An Empirical Investigation of the Relationship Between Success in Mathematics and Visual Programming Courses

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ABSTRACT

Many universities do not have prerequisites for the introductory computer visual programming course. Therefore, faculty and students do not have any means of predicting the student’s performance in this course. This research addresses this issue. Past research and accepted theory are presented to show the cognitive requirements for success in a first procedural programming course to be similar to those required for success in a mathematics course. Such research is lacking for visual programming. This research shows similar correlations between math courses and visual programming courses. Significant positive correlations were found between grades from Freshmen mathematics courses, ACT math scores, SAT math scores and grades from a Sophomore introductory visual programming course. This indicates that students who perform well in Freshman level Math courses, possess the cognitive characteristics required to perform equally well in Sophomore level visual programming classes. We can predict that students who perform well in math courses will perform equally well in a visual programming course.

Keywords: cognitive development, prerequisites, programming languages, procedural programming, visual languages, mathematics, business mathematics.

1. INTRODUCTION

There is a need to have prerequisites for programming courses to ensure that those who enroll have the necessary cognitive skills to be successful. A strong mathematics background predicts success in procedural programming (Alspaugh, 1970; Ricardo, 1983; Ignatuk, 1986). Studies have shown that math scores on the Scholastic Aptitude Test (SAT-M) and the American College Testing program (ACT) correlate with procedural programming course grades (Renk, 1987; Ott, 1989). Several other studies have shown a relationship between mathematics proficiency and success in procedural programming (Taylor and Mounfield, 1991). These studies support the practice of mathematics prerequisites for computer courses (Ralston, 1984; Saiedian, 1992).

However, there is no research to show whether this is true or not with visual programming. The purpose of this study is to investigate whether, like with procedural programming, there is a relationship between mathematics proficiency and success in visual programming. The Null hypothesis used in this research is: “there is no relationship (predictability) between success in Math courses and success in a Visual Programming course.”

1.2 Definitions of Procedural and Visual Programming

A procedural programming language is characterized by three properties: the sequential execution of instructions, the use of variables representing memory locations, and the use of assignment to change the values of variables (Louden, 1993). An example of such a language is COBOL. The instructions consist of three structure types: sequential, decision, and iteration. The instructions are placed in modules or subroutines with the data declarations kept separately from the procedure code.

Visual programming, such as Visual Basic, consists of visual objects that contain procedural code. An object can be loosely described as a collection of memory locations together with all the operations that can change the values of these memory locations (Louden, 1993). Data declarations, data definitions and program instructions are
all under one identifier, which is known as an object. The
language characteristic of Visual programming is the
manipulation of visual objects on a computer screen.

Visual Basic evolved from and is an enhancement of
regular procedural BASIC (Pietromonaco, 2002; Shelly &
Cashman & Quasney, 2003). Visual Basic has the code for
the procedural structures of sequence, iteration, and
selection with the added features of visual object-oriented
components. The visual components, such as a button, are
known as objects. They have properties and event
procedures (Nelson, 1993). Visual Basic “public” and
“private” procedures are like OOP public and private
methods. Visual objects encapsulate properties and event-
procedures (Schneider, 1999). Such characteristics are
lacking in procedural languages, therefore making visual
programming different from procedural programming.

The literature supports the idea that Visual Basic is a type
of Visual Programming language, different from
procedural (Buchner, 1999; Grehan, 1996a, 1996b; Llewellyn & Stanton & Roberts, 2002; Oz, 2002; Potter,
2003; Spain, 1996; Stair and Reynolds, 2001). One
academic text book describes Visual Basic as an OOP
language, rather than a third generation procedural
language like BASIC, C, COBOL, Pascal (O’Brien, 2004).
Visual Basic supports a syntax that looks a little object-
oriented (Holtzman, 1996; Bradley & Millsapough, 2003).
A report describes the extent to which object-oriented
(OO) programming can be performed in Visual Basic (Kai & McKim, 1998).

There is a distinction between procedural languages and
Visual Basic. “In procedure-oriented languages, the
emphasis of a program is on how to accomplish a task. The
programmer must instruct the computer every step of the
way. The programmer determines and controls the order in
which the computer should process the instructions.
Object-oriented/event-driven programming languages
emphasis is on the objects included in the user interface
(such as buttons) and the events (such as clicking) that
occur when those objects are used. Visual Basic is an
object-oriented/event-driven programming language.”
(Zak, 1999).

“To stress that Visual Basic is fundamentally different
from traditional programming languages, Microsoft uses
the term project, rather than program, to refer to the
combination of programming instructions and user
interface that makes a Visual Basic application possible”
(Schneider, 1999). With its object-oriented methods and
procedures, visual basic and other "visual" programs
require a different mindset from the common in-line
programming languages (Shirer, 2000).

2. LITERATURE REVIEW

Procedural programming, math skills and several cognitive
abilities such as general reasoning and analytic processing
have a positive correlation (Fletcher, 1984). Three studies
have shown relationships between success in procedural
programming, mathematics proficiency, and Piaget's
cognitive development (Caflolla, 1987; Azzedine, 1987;
and Werth, 1985). These relationships may be due to the
usage of the same area of the brain. Studies have shown
both procedural programming performance and math
ability correlating to the left hemisphere of the brain (Losh,
1984; Ott, 1989; Rotenjberg and Arshavsky, 1997).

2.1 Piaget’s Cognitive Development Theory

Piaget’s cognitive theory consists of three development
levels (Piaget, 1972; Epstein, 1990): pre-operational,
concrete, and formal operations. The first cognitive level,
pre-operational, is a very low level of thinking. Such a
person can use symbols from visual and body sensation to
represent objects but has problems with reversing actions
mentally (Biehler and Snowman, 1986, p. 62). For
example, that person fails to failure to recognize that the
amount of water remains the same when poured from a tall
thin glass to a short wide glass. At the next level, concrete,
a person can understand conservation of matter and
classification/generalization; i.e. conclude that all dogs are
animals and not all animals are dogs. However, such a
person is unable to comprehend mathematical ratios
(Barker and Unger, 1983). The final and highest cognitive
development level defined by Piaget is formal operation.
The ability to deal with abstractions, form hypotheses,
solve problems systematically, and engage in mental
manipulations characterizes this cognitive level (Biehler
and Snowman, 1986, p. 63). Biconditional reasoning, such
as “if and only if” logic, is a precondition to formal
operational reasoning (Lawson, 1983). Procedural
programming logic uses biconditional reasoning.

Piaget's theory indicates that formal operational thinking
abilities develop around age 12 (Chiapetta, 1976). It is at
this age that some students begin to move from concrete
thinking to logic/abstract thinking. Several studies have
shown that formal operations, such as abstractions and
logical thinking, develop at different ages or not at all in
people (Griffiths, 1973; Schwebel, 1975; Pallrand, 1979;
Bastian et al., 1973; Epstein, 1980). Many high school and
college students fail to attain full formal operational
thinking (Griffiths, 1973; Renner and Lawson, 1973;
Renner et al, 1978; Schwebel, 1972, 1975). This also
applies to adults. Research has shown that a majority of
adults fail at many formal operational tasks (Petrushka,
1984; Sund, 1976).

2.2 Cognitive characteristics of Computer
Programming

Research suggests that procedural programming deals with
high cognitive abilities such as problem solving and
Piaget's cognitive formal operations (Dalbey and Linn,
have shown that formal operational reasoning ability is
necessary for success in procedural computer
programming/logic (Caflolla, 1987; Fletcher, 1984; Little,
1984; Ricardo 1983; Azzedine, 1987; Barker and Unger,
1983; Barker, 1985).
Since procedural programming skills are related to logical reasoning (Cafolla, 1987; Flok, 1973; Foreman, 1988, 1990), low cognitive development thinkers are unable to do programming in light of Piaget's theory of cognitive development. This is consistent with Little's (1984) study. That study showed students who tested high in formal operations, scoring higher on programming and logical thinking measures than students who were concrete thinkers (a Piaget's lower level of cognition).

Cafolla (1987) found, "... some people of college age have difficulty learning procedural programming. This suggests that the cognitive skills needed to learn procedural programming develop later or perhaps never, in some.” There are those who lack or have limited cognitive skills to learn procedural programming (Becker, 1982).

Cognitive development is a factor in determining one's ability to learn procedural programming (Folk, 1973). Those who reach Piaget's formal operational stage, have the mental tools needed to understand programming. They have an abstract learning style that helps them learn programming (Hudak and Anderson, 1990).

Two recent studies have shown that object-oriented programming also requires formal operational reasoning ability (White, 2001; 2002). Is this also true for visual programming? Does visual programming success require formal operational cognitive development just like procedural and object-oriented programming do?

2.3 Math as an Indicator
The learning of complex, abstract concepts found in mathematics appears to require Piaget’s formal operation cognitive level (Pallrand, 1979; Parrino, 1981; Niaz, 1989; Nasser, 1993; Wolfe, 1999). Therefore, math is a good indicator of having the required cognitive development level to learn procedural programming. However, math grades from high school or college courses may be less accurate due to different instructors, different books, different tests, and different grading standards (White, 2003). The grades may not be comparable.

3. METHOD
To do research that will verify a math course to be beneficial, is difficult. It is infeasible to "randomly” assign students to different semester programming and math courses. Students have a set curriculum of courses to follow. The best way to research a math course prerequisite that indicates the required cognitive skills, is to correlate math course grades with programming course grades (White, 2003).

This study was done independent of instructors or math course locations. The intervening variables of different instructors and locations were not controlled or held constant. The math courses were taken with different instructors at various state universities and community colleges.

3.1 Data
A request to a state university Registrar’s Office was made for all students who took the first Computer Information Systems (CIS1) programming course for the past 3 years. Each record contained the CIS1 grade, three Freshmen Mathematics, and ACT/SAT math scores. The study did not consider Math equivalent courses taken. For example: it is possible that students who took College Calculus did not have this math grade considered in the evaluation. The sample size was 837 records.

Grades were given the values of 4 for an “A,” 3 for a “B,” 2 for a “C,” 1 for a “D,” and 0 for an “F.” Grades of “W” were treated as missing when performing correlations and step-wise regression. “W” grades were considered when evaluating grade distributions with math courses serving as a filter. If a course was repeated, the first grade was used. Using the second grade would have induced inflated grades due to familiarity of course content. However, many grades were either missing or were “Withdraw”. These grades were dealt with as exclude cases pair wise in the statistical analysis.

The sequence in taking the courses was ignored. Research has shown that Math and Programming courses do not improve/change cognitive development nor ability (Kurland et al., 1986; Flores, 1985; Platt, 1990; Shaw, 1984; Ignatuk, 1986; Mains, 1997; Kim, 1995; Priebe, 1997)

3.2 Variables
The dependent variable was a Sophomore level CIS Introductory Programming course (CIS1) using Visual Basic.

The five independent variables were:
(1) a Freshmen College Algebra course (Math1). This course covered linear equations, inequalities, word problems, functions, and logarithms.
(2) a Freshmen Mathematics for Business and Economics I course (Math2). This course covered college algebra and finite mathematics. College Algebra (Math1) was a prerequisite.
(3) a Freshmen Mathematics for Business and Economics II course (Math3). This course covered college finite mathematics and elementary differential calculus. College Algebra (Math1) was a prerequisite.
(4) the SAT math score
(5) the ACT math score

3.3 Statistics
Descriptive statistics and a correlation matrix of all variables were obtained. The independent variables were the three math courses and ACT/SAT scores. The math course grades (independent variable) of a grade of “2” (a grade of “C”) or better, was used as a filter. This showed
the changes in grade distribution of the visual programming course (dependent variable) when the math course was set as the criteria. Class GPA’s were calculated before and after filtering.

4. RESULTS

Over the last three years the percentage of D’s and F’s assigned in the CIS1 course, was 23%. When “W’s” were considered, the percentage of D’s, F’s, and W’s jumped to 34%. Table 1 shows that as the level of the prerequisite math course increased, so did the CIS1 class GPA, while poor grades of D’s, F’s, and W’s decreased. However, the number of students decreased since many had not yet taken the math courses. That is, out of the entire population of students enrolled in the programming course, 34% of them made a D, F or W in the programming class. Of the students enrolled in the programming course who had already passed the Math3 course with grade C or better, only 25.3% of them made a D, F or W in the programming class. As shown in the literature for procedural programming (Taylor and Moundifled, 1991; Ott, 1988; Renk, 1987), this visual programming course had significant correlations with math course grades and ACT/SAT scores as shown in Table 2. The highest correlations were with Math2 and Math3. With only one exception (SAT_Math and Math1), all variables correlated at the .05 confidence level.

It is interesting to note that the ACT/SAT scores also correlated at the confidence level of .05, yet the correlations were small when compared to the math courses. This may be due to a difference between students’ ability, as indicated by the ACT/SAT scores, and a willingness to perform, as indicated by math course grades.

5. DISCUSSION

The statistical relationships shown in the data analysis are sufficient to allow us to draw several conclusions and inferences. First, it is reasonable to conclude that the Freshman Math course, while insufficient to improve a student’s analytical and logical thinking skills, it is quite efficient and effective in assessing those skills. Second, from this and prior research it is clear that analytical and logical thinking skills are necessary to perform successfully in Math as well as in procedural, object-oriented, and visual programming courses. Third, it is clear that successes in the Freshman Math courses are a fairly good predictor of potential success in a Sophomore visual programming class.

The significance and practical usefulness of this research lies in the fact that we can predict the potential for success in a visual programming class from the student’s performance in the Freshman Math class. This can be an invaluable tool in advising students as to whether they should pursue a visual programming class or not. We will be serving our students tremendously if we can advise them whether it will be fruitful for them to pursue computer programming courses or not, since their potential for success in computer programming can be predicted from their performance in their Freshman Math class. Most universities teach computer programming during the Sophomore year after the student has had a Freshman Math course. The performance in the Freshman Math course can be used as an advising tool. Students who failed or performed poorly in the Math class can be advised that based on the results of this study, they would be unlikely to perform well in a visual programming course. Therefore a fixed Math prerequisite to the programming class, will do the students a service by not allowing them to enroll in a Programming course that statistically they would be expected to perform very poorly in. A coursework that does not involve computer programming can be arranged for a student who performs badly in the Freshman Math class. The effect on the students will be that we can advise them better as to what courses they are expected to perform well in, therefore guiding them towards a degree plan that they are cognitively capable of. The effect on faculty and university administration will be that students who are enrolled in computer programming classes will be expected to perform well and the success rates in those classes will increase. This ability to predict the student’s success will be a win-win situation for both the student and the academic institution.

<table>
<thead>
<tr>
<th>Table 1. CIS1 Grades</th>
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</thead>
<tbody>
<tr>
<td>Math Prerequisite of grade “C” or better</td>
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<tr>
<td>No Prereq</td>
</tr>
<tr>
<td>Class GPA</td>
</tr>
<tr>
<td>Grades % D’s &amp; F’s (no W’s considered)</td>
</tr>
<tr>
<td>% of D’s, F’s, &amp; W’s</td>
</tr>
<tr>
<td>Total N of students</td>
</tr>
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Table 2. Correlations

<table>
<thead>
<tr>
<th></th>
<th>CIS1</th>
<th>MATH1</th>
<th>MATH2</th>
<th>MATH3</th>
<th>ACT_MAT</th>
<th>SAT_MAT</th>
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<tbody>
<tr>
<td>CIS1</td>
<td>Pearson</td>
<td>.199**</td>
<td>.370**</td>
<td>.333**</td>
<td>.142*</td>
<td>.135*</td>
</tr>
<tr>
<td>Sig. (2-tailed)</td>
<td></td>
<td>.000</td>
<td>.000</td>
<td>.000</td>
<td>.045</td>
<td>.011</td>
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<tr>
<td>N</td>
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<td>348</td>
<td>469</td>
<td>199</td>
<td>350</td>
</tr>
<tr>
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<td>1.000</td>
<td>.318**</td>
<td>.209**</td>
<td>.317**</td>
</tr>
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<td>.318**</td>
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<td>.299**</td>
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<tr>
<td>MATH3</td>
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<tr>
<td>Sig. (2-tailed)</td>
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<td>287</td>
<td>534</td>
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<tr>
<td>ACT_MAT</td>
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<td>.142*</td>
<td>.317**</td>
<td>.299**</td>
<td>.304**</td>
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<tr>
<td>Sig. (2-tailed)</td>
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<td>.001</td>
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<td>Sig. (2-tailed)</td>
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</table>

** Correlation is significant at the 0.01 level (2-tailed).
* Correlation is significant at the 0.05 level (2-tailed).

6. REFERENCES


AUTHOR BIOGRAPHIES

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