Application of Visual Cues on 3D Dynamic Visualizations for Engineering Technology Students and Effects on Spatial Visualization Ability: A Quasi-Experimental Study

Petros Katsioloudis, Vukica Jovanovic, and Mildred Jones
Old Dominion University

Abstract

Several theorists believe that different types of visual cues influence cognition and behavior through learned associations; however, research provides inconsistent results. Considering this, a quasi-experimental study was done to determine if there are significant positive effects of visual cues (color blue) and to identify if a positive increase in spatial visualization ability for students in engineering technology courses is observed. According to the results of this study it is suggested that the use of the specific visual cue (color blue) provides no statistically significant higher scores versus the treatment that did not utilize any visual cues.

Introduction

There are several reasons for exploring the potential of color information and its effects on improving spatial visualization ability. Color is one of the fundamental properties of objects and is detected preattentively with other primary properties like brightness and line orientation (Enns & Rensink, 1991; Treisman, 1986). Even though the role of color in object constancy and depth perception is clear, the value of adding redundant color as spatial stimuli has attracted very little attention (Alington, Leaf & Monaghan, 2001). According to Mehta and Zhu (2009) a large amount of research has been done in this domain; however, the psychological processes through which color operates have not been fully explored. As a result, the field has observed certain conflicting results. To add to the related body of knowledge the following study was conducted.

The following was the primary research question:

Is there a difference in spatial visualization ability, as measured through technical drawings, among the impacts of visual cues (adding blue color) on dynamic visualizations for engineering technology students?

The following hypotheses will be analyzed in an attempt to find a solution to the research question:

H₀: There is no difference in spatial visualization ability, as measured through technical drawings, among the impacts of visual cues (adding blue color) on dynamic visualizations for engineering technology students.

H₁: There is an identifiable difference in spatial visualization ability, as measured through technical drawings, among the impacts of visual cues
(adding blue color) on dynamic visualizations for engineering technology students.

Review of Literature

Spatial Ability

According to Hegarty and Waller (2005), spatial ability is a collection of cognitive skills that allow the learner to relate within his/her environment. Developed through spatial cognition, spatial ability can be described as the ability to form and retain mental representations of a stimulus, or mental model, and can be used to see if mental manipulation is possible (Carroll, 1993; Höffler, 2010). This ability has been recognized as an individual ability, somewhat autonomous of general intelligence (Höffler, 2010). According to several studies, it has been suggested that individuals with higher spatial abilities have a wider range of strategies to solve spatial tasks (Gages, 1994; Lajoie, 2003; Orde, 1997; Pak, 2001). Spatial abilities, specifically visualization, play a critical role in the success of a variety of professions, such as engineering, and other technical, mathematical, and scientific professions.

Spatial Ability used in Engineering Education

According to Contero, Company, Saorín & Naya (2006) shifting from a teacher-centered to a student-centered education paradigm in engineering education requires teachers to put an emphasis on spatial reasoning. Known as a critical engineering skill, spatial ability has been identified as having a positive correlation with learning achievements in engineering education (Mayer & Sims, 1994; Mayer, Mautone & Prothero, 2002). Ferguson (1992) defines engineering drawings as the process where a concept is taken from a learner’s mind and articulated through drawings to another person’s mind, thus transferring an object from a 2D to a 3D representation of the object. These physical object manipulations, done through freehand sketching on paper and/or computer-aided sketching, can improve the spatial ability of freshmen engineering students (Martín-Gutiérrez, et al., 2010). In engineering courses descriptive geometry, orthographic views, and three-dimensional modeling have all been employed as a means to improve learners’ spatial abilities (Martín-Gutiérrez, Gil, Contero & Saorín, 2013). More general Spatial Visualization encompasses the mental alteration of an object through a sequence of adjustments. It is considered a key factor in the success of engineering students (Ferguson, Ball, McDaniel, & Anderson, 2008).
Spatial Visualization

Spatial visualization can be defined as “the ability to mentally manipulate, rotate, twist or invert a pictorially presented stimulus object” (McGee, 1979, p. 893). Strong & Smith (2001) suggest a definition of spatial visualization as “the ability to manipulate an object in an imaginary 3-D space and create a representation of the object from a new viewpoint” (p. 2). Researchers in engineering education, the U.S. Department of Labor, as well as major industry representatives have called for the improvement of spatial visualization ability in engineering and technology students (Ferguson, et al., 2008). Over the past two decades there has been an increased sense of urgency on spatial visualization as a primary focus in engineering education, as reported in journal articles and conference proceedings (Marunic & Glazar, 2013; Miller & Bertoline, 1991). In a recent research study, Branoff & Dobelis (2012) discovered a relationship between reading engineering drawings and visualization ability. Sorby & Baartmans's (2000) research on an introductory course, constructed to enhance 3D spatial visualization skills, revealed statistically significant gains in scores and higher retention in first-year engineering students than those who did not take the course. In the matter of engineering student retention, research suggests positive correlations between spatial visualization ability and the retention and completion of degree requirements for engineering and technology students (Brus, Zhoa & Jessop, 2004; Sorby, 2001). In conjunction with the positive correlations related to retention for engineering students (Brus, et al., 2004; Sorby, 2001), several studies suggest that dynamic visualizations, as opposed to static visualizations, have more benefit for students with advanced spatial skills, such as engineering students (Huk, 2006; Lewalter, 2003). Wu & Shah (2004) suggest that dynamic visualizations and 3D animations offer an environment that supports a learner’s incomplete mental model.

Dynamic Visualizations

Today, with the introduction of computer-based design tools (CAD), dynamic visuals are used in place of, or in addition to, static visuals, such as pictures. Research suggests that dynamic visualizations enhance the learning process for learners with high spatial ability (Huk, 2006; Lewalter, 2003). Research has suggested that dynamic visualizations in learning may improve spatial ability in learners with low spatial ability, and may, in fact, have a compensating effect for the low spatial ability learners (Hegarty & Kriz, 2008; Höffler, 2010; Huk, 2006; Mayer & Sims, 1994). Hegarty and Kriz (2008) suggest that dynamic visualizations act as a “cognitive prosthetic” for learners possessing low spatial ability. Hays (1996) found a statistically significant interaction of spatial ability with learners who possess low spatial ability. In this study, the learners receiving animations made greater gains than those receiving no animations.
Visual Cues and Color

Cuing may also enhance a learner’s experience when related to visualizations and text that allow the learner to integrate representations resulting in a deeper understanding of the representative content (Mayer, 2009). Kühl, Scheiter, and Gerjets, (2012) found that cuing significantly increased a learner’s recall in spatial visualization. Lambert, Roser, Wells, and Heffer (2006) found that cuing resulted in rapid orienting by peripheral onsets, as well as target location and specific features, such as color. According to Seddon and Shubber (1984), color in spatial ability, specifically rotation, may assist learners with following the path taken by each part of the structure during rotation. In a research study of color influence on visual memory, Borges, Stepnowsky, and Holt (1977) found that recognition memory in subjects was 5-10% better for colored images than the black and white versions of the same images. Alington, et al. (2001) study suggests that color improved performance in men and women in relation to spatial visualization. Color theorists believe that color influences cognition and behavior through learned associations (Elliot, Maier, Moller, Friedman, & Meinhardt, 2007). However, research provides inconsistent results when using visual cues like color (Seddon & Shubber, 1985). For example the amount of color may have an effect on the results when comparing color versus monochrome. Too much color, however, may have an adverse effect on the subjects when comparing color versus monochrome (Seddon & Shubber, 1985).

For this specific study, the authors had to decide which color to use for the visual cue treatment groups ($n_2 \& n_3$). Previous research has suggested that red and blue have different associations within the cognitive domain. More specifically, in a study conducted by Mehta and Zhu (2009), the colors red and blue were compared towards effects on cognitive task performance. Red is often associated with dangers and mistakes [e.g., errors that are circled with a red ink pen, stop signs, and warnings] (Elliot et al., 2007). In disparity, blue is often associated with openness, peace, and tranquility [e.g., ocean and sky] (Kaya & Epps, 2004). In addition, a study conducted by Elliot et al., (2007), revealed that significantly more participants chose the blue (66%) versus red (34%) color when the task was described to be creative [$\chi^2 (1) = 7.12, P < 0.01$]. The same pattern of results emerged when the task was described to be detail-oriented, i.e., more people thought the blue (74%) versus red (26%) background color would enhance their performance even on the detail-oriented task [$\chi^2 (1) = 15.06, P < 0.001$], (Elliot et al., 2007). For this specific study, we chose the color blue.

Methodology

A quasi-experimental study was selected as a means to perform the comparative analysis of spatial visualization ability during the Fall of 2014. The study was conducted in an engineering graphics course, MET 120 (Computer
Aided Drafting), as a part of the Engineering Technology program. The participants from the study are shown in Figure 1. Using a convenience sample, there was a near equal distribution of the participants between the three groups.

![Figure 1. Research Design Methodology](image)

The engineering graphics course emphasized hands on practice using 3D AutoCAD software in the computer lab, along with the various methods of editing, manipulation, visualization, and presentation of technical drawings. In addition, the course included the basic principles of engineering drawing/hand sketching, dimensions, and tolerance principles.

The students attending the course during the Fall Semester of 2014 were divided into three groups. The three groups \((n1=24, n2=21\) and \(n3=22\), with an overall population of \(N = 67\) were presented with a visual representation of an object (visualization) and were asked to create a sectional view. The first group \((n1)\) received a dynamic 3D printed dodecahedron visualization, self rotated at 360 degrees on the top of a motorized base at about 4 rounds per minute (slow rotation was used to prevent optical illusion and distortion of the original shape) during the creation of the sectional view (see Figure 2). The second group \((n2)\) received the same dynamic 3D printed dodecahedron visualization, also self rotated at about 4 rounds per minute at 360 degrees on the top a motorized base at about 4 rounds per minute with students wearing blue glasses (see Figure 3); thus, it created a blue background around the visualization during the creation of the sectional view. The third group \((n3)\) received a blue, shaded PC developed, dynamic 3D dodecahedron visualization, also self rotated at about 4 rounds per minute at 360 degrees at about 4 rounds per minute (see Figure 4). Since color was used as a part of the study treatment, and to prevent bias with color blind students, all participants were presented with a power point slide that had three color filled circles (red, blue and yellow) and were asked to report on a piece of
paper the three colors. No students were identified as color blind since everyone stated the correct colors.

Figure 2. Dodecahedron 3D Printed Dynamic Visualization

Figure 3. Blue glasses treatment used for Group 2
In addition, all groups were asked to complete the Mental Cutting Test (MCT) (CEEB, 1939) instrument 2 days prior to the completion of the sectional view drawing in order to identify the level of visual ability and show equality between the three groups. The MCT was not used to account for spatial visualization skills in this study. Its only purpose was to establish a near to equal group dynamic based on visual ability, as it relates to Mental Cutting ability. According to Nemeth and Hoffman (2006), the MCT (CEEB, 1939) has been widely used in all age groups, making it a good choice for a well-rounded visual ability test. The Standard MCT consists of 25 problems. The Mental Cutting Test is a sub-set of the CEEB Special Aptitude Test in Spatial Relations, and has also been used by Suzuki (2004) to measure spatial abilities in relation to graphics curricula (Tsutsumi, 2004).

As part of the MCT test, subjects are given a perspective drawing of a test solid, which is to be cut with a hypothetical cutting plane. Subjects are then asked to choose one correct cross section from among 5 alternatives. There are two categories of problems in the test (Tsutsumi, 2004). Those of the first category are called pattern recognition problems, in which the correct answer is determined by identifying only the pattern of the section. The others are called quantity problems, or dimension specification problems, in which the correct answer is determined by identifying, not only the correct pattern, but also the
quantity in the section (e.g. the length of the edges or the angles between the edges) (Tsutsumi, 2004).

Upon completion of the MCT, the instructor of the course placed two identical models of the dynamic 3D dodecahedron for groups $n_1$ & $n_2$ in a central location in two different classrooms ($n_2$ also received blue glasses). The instructor also projected the dynamic 3D PC generated visualization in a third room, where the three groups were asked to create a sectional view of the dodecahedron (see Figure 5). This process takes into consideration that research indicates a learner’s visualization ability and level of proficiency can easily be determined through sketching and drawing techniques (Contero, Company, Saorin, & Naya, 2006; Mohler, 1997). The students placed in the first group ($n_1$) were able to approach the visualization and observe from a close range. Students placed in the second group ($n_2$) also had the privilege of close observation, but had to wear and keep on the blue glasses throughout the whole treatment. The computer generated dynamic visualization was presented to the third group ($n_3$) on a projector and they had the same time and lighting conditions as everyone else in order to create a sectional view of it.

![Figure 5. Sectional View of Dodecahedron](image)

The engineering drawing used in this research was a sectional view of the dodecahedron (see Figure 5). Sectional views are very useful engineering graphics tools, especially for parts that have complex interior geometry, as the sections are used to clarify the interior construction of a part that cannot be clearly described by hidden lines in exterior views (Plantenberg, 2013). By taking
an imaginary cut through the object and removing a portion, the inside features
could be seen more clearly. Students had to mentally discard the unwanted
portion of the part and draw the remaining part. The rubric used included the
following parts: 1) use of section view labels; 2) use of correct hatching style for
cut materials; 3) accurate indication of cutting plane; 4) appropriate use of cutting
plane lines; and 5) appropriate drawing of omitted hidden features. The
maximum score for the drawing was 6 points.

Data Analysis

Analysis of MCT Scores

The first method of data collection involved the completion of the MCT
instrument prior to the treatment to show equality of spatial ability between the
three different groups. The researchers graded the MCT instrument as described
in the guidelines by the MCT creators. A standard paper-pencil MCT was
conducted, in which the subjects were instructed to draw intersecting lines on the
surface of a test solid with a green pencil before selecting alternatives. The
maximum score that could be received on the MCT was 25 and, as it can be
seen in Table 1, n1 had a mean of 14.45, n2 had a mean of 12.75, and n3 had a
mean of 13.25. A one-way ANOVA was run to compare the mean scores for
significant differences, as it related to special skills among the three groups.
There was no significant difference between the three groups as far as spatial
ability, as measured by the MCT instrument (see Table 2).

Table 1

<table>
<thead>
<tr>
<th>MCT Descriptive Results</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
</tr>
<tr>
<td>N          Mean  SD    Std. Error  95% Confidence Interval for Mean</td>
</tr>
<tr>
<td>Lower Bound  Upper Bound</td>
</tr>
<tr>
<td>-------------</td>
</tr>
<tr>
<td>3D Printed (n1)   24   14.45  4.564  .847  12.71  16.18</td>
</tr>
<tr>
<td>3DPrinted Blue (n2)  21  12.75  4.561  .931  10.82  14.68</td>
</tr>
<tr>
<td>PC Blue Image (n3)  22  13.25  4.046  .826  11.54  14.96</td>
</tr>
<tr>
<td>Total          67   13.55  4.412  .503  12.54  14.55</td>
</tr>
</tbody>
</table>
Table 2

\[
\begin{array}{|c|c|c|c|c|c|}
\hline
\text{Quiz} & \text{SS} & \text{df} & \text{MS} & \text{F} & \text{p} \\
\hline
\text{Between Groups} & 40.918 & 2 & 20.459 & 1.053 & 0.354 \\
\text{Within Groups} & 1438.172 & 65 & 19.435 \\
\text{Total} & 1479.091 & 67 \\
\hline
\end{array}
\]

* Denotes statistical significance

Analysis of Drawing

The second method of data collection involved the creation of a sectional view drawing. As shown in Table 3, the group that used the 3D Printed Model, and wore the blue glasses as visual aid \((n = 21)\), had a mean observation score of 3.26. The groups that used the PC computer generated model, and used no blue glass visual \((n = 24)\), and the PC generated blue shaded image \((n = 22)\), had lower scores of 3.17 and 3.00 respectively. A one-way ANOVA was run to compare the mean scores for significant differences among the three groups. The result of the ANOVA test, as shown in Table 4, was not significant, \(F(2, 62) = 6.525, p < 0.802\). The data was dissected further, through the use of a post hoc Tukey’s honest significant difference (HSD) test. As it can be seen in Table 5, the post hoc analysis shows no statistically significant difference between the 3D printed Blue vs. PC Model \((p < 0.968, d = 0.96)\) and the 3D Printed Blue vs. PC Blue Image \((p = 0.792, d = 0.263)\), with PC Blue Image vs. PC Model being equal and higher than the first one in both cases \((p=.792, d=.263)\).

Table 3

\[
\begin{array}{|c|c|c|c|c|c|c|}
\hline
\text{N} & \text{Mean} & \text{SD} & \text{Std. Error} & 95\% \text{ Confidence Interval for} & \text{Lower Bound} & \text{Upper Bound} \\
\hline
3D Printed & 24 & 3.17 & 1.465 & 0.299 & 2.55 & 3.79 \\
3D Printed Blue & 21 & 3.26 & 1.046 & 0.240 & 2.76 & 3.77 \\
PC Blue Image & 22 & 3.00 & 1.272 & 0.271 & 2.44 & 3.56 \\
Total & 67 & 3.14 & 1.273 & 0.158 & 2.82 & 3.45 \\
\hline
\end{array}
\]
Table 4

Sectional View Drawing ANOVA Results

<table>
<thead>
<tr>
<th>Quiz</th>
<th>SS</th>
<th>df</th>
<th>MS</th>
<th>F</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Between Groups</td>
<td>.736</td>
<td>2</td>
<td>.368</td>
<td>.222</td>
<td>.802</td>
</tr>
<tr>
<td>Within Groups</td>
<td>103.018</td>
<td>62</td>
<td>1.662</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Total</td>
<td>103.754</td>
<td>64</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

* Denotes statistical significance

Table 5

Sectional View Drawing Tukey HSD Results

<table>
<thead>
<tr>
<th>Visual Aids (1 vs. 2 vs. 3)</th>
<th>Mean Diff. (1-2)</th>
<th>Std. Error</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>2 vs 1 3D Printed Blue vs. 3D Printed</td>
<td>.096</td>
<td>.396</td>
<td>.968</td>
</tr>
<tr>
<td>2 vs 3 3D Printed Blue vs. PC Blue Image</td>
<td>.263</td>
<td>.404</td>
<td>.792</td>
</tr>
<tr>
<td>3 vs 1 PC Blue Image vs. 3D Printed</td>
<td>.263</td>
<td>.404</td>
<td>.792</td>
</tr>
</tbody>
</table>

* Denotes statistical significance

Discussion

This study was done to determine significant positive effects of visual cues (color blue) and to identify a positive increase of spatial visualization ability for students in engineering technology courses. In particular, the study compared the use of different visual models: a 3D printed solid dynamic visualization with the addition of blue glasses to add blue color background around the model, a 3D computer generated blue shaded dynamic visualization, and a 3D printed dynamic visualization with no additional visual cue treatment. It was found that the use of visual cue (color blue) provided no statistically significant higher scores versus the treatment that did not utilize any visual cues.

While not statistically significant, the students who received treatment using the 3D printed Dynamic visualization, with the addition of the blue glasses visual cue, outperformed their peers who received treatment from the other two types of visualizations. Previous research supports that the effect of color on those with high spatial ability may result in little benefit, as high spatial ability learners develop mental models on shape alone. Khooshabeh and Hegarty (2008) suggested that color affects the performance of learners with low spatial ability more so than those with high spatial ability.
Strong and Smith (2002) reported that variations in technologies used for educating students may include application of texture, color, and lighting to 3D models which may significantly impact spatial ability. In a research study by Khooshabeh and Hegarty (2008), it was determined that color affected the performance in participants with low spatial ability, but did not show any statistically significance in students who already possess high spatial abilities as in engineering courses. This is mainly due to the high spatial ability learner using more schematic spatial mental representations where as the low spatial ability learn tend to use both visual and spatial information in performing tasks (Khooshabeh & Hegarty, 2008). Due to the findings in this study and the relatively high scores recorded from the MCT given to the participants prior to the treatment, the researchers believe that the population used (engineering technology students) did not demonstrate a statistically significant difference in spatial abilities from the addition of the color, due to the fact that spatial abilities were well developed in this population.

**Limitations and Future Plans**

In order to have a more thorough understanding of the use of visual cues used by engineering technology students during the creation of sectional views of 3D dynamic visualizations, and to understand the implications for student learning and spatial ability, it is imperative to consider further research. Future plans include, but are not limited to:

- Repeating the study to verify the results by using additional types of visual cues.
- Repeating the study using a different population such as technology education, science, or mathematics students.
- Repeating the study by comparing male versus female students, as it has been suggested that males tend to do better on spatial ability tasks than females (Carriker, 2009).
- Repeating the study with different populations to identify whether individuals with low spatial abilities, can benefit from the use of additional visual cues, such as color.

**References**


CEEB. (1939). *Special Aptitude Test in Spatial Relations (MCT).* Developed by the College Entrance Examination Board, USA, 1939.


**Note**

Preliminary results of this study were presented at the 2016 ASEE Midyear Conference proceedings in Daytona Beach, FL.

**About the Authors**

**Petros J. Katsioloudis** is an Associate Professor, the Industrial Technology Program Leader and Chair of the Department of STEM Education and Professional Studies, Old Dominion University, Norfolk, VA. His research focuses on improving teacher and student performance in STEM education, technical visualization and enhancing the development of a national STEM-educated workforce.

Email: pkatsiol@odu.edu

**Vukica Jovanovic** is an Assistant Professor of Mechanical Engineering Technology Department, Frank Batten College of Engineering and Technology, Old Dominion University, Norfolk, VA. Her areas of interest include the following: Mechatronics, Product Lifecycle Management, Digital Manufacturing, Engineering Collaboration, RFID, and Assembly Systems.

Email: v2jovano@odu.edu

**Mildred Jones** is a graduate student in the Department of STEM Education and Professional Studies at Old Dominion University. Her research focuses on technical visualization.

Email: mjones@odu.edu