Individual Differences and Impact of Gender on Curvature Redirection Thresholds

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ABSTRACT
To enable real walking in a virtual environment (VE) that is larger than the available physical space, redirection techniques that introduce multisensory conflicts between visual and nonvisual cues to manipulate different aspects of a user’s trajectory could be applied. When applied within certain thresholds, these manipulations could go unnoticed and immersion remains intact. Research effort has been spent on identifying these thresholds and a wide range of thresholds was reported in different studies. These differences in thresholds could be explained by many factors such as individual differences, walking speed, or context settings such as environment design, cognitive load, distractors, etc.

In this paper, we present a study to investigate the role of gender on curvature redirection thresholds (RDTs) using the maximum likelihood procedure with the classical two-alternative force choice task. Results show high variability in individuals’ RDTs, and that on average women have higher curvature RDTs than men. Furthermore, results also confirm existing findings about the negative correlation between walking speed and curvature RDTs.

CCS CONCEPTS
• Human-centered computing → Virtual reality: Empirical studies in HCI. • Applied computing → Psychology:

KEYWORDS
walking, virtual reality, multisensory conflicts, gender difference

1 INTRODUCTION
It has been shown that the most immersive way to explore a VE is to really walk in it [Usch et al. 1999]. However, the problem with real walking arises when the VE is much larger than the available physical space. A solution to this problem was proposed where a sensory conflict between the visual information provided through the head-mounted display (HMD) and nonvisual information (vestibular and proprioceptive) is introduced [Razzaque et al. 2001]. This conflict, quantified as redirection gain, results in different virtual and real trajectories, for example: the user walks on a different path curvature in real life than in the VE (curvature redirection), the user walks slower or faster in real life than in the VE (translation redirection), and the user rotates slower or faster in real life than in the VE (rotation redirection). Nevertheless, there is a limit to this sensory conflict, beyond which it is noticeable to the user and a break in immersion occurs. Research effort has been spent on identifying this limit, so-called threshold. However, results found by different groups have quite high variance. For example, curvature thresholds found by different research groups vary considerably: 5 m [Grechkin et al. 2016], 10 m [Nitzsche et al. 2004] and 22 m [Steinicke et al. 2010]. An explanation for this high variance could be that these experiments had different setups and did not control for factors that affect the thresholds such as speed [Neth et al. 2011] or environment design [Hodgson et al. 2014]. However, one significant factor could be the high variance in individuals’ ability to detect the sensory conflict. This ability has been proposed to correlate with one’s visual dependence [Nguyen et al. 2016]. More specifically, stronger reliance on visual information could lead to higher threshold for detecting this conflict. Since a number of past research has reported gender difference in visual dependence, it could be hypothesized that there is also gender difference in the ability to detect the multisensory conflict introduced during redirected walking.

2 RELATED WORK
In different tasks related to spatial ability, it has been shown that women tend to have higher visual dependence than men. In a rod-and-frame test where subjects have to align a rod in a tilted frame to the gravitational vertical axis, women are more affected by the frame orientation [Tremblay et al. 2004] [Barnett-Cowan et al. 2010]. In an optokinetic drum, women perceive self-motion induced by a rotating pattern, so-calledvection, faster and with stronger...
in the context of redirected walking, there has been research effort to investigate the effect of gender on RDTs [Bruder et al. 2009]. Even though no significant effect of gender on RDTs was found, they remarked that the number of subject was too small. Additionally, walking speed was not controlled in their experiment, which may have lead to additional noise in the result.

Given the well established gender difference in visual dependence and the limitations in terms of sample size and controlled factors of the existing study, we performed a larger scale study to investigate the effect of gender on curvature RDTs where walking speed is controlled. We also investigate if the effect of walking speed on curvature RDTs could be reproduced using a non-visual method of speed regulation. Different protocols for threshold identification are also discussed.

3 METHODOLOGY
3.1 Threshold identification
Threshold identification refers to the process of identifying the stimulus level at which a user can correctly detect the stimulus a predefined percentage of the time, e.g. 75%. The result from such detection task is conventionally modeled as a psychometric function, which describes the relationship between the stimulus level (x-axis) and the chance of a certain response (y-axis). Depending on the question posed to the user, there are two types of psychometric functions. In a yes/no task, the user is exposed to one stimulus per trial and is asked the question: “Did you detect the stimulus?”. The y-axis of the psychometric function in this case represents the percentage of “yes” responses, ranging from 0 at zero stimulus level to 1 at significantly high stimulus level. In a classical two-alternative forced-choice task (2AFC), the user is presented two options concurrently or sequentially in 2 intervals, only one of which contains the stimulus, and answers to the question: “In which interval was the stimulus present?”. In this case, the y-axis of the psychometric function represents the percentage of correct responses, and ranges from 0.5 at zero stimulus level to 1 at significantly high stimulus level. Existing RDTs studies employed a variant of the classical 2AFC method where alternative questions such as: “Did you walk to the left or the right?” were presented. The psychometric curve derived from this variant is similar to the one from a yes/no task, where the y-axis ranges from 0 to 1. Generally, the 2AFC method is preferred over the yes/no method due to the fact that the yes/no method often contains response bias, e.g. a tendency to answer “yes” when the stimulus is not perceivable. Therefore, originally in our pilot study, we employed the same 2AFC variant task as other existing RDTs studies for our experiment. However, one pilot subject consistently experienced the sensation of going left when the gains in both directions are sufficiently small, and thus keeping answering “left”. This resulted in a right curvature radius for his left threshold. From an application point of view, this subject can never be redirected left because his left threshold is a right curve. When the same pilot subject was tested again with the classical 2AFC task, the left threshold were found to be a ~13m radius left curve. This finding suggests that the 2AFC variant task is not robust against this type of bias. As a result, for the identification of our curvature RDTs in this experiment, we use the classical 2AFC task.

Independent of the task used, the threshold identification process revolves around identifying the psychometric function. There are two main types of methods used for threshold identification. The first type, the constant stimuli method (CSM), fits the whole psychometric function based on users’ responses at a pre-selected range of stimuli. In order to obtain a well-fitted curve, many stimulus levels are required and a high number of repetitions is usually needed at each tested stimulus level. For a threshold identification application, this method may not be very efficient since not the whole curve, but only a certain point on the curve or the slope of the curve is normally of interest. The second type, the adaptive method, tackles the efficiency problem of the CSM by using an adaptive procedure which selects the next stimulus level based on the previous response(s) of the users. This type of method does not identify the whole psychometric function, but only some of its parameters such as threshold, slope, guessing rate, etc. One category of adaptive methods is called the maximum likelihood procedure in which the optimal placement of the next stimulus is determined by fitting a parametric model of the psychometric function using data collected from all previous trials. Depending on stimulus placement strategies, stopping conditions and threshold final estimation methods, there is a wide variety of maximum likelihood procedures, of which the Bayesian method is recommended when only the threshold needs to be identified [Ttreutlein 1995]. While the CSM is simpler to implement, more trials are required leading to longer exposure time of the subjects in the VE. Therefore, in our experiment, we adopt a Bayesian method called QUEST [Watson and Pelli 1983].

3.2 Speed regulation
In an experiment by Neth et al., walking speed has been shown to have a negative correlation with curvature RDTs [Neth et al. 2011]. In that experiment, walking speed was controlled by the subjects following and maintaining a distance to an object moving at constant speed. However, it is not known how such task distracts the users visually from the main task of detecting curvature redirection, and consequently how it affects RDTs. Therefore, in our experiment we control the users’ speeds by having them following certain audio step rhythms. Dean studied the energy expenditure of human walking and derived a relationship between walking speed, stepping frequency and height [Dean 1965]. The equation describing this relationship is used for generating the aforementioned step rhythms.

4 USER STUDY
4.1 Experiment Design
The experiment scene contains an empty room with four surrounding walls and a red target located 7.5m from the starting position of the user. Before the experiment, users’ heights are collected and their personal step rhythms are generated. Each user has to walk with two speed conditions: 0.75m/s and 1.25m/s (similar to the conditions in [Neth et al. 2011]) regulated by the step rhythms. We also identify the left and right thresholds separately for each user, resulting in four threshold values of four psychometric functions to be found per user. Speed and curvature direction are therefore within-subject variables.
Using the 2AFC method, in each trial each user walks to the red target two times. In only one of the two walks, a curvature gain is applied such that the subject walks on a curve in real life while seeing that he/she walks straight in the VE. The curvature gain is defined as the inverse of the radius of the curve that the subject walks on. The order is randomized between trials. Four separate “QUESTs” handle the four psychometric functions and calculate the next curvature gain value to be tested for each function. These four next values are selected randomly to be presented in the next trial. This process is called interleaving, commonly used to reduce training and adaption effect. For each “QUEST”, 40 trials are required, resulting in a total of 160 trials per user. During the trials, the process of updating the “QUESTs” with users’ responses and computing the next curvature gain value is automated. Users respond to the 2AFC question using a built-in eye tracker.

4.2 Experiment Setup
The experiment setup consists of an Oculus DK2 with a built-in SMI eye tracker and an Intersense IS-1200 inside-out optical tracking system mounted on top, providing 6-DOF positional tracking at 180Hz. A cover is added in front of the headset to prevent users from seeing the floor. The scenes are optimally designed in Unity to run at the HMD’s maximum frame rate 75Hz. The whole setup is powered by a backpack-mounted notebook. The available tracking space is 12m x 6m. The users also wear noise cancelling headphones, where the step regulating rhythms are played.

4.3 Participants and Procedure
61 subjects (aged from 18-35 (M=25.1, SD=3.9), 30 male and 31 female, right-handed, with normal or corrected-to-normal vision, no vestibular dysfunction or injured) were recruited through the university market place. The experiment lasted about 90 minutes including instruction and set up time and was divided into two sessions. The subjects were paid 15CHF/hour for their contribution. In the first session, the subjects were informed about the purpose of the experiment and signed the consent form to participate. Subjects provided basic information such as age, handedness, and gaming hours per week. They were informed about the risk of motion sickness and filled out Kennedy’s simulator sickness questionnaire (SSQ) [Kennedy et al. 1993].

In both sessions, there was a significant increase in score in the after condition. To investigate the effect of gender on this increase in SSQ score, we fitted a linear mixed model which includes gender as a fixed factor and subject as a random factor: ScoreIncrease ~ 1 + Gender + (1|Subject). Results showed no significant effect of gender on the increase in SSQ score.

For each subject, four threshold values were obtained corresponding to the two speeds and two curvature directions: left slow, left fast, right slow and right fast. The distribution of the 240 obtained threshold values (albeit independency between data points as each subject contributes four data points) is shown in Figure 1, serving only as an overall picture. In general, there is a high variability in curvature gain thresholds, ranging from 0.0237 (equivalent to a radius of 42.2m) to 0.1994 (equivalent to a radius of 5.2m) and a median of 0.0976 (equivalent to a radius of 10.24m), interquartile range 0.07 to 0.1336.

5 RESULTS AND DISCUSSION
Out of 61 subjects, one female subject reported motion sickness after 10 minutes and could not continue. 60 remaining subjects could complete the whole experiment. Two-sample t-tests were conducted to compare the SSQ scores before and after each session. In both sessions, there was a significant increase in score in the after condition. To investigate the effect of gender on this increase in SSQ score, we fitted a linear mixed model which includes gender as a fixed factor and subject as a random factor: ScoreIncrease ~ 1 + Gender + (1|Subject). Results showed no significant effect of gender on the increase in SSQ score.

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Although gender is correlated with RDTs, there is quite high variance in RDTs within each gender. Therefore, a measure of individual visual dependence could potentially be a better predictor for RDTs. Nevertheless, this hypothesis remains to be investigated in future experiments.

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