}_{i} &= 1 \text{ if } \mathbf{Y}_{i}^{*} < \boldsymbol{\mu}_{1} \\ \mathbf{Y}_{i} &= 2 \text{ if } \boldsymbol{\mu}_{1} \leq \mathbf{Y}_{i}^{*} \leq \boldsymbol{\mu}_{2} \\ \mathbf{Y}_{i} &= 3 \text{ if } \mathbf{Y}_{i}^{*} > \boldsymbol{\mu}_{3} \end{aligned}$$

The ordered logit technique uses the observations on Y to determine the parameter vector  $\beta$  and the threshold values  $\mu_1$ ,  $\mu_2$ , and  $\mu_3$  so as to be able to subsequently estimate Y<sup>\*</sup> and predict Y for specific configurations of X.

In order to ensure that enough responses have been given to each response category of each attribute so that statistically significant coefficients can be obtained, responses to the perception-related questions (4 to 10) are aggregated into three categories: "negative" (-3 to -1), "neutral" (0) and "positive" (+1 to +3). As concerns the demographic attributes, the age and frequency of visit parameters are grouped into three categories: "under 30", "31-50" and "over 51", and "rarely" (less than once a month), "intermittently" (once a month) and "regularly" (at least once a week), respectively. No further grouping has been done to the responses to the gender question, as this is binary. As such, the parameters used in the statistical analysis, with their respective categories, are:

- the seven perception-related parameters, i.e. visual environment (VisEnv), ease of moving around (MovAround), comfort (Comf), safety (Saf), crossing positioning (CrossPos), crossing waiting time (WaitTime), and crossing capacity (CrossCap), each one having three categories (1=negative, 2=neutral, 3=positive);
- and the three demographic parameters, namely respondent's age (Age, 1=under 30, 2=30-50, 3=over 50), respondent's gender (Gen, 1=male, 2=female) and frequency of visit (Freq, 1=rarely, 2=intermittently, 3=regularly).

Regression is carried out by taking each one of the seven perception-related attributes as the dependent variable and fitting a model with the remaining perception-related and the three demographic parameters as independent variables. No models are fit with the demographic attributes as dependent variables, as these may be considered constant and can thus only be analysed as causes and not as consequences.

It should be noted that while the attributes are used as ordinal variables in the left hand side of each model, it has been decided that only binary variables (zero-one) be used as independent variables in the right hand side to ensure better readability of the effects. Namely, if ordinal variables were included in the right hand side, then the effect of an independent variable on the dependent one would not only be expressed by the coefficient, but would also depend on the value of the independent variable itself (such that a value of "2" would mean that the effect would be doubled). Zero-one

variables on the other hand ensure that the effect only depends on the value of the coefficient, as the effect is be zero if the variable is zero, and is equal to the coefficient is the variable is one.

In order to meet this condition, each ordinal variable is replaced by a series of binary ones in the right hand side of each model, depending on the number of categories of the corresponding attribute. Considering the fact that the number of variables coming into a model for each attribute should be n-1, where n is the number of categories of the attribute, two variables are introduced for each perception-related parameter, another two each for "Age" and for "Freq", and another one for "Gen". With respect to the notation, the postfixes (\_2, \_3) denote the cases where the respective attributes take the value indicated (e.g. the \_VisEnv\_2 zero-one variable denotes the VisEnv=2 category). The attributes considered and the respective variables used in the models are listed in Table 1.

Attribute	Ordinal var.	Categories	Binary independent var.
Visual environment	VisEnv	1=negative	_VisEnv_2
		2=neutral	_VisEnv_3
		3=positive	
Ease of moving around	MovAround	1=negative	_MovAround_2
		2=neutral	_MovAround_3
		3=positive	
Comfort	Comf	1=negative	_Comf_2
		2=neutral	_Comf_3
		3=positive	
Safety	Saf	1=negative	_Saf_2
		2=neutral	_Saf_3
		3=positive	
Crossing positioning	CrossPos	1=negative	_CrossPos_2
		2=neutral	_CrossPos_3
		3=positive	
Crossing waiting time	WaitTime	1=negative	_WaitTime_2
		2=neutral	_WaitTime_3
		3=positive	
Crossing capacity	CrossCap	1=negative	_CrossCap_2
		2=neutral	_CrossCap_3
		3=positive	
Respondent's age	Age	1=under 30	_Age_2
		2=30-50	_Age_3
		3=over 50	
Respondent's gender	Gen	1=male	_Gen_2
		2=female	
Frequency of visit	Freq	1=rarely	_Freq_2
		2=intermittently	_Freq_3
		3=regularly	

	TABLE 1	Attributes	and	variables
--	---------	------------	-----	-----------

#### 4.2 Modelling results

The STATA 10 statistical software package is used to perform the series of regressions and estimate the coefficients of the resulting ordered logit models. The results are shown in Table 2, where in each of the sub-tables the dependent ordinal variable is shown at the top-left cell. It should be noted that the tables of the two regressions using the VisEnv and Saf attributes as the dependent variables have not been included in Table 2, as the models did not converge. The two attributes have not been included in the other models as independent variables either, as they have either resulted in non-convergence, or in statistically insignificant models as whole. The reason behind this is that almost all respondents gave positive ratings to both visual environment and safety (hence the higher average ratings shown in

Figure 4a) regardless of their responses to the other questions, thus making an estimation of their interrelationships with the other attributes impossible. While from a purely statistical perspective this implies that visual environment and safety have no effect on the other attributes, this conclusion cannot be accepted from a logical perspective, as several interrelationships can be identified in practice (e.g. crossing capacity and safety, safety and age, visual environment and comfort, etc). The occurrence in question has thus been attributed to potential lack of clarity in the questions' formulation, or to the absence of benchmarks as to what constitutes a pleasant and safe environment, which can be established through comparison with a different site.

Looking at the other five models in Table 2, it can be seen that they are statistically significant as whole, as the null hypothesis for each model that all its coefficients are equal to zero is rejected at the 5% level in all models (Prob<chi2 smaller than 0.05). For the interpretation, the values of the coefficients (Coef. column) and their statistical significance to the 5% level through their p-values (P>|z| column) are examined.

(a)							
Ordered logist	ic regressior	1		Number	r of obs =	202	
				LR ch	i2(13) =	70.14	
Log likelihood	= -144 61498	2		Prop ,	$P R^2 =$	0.0000	
		,					
MovAround	Coef.	Std. Err.	Z	₽>   z	[95% Conf.	Interval]	
Comf 2	1.249796	.4153594	3.01	0.003	.435707	2.063886	
Comf 3	2.03011	.4639469	4.38	0.000	1.120791	2.939429	
_CrossPos_2	1.795007	.6092461	2.95	0.003	.6009071	2.989108	
_CrossPos_3	2.065914	.5748088	3.59	0.000	.9393097	3.192519	
_WaitTime_2	4637023	.8038814	-0.58	0.564	-2.039281	1.111876	
WaitTime_3	.063/535	./910835 5212457	-0.27	0.936	-1.486/42	2014607	
_CrossCap_2	4390324	5146418	-0.27	0.788	- 569647	1 447712	
_01000004p_0   Age 2	-1.013645	.4731191	-2.14	0.032	-1.940941	0863483	
Age 3	-1.270391	.5493713	-2.31	0.021	-2.347139	193643	
_Gen_2	.4088403	.3291786	1.24	0.214	2363379	1.054018	
_Freq_2	.4815159	.4458645	1.08	0.280	3923625	1.355394	
_Freq_3	5226736	.4202032	-1.24	0.214	-1.346257	.3009096	
/cut1	- 1268073	8104771			-1 715313	1 461699	
/cut2	1.532585	.8221726			0788434	3.144014	
(1)							
(b)							
(b) Ordered logist	ic regressior	1		Numbe	r of obs =	202	
(b) Ordered logist	ic regressior	1		Numbe LR ch	r of obs = i2(13) =	202 58.15	
(b) Ordered logist	ic regressior = -191 77096	1		Number LR ch: Prob >	r of obs = i2(13) = > chi2 = R2 =	202 58.15 0.0000 0.1316	
(b) Ordered logist Log likelihood	ic regressior	i 5		Number LR ch: Prob > Pseudo	r of obs = i2(13) = > chi2 = o R2 =	202 58.15 0.0000 0.1316	
(b) Ordered logist Log likelihood Comf	ic regression = -191.77096 	Std. Err.	Z	Number LR ch: Prob > Pseudo P> z	r of obs = i2(13) = > chi2 = o R2 = [95% Conf.	202 58.15 0.0000 0.1316 Interval]	
(b) Ordered logist Log likelihood  Comf   	ic regression 	5 Std. Err.	z 2.38	Number LR ch: Prob 2 P> z  0.017	r of obs = i2(13) = chi2 = o R2 = [95% Conf. .2355013	202 58.15 0.0000 0.1316 Interval] 2.441854	
(b) Ordered logist Log likelihood 	ic regression 	5 Std. Err. .5628555 .5344686	z 2.38 4.27	Numbe: LR ch: Prob > Pseudo P> z  0.017 0.000	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197	202 58.15 0.0000 0.1316 Interval 2.441854 3.332275	
(b) Ordered logist Log likelihood Comf   MovAround_2   MovAround_3   CrossPos_2	ic regression 	Std. Err. .5628555 .5344686 .577844	z 2.38 4.27 0.19	Number LR ch: Prob > Pseudo P> z  0.017 0.000 0.848	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression 	Std. Err. .5628555 .5344686 .577844 .5391777	z 2.38 4.27 0.19 0.01	Number LR ch: Prob > Pseudo P> z  0.017 0.000 0.848 0.995	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression 	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637	z 2.38 4.27 0.19 0.01 -0.57	Numbe LR ch: Prob 2 Pseudo P> z  0.017 0.000 0.848 0.995 0.569	r of obs = i2(13) = > chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression = -191.77096 Coef. 1.338678 2.284736 .1107058 .0031667 4521513 1338138 7274247	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 530367	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39	Number LR ch: Prob 2 Pseudo 0.017 0.017 0.000 0.848 0.995 0.569 0.864 0.164	r of obs = i2(13) = > chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 -302022	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597 1.402042 1.76672	
(b) Ordered logist Log likelihood 	ic regression = -191.77096 Coef. 1.338678 2.284736 .1107058 .0031667 .4521513 -1338138 .7374347 1.157451	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 .5303857 4858179	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38	Number LR ch: Prob 2 Pseudo 0.017 0.000 0.848 0.995 0.569 0.864 0.164 0.017	r of obs = i2(13) = > chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 -3021022 2052652	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597 1.402042 1.776972 2.109636	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression = -191.77096 Coef. 1.338678 2.284736 .107058 .0031667 4521513 1338138 .7374347 1.157451 579616	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 .5303857 .4858179 .3491055	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38 -1.66	Number LR ch: Prob 2 Pseudo 0.017 0.000 0.848 0.995 0.569 0.864 0.164 0.164 0.017 0.097	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 3021022 .2052652 -1.26385	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597 1.402042 1.776972 2.109636 .1046181	
(b) Ordered logist Log likelihood 	ic regression = -191.77096 Coef. 1.338678 2.284736 .1107058 .0031667 4521513 1338138 .7374347 1.157451 579616 6192661	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 .5303857 .4858179 .3491055 .4184308	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38 -1.66 -1.48	Number LR ch: Prob 2 Pseudo 0.017 0.000 0.848 0.995 0.569 0.864 0.164 0.164 0.017 0.097 0.139	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 3021022 .2052652 -1.26385 -1.439375	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597 1.402042 1.776972 2.109636 .1046181 .2008432	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression = -191.77096 Coef. 1.338678 2.284736 .1107058 .0031667 -4521513 -1338138 .7374347 1.157451 579616 6192661 .3131166	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 .5303857 .4858179 .3491055 .4184308 .2833993	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38 -1.66 -1.48 1.10	Number LR ch: Prob 2 Pseudo 0.017 0.000 0.848 0.995 0.864 0.164 0.017 0.097 0.139 0.269	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 3021022 .2052652 -1.26385 -1.439375 2423357	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.105993 1.402042 1.776972 2.109636 .1046181 .2008432 .8685689	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression 	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 .5303857 .4858179 .3491055 .4184308 .2833993 .3707309	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38 -1.66 -1.48 1.10 -0.84	Number LR ch: Prob 3 Pseudo 0.017 0.000 0.848 0.995 0.864 0.164 0.017 0.097 0.139 0.269 0.403	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 3021022 .2052652 -1.263857 -1.23357 -1.036793	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.105993 1.402042 1.776972 2.109636 .1046181 .2008432 .8685689 .4164458	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression 	Std. Err. 5628555 5344686 577844 5391777 7937637 7836143 5303857 4858179 3491055 4184308 2833993 3707309 3502757	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38 -1.66 -1.48 1.10 -0.84 1.30	Number LR ch: Prob 3 Pseudo 0.017 0.000 0.848 0.995 0.864 0.164 0.017 0.097 0.864 0.164 0.017 0.097 0.139 0.269 0.403 0.194	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 3021022 .2052652 -1.263652 -1.263657 -1.26357 -1.26357 -1.26357 -1.036793 2313111	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597 1.402042 1.776972 2.109636 .1046181 .2008432 .8685689 .4164458 1.141744	
(b) Ordered logist Log likelihood ———————————————————————————————————	ic regression 	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 .5303857 .4858179 .3491055 .4184308 .2833993 .3707309 .3502757 .9096525	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38 -1.66 -1.48 1.10 -0.84 1.30	Number LR ch: Prob 2 Pseudo 0.017 0.000 0.848 0.995 0.864 0.164 0.017 0.097 0.864 0.164 0.017 0.097 0.269 0.403 0.194	r of obs = i2(13) = chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 3021022 .2052652 -1.26365 -1.439375 2423357 -1.036793 2313111	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597 1.402042 1.776972 2.109636 .1046181 .2008432 .8685689 .4164458 1.141744 3.498113	
(b) Ordered logist Log likelihood —————————————————————————————————— MovAround_2   	ic regression 	Std. Err. .5628555 .5344686 .577844 .5391777 .7937637 .7836143 .5303857 .4858179 .3491055 .4184308 .2833993 .3707309 .3502757 .9096525 .9301734	z 2.38 4.27 0.19 0.01 -0.57 -0.17 1.39 2.38 -1.66 -1.48 1.10 -0.84 1.30	Numbel LR ch: Prob 3 Pseudo P> z  0.017 0.000 0.848 0.995 0.569 0.864 0.164 0.017 0.097 0.139 0.269 0.403 0.194	r of obs = i2(13) = > chi2 = p R2 = [95% Conf. .2355013 1.237197 -1.021848 -1.053602 -2.0079 -1.66967 3021022 .2052652 -1.26385 -1.439375 2423357 -1.036793 2313111 0676593 1.407603	202 58.15 0.0000 0.1316 Interval] 2.441854 3.332275 1.243259 1.059936 1.103597 1.402042 1.776972 2.109636 .1046181 .2008432 .8685689 .4164458 1.141744 3.498113 5.053816	

TABLE 2	Ordered	logistic	regression	models
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TABLE 2	(continued)
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Log likelihood		1		LR chi Prob >	chi2 =	202 41.32 0.000
	= -141.24843	3 		Pseudo	R2 =	0.127
CrossPos	Coef.	Std. Err.	Z	₽> z	[95% Conf.	Interval
MovAround 2	.0690453	.5223259	0.13	0.895	9546947	1.09278
MovAround 3	1.187012	.5198409	2.28	0.022	.1681427	2.205882
Comf 2	4973046	.425519	-1.17	0.243	-1.331307	.336697
Comf 3	.0286856	.4671069	0.06	0.951	8868271	.944198
WaitTime 2	1.575098	.7554797	2.08	0.037	.0943854	3.05581
WaitTime 3	2.108894	.7431241	2.84	0.005	.6523974	3.5653
CrossCap 2	.2004185	.5579143	0.36	0.719	8930734	1.2939
CrossCap 3	1375428	.5342588	-0.26	0.797	-1.184671	.909585
Age2	.7152615	.4002679	1.79	0.074	0692492	1.49977
_Age_3	.444675	.4978483	0.89	0.372	5310897	1.4204
_Gen_2	080589	.338662	-0.24	0.812	7443543	.583176
_Freq_2	0896156	.4653678	-0.19	0.847	-1.00172	.822488
_Freq_3	-1.07301	.4072423	-2.63	0.008	-1.87119	274829
+						
/cut1	319948	.7919441			-1.87213	1.23223
/cut2	1.506265	.8077501			0768956	3.08942
Log likelihood	= -130.08093	3		LR chi Prob > Pseudo	chi2 = R2 =	52.7 0.000 0.168
WaitTime	Coef.	Std. Err.	z	₽>   z	[95% Conf.	Interval
MovAround 2	0431655	.5349036	-0.08	0.936	-1.091557	1.00522
MovAround 3	.4344871	.537156	0.81	0.419	6183193	1.48729
Comf 2	.456753	.4485614	1.02	0.309	4224111	1.33591
Comf 3	.1335155	.4470725	0.30	0.765	7427306	1.00976
_CrossPos_2	.5080314	.6296382	0.81	0.420	7260368	1.742
_CrossPos_3	1.33566	.5987142	2.23	0.026	.1622018	2.50911
_CrossCap_2	.1729429	.4814151	0.36	0.719	7706134	1.11649
_CrossCap_3	1.998022	.4864742	4.11	0.000	1.04455	2.95149
_Age_2	.3880264	.4234662	0.92	0.360	4419522	1.21800
_Age_3	1.34842	.52/241/	2.56	0.011	.3150455	2.381/9
_Gen_2	3013085	.3441275	-0.88	0.381	9/5/86	.3/3169
_Freq_2	258104	.4239173	-0.61	0.543	-1.08896/	.5/2/58
Freq_3	.2362618	.4328494	0.55	0.585	6121075	1.08463
/cut1	3757227	.7801506			-1.90479	1.15334
	2.248294	.8072543			.6661049	3.83048
/cut2						
/cut2   (e) Ordered logist Log likelihood	ic regression = -154.24334	1		Number LR chi: Prob > Pseudo	of obs = 2(13) = chi2 = R2 =	20 62.0 0.000 0.167
/cut2   (e) Ordered logist Log likelihood CrossCap	ic regression = -154.24334 Coef.	1 Std. Err.	Z	Number LR chi: Prob > Pseudo P> z	of obs = 2(13) = chi2 = R2 = [95% Conf.	20 62.0 0.000 0.167 Interval
/cut2   (e) Drdered logist Log likelihood CrossCap   MovAround 2	ic regression = -154.24334 Coef.	1 Std. Err. .4901582	z 0.62	Number LR chi: Prob > Pseudo P> z  0.534	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433	20 62.0 0.000 0.167 Interval
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround_2   MovAround_3	ic regression = -154.24334 	1 Std. Err. .4901582 .5004164	z 0.62 1.25	Number LR chi: Prob > Pseudo P> z  0.534 0.213	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283	20 62.0 0.000 0.167 Interval 1.26584 1.60436
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround_2   MovAround_3   	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072	4 Std. Err. .4901582 .5004164 .4148936	z 0.62 1.25 1.93 2.25	Number LR chi Prob > Pseudo P> z  0.534 0.213 0.054 0.054	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283 0141693 125622	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround 2   MovAround 3   Comf_2 2   Comf_3	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 220177	4 Std. Err. .4901582 .5004164 .4148936 .4403051	z 0.62 1.25 1.93 2.25	Number LR chi' Prob > Pseudo 0.534 0.213 0.054 0.025 0.710	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283 0141693 .1256832	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218 1.85164
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround_2   MovAround_3   _Comf_2   _Comf_3   _CrossPos_2	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271	4 Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .604072	z 0.62 1.25 1.93 2.25 0.36	Number LR chi: Prob > Pseudo 0.534 0.213 0.054 0.025 0.718 0.021	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283 0141693 .1256832 -1.046481 -1.256720	20 62.0 0.000 Interval 1.26584 1.60436 1.61218 1.85164 1.51873
/cut2   	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271 -0631554 234455	4 Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 7571467	z 0.62 1.25 1.93 2.25 0.36 -0.10 2.00	Number LR chi: Prob > Pseudo 0.534 0.213 0.054 0.025 0.718 0.918 0.022	of obs = 2(13) = chi2 = R2 = [95% Conf. 	20 62.0 0.000 Interval 1.26584 1.60436 1.61218 1.85164 1.51873 1.13241
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround 2   MovAround 3   _Comf 2   _Comf 3   _CrossPos_2   _CrossPos_3   WaitTime_2   MovArime_2	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271 0631554 2.340458 3.300066	4 Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 .7571467 7450346	z 0.62 1.25 1.93 2.25 0.36 -0.10 3.09 4.56	Number LR chi: Prob > Pseudo D> z  0.534 0.213 0.054 0.025 0.718 0.918 0.918 0.002	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 0141693 .1256832 -1.046481 -1.258729 .8564781 1 93625	20 62.0 0.000 1.167 1.26584 1.60436 1.61218 1.85164 1.51873 1.13241 3.822443
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround_2   MovAround_3   Comf_2   Comf_3   CrossPos_2   CrossPos_3   WaitTime_2   WaitTime_3   Marce 1	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271 0631554 2.340458 3.399866 -5829074	4 Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 .7571467 .7450346 .4438523	z 0.62 1.25 1.93 2.25 0.36 -0.10 3.09 4.56 -1.31	Number LR chi: Prob > Pseudo 	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 0141693 .1256832 -1.046481 -1.258729 .8564781 1.939625 -1.452842	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218 1.85164 1.51873 1.13241 3.82443 4.86010 287027
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround 2   _Comf_2   _Comf_3   _CrossPos_2   _CrossPos_3   _CrossPos_3   _WaitTime_2   _WaitTime_3   _Are_2	<pre>ic regression = -154.24334 Coef3051491 .6235698 .7990072 .9886653 .23612710631554 2.340458 3.3998665829074 -1.340833</pre>	A Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 .7571467 .7450346 .4438523 .5024737	z 0.62 1.25 1.93 2.25 0.36 -0.10 3.09 4.56 -1.31 -2.67	Number LR chi: Prob > Pseudo 0.534 0.213 0.054 0.025 0.718 0.918 0.002 0.000 0.189 0.008	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283 0141693 .1256832 -1.046481 -1.258729 .8564781 1.939625 -1.452842 -2.325663	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218 1.85164 1.85164 1.38743 4.86010 .287027 .356002
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround_2   MovAround_3   Comf_2   Comf_3   CrossPos_2   CrossPos_2   CrossPos_2   MaitTime_2   WaitTime_3   Age_2   Age_2   Age_3	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271 0631554 2.340458 3.399866 5829074 -1.340833 1118/277	4 Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 .7571467 .7450346 .4438523 .5024737 3249769	z 0.62 1.25 1.93 2.25 0.36 -0.10 3.09 4.56 -1.31 -2.67 0.34	Number LR chi: Prob > Pseudo 0.534 0.023 0.054 0.025 0.718 0.002 0.000 0.189 0.008 0.731	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283 0141693 .1256832 -1.046481 -1.258729 .8564781 1.939625 -1.452842 -2.325663 -5251003	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218 1.85164 1.51873 1.13241 3.82443 4.86010 .287027 356002 748755
/cut2   	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271 0631554 2.340458 3.399866 5829074 -1.340833 .1118427 -2523768	4 Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 .7571467 .7450346 .4438523 .5024737 .3249769 4118915	z 0.62 1.25 1.93 2.25 0.36 -0.10 3.09 4.56 -1.31 -2.67 0.34 -0.61	Number LR chi: Prob > Pseudo D-534 0.534 0.213 0.054 0.225 0.718 0.918 0.918 0.002 0.000 0.189 0.008 0.731 0.540	of obs = 2(13) = chi2 = R2 = [95% Conf. 	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218 1.85164 1.51873 1.13241 3.82443 4.86010 .287027 356002 .748785 5554015
/cut2   	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271 0631554 2.340458 3.399866 5829074 -1.340833 .1118427 2523768 6968797	4 Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 .7571467 .7450346 .4438523 .5024737 .3249769 .4118915 .40553.48	z 0.62 1.25 0.36 -0.10 3.09 4.56 -1.31 -2.67 0.34 -0.61 -1.72	Number LR chi: Prob > Pseudo 0.534 0.213 0.054 0.025 0.718 0.918 0.002 0.000 0.189 0.008 0.731 0.540 0.086	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 0141693 .1256832 -1.046481 -1.258729 .8564781 1.939625 -1.452842 -2.325663 5251003 -1.059669 -1.491713	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218 1.85164 1.51873 1.13241 3.82443 4.86010 .287027 356002 .748785 .554915
/cut2   (e) Ordered logist Log likelihood CrossCap   MovAround 2   MovAround 3   _Comf_2   _Comf_3   _CrossPos_2   _CrossPos_3   _CrossPos_3   _MaitTime_2   _MaitTime_3   _Age_3   _Freq_2   _Freq_3	ic regression = -154.24334 Coef. .3051491 .6235698 .7990072 .9886653 .2361271 0631554 2.340458 3.398866 5829074 -1.340833 .1118427 2523768 6968797	A Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .6099978 .7571467 .7450346 .4438523 .5024737 .3249769 .4118915 .4055348	z 0.62 1.25 1.93 2.25 0.36 -0.10 3.09 4.56 -1.31 -2.67 0.34 -0.61 -1.72	Number LR chi: Prob > Pseudo P> z  0.534 0.213 0.054 0.025 0.718 0.002 0.000 0.189 0.008 0.731 0.540 0.086	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283 0141693 .1256832 -1.046481 -1.258729 .8564781 1.939625 -1.452842 -2.325663 5251003 -1.059669 -1.491713	20 62.0 0.000 0.167 Interval 1.26584 1.60436 1.61218 1.85164 1.85164 1.382443 4.86010 .287027 356002 .748785 .554915 .09795
/cut2   (e) Drdered logist Log likelihood CrossCap   MovAround_2   MovAround_3   Comf_3   CrossPos_2   CrossPos_3   WaitTime_2   MaitTime_3   Age_2   Age_3   CFreq_2   Freq_3   / cut1	<pre>ic regression = -154.24334</pre>	A Std. Err. .4901582 .5004164 .4148936 .4403051 .654404 .609978 .7571467 .7450346 .4438523 .5024737 .3249769 .4118915 .4055348 .8600934	z 0.62 1.25 1.93 2.25 0.36 -0.10 3.09 4.56 -1.31 -2.67 0.34 -0.61 -1.72	Number LR chi: Prob > Pseudo 0.534 0.213 0.054 0.025 0.718 0.002 0.000 0.189 0.008 0.731 0.540 0.086	of obs = 2(13) = chi2 = R2 = [95% Conf. 6555433 3572283 0141693 .1256832 -1.046481 -1.258729 .8564781 1.939625 -1.452842 -2.325663 5251003 -1.059669 -1.491713 9117492	20 62.0 0.000 0.167 1.26584 1.60436 1.61216 1.85164 1.85164 1.51873 1.13241 3.82443 4.86010 .2356002 .748785 .554915 .09795

Table 2a shows the resulting model with the ease of moving around attribute (MoveAround) as the dependent variable. As can be seen, the coefficients of the variables \_Comf\_2, \_Comf\_3, \_CrossPos\_2, \_CrossPos\_3, \_Age\_2 and \_Age\_3 are found to be statistically significant (<0.05) and are all positive. With respect to the first four, this implies that the attributes of comfort and positioning of crossings have positive effects on the ease of moving around the area, such that pedestrians who rate the comfort and the positioning of crossings in a street space positively are also likely to find it easy to move around. It is additionally noteworthy that the effects increase with higher ratings of both attributes, and that while the effect of the crossings positioning is higher than that of the comfort in neutral ratings, it is roughly equal in the case where both attributes are rated positively. On the other hand, the negative coefficients of \_Age\_2 and \_Age\_3 indicate an inverse relationship between age and the ease of moving around. In fact, it seems that the ease of moving around is likely to be looked at less favourably by older people, as already suggested by the histogram of Figure 4c.

Considering the model with the comfort attribute (Comf) as the dependent variable (Table 2b), the variables \_MoveAround\_2, \_MoveAround\_3 and \_CrossCap\_3 have statistically significant coefficients. In the case of the former two (MoveAround), the finding of the first model can be identified as a two-way effect, as it appears that the ease of moving around has an effect on the pedestrian comfort, with positive coefficients increasing with higher ratings. As concerns the other attributes, the positive coefficient of \_CrossCap\_3 indicates that highly-rated crossing capacity is likely to also lead to a positive perception of comfort.

Taking the positioning of crossings (CrossPos) as the dependent variable (Table 2c), the variables \_MoveAround\_3, \_WaitTime\_2, \_WaitTime\_3 and \_Freq\_3 have statistically significant coefficients. The positive coefficient of the first variable demonstrates that the positively rated ease of moving around the space is likely to positively affect the rating of the crossings' positioning. The positive coefficients of the next two variables indicate that the waiting time has a similar effect, which in fact becomes stronger with increased positive perception of waiting time (the coefficient of \_WaitTime\_3 is higher than \_WaitTime\_2). As such, it appears that pedestrians who find it easy to move to the crossings and do not find that they are waiting too long to cross are also likely to be happy with the positioning of the crossings. On the other hand, the negative coefficient of \_Freq\_3 suggests that the positioning of the crossings is looked at more critically by regular visitors to the area.

The interrelationship between the positioning of the crossings and the perceived waiting time is further found to be two-way in the model where waiting time (WaitTime) is the dependent variable (Table 2d), as \_CrossPos\_3 has a positive statistically significant coefficient. Additional effects on the waiting times that can be identified are that of the crossing capacity, given the positive significant coefficient of \_CrossCap\_3, and that of the older age, given the also positive coefficient of \_Age\_3. These indicate that pedestrians may perceive waiting at crossings more favourably if crossings appear well-positioned and of sufficient capacity, or if the pedestrians are over 50 years old.

Lastly, interesting conclusions can be drawn from the last model, where the crossing capacity (CrossCap) is the dependent variable (Table 2e), and where the variables \_Comf\_3, \_WaitTime\_2, \_WaitTime\_3 and \_Age\_3 have statistically significant coefficients. The positive coefficient of \_Comf\_3 implies that pedestrians who rate comfort positively are likely to look more favourably on crossing capacity, indicating that the interrelationship between the two parameters is two-way. The positive coefficients of the WaitTime variables, in conjunction with the model of Table 2d, show that the interrelationship of the waiting time at crossings and of the perceived capacity of the crossings is also two-way, with higher ratings of the former having a higher effect of the latter. On the other hand, the negative coefficient of \_Age\_3 suggests that older pedestrians (over 50) perceive the capacity of the crossings less favourably.

In summary, the models suggest that there is interdependence between many of the parameters constituting the pedestrians' perception of public spaces. The different relationships are summarised in Table 3, where the nature of each dependence (positive, negative or none) and its magnitude are indicated.

	TABLE 5 Interdependence of pedestrian perception parameters							
		Dependent variable						
		Ease of movement	Comfort	Crossing positioning	Crossing waiting time	Crossing capacity		
	Ease of movement		+ +	+	•	•		
le	Comfort	+ +				+		
Independent variabl	Crossing positioning	+ +	•		+	•		
	Crossing waiting time			+ +		+ + +		
	Crossing capacity		+		+			
	Age	—	•	•	+	Ι		
	Gender							
	Frequency of visit			_				

TABLE 3 Interdependence of pedestrian perception parameters

#### 5 CONCLUSIONS

In light of the shift in focus in urban street design, this paper examined the topic of the evaluation of the pedestrian experience in public spaces and presented a new method based on a questionnaire for use in on-street surveys, building on comprehensive pedestrian evaluation systems. Collecting data from respondents in London's South Kensington area, which has recently undergone major redevelopment to a more pedestrian-oriented street design, the study found that the new design generally attracted positive comments, but that there is room for improvement in terms of the pedestrian comfort. Then, the study went on to conduct statistical analysis to the data to investigate the interrelationships between the several aspects constituting the pedestrian experience. By performing a series of ordered logistic regressions, it was found that there are strong positive dependences between comfort and ease of movement, between the perceived positioning of the crossings and the ease of movement, and between the pedestrian's age and the ease of movement and perceived crossing capacities, and between the frequency of visit to an area and the perceived crossing positioning.

The results highlight the need for putting special attention into the different aspects of public space at the design stage, since the way in which pedestrians perceive a particular aspect may be influenced by altering other aspects rather than just the one in question. For example, if designers were interested in improving the waiting times of crossings, they may want to consider improving the capacity of the crossings instead, since this is likely to improve the perception of the waiting times by pedestrians without the need of modifying the signal programmes.

While the study has thrown some light into the topic of assessing the pedestrian experience, research in this direction continues. Further work will primarily concentrate on collecting data from more sites in order to investigate the effects of the visual environment and safety parameters, but also to observe the performance of the new questionnaire in a more generic context. Additionally, through comparative work with comprehensive pedestrian audit systems, such as PEDS and PERS, predictive models will be derived which will allow for more detailed assessment of the pedestrian experience in situations with limited data availability. Finally, the relationship of the pedestrian experience with the behaviour of other road users, such as vehicle drivers and cyclists, will be investigated.

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