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Results of a user study on 2D hurricane visualization

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RESULTS OF A USER STUDY ON 2D HURRICANE VISUALIZATION

By

Joel Paul Martin

A Thesis
Submitted to the Faculty of
Mississippi State University
in Partial Fulfillment of the Requirements
for the Degree of Master of Science
in Computer Engineering
in the Department of Electrical and Computer Engineering

Mississippi State, Mississippi

August 2008

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2008

RESULTS OF A USER STUDY ON 2D HURRICANE VISUALIZATION

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We present the results from a user study looking at the ability of observers to mentally integrate wind direction and magnitude over a vector field. The data set chosen for the study is an MM5 (PSU/NCAR Mesoscale Model) simulation of Hurricane Lili over the Gulf of Mexico as it approaches the southeastern United States. Nine observers participated in the study. This study investigates the effect of layering on the observer's ability to determine the magnitude and direction of a vector field. We found a tendency for observers to underestimate the magnitude of the vectors and a counter-clockwise bias when determining the average direction of a vector field. We completed an additional study with two observers to try to uncover the source of the counter-clockwise bias. These results have direct implications to atmospheric scientists, but may also be able to be applied to other fields that use 2D vector fields.

Key words: User study, visualization, weather, Hurricane Lili, 2D vectors, layering

DEDICATION

To Mom, Dad and David.

ACKNOWLEDGMENTS

I would like to thank the members of my committee: Dr. Moorhead for his continued support through this process, Dr. Swan for the tremendous help with setting up, running, and analyzing the user study, and Dr. Jankun-Kelly for never missing an opportunity to ask me how my thesis was progressing.

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Permission to use Figure 2.1 was given by Mahnas Jean Mohammadi-Aragh.

TABLE OF CONTENTS

DEDICATION	ii
ACKNOWLEDGMENTS	iii
LIST OF FIGURES	vi
CHAPTER	
1. INTRODUCTION	1
2. BACKGROUND	2
3. EXPERIMENTAL TASK AND SETTING	5
3.1 Experimental Design and Procedure	5
3.2 GUI and Software	10
3.3 Independent Variables	11
3.3.1 Layers	11
3.3.2 Selection	11
3.3.3 Shapes	11
3.4 Dependent Variables	13
3.4.1 Magnitude Error	13
3.4.2 Direction Error	13
3.4.3 Response Time	14
3.5 Observers	14
4. RESULTS AND DISCUSSION	16
4.1 Wind Magnitude	17
4.2 Wind Direction	23
4.3 Directional Error Study	24
5. CONCLUSIONS AND FUTURE WORK	27
5.1 Future Work	28

REFERENCES 29

APPENDIX

A. INSTITUTIONAL REVIEW BOARD APPROVAL 31

B. CONSENT FORM 33

C. QUESTIONNAIRE 36

LIST OF FIGURES

2.1	Hurricane Lili’s track in Google Earth. Track data is from the National Hurricane Center [7]. Image credit: Mahnas Jean Mohammadi-Aragh. . . .	2
3.1	Point selection circle that the observer would see. The selected point is at the center of the red/white circle.	7
3.2	Area selection box that the observer would see. The selected area is the area included in the black/white box.	7
3.3	The user study GUI in training mode. Note the correct answer on the right side of the screen. This was removed once the test began.	8
3.4	Each of the four basic types of images. Combinations of these four formed the eight image types used in the study. The Images are: (top left) glyphs only (top right) glyphs + state lines (bottom left) glyphs + pressure contours (bottom right) glyphs + color map.	12
4.1	Magnitude Error vs. Observers.	18
4.2	Magnitude Error vs. Question Type, Layers.	19
4.3	Magnitude Error vs. Selection Area Size.	20
4.4	Magnitude Response Time vs. Question Type, Layers.	21
4.5	Magnitude Response Time vs. Selection Area Size.	21
4.6	Magnitude Response Time vs. Contour Lines, Color Map, State Lines for the area questions (360 trials).	22
4.7	Directional Error vs. Observers (880 trials). Observers 10 and 11 saw reversed glyphs (Figure 4.9).	23
4.8	Directional Error vs. Selection Area Size.	24

4.9 The glyph that is produced by RIP is shown on the left. In order to prove that the counter-clockwise bias is caused by the shape of the glyph, we implemented a small study (2 observers) where the glyph shape is flipped, as shown on the right. 25

4.10 Directional Error vs. Glyph Orientation (880 trials). 26

CHAPTER 1

INTRODUCTION

In this paper, we report the design and results of a user study that investigated the current visualization methods used in the weather modeling community and determined their efficacy. The study used model output simulating Hurricane Lili (2002). This study concentrated on the ability of observers to integrate both the magnitude and the direction of vector fields over an area and to determine the effects of layering in a 2D vector field. Much of the work described herein was published in [8].

CHAPTER 2

BACKGROUND

Hurricane Lili began forming on the west coast of Africa on September 16, 2002 and became a hurricane on September 30 over Cayman Brac and Little Cayman Islands. As Lili approached the southeastern United States, it intensified, reaching a maximum wind speed of 125 knots (category four). However, the storm unexpectedly weakened in the 13 hours before landfall, becoming a category two hurricane. Lili made landfall in the United States on October 3 near Intracoastal City, LA with a maximum wind speed of 80 knots [7]. Lili's path as it approached the U.S. can be seen in Figure 2.1.

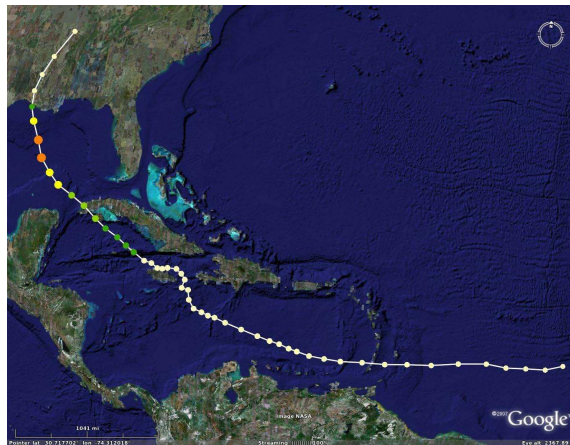


Figure 2.1

Hurricane Lili's track in Google Earth. Track data is from the National Hurricane Center [7]. Image credit: Mahnas Jean Mohammadi-Aragh.

The Hurricane Lili data used in this study were generated by Zhang et al. [12] using the Fifth-Generation NCAR / Penn State Mesoscale Model (MM5). They attribute the rapid weakening of Hurricane Lili to a dry air infusion that can be seen in the MM5 model when output from the *Aqua* Moderate Resolution Imaging Spectroradiometer (MODIS) satellite is included in the model run. To study their simulation results, these domain experts created imagery using Read/Interpolate/Plot Version 4 (RIP) [10], a visual analysis package that is a de facto standard in that community. RIP uses a domain-specific glyph to visualize flow.

The past two decades have seen a wide variety of new visualization techniques. Texture-based methods [6] are able to produce high-resolution and visually pleasing results. However, geometry-based techniques such as streamlines and hedgehogs remain the most common visualization techniques used to investigate real-world flow phenomena [11]. For the most part, there is only anecdotal or assumptive understanding of the effectiveness of these techniques. The visualization research community is beginning to accept that the efficacy of most visualization techniques needs to be verified by user studies [4].

Some flow visualization user studies have been reported. Laidlaw et al. [5] dealt with detecting and identifying critical points. They also studied an advection task in which the observer estimated where a particle in a flow field would move. This was done using several common flow visualization techniques including arrow glyphs and line-integral convolution (LIC). They found that arrows on a regular grid were generally less effective than the other techniques they tested. Andryscio [1] only dealt with the advection of a particle, but included streamlines and pathlines. Also, the observers were asked to advect

a particle back to its origin. Unfortunately, Andrysco was unable to find statistical significance, which was contributed to the test methodology and observers that did not take the tasks seriously.

This weather and flow visualization study is focused on domain-specific 2D glyphs, a simple and intuitive technique still in widespread use. While some weather researchers have started to consider 3D visualization approaches, many continue to rely on 2D methods. In this study, we are interested in investigating the overall performance of these glyphs coupled with one or more layers of scalar data. This will pave the way for future work on other flow visualization techniques.

CHAPTER 3

EXPERIMENTAL TASK AND SETTING

To encourage weather researchers to start using visualization methods that we felt had greater efficacy, we designed a user study to test some potential improvements to their approach. We were particularly interested in how image layering affected an observer's performance. For this study, we defined layering as the number of different techniques (contours, color map, state boundaries) that are added to a base image. Our hypothesis was that as more layers are added, the observer's ability to interpret the base image will be degraded. Additionally, we were interested in an observer's ability to mentally compute the average direction of vectors over an area.

3.1 Experimental Design and Procedure

The prevailing wind magnitude and direction is an important aspect of weather understanding and forecasting. In this experiment, we presented observers with vector fields denoted by glyphs (Figure 3.1) that indicated both wind magnitude and wind direction. While not common in the visualization community, we used the vector glyphs that were used in [12] and are common in the weather modeling community (e.g., of the nine articles in the January 2008 issue of Monthly Weather Review that contained glyphs for wind, seven of them used the glyph style shown in Figure 3.1) [9]. All of the data sets we used

showed the Gulf of Mexico region during the 24 hours prior to the landfall of Hurricane Lili, hence all of the data sets contained a spiral wind field and eye structure (e.g., Figure 3.1). Note this is the period when Lili rapidly weakened from a category four to a category one hurricane on the Saffir-Simpson Scale. We examined observer performance on two different sub-tasks:

- **Magnitude Estimation:** We asked observers to estimate the wind speed at either a point (Figure 3.1) or over a rectangular area (Figure 3.2). The glyphs encoded the wind speed by the number and type of extensions drawn on the right-hand side of the glyph; observers always saw this encoding given in the legend shown in Figure 3.3. In the point case, the observer's task was to estimate the magnitude at the specified location using the glyphs immediately surrounding the marker. In the area case, the observer's task was to average the wind speed of all of the glyphs contained within the selection area.
- **Direction Estimation:** We asked observers to estimate the wind direction at either a point (Figure 3.1) or over a rectangular area (Figure 3.2). The glyphs encoded the wind direction using a weather-vane metaphor: we imagine that we mount the tip of the glyph on a rigidly-fixed, rotating point; the wind catches the 'flags' along the back of the glyph and rotates the glyph to point along with wind direction (Figure 3.3). As with all low pressure systems in the northern hemisphere (hurricanes included), the wind field rotates in a counter-clockwise direction. This can also be seen in Figure 3.3. In the point case, the observer's task was to estimate the wind direction at the specified point using the surrounding glyphs. In the area case, the observer's task was to visually integrate the directions of all of the glyphs contained in the selection area and report the average direction. Note that two glyphs of equal magnitude that point in opposite directions cancel each other out.

The motivation behind these tasks is that it is important for atmospheric scientists to understand the magnitude and direction of wind, both at discrete points, as well as over extended areas. This is different from the tasks presented by Laidlaw et al. and Andryscio because they tested the observer's ability to determine where a single particle would go rather than determining the observer's understanding over an area [1, 5]. While in the most general case atmospheric scientists might be interested in various shapes of extended

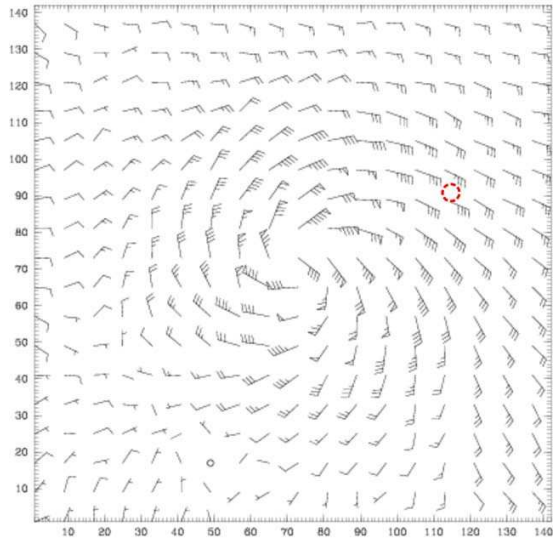


Figure 3.1

Point selection circle that the observer would see. The selected point is at the center of the red/white circle.

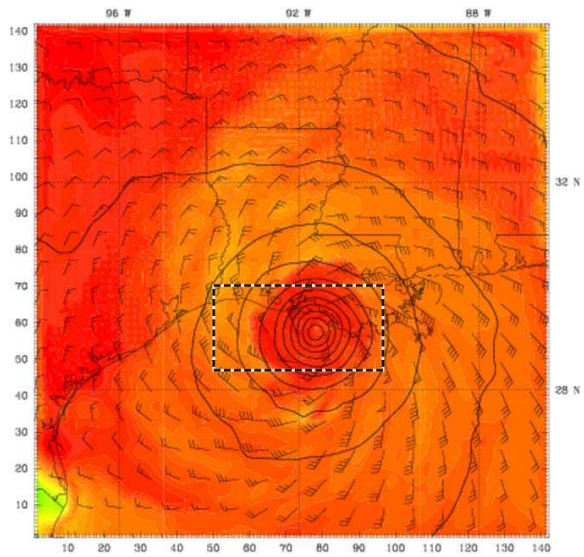


Figure 3.2

Area selection box that the observer would see. The selected area is the area included in the black/white box.

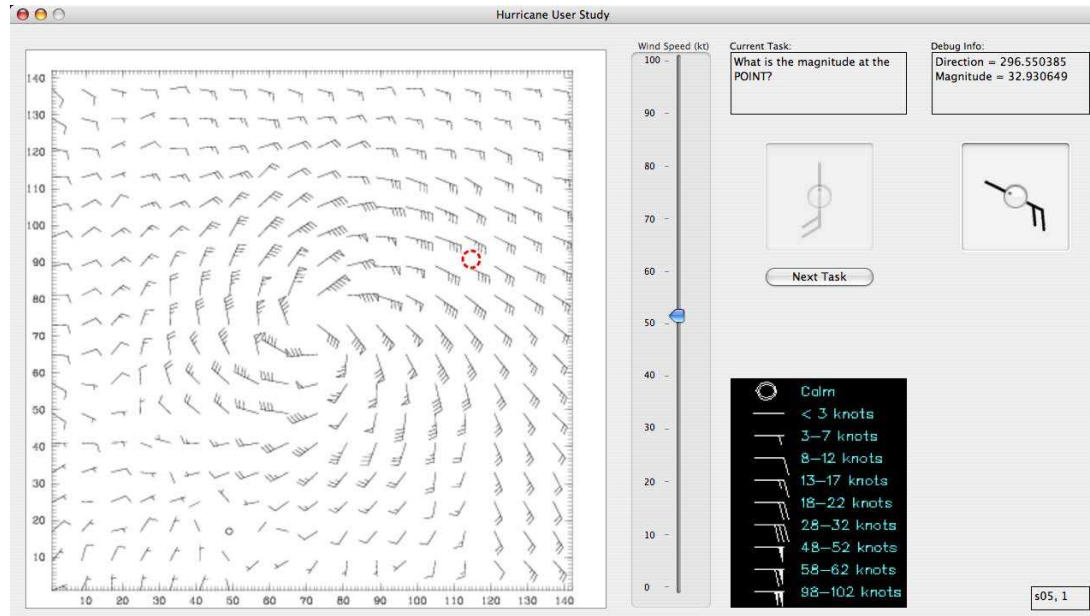


Figure 3.3

The user study GUI in training mode. Note the correct answer on the right side of the screen. This was removed once the test began.

areas, for tractability in this experiment we only examined rectangular areas of certain sizes. Imagery for this study was created using Read/Interpolate/Plot Version 4 (RIP) [10] and was based off the images included in [12].

During the training and test, observers were in a conference room with only the test administrator(s). The observer was asked to sign a consent form and answer a general questionnaire, as well as given an incentive for completing the study (\$10). The observer was also asked to turn off their cell phone or set it to silent.

The test was administered on an Apple PowerBook laptop with a 15.2 inch screen running at the screen's native resolution of 1280x854. Observers were free to move the laptop to a comfortable position. Most observers placed the laptop within an arm's length,

but no specific measurements were taken of this. A Logitech V270 Bluetooth mouse was placed beside the computer. The touchpad was still active for observers that wanted to use it. As many potential distractions were removed from the screen as possible (e.g., the screen saver was turned off, icons removed, etc.).

Observers were given training on the testing software based on a bulleted training script. There were 15 specific points that the administrator covered with the observer during training. The test administrator walked the observer through four training images (two point and two area). These locations and images were randomly selected at the same time the images for the trials were selected. The observer was taught the meaning of the glyphs, instructed on the use of the two sliders (magnitude and direction), and instructed to only use the glyphs in determining their answers (i.e. they were not to use contours, color map or state lines to answer the questions). Observers were given opportunities to ask questions before beginning the test.

The observer then began the test. The test administrator watched the observers closely during the first few questions to see if their answers were reasonable. Some continued to ask questions leading to response time outliers, but magnitude and direction were generally unaffected by this. Had an observer's responses been significantly different than expected, a new training file could have been generated and the test restarted. This, however, was never necessary.

The observer was given an opportunity for a break after every 10 questions. There was no set length for these breaks and the observer could continue at any point. The magnitude and direction the observer entered for each task was recorded, along with the time it took

to complete each subtask. The test administrator recorded notes about events that might affect results (e.g. interruptions or additional questions). Each observer completed 80 tasks, where one task was answering the magnitude and direction for one image.

3.2 GUI and Software

To implement the study, several scripts and a GUI were written. The scripts were written in Perl; the GUI is a Cocoa application. The hurricane images presented to the observer were 607x607 pixels. The full GUI occupies 1269x716 pixels.

To generate the list of tests an observer would encounter, a script randomly selects a data set, time step, and region of interest. The region selected is restricted so that it is not within 100 pixels of the edge of the data (in screen-space). The test conditions (point/area, color map, state lines, pressure contours) are randomly ordered, but balanced so that there is an equal number of each test condition for each observer. A second script determines the ground truth for each selected location by examining the input data for the imagery.

During a test, the GUI (Figure 3.3) displays the pre-determined tests to the observer. It records the magnitude and direction that the observer selects, along with the time it takes the observer to make their decisions. The output is cleaned and formatted before inputting it into the statistics packages Minitab, SPSS, and R. Minitab was used for publication-quality graphics and statistical tests, SPSS for statistical tests, and R for exploratory and preliminary analysis graphics.

3.3 Independent Variables

3.3.1 Layers

All of the images contained wind glyphs. Additionally, the images contained zero to three additional layers which we hypothesized would reduce the observer's ability to determine the magnitude and direction of the glyphs. The layers were pressure contour lines, state boundary lines, and a color map of relative humidity, for a total of eight different layer combinations. Figure 3.4 shows the four base images.

3.3.2 Selection

The selection for each question was either a point or an area (Figures 3.1 and 3.2). For a point, we told observers to use the four vectors surrounding the designated point to determine its value. For an area, we told observers to mentally integrate all of the vectors surrounded by the box. For black and white images, a red/white indicator was used; for color images, a black/white indicator was used. These can also be seen in figures 3.1 and 3.2. These colors were picked because they provided good contrast when they were applied to the image.

3.3.3 Shapes

For area selections, three shapes were chosen: square, horizontal rectangle, and vertical rectangle. Two sizes of each shape were presented to the observer (sizes are in screen-pixels):

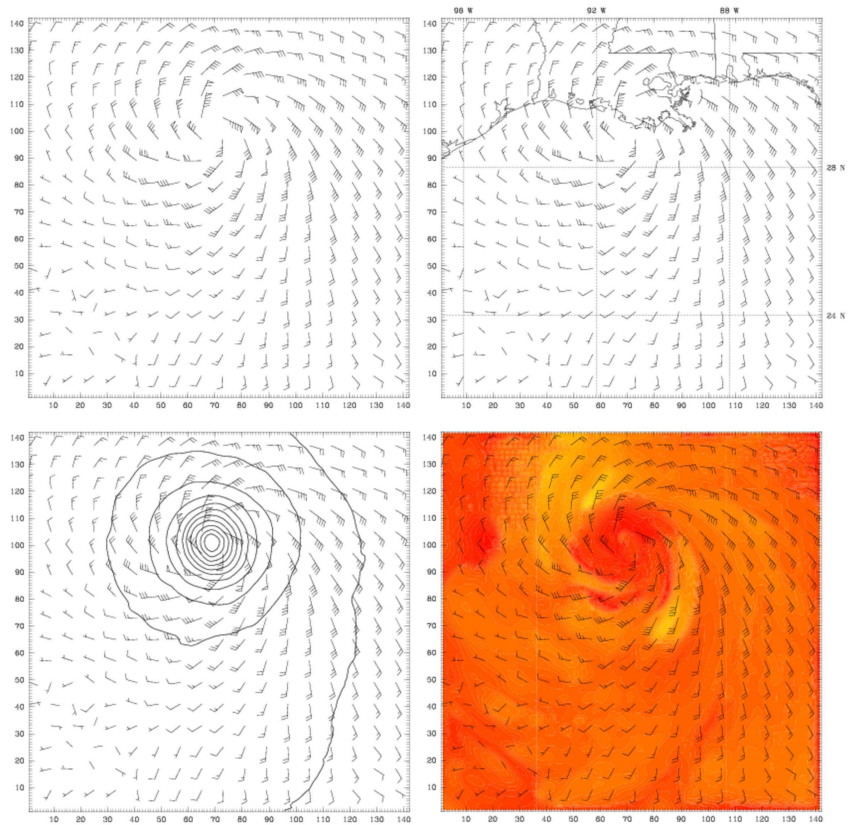


Figure 3.4

Each of the four basic types of images. Combinations of these four formed the eight image types used in the study. The Images are: (top left) glyphs only (top right) glyphs + state lines (bottom left) glyphs + pressure contours (bottom right) glyphs + color map.

- Square: 100x100 and 175x175
- Vertical rectangle: 50x150 and 100x200
- Horizontal rectangle: 150x50 200x100

3.4 Dependent Variables

There were two main measured quantities: wind magnitude (velocity in knots) and wind direction (in degrees). For each, we have a ground truth measurement and the observer's measurement, as well as the observer's response time. From these, we calculated four dependent measures: (1) magnitude error, (2) magnitude error response time, (3) direction error, and (4) direction error response time.

3.4.1 *Magnitude Error*

The magnitude error is computed as the difference in the ground-truth-magnitude and the observer-magnitude. A magnitude error of 0 means the observer has chosen the ground truth wind speed. A positive magnitude error means the observer has overestimated the ground truth wind speed, while a negative magnitude error means the observer has underestimated the ground truth wind speed.

3.4.2 *Direction Error*

The direction is collected in degrees; $0^\circ / 360^\circ$ is due north; degrees are left-handed, meaning they increase clockwise. For processing purposes, the direction is converted to $-180^\circ \leq \text{direction} \leq +180^\circ$, where 0° is still north. The direction error is the difference in the ground-truth direction and the observer-direction. A direction error of 0° means the

observer has chosen the correct ground truth direction. A positive direction error means the observer has overestimated the ground truth direction in a positive left-handed sense; i.e. the observer's direction is clockwise from the ground truth direction. A negative direction error means the observer has underestimated the ground truth direction in a negative left-handed sense; i.e. the observer's direction is counter-clockwise from the ground truth direction.

3.4.3 Response Time

Response time is defined as the number of microseconds from the time the observer is shown the question until the observer presses the 'next task' button. Separate values are collected for both magnitude and direction.

3.5 Observers

Initially, we had planned to recruit observers who were Mississippi State University students and faculty in the Broadcast and Operational Meteorology programs. However, as the study design progressed, we determined that a strong background in weather sciences was not necessary to complete the study. Researchers from various fields commonly look at vector visualizations and should be capable of answering the questions we were asking. As a result, the observers recruited for this study include students and faculty with weather, visualization, or computational fluid dynamics backgrounds. Nine observers participated in the main study, comprising six males and three females. In addition, as described in Chapter 4 (Results and Discussion), we recruited two additional male observers to study

reversed glyphs. Thus a total of eleven observers participated, but unless otherwise specified, all of the results are based on the first nine observers. None of the observers reported color blindness in the questionnaire. Since observers were not interpreting the colors in the images, we did not perform tests to verify their responses.

CHAPTER 4

RESULTS AND DISCUSSION

Each of the nine observers who participated in the main study completed 80 trials, for a total of 720 completed trials. Each trial produced a value for magnitude error, magnitude response time, directional error, and directional response time. We analyzed the data using standard error plots and univariate analysis of variance (ANOVA). For the ANOVA, we modeled our experiment as a repeated-measures design that considers *observer* a random variable and all other independent variables as fixed. The distributions on which ANOVA analysis is based assume that, for each tested effect, the data is normally distributed and the variance is homogenous. Howell says that for repeated-measures designs such as the ones we report here, these two assumptions are usually violated. Therefore, following the recommendations of Howell (p. 486) [3], for each tested effect we applied the Huynh and Feldt correction ϵ ; when the F -test is conducted, the degrees of freedom are multiplied by ϵ . This results in a more conservative test that corrects for the degree to which the ANOVA assumptions are violated. For our analysis, we applied the Huynh and Feldt correction whenever the data was completely balanced (and thus it was possible to calculate ϵ using SPSS). However, some of our F -tests were over unbalanced data, and thus we were not able to calculate this correction for every F -test. Therefore, we do not report Huynh and Feldt-corrected F -tests in this section. However, whenever we were able to calculate the

more conservative Huynh and Feldt correction we did so, and we did not find any cases where the Huynh and Feldt correction changed the outcome of an F -test from significant to non-significant.

We processed outliers in the data with the procedure described by Barnett and Lewis [2]. We determined outliers by examining histograms that summarized each dependent measure; for magnitude error and directional error the histograms showed symmetric normal distributions, while for response times the histograms showed skewed normal distributions. We determined outliers on a case-by-case basis, by examining the tails of the distributions and noting values that appeared after conspicuous gaps in the histogram. Each outlier was replaced by the median of the remaining values in the experimental cell. Given that outliers are considered mistaken values, this procedure improves the calculation of means, standard errors, and the sums-of-squares terms used in ANOVA, which would otherwise be inappropriately influenced by the outlying values.

4.1 Wind Magnitude

In the wind magnitude data, there were 31 outliers that needed to be processed. This is 31 of 720 responses (4.3% of the values). While this is quite a few outliers, it reflects the difficulty of determining the magnitude, particularly over areas. There is negative bias as all observers tended to underestimate wind speed by an average of 4.0 knots; this underestimation is significantly different from zero ($F(1, 8) = 2629, p < .000$). Figure 4.1 shows the magnitude error results for each observer. All observers underestimated the magnitude, from an average of 1.8 knots for observer 2 to an average of 5.7 knots for observer 5.

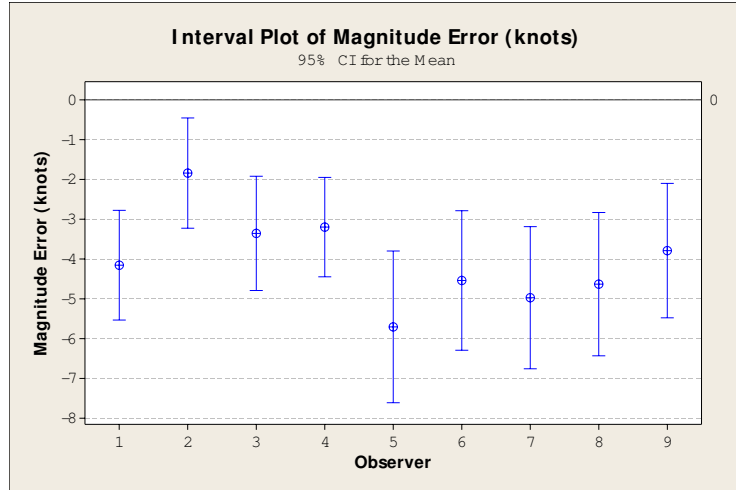


Figure 4.1

Magnitude Error vs. Observers.

Figure 4.2 shows the main results for magnitude error. In the results, we encoded *layers* as a three-digit binary number where the digits indicate, from left to right, the presence or absence of contour lines, color map, and state lines (“1” indicates presence and “0” indicates absence).

Observers were considerably more accurate with points than with areas ($F(1, 8) = 151.2, p < .000$). This is expected because integrating over areas should be more difficult than determining the magnitude at a single point. It also shows that the overall underestimation trend comes mostly from area tasks. Because the standard error bars overlap within both area and point questions, there is no evidence that the number of layers make a systematic difference in observers’ ability to determine the magnitude; analysis showed no effect of layers ($F(7, 56) = .392, p = .903$), nor any interaction with question type ($F(7, 56) = .894, p = .518$). For us, these are somewhat negative results, as they do not

support our initial hypothesis that increasing the number of layers in the imagery would make the task more difficult. It is possible that the specific portion of the data set queried, and perhaps the queried value itself, make a much larger difference than the number of layers.

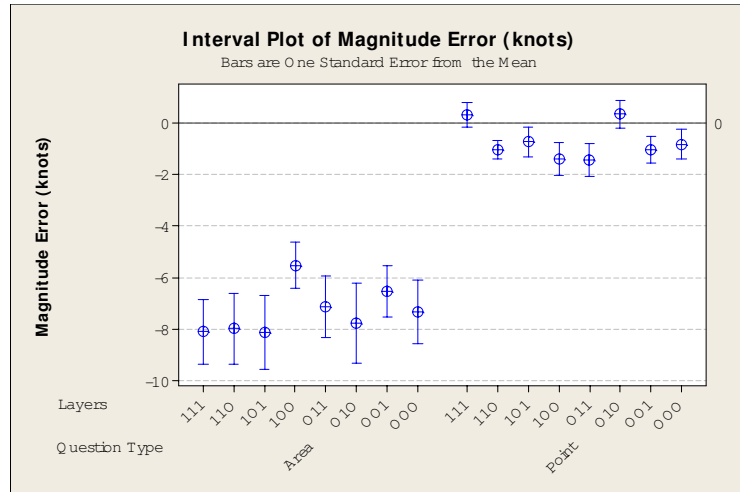


Figure 4.2

Magnitude Error vs. Question Type, Layers.

Figure 4.3 shows that the task gets harder as the size of the selection area increases ($F(2, 16) = 84.5, p < .000$). Observers increasingly underestimate the magnitude for larger squares or rectangles. We did not find any effect of selection area shape; observers gave equivalent results for squares, horizontal rectangles, and vertical rectangles.

We also recoded and analyzed response times for each magnitude trial. Figure 4.4 shows that the average response time for areas was greater than the time for points ($F(1, 8) = 26.8, p = .001$), which is further evidence that the area questions were harder than the

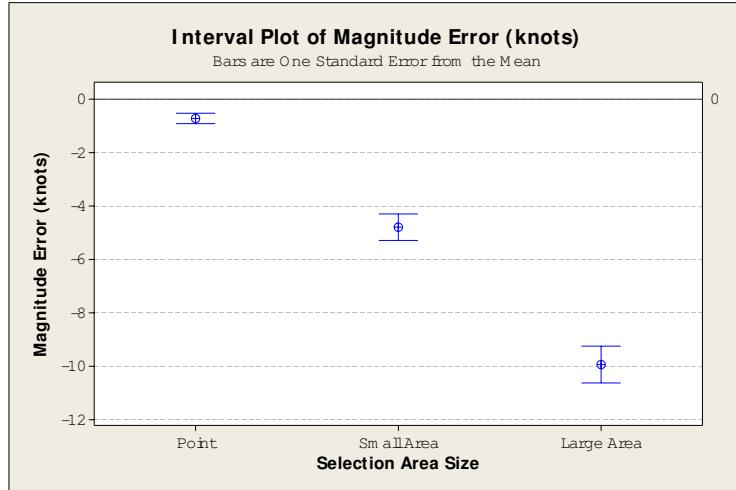


Figure 4.3

Magnitude Error vs. Selection Area Size.

point questions. We found an overall main effect of layer ($F(7, 56) = 2.12, p = .057$), which is further analyzed below. We found a trend towards an interaction between layer and question type ($F(7, 56) = 1.9, p = .090$). Figure 4.5 shows that observers took longer to answer with larger areas ($F(2, 16) = 25.4, p < .000$), which is further evidence that larger areas were more difficult. We found no response time differences for different area shapes.

Figure 4.4 suggests a layer effect for areas, but not for points. Indeed, there was a main effect of layers for the area trials ($F(7, 56) = 2.75, p = .016$). Figure 4.6 shows the response times for the absence (“0”) and presence (“1”) of each layer for the 360 area trials. We found main effects for the presence or absence of contour lines ($F(1, 8) = 12.3, p = .008$), color maps ($F(1, 8) = 4.74, p = .061$), and state lines ($F(1, 8) = 18.1, p = .003$), but no interaction effects. Contrary to our hypothesis, observers were *faster* in the presence

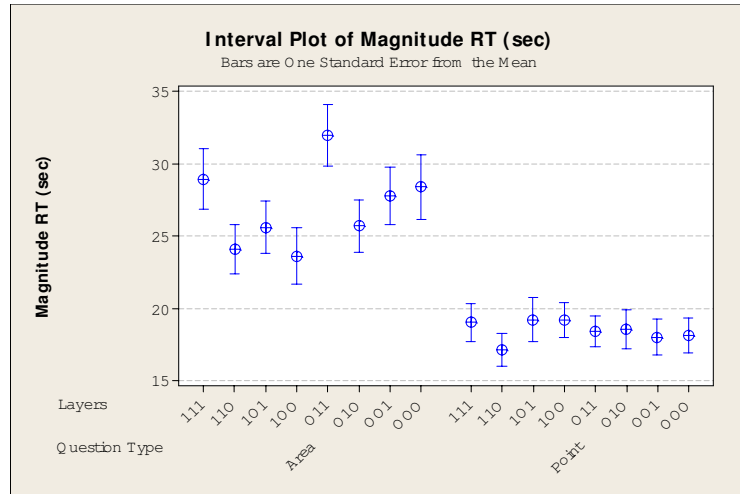


Figure 4.4

Magnitude Response Time vs. Question Type, Layers.

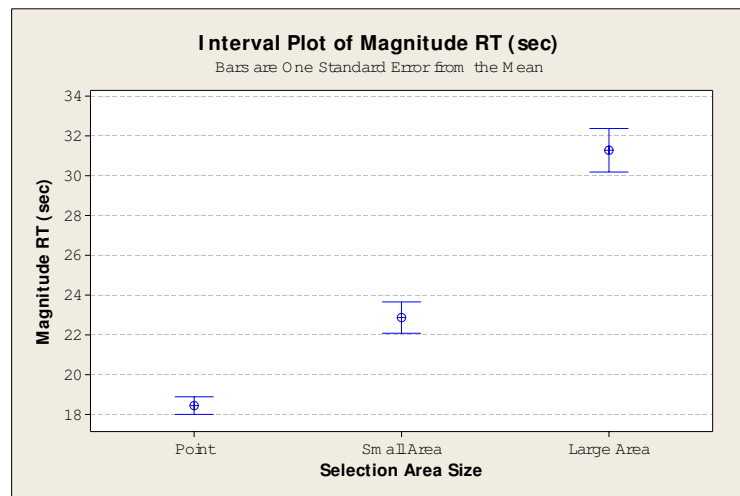


Figure 4.5

Magnitude Response Time vs. Selection Area Size.

of contour lines. We conjecture that the contour lines helped observers estimate the wind speed because they frame the wind fields (Figure 3.4). As we expected, observers were slower when color maps and especially state lines were present, which indicates that these layers made the task more difficult. Interestingly, although color maps are much more visually salient than the state lines (Figure 3.4), the magnitude of the effect for color maps (1.33 seconds) was much smaller than for state lines (3.11 seconds). This may occur because the relative humidity shown by the color maps also follows the wind field, while the state lines are completely arbitrary with respect to the wind field. In addition, note that these results are only for response time — the lack of magnitude error results indicates that observers' accuracy was not effected by the layers.

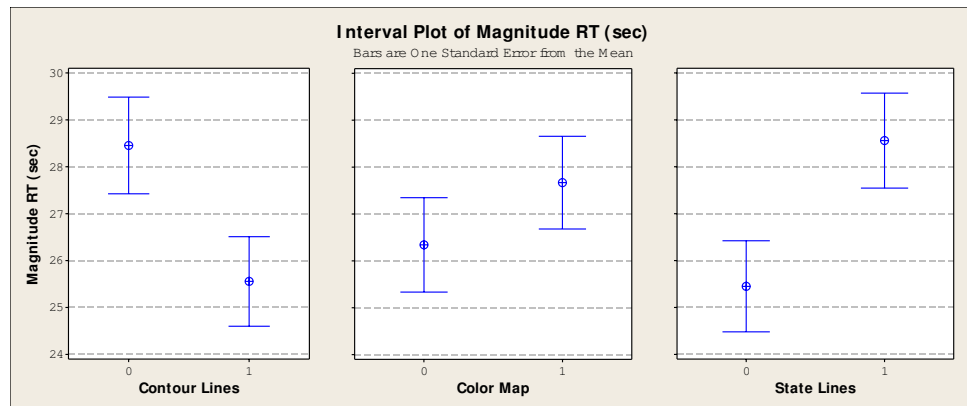


Figure 4.6

Magnitude Response Time vs. Contour Lines, Color Map, State Lines for the area questions (360 trials).

4.2 Wind Direction

The second dependent measure in this study was wind direction. Out of 720 wind direction responses, 27 (2.9%) outliers were removed. The directional error shows a negative (counter-clockwise) bias of -6.1° , which is significantly different from zero ($F(1, 8) = 41.8, p < .000$). Figure 4.7 shows the directional error results for each observer; observers 10 and 11 saw “flipped” glyphs and are discussed in more detail later. All observers showed a negative bias, which ranged from an average of -3.41° for observer 2 to an average of -10.84° for observer 8.

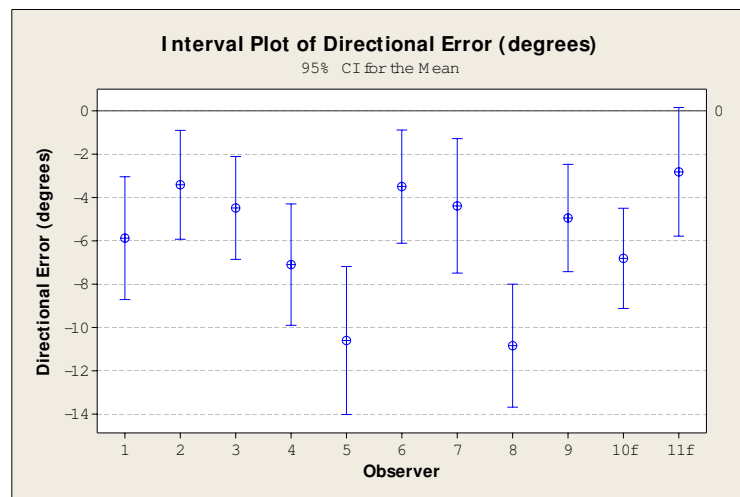


Figure 4.7

Directional Error vs. Observers (880 trials). Observers 10 and 11 saw reversed glyphs (Figure 4.9).

Figure 4.8 shows that the magnitude of the negative bias increased as the size of the shape increased ($F(2, 16) = 16.9, p < .000$). We did not find any effects of area shape on directional error.

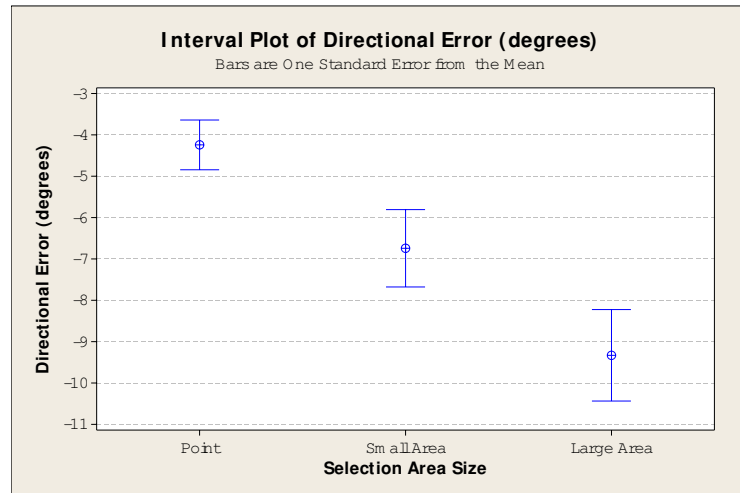


Figure 4.8

Directional Error vs. Selection Area Size.

We also analyzed the response time for directional error. Our only response time finding was that it took observers longer to enter the direction for area questions (11.3 seconds) than for point questions (10.0 seconds) ($F(1, 8) = 6.08, p = .039$).

4.3 Directional Error Study

Considering that these glyphs are commonly used by the weather science community, it is interesting that observers consistently (1) underestimated the wind magnitude shown by the glyphs, and (2) showed a consistent counter-clockwise bias when estimating wind

direction. We hypothesized that the counter-clockwise bias was likely due to the glyphs' asymmetric visual design, where the wind-speed flags are always on the left (Figure 4.9). To quickly test this hypothesis, we flipped the glyph orientation (Figure 4.9), and ran two additional observers through exactly the same protocol. We expected that this would give us a clockwise bias with a similar magnitude as the RIP's regular glyphs did. Figures 4.7 and 4.10 show the results. Contrary to our hypothesis, our two observers still displayed a counter-clockwise bias with the flipped glyphs, and the magnitude of the bias is comparable to what we found with our first nine observers. The difference between the groups is not significant ($F(1, 878) = 1.41, p = .235$). Further study will be required to determine what, if any, role glyph shape plays in this directional bias.

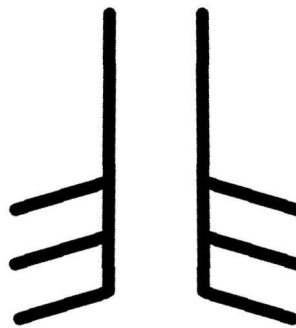


Figure 4.9

The glyph that is produced by RIP is shown on the left. In order to prove that the counter-clockwise bias is caused by the shape of the glyph, we implemented a small study (2 observers) where the glyph shape is flipped, as shown on the right.

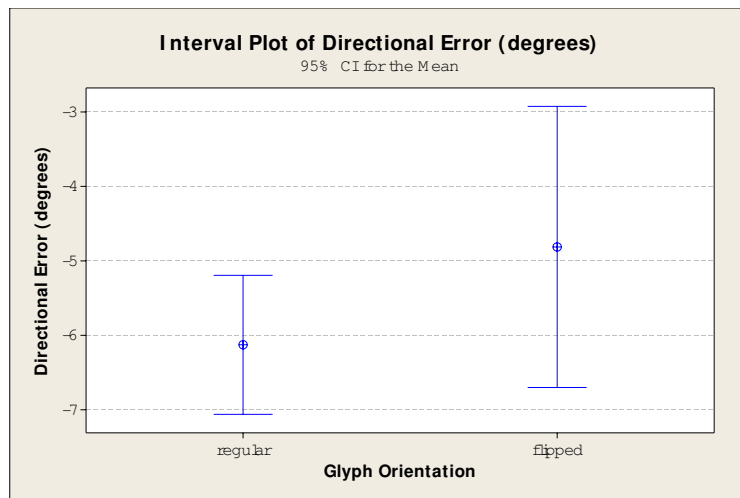


Figure 4.10

Directional Error vs. Glyph Orientation (880 trials).

CHAPTER 5

CONCLUSIONS AND FUTURE WORK

Our experiment has empirically verified some expected results, as well as revealed some surprising and unexpected results. Among the expected results is that determining the magnitude and direction of wind speed over an area is a harder task than at a point. Another unsurprising result is that the larger the area that needs to be mentally integrated, the harder the task and the longer it takes.

This experiment revealed three surprising findings. First, although some data layers (state lines) increased the response time for wind speed estimation over areas by as much as 3.11 seconds, overall we did not find that additional data layers made the glyphs more difficult to read. Second, observers underestimated wind speed by an average 4.0 knots; and their underestimation became worse as the area over which they were estimating increased. Third, observers showed an average counter-clockwise bias of -6.1° in wind direction, and this bias also became worse as the area over which they were estimating increased. The wind direction bias cannot be entirely explained by the asymmetric nature of the glyph shape.

As discussed in the introduction, the glyphs studied here are widely used in the weather science community. Our unexpected positive finding is that these glyphs can be reliably read in the presence of additional data layers. However, our unexpected negative find-

ing is that observers reliably underestimated wind speed and showed a bias in estimating wind direction. Because understanding wind speed and direction are fundamental tasks in weather data analysis, this is a potentially serious finding.

5.1 Future Work

It is well-known in the visualization field that the design space for any glyph-based vector field visualization technique is very large. A fruitful goal for future work would be to study related glyph techniques to see if they exhibit the same sorts of biases we found here. Another fruitful goal would be to tweak the parameters of the current glyphs, which are widely used, to see if they can be improved. For example, perhaps the wind speed underestimation could be reduced if the flags that denote wind speed were made larger (Figure 3.3).

Finally, while a strength of the current study is that it used real-world data, it is possible that the nature of the dataset influenced the results. In particular, we suspect that the directional bias might have arisen because every dataset showed a counter-clockwise circular wind field around a northern hemisphere tropical cyclone. We would like to repeat the study with a southern hemisphere tropical cyclone data set, where the wind field would rotate in a clockwise direction.

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APPENDIX A
INSTITUTIONAL REVIEW BOARD APPROVAL



**CONTINUING REVIEW
REPORT FORM**

RECEIVED
SEP 14 2007

PI Name: Robert Moorhead
Address: GRI
MS 9627
Advisor:

IRB Docket #06-239
Project Title: HPCVI Hurricane Visualization
User Study
Expiration date: 10/15/07

Funding status: Awarded List agency: DoD
 Pending Not awarded No cost study
Other: award expired; still running subjects

1. Number of subjects involved in the project (to date, if ongoing): 5

2. Were there any adverse events or unanticipated problems?

YES, Provide a detailed statement:
 NO

3. Number of subjects who withdrew from the project: 0

4. Were there any complaints regarding the project?

YES, Provide a detailed statement: The original experiment was too long (75 minutes), so we adjusted it such that it is now only about 50 minutes in most cases.
 NO

5. Is there any new information since the last IRB review that might impact the Board's understanding of the risks or benefits?

YES, Provide a detailed statement:
 NO

6. Has the project been modified since the last IRB review?

YES, Have all changes been submitted for IRB approval: YES NO
 NO

7. Are you still collecting data? YES NO

8. Are your remaining activities confined to data analysis? YES NO

9. Projected end of project (data analysis complete): 3/31/08

10. Please attach a current consent form.

Robert Moorhead
Principal Investigator

9/13/07
Date

Research Advisor (if applicable)

Date

OFFICE USE ONLY	
<input type="checkbox"/> Administrative	Approved: <u>Christine Wall</u> Date <u>9-26-07</u>
<input checked="" type="checkbox"/> Expedited	
<input type="checkbox"/> Full Board	New expiration date: <u>10/15/08</u>

965 ✓

APPENDIX B
CONSENT FORM

Consent Form

Title of Study: HPCVI Hurricane Visualization User Study

Study Site: HPC² – Room: TBD

Name of Researcher(s) & University affiliation:

Dr. Robert J. Moorhead	Mississippi State University rjm@gri.msstate.edu , 662-325-2850
Dr. Ed Swan	Mississippi State University swan@cse.msstate.edu , 662-325-7507
Joel Martin	Mississippi State University jmartin@gri.msstate.edu , 662-325-8804

What is the purpose of this research project?

The purpose of this research is to compare different visualization methods of different aspects of hurricanes such as wind velocity, mixing ratio, and rainfall in order to create a catalogue of the best ways to visualize hurricanes.

How will the research be conducted?

For this experiment, you will be asked to watch short video clips of different hurricane visualizations on a computer and answer a short list of questions about the hurricanes. The accuracy of your answers will be used to judge the effectiveness of the different visualizations. We expect that the whole process will take no more than 45 minutes to complete. If you have further questions regarding the procedure, please ask one of the researchers. However, answers to questions that may affect the results of the experiment will be deferred until after you have completed the procedure.

Are there any risks or discomforts to me because of my participation?

The risks involved in this experiment are no more than those associated with typical computer use. It is also important to note that your academic rating will not be affected in any way by your participation or nonparticipation.

Does participation in this research provide any benefits to others or myself?

This experiment may improve meteorologists' ability to understand and communicate data about hurricanes, which will make predicting the behavior of hurricanes more accurate. You will benefit from this experiment from exposure to the different visualization and the meteorological phenomenon that is being visualized.

For your participation you will be paid a sum of \$10. You can discontinue or withdraw your consent to participate in this experiment at any time. You will not be coerced in any other way to participate in this experiment.

Will this information be kept confidential?

The data collected will be analyzed and the results may be published. The data and all related information will not be associated with your name or identity outside the context of the study or in any published results. In particular, data will be indexed by number; at no time will your name be associated in any way with this number on any computer system. We will keep only one written list linking names and numbers, and this list will not be put in computerized format, nor will it be revealed to anybody outside the study. The consent forms and the list linking the names with an id number will be stored in a locked filing cabinet. The filing cabinet will be in Room A209 of the HPC2 during the data collection. The confidential material will be moved to Room 207A of Simrall Hall after the data is collected.

The information will not be disclosed to anyone other than the researchers listed above until it is made anonymous. If you have any questions about confidentiality, please ask one of the researchers.

*Also, please note that these records will be held by a state entity and therefore are subject to disclosure if required by law.

Who do I contact with research questions? If you should have any questions about this research project, please feel free to contact any of the 3 researchers.

*For additional information regarding your rights as a research subject, please feel free to contact the MSU Regulatory Compliance Office at 662-325-5220.

What if I do not want to participate?

Please understand that your **participation is voluntary**, your **refusal to participate will involve no penalty or loss** of benefits to which you are otherwise entitled, and you **may discontinue your participation** at any time without penalty or loss of benefits. You are not required to answer any specific questions asked of you during the study.

You will be given a copy of this form for your records.

Statement of Consent

I have read the above information. I have had the opportunity to ask questions, and they have been answered. I understand that I do not have to fill out this form and that I do not have to participate in this research. I voluntarily consent to participate in the research. I am at least 18 years old.

Participant: _____ Signature: _____ Date: _____

Investigator Signature: _____ Date: _____

APPENDIX C
QUESTIONNAIRE

