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B.S (Columbia University) 2009

THESIS

Submitted in partial satisfaction of the requirements for the degree of

DOCTOR OF PHILOSOPHY

in

Mechanical and Aerospace Engineering

in the

OFFICE OF GRADUATE STUDIES

of the

UNIVERSITY OF CALIFORNIA

DAVIS

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2017

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June 2017  
Mechanical and Aerospace Engineering

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## **Abstract**

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## Acknowledgments

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# Chapter 1

## Introduction

### 1.1 Overview

The motivation for the research proposed stems from the design of complex human-machine interfaces, particularly cockpits for spacecraft and high performance aircraft. A common problem is that the design (or modernization) of a cockpit will go into flight before evaluation pilots are completely satisfied with the design process, due to time and budget concerns. The remaining deficiencies of the cockpit design are expected to be overcome through additional crew training. These deficiencies represent a cost throughout the lifetime of the vehicle in training, and “unknown” deficiencies of the human-machine system are often discovered too late in accident reports. To address these issues, we have designed and prototyped a system to allow pilot evaluations to occur earlier and more frequently in the design process. The aim of our system is to give a non-functioning geometric mockup the interactivity of a simulator. The mockup, which is used in the early stage of the design process, can now be used for more extensive cockpit evaluations, with a quicker iteration time than a simulator requires. Additionally, our evaluation of this prototype system has led to a novel metric for quantitative evaluation of cockpit design based on hand tracking of the pilots’ motion.

The prototype system, called the “Rapidly Reconfigurable Research Cockpit”, consists of a virtual reality (VR) headset, optical hand tracking device, and 3D printed cockpit surfaces (Figure 1). The VR head-mounted display (HMD) provides a virtual view

of the cockpit environment. A challenging aspect of immersive VR simulators like this, however, is to provide an appropriate tactile feel of the objects being interacted with (in our case, the instruments of a cockpit). We solve this by using 3D-printed instruments co-located with the virtual environment, which is typically called “passive haptics”. The instruments remain inert and non-functioning in the real world, but appear fully functional in the virtual world presented to the user. The user can interact with the virtual cockpit (i.e. pressing buttons) with the use of an optical based hand tracker, which measures the position of the pilots’ hand. This position is used to determine if the user is activating a button. Measurements from the hand tracker are used to display the essential visual feedback of hand position in the virtual world, as normally the use of an HMD blocks the view of one’s own body.

We have used this prototype system to conduct pilot studies to shape the direction of the research plan. The three main tracks of the research can be summarized in these aims:

1. Create a virtual reality system with passive haptics and hand tracking that allows for evaluation of early design cockpits.
2. Evaluate the performance, quantitatively and qualitatively, of the virtual reality environment and user interface, identifying the best practices for using such a system.
3. Develop a quantitative metric for “task familiarity” using data recorded from hand tracking whilst using the virtual cockpit, and investigate how the model changes based on factors such as workload, cockpit layout, environment, training level, etc.

The first aim has been completed, while significant progress has been made on the second. The final aim is in early stages of study, but results to date are promising.

Three main experiments have been planned (one of which is completed) to support these aims. The first experiment asked subjects to repeatedly target one of four buttons on the panel under different conditions, testing differences between real versus virtual world as well as passive haptics versus no passive haptics (mid-air reaching). We recorded time to completion and accuracy to the button center. It was found that the virtual world caused

a slight but significant degradation in time for the reaching movement, but no significant change in accuracy. The accuracy requirement is more important for our application of safety-critical human-machine interfaces, where accuracy requirements often prevail over movement time. Experiment 2 will have users complete a “Fitts’ Circle” (page 10, Figure 4) in different conditions to again test the passive haptics versus mid-air reaching. The recorded 3D hand trajectory data will also be used to develop the metric for Aim 3 (which is being called the “task familiarity” metric or model in this report). By using the Fitts’ circle for Experiment 2, we can use the Fitts’ throughput as a truth measure of familiarity to train and test the model upon. This model will then be tested in a more realistic cockpit situation in Experiment 3. Additionally, the second experiment will undertake a more extensive subject survey to better gauge how users find the environment, especially the difference between passive haptics and mid-air reaching.

The goal of the metric of “task familiarity” is to quantifiably measure how familiar the pilot is with the current cockpit design and task. The metric will be developed by creating a statistical characterization of the user’s 3D button-to-button trajectory movements while evaluating a cockpit design in our system. The analysis of each trajectory will create a set of attributes based on current models and metrics found in literature, discussed in detail later. We can then collect a set of trajectories for a particular experimental condition or subject demographic (i.e. pilots vs. not pilots), and by analyzing the set of trajectories measured, a predictive model for these parameters can be developed. For experiment 2, the major experimental condition will be the Fitts’ throughput measure used as an analog for training level. An analysis of data from a previous preliminary study has supported this approach, but Experiments 2 and 3 will further develop and test this model. If the “task familiarity” model can predict training level and model the training curve for a pilot in a new interface, it could provide a way to differentiate between a poorly designed cockpit and an evaluation pilot who has simply not had enough time to fully train to the new cockpit.

To frame this proposed research, the next chapter will summarize previous work relevant to this research proposal. The first few sections deal with virtual reality: different haptic environments and evaluation of virtual environments. Next input device evaluation research such as Fitts’ Law is discussed, highlighting work that has been performed eval-

uations in virtual environments. The final section discusses the trajectory models that we propose using as a basis for extension into a “task familiarity” metric.

## 1.2 Literature Review

### 1.2.1 Virtual Tactile Environments

One of the challenges of virtual environments has been presenting the user with a tactile sensory input that matches the visual sensory inputs they are experiencing. Many haptic glove products have been created that attempt to simulate this with motor vibrations. Since they are anchored to your body, they cannot provide any external stopping force, so the simulation can break down when interacting with rigid objects. Other devices use a stylus mounted to a 6-DOF actuator that can simulate forces, but these have been found to be best suited for surgical tasks, when a tool would be used in the real scenario as well. The focus of this work is the idea of *passive haptics*, whereby we recreate the tactile environment to provide the sense of touch and external forces. The disadvantage is that the tactile environment cannot dynamically change during the simulation. Some researchers have attempted to create robotic passive haptics (McNeely & others, 1993; Tachi, Maeda, Hirata, & Hoshino, 1994), which use a robotic arm to bring the appropriate tactile experience to the user as they reach into the virtual world. This quickly becomes complicated and expensive. Others have tried approaches to morph the virtual world to adapt to a generic passive haptic device (Kohli, 2009; Kohli, Whitton, & Brooks, 2012), which can work only for small discrepancies. Fortunately, the adaptability (during simulation) of the passive haptic environment is not a requirement for the use case of a pilot evaluating a new cockpit interface. We still retain the ability to quickly change the environment via 3D printing new parts in between simulation sessions.

The first appearance of work similar to ours was by Schiefele et al. (1998) in Germany, which also used a head-mounted display (HMD) to replace the visuals of a flight simulator (though technology has drastically changed in HMD’s since their work). They recreated only the positions of the panels with a flat plate and then used a magnetic hand tracker for reading the pilot input. A short usability study was presented, and they found

that users could activate switches, knobs and dials in less time with the physical panel than without (i.e. reaching into mid-air). Insko (2001) investigated the impact of passive haptics on generic virtual environments. It was found that the passive haptics increased the sense of presence felt by users. An experiment had two groups complete a maze in a virtual environment, one group with passive haptics on boundaries, and one without. After training the passive haptic group bumped into far fewer obstacles than the group without.

Borst and Volz (2005) created a “mixed” haptic virtual environment, whereby they combined an active haptic glove with a passive haptic panel. The panel provided the stopping force and depth location of the interface, while the glove haptics provided the finer details of the buttons, switches and dials. They compared subjects performing tasks on the panel using the glove by itself (no panel), the passive haptics by itself, as well as their mixed condition. They found the glove by itself consistently performed worse than the other two, but overall no significant difference existed between passive haptics and their mixed haptics.

### 1.2.2 Virtual Environment Evaluation

It is notoriously hard to quantify the effectiveness of a virtual environment (VE). One can report on quantitative measures of the virtual reality (VR) setup such as latency, field of view, resolution and others, but in the end the only useful measure is whether the user finds the virtual environment convincing. This can also change depending on the user – one user could find a VE convincing while another finds it intolerable. The effectiveness of a VE is often linked to the sense of *presence*, defined “as the subjective experience of being in one place or environment, even when one is physically situated in another” (Witmer & Singer, 1998). Sheridan (1992) argued that presence is a subjective sensation that necessitates a subjective measure. It is important to quantify this as increased presence has been shown to improve task performance (Youngblut & Huie, 2003). Two of the more popular post-questionnaires that have been developed are the Presence Questionnaire (PQ) (Witmer & Singer, 1998) and the Slater-Usoh-Steed (SUS) questionnaire (Slater, Usoh, & Steed, 1994). The former (PQ) consists of 32 questions that cover various factors of presence such as sensory, control, distraction and realism. The SUS questionnaire consists of only 6 questions and is focused on the feeling of presence. Nystad & Sebok (Nystad & Sebok,

2004) found the PQ to be more sensitive to technology and interaction methods, while the SUS fits better to the definition of presence. This suggests PQ may be better suited for our work when comparing between conditions, but SUS could be useful as a general evaluation of the overall system. The increased presence with passive haptics found by Insko (2001) is an important finding we aim to replicate for our virtual environment.

We also expect that our system will reduce the amount of arm fatigue experienced by the operator compared to a mid-air pointing virtual environment (i.e. no tactile workstation present). This common phenomenon in virtual environments is often called gorilla arms, as people feel that their arms are too heavy to support after prolonged use reaching out into the void. A measure of this arm fatigue is not included in the standard presence questionnaire. The literature does not contain a large volume of experiments that measure arm fatigue. Qualitative measures include the Likert scale, NASA Task Load Index (NASA TLX), the ISO-9241-9 “Assessment of Comfort and Fatigue” and the Borg scales (RPE and CR10). The Borg CR10 performs favorably to the other scales for measures of exertion (Grant et al., 1999). Recently, a quantitative approach has been proposed that uses optical tracking of the arm and wrist which validated well against a Borg CR10 scale (Hincapi-Ramos, Guo, Moghadasian, & Irani, 2014). Adapting their software to our system would be out of scope (and requires more upper-limb measurement than we are currently capable of) but the validation against the Borg CR10 scale is promising.

### 1.2.3 Input Device Evaluation

As we endeavor to create the input method for interacting with the passive haptic cockpit panel, we must ask how to evaluate and report on the input device we have created. For decades one of the most important tools for the Human-Computer Interface (HCI) community has been Fitts’ Law. It has been used to guide both input device inventions (i.e. mice, trackballs, joysticks, etc.) as well as interface design guidelines. A summary of the field is presented here, with notable contributions and similar work following.

## Fitts' Law

Paul J. Fitts (1954) originally reported on his experiments of targeted reaching movements between varying target sizes, and compared the linear trend he discovered to information theory models. He postulated that the entire control system of the human, from motor control to visual and proprioceptive feedback, can process information at a certain maximum rate, and the bandwidth of the channel must be derivable from the physical parameters of the target. Fitts' adapted Shannons' Theorem (Shannon, 1949) from information theory for the bandwidth of a channel in the presence of a signal and noise. The distance to the target ( $D$ ) is considered the signal, and the width of the target ( $W$ ) is considered to be the noise (Figure 2). This analogy allowed Fitts' to create a formula for the *index of difficulty* (ID) of a target:

$$ID = \log\left(\frac{2D}{W}\right) \quad (1.1)$$

It was found experimentally that the index of difficulty is linearly correlated to *movement time* (MT), the time taken to move between the start and the target. This leads to the equation that is now known as Fitts' Law:

$$MT = a + b \cdot ID \quad (1.2)$$

where  $a$  and  $b$  are the parameters of the linear regression, which are determined experimentally. A typical experiment consists of subjects performing movements to a range of targets with varying index of difficulty, and recording their movement time. The data is then fit to a linear regression, as shown in Figure 3.

Since its original formulation, refinements to the model have focused on the formula for index of difficulty, ID, and the definition of its dependencies, width and distance. Crossman (1957) suggested a very important refinement called the adjustment for accuracy. This provides a correction to the ID to compensate for the task which is actually performed by the subject as opposed to the task that was asked to be performed. For example, if a subject is presented with two large targets close together ( $D \ll W$ ), then he/she tends to aim towards the inner edges of the targets instead of the center, effectively using a much

smaller target width. The adjustment for accuracy replaces the width,  $W$ , with the *effective width*,  $W_e$ . The basis for this adjustment is that the information theory basis for Fitts' Law expects that the “noise” of the target width induces a 4% error rate (I. S. MacKenzie, 1992). If more or less than 96% of the target hits are successfully within the target, then the target width must be adjusted to fit the target error rate of 4%. Motor tasks have been shown to be a Gaussian distribution (Crossman & Goodeve, 1983; Welford, 1968; Woodworth, 1899), thus 96% of target endpoints will be within 4.133 standard deviations ( $\sigma$ ).

$$W_e = 4.133 \sigma$$

It is best to calculate the effective width for each subject in each condition as it is sensitive to both between and within subject variability. The effective width can then be used to calculate an *effective index of difficulty*,  $ID_e$ . The effective width described here is for one dimensional targets, but there are extensions to 2D targets as well (Murata, 1999).

The form of the logarithmic term in the index of difficulty equation (Eq. 1) has encountered some debate over the years. Fitts did not describe the reasoning for including the factor of two in his original formulation, and Welford (1968) suggested a modification that provided a better fit for low values of index of difficulty. The Welford modification simply adds

$$ID = \left( \frac{D}{W} + 0.5 \right)$$

A more recent variation noted by I.S. MacKenzie (1989) suggests that by comparing to Shannons' Theorem as Fitts originally did, the form should be,

$$ID = \left( \frac{D + W}{W} \right) = \left( \frac{D}{W} + 1 \right)$$

which has been widely adopted in literature as the Shannon's Formulation. Aside from its similar structure to Shannon's Theorem, it also provides the benefit that it cannot produce a negative index of difficulty.

The first use of Fitts' Law in the human-computer interface (HCI) community was with Card et al. (1978), comparing the use of a mouse, joystick and keys for text selection.

This led to an extensive use of Fitts' Law in the HCI field, both as a way to evaluate input devices and a design law for user interfaces.

Best practices of using Fitts' Law have been developed, and the standard ISO-9241-9:2000 (International Organization for Standardization, 2000) provided some guidelines for the use of Fitts' Law in evaluation of input devices. Most notably, it was suggested to use the ISO-9241 "Fitts' circle" for evaluations of 2D tasks (Figure 4). This input pattern has been adopted widely in literature, and provides a standard for comparing across studies.

Soukoreff & I.S. MacKenzie (2004) present a review of the best practices for designing and performing a Fitts' law experiment, including the use of the Shannon Formulation and the ISO-9241 "Fitts' circle". They also advise the use a large range of ID values, ideally from 2 to 8 bits. Many experiments achieve this by varying distance and width, but Guiard (2009) points out the inconsistency of designing for a range of ID based on two factors, and shows that by holding one fixed lowers the regression error, while reducing experiment complexity. Wobbrock (2011) provided an independent validation of this, showing that significant effects remained when retroactively removing the varying diameter of targets.

When used to evaluate input devices, the metric of comparison is often *throughput*. Throughput (TP) is defined as the index of difficulty (ID) over the movement time (MT).

$$TP = \frac{ID}{MT}$$

The units of throughput are bits per second, and it represents the information capacity of the input device. It has been validated to be invariant to subjects prioritizing speed or accuracy (I. S. MacKenzie & Isokoski, 2008). A higher throughput indicates that the human is able to convey more information through one input device versus another.

The debate on which form and refinement of Fitts' Law is the best to use continues to this day (Drewes, 2010; Hoffmann, 2013), and this work will not attempt to settle this debate. However, the understanding of the origin of Fitts' Law, best practices and the many forms of the equations will greatly influence the design and success of the experiments performed. Since no form of the equations have been rejected outright in the literature, it is recommended to perform analysis with all versions and note if any significant differences

in major conclusions result (Soukoreff & MacKenzie, 2004).

### Dimensionality of Fitts' Law

There are two dimensionalities to consider when discussing Fitts' Law experiments: the dimensionality of the targets or task, and the dimensionality of the movement. A 1D task would have the targets aligned along a single axis, whereas a 2D task would have them located on a plane (such as the Fitts' circle in Figure 4). The movement dimensionality is how many dimensions the subject is allowed to move within between the targets. For example, the original Fitts' Law task was a 1D task as the targets were separated in a line and no discretion was used in the transverse direction. The movement was effectively 2D, along the targets and up/down from the table.

The popularity of Fitts' Law in the HCI field has led many researchers to focus on the use of Fitts' Law on a 2D computer screen controlled by a mouse. This is a 2D task, with a constrained 2D movement. Our use of Fitts' Law will be a 2D task, since the buttons are configured on a plane on the cockpit panel, with unconstrained 3D movement. There are some studies that present this task, especially with the advent of touchscreen devices. Some have proposed minor variations to the formulation of Fitts' law to adjust for touchscreens (Bi, Li, & Zhai, 2013; Sears & Shneiderman, 1991), but others have found the original formulation to perform well (Scott MacKenzie, 2015). There are also studies discussed in the next section that perform this task in a virtual environment.

Refinements have been proposed to account for the possible effects of extending the task dimensionality to 2D, such as the bivariate target model (Accot & Zhai, 2003; I. S. MacKenzie & Buxton, 1992; Wobbrock et al., 2011). Some work has also explored this for 3D tasks, introducing a trivariate model (Grossman & Balakrishnan, 2004) and corrections for 3D placement of targets (Cha & Myung, 2013; Murata & Iwase, 2001). These experiments place targets in a truly 3D space, with of course 3D motion required of the user.

## Applying Fitts' to Virtual Environment

Fitts' Law has been used in evaluating virtual environments and their input devices. Chun et al. (2004) evaluated a set of 3D stereo displays with a single haptic-enabled stylus using a Fitts' tapping task. They did not have well-fit regressions, but this could have been due to a small range of ID values used (2-3), targets being placed in 3D space (a 3D task with 3D movement) with random order, and averaging across subjects before completing the regression. Teather & Stuerzlinger (2011) performed a similar task and also found an underperformance of the regression in the fully 3D task. One condition performed a mid-air 2D planar task with 3D movement, and no significant difference in throughput was found compared to the same task with constrained 2D motion. Liu et al. (Liu, Van Liere, Nieuwenhuizen, & Martens, 2009) performed a planar multi-directional Fitts' task with a stereoscopic display, and compared virtual world to real world results, finding movement time twice as long in the virtual condition. The trajectories were analyzed to determine where the extra movement time was spent, which is discussed in a later section (Trajectories in Virtual Environments).

One of the few experiments that utilize Fitts' Law in an immersive HMD was performed by Kohli et al. (2012) who were focused on evaluating discrepant (warped virtual space) virtual environments to fool the user of the nature of the physical world. They had the subjects perform an ISO-9241 Fitts' circle on a panel placed in front of them, and in some conditions the panel was rotated in the virtual world but not in the real world to create the discrepancy. The subjects' hand movement was warped to compensate for this.

Coelho & Verbeek (2014) evaluated the LeapMotion hand tracker (the same technology we are using) in a 3D Fitts' task (with visual feedback on a computer screen), but did not perform a Fitts' model analysis, comparing only time. Seixas et al. (2015) performed an ISO 9241-9 Fitts' circle task with the LeapMotion. Their goal was to compare a gesture based selection technique, and they provide a Fitts' throughput comparison between that and direct mid-air selection.

### 1.2.4 Human Targeting Motion Trajectories

#### Trajectory Models

Fitts' Law provides a validated model for predicting movement time of a human movement, but this only provides part of the story. Analyzing overall time of movement simplifies the complex trajectory that a human forms in their movement, but it leaves us with only one measure of performance. With the use of the hand tracker we can quantify a complete 4D dataset of the reaching movement – 3D position and time. This allows us to investigate a more complete picture of the motion of the subject than has been technologically possible until the last few years. The trajectory of a human arm motion in a point-to-point multi-joint arm movement can take an infinite number of paths due to the complex joint structure of the human arm. It has been observed in many experiments, however, that humans tend to follow a common trajectory – a roughly straight spatial path, with a bell-shaped velocity profile. This observation has led researchers to create mathematical models to predict the trajectory.

The *minimum jerk* model was first proposed by Flash & Hogan (1985), and provided very good agreement to experimental data. Their hypothesis for the model is that the human hand trajectory in point-to-point movements follows the trajectory that supplies the minimum amount of jerk (time derivative of acceleration). This provides a simple optimization criterion,

where the  $x$ ,  $y$ , and  $z$  are the position coordinates, and the integral covers the time period of interest,  $t$ . The model has remained popular due to its simplicity, and ability to provide a closed form solution. It is a purely kinematic model, meaning that it does not require any knowledge of the human arm itself. The model can be solved for an optimized trajectory with only the initial and final conditions. If Eq. 2 is solved with initial and final conditions of zero velocity and acceleration, then this provides a closed form and decoupled solution, presented here for  $x$ , but  $y$  and  $z$  take the same form:

where  $x_0$  and  $x_f$  are initial and final position, respectively, and  $t_f$  is the time of the movement. A plot of the solution of Eq. 3 for position and velocity is shown in Figure 5.

The minimum jerk model performs well for many trajectories, however since it relies only on the kinematic variables, it does not take into account arm length, external force or load, or range of motions. Uno et al. (1989) proposed his *minimum torque* model based on the theory that humans must plan their movement based on a physical variable, and found that minimizing torque at each joint provided a good fit to experiment. The minimum torque model was refined to the minimum torque-change and renamed the *minimum commanded torque* model (Nakano et al., 1999), and the optimization criterion is presented here.

$$C_T = \frac{1}{2} \int_0^{t_f} \sum_{i=1}^N \left( \frac{d\tau_i}{dt} \right)^2 dt$$

where  $\tau_i$  is the torque at the  $i$ -th joint.  $N$  is the number of joints and  $t$  the time of the movement. Since jerk is dependent on torque, it is not surprising that the trajectories predicted using this model are quite similar to those of the minimum jerk model (Breteler, Meulenbroek, & Gielen, 2002). The differences are that due to the modeling of the kinematics of the arm the minimum torque-change model predicts curvature in the trajectory positional data (Figure 6). As shown in Figure 5, the minimum jerk predicts a symmetric velocity profile, scaled by time and distance. Several experiments have shown that the deceleration phase (time from peak velocity to target) is longer than the acceleration phase (time to peak velocity) (C. L. MacKenzie, Marteniuk, Dugas, Liske, & Eickmeier, 1987; Milner, 1992; Uno et al., 1989). The minimum torque-change model more accurately depicts this velocity asymmetry.

C. L. MacKenzie (1987) showed that there was correlation between the target size and the time of the deceleration phase for a Fitts' task, suggesting that subjects will adapt their trajectory for accuracy requirements. Our application to a cockpit environment is one where accuracy demands are often more important than time demands for manipulation of instruments.

## **Adaption to Novel Dynamics/Environments**

One of the assumptions of the minimum jerk and torque models is that the subject is performing rapid aimed movements (i.e. well trained and no cognitive workload), much like the requirement of Fitts' Law. This implies that the trajectories do not match the model as well when the subject is learning the pattern of movement, or introduced to a novel situation. Most articles do not report on the learning phase and report data after training effects have worn off. The learning phase is something we are interesting in characterizing for the development of our trajectory based training level model. There are a few studies that have investigated the adaptation of subjects to novel dynamics, by adding artificial forces to an end effector that is moved by the subject (Conditt, Gandolfo, & Mussa-Ivaldi, 1997; Scheidt, Reinkensmeyer, Conditt, Rymer, & Mussa-Ivaldi, 2000). These show that subjects initially experience deviations from the minimum torque model, but will approach the optimal movement over a number of trials.

## **Trajectories in Virtual Environments**

One of the earliest experiments to measure trajectories of human targeting motions (Woodworth, 1899) discovered that there are two distinct phases of the targeting motion: a ballistic phase and a corrective phase. Subjects typically perform the bulk of their movement in the ballistic phase, often considered more of an open-loop or feed-forward mechanism, and as they near the target switch to a feedback, closed-loop, corrective phase (Elliott, Binsted, & Heath, 1999).

Experiments in virtual environments have shown that users typically spend more time in the corrective movement phase than when performing the same task in a non-virtual environment (Liu et al., 2009). However, this work was completed for a mid-air interaction, so it is an open question whether the introduction of passive haptics would reduce the corrective phase penalty for operating in a virtual environment. We can quantitatively investigate this question with the use of our virtual environment.

## Other Trajectory Metrics

Our work so far has focused on the use of the minimum jerk and its derived models, but there have been other metrics developed in the literature to evaluate the trajectory of a targeting motion. Meyer (1988) developed a parsing criterion to divide the trajectory into submovements and defined three types: return from overshoot, accelerate due to undershoot, and slowing towards target. MacKenzie et al. (2001) presented a set of accuracy measures for targeting motions based on trajectory information. They were developed to compliment a Fitts' input device evaluation. The measures are named target re-entry, task axis crossing, movement direction change, orthogonal direction change, movement variability, movement error, and movement offset. They quantify types of movement that are not along the ideal straight-line path from start to goal.

### 1.2.5 Summary

The combination of immersive virtual reality simulator with an inert passive haptic has not been extensively tested in the virtual reality literature. Other approaches at immersive haptics have focused on creating a dynamic environment, but our application of a static cockpit does not need to change geometrically during the simulation (not to be confused with our re-configurability between simulations). The integration of new hand tracking technologies is a further innovation, allowing us to provide interactivity with the passive haptics.

To evaluate this new virtual environment, we will use two validated measures – presence and Fitts' Law. Fitts' Law is a well validated and actively used model for evaluating both input devices and interfaces. To date there have been only a few uses of Fitts' Law in immersive head-mounted display virtual environments. As of our knowledge, no one has performed a Fitts' Law experiment in a virtual environment using the new generation optical hand trackers (i.e. like the LeapMotion we are using). A Fitts' Law characterization is an important step to take to evaluate the system we have created, and provides a standard means of comparison across previous and future literature.

The minimum jerk and minimum torque change trajectory models have been de-

veloped to answer the fundamental questions of how humans plan their reaching movements. The validation of these models to experimental data in the literature supports these theories. As opposed to matching experimental data with theory, our approach will utilize the minimum jerk and minimum torque change models to evaluate the performance, or “task familiarity” of subjects. This leads to a quantifiable metric of the pilots’ familiarity with the cockpit under review, which would be an invaluable new tool for cockpit designers.

### 1.3 Hypotheses

This section serves as a brief summary of the major hypotheses we are aiming to test with this work that has not been fully addressed in the literature. This list highlights the motivations for the experimental designs and modeling objectives presented in the next two sections.

- The use of an immersive virtual reality environment with optically tracked hands compared to the real world has no effect on the accuracy of users selecting cockpit-sized buttons
- The use of passive tactile feedback in a virtual environment compared to no tactile feedback (mid-air reaching) has an effect on...
  - ...the performance (time and accuracy) of users selecting buttons
  - ...the presence sensed by the users
  - ...the Fitt’s throughput of the users
  - ...the button-to-button trajectory of the users
- The similarity of the recorded hand trajectory of the user to the minimum jerk or torque-change model is correlated to the Fitts’ throughput of the subject.
- The similarity of the recorded hand trajectory of the user to the minimum jerk or torque-change model is affected by...
  - ...the training level of the user.

- ... the workload of the user.
- ... the environmental conditions of the user.

# Chapter 2

# Prototype Design

## 2.1 Overview

In this chapter the progression of the Rapidly Reconfigurable Research Cockpit prototype is described. Three major versions of the prototype are outlined here and explained how they evolved from each other. The first prototype was not used in a formal experiment, but two versions based on the second prototype were used for Experiment 1 (Chapter 3) and 2 (Chapter 4). The final prototype was used in the third experiment (Chapter 5).

## 2.2 Requirements

In this section, the requirements we developed for the Rapidly Reconfigurable Research Cockpit (R3C) prototype are outlined. These requirements guided the development of the prototypes, based on the technology available.

**Rapid prototyping of physical mockup.** The main motivation for this system is that it enables rapid redesign of cockpit systems undergoing design evaluations. To permit this, the R3C system should allow for rapid changes in the physical world without needed to reconfigure the software heavily, or reconfigure any hardware for interaction.

**Minimal setup and calibration.** Another goal of the system is that it can be used “on top” of existing mockup systems. One consequence of this means that as much as

possible, the hardware used must be easy to set up and able to use in constrained spaces. This eliminates some of the high-end head mounted displays that use complicated head tracker systems requiring precise calibration. Additionally, this eliminates many of the past approaches to hand tracking which use multiple cameras precisely calibrated, or magnetic sensors in a well-defined magnetic field. The minimal setup also leads to the need for hand tracking, as the recognition of the user input must be done without outfitting the mockup with working input actuators (i.e. buttons).

**Immersive head mounted display.** The visual system needed.

**Unobtrusive hand tracking.** The hand tracking.

## 2.3 Prototype Development

### 2.3.1 First Prototype

The first prototype was the proof of concept of the system, and before it could be used for research studies, it was quickly outdated with new technology and software updates. It utilized the first generation Oculus Rift Development Kit and the first generation of software for the LeapMotion hand tracker. The basis of the software developed during this prototype phase continued throughout all of the prototypes.

#### Rendering Engine

We are using a NASA developed rendering engine named EDGE to provide the visuals for the virtual scene rendered in the head-mounted display (HMD). EDGE is highly customizable and extendable through C/C++ plugins, Tcl scripts, or networking functions. Many of the integrations described in the prototypes were created as EDGE plugins.

#### Virtual Reality Headset

The head-mounted display (HMD) we used for the first prototype was an Oculus Rift Development Kit. The lightweight headset provides an immersive virtual reality experience by combining a wide field-of-view scene with accurate head tracking, giving a

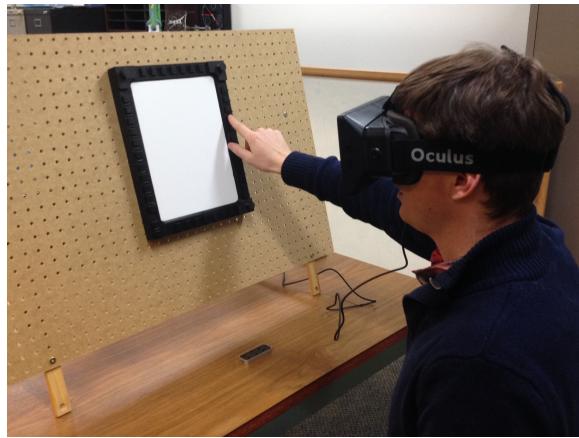
stable virtual world. The small display (cell phone sized) is viewed through a single set of lenses, and the barrel distortion of the lens is corrected for in the rendering engine by a pincushion distortion of the rendered scene. The orientation of the head is tracked using internal sensors (accelerometer, gyroscope and magnetometer). The software developers kit (SDK) exposes the head orientation and provides the appropriate distortion for a rendered scene (which is done for each eye), and this has been developed into an EDGE plugin. The first generation Oculus Rift Development Kit has a 1280x800 LCD screen (with 640x800 for each eye) with a refresh rate of 60Hz. The head orientation sensors do not support position, and experienced drift after a few minutes.

### 3D Printed Instruments

A central tenant of our technical approach is to use physical, geometrically accurate instrument shapes that provide no functionality (i.e. no screens, no working buttons, etc.). This is intended to imitate the fidelity of a typical early design stage mockup. In order to achieve this we have produced 3D-printed “instruments” that can be easily rearranged on a panel mount (pegboard). Since the devices are rapidly prototyped, they can be redesigned in a much smaller time frame than typical simulator instruments. By using the geometrically accurate instruments at the correct cockpit locations, the user is provided with accurate tactile and proprioceptive feedback without the need for entering the challenging field of virtual haptic feedback.

To use an instrument in the R3C system, it needs two versions of the 3D model. The first is the version for the rendering engine, which requires textures and possibly work to reduce the number of polygons so that rendering is not slowed down by the model. The second is for the 3D printing, which depending on the 3D printer used, will often require small changes to allow for a successful print. Both of these require a surface mesh model. Often, the original version exists as a CAD (Computer-Aided Design) model which typically does not describe the surface mesh directly but the ... The difficulty of the conversion from CAD to 3D model mesh can depend on the model, and the requirements of the rendering engine.

Our demo instrument was developed based on a large multi-function display with



**Figure 2.1**

edge keys.

### **Hand Tracking**

The LeapMotion hand tracker was selected as it provided an unobtrusive method for hand tracking. The original version of the software, used in this prototype, provided information on the location and orientation of each fingertip. It also gave the position and orientation of the palm, but did not provide any details on the joints or which finger of the hand the fingertip belonged to.

### **Button Recognition**

With a working hand tracking system, the next step was to develop a method to determine when users were pushing a button. The button recognition system was developed within the EDGE engine agnostic to the hand tracker, so that future prototypes could use a different hand tracker if a new device were released.

The button detection algorithm is a simple collision detection model. A rectangular box is defined that extends outside the button, including a tolerance zone to account for misalignment and poor tracking. When a fingertip enters and stays in the box for approximately 150ms<sup>1</sup> then a button event is triggered. The purpose of the delay is to account for false positives when a user might accidentally enter the box without intention to press the button. The advantage to using the optical tracking to determine when a user has selected

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<sup>1</sup>This time is configurable and was often tweaked for each experiment.

a button is that it has the potential to significantly reduce the complexity of the system. If the user interactions with the panel can be determined solely by tracking his/her hands from the external sensor, then the cockpit panel needs only to provide physical feedback, and does not require any wiring.

### Challenges

The hand tracking caused two issues with the use of the prototype. The first was simply that the reliability of the tracking caused many instances of dropped tracking, causing fingers to disappear unexpectedly. The second was the conversion between hand tracker coordinates and instrument coordinates. The registration between the virtual and physical worlds was difficult to get right with this configuration. Even with precise alignment of the hand tracker, the fingertip positions did not always align properly with the buttons in the virtual world when a subject had their finger on and felt a physical button. Since the hand tracking was not very reliable, many users would almost completely ignore its input and find the buttons by proprioception and tactile feedback. This then led to confusion as the button would not register a button press but the strong tactile feedback meant they would not leave the physical button to hunt for the virtual button location.

Another challenging portion of the registration was that since the Oculus Rift did not have head positioning, the head position would have to be set manually. If a subject were taller, shorter or sat further forward or back, then the visuals would not appear with the correct depth perspective as the panel actually was positioned at. This caused subjects to initially reach too short or too far for the panel.

#### 2.3.2 Second Prototype

A number of improvements were made in the second generation of the prototype. Due to software upgrades, the hand tracking became more robust and provided more complete information about the entire hand position. The visuals were upgraded to the second Development Kit of the Oculus Rift, which provided improved visual resolution but more importantly an external head tracker (camera) which enabled head movement and virtually eliminated drift of the internal sensors. A major focus of this version of the development

work for this prototype was to improve reliability and registration. The reliability was partially improved with the new hand tracking software, but other countermeasures developed are described as well. Registration refers to the alignment between the physical world and the virtual world.

## **Virtual Reality Headset**

The head tracking provided in the newest development kit version gives a significant improvement over the original Oculus Rift in the registration between the real world and the virtual world. The head-tracking camera is mounted on a known location on the panel (see Figure 3.1), and the tracking software provides a measurement of the relative location between the head and the camera. The virtual world can then be rendered relative to the location of the camera.

## **Hand Tracking**

The upgrade to the hand tracking software provided two features that were integrated. The first was an improvement in the fidelity of the tracking, including full information on the joints of the hand. The second was the introduction of a “head mounted” mode, which allowed for tracking to be optimized for looking down at a hand.

The tracking upgrade adds the location and orientation of the entire hand at a skeletal level. This means all the joints are tracked and the bones between them. The upgraded view is shown in Figure ???. This provided a more immersive feeling than the floating fingers of the original prototype.

The head-mounted mode was introduced to provide hand position in a virtual environment by way of mounting it on the head-mounted display itself. Upon initial testing of this feature, two observations were made. The tracking was much improved when looking down at the hands, seemingly providing more robust and reliable tracking data. However, using it mounted to the head-mounted display caused a larger disconnect between the real world and physical world. The registration relies on a known, rigid connection between the hand tracker and the location of the instruments. When the tracker was mounted on the head-mounted display, the transformation between LeapMotion coordinates and panel

coordinates relied on knowing the location of the head. Although this information could be obtained from the newer model HMD which included head tracking, it was simply not precise or stable enough to use as the base for the hand tracking coordinates. This led us to develop a mount that would hold the LeapMotion upside down over the panel and instruments. This provided the best of both worlds, with the reliability of the face-down tracking and the rigid connection to the panel and instrument location.

### **Button Recognition Countermeasures**

Two countermeasures were introduced in this prototype that aided users in learning the button recognition algorithms of the hand tracker. A simple visual indication was added when a user entered the zone in front of the button. This was typically implemented with the button itself being highlighted or changing color. With this crucial feedback, new users were able to learn the tolerances of the button zone and understand when the button was going to be ‘pressed’. For experienced users, the feedback provided confirmation that they were in the zone and were not waiting for it to register incorrectly. The second countermeasure was the addition of an aural feedback upon the button press being registered. A simple “click” noise was played when the button press event occurred, confirming the button press to the user on a quicker cognitive channel than processing if the expected change on the instrument occurred. The visual feedback of the highlighted button went away at the press event too, providing feedback in both visual and aural channels.

### **Capacitive Touch Sensors**

The capacitive touch sensors were initially developed as a countermeasure to the problems encountered with the hand tracking from the original prototype. As the hand tracking became more robust with the new software from the manufacturer, the countermeasure was not as important, but the sensors played a new role in the prototyping: validating the accuracy and providing calibration for the hand tracker.

In order to be able to activate the buttons reliably when the hand tracking was degraded or dropping out, it was decided to investigate capacitive touch sensors. To stay with the goal of minimal setup, the original capacitive touch sensors were developed with



**Figure 2.2:** Capacitive touch array button. Each row and column of diamond pads are connected as one electrode each and together can provide location information of where the user presses.

copper tape electrodes placed on the top of the 3D printed instrument buttons. These electrodes are shown in Figure ???. A Freescale MPR121 capacitive touch sensor was used to read the electrodes and communicate with an Arduino which sent touch events over a serial communication line to the computer. These serial events were read by the rendering engine to trigger events when a copper pad was touched.

Since touch accuracy can be important in safety-critical applications, we also desired the ability to record where on the button the finger press was located. This would help to determine if we had any registration biases in our system. To accomplish this, a custom printed circuit board was developed that provides an electrode array of 5 rows and 5 columns over a 1-inch by 1-inch square. This board is show in Figure 3.3, where it is mounted in a 3D printed instrument. The capacitive state of each row and column can provide a measurement of the center of the finger press on the grid created by the rows and columns. With this configuration finger-location accuracy of under 0.1-inch can be achieved. The location of the finger press can help provide a measure of the accuracy of the registration between the optical sensors and the real world location.

## Calibration

The hand tracking provided by the LeapMotion provided precise and repeatable measurement of the hand positions throughout its tracking volume. The accuracy as the hand got further away from the sensor, however, was insufficient. Put another way, the position of the hand in the virtual world was offset from the true button position in the physical world, yet the offset was consistent between movements. This led us to develop a calibration to provide a more accurate registration between the virtual and physical hand

positions. With the addition of the capacitive touch sensors, a calibration mechanism was made possible since it allowed for well known positions in the physical world to be translated to the virtual world. The location of the buttons are well known a priori from the dimensions of the physical world setup. This means that when a finger was placed on a physical button with a capacitive sensor, the known position can be recorded along with the “measured” position of the hand tracker. After collecting enough points, the calibration can correct for the measured offset and provide an accurate registration between the real and virtual worlds. The mathematical basis for this calibration is described here.

The calibration works on the assumption that there exists a transformation matrix,  $\mathbf{T}$ , that can solve the following equation:

$$\vec{x}_{known} = \mathbf{T}\vec{x}_{measured} \quad (2.1)$$

The  $\vec{x}_{known}$  and  $\vec{x}_{measured}$  correspond to the known location of a calibration point (button) and the location measured by hand tracker, respectively. Since we have that information, it is only required to solve for the transformation matrix itself.

Using a simple least squares approach to find the coefficients of the matrix, the registration between the virtual and physical worlds is vastly improved. The transformation matrix is not constrained to a simple rotation (i.e. not assumed orthogonality or other special properties) so the solution is found by expanding and solving the general homogenous coordinates transformation matrix.

$$\begin{bmatrix} x_{known} \\ y_{known} \\ z_{known} \\ 1 \end{bmatrix} = \begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} \\ T_{21} & T_{22} & T_{23} & T_{24} \\ T_{31} & T_{32} & T_{33} & T_{34} \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_{measured} \\ y_{measured} \\ z_{measured} \\ 1 \end{bmatrix} \quad (2.2)$$

Typical least squares approaches would attempt to find  $\mathbf{x}_{measured}$  in Eqn. 2.1, however we desire to find the matrix  $\mathbf{T}$  itself. It can be shown that expanding the matrix equation (Eqn. 2.2) for multiple points (i.e.  $\vec{x}_{known,1}, \dots, \vec{x}_{known,n}$  and  $\vec{x}_{measured,1}, \dots, \vec{x}_{measured,n}$ ) and then collecting like terms will convert the problem into three different least squares problems. They are shown here, dropping the subscripts to  $k$  and  $m$  for known and measured.

$$\begin{bmatrix} x_{k1} \\ x_{k2} \\ \dots \\ x_{kn} \end{bmatrix} = \mathbf{X}_M \begin{bmatrix} T_{11} \\ T_{12} \\ T_{13} \\ \dots \\ T_{14} \end{bmatrix}, \quad \begin{bmatrix} y_{k1} \\ y_{k2} \\ \dots \\ y_{kn} \end{bmatrix} = \mathbf{X}_M \begin{bmatrix} T_{21} \\ T_{22} \\ T_{23} \\ \dots \\ T_{24} \end{bmatrix}, \quad \begin{bmatrix} y_{k1} \\ y_{k2} \\ \dots \\ y_{kn} \end{bmatrix} = \mathbf{X}_M \begin{bmatrix} T_{31} \\ T_{32} \\ T_{33} \\ \dots \\ T_{34} \end{bmatrix},$$

where  $\mathbf{X}_M = \begin{bmatrix} x_{m1} & y_{m1} & z_{m1} & 1 \\ x_{m2} & y_{m2} & z_{m2} & 1 \\ \dots \\ x_{mn} & y_{mn} & z_{mn} & 1 \end{bmatrix}$

At least 4 points are needed to solve this system, and it has been found that a calibration with small least squares residuals can be achieved with 10-20 well chosen points. The calibration matrix is then used to find the offset of the tip of the index finger for each hand, and this offset is used to move the entire hand.

### Capacitive Touch Arrays

- second generation Oculus prototype
- Big upgrade due to head tracking

#### 2.3.3 Third Prototype

The third prototype was developed for the design evaluation experiment (Chapter ??). With the improvements of .

- Return to pure 3d printed instruments
- Calibration with touchscreen
- Calibration on a single plane problem

## 2.4 Lessons Learned and Future Work

- IR interference
- Small amounts of training make a big difference
- Conversely, not explaining fully how it works can lead to good performance/feedback
- hand tracking difficulties can be very frustrating

# **Chapter 3**

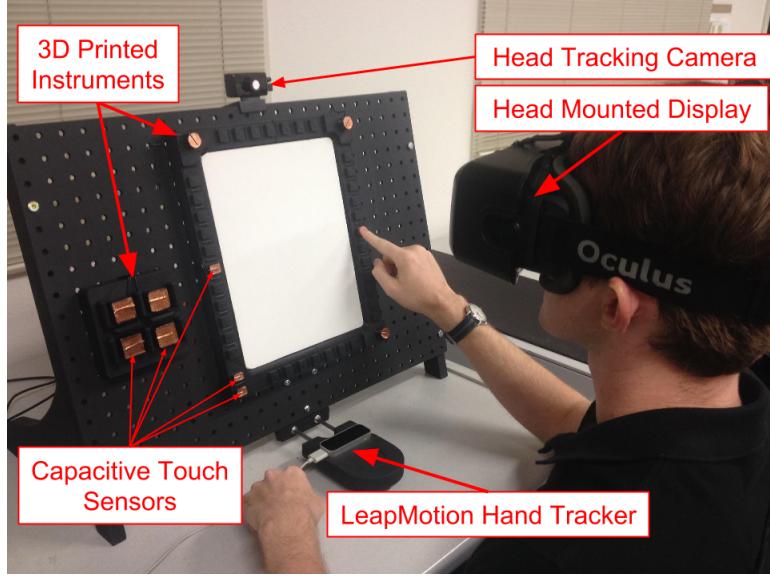
## **Pointing Experiment**

This chapter was originally published in the conference proceedings for AIAA Modeling and Simulation Technologies 2015.

### **3.1 Introduction**

The underlying motivation for this project is to bring high-fidelity simulator response to an earlier stage in the cockpit design cycle, where traditionally, the focus is on layout of displays and controls in a minimally-functional mockup. With the R3C concept, system functionality can be experienced from the start, in the virtual environment via a head-mounted display (HMD), while the user interacts with the cockpit systems by touching the surfaces of a low-fidelity (3-D printed) environment. The user receives the tactile feedback of touching a button or edge-key, and via hand-tracking technology, sees the response in the virtual displays. Required changes to the user-interface environment or even “what-if” ideas can be made quickly and at low cost by modifying the 3-D printed surfaces, and then reflecting these physical changes in the virtual visual scene. By creating a high-fidelity virtual simulator on top of a low-fidelity geometric mockup, this approach may allow more design/layout iterations at lower cost, which could lead to a final design that is closer to optimum than would be feasibly attained with traditional approaches.

We have developed a proof of concept of the R3C system, which was used to perform an initial, simple, targeting study to validate our approach, and which will be



**Figure 3.1:** Prototype R3C System

the basis for future work. Figure 3.1 shows an over-the-shoulder view of a user of the R3C prototype with annotations to call out the major components of our current technical approach. The user is wearing an immersive virtual reality head mounted display (Oculus Rift Development Kit 2), which presents them a virtual scene (shown in Figure 3.2) that is spatially stabilized with respect to the physical instrument panel. As the user reaches out toward the inert 3D printed instruments, a LeapMotion hand tracker reads the position of each finger and the pose (attitude and configuration) of the hand. A simple collision detection algorithm, described below, uses this information to determine when a user has touched a button. Since it is important for the user to visually track their own hand during pointing and actuation tasks, we also use the hand tracker to render the dynamic position of the hand in the virtual scene (Figure 3.2). On certain buttons we have added copper pads that are connected to a capacitive touch sensor as a secondary sensor for button actuation. The camera mounted above the panel is an infrared camera that senses the position of the HMD. The display on the screen can be configured as desired; it is currently showing an external camera view. The behaviors of each button are controlled in software, and can be easily reconfigured.

The use of virtual or augmented reality in aerospace has been extensive, starting with the first flight simulators, and even in the cockpit in research studies[11, 1]. Significant



**Figure 3.2:** Virtual view of user from Figure 3.1

recent work investigates virtual reality tools in a simulator[13, 16, 32]. However, our concept of merging an accurate but low-cost tactile environment with the high-fidelity virtual view is so far untested. Accurate haptic feedback has been a goal of virtual reality (VR) researchers since the emergence of VR. Providing dynamic haptic feedback, however, still proves challenging to date[31, 21]. Our approach allows accurate haptic feedback for the case of a static workstation, such as found in a real cockpit. Combined with 3D printing and our virtual simulator overlay, this provides an inexpensive platform to create a functioning simulator, able to adapt quickly to large-scale design changes.

## 3.2 Technical Approach

This section details the components of our prototype and how they are integrated into the R3C system, as well as rational for selection, improvements and lessons learned.

### 3.2.1 3D Printed Instruments

A central tenant of our technical approach is to use physical, geometrically accurate instrument shapes that provide no functionality (i.e. no screens, no working buttons, etc.). This is intended to imitate the fidelity of a typical early design stage mockup. In order to achieve this we have produced 3D-printed “instruments” that can be easily rearranged on a panel mount (pegboard). Since the devices are rapidly prototyped, they can be redesigned in

a much smaller time frame than typical simulator instruments. By using the geometrically accurate instruments at the correct cockpit locations, the user is provided with accurate tactile and proprioceptive feedback without the need for entering the challenging field of virtual haptic feedback.

### 3.2.2 EDGE Rendering Engine

We are using a NASA developed rendering engine named EDGE to provide the visuals for the virtual scene rendered in the head-mounted display (HMD). EDGE is highly customizable and extendable through C/C++ plugins, Tcl scripts, or networking functions. Existing dynamic simulations (e.g. Matlab) can also be integrated.

### 3.2.3 Oculus Rift

The Oculus Rift virtual reality head mount display is currently only available as a developer kit. The lightweight headset provides an immersive virtual reality experience by combining a wide field-of-view scene with accurate head tracking, giving a stable virtual world. The small display (cell phone sized) is viewed through a single set of lenses, and the barrel distortion of the lens is corrected for in the rendering engine by a pincushion distortion of the rendered scene. The orientation of the head is tracked using internal sensors as well as an infrared camera to give an accurate pose for rendering the scene. The software developers kit (SDK) exposes the head position and orientation and provides the appropriate distortion for a rendered scene (which is done for each eye), and this has been developed into an EDGE plugin. The technology of these head mounted displays is improving rapidly, driven by technology developed for the cell phone and demand from the gaming industry.

The head tracking provided in the newest development kit version gives a significant improvement over the original Oculus Rift in the registration between the real world and the virtual world. The head-tracking camera is mounted on a known location on the panel (see Figure 3.1), and the tracking software provides a measurement of the relative location between the head and the camera. The virtual world can then be rendered relative to the location of the camera.

### 3.2.4 LeapMotion Hand Tracker

The LeapMotion hand tracker is a small, consumer orientated optical hand tracker. It uses two infrared cameras as the source of its proprietary tracking algorithm. As of version 2.0 of their SDK, they expose to the application developer the positions of all the joints and bone positions of the hand. An EDGE plugin was developed in-house at UC Davis to position the nodes in the scene from this information, which is the source for rendering the virtual hand as well as the button detection algorithm.

We decided to use the LeapMotion as it provides logistical simplicity compared to professional motion capture devices that require multi-camera setups with lengthy calibrations as well as high cost. A main goal of our system is that it can be used without much setup in the physical world, leading us to the new small optical trackers. There are other promising devices (i.e. Kinect2 and other 3D camera systems) that could be used in the future.

### LeapMotion Calibration

Early in the integration process, we discovered that the LeapMotion provided precise and repeatable measurement of the hand positions throughout its tracking volume. The accuracy as the hand got further away from the sensor, however, was insufficient. Put another way, the position of the hand in the virtual world was offset from the true button position in the physical world, yet the offset was consistent between movements. This led us to develop a calibration to provide a more accurate registration between the virtual and physical hand positions. Since we accurately know the position of the panel and instruments, it is only required to solve the following linear algebra equation for the transformation matrix,  $\mathbf{T}$ :

$$\vec{x}_{known} = \mathbf{T}\vec{x}_{measured} \quad (3.1)$$

Using a simple least squares approach to find the coefficients of the matrix, the registration between the virtual and physical worlds is vastly improved. The transformation matrix is not constrained to a simple rotation (i.e. not assumed orthogonality or other special properties) so the solution is found by expanding and solving the general homogenous

coordinates transformation matrix.

$$\begin{bmatrix} x_{known} \\ y_{known} \\ z_{known} \\ 1 \end{bmatrix} = \begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} \\ T_{21} & T_{22} & T_{23} & T_{24} \\ T_{31} & T_{32} & T_{33} & T_{34} \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_{measured} \\ y_{measured} \\ z_{measured} \\ 1 \end{bmatrix} \quad (3.2)$$

Typical least squares approaches would attempt to find  $\mathbf{x}_{measured}$  in Eqn. 3.1, however we desire to find the matrix  $T$  itself. It can be shown that expanding the matrix equation (Eqn. 3.2) for multiple points (i.e.  $\vec{x}_{known,1}, \dots, \vec{x}_{known,n}$  and  $\vec{x}_{measured,1}, \dots, \vec{x}_{measured,n}$ ) and then collecting like terms will convert the problem into three different least squares problems. They are shown here, dropping the subscripts to  $k$  and  $m$  for known and measured.

$$\begin{bmatrix} x_{k1} \\ x_{k2} \\ \dots \\ x_{kn} \end{bmatrix} = \mathbf{X}_M \begin{bmatrix} T_{11} \\ T_{12} \\ T_{13} \\ T_{14} \end{bmatrix}, \quad \begin{bmatrix} y_{k1} \\ y_{k2} \\ \dots \\ y_{kn} \end{bmatrix} = \mathbf{X}_M \begin{bmatrix} T_{21} \\ T_{22} \\ T_{23} \\ T_{24} \end{bmatrix}, \quad \begin{bmatrix} z_{k1} \\ z_{k2} \\ \dots \\ z_{kn} \end{bmatrix} = \mathbf{X}_M \begin{bmatrix} T_{31} \\ T_{32} \\ T_{33} \\ T_{34} \end{bmatrix},$$

where  $\mathbf{X}_M = \begin{bmatrix} x_{m1} & y_{m1} & z_{m1} & 1 \\ x_{m2} & y_{m2} & z_{m2} & 1 \\ \dots \\ x_{mn} & y_{mn} & z_{mn} & 1 \end{bmatrix}$

At least 4 points are needed to solve this system, and it has been found that a calibration with small least squares residuals can be achieved with 10-20 well chosen points. The panel and instrument edges can provide well-known points, as well as button locations. The calibration matrix is then used to find the offset of the tip of the index finger for each hand, and this offset is used to move the entire hand.

## LeapMotion Mounting Position

The LeapMotion software recently (as of version 2.2) began officially supporting a “VR Mode” which allowed users to mount the device on the front of a VR HMD and track the hands relative to the head. This was tested in our system and although we experienced better tracking, the hand would not stay stable relative to the panel during head movement. This led us to develop a mount that would hold the LeapMotion above the panel looking down, and putting the software in “VR Mode” we were able to achieve better tracking but with a fixed known position, the hand remained stable relative to the panel.

### 3.2.5 Button Detection Algorithm

The button detection algorithm is a simple collision detection model. A rectangular box is defined that extends outside the button, including a tolerance zone to account for misalignment and poor tracking. When a fingertip enters and stays in the box for 50ms then a button event is triggered. An event is also triggered when a finger enters the box, which we use to change the color of the button to indicate proper alignment to the user. The advantage to using the optical tracking to determine when a user has selected a button is that it has the potential to significantly reduce the complexity of the system. If the pilot interactions with the panel can be determined solely by tracking his/her hands from the external sensor, then the cockpit panel needs only to provide physical feedback, and does not require any wiring. Of course, the primary drawback of this approach is that the efficacy is limited to the accuracy and reliability of the hand tracker, and the algorithms tracking this movement for button inputs. Nonetheless, as we show in our experiment, the LeapMotion coupled with this algorithm can be used effectively.

For the experiment described below, the collision detection box was set to extend 0.5 inches inward and outward from the button, including a 0.1-inch tolerance on all dimensions. Aslandere [32] found no significant effect with button selection ability based on changing this volume in a purely virtual environment, so we did not change the collision detection box size for this first test of the R3C (although it is easily modified in software).



**Figure 3.3:** Capacitive touch array button. Each row and column of diamond pads are connected as one electrode each and together can provide location information of where the user presses.

### 3.2.6 Capacitive Touch Sensors

For this early research phase, we needed a baseline method to determine if a button was touched, for comparison with the LeapMotion IR camera hand tracking method. A simple method was first developed using copper tape electrodes on top of certain buttons (as can be seen in Figure 3.1). These electrodes are connected to a Freescale MPR121 capacitive touch controller that registers when a finger touches them. The capacitive touch controller is connected to an Arduino, acting as a serial communicator between the microchip and the computer. This setup provides a binary state of whether a finger is touching the button. This is used to help determine whether the button detection algorithm using the hand tracker is working correctly.

Since touch accuracy can be important in safety-critical applications, we also desired the ability to record where on the button the finger press was located. This would help to determine if we had any registration biases in our system. To accomplish this, a custom printed circuit board was developed that provides an electrode array of 5 rows and 5 columns over a 1-inch by 1-inch square. This board is shown in Figure 3.3, where it is mounted in a 3D printed instrument. The capacitive state of each row and column can provide a measurement of the center of the finger press on the grid created by the rows and columns. With this configuration finger-location accuracy of under 0.1-inch can be achieved. The location of the finger press can help provide a measure of the accuracy of the registration between the optical sensors and the real world location.

### 3.2.7 Summary and Lessons Learned

We've discovered a few qualitative effects of various technologies and methods throughout the time we've spent building this prototype, resulting in the following "lessons learned".

The registration between the virtual world and real world was a concern in the first prototype that did not include the head tracking setup of the Oculus Rift DK2, and it required precise measurement between the location of the user (including seat height) and the desk and panel. This would cause problems when the head orientation yaw sensors drifted as well, where the user would not be lined up with the panel after a few minutes of use. The upgrade to the head-tracking camera to anchor the virtual world to the real world greatly improved both stability and accuracy, and also significantly improved the feeling of presence in the virtual environment.

The registration between the real world and the hand tracker position in the virtual world is even more important, however. Since the user can feel the proper place for their finger when the finger arrives at the target, if the hand tracker is not lined up they may not be able to activate the button detection algorithm easily, or at all if there is physical interference. This problem created the need for the calibration scheme, and having the calibration is essential for using the LeapMotion in a task where the real and virtual world needs to be aligned. This problem also highlights the importance of testing people who are new to the system. New users were fixated on placing their finger where the physical feedback indicated the button was and did not adjust for misalignments with the hand tracker, while expert users learned to ignore the physical feedback and activated the hand tracker by finding the misaligned location. Additionally, we discovered that hand pose and speed can influence the performance of the hand tracker, such that expert users can perform in the system quicker and with greater accuracy. This is discussed again later in the context of the experiment performed.

The purely optical button detection algorithm is prone to failures, especially without a complex algorithm performing heuristics on the movement to determine the user's desired selection. The simple collision detection model works well, but is difficult for the

user without the loop-closing visual indication of when the detection box is activated. In our case we added the button changing color. It was also important to have a slight delay in activation, as this not only provides a failsafe for false positives, but it also adds a way to explore the tactile surface without consequences.

The latency of the entire R3C system has not yet been quantified, and there are two significant aspects where this might be interesting: from hand movement to hand tracker measurements through the rendering pipeline to the image shown to the user; as well as from head movement through to the image shown to the user. However, latency is clearly low enough to make measuring it low priority, as the hand movement delay in the virtual world is unnoticeable to the average user. Additionally, the latency of the Oculus Rift head mounted display is drastically improved from earlier technologies such that it does not cause a concern. Finally, the manufacturers of both LeapMotion and Oculus are actively improving on the latency on both fronts.

We also initially found tracking performance degradation when the finger approaches the instrument panel that was due to optical interference with the panel objects (i.e. it was happening only when the panel was present). This was improved when we moved the LeapMotion to the top looking down, but the larger effect was found when looking at the image captured by the LeapMotion infrared cameras, discovering our black 3D printed instruments were highly reflective in infrared and showing up the same color as the hand. Applying a matte finish greatly improved this, but it is still a problem we are dealing with to date. We have also found better results when controlling entire backdrop of the LeapMotion field of view with a dark matte material, as this helps provide a greater contrast between the hand and the background. There still remains work to improve this, as a few subjects in the experiment indicated they found degraded tracking with the panel present.

### 3.3 Experimental Methods

#### 3.3.1 Experiment Goals and Motivation

We performed a brief pilot study to validate our technical approach and take some measurements of the effects of using the R3C system. Specifically, we were interested

whether a new user could accurately target the correct button in an instrument panel, and how the different technologies used affected the targeting task. There were three main effects of the targeting task under study in this experiment:

- The effect of the use of the VR HMD versus no VR HMD (i.e. virtual vs. real world).
- The effect of the physical panel versus no physical panel present (i.e. having the tactile feedback vs. no tactile feedback)
- The effect of the optical tracking button detection versus the capacitive touch button detection

The use of the virtual reality headset also implies that subjects have to rely on the visual feedback from the hand tracker virtual hand to target the button.

We were concerned about the effect of both touch-selection accuracy and movement time, but the context for our design is an aerospace cockpit, where accuracy in selecting the intended button is typically paramount over movement time. Previous research in targeting tasks in virtual environments has not always reported on success rates, indicating incorrect trials were re-performed. We did not do this, as we wanted to ensure we recorded the success rate.

Previous work has investigated various 3D pointing tasks in the real and virtual world without tactile feedback[20], or with tactile feedback but no immersive virtual reality[34], or with virtual haptic feedback[8], or pure virtual worlds[4, 12]. Our work differs in the use of the tactile feedback provided by the panel combined with an immersive virtual world.

### 3.3.2 Experiment Design

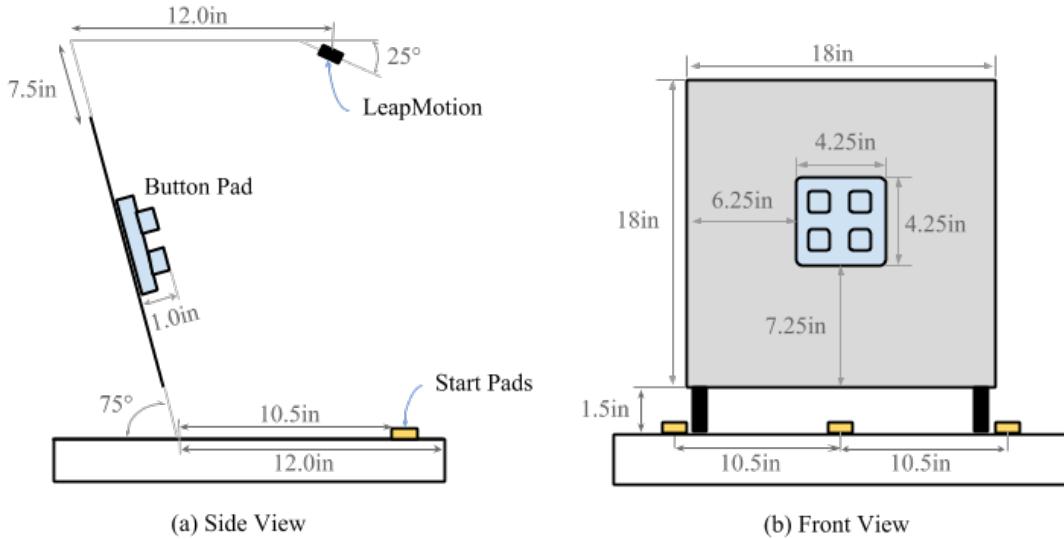
Figure 3.4 shows a diagram of the experimental setup. The panel was configured with a single four-button keypad, arranged in a 2x2 grid. Three different starting pads were placed on the desk near the edge. The buttons on the keypad were  $1'' \times 1''$  and were equipped with the capacitive touch sensor arrays as described above, allowing the measurement of position of finger press. The start pads were capacitive touch copper tape electrodes, also

$1'' \times 1''$ , but had no position detection capability. The participants were asked to start with their index finger on the proper start pad and then target the correct button on the panel. Audio prompts were given through headphones, which indicated the starting pad and panel button goal before each task. Subjects were instructed to target the center of the correct button, but were given no instructions on speed. Subjects targeted the button using their index finger of their dominant hand, which was the right hand for all subjects. Each participant performed 48 targeting tasks under each of the four different conditions:

1. Wearing the HMD, button selection registered by capacitive sensors, panel physically present
2. Wearing the HMD, button selection registered by hand tracking sensors, panel physically present
3. Wearing the HMD, button selection registered by hand tracking sensors, panel not physically present
4. Not wearing the HMD, button selection registered by capacitive sensors, panel in position

Before each condition the subjects were given one set of 12 trials for familiarization of that particular condition. During the “no panel” condition, the hand tracker remained mounted at the same point, but the panel was moved out of the way, such that the subjects were targeting the buttons in the virtual world only. The button selection in every condition was indicated by a sound played in the headphones when the appropriate sensor measured a button press. In addition, during the optical sensor trials, the button color changed color to indicate they had entered the selection zone (as described in Section 3.2.5). The tasks would time out after 10 seconds. The task was not repeated for an incorrect button selection or timing out, instead the subject moved on to the next task. The condition order was randomized for each subject, and the tasks were performed in a pseudo-random order.

The subjects were all briefed on the technology from a prewritten speech explaining how to activate each of the sensors. No information was given on how to best perform the task or advice on working with the various sensors.



**Figure 3.4:** Diagram of Experiment Setup

The dependent variables measured were time from starting pad to trial goal button, accuracy measured on the capacitive touch sensors (except for condition 3 where this was not possible), as well as success rate. The raw data from the LeapMotion hand tracker was also recorded for each trial, allowing for post analysis of hand trajectory.

### 3.3.3 Subject Pool

The experiment was performed with 8 subjects, recruited from the university student population, all with no or limited previous exposure to the R3C setup. All subjects indicated they had either never used a virtual reality headset or briefly tried them once or twice. Ages ranged from 18-31 with 6 male, 2 female. They all had correctable vision to 20/20 and none indicated difficulty seeing the image in the virtual reality HMD. After the experiment, every subject indicated no motion sickness, and only mild eyestrain was reported.

## 3.4 Results

The time was recorded for each trial from release of starting pad to the successful button press using the correct button registration. Trials which timed out or where the

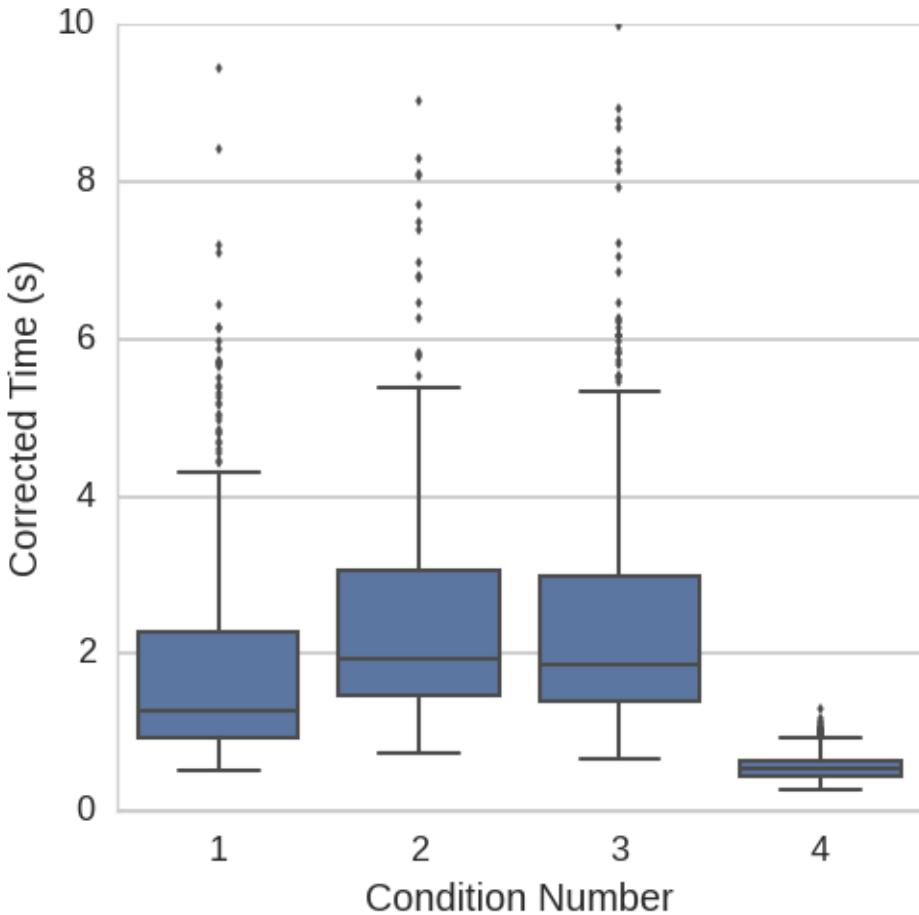
Condition	Mean Corrected Time (s)	Average Speed (in/s)	Success Rate
1. HMD, Capacitive, Panel	1.88 ( $\sigma = 1.47$ )	14.51 ( $\sigma = 7.32$ )	98.6%
2. HMD, Optical, Panel	2.43 ( $\sigma = 1.42$ )	9.80 ( $\sigma = 4.61$ )	97.4%
3. HMD, Optical, No Panel	2.49 ( $\sigma = 1.68$ )	10.21 ( $\sigma = 5.17$ )	96.8%
4. No HMD, Capacitive, Panel	0.59 ( $\sigma = 0.18$ )	34.04 ( $\sigma = 9.16$ )	99.4%

**Table 3.1:** Mean results across subjects. Standard deviations are reported as  $\sigma$ .

incorrect button was targeted were discarded for the time analysis. The time for each trial was corrected for the varying distances of the movements by multiplying by the average distance (18.4 in) over the straight-line distance between the start and goal button for that trial (which varied from 15.5in to 20.5in). A boxplot of the distribution of the corrected time grouped by condition across subjects is shown in Figure 3.5. The effects of each were determined using a two sample T-test on the corrected time measure between the appropriate conditions. The effect of the use of the HMD between condition 1 and 4 (both using capacitive and panel) was significant ( $p < 0.01$ ), and on average caused the corrected time to drop 1.31 seconds. The optical detection algorithm also has a negative effect on corrected time ( $p < 0.01$ ), comparing conditions 1 and 2, but the drop in performance was only 0.55 seconds. However, it was found that the panel had no significant effect on performance across all subjects (conditions 2 and 3).

When comparing the results of the corrected time within subjects it was found that 3 subjects had a significant positive effect with the panel while only 1 subject had a significant negative effect, the rest showing no effect. Similarly, the optical tracking versus capacitive tracking had no effect on 3 subjects, while the rest display the same as the between subjects finding of negative effect.

The selection of the correct button was performed almost without error, which is a promising result for new users of the system. There was no significant effect between the different conditions on success rate. Over 8 subjects and 1517 trials, only 12 trials selected the wrong button and only 12 trials timed out, giving an overall 98.4% observations made

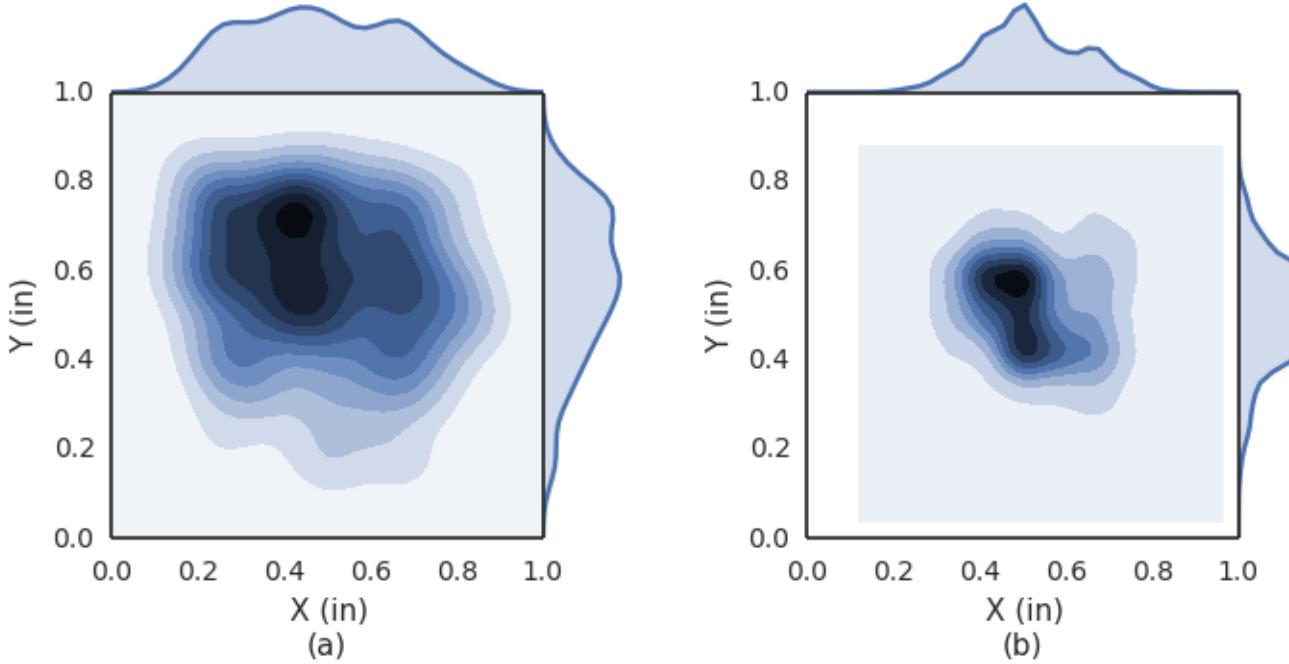


**Figure 3.5:** Boxplot of normalized time distribution by condition. Condition numbers are:

1. HMD, Capacitive, Panel;
  2. HMD, Optical, Panel;
  3. HMD, Optical, No Panel;
  4. No HMD, Capacitive, Panel.
- The median is indicated by the line in the box, the boxes contain the inter-quartile range (IQR), and the whiskers extend to 1.5\*IQR with outliers plotted.

during the experiment indicate that at least some of the wrong button selections were due to misheard prompts or loss of attention.

The accuracy of the button press itself was also measured, and Figure 3.6 shows the distribution of the locations registered by the capacitive touch sensors on the first press of the trial (over the first 100ms). In the plots,  $(X,Y) = (0,0)$  corresponds to the bottom left corner of the button. The two conditions shown (1 and 4) were the two trials where capacitive touch registered the button selection. There is a smaller distribution for the no HMD condition (Figure 3.6(b)), but with the HMD (Figure 3.6(a)) subjects were still



**Figure 3.6:** Finger press location 2D distribution for (a) HMD, Capacitive, Panel and (b) No HMD, Capacitive, Panel. The distribution is calculated as a Gaussian kernel density estimate. The mean ( $x, y$ ) location and standard deviation for each is (a)  $(0.48 \pm 0.18, 0.56 \pm 0.17)$  and (b)  $(0.52 \pm 0.12, 0.51 \pm 0.12)$ .

within error to the center of the button. It should be noted again here that in the HMD condition, subjects had to rely on the virtual hand from the hand tracker for any visual feedback of hand position.

A Fitts law analysis of the data was not the goal with the experiment design, and since we did not vary the button width a Fitts analysis could only be performed with varying distance. The index of difficulty would only range from 4.0 to 4.5, providing an insufficient range for a proper Fitts analysis. Including extra parameters for the 3D nature of the task [24, 7] did not improve the fit, especially considering the symmetry of the setup. Furthermore the time recorded of the tasks were typically more than in a Fitts' task, since many subjects spent time honing based on the accuracy instructions.

### 3.5 Discussion

The start pads and hand tracker were located relative to each other in such a way that the hand would not be in view of the tracker at the beginning of the task, thus the user would have to wait for the hand tracker to acquire the hand when in view. This meant for the trials using the HMD the subject would typically have an open loop (blind) ballistic phase followed by the closed loop homing phase once the virtual hand appeared. Across subjects for all HMD conditions this hand acquisition by the tracker took on average 660ms ( $\sigma = 649\text{ms}$ ) from leaving the start pad. An expert user who performed the experiment was able to get the hand to appear in the tracker in 330ms ( $\sigma = 133\text{ms}$ ). This could indicate that training on how to get the best hand tracker performance could improve results. The same expert also only experienced a 416ms slowdown between using the HMD and not using the HMD (conditions 1 and 4), compared to the 1.31s slowdown measured across the 8 new subjects.

We had originally hypothesized that having the panel would improve the selection performance of the button, but for this task it had no effect. This may turn out to be due to the instructions given, which called attention to accuracy. It is also possible that while the panel did not improve targeting performance in this task, it may provide a benefit to the subjective feeling of presence in the virtual world, which may be beneficial in a more demanding situation with multiple tasks.

Another variable that was not measured was fatigue. Having the panel as a backstop likely decreases the amount of fatigue for a repeated targeting task, as floating the arm in free space to target a button in a purely virtual environment with no panel can cause a buildup of arm fatigue. If the virtual world and physical world are correctly aligned, then the finger can rest on the button while the hand tracker works to detect a button press.

Overall, the lack of an effect of having the panel there could also be interpreted as a positive result. As discussed before, a significant problem we've been working on is the improvement of tracking performance when the hand nears the panel. The results indicate that this problem has been mitigated at least as much to provide performance on par with the purely virtual world.

### **3.6 Future Work**

The promising results from the targeting pilot study indicate that users had no trouble accurately selecting the 1 inch square buttons. This would be a fairly large button area for an aerospace cockpit. A future study would investigate the limits of this system in terms of button size, and if smaller buttons pose any significant effect in accuracy. By varying the button size, this would provide a better experimental design to perform a Fitts Law analysis of the R3C system.

As previously discussed, an expert user can handle the inaccuracies of the system better than a novice, thus implying that the pilot study results did not reach an asymptote in training effects, or that there are long term training effects. The pilot study did not provide any training to subjects so as to not bias a certain condition, but future experiments may include a longer training session before subjects perform tasks in the R3C setup.

### **3.7 Conclusion**

Although this initial application of the R3C system is to aviation/space vehicle cockpits, any sufficiently complex human-system interface could be designed with this system, such as telerobotics, air traffic control, robotic surgery, etc. The technology that supports our proof of concept system is rapidly improving, and further iterations of our technology integration will provide higher fidelity and an easier user experience.

We have showed that new users can accurately select buttons in a simple targeting task in our system. All of our measured effects showed no significance on correct button selection. The use of the head mounted display to provide visuals of hand position provides a small time penalty in button selection. The use of the physical panel provided no significant effect in time compared to having the subjects target purely virtual buttons. Our optical tracking algorithm had a slight negative effect in time compared to using the capacitive touch sensors. We are hopeful that training and improvements in the technology will reduce the performance gap.

# Chapter 4

## Passive Haptics Experiment

### 4.1 Introduction

Passive haptics is a term that has been used to describe a variety of technologies or techniques to provide the sense of touch to a user of a virtual environment. It is often defined by its distinction from active haptics, which simulate the sense of touch with energy exchange, typically electromechanical. A common active haptic technology used in immersive virtual environments is a haptic glove, which utilizes small motors at the fingertips. In contrast, passive haptics often utilize proxy objects placed in the physical world to co-incide with the virtual environment experience. The proxy objects can be simple or complex. They can be colocated and accurate with the virtual world or purposefully designed to trick the user. In our paper, we utilize a simple colocated passive haptic device and measure its effect on the presence and performance of subjects using a 2D panel in a 3D immersive virtual environment.

The advantages to using a simple passive haptic can be easily understood: less cost and complexity compared to most active haptic solutions. However, the disadvantage comes with its inflexibility. Due to their nature, passive haptics often have to be purpose built for a single or limited experience. While past research has aimed to address this, by either actively positioning a proxy object or simplifying the proxy object to fool the user, our application does not suffer from this limitation. The motivation for our research comes from the application of designing aerospace cockpits, complex human-machine interfaces

where the user is stationed at their workspace. For the purpose of evaluating a cockpit design, the user does not need a dynamic tactile environment. Furthermore, many cockpit design processes already create a physical mockup which can provide the passive haptics for this evaluation.

We present our findings in testing passive haptics versus no haptics in an immersive virtual reality environment. Using a head-mounted display and a hand tracker, the subjects performed the same Fitts' Law style task under these two haptic conditions. The passive haptics was a flat surface placed at an angle on a desk in front of their seating area. Their performance on the Fitts' task was recorded as well as their responses to a presence survey, a self reported arm fatigue score and a general questionnaire.

## 4.2 Background

### 4.2.1 Haptics

Passive haptics has been a topic of research since the early immersive virtual environments. Robotic passive haptics were used to ameliorate the inflexibility of a proxy object by utilizing a robotic arm to position the proxy object in the virtual environment where the user was reaching[33, 23]. Insko[17] found increased presence using passive haptics for a maze, and also found that subjects trained with the passive haptics performed better after they were removed than the group that never used them. Another track of work combined active haptics with passive haptics, using a haptic glove with a physical panel to create mixed haptics [3]. Again, performance was increased with the haptics, but minimal differences were found between using the mixed haptics and the passive haptics alone. Similar to our motivation and work, Schiefele et al.[27] replaced a cockpit panel with a flat panel in an immersive head-mounted virtual environment, and found that users could activate buttons and switches in less time with the panel present than without. While much of the research involving passive haptics indicates an increase in the presence of the user, some have questioned whether active haptics provides benefits. Pontonnier et al.[26] discovered that subjects had decreased presence ratings in a virtual assembly task when using a haptic glove, versus both a real environment and a virtual environment without

haptics. We build on this previous work by investigating the effects of passive haptics with the latest virtual environment technology, as well as performing a complete Fitts' Law characterization between no haptics and passive haptics.

#### 4.2.2 Fitts' Law

Fitts' originally devised a relationship between movement time and the distance and size of targets for a human performing rapid aimed movements[10]. This has since become known as Fitts' Law, and later work has refined the index of difficulty (ID) as:

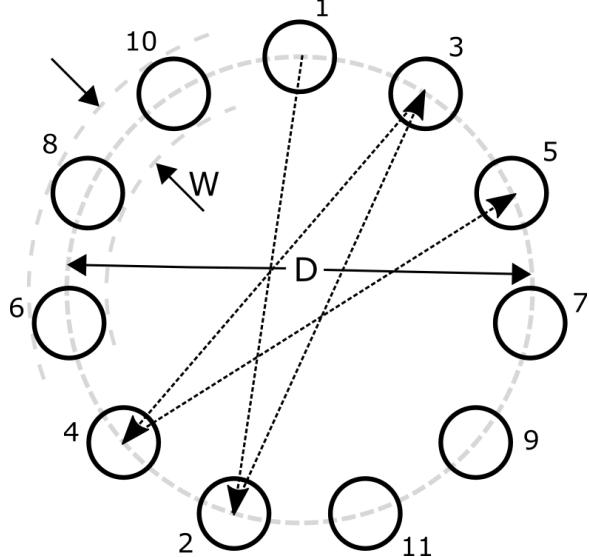
$$ID = \log_2 \left( \frac{D}{W} + 1 \right) \quad (4.1)$$

where  $D$  is the distance to the target from the starting location and  $W$  is the width of the target. This formula for index of difficulty is known as the Shannons' formulation[22].

Commonly, the index of difficulty is related to movement time (MT) through a linear regression. However, in this work we are concerned with the measurement known as throughput (TP). Throughput has been recommended as the dependent measure for comparisons between experimental conditions[30]. It can be thought of theoretically as the rate of information the human can input with the particular experimental setup or input device. It is defined mathematically as the index of difficulty over the movement time, and has the units of “bits per second” (bps).

$$TP = \frac{ID}{MT} \quad (4.2)$$

The use of Fitts' Law as a tool for human-computer interface research began with the research of Card et al.[6] for the evaluation of different input devices for text entry. The ISO 9241-9 standard was published in 2000 with guidance on using Fitts' Law as an evaluation of pointing devices[18]. Along with the ISO standard, there have been calls to standardize the use of Fitts' Law so that results can be compared across literature[30]. For a 2-dimensional task (where all the buttons exist on a single plane), it is recommended to use the circle layout as shown in Figure 4.1. This layout is referred to as the “Fitts' circle” within this article.



**Figure 4.1:** Fitts' Circle diagram.  $D$  is the distance between targets and  $W$  is the width of the targets.

In this work, we use the effective width for the targets, which is defined as:

$$W_e = 4.133\sigma \quad (4.3)$$

where  $\sigma$  is the standard deviation of the end point positions. This is known as the adjustment for accuracy[35]. This correction accounts for the performance of the subject, especially on lower index of difficulty conditions where they may aim for the inside edge of a target. Hence, the use of the effective width provides the index of difficulty for the task that the subject performed, not the task presented to them. The effective width is calculated per subject per distance and width configuration, and subsequently used in the index of difficulty equation.

$$ID_e = \log_2 \left( \frac{D}{W_e} + 1 \right) \quad (4.4)$$

Previous work of evaluating virtual environments and input devices with Fitts Law include evaluating 3D tasks [8, 20, 34]. Kohli et al. [19] used a passive haptic environment with a head-mounted display, but evaluation focused on their warped virtual space technique. Recently, some reports have evaluated the Fitts performance of the LeapMotion hand tracker [9, 28], but none have combined hand-tracking with the use of a head-mounted display.

#### 4.2.3 Presence

The feeling of presence is often specified as a goal of a virtual environment. A definition from Witmer and Singer[36] reads:

“*Presence* is defined as the subjective experience of being in one place or environment, even when one is physically situated in another.”

Sheridan [29] argued that presence is a subjective sensation that necessitates a subjective measure. Increased presence has been shown to lead to increased performance in a virtual environment task[38]. We use a version of the Presence Questionnaire (PQ) proposed by Witmer & Singer[36]. The post-questionnaire consists of 16 questions with a seven-point Likert scale. Nystad & Sebok [25] found the PQ to be sensitive to technology and interaction methods, so it was chosen to investigate the differences between the two conditions.

#### 4.2.4 Arm Fatigue

Despite the concern of arm fatigue in virtual environments[5], it was surprising that most results in literature were anecdotal or for mitigations without quantification of the fatigue. Since fatigue is a subjective quantity, it can be hard to measure it between subjects, and sometimes even within. The negative impact of arm fatigue on using virtual environments makes it worth investigating. The arm fatigue scale used within this experiment is a Borg Rating of Perceived Exertion (RPE) scale that ranges from 6-20[2]. Hincapie-Ramos et al.[15] proposed a model for quantifying and predicting the amount of arm fatigue that correlated well with a Borg scale. Although there does not exist a standard for arm fatigue measurements in virtual environments, we hope that future researchers will consider including arm fatigue ratings in their experiments.

### 4.3 Methods

The purpose of the experiment described within this paper is to answer the following research questions:

1. Will throughput be higher with passive haptics?

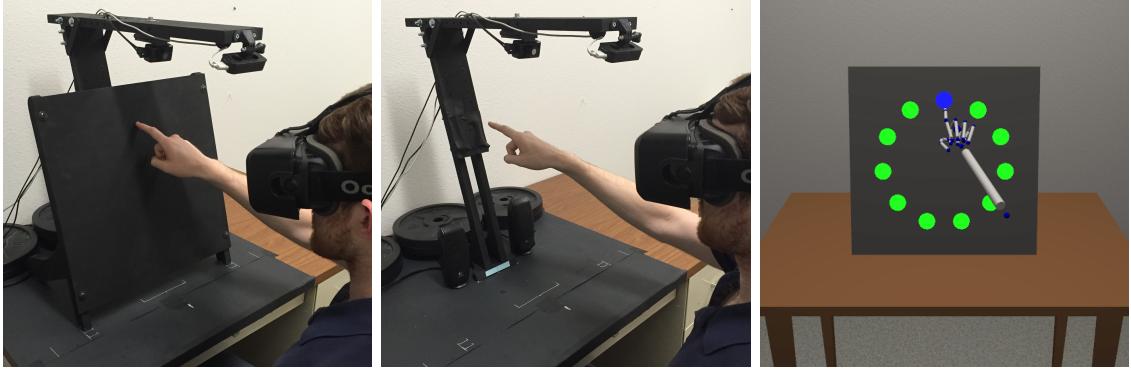
2. Do subjects learn the task quicker with passive haptics?
3. What are the differences between the formation of reaching motion trajectories with passive haptics?
4. Do subjects have lower arm fatigue with passive haptics?
5. Do subjects feel more presence with passive haptics?

#### 4.3.1 Experimental Setup

For the experimental setup, subjects were seated at a desk with a blank panel mounted on an angle in front of them. The plywood panel (45cm x 45cm) was used to provide only the backstop of the virtual buttons for the “Passive Haptics” condition. The button selection is registered by the subject moving their index finger into a hover zone (cylinder for the circle buttons) in front of the button that extends outward 0.5in. Their entrance into the hover zone is indicated to them by the button changing color. A successful button press is registered after 150ms, and is indicated by the color turning off and a button click noise being played over speakers.

The equipment used consists of an Oculus Rift DK2 (Development Kit 2) head-mounted display (HMD) and a LeapMotion hand tracker. The low-persistence OLED display of the HMD has a resolution of 1920x1080, with a refresh rate of 75Hz. The field of view is approximately 100°. It utilizes internal sensors (accelerometers and gyros) and an external infrared camera for head tracking.

The LeapMotion is a markerless hand tracker which utilizes dual infrared cameras to calculate the position of hands and fingers in view. This position is used to provide an image of the hand position in the virtual environment, as well as for determining when a button is pressed. Instead of using the LeapMotion in its original face-up configuration, it was mounted above the working area and pointed down. Our pilot studies indicated that hand tracking from the LeapMotion was improved utilizing this face-down setup with the software using the head mounted configuration. However, the hand tracker could not be mounted on the head mounted display as it required a fixed position relative to the passive



(a) Passive Haptics condition (b) No Haptics condition (NH) (c) View of virtual world (same (PH) for both)

**Figure 4.2:** Experimental conditions and view of virtual environment.

haptics to maintain appropriate registration between the virtual world and the passive haptics.

A custom calibration scheme was developed for the hand tracker as the initial registration between physical and virtual worlds was not very accurate. Despite the inaccuracy, the LeapMotion software was very precise, so after performing the calibration the registration was kept stable. The calibration performed a least squares calculation to solve for a transformation matrix between known real world locations and the reported location from the LeapMotion.

#### 4.3.2 Experimental Task

The experimental task was a Fitts' circle in the virtual environment, performed by subjects in two haptic conditions. The subjects were seated at a desk for the experimental task, and the circle was located on a panel mounted on the desk. The two haptic conditions were “No Haptics (NH)” and “Passive Haptics (PH).” These conditions are pictured in Figure 4.2. For the “Passive Haptics” condition a physical panel was co-located with the panel in the virtual world, which was removed for the “No Haptics” condition.

There were no differences to the task itself or the method of button activation. The only difference between the conditions was the removal of the physical panel. The hand tracker remained in the same location, preserving the location of the buttons in the virtual

environment. The dimensions of the virtual world were no different for either condition. In fact, there was no change to the software between the conditions.

Subjects performed the Fitts' circle for three different distances (20cm, 30cm, 40cm) and five different button widths (5mm, 10mm, 15mm, 20mm, and 25mm). These configurations were chosen to span a wide range of indices of difficulty (3.2—6.4). For each configuration of distance and width, subjects had to complete the full pattern of 11 buttons three times consecutively. This set of 30 movements<sup>1</sup> for a single configuration is referred to as a single trial for a subject. The distance was kept constant for the consecutive trial until all five button widths were complete. The distances were presented in either smallest to largest or vice versa, which was counterbalanced among subjects.

This set of 15 trials was repeated for each haptics condition and the order was kept the same within subjects. The sequence that the two conditions were presented to each subject was also counterbalanced. The dependent measures will be tested for interaction with sequence to determine if the results were affected by condition order.

#### 4.3.3 Experimental Design

As described in the previous section, the experiment was performed with a within-subjects design. Subjects were asked to complete the same experimental task for both conditions of haptics: Passive Haptics (PH) and No Haptics (NH). The two different haptic conditions are the main independent variables.

It was expected we would find a large amount of skill transfer between the two conditions, so the order in which subjects performed the two conditions was counterbalanced. This created a second independent variable that is between subjects. The subjects who performed the conditions with the PH being their first condition were one group, and the subjects who performed NH as their first condition were a second group. We call this grouping “sequence” and refer to the two groups as “PH First” and “NH First”.

For a Fitts' Law evaluation, it is often recommended that the data collection only begins when the subject is fully trained on the task. However, one of the goals of the experiment is to investigate the learning rate of the subjects. For that reason, the

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<sup>1</sup>10 movements between the 11 targets

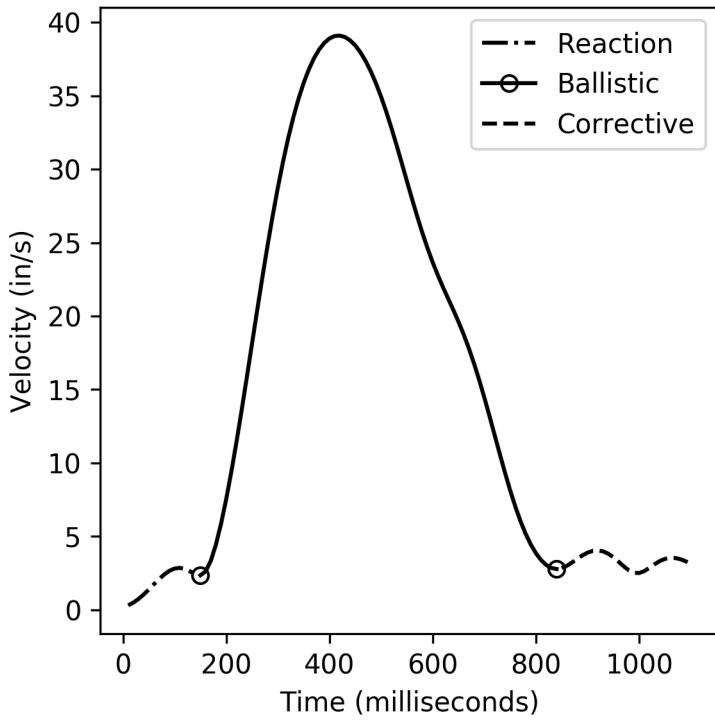
subjects were given no separate training time for the task or virtual environment. In lieu of collecting the data from fully trained subjects, the throughput analysis will be carried out on movements that are determined to be composed of mostly ballistic motion. The filtering parameters were determined post-hoc from the trajectory recordings. Their development and parameter selection are discussed in the results.

#### 4.3.4 Dependent Measures

The main dependent measure is the Fitts' throughput, measured through the movement time between button presses. Additionally, the trajectory of each movement is recorded from the hand tracker for analysis. To determine the arm fatigue, subjects were asked to rate their arm fatigue on the Borg scale from 6 to 20. The scale was presented with anchors as shown in Appendix A.1 (Table A.10). The arm fatigue rating was collected at the beginning of each condition, and then after every other configuration of distance and width combination (and after the final trial due to an odd number of trials). At the completion of each condition the subject was given a presence questionnaire. At the end of the experiment, an additional condition comparison survey was given to ask for opinions on the two haptic conditions.

#### 4.3.5 Trajectory Phases

Human reaching movements have long been known to consist of two distinct phases[37]: a ballistic phase and a corrective phase. The ballistic phase is an open-loop, gross adjustment towards the target, while the corrective phase is more refined movement honing in on the target. To separate the trajectories into the various phases, a simple algorithm was developed. First, the local minima are found throughout the velocity profile of the movement to separate the movement into various submovements. The “ballistic phase” is then classified as the submovement which contains the peak velocity of the entire movement. The various submovements after the ballistic phase are classified as the “corrective phase”. Any movement before the ballistic phase is classified as a “reaction time”. An example of the results of this classification is shown in Figure 4.3. This mathematical definition does break down for certain cases where a subject might have two submovements in the ballistic

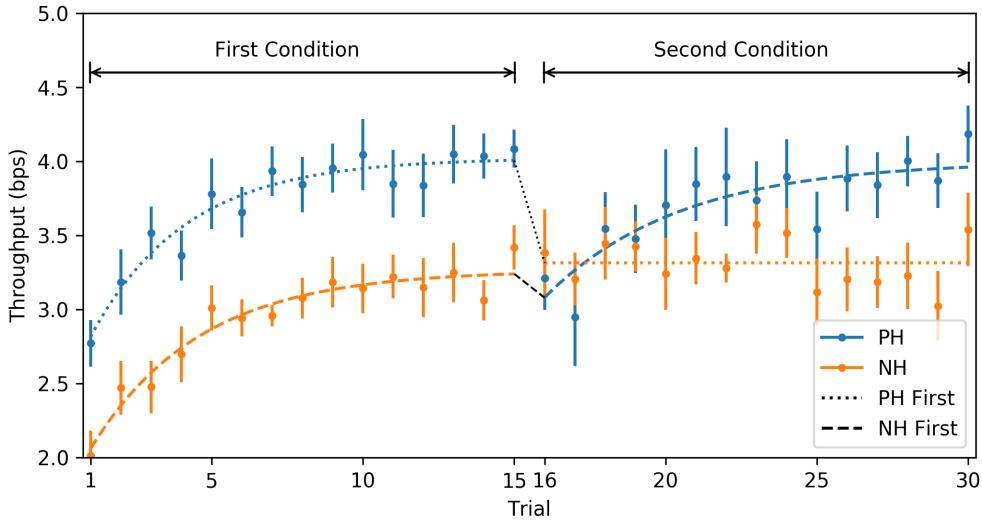


**Figure 4.3:** Example trajectory with three phases indicated.

phase due to a mid-course correction or similar, however one of the main purposes of this classification for this experiment was to find movements which are appropriate to use for the Fitts' Law calculations. Therefore, movements with multiple submovements are likely to not be ballistic, well-learned movements.

#### 4.3.6 Trajectory Filtering

We used a low-pass filter on the trajectory recordings to reduce the amount of noise. The LeapMotion processes data at a variable frequency, thus creating a variable rate for recording. The frequency typically varies from about 100Hz to 120Hz. To perform the filtering, the data was first resampled to a fixed rate of 100Hz. The filter used is a fourth-order Butterworth filter with a cut-off frequency of 5Hz. The cut-off frequency was chosen as voluntary hand movements have been shown to be below such a rate[].



**Figure 4.4:** Throughput per trial. The learning curve exponential fit is given by Eqn. 4.5 with parameters from Table 4.1.

## 4.4 Results

### 4.4.1 Participants

Twenty (20) subjects were recruited from the UC Davis engineering student population, both undergraduate and graduate students. The age range was 19-29 ( $M = 22.95, \sigma = 3.0$ ) with 16 males and 4 females. The genders were balanced amongst the counterbalanced sequence groups. All subjects indicated either less than one hour or no prior experience with virtual reality.

### 4.4.2 Throughput

The throughput is calculated per movement using Equation 4.2 and meaned per trial before being meaned per subject and condition. Throughput is used to investigate the first two research questions: do the subjects learn quicker and does their throughput performance improve with passive haptics?

## Rate of Learning

The average throughput for each trial, separated by haptics condition, is shown in Figure 4.4. An exponential rise to a learned state is fit to each haptics condition to model the learning curve of the subjects. The equation used is given as:

$$TP(T) = TP_{\infty} - (TP_{\infty} - TP_0)e^{(-T/\tau)} \quad (4.5)$$

where  $T$  is the trial number,  $TP_{\infty}$  is the asymptotic learned value of throughput,  $TP_0$  is the initial value at  $T = 0$  and  $\tau$  is the time constant. The time constant is the amount of trials for the throughput to rise 36% of the difference between the fully learned throughput ( $TP_{\infty}$ ) and the initial throughput ( $TP_0$ ). The parameters of the fit for each condition is shown in Table 4.1, as well as the standard error of the estimate (SEE). The value of throughput from the regression fit as a percentage of the fully learned value ( $TP_{\infty}$ ) for representative trials are listed in Table 4.2.

The rate of learning is very similar amongst both groups in the first condition. For the first condition it can appear that the subjects performing NH learned quicker than their counterparts performing PH by looking at the time constants. However, as the values in Table 4.2 show, the NH First group started their first trial at a lower percentage of their learned state. The shape of the learning curves are very similar and both groups reached approximately 90% of their fully learned state by the 5<sup>th</sup> trial.

The learning curves are quite different for the second condition. The NH condition does not have a learning curve at all, with a straight line being a better fit than the exponential function. This indicates the transfer of training from the PH condition allowed the group who did PH first to immediately perform in NH at the same level as the fully learned state of the subjects who learned NH in their first condition. This transfer of training to the second condition did not occur as strongly for the group who did NH first. Their initial performance of PH did start out at a slightly higher level than the subjects who did PH first (2.8 bps vs 2.4 bps), but after 5 trials this difference had converged (3.6 bps vs 3.7 bps).

These results indicate that subjects did not learn faster with Passive Haptics. The only differences between learning rates is the positive transfer of training from performing

Sequence	Condition	Haptics	$TP_{\infty}$	$TP_0$	$\tau$	SEE
PH First	First	PH	4.0	2.4	3.2	0.58
PH First	Second	NH	3.3	3.3	0.0	0.65
NH First	First	NH	3.3	1.7	3.6	0.49
NH First	Second	PH	4.0	2.8	4.5	0.78

**Table 4.1:** Exponential fit parameters of Eqn. 4.5. Curves are shown in Figure 4.4.

Sequence	Condition	Haptics	$TP_1$	$TP_5$	$TP_{10}$	$TP_{15}$
PH First	First	PH	70.0%	91.5%	98.3%	99.6%
PH First	Second	NH	100.0%	100.0%	100.0%	100.0%
NH First	First	NH	63.0%	87.8%	97.0%	99.2%
NH First	Second	PH	77.0%	90.6%	97.0%	99.0%

**Table 4.2:** Percentage of fully learned state for various trials for each group and condition from learning model fit.  $TP_i$  is  $TP(i)/TP_{\infty}$  using Eqn. 4.5. Trial 15 is the final trial for each condition.

Passive Haptics first and No Haptics second. The answer to the research question *Do subjects learn the task quicker with passive haptics?* appears to be that the passive haptics does not make subjects learn faster, but they are able to learn the task quicker without passive haptics afterward. It does appear that the fully learned state is different between the haptic conditions, which is investigated further in the next sections.

### Ballistic Movement Filtering

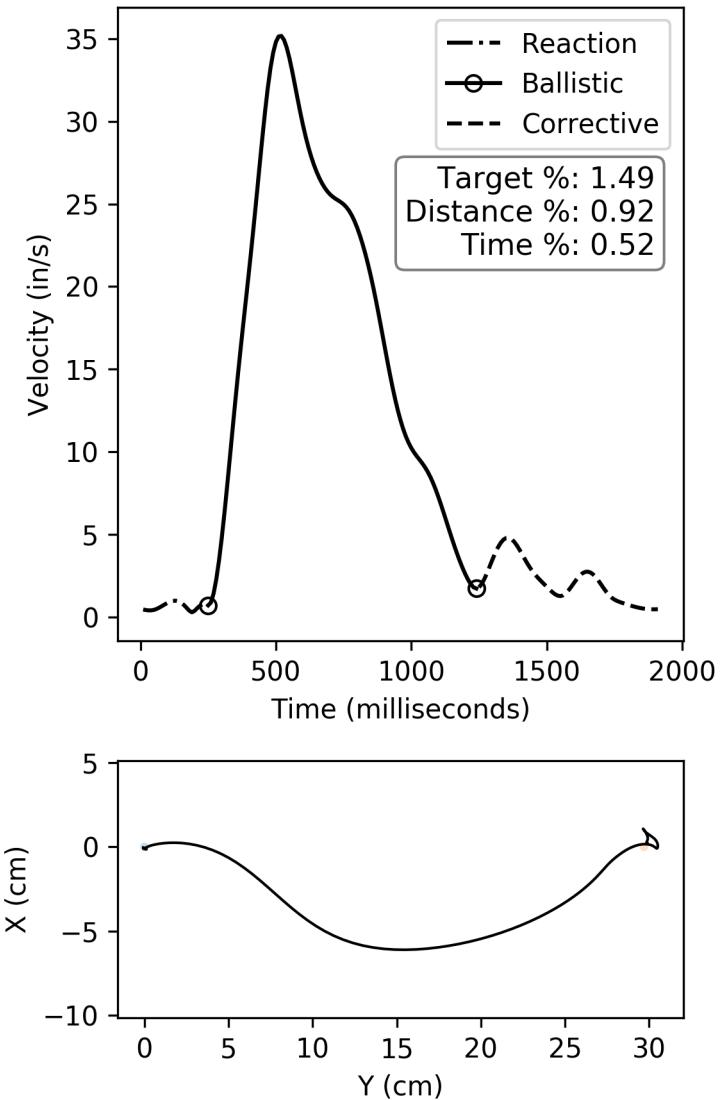
Before they could be used for the Fitts' Law analysis the trajectories were filtered so that only movements which were direct to target were included. A well-learned movement appropriate for Fitts' Law is one which moves directly towards the target and does not have much extraneous movement or idle time beyond what is required to complete the task. A

number of deviations from these well learned movements were observed in the dataset which we aimed to filter out for the final throughput calculations. We describe in this section the three metrics developed that were used to determine whether a movement was direct to target.

The first metric was the ratio of path distance travelled in the ballistic phase to the distance between the targets for that movement (i.e. 20cm, 30cm or 40cm). Ideally, the ballistic portion would cover the majority of the distance between the targets. A movement which covered too little of the target distance could mean the subject slowed or stopped in the middle of movement, and one that covered more could mean the subject overshot or had an indirect trajectory. A sample movement that gets flagged by this filter is shown in Figure 4.5, which has a ratio of 1.49. This is an example of a movement that does seemingly have a direct movement toward the target, but includes a large deviation perpendicular to the movement axis. This deviation could mean the subject initially aimed their ballistic portion in the wrong direction but performed a correction during the movement. The limits for this filter were chosen as having a ratio between 0.90 and 1.10, i.e. within 10% of the target distance. This filtered out 4577 of the 17970 movements.

The second filtering metric was also based on the ballistic phase path distance. For this metric it was compared to the total path distance the subject travelled for their entire movement. This filter targeted movements where after the ballistic phase the subject moved away from the target, or had a smaller but significant movement before the main ballistic movement. An example is shown in Figure 4.6 which shows a movement which passed the first metric (the ballistic phase distance ratio to the target distance was 1.03), but the ballistic phase was only 56% of the total distance travelled during that movement. This was a common problem where a subject would have a false start for the next target and move away from the current target before their button press was activated. The threshold for the filter was set at 0.80, leaving 4264 movements lower than the threshold, though only 1675 were unique from the target distance filter.

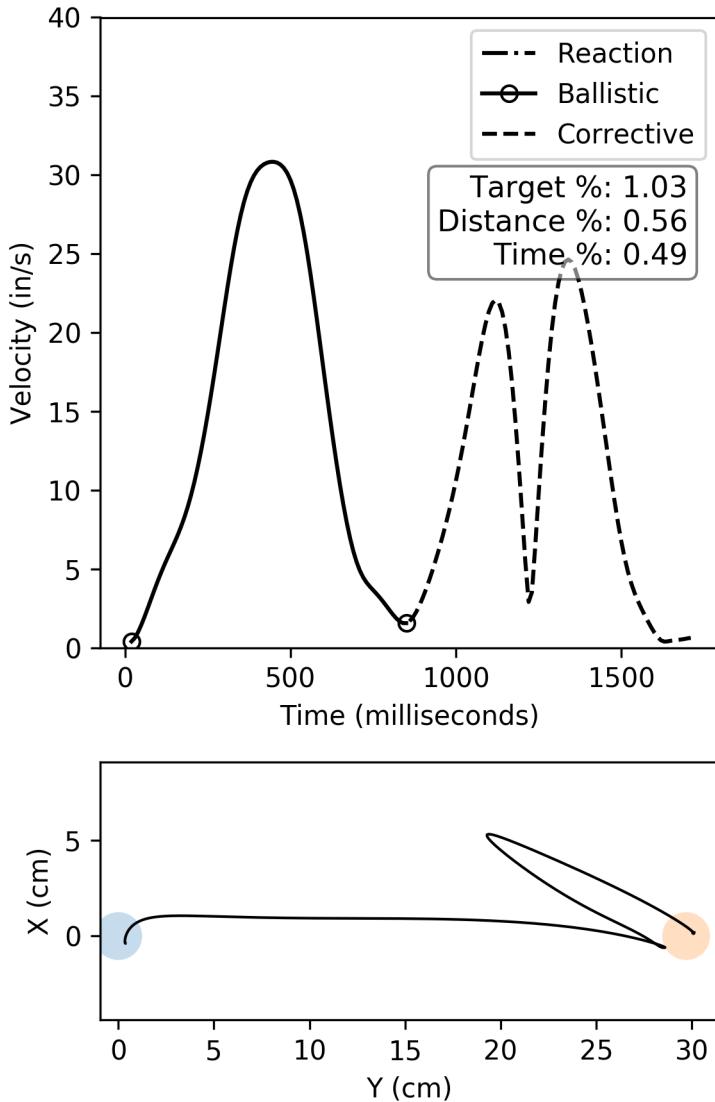
The last filter took a time based approach, and looked at the ratio of the time spent in the ballistic phase over the total movement time. This filter removed movements where the subject spent an inordinate amount of time either before or after the ballistic phase.



**Figure 4.5:** An example of a movement with a large ballistic distance to target distance ratio. The bottom plot shows the projection of the trajectory on the plane of the Fitts' circle.

If they were not moving during the non-ballistic phases, it would not have been caught by the distance-based filters either. The threshold of 0.40 meant that 4496 movements were filtered, however only 873 of those are unique of the other two filters. Figure 4.7 illustrates a movement where the subject waited before initiating the ballistic movement. Since the subject did not move during this idle time, the other two filters did not flag this movement.

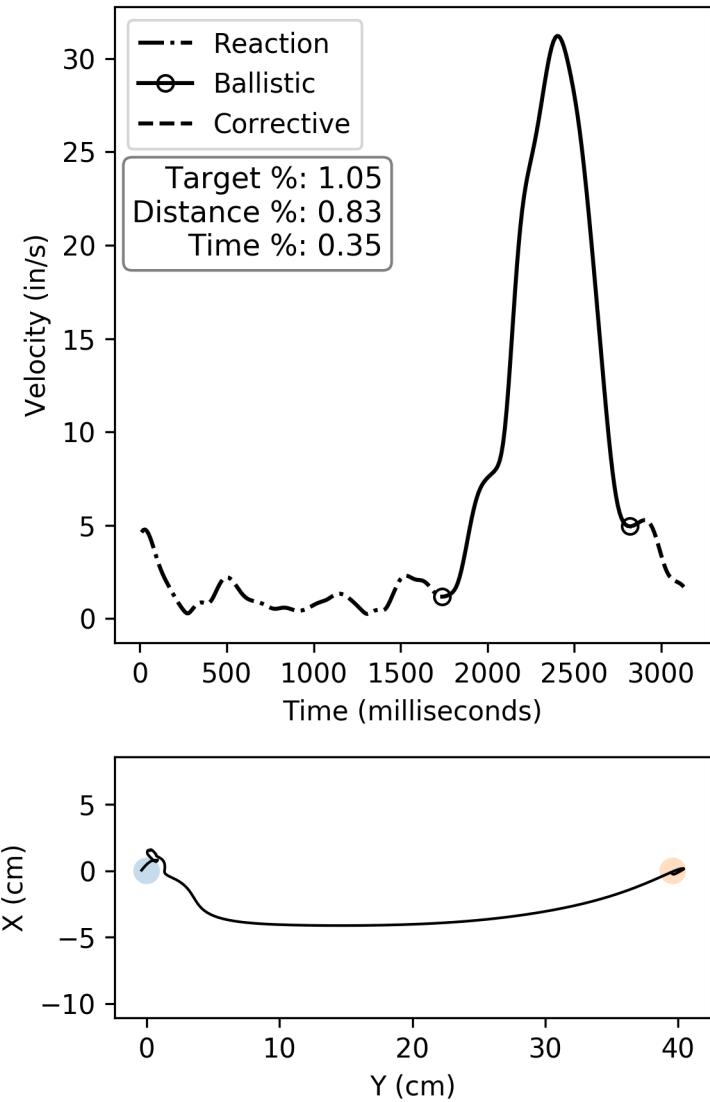
The combination of these three filters led to 7125 of 18000 movements being filtered out, leaving 60% of the movements. For each of the metrics, the threshold was determined



**Figure 4.6:** An example of a movement with a small ballistic distance to total path distance ratio. The bottom plot shows the projection of the trajectory on the plane of the Fitts' circle.

by investigating the distribution and looking at sample movements on either end of the threshold to determine if it was an appropriate value.

The final check before performing the Fitts' calculation was to ensure that each trial had enough data for the adjustment for accuracy calculation. The adjustment for accuracy is based on the distribution of endpoint data from a single trial (which is one distance and width configuration). One trial consists of 30 consecutive movements, but the ballistic filtering could diminish the amount remaining in each, so a trial was only included



**Figure 4.7:** An example of a movement with a small ballistic time to total time ratio. The bottom plot shows the projection of the trajectory on the plane of the Fitts' circle.

if at least half (15 of 30) of the movements were considered to be good movements by the ballistic filters. This means that movements from a trial that did not have enough good movements were also filtered out from the Fitts' calculation. Not only is this important to make sure the adjusted width is valid, it also removes trials where the subject likely did not reach a fully learned state, as most of their movements were not primarily ballistic movements direct to target. On average, 11 of 15 trials per condition from each subject ( $M = 10.98, \sigma = 3.25$ ) had enough good movements to be included. This left 9218 move-

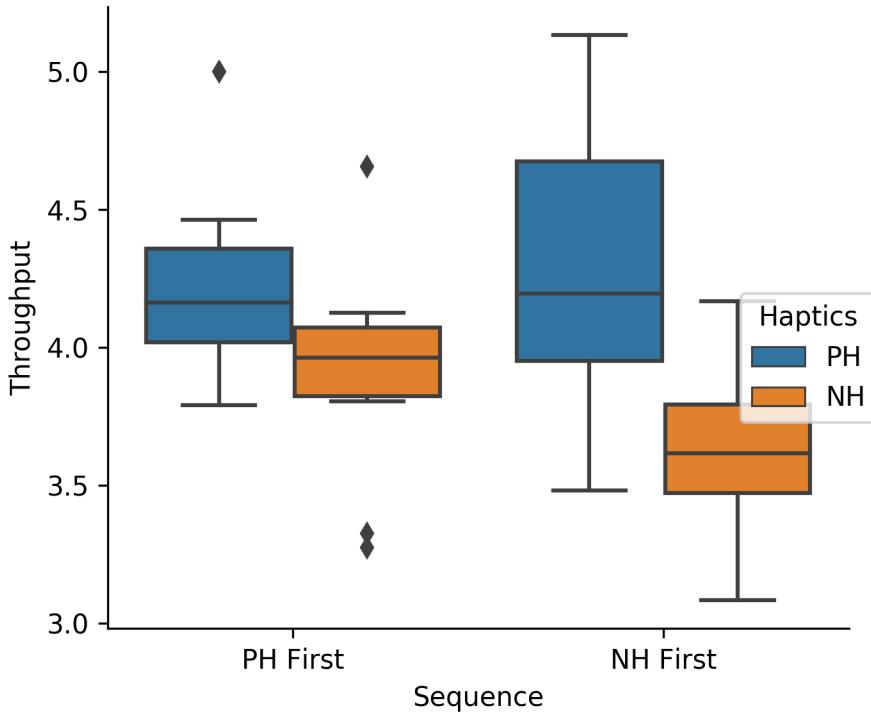
ments for the Fitts' calculation, just over half of the total movements (51.3%). Slightly more movements were filtered from the No Haptics condition, with 47.1% of NH movements left after the filtering, compared to 55.5% of the PH movements.

## Throughput

The throughput was found to be higher in the PH condition, at 4.25 bps compared to the 3.76 bps of the NH condition. A two-way mixed ANOVA was performed to determine the effect of haptics condition. Since we already expected order effects and have seen them with the transfer of training seen in the Rate of Learning section, the sequence the subjects performed the haptic conditions was included as a between subjects factor. The effect of haptics was found to have a significant effect on the throughput ( $F(1, 18) = 35.59, p < 0.001$ ) between the PH condition ( $M = 4.25, \sigma = 0.44$ ) and the NH condition ( $M = 3.76, \sigma = 0.38$ ). There was no effect on throughput based solely on sequence group ( $F(1, 18) = 0.53, p = 0.47$ ), but there was a marginally significant interaction effect between the sequence and haptics ( $F = (1, 18) = 4.48, p = 0.048$ ).

As can be seen in Figure 4.8, this marginal interaction effect appears to indicate that both groups had improved performance but the PH First group performed better at the NH condition. A post-hoc repeated measures ttest between haptic conditions for the subjects who performed PH First was significant ( $t(9) = 4.62, p < 0.001$ ), with the PH condition ( $M = 4.23, \sigma = 0.34$ ) outperforming the NH condition ( $M = 3.91, \sigma = 0.40$ ). The mean of the differences between subjects was 0.32 bps. The group of subjects who performed NH first also had a significant effect in the ttest ( $t(9) = 3.96, p < 0.001$ ) between the PH condition ( $M = 4.28, \sigma = 0.54$ ) and the NH condition ( $M = 3.62, \sigma = 0.32$ ). There was a higher mean of differences (0.66 bps) with the NH First group than the PH First group. These post-hoc tests confirm that both groups had a significant effect of haptics, though the NH First group had a larger difference between the conditions.

It is worth noting that without using the ballistic filter, the major conclusions found do not change. The only major difference is the magnitude of the throughput and size of the differences. The statistical tests have the same results as well. The results by haptics for both filtered and unfiltered are shown in Table 4.3. These results indicate



**Figure 4.8:** Throughput boxplot by haptics and sequence.

	Filtered		Unfiltered	
Haptics	Mean	SD	Mean	SD
PH	4.25	0.44	3.86	0.50
NH	3.76	0.38	3.25	0.37

**Table 4.3:** Throughput scores by haptics condition.

that subjects do have higher throughput with passive haptics, answering our first research question.

#### 4.4.3 Trajectory Phases

As described in Section 4.3.5, each movement of the subjects can be dissected into three distinct phases: reaction time, ballistic phase, and corrective phase. We have already seen that, overall, the subjects took more time to complete a movement without the passive haptics in place with the results of throughput. In this section we investigate the differences

in time spent in the three phases. We report here the means of time spent in each of the three phases. Times reported are all milliseconds. The results are shown for the filtered and unfiltered results, where the filtered results only include movements that were deemed purely ballistic by the filtering methods in Section 4.4.2. The unfiltered results include all movements. The phases were meaned per subject first, and then by condition. The time spent in each phase is listed in Table 4.4. Each phase was tested for the effect of haptics and sequence through a mixed two-way ANOVA. The statistical tests reported here are for the filtered results, but the significances do not change between filtered and unfiltered.

The reaction time had no effect by haptics ( $F(1, 18) = 0.32, p = 0.58$ ) or sequence ( $F(1, 18) = 0.001, p = 0.98$ ). However, for the interaction between the two a significant effect was found ( $F(1, 18) = 18.56, p < 0.001$ ). The interpretation of this interaction effect without main effect significance is that the first and second condition had a different reaction time, without dependence on the haptics condition or group. The mean reaction time in the first condition for both groups was 130.9 milliseconds ( $\sigma = 41.0$ ), but in the second condition it was just over 20 milliseconds faster, with an average of 109.1 milliseconds ( $\sigma = 36.6$ ). The reaction time for both conditions was lower than generally accepted values for reaction time to a visual or aural stimulus[]]. This is not surprising, as the task was a serial task which the subjects would learn the pacing of throughout the experiment. It would be expected that they were able to learn to anticipate the activation of a button (which was also the start of the next movement) as it would activate 160 msec after the subject entered the zone of the previous button. This interaction effect indicates that subjects did learn how to anticipate the activation event independent of the haptics or the order they performed the conditions.

The ballistic phase time had a significant effect of haptics ( $F(1, 18) = 24.14, p < 0.001$ ) between PH ( $M = 772.6, \sigma = 67.7$ ) and NH ( $M = 719.5, \sigma = 60.0$ ). There was no effect of sequence or the interaction effect between haptics and sequence. Since the ballistic phase should be mostly independent of the use of passive haptics, it was not expected to see the ballistic phase have an effect of haptics. It is unclear the exact mechanism that led to this, but it could be an artifact of the passive haptics causing the subjects to learn the movement, and thus allowing them to move quicker. There is little difference between the filtered movements and unfiltered movements results, the difference of the means were

		Filtered			Unfiltered		
		Mean	SD	p	Mean	SD	p
Reaction Time	PH	118.58	45.32	0.57	241.57	205.39	0.33
	NH	121.44	34.80		291.97	169.19	
Ballistic Phase	PH	719.54	59.99	< 0.001	679.48	44.79	< 0.001
	NH	772.57	67.62		727.53	60.01	
Corrective Phase	PH	199.13	48.10	< 0.001	564.73	253.56	< 0.001
	NH	256.58	42.08		837.22	293.15	

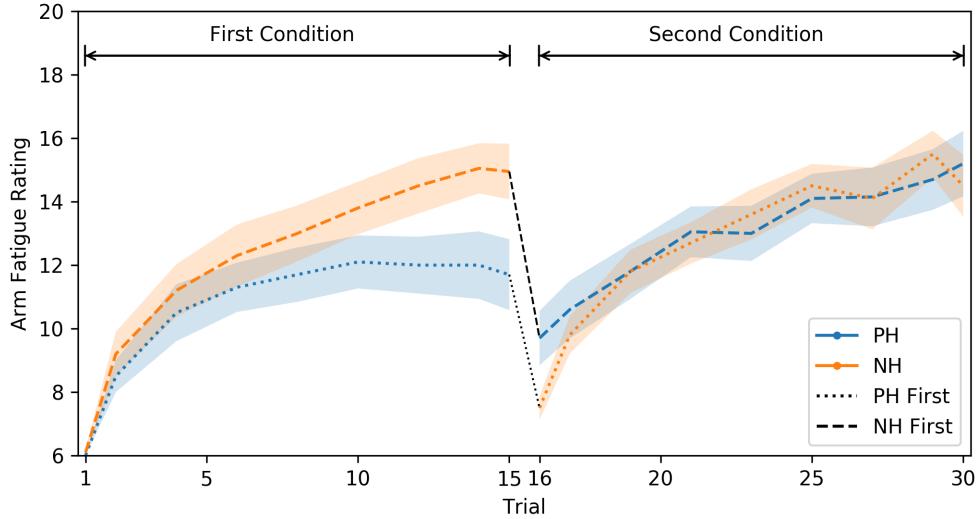
**Table 4.4:** Time in each movement phase by haptics conditions.

within a few milliseconds. Although the ballistic phase had lower movement times with the unfiltered movements, this is due to including movements where the ballistic was not the complete movement, reducing the mean.

The corrective phase time also had a significant effect of haptics ( $F(1, 18) = 22.46, p < 0.001$ ) between NH ( $M = 256.6, \sigma = 42.1$ ) and PH ( $M = 199.1, \sigma = 48.1$ ). There was no effect of sequence or the interaction effect between haptics and sequence. This result was expected as one of the main benefits of the passive haptics is that the subject does not have to ‘find’ the target along one dimension. There was a more noticeable difference between the results of the unfiltered and filtered movements for the corrective phase. The corrective phase time was much higher in both conditions for the unfiltered, with PH having a mean of 564.7 ( $\sigma = 253.6$ ) and NH having a mean of 837.2 ( $\sigma = 293.2$ ).

#### 4.4.4 Arm Fatigue

The subjects were asked for a rating of their arm fatigue every other trial, as well as before the first trial and after the last trial of each condition. One trial lasted for 30 movements and consisted of a single distance and width configuration. The scale ranged from 6 to 20, and subjects were allowed to record decimal ratings. The full scale



**Figure 4.9:** Arm Fatigue by Trial.

with anchors are shown in Appendix A.1 (Table A.10). The average rating at each trial, separated by haptics condition, is shown in Figure 4.9.

There is an evident difference between the two haptic conditions for the first condition performed, with the NH condition subjects accumulating more fatigue throughout the trials. At the end of the first condition, the subjects who performed the PH condition rated their arm fatigue 3.25 points lower on average than the NH condition subjects. Both groups have a similar rate of recovery, but the second condition quickly converges and shows no apparent difference between the two haptics conditions.

A within-subjects repeated measure (haptics) with two between-subjects measures (sequence and trial) ANOVA was performed to test the significance of haptics and the interaction effect of haptics and sequence. The interaction effect of haptics of sequence was found to be significant ( $F(1, 18) = 22.6, p < 0.001$ ) as well as the main effect of haptics ( $F(1, 18) = 5.47, p = 0.03$ ). As a result of the significant interaction effect, a post-hoc ANOVA with haptics and trial as the two within subjects repeated measures was run on both sequence groups. The NH First group had no effect due to haptics ( $F(1, 9) = 2.08, p = 0.18$ ), consistent with the observations from Figure 4.9. The subjects rated the same trial between conditions an average of only 0.69 ( $\sigma = 2.1$ ) points higher for the PH condition, which was their second condition. The PH First group did have a significant effect due to haptics

Haptics	Score	Std. Dev	Cronbach's alpha
PH	77.7	9.56	0.718
NH	71.0	9.70	0.711

**Table 4.5:** Presence Score Summary

( $F(1, 9) = 42.37, p < 0.001$ ). This group rated the PH condition an average of 2.0 ( $\sigma = 2.0$ ) points lower within trials.

These results show that the subjects only had reduced arm fatigue using the passive haptics for the first condition. For all other conditions the arm fatigue ratings reached the same level by the end of the condition.

#### 4.4.5 Presence

The presence survey was administered after each haptics condition. The questions had a 7-point Likert scale response with anchors at either end and the middle. The score given in this section is a sum of the responses, on a scale of 1 to 7, where a higher score indicates higher presence. A few questions were asked with an inverted scale (i.e. a score of 1 indicated higher presence) and were reversed before the score was calculated. The internal consistency of the presence questionnaire was tested per condition using Cronbachs' alpha, and was found to be consistent in both conditions ( $\alpha = 0.72$  and  $\alpha = 0.71$  for PH and NH, respectively). The full survey questions and average responses per condition are listed in Table 4.6. The item total correlation (the correlation between the questions' score and the total score) is also listed for each question.

The average scores per condition are given in Table 4.5. The presence scores were tested with a mixed within subjects repeated measures (haptics) and between subjects measures (sequence) ANOVA. The score had a marginally significant effect between haptics conditions ( $F(1, 18) = 6.08, p = 0.024$ ), with Passive Haptics having a slightly higher mean ( $M = 77.7, \sigma = 9.56$ ) than No Haptics ( $M = 71.0, \sigma = 9.70$ ). There was no significant effect of sequence ( $F(1, 18) = 4.01, p = 0.58$ ) nor for the interaction effect between sequence and haptics ( $F(1, 18) = 0.71, p = 0.41$ ).

Question	NH		PH	
	Score	ITCorr	Score	ITCorr
1. How much were you able to control events?	4.8	0.59	5.3	0.74*
2. How natural did your interactions with the environment seem?	4.3	0.45	5.0	0.81*
3. How much did the visual aspects of the environment engage you?	5.4	0.36	5.3	0.65
4. How much did the auditory aspects of the environment engage you?	6.2	0.32	6.0	0.19
5. How much did the tactile (sense of touch) aspects of the environment engage you?	2.6	0.31	5.6	0.53
6. To what extent did you associate the computer generated arm and hand with being "your body" while in the virtual environment?	4.5	0.35	5.0	0.65
7. How natural was the mechanism which controlled movement through the environment?	4.0	0.48	4.5	0.72*
8. How much did your experiences in the virtual environment seem consistent with your real-world experiences?	4.0	0.49	4.5	0.66
9. How involved were you in the virtual environment experience and the task you were performing?	5.8	0.50	5.8	0.22
10. How distracting was the control mechanism? <sup>†</sup>	4.0	0.52	4.7	0.29
11. How much delay did you experience between your actions and expected outcomes? <sup>†</sup>	4.7	0.38	4.9	0.04
12. How quickly did you adjust to the virtual environment experience? <sup>†</sup>	2.9	-0.08	2.1	-0.31
13. How proficient in moving and interacting with the virtual environment did you feel at the end of the experience?	4.9	0.65	5.2	0.25
14. How much did the control devices interfere with the performance of assigned tasks?	4.0	-0.40	4.0	-0.45

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Question	NH	Neither	PH
<b>Q1.</b> In which condition did you feel you performed faster?	1	1	18
<b>Q2.</b> In which condition did you feel you performed more accurately?	3	2	15
<b>Q3.</b> Did you feel that your arm fatigued more or quicker in one condition over the other?	13	3	4
<b>Q4.</b> Did you feel that you were actually in the virtual room more in one condition over the other?	3	4	13
<b>Q5.</b> Which condition did you prefer?	0	2	18

---

**Table 4.7:** Condition comparison survey summary of results.

The questions that correlated the strongest with the presence score were questions 1, 2 and 7: all asking about control mechanisms and how natural they felt. The PH condition found these to be significant ( $p < 0.001$ ). The question with the biggest difference between the two conditions was question 5 which asked directly about the tactile aspects of the environment, which the PH condition scored 3 points higher on average. After that, the questions with the largest difference were questions 1, 2, 10 and 15: all questions asking about control mechanisms again, and all providing a larger score with the PH condition.

#### 4.4.6 Condition Comparison

The condition comparison survey asked the subjects five questions directly comparing the two conditions. The questions and a summary of the responses are shown in Table 4.7. The subjects overwhelmingly responded that they preferred the Passive Haptics (PH) condition (Q5), with all but two subjects choosing it. In fact, no subject preferred

the No Haptics (NH) condition, the two subjects who did not choose PH responded that neither was more preferred.

The other questions had responses similar to the results from the other sections. The majority of subjects responded that they were more accurate and faster in the PH condition (Q1 and Q2), which agrees with the throughput results. The subjects who chose Neither or NH were usually not actually faster or more accurate in the NH condition. In fact, only two subjects had a throughput that was higher in the NH condition, and neither subject chose NH as the condition they performed faster in, though one did say they performed more accurately in the NH condition.

Question 4 asked subjects directly about their feeling of presence, and 13 subjects chose PH, with the remaining split between 4 saying neither and 3 saying NH. The results of the presence questionnaire suggested that subjects felt more present with the passive haptics, which this agrees with. The arm fatigue question (Q3) was worded to ask which condition they felt provided either more or quicker arm fatigue, and 13 subjects felt this was the case with the NH condition. The remaining were split between neither (3) and PH (4). The results of the arm fatigue questionnaire during the experiment found a similar result as the condition comparison survey.

## 4.5 Discussion

After they performed their first condition the experimenter would reposition the panel and explain their next condition. This meant that subjects who started with the panel (PH) would then see the panel be removed and have the no panel (NH) condition explained to them, and vice-versa. Subjects often made comments during this change about how they were happy for the panel to be added, or concerned about losing it. For example, one subject upon the placement of the panel commented “this is definitely going to be better.” Another who was transitioning from PH to NH simply remarked, “Oh, no!” One of the subjects who felt that they performed faster and more accurately without the panel (NH condition) in the final questionnaire explained afterward that this was due to often needing to press the button twice in the PH condition, which we believe was likely due to their lack

of adaption to the button detection algorithm.

The PH condition is a similar setup to one condition tested in Kohli et al.[19]. They found a throughput value of  $\tilde{6}$ bps for this condition, compared to our result of 4.25bps, which was significantly lower. Their setup used a marked fingertip tracker instead of our markerless approach, and the target activation occurred with low latency upon contacting their passive haptics. Both of these could be the reason for their higher throughput. Seixas et al. [28] performed a similar condition to our NH condition, using a LeapMotion to perform a Fitts' circle. The major difference was the visual feedback was given on a computer monitor, and offset from the physical location of the hand. Their results found a throughput of  $\tilde{2.9}$ bps, much lower than our 3.76bps. This performance improvement in our experiment is likely due to the additional immersion of having a head-mounted display.

## 4.6 Conclusion

We found that the use of a passive haptic for a 2D targeting task caused a significant increase in Fitts' throughput. The subjects did not learn the task any quicker with passive haptics, but a positive transfer of training existed for subjects who used passive haptics and then later performed the same task without passive haptics. Upon investigating the trajectories of the movements, it was found that the passive haptics reduced the amount of time subjects spent in the corrective phase. Subjects reported lower arm fatigue when using the passive haptics, though only for the first condition performed. The Presence Questionnaire (PQ) score was marginally significantly higher for the passive haptics condition. Subjects did overwhelmingly report that they preferred the passive haptics condition.

# Chapter 5

## Design Evaluation Experiment

### 5.1 Introduction

The final experiment combines the lessons of the previous experiment to investigate the use of the Rapidly Reconfigurable Research Cockpit (R3C) in a design evaluation study. The goal of this experiment is to determine if the R3C system can be used in the place of a more traditional evaluation tool. As previous chapters have discussed, there are a number of self-evident advantages to using the R3C system. However, there remain some technical limitations to the technology that could hinder adoption. We found that a button targetting task took more time in our virtual environment than in the real world (Chapter 3). The following experiment (Chapter 4) found that a Fitts' Law task produced a higher throughput using a passive haptics layer, mitigating some of the time increase of targeting buttons in a cirtual environment. In the experiment described in this chapter, we used the R3C system as the simulation tool for a design evaluation study of a cockpit instrument. The purpose in undergoing this evaluation study is to understand if these limitations would interfere with the metrics that might be used in evaluating a new cockpit design.

We designed an experiment which asks for feedback from subjects who take the role of design evaluators for a cockpit instrument. The subjects were divided into two groups: one group used an R3C setup to operate the instrument, while the other used a more traditional setup: a touchscreen simulator of the instrument. This seperation of groups will allow a comparison of the feedback from subjects between groups. Both groups

**Figure 5.1:** Simulator Workstation

evaluated the same two instrument designs, and subjects were asked to provide feedback using the same questionnaires. We hypothesize that the R3C system could be used in place of a traditional simulator if the two groups provide similar responses to the designs. Additionally, we utilized common quantitative metrics to evaluate performance to determine if the conclusions that would be drawn from these change between groups.

## 5.2 Methods

### 5.2.1 Simulator Setup

The simulator workstation as configured for each group is shown and annotated in Figure 5.1. It was designed to have as much as possible to be the same between the two configurations. The joystick and instrument were positioned in the same location for each group. Neither group had out the window visuals, relying only on the attitude indicator on the instrument. For the Virtual Reality (VR) group, the visuals showed a plain interior of a cockpit, but the out-the-window view was black. Both groups had an aural indication (a click noise of a button being pressed) when a button was activated on the instrument, using the speakers mounted behind the instrument panel.

Beyond the VR group using a virtual reality headset for the visuals, the main difference between the two groups was the method for pressing the buttons on the instruments. The VR group used the hand tracker activated system previously described in Chapter [chapter 2](#). For this experiment, the buttons were configured to highlight a blue color when the hand tracker registered a finger within the zone. The zones were extended 0.1in around the border of the button, raised a height of 0.5in above the surface of the button. When the button was activated after the 150 millisecond delay, the highlight would disappear and the button in the virtual world would move inwards as if it were being pushed in<sup>1</sup>, the press sound would play, and the behavior on the instrument associated with pressing that button would occur. A separate release sound would play when the finger left the zone after a

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<sup>1</sup>Of course, the physical button could not and did not move.

successful press, and for the VR group the button would move back to its starting position.

The Touchscreen (TS) group used a 10.1 inch capacitive touch screen with resolution of 1024x600. The active area of the screen was 8.8in by 5.1in, with outside dimensions of 10.4in by 6.7in. The two instruments were rendered in a web browser using standard HTML elements. Javascript press and release events were used to simulate the same behavior as described for the VR group, except for the highlighting before a button press. The visuals of the tracker were rendered on top of the browser window with the same OpenGL rendering code used for the VR group.

### 5.2.2 Task Design

Based on the technology available for the simulator base, a number of requirements were laid out that would guide the design of an appropriate task and instrument designs. The instrument and task required:

- Flight task using a standard joystick
- Additional task that requires use of multiple buttons on the instrument
- Able to develop simulator for both touchscreen and R3C setup
- Able to design two different layouts with one design having distinct flaws
- Simple design, yet complex enough task to have sufficient workload
- Operationally relevant tasks analogous to those required in a cockpit

Ultimately, we designed a task that required number and letter inputs using the buttons, while simultaneously flying a pitch disturbance profile.

### Tracking Task

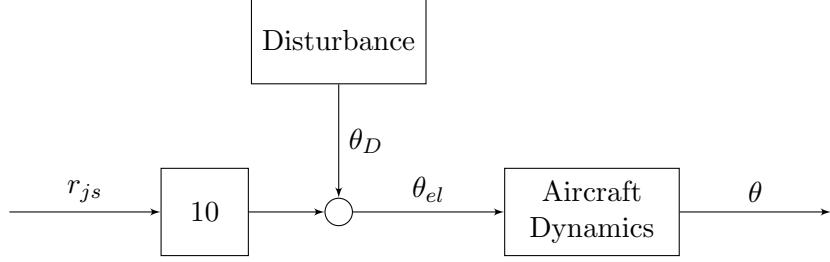
The tracking task display was a standard attitude indicator display, shown in Figure 5.2. Each tick corresponds to 1 degree in the dynamics simulation, with major ticks at intervals of 5 degrees. The attitude indicator was rendered to the size of 3.4 inches



**Figure 5.2:** Attitude Indicator Display

square on the instrument. Subjects controlled the one-dimensional (pitch only) task using a joystick with their left hand. The joystick is pictured in Figure 5.1.

The flight dynamics model of the simulator was a stability derivative based model for a Boeing 747 in low altitude flight. The block diagram of the dynamics is shown in Figure 5.3. The dynamics model was updated and recorded at a rate of 125Hz. The output of the joystick,  $r_{js}$ , varies from  $-1.0$  to  $1.0$ , and the gain of  $10^\circ$  was chosen to ensure the pilot had enough control authority to complete the task. The flight condition is listed as “Flight Condition 2” in NASA CR-2144[14]. This dynamics model was chosen due to availability. The specifics of the dynamical model were not important other than providing a response similar to an aircraft in flight. The model was linearized from sea-level flight at an airspeed of 335 ft/s. The configuration of the airplane had the gear up, no flaps, total weight of 654,000 lbs, and an angle of attack of  $7.3^\circ$ . The transfer function of the aircraft dynamics is given as:



**Figure 5.3:** Tracking Task Dynamics Block Diagram

$$\frac{\theta}{\theta_{el}} = \frac{-0.572(s + 0.553)(s + 0.0396)}{(s^2 + 2\zeta_1\omega_1 + \omega_1^2)(s^2 + 2\zeta_2\omega_2 + \omega_2^2)} \quad (5.1)$$

$$\omega_1 = 0.0578$$

$$\zeta_1 = 0.0160$$

$$\omega_2 = 1.12$$

$$\zeta_2 = 0.798$$

The disturbance model is based off the model developed in SweetRef[?]. It is designed to provide a broad spectrum of frequencies that the human controller needs to respond to. The disturbance is a sum of sines described by:

$$\theta_D = K \sum_{i=1}^{12} \left[ a_i \left( \frac{2\pi k_i}{240} \right) \sin \left( \frac{2\pi k_i}{240} t + \phi_i \right) \right] \quad (5.2)$$

The  $k_i$  terms are given as,

$$k_1 = 7, \quad k_2 = 11, \quad k_3 = 16$$

$$k_4 = 25, \quad k_5 = 38, \quad k_6 = 61$$

$$k_7 = 103, \quad k_8 = 131, \quad k_9 = 151$$

$$k_{10} = 181, \quad k_{11} = 313, \quad k_{12} = 523$$

The amplitude terms are  $a_i = 0.5$  for  $i \leq 6$  and  $a_i = 0.05$  otherwise. The phase terms,  $\phi_i$ , were randomly selected on the  $(-\pi, \pi)$  interval ensuring a uniform distribution. This random selection was pre-calculated for each trial, however the order was repeated for each subject so there was no between subject variance in the disturbance signal. Furthermore, each subject received the same sequence of disturbance signals for each instrument design. The disturbance amplitude,  $K$ , was chosen such that the root-mean square (RMS) of the

signal was 3.5 degrees. The value of the was chosen through pilot studies to ensure the task was challenging but not overwhelming.

### Prompting Task

The prompting task was designed to be both a realistic task for a cockpit as well as a demanding task when done in addition with the tracking task. The task developed required the subjects to read and memorize a short string of characters and enter it back using the buttons on the instrument. To limit the task physically (by number of buttons) and mentally, the characters used were the number 1 through 6 and the letters A through F. The prompts were 4 characters long and once the subject started entry the prompt would disappear, forcing them to hold it in short term memory.

The sequence of the prompts was separated into 10 second “windows”. The prompt would appear randomly between 2 and 3 seconds of the start of the window. From the time of appearance, subjects were given seven seconds until timeout. When the subject pressed the first button of the prompt, the prompt itself was cleared and asterisk symbols (\*) were shown in place of the prompt for each button entry by the subject. If the subject ran out of time, the text in entry area would return to black. Although subjects were briefed on the timeout and given practice to learn the pace, no warning or indication of time left was shown during the trials. Whether they completed the prompt within the time limit, or they timed-out, this process was repeated every 10 seconds. This meant that subjects had at least 3 seconds of time with no prompt.

The prompts themselves were always composed of three numbers followed by a letter or three letters followed by a number. This structure was decided upon to provide a consistent pattern, yet still utilize both letters and numbers in every prompt. The prompts were randomly chosen but were not allowed to have repeat numbers or letters. The selection of letters or numbers as the first three characters was randomly chosen as well, with an equal weight to each.

### 5.2.3 Instrument Designs

The two different designs used were developed to be both realistic as a cockpit instrument design that would be under consideration, yet still have one design with flaws that would be found in a design evaluation. We developed a ‘Keypad’ design with the prompting task button keys on the right side and the tracking task on the left, and an ‘Edgekey’ design with the prompt buttons split on either side of the tracking task display. The tracking task display was the same size on the display for both designs. The prompting task text was placed below the tracking task display, and the same font, size and color was used for both designs. The prompting task text font was approximately 0.62in tall. These were kept consistent to limit the number of possible variables between the two designs. The prominent difference is the placement and behavior of the buttons which is described in this section.

The Keypad design is pictured in Figure 5.4. The buttons are 1in by 0.75in, with about 0.26in between buttons horizontally and 0.38in vertically. Each button has the label directly on the top of the button. The 3D-printed instrument used for the VR group had the buttons raised a height of 0.31in from the surface of the instrument. The button labels were also raised to provide a tactile feedback. The font was approximately 0.36in tall, and the labels were embossed above the button surface 0.05in.

The Edgekey design is pictured in Figure 5.5. In this design, there is not a single button for every number and letter. Instead, the bottom button on either side would switch the behavior (and labels) of the remaining six buttons from being 1 through 6 to A through F. In other words, the bottom “switching” buttons would change the rest of the buttons from the numbers to the letters, and vice-versa. The labels were placed offset from the button on the “screen” portion of the instrument, allowing them to change dynamically. The fonts were approximately 0.32in tall, and were blue on the screen. The buttons are slightly smaller in this design, at 0.76in by 0.55in. A smaller button size was needed to fit the labels and the buttons side by side. The spacing between buttons vertically is the same as the Keypad design at 0.38in. The center to center distance between the two sides of the button rows is 7.3in. The 3D-printed instrument version had raised nubs on each

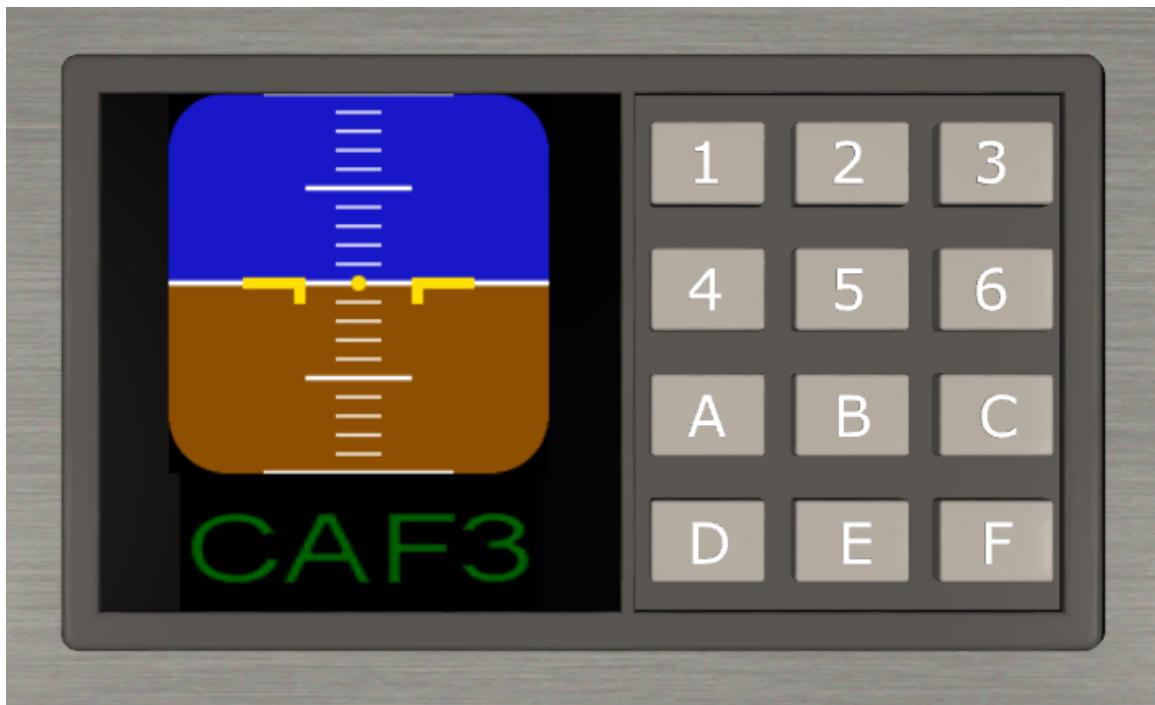


Figure 5.4: Keypad Design

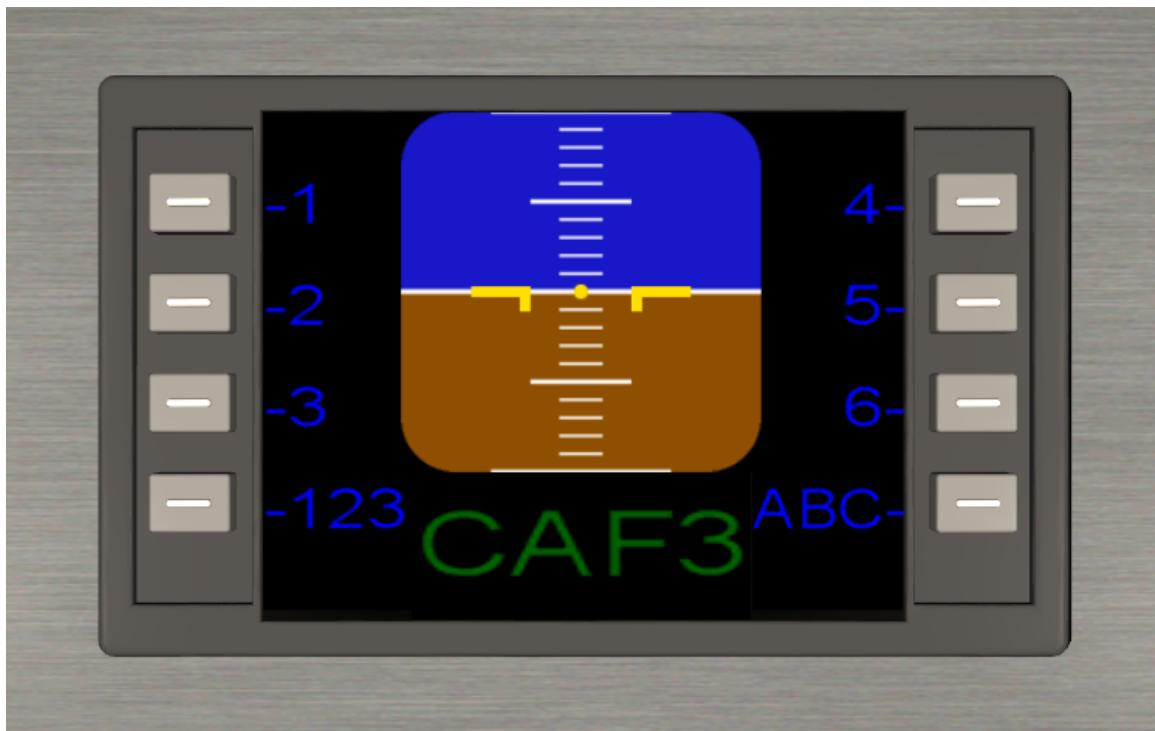


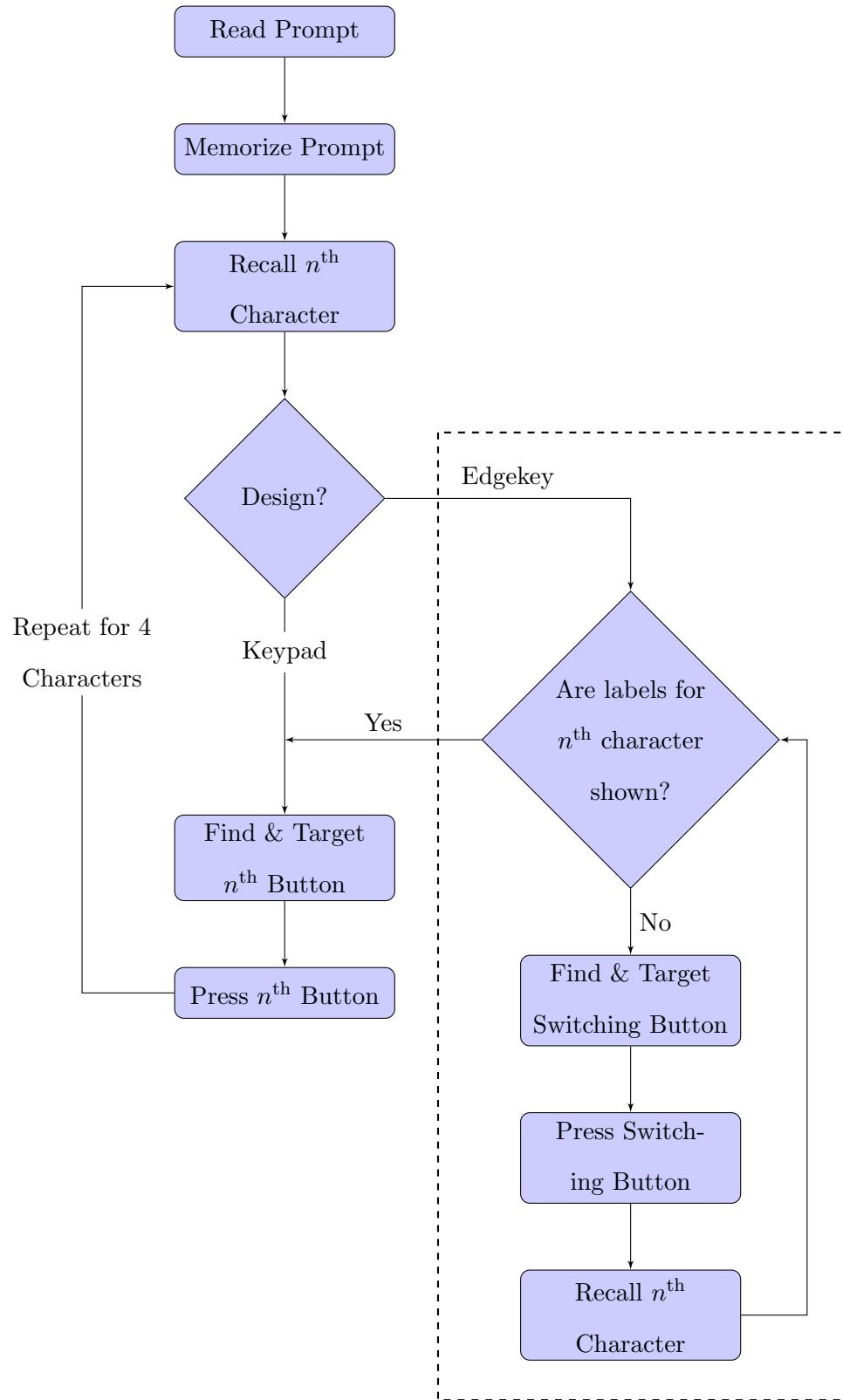
Figure 5.5: Edgekey Design

button covering half the width, 0.08in tall and raised 0.05in. As with the Keypad design, the buttons had the same height of 0.31in from the surface.

While some of the more subtle differences were expected to be noted by the evaluation study (e.g. having smaller buttons, different position of the flight task), the major flaw designed into the Edgekey design was the switching key to change from letters to numbers and back. This additional action fundamentally changed the demands of the task, as the subjects now had to press this additional button to change labels at least once per prompt. The cognitive workflow of the subject is diagrammed in Figure 5.6. The additional mental effort of the Edgekey design is shown in the dashed box, where the subject has to verify the state of the instrument and possibly press the switch button before they press the buttons of the prompt. Since the prompts were kept as a consistent format of three of the same type and fourth of the other, this extra work was easily skipped for most buttons, and anticipated between the third and fourth button. There was no guarantee that the next prompt would start with the instrument in the correct state for the new prompt, so there was always an additional cognitive load in determining whether a switch was necessary at the beginning of the prompting window, which would be accompanied with the physical effort if the switch was needed.

#### 5.2.4 Experiment Design

Subjects were divided into the two groups, Touchscreen (TS) and Virtual Reality (VR). The overall sequence of the experiment started with a training session on the simulator and the task, followed by an evaluation session for each of the two designs, finishing with questionnaires asking subjects to evaluate the two designs. The timeline of the experiment was the same for each subject, except for counterbalancing the order that the designs were evaluated. The training portion started with a slide deck explaining the tasks, the simulator that the subject was using (depending on which group they were in), and the functionality of the two designs they were to evaluate. Next, they performed practice trials with just the tracking task and then just the prompting task. The practice trials of the tracking task were 60 seconds long and repeated until the subjects' performance had flatlined. This took between three to six trials for each subject.



**Figure 5.6:** Prompting Task Flowchart of Cognitive Work for Each Design. Extra work of Edgekey design enclosed in dashed line box.

For the evaluation sessions with each design, they performed six trials with both tasks. The first three were a minute long, and were considered practice trials, and not included in the data analysis, though this was not communicated to the subjects. The following three trials were two minutes each, and were used for the analysis. Each evaluation session concluded with a two minute trial of just the tracking task without the prompting task. This was included to investigate if the subject had improved or fatigued at the tracking task throughout the experiment.

The independent variables of the experiment are Group and Design. The Group is the simulator the subject used, a between subjects factor, and either TS or VR. The Design is a within subjects factor, the two instrument designs that every subject evaluated — Edgekey and Keypad.

### 5.2.5 Dependent Measures

The dependent measures were chosen to evaluate the performance of each task individually as well as the workload of the subject. For the tracking task, the root-mean square error (RMSE) was calculated for each trial. The error in this case is simply the pitch shown to the subject, the output of the flight model described in Section 5.2.2.

The prompting task has two dependent measures, for speed and accuracy. For speed we consider the *response time*, defined as the time between when the prompt is first shown to the subject and when they press the first button of their response entry. The accuracy is measured by how many prompts they complete correctly. Twelve prompts are shown to the subject within each trial. The response time was meaned per trial first and then per design for each subject, and the number of correct prompts is meaned per design for each subject.

A NASA Task Load Index (TLX) survey was administered after they completed each design to measure the workload of the subject. The TLX survey asks for a rating of their workload between 0-100 for the following subscales: Mental Demand, Physical Demand, Temporal Demand, Performance, Effort, and Frustration. Our implementation allowed selection of the ratings within increments of 5, and included anchors of “Low” and “High” at the extrema of 0 and 100, respectively (except for Performance, which uses

“Good” and “Bad”). The midpoint was also visually indicated with a larger tick. The ranked pairs modification was used and completed for both times the subject took the survey. This modification asks the subject, for each of the pairwise combinations of subscales, which they felt contributed more to their workload. The number of times they select each subscale is used a weight to calculate a weighted mean for the total TLX score.

Finally, the subjects were given a questionnaire asking for their feedback on each instrument design. For each design, the subjects were asked the following questions:

- Please comment on any difficulties you had performing the prompting task with this design especially in contrast to the other design.
- Please comment on anything you liked in this design.
- Please comment on anything you did not like in this design.
- Any other comments?

Additionally, the following questions were asked:

- Which instrument design did you prefer? Why?
- Did you experience any physical fatigue during the experiment? Where?
- Any other comments?

An open form text box was used for the response field for each of these questions.

In a standard design evaluation study, the feedback received from the users in this questionnaire (and other debriefing interviews) would often be the main source for carrying out re-design. The purpose of this feedback in this experiment is to determine and document in which ways does this feedback differ. For example, if most subjects in one group noted issues with the size of a button, while no one in the other group found an issue with that button, this would indicate that using this VR system may not highlight the same issues regarding button sizes. The groups were purposely left ambiguous in the example, as it does not matter which group found the flaw and which group did not comment on it. Although we could postulate as to which group are “correct” in their evaluation of the instrument, it

is not a useful exercise, as the only result is to document what potential differences could arise so that users of this system can be aware.

With that goal in mind, the analysis of the feedback questions seeks to find differences between the groups. The sentences from the open form responses were first separated into single feedback comments, and summarized using common language. If a single subject repeated the same comment in the answers to multiple questions, they were only counted once. Each of these simplified feedback comments were assigned to a category or overall summary of their feedback. This process was completed separately for each group. We aim to look for feedback that is unique to a certain group or feedback that receives a higher frequency of comments in one group. This will provide a summary of where the groups provide the same feedback and where they provide differing feedback.

### 5.2.6 Hypotheses

The main hypothesis of this experiment is that the use of a VR/R3C simulator will not affect the conclusions of a design evaluation study, compared to a traditional touchscreen simulator. We do expect that some of the dependent measures may have a significant difference in Group or a significant difference in Design. The more important measure for us, however, is the interaction effect. This will test if the change between Designs is similar for the two Groups. If this is the case, then it may indicate that an evaluation study using one of these simulators could draw differing conclusions of an evaluation study using the other. Statistically, we will test the hypothesis that there exists no interaction effect between Group and Design for any of our dependent measures.

Additionally, the two tracking only trials performed at the end of each evaluation session, as well as the final tracking only training trial, will be used to investigate if the subjects were still learning the tracking task. The concern if subjects became more trained in the tracking task is that it could lower their attentional needs to that portion of the task, causing a change in performance on the prompting task that was not due to the design change. These hypotheses are enumerated here:

H1. The tracking task RMSE will have no interaction effect between Group and Design

H2. The prompt response time will have no interaction effect between Group and Design

H3. The number of correct prompts will have no interaction effect between Group and Design

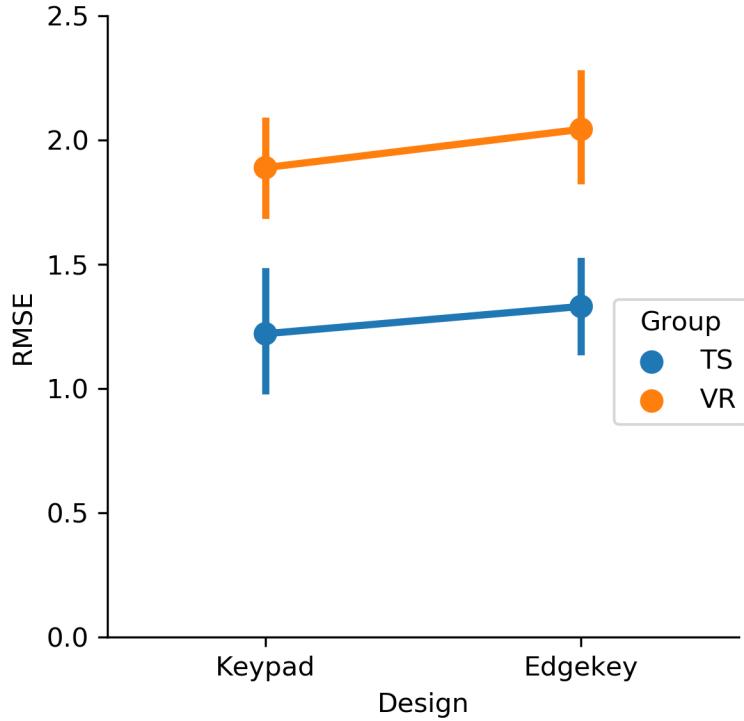
H4. The NASA TLX scores will have no interaction effect between Group and Design

H5. The tracking task RMSE for the last training trial and the tracking only trials will not change throughout the experiment

### 5.2.7 Statistical Tests

The quantitative dependent measures are tested with a two-way ANOVA, with one within subjects factor (Design) and one between subjects factor (Group). The Design factor contains two levels, the two designs each subject tested, Edgekey and Keypad. The Group factor also contains two levels, the VR (Virtual Reality) group and the TS (Touchscreen) group. When the ANOVA showed significance in the interaction test, post-hoc repeated measured t-tests were undertaken to determine the significance of Design within each Group. Independent samples t-tests were used to test the significance of Group within each Design. The last hypothesis testing the effects of learning on the trials with only the tracking task will be tested with a two-way ANOVA, with the Group as a between subjects factor, and the trial number as a within subjects factor. The trial number is chronological in the order the subjects performed them. The first trial was the last tracking only training trial, and the next two were tracking only trials at the end of each design evaluation.

Statistical significance level was corrected using the Bonferroni correction considering the 5 hypotheses being tested. All effects were considered statistically significant at the 0.01 level ( $\alpha = 0.05/5 = 0.01$ ). Effects which have a significance level between  $0.05 < p < 0.01$  are considered to be marginally significant.



**Figure 5.7:** Factor Plot of RMSE

## 5.3 Results

### 5.3.1 Demographics

Twenty-three subjects were recruited from the UC Davis engineering undergraduate and graduate student population. Twelve subjects were placed in the VR group, and the remaining eleven in the TS group. The mean age was 21.0 ( $\sigma = 3.14$ ), with 19 male and 4 female subjects. The genders were balanced between the two groups. Most subjects had no flight experience (two were student pilots), and all of the VR group subjects indicated that they had less than one hour of experience using virtual reality headsets. It should be noted that the subjects are not the beneficial population of the research. The task and experiment was designed with this in mind and mitigated through training and the simplicity of the task design.

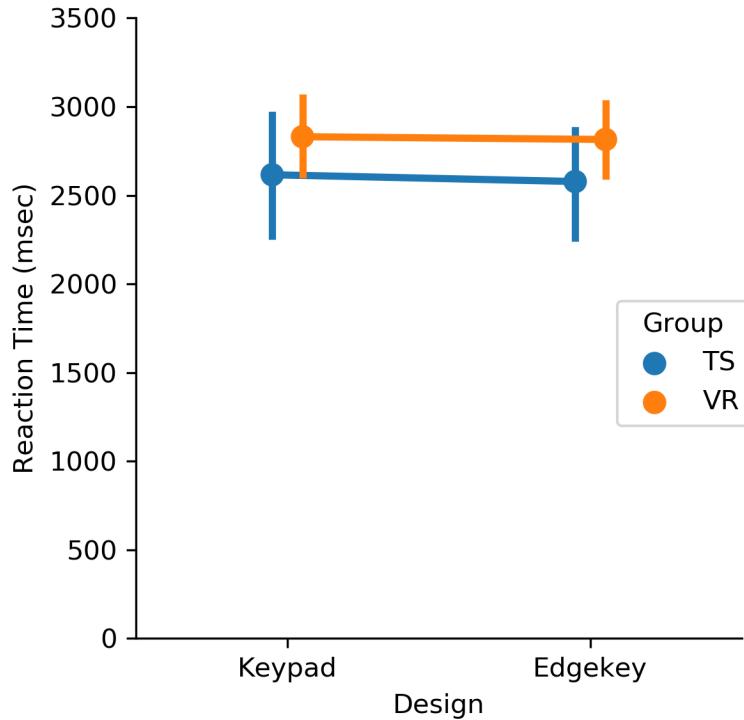
### 5.3.2 Performance Measures

#### Tracking Task RMSE

The performance of the tracking task was measured using the root-mean square error (RMSE) of the pitch. The effect of Group yielded an  $F$  ratio of  $F(1, 21) = 21.4, p < 0.001$  indicating a significant difference between VR ( $M = 1.28\text{deg}, \sigma = 0.38\text{deg}$ ) and TS ( $M = 1.97\text{deg}, \sigma = 0.38\text{deg}$ ). In both groups, subjects were performing the tracking task using the same joystick. The most direct factor that could contribute to the decreased performance in the tracking task for the VR group is the loss of visual acuity in the tracking task display due to the technical limitations of the VR head-mounted display. Indirectly, the additional workload of the prompting task could be taking attention away from the tracking task. The effect of Design indicated a marginally significant difference ( $F(1, 21) = 5.94, p = 0.024$ ) for the tracking task RMSE between Keypad ( $M = 1.57\text{deg}, \sigma = 0.51\text{deg}$ ) and Edgekey ( $M = 1.70\text{deg}, \sigma = 0.52\text{deg}$ ). The only change in the tracking task display between the two instrument designs is a small change in position. It moves from being on the left side for the Keypad to the middle for the Edgekey. Since there was no change otherwise, this suggests that any difference on the tracking task performance between the designs would be related to additional workload from the prompting task. The interaction effect was not significant ( $F(1, 21) = 0.17, p = 0.69$ ).

We can investigate the trials where the subjects were only doing the tracking task to further investigate the change in performance between the two groups. The subjects ran a single trial that was just the tracking task without the prompting task at the end of each evaluation session. These trials were included to be used as a test of the assumption that the subjects were no longer learning, but can also be used as a test of the Group factor on the tracking task performance. The effect of group on RMSE for the tracking-only trials yielded a marginally significant difference ( $F(1, 21) = 4.81, p = 0.039$ ) between the VR Group ( $M = 1.32, \sigma = 0.50$ ) and the TS Group ( $M = 0.91, \sigma = 0.43$ ). There was no significant difference for the effect of design ( $F(1, 21) = 0.068, p = 0.80$ ). The interaction effect between group and design was also not significant ( $F(1, 21) = 3.21, p = 0.087$ ).

Although the tracking only trials found a marginally significant difference for the



**Figure 5.8:** Factor Plot of Response Time

group, the difference was much more distinct for the trials with both tasks. This indicates that when the subjects were focused on the single task, they were able to mitigate most of the visual resolution differences between using a touchscreen and the virtual reality screen. Additionally, the marginally significant difference between the designs for the trials with both tasks was reduced to no significance when the additional prompting task was removed. This also points to the additional workload of the prompting task causing a performance drop on the tracking task. The factors leading to the added workload of the prompting task are investigated in the next performance measures discussed.

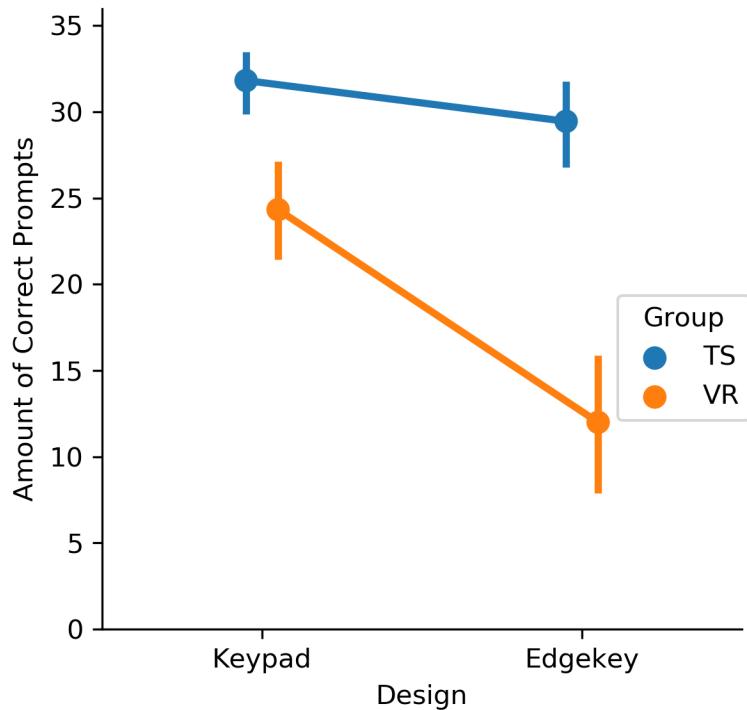
### Prompt Response Time

The first measure of the prompting task is the response time of the subject. The response time is defined as the time from the prompt is shown to each subject until they press the first button of the prompt. For the Edgekey design, it would be possible that the subject had to start with the switching button if the new prompt did not start with the same mode (letters or numbers) as the previous prompt (see Figure 5.6). Since this button

would not clear the prompt when it was pressed, it is not considered the first button of their entry. However, this would still require an additional movement of the subject, adding additional time. For this reason, the prompts which required the subject to start with the switch key are filtered out of this analysis. After filtering, 885 of the total 1700 prompts recorded for the Edgekey design were kept.

The response time was unique among the dependent measures, as all tests were insignificant. The effect of group yielded an  $F$  ratio of  $F(1, 21) = 1.19, p = 0.29$  indicating no significant difference between VR ( $M = 2812\text{msec}, \sigma = 383\text{msec}$ ) and TS ( $M = 2594\text{msec}, \sigma = 567\text{msec}$ ). One factor that could influence the response time between groups is the additional time to activate a button in the VR environment versus the touchscreen. The touchscreen subjects were using a familiar interface for activating the buttons, while the VR subjects needed to activate the button with the virtual hand. However, a large portion of the response time for the subject is their cognitive processing of the prompt – recognizing the new prompt has appeared, reading it, then memorizing it. Beyond potential differences in the visual environment, the cognitive portion should not take more time for one group or the other. A potential reason that there could be a lower than expected difference between the group means is that some VR subjects learned to keep their hand closer to the instrument so that the hand tracker could keep it in view. When the hand tracker lost view of the hand, the re-acquisition time could be significant, so holding it close to the instrument would prevent this from happening. This issue comes up again when looking at the subjects' response to questions about fatigue.

The effect of design was also insignificant ( $F(1, 21) = 0.68, p = 0.42$ ) between Keypad ( $M = 2728\text{msec}, \sigma = 512\text{msec}$ ) and Edgekey ( $M = 2687, \sigma = 471\text{msec}$ ). The biggest difference between the two designs is the switching key on the Edgekey design. As described above, the need for an additional switch press before the first prompt button was filtered out, so we are only comparing prompts where the first button was available right away to the subject. Even though the physical requirements were filtered out, subjects still need to verify that the labels are in the correct state for starting entry. Since the Edgekey design had more time pressure due to the need for the switch key, subjects could have learned to respond quicker to adapt for this. However, these differences in the design did



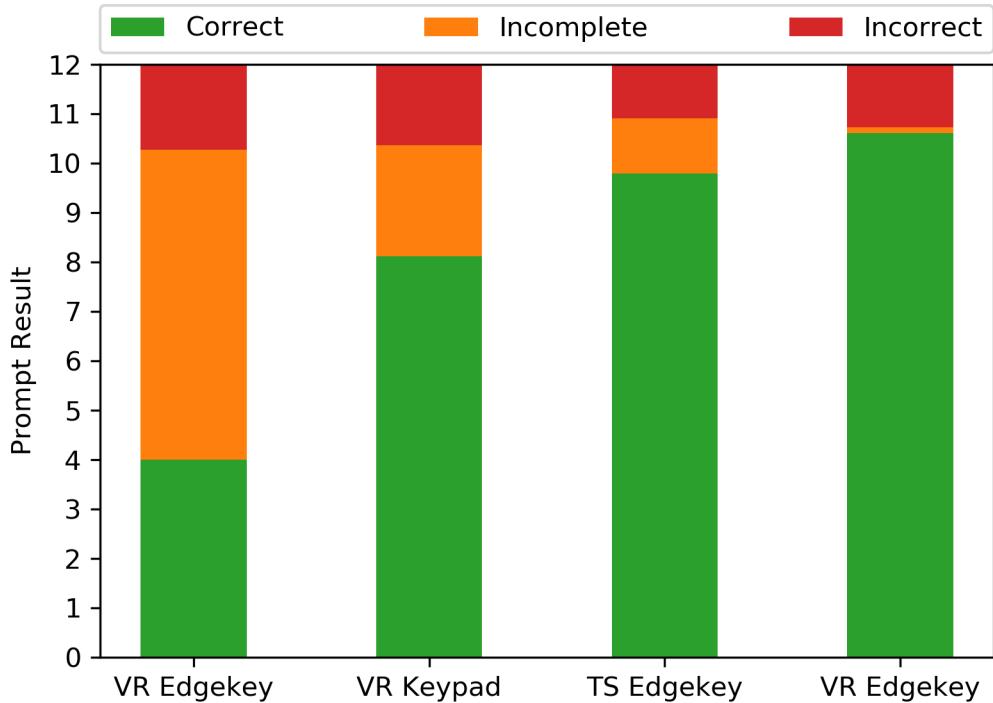
**Figure 5.9:** Factor Plot of Correct Prompts

not appear to have a significant effect on the response time. Finally, the interaction effect was not significant ( $F(1, 21) = 0.001, p = 0.96$ ).

### Prompts Correct

The second measure of the prompting task is the accuracy of the subjects in correctly completing the prompt. To get the prompt correct includes two important components for the subject. First, they must remember the prompt as they enter it, and second, they must be able to physically press the buttons within the seven second response window. For the statistical test, we are using the count of how many prompts each subject completed successfully per trial. Among the incorrect prompts, we can differentiate between whether the subject entered the prompt incorrectly (failure to remember the prompt) or whether the subject ran out of time (failure to physically press the buttons). These counts are reported to help analyze the results, but are not used in the statistical tests. There were 12 prompts per trial, and every subject completed three trials for each design.

The number of correct prompts had a significant interaction effect between group



**Figure 5.10:** Result of prompts

and design ( $F(1, 21) = 27.8, p < 0.001$ ), meaning the main effects must be interpreted with the post-hoc tests as well. Both main effects were significant, the effect of group yielded an  $F$  ratio of  $F(1, 21) = 43.9, p < 0.001$  while the effect of design yielded an  $F$  ratio of  $F(1, 21) = 64.1, p < 0.001$ .

For the effect of design on the VR group, the repeated measured t-test indicated a significant difference ( $t(11) = 8.0, p < 0.001$ ) between the Keypad ( $M = 8.11, \sigma = 1.62$ ) and the Edgekey ( $M = 4.00, \sigma = 2.37$ ). The TS group had a marginally significant difference ( $t(10) = 2.28, p = 0.045$ ) between Keypad ( $M = 10.6, \sigma = 0.96$ ) and the Edgekey ( $M = 9.82, \sigma = 1.38$ ). These results indicate that both groups had trouble with the additional time pressure caused by the Edgekey design requiring the use of the switch key. The TS group performed a lot closer to their performance in the Keypad design, however, only getting approximately 1 fewer prompt correct. The VR group had much more difficulty in the Edgekey design, correctly completing about half as many as they completed in the Keypad design. However, they had more difficulty in both designs compared to the TS group.

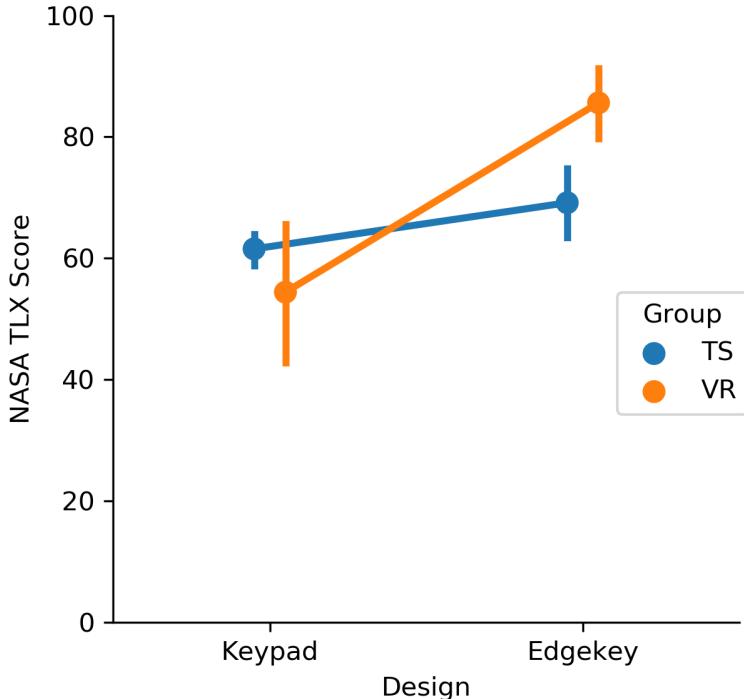
This agrees with the post-hoc tests for differences between groups within each design. These tests had significant effects for both the Keypad design ( $t(21) = 4.44, p < 0.001$ ) between the VR group and the TS group, and the Edgekey design ( $t(21) = 7.05, p < 0.001$ ) between the VR group and the TS group. The main effect of group clearly has a meaningful effect, which found the VR group ( $M = 6.05, \sigma = 2.88$ ) had significantly fewer correct prompts than the TS group ( $M = 10.2, \sigma = 1.2$ ). This difference is largely due to subjects not being able to complete the prompt. Figure 5.10 shows the breakdown of the mean result of each trial for each group and design.

Across all groups and designs, very few prompts were completed that were incorrect, and most of the difference in number completed correctly is due to the incomplete prompts. A contributing factor for this would be the method of button activation used for the VR group combined with the time pressure. Another contribution would be the limitations of the hand tracker. When the hand tracker lost tracking or gave bad information, it became hard or impossible for the subject to activate a button until the hand tracker returned to normal. When this happened in the middle of a prompt, the amount of time it took to recover from the bad tracking would lead to a timeout on the prompt entry, causing an incomplete prompt. The variance of number correct was also much larger in the VR group, which could be caused by some subjects adapting to the unfamiliar VR environment more rapidly.

## NASA TLX

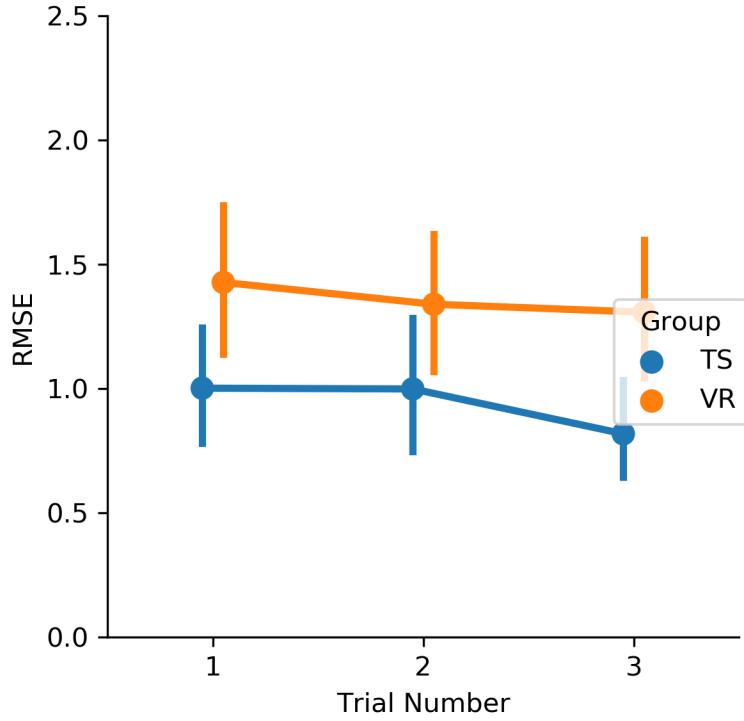
After the subject completed their trials for each design, they filled out a NASA TLX workload survey. Their scores, weighted means by the pairwise comparisons, are used here as a measure of their self-reported workload. The interaction effect between group and design was found to be significant ( $F(1, 21) = 8.25, p < 0.001$ ). The main effects showed a significant difference in design ( $F(1, 21) = 23.6, p < 0.001$ ), but not in group ( $F(1, 21) = 1.69, p = 0.21$ ). This could mean that the group did not affect the TLX score, but in the presence of an interaction effect, the post-hoc tests guide the interpretation.

The repeated measures t-tests indicated significance between designs for the VR group ( $t(11) = -4.20, p = 0.001$ ) between the Keypad design ( $M = 54.4, \sigma = 20.4$ ) and the



**Figure 5.11:** Factor Plot of NASA TLX

Edgekey ( $M = 85.6, \sigma = 11.2$ ) There was a marginally significant difference between designs for the TS group ( $t(10) = -2.72, p = 0.02$ ) between the Keypad design ( $M = 61.5, \sigma = 4.46$ ) and the Edgekey ( $M = 69.2, \sigma = 10.1$ ). The effect of design was much stronger in the VR group, but both groups indicated respectively higher workload on the TLX scores for the Edgekey design. This follows from the experimental design which predicted that the Edgekey design would be more difficult. One factor that could have contributed to a larger difference in scores for the VR group could be the increased difficulty subjects had in completing the prompt, as seen in the results of the number of incorrect and incomplete prompts for the VR group using the Edgekey design (Figure 5.10). The effect of group was not shown to be significant in the ANOVA analysis, but the independent samples t-test showed a significance for the Edgekey design ( $t(21) = 3.69, p < 0.01$ ) between the VR Group ( $M = 85.6, \sigma = 11.2$ ) and the TS Group ( $M = 69.2, \sigma = 10.1$ ). With the Keypad design, The effect of group was not significant ( $t(21) = -1.13, p = 0.27$ ) between VR ( $M = 54.4, \sigma = 20.4$ ) and TS ( $M = 61.5, \sigma = 4.46$ ). These tests further illustrate that the VR group found a higher workload for the Edgekey design specifically, as both groups



**Figure 5.12:** Factor Plot of RMSE for Tracking Only Trials

rated the workload in the Keypad design similarly.

### Tracking Task Learning

Throughout the experiment the subjects did trials with only the tracking task, instead of both the tracking task and the prompting task. Initially, they performed a number of training trials at the beginning with only the tracking task, and then after each evaluation session there was a single trial of just the tracking task. In this section we will test the RMSE of their final training trial and the two after-evaluation trials for any significant learning effects. The trial number is chronological throughout the timeline of the experiment for each subject. This means that due to the counterbalancing, the second and third trial are done with different designs based on the subject. Since the visual environment of the tracking task was quite different for each group, the Group factor is included as a between subjects factor.

The two-way ANOVA found Group to be a marginally significant factor ( $F(1, 21) = 4.94, p = 0.37$ ) between the VR Group ( $M = 1.36, \sigma = 0.51$ ) and the TS Group ( $M =$

$0.94, \sigma = 0.43$ ). The effect of group on the tracking task was already established, so the marginal significance found here is not unexpected. Trial number was found to have no significant effect ( $F(1, 21) = 3.65, p = 0.069$ ) between the three trials. The means of the three trials, in order, are  $1.23^{circ}(\sigma = 0.54^{circ})$ ,  $1.18^{circ}(\sigma = 0.51^{circ})$ , and  $1.07^{circ}(\sigma = 0.51^{circ})$ . Even though the statistical test indicates no significance, the means do decrease as trial number increases. This combined with the large variance suggest that some subjects were experiencing some training effects, but overall the effect of training is not significant. The interaction effect of Group and trial number had no significance ( $F(1, 21) = 0.16, p = 0.69$ ).

## Summary

A summary of the significance results from the ANOVA and post-hoc t-tests for all the performance measures are shown in Table 5.1. The significance is indicated by ‘\*’ for  $p < 0.01$ , ‘+’ for  $0.01 < p < 0.05$ , and ‘-’ for no significance. For the measures with significant interaction effect, the post-hoc t-tests are shown per group and per design.

### 5.3.3 Design Feedback

As discussed in Section subsection 5.2.5, the long-form feedback questions were synthesized and summarized into categories. The categories and the counts of comment occurrence for each group is summarized in Table 5.2. Categories which only received one comment are not included in this table in interest of brevity, the full table is shown in Appendix A.23.

By far the issue that received the most feedback was the difficulty of using the switch key (Edgekey, Switch Difficult). Most of the complaints stated the extra difficulty of having to press another button. Some of the other complaints from this category were: it took extra time (with no extra time given), it added to the mental demands of the task, and it was difficult to see which mode the instrument was in. Both groups disliked the switch key, and mentioned it just as frequently.

“Switching from numbers to letters was hard, especially if I was trying to compensate for turbulence and was struggling at the time.” (TS Subject)

“I did not like how much extra work it was. It took so much extra focus that I forgot I was flying with the joystick” (VR Subject)

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ANOVA			
	Group	Design	Group:Design (Interaction)
Tracking RMSE	*	+	-
Response Time	-	-	-
Prompts Correct	*	*	*
NASA TLX	-	*	*

t-tests				
	Design	Group		
	VR Group	TS Group	Keypad Design	Edgekey Design
Prompts Correct	*	+	*	*
NASA TLX	*	+	-	*

**Table 5.1:** Statistical Significance Test Results. ‘\*’ indicates significance at the  $p < 0.01$  level, ‘+’ indicates marginally significant ( $0.01 < p < 0.05$ ), and ‘-’ indicates no significance.

Many subjects noted the familiarity of the Keypad design (Keypad, Familiar) and that having the buttons close together (Keypad, Buttons Proximal) as things they like about that design. The familiarity was noted more often for the TS Group, but both were some of the more frequent comments within each group.

One comment about the Edgekey design that got more frequent mentions from the TS Group was that they found having the flight task in the middle of the display, centered between the buttons, was preferred (Edgekey, Centered Flight Task Better). The subjects who chose the Edgekey as their preferred design nearly unanimously cited this as their reason for their preference<sup>2</sup>. The comments that fed into this category also included subjects who noted the difficulty of splitting their focus back and forth with the Keypad design. Interestingly, two of the TS Group subjects noted that they would have found the Keypad easier if they had tactile feedback to guide their input. This could suggest that the

<sup>2</sup>The one holdout did not explain why they preferred the Edgekey design.

Topic	Feedback Summary Category	VR Group	TS Group
Edgekey	Switch Difficult	14	12
Keypad	Familiar	6	11
Edgekey	Centered Flight Task Better	3	13
Keypad	Buttons Proximal	6	7
Keypad	Buttons Always Visible	5	5
Other	Hand Tracking Issues	9	0
Edgekey	Hand Blocks View	3	4
Fatigue	Prompting Arm	4	1
Edgekey	Clean Design	3	2
Fatigue	Fatigue from Joystick	0	4
Edgekey	Easier	0	4
Keypad	Buttons Confusable	0	4
Other	Colors Disliked	2	2
Fatigue	Eye Fatigue	3	0
Keypad	Easy Focus Switch	2	1
Keypad	More Mistakes	1	2
Edgekey	Accuracy Worse	1	2
Keypad	Buttons Bad Layout	2	0

**Table 5.2:** Counts of Design Feedback Comments per Group. Sorted by sum of comments.

reason the VR Group subjects did not find the centered flight task advantageous is because with the tactile feedback of the 3D-printed instruments they were able to keep visual focus on the left half of the screen in the Keypad design, thus not seeing benefit from the centering of the flight task display.

“[The Edgekey design] forced me to pay more attention to what I was typing, this wouldn’t have been a problem if the keypad was a physical device that allowed me to locate the numbers and letters without looking, much like the dots on a computer keyboard.” (TS Subject)

“I like that the flight control was cent[e]red, so you could see it even when you were looking at the buttons.” (VR Subject)

The most notable exceptions to providing similar feedback between groups are the categories that relate to fatigue issues. Many subjects in the TS group noted fatigue caused from using the joystick, yet none in the VR group did, despite using the same joystick setup, and seated in the same location. The VR group did note more fatigue in their other arm that was used for the prompting task. This fatigue seemed to be caused by the additional effort needed to have the hand tracker recognize the hand. For example, one subject wrote:

“My right wrist was somewhat fatigued. Though I think this is mostly from positioning my hand for the simulator to recognize my input.” (VR Subject)

Some of this additional effort was due to subjects learning to hold their prompting task hand “hovering” while waiting for the next prompt. This was done to keep the hand in view of the hand tracker as when the hand leaves the field of view, the re-acquisition will slow down the entry of first button. Many subjects organically learned this, and kept their arm in front of the instrument between prompts.

Similar to the fatigue issues being different, there were some comments that were due to the technology being used more-so than the designs themselves. Obviously, the subjects who noted difficulty using the hand tracker, or the one subject who mentioned touchscreen issues, are specific to the simulator technology the used. However, some of the other categories had comments that may have been indirectly caused by the different technologies and their limitations. For example, some subjects noted the keypad design caused them to make more mistakes. For the TS Group, this was due to the touchscreen being too responsive to the button presses:

“[S]ince I was able to go more quickly with this layout, I had more mistakes in the entry.” (TS Subject)

One subject in VR who complained of more mistakes in the Keypad design, identified a common problem caused by the hand tracker. When the hand tracker was having registration issues it would sometimes mistakenly place the other fingers in the activation zone of the buttons underneath the one being targeted, causing multiple buttons to be pressed in a short period of time.

“There’s more unintended register since other fingers might trigger the buttons.”  
(VR Subject)

Although only one subject noted this, it was observed happening to many subjects. In fact, for the VR group, eight of the twelve subjects had the wrong button register within 200 milliseconds of the last button in the Keypad design. In the other designs and groups this happened to only one or two subjects.

## 5.4 Discussion

The motivation of this experiment was to determine the differences between using an R3C simulator system and a traditional simulator system to perform a design evaluation experiment. We had two groups of subjects perform the same evaluation task on two different designs of a cockpit instrument, one group using the R3C system and the other a touchscreen system. The evaluation task included a pitch disturbance tracking task and a call and response prompting task. In addition to the quantitative performance measures of the task, subjects were asked for their feedback on the two designs at the conclusion of the experiment.

The results are summarized using their two independent variables: Group and Design. Group, a between subjects factor, refers to the technology the subject used: either Virtual Reality/R3C (VR) or Touchscreen (TS). Design is a within subjects factor, and is the instrument design the subject was evaluation: Edgekey or Keypad.

The VR Group had worse performance than the TS Group with the RMSE of the tracking task. Subjects from both groups had a marginally significant difference in tracking

task performance due to Design, with subjects performing better with the Keypad design. It was also shown that, on control trials that had only the tracking task (no prompting task), the effect of Group was reduced to marginally significant. The response time of the prompting task had no significant effect based on Group nor Design. Neither of these two previous measures had interaction effects between Group and Design. The number of correct prompts had a significant interaction effect. While the TS Group was able to complete significantly more prompts correctly overall than the VR group (averages of 10.2 vs. 6.1, respectively) the VR group had a significant effect with the Design and the TS group only had marginal significance. This interaction can be clearly seen in the factor plot of correct prompts (Figure 5.9). The NASA TLX workload scores also had an interaction effect between Group and Design. The TLX scores for the VR group had a significant effect in Design, with subjects rating the Edgekey design over 30 points higher than the Keypad design (averages of 54.4 to 85.6, respectively). However, like the number of prompts correct measure, the TLX score was found to be only marginally significant for the TS group, rating the Keypad at 61.5 to the Edgekey's 69.2.

Our results suggest that tasks or performance measures which are dominated by a cognitive portion, such as the prompt response time, provide similar results. Tasks which rely on visual resolution or time pressured responses may not produce the same results between designs using the R3C system. None of the effects reversed slope between designs, however, and the only change is in magnitude of the effect. In fact, for both the number of prompts correct and the workload ratings, which had significant interaction effects, the use of the VR system amplified the effect of design within the groups from a marginally significant effect to a significant effect.

The results of the subjective feedback analysis found that there was no omission of major feedback items on the design of the two instruments from either group. The only feedback comments that did not transfer were the fatigue issues, and of course technology-specific issues. We did discover that some issues were mentioned at differing frequencies, which is to say, one group would have more subjects mention it than the other. These results suggest that the use of the R3C system for receiving feedback from a design would be appropriate.

Many design evaluation studies would be concluded with both paper questionnaires as well as open interviews to receive the feedback from the subject. Our experimental design avoided the use of the interview for two reasons. First, since our subjects were not subject domain experts or experienced evaluators, we wanted to ensure that the prompting of the questions were consistent. Second, the primary goal of the design feedback for this experiment was not to evaluate the designs, but rather to compare evaluations. The use of a proctor interviewing the subjects could introduce accidental bias into the responses of the subjects. This can often be useful when evaluating a new interface, for example, an interviewer could ask subjects about a flaw they had not mentioned yet to determine if they did not notice it or did not care about it. However, in our case, we forgoed this additional information to ensure no bias was introduced in the collection of their opinions.

This was a limited study of the utility of VR/R3C for design evaluation purposes. The task and instrument design was kept simple in nature for this study in order to limit the amount of confounding variables as well as keep it easy to learn for the subject population. Future studies could investigate this system in a more involved design study, with multiple instruments or designs, or more complex behavior in the cockpit. At this point, it would become more essential to use subject domain experts (i.e. experienced pilots) in order to validate these results.

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# Appendices

# Appendix A

## Result Tables

### A.1 Passive Haptics Experiment

Sequence	Haptics	Mean	Std. Dev.
NH First	—	3.877	0.5265
PH First	—	3.99	0.3684
—	NH	3.71	0.3805
—	PH	4.157	0.412
PH First	NH	3.879	0.3892
PH First	PH	4.1	0.3284
NH First	NH	3.541	0.3007
NH First	PH	4.213	0.4934

**Table A.1:** Throughput Means

Factor	F ratio	p value
Sequence	0.5297	0.4761
Haptics	35.53	$1.221 \times 10^{-5}$
Sequence:Haptics	9.062	0.007516

**Table A.2:** Throughput ANOVA

Sequence	Haptics	df	t	p value
PH First	—	9	2.547	0.03135
NH First	—	9	5.503	0.0003787
—	PH	18	-0.602	0.5547
—	NH	18	2.176	0.04309

**Table A.3:** Throughput t-tests

Sequence	Haptics	Mean	Std. Dev.
NH First	—	77.5	10.56
PH First	—	71.2	8.764
—	NH	71	9.701
—	PH	77.7	9.565

**Table A.4:** Presence Score Means

Factor	F ratio	p value
Sequence	4.089	0.05828
Haptics	6.079	0.02396
Sequence:Haptics	0.7164	0.4084

**Table A.5:** Presence Score ANOVA

Haptics	Cronbach's alpha
No Haptics	0.9139
Passive Haptics	0.9191

**Table A.6:** Presence Score Cronbachs alpha

Sequence	Haptics	Mean	Std. Dev.
NH First	—	15.07	2.948
PH First	—	13.1	3.523
—	NH	14.72	2.854
—	PH	13.45	3.762
PH First	NH	14.5	3.064
PH First	PH	11.7	3.529
NH First	NH	14.95	2.773
NH First	PH	15.2	3.259

**Table A.7:** Arm Fatigue Ratings Means

Factor	F ratio	p value
Sequence	2.119	0.1627
Haptics	9.717	0.00595
Sequence:Haptics	13.9	0.001538

**Table A.8:** Arm Fatigue Ratings ANOVA

Sequence	Haptics	df	t	p value
PH First	—	9	-4.583	0.001323
NH First	—	9	0.4596	0.6567
—	PH	18	-2.304	0.03336
—	NH	18	-0.3443	0.7346

**Table A.9:** Arm Fatigue Ratings t-tests

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Please circle the number that best describes the amount of fatigue you feel in your arm. Try to consider the overall sensation of fatigue, soreness, and exertion level. Consider the arm to include shoulder muscles that you use to move the arm.

- 
- 6 No fatigue
  - 7 Very, very lightly fatigued
  - 8
  - 9 Very lightly fatigued
  - 10
  - 11 Lightly fatigued
  - 12
  - 13 Somewhat fatigued
  - 14
  - 15 Heavily fatigued
  - 16
  - 17 Very heavily fatigued
  - 18
  - 19 Very, very heavily fatigued
  - 20 Maximal fatigue
- 

**Table A.10:** Borg RPE Scale as used

## A.2 Design Evaluation Experiment

Group	Design	Mean	Std. Dev.
TS	—	1.277	0.3789
VR	—	1.967	0.378
—	Edgekey	1.704	0.5188
—	Keypad	1.57	0.5068

**Table A.11:** RMSE Means

Factor	F ratio	p value
Group	21.42	0.000145
Design	5.944	0.02374
Group:Design	0.1669	0.687

**Table A.12:** RMSE ANOVA

Group	Design	Mean	Std. Dev.
TS	—	2595	567.3
VR	—	2813	383.2
—	Edgekey	2688	471.1
—	Keypad	2729	512.5

**Table A.13:** Response Time Means

Factor	F ratio	p value
Group	1.199	0.2859
Design	0.6814	0.4184
Group:Design	0.00184	0.9662

**Table A.14:** Response Time ANOVA

Group	Design	Mean	Std. Dev.
TS	—	10.2	1.242
VR	—	6.056	2.889
—	Edgekey	6.768	3.525
—	Keypad	9.304	1.831
VR	Edgekey	4	2.37
VR	Keypad	8.111	1.616
TS	Edgekey	9.788	1.393
TS	Keypad	10.61	0.964

**Table A.15:** Correct Prompts Means

Factor	F ratio	p value
Group	43.56	$1.552 \times 10^{-6}$
Design	63.93	$8.309 \times 10^{-8}$
Group:Design	26.89	$3.872 \times 10^{-5}$

**Table A.16:** Correct Prompts ANOVA

Group	Design	df	t	p value
VR	—	11	8.039	$6.234 \times 10^{-6}$
TS	—	10	2.287	0.04526
—	Keypad	21	4.441	0.0002262
—	Edgekey	21	7.053	$5.839 \times 10^{-7}$

**Table A.17:** Correct Prompts t-tests

Group	Design	Mean	Std. Dev.
TS	—	65.35	8.535
VR	—	70.01	22.65
—	Edgekey	77.74	13.4
—	Keypad	57.83	15.18
VR	Edgekey	85.61	11.21
VR	Keypad	54.42	20.4
TS	Edgekey	69.15	10.06
TS	Keypad	61.55	4.468

**Table A.18:** NASA TLX Means

Factor	F ratio	p value
Group	1.688	0.208
Design	23.57	$8.455 \times 10^{-5}$
Group:Design	8.252	0.009113

**Table A.19:** NASA TLX ANOVA

Group	Design	df	t	p value
VR	—	11	-4.205	0.001474
TS	—	10	-2.718	0.02164
—	Keypad	21	1.132	0.2703
—	Edgekey	21	-3.693	0.001351

**Table A.20:** NASA TLX t-tests

Group	Trial	Mean	Std. Dev.
TS	—	0.9402	0.425
VR	—	1.359	0.5176
—	1	1.225	0.5421
—	2	1.177	0.5144
—	3	1.074	0.5053
VR	1	1.428	0.5754
VR	2	1.34	0.4877
VR	3	1.308	0.5244
TS	1	1.002	0.4223
TS	2	0.9994	0.5038
TS	3	0.8188	0.3488

**Table A.21:** Tracking Only Trials RMSE Means

Factor	F ratio	p value
Group	4.937	0.0374
Trial	3.649	0.06987
Group:Trial	0.1621	0.6913

**Table A.22:** Tracking Only Trials RMSE ANOVA

Topic	Feedback Summary Category	VR Group	TS Group
Edgekey	Accuracy Worse	1	2
Edgekey	Busy Design	0	1
Edgekey	Centered Flight Task Better	3	13
Edgekey	Clean Design	3	2
Edgekey	Easier	0	4
Edgekey	Familiar	1	0
Edgekey	Hand Blocks View	3	4
Edgekey	Labels Easy to Read	1	0
Edgekey	Labels Hard to Read	1	0
Edgekey	Switch Difficult	14	12
Edgekey	Switch Neutral	0	1
Fatigue	Eye Fatigue	3	0
Fatigue	Fatigue from Joystick	0	4
Fatigue	Prompting Arm	4	1
Keypad	Buttons Always Visible	5	5
Keypad	Buttons Bad Layout	2	0
Keypad	Buttons Confusable	0	4
Keypad	Buttons Proximal	6	7
Keypad	Easy Focus Switch	2	1
Keypad	Familiar	6	11
Keypad	Labels Good	1	0
Keypad	Labels Hard to Read	1	0
Keypad	More Mistakes	1	2
Keypad	More Successful	0	1
Keypad	Tracking Easier	1	0
Other	Colors Disliked	2	2
Other	Hand Tracking Issues	9	0
Other	Prompt Location Bad (Both)	1	1
Other	Single Finger Hard	1	1
Other	Touchscreen Issues	0	1

**Table A.23:** Full Feedback Comments by Category.