

# An Intelligent 3D User Interface Adapting to User Control Behaviors

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## ABSTRACT

The WALK mode is one of the most common navigation interfaces for 3D virtual environments. However, due to the limited view angle and low frame rate, users are often blocked by obstacles when they navigate in a cluttered virtual scene with such a mode. Intelligent 3D navigation interfaces with assisting mechanisms, such as motion planning methods or virtual force fields, have been proposed in the literature to improve navigation efficiency. Nevertheless, the applicability of these methods is subject to individual discrepancy, and the control parameters of these methods are usually determined by empirical means. In this paper, we propose an intelligent navigation interface with a personalizable assisting mechanism. We have designed two methods, simulation experiment and dynamic adjustment, to find the best control parameters for composing artificial forces for an individual in an off-line and on-line manner, respectively. The simulation experiment method searches for the optimal control parameters for a user in a systematic manner while the dynamic adjustment method makes the assisting mechanism adaptive to user control behaviors as well as environmental variations in real time. Our experiments show that both methods can further improve the navigation efficiency for a wider range of users.

## Categories and Subject Descriptors

H.5.1 [Information Interfaces and Presentation]: User Interface – *Graphical User Interface, Input Devices and Strategies*.  
I.3.6 [Computer Graphics]: Methodology and Techniques - *Interaction Techniques*.

## General Terms

Algorithms, Design, Experimentation, Human Factors.

## Keywords

Intelligent 3D interface, Artificial Force Field, Personalized User Interface control, Adaptive Assisting Mechanism.

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## 1. INTRODUCTION

Virtual Reality (VR) techniques adopt interactive 3D user interfaces to allow a user to have realistic experiences in virtual scenes. Such 3D user interfaces have a wide range of applications in computers, industries, medicine, military, and education. For example, one can find VR applications in Computer-Aided Design (CAD), flight simulation, virtual museums, and molecular structure analysis, and so on. Due to the constraints of 2D input and display devices, the design of a good user interface on a typical personal computer become a crucial and challenging problem.

Today 3D contents are becoming prevalent on the web via 3D browsers such as VRML browsers. The common navigation modes on such a browser include WALK, PAN, TURN, FLY, ROLL, and so on. The WALK mode is usually the default mode that allows a user to walkthrough a 3D scene. However, due to the limited field of view that can be displayed on a screen, a viewpoint might easily get stuck at a corner of a scene since the body of the avatar collides with the environmental obstacles. It usually takes several maneuvers for a user to escape from this type of difficult region. In order to overcome this problem, much previous work has proposed assisting mechanisms, such as motion planning algorithms[8][11] and virtual forces[10], to improve navigation control.

However, experiments show that user discrepancy on the familiarity of controlling 3D interfaces greatly influences the effects of such assisting mechanisms. Novice users might need more guidance from the system while expert users might want to have full control of the interface. Therefore, it is more desirable to have a system that can adapt to an individual user's control behavior and adjust the degrees of assistance accordingly. In this paper, we propose two mechanisms to make the intelligent 3D user interface with virtual forces adaptive to users' control behaviors in order to further improve the navigation performance.

The rest of the paper is organized as follows. In the next section, we will review the work pertaining to intelligent 3D user interfaces. We will briefly review the assisting mechanism of virtual force field to identify the control parameters that can be tuned to fit users' control behaviors. Next we will propose a mechanism to find the optimal control parameters by simulation experiments. We will then present another mechanism to adjust these control parameters in an on-line manner. Experimental results will be presented and discussed in the following section. Finally, we will conclude the paper at the end.

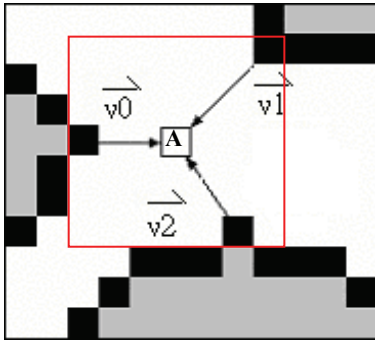


Figure 1. Example of how to compute the virtual force for a viewpoint A according to the boundary cells of nearby obstacles.

## 2. RELATED WORK

3D user interface design has attracted much attention in the virtual reality research community. Various kinds of devices have been invented to facilitate the control of 3D scenes. For example, Head Mounted Device (HUD), data gloves, 3D tracking, and haptic devices are all examples of hardware interfaces for 3D control. However, for most desktop computers, 2D mice remain the most accessible control devices [6][13]. However, it is a challenge to use a 2D mouse to control a 3D scene for a novice user since the controllable degrees of freedom (DOF) are limited by the DOF (2) of a mouse at a time. In addition, the mapping between them may not be very intuitive. Therefore, several researches have made proposals to improve the 3D control interface, especially for 3D rotation [2][4][7][14].

Although many intelligent user interfaces have been proposed in the literature, most of them are not for 3D manipulation [12]. Exceptions include using motion-planning techniques [8] to provide task-level controls. For example, Drucker and Zeltzer[3] argue that a task-level viewpoint control is crucial for exploring virtual scenes such as virtual museums since the users should be allowed to concentrate on scene viewing instead of being distracted by low-level navigation control. In addition, Artificial Intelligence (AI) techniques have been used to improve the efficiency of 3D navigation, especially for the WALK mode, at a control level. These techniques include motion planning algorithms [11] and virtual force field methods [10][5][16][17]. In the motion planning approach, efficient path planners are evoked to generate a collision-free path to take the viewpoint past the difficult region where the user got stuck. In the virtual force approach, a repulsive force from obstacle boundary is computed to influence the viewpoint motion such that it is less likely to collide with the environmental obstacles. In other words, the force field approach aims to prevent collisions while the motion planning approach can generate a feasible path for the viewpoint to the desired goal when collisions do happen.

Since user discrepancy is more significant for 3D interfaces, personalization of 3D interfaces is an important problem although not much work has been done in this direction. In [9], a reinforcement learning method has been employed to realize user interface customization for manipulating 3D objects in a virtual environment.

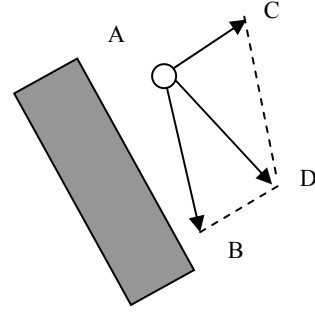


Figure 2. Incorporating virtual force into a control loop.

## 3. VIRTUAL FORCE FIELD

In this section, we will first review how virtual force is incorporated into the control loop of user navigation and then identify the control parameters for composing the virtual force.

### 3.1 How Virtual Force Works

In order to compute the virtual force field, a discrete representation of the virtual world is commonly used. The objects in the world are usually projected into obstacle regions in a 2D grid. The boundary cells of obstacles generate repulsive forces (inversely proportional to the distance from the viewpoint) to influence the nearby cells. Each free cell in the grid will be given a virtual force vector composed from the forces generated by the boundary of all nearby obstacles around the viewpoint. For example, in Figure 1, the viewpoint A with an influence region of  $7 \times 7$  will be affected by cells  $v_0$ ,  $v_1$ , and  $v_2$ . If the virtual world layout is static, the virtual force field can be computed in a preprocessing step and stored in a 2D array for run-time lookups.

In the regular control loop of a navigation interface, a user's mouse input (dragged vector) is translated into an offset vector for updating the next configuration of the viewpoint ( $\overline{AB}$  as in the example of Figure 2). With the influence of the virtual force field, a force vector is computed for the current configuration A and transformed into a difference vector ( $\overline{AC}$ ). Then this vector is used to compose the final updating vector ( $\overline{AD}$ ) for the next viewpoint configuration.

### 3.2 Force Field Adjustment

How the static force field is applied to affect a user's navigation is determined by the interaction with the user. In [10], the virtual force is computed dynamically according to the following formula.

$$\vec{F}'_p = \mu_{user} \times \mu_v \times \mu_d \times \vec{F}_p. \quad (1)$$

$\vec{F}_p$  is the static force precomputed for a give point  $p$  while  $\vec{F}'_p$  is the force adjusted according to user interaction and control history. The adjustment parameters,  $\mu_{user}$ ,  $\mu_v$ , and  $\mu_d$ , are defined as follows.  $\mu_v = kv$ , where  $\mu_v$  is the velocity parameter,  $v$  is the magnitude of the current velocity, and  $k$  is a constant.  $\mu_d^n = \eta \times \mu_d^{n-1}$ , where  $\mu_d^n$  is the directional parameter for step  $n$ , and  $\eta$  is determined by the dot product of  $\vec{F}_p$  and the forward

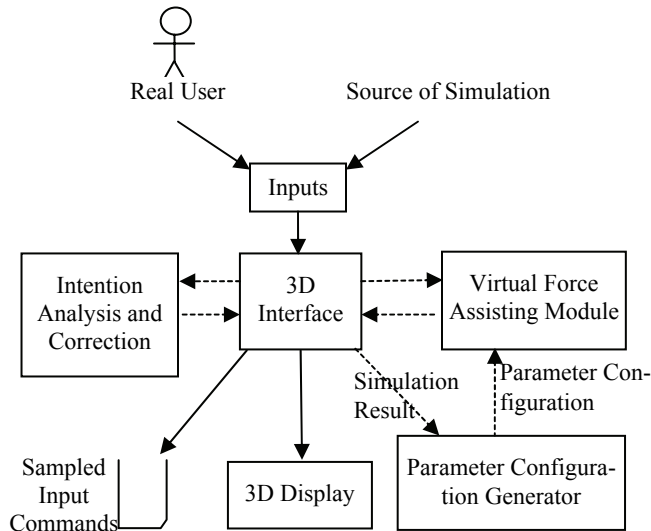


Figure 3. Overall system architecture.

facing vector. When the dot product is greater than zero,  $\eta = m$ , where  $m$  is a magnifying factor greater than 1. When the dot product is less than or equal to zero,  $\eta = n$ , where  $n$  is a shrinking factor less than 1.  $\mu_{user} = l$  is a user-preference factor specified by the user. The greater value of  $\mu_{user}$  means stronger force influence.

In [10],  $(l, k, m, n)$  are four fixed control parameters, called *parameter configuration*, that are determined by empirical means. However, it was reported that the improvement of navigation efficiency with such assisting mechanisms is greatly affected by scene variations and user characteristics. There is no single parameter configuration that is the best for everyone. Even if we want to find a good configuration for most people, tedious experiments need to be done. Therefore, it is highly desirable to have an automatic mechanism to find the best parameter configuration for applying virtual forces to a given user.

#### 4. DESIGNING ADJUSTABLE VIRTUAL FORCES

According to the experience of the above experiments, we know that the virtual force needs to be adaptive to a user in order to achieve the best improvement. Hence, we have designed two assisting mechanisms, *simulation experiments* and *dynamic adjustment*, which can be used to find the best parameter configuration in an off-line and on-line manner, respectively. The overall system architecture proposed in this paper is shown in Figure 3. The inputs to the system could be from a real user or from simulated data. The intention analysis and correction module is used to modify the simulated inputs so that the user's high level intention is kept. The parameter configuration generator is used to come up with a good configuration according to the simulation result. Then the virtual force assisting module uses this parameter configuration to compute the virtual force for updating the next viewpoint location.

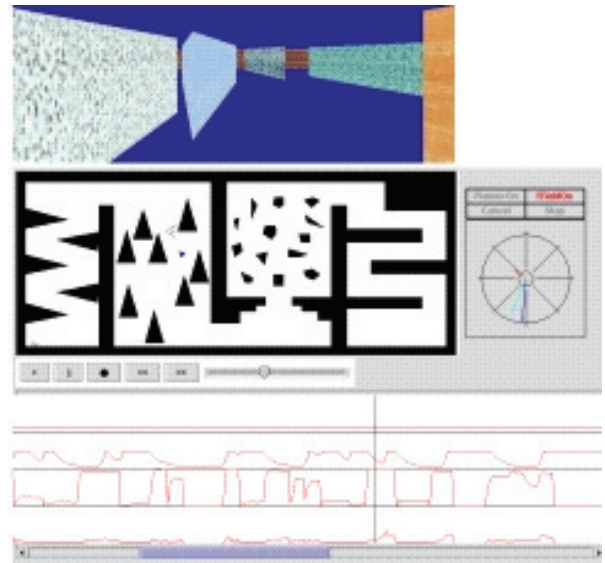


Figure 4. Graphical interface of the experiments. The upper part is the 3D display module while the middle portion is the 2D layout map. The lower portion shows the trending of the virtual force.

The simulation experiment method aims to search for the best parameter configuration for a user in a specific world in an off-line manner. We first ask the user to navigate through the world by accomplish a given task and record how the user controls the interface interactively under the influence of virtual force. We then try to simulate the user's input in an off-line manner by feeding the sampled input commands to the interface. In each simulation experiment, we use a fixed parameter configuration for the whole course and record the number of steps taken to reach the goal location. The optimal parameter configuration can be found after all possible configurations have been exhaustively tested.

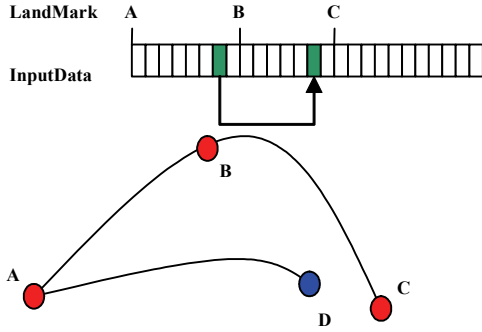
The dynamic adjustment method is an on-line version of the simulation experiment. The main difference is that the scale is smaller in order to make it feasible for on-line applications. In this mechanism, the simulation experiment is done only for a segment of a few steps in the past control inputs. A greedy algorithm is used, and only neighboring parameter configurations are evaluated to find a better configuration for the next step. Although we only move the parameter configuration one step at a time, the configuration will converge to the best value gradually after a few steps if there exists a unique optimal one.

We will describe the above two methods in more details in the next two sections and compare the effectiveness of the two methods in Section 7.

### 5. SIMULATION EXPERIMENT METHOD

#### 5.1 Recording Navigation Process

From [10] we know that the virtual force field mechanism can improve navigation efficient by reducing the number of steps taken to reach the goal from a given initial location. However, in order to understand how the virtual force helps in various occasions, we have to record the input commands during navigation and analyze how the virtual forces are computed for these inputs. In order to record the user inputs, we have implemented an inter-



**Figure 5. Using milestones to modify the short-term goal for simulation experiments. When the viewpoint moves from A to D, the simulation might skip milestone B due to a closer milestone C.**

face, as shown in Figure 4, to collect a user’s inputs (dragged vectors) as well as the navigation trajectory under the influence of virtual forces. With this raw data, the user’s navigation process can be simulated and reproduced.

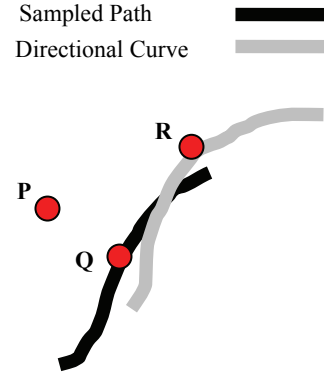
## 5.2 Simulation Experiment

The effectiveness of the virtual force mechanism is determined by the four-parameter configuration  $(l, k, m, n)$ , and it differs greatly among users. We aim to find a personalized configuration providing an optimal navigation performance for a given user. However, we do not want the user to repeat the navigation experiment in person on the other hand. In fact, a user’s interface control behavior might change when a user gets familiar with a specific scene. Therefore, we propose to use simulation to repeat the experiments to search for the optimal configuration. By optimal configuration, we mean the parameter configuration that can produce the virtual forces to assist the user to take the least number of steps to reach the goal location.

## 5.3 Intention Analysis and Input Correction

Since we have recorded the original input commands (dragged vectors) on the interactive interface, we can play back the navigation if all parameters remain the same. However, once we start to modify the parameter configuration, we are altering the virtual force applied to the interaction and the resulting path might not be the same as the original one. In a real experiment, a user can adjust the mouse input with visual feedback when the influence of virtual forces changes. However, in a simulation, if we keep the original inputs, the simulation might not generate productive or even useful movements. Instead, what we have to keep is the high-level intention behind the user’s input. Therefore, we propose two methods, called *milestone* and *directional curve*, to adjust the input command dynamically according to the current viewpoint location.

The first method creates milestones along the path by sampling points that the user passes through in the real navigation experiment. The milestones divide the input commands into segments. We assume that these milestones are the mid-term goals in each segment listed in sequence that reflect the user’s high-level intention. During the simulation, if the closest milestone is changed to one in a future segment, then all input commands in this segment are skipped. For example, in Figure 5, when the input commands are still in the segment around milestone A and the viewpoint has



**Figure 6. Using directional curve to modify orientation for simulation experiments.**

been moved to location D, which is closer to milestone C than to milestone B. In this case, all input commands behind milestone C will be skipped and next command will start in the segment containing milestone C.

The second method, called directional curve, is designed to fix the orientation problem of the input commands. When the viewpoint location deviates from the original path due to different virtual force influence, the input commands also need to be adjusted to make the viewpoint move toward the right direction. When a user navigates, he/she controls the interface according to the scene in front of his/her viewpoint. One can imagine that there exists a short-term goal point in front of each viewpoint location. We call the list of virtual points along the path the directional curve. We can use the points along this curve to adjust the moving direction of the viewpoint. For example, in Figure 6, we assume that the current viewpoint location is at P while the closest point along the sampled path is Q. Assume that the corresponding point on the directional curve for Q is point R. Then the vector PR becomes the adjusted direction for the viewpoint. In other words, we are using an imaginary point in front of the viewpoint along the navigation path to guide the viewpoint movement in the simulation experiments.

The low-level input commands are almost meaningless and cannot be used in simulation if it lacks the intention behind them. The two methods above are designed to maintain the high-level intention of a human user in the simulation by analyzing the navigation path traced by the user. As we will show in later sections, the simulation experiments can run by itself with these two methods even if the virtual forces vary significantly.

## 6. DYNAMIC ADJUSTMENT METHOD

Although the simulation experiments can find the optimal parameter configuration automatically, it can only be done in an off-line manner due to the time-consuming simulation process. For example, if we divide the range of each parameter into  $K$  intervals, the total number of simulation experiments need to be done is  $K^4$  before we can find the optimal one in a brute-force search. In addition, the optimal configuration may only be good for a specific scene even though it is likely to be a good starting point for the user. Consequently, we investigate the possibility of bringing

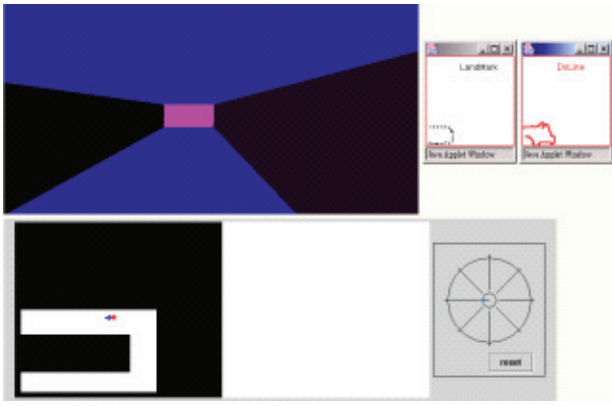


Figure 7. Simple experimental scene for testing the effectiveness of intention analysis.

the experiments on-line with a smaller scale. This method is called the dynamic adjustment method.

### 6.1 Window sampling for on-line search

In the dynamic adjustment method, we use a sliding window of size  $N$  and maintain the input commands for the latest  $N$  steps. For example, assume that the user has taken  $K$  steps, and the system will record the command inputs for the  $N$  steps between  $\text{MAX}(K-N+1, 0)$  and  $K$ . The system will use the last  $N$  steps to run the simulation experiment and try to obtain a better configuration, if exists, among the neighbors for the next step. In other words, we will simulate the last  $N$  steps with different sets of parameters to see which one takes the least number of steps to reach a goal if it were used. For example, assume that the parameter configuration is denoted by  $(A_1, A_2, A_3, A_4)$ . Then the neighboring configurations will be  $(A_1 - \Delta_1, A_2, A_3, A_4)$ ,  $(A_1 + \Delta_1, A_2, A_3, A_4)$ ,  $(A_1, A_2 - \Delta_2, A_3, A_4)$ ,  $\dots$ ,  $(A_1, A_2, A_3, A_4 + \Delta_4)$ . The best neighbor will be selected as the configuration for the next step. The configuration will converge to a local optimal one (if not global) gradually if the user keeps his/her navigation habits. Since the parameters are not specifically designed for a scene, the system should be able to adjust the configuration to a better one for any scenes in an on-line manner. In addition, when a user gets familiar with a scene or the navigation control, the degree of assistance that the user prefers might also change over time.

### 6.2 Adding the momentum mechanism

Although the dynamic adjustment mechanism described above is able to adjust the parameters toward a better one, the user may feel inconsistent if the parameters are changed too often simply because of scene variation. If we adjust the parameters according to the frame rate, which is typically above ten frames per second, the fluctuation on the forces may be too frequent for a user to control the interface in a consistent way. Therefore, we have added an inertia mechanism into the system such that unless enough momentum has been accumulated for a specific change, the on-line system will not adjust the configuration immediately. We keep a counter for each neighbor of the current parameter configuration and increment the counter when we attempt to visit it every time. A change is actually made only if the counter has reached a predefined threshold value, say 3. In other words, unless the changes in that direction have accumulated enough momentum, we do not make the move on the parameter changes.

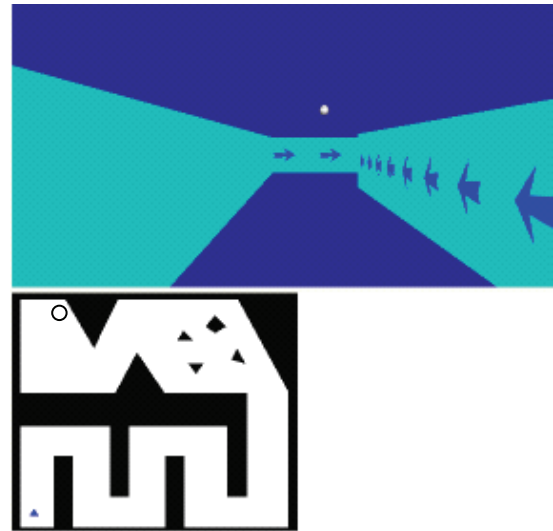


Figure 8. The 3D scene used in the navigation experiments. The circle on 2D map is the goal location.

With such a mechanism, we can maintain a more controllable interface by avoiding frequent fluctuation on the configuration as well as the resulting virtual forces.

## 7. EXPERIMENTS AND DISCUSSIONS

We have implemented the aforementioned virtual force adjustment mechanism in Java and integrated it into the 3D VRML [15] browser implemented by Blaxxun [1].

### 7.1 Effects of Using Intention Analysis

A key function in the simulation experiments is the intention analysis mechanism. In order to verify the effect of the intention analysis methods, we have done some experiments on the scene shown in Figure 7. The scene is kept simple intentionally in order to screen out other factors. In three simulation experiments with different parameter configurations, the viewpoint all gets stuck at the corner when no intention analysis and adjustment are used. On the other hand, if the intention analysis mechanism is used, the viewpoint can successfully reach the goal at the other end of the U-shaped hallway in all experiments. This shows that the high-level intention analysis is crucial for making the simulation feasible.

### 7.2 Comparisons of Simulation Experiments and Dynamic Adjustment Methods

Ten subjects are invited to do the experiment of comparing navigation efficiencies under different assisting mechanisms. Two of them (users A and B) are familiar with the control interface of VRML browsers while another two subjects (C and D) are familiar with 3D game controls. Three of the subjects (E, F, and G) have a good command of computers but not 3D interfaces. Another two subjects (H and J) only use computers occasionally (once a month) and the last one (I) never uses computers. We used the scene shown in Figure 8 for each subject to do the experiments with the assisting mechanisms of simulation experiments and dynamic adjustment.

In the simulation experiments, we first record the user's input commands without any assisting mechanisms. Then we ask the

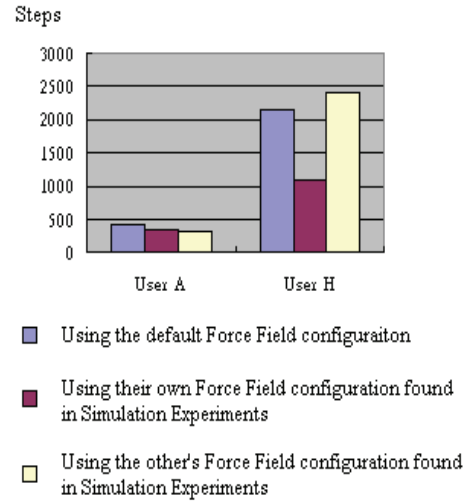
**Table 1. The number of steps taken to reach the goal location with four different mechanisms**

User	No Assisting Mechanisms		Using Default Parameter Configuration		Using Simulation Experiments		Using Dynamic Adjustment	
	Step	%	step	%	step	%	step	%
A	779	100	407	52.2	328	42.1	325.6	41.8
B	902	100	416	46.1	320	35.5	424.7	47.1
C	1838	100	491	26.7	548	29.8	555.7	30.2
D	1199	100	458	38.2	607	50.6	614.1	51.2
E	834	100	510	61.2	406	48.7	439.9	52.7
F	1715	100	840	49.0	792	46.2	585.8	34.2
G	1338	100	950	71.0	716	53.5	735.4	55.0
H	4111	100	987	24.0	771	18.8	803.9	19.6
I	7129	100	2143	30.1	1096	15.4	2150.4	30.2
J	1927	100	1061	55.1	725	37.6	963.4	50.0
Avg	2177.2	100	826.3	45.4	630.9	37.8	759.9	41.2

user to do two runs for each of the following two parameter configurations. One is the default parameter configuration, and the other one is the optimal configuration found for the user in the simulation experiments. We record the numbers of steps taken to reach the goal in each run and compute their average for each configuration. The experimental results are summarized in first three columns in Table 1. The number of steps of the runs without any assistance is used as a base (100%) to compare with other mechanisms. The second column shows the numbers of steps with the default configuration chosen carefully while the third column shows the performance with the optimal configuration. From these data, we have verified the effectiveness of using virtual force field as the assisting mechanism because the average number of steps is reduced to 45.4% with the default set of parameters. We also found that the performance can be further improved by 8% if we use the optimal configuration found in simulation experiments. This effect seems to be more significant for novice users such as subject I, whose number of steps is reduced by around 15% if the optimal configuration is used.

In the experiment of dynamic adjustment, we choose three initial configurations for each subject to run. For each of the three initial configurations, we ask the subject to repeat the navigation experiments three times and take the average of the night experiments at the end. The result is shown in the fourth column in Table 1. We do not expect the performance of the dynamic adjustment mechanism to be better than simulation experiments since its advantages are on the on-line characteristic and flexibility to adapt to the world. We found that the numbers of steps for four users (A, C, D, and H) are almost the same as the ones using the optimal configuration. One of them (Subject F) even has better performance with the dynamic adjustment mechanism.

In order to compare user discrepancy on the optimal configuration, we have chosen two subjects (A and H) to do the comparisons. Each subject is asked to run the experiments with the optimal



**Figure 9. Comparing the effects of different parameter configurations on different users**

configuration for the other subject. For example, user A is asked to run the optimal configuration for user H and vice versa. The result is shown in Figure 9. We found that the sensitivity of each user to the virtual force configuration might vary significantly. For novice users such as user H, using other people optimal configuration might be even worse than using the default configuration. Therefore, we can conclude that user discrepancy does exist in adopting the assisting mechanisms and the adaptive interface can greatly benefit novice users in providing just-in-time navigation assistance.

## 8. CONCLUSION AND FUTURE WORK

In our previous work, we have shown that the assisting mechanism with virtual force can greatly improve navigation efficiency by reducing unnecessary collisions with environmental obstacles. However, the discrepancy between users in adopting the intelligent interfaces has inspired the work reported in this paper. We have designed two adaptive mechanisms to find a set of good parameters automatically for a specific user. Our experiments show that the simulation experiments can find the best virtual force parameters but the simulation can only be done in an off-line manner due to the time-consuming search process. The on-line dynamic adjustment method, on the other hand, does not require any prior analysis and could be as effective as using a fixed optimal configuration found in simulation experiments. The adaptiveness of such a mechanism to scene variations and personal differences also make it attractive in more 3D applications.

In order to simplify the system experiments, we only study the effectiveness of virtual forces in the current system. However, we know that the path planning method is also effective in guiding a user when a collision does happen. Therefore, in the future we will try to incorporate the automatically generated escape path into the virtual force field to make the navigation even more efficient. In addition to improving the navigation efficiency passively by providing virtual forces for a given scene, we are also working in the direction of redesigning a virtual scene with a better layout or more informative direction signs at appropriate places. We believe that a good assisting mechanism on a well designed scene

will provide an efficient and pleasant navigation for general way finding.

## 9. ACKNOWLEDGEMENT

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