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1. In two thirds of the cases, students taking courses by distance education outperformed their student counterparts enrolled in traditionally instructed courses;
2. The overall effect size \( d^+ \) was calculated as 0.37 standard deviation units (0.33 < 95% confidence interval < 0.40); and
3. This effect size of 0.37 indicates the mean percentile standing of the DE group is at the 65th percentile of the traditional group (mean defined as the 50th percentile).
Differences Between Traditional and Distance Education Academic Performances: A meta-analytic approach

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Abstract

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Background

Introduction

The Telecommunications Revolution of the last two decades of the Twentieth Century has changed all aspects of life, public and personal. The Internet truly has cast a worldwide Web of almost instantaneously active, fiber optic strands that bind together the practical worlds of business and commerce, and facilitates the exchange of views in the various academic and non-academic disciplines. In response to this burgeoning exchange of ideas, education systems (mainly in the industrialized countries and at higher levels) have pursued new methods of delivering education.

Distance Education

From time immemorial, teacher-lecturing/student-listening was the primary mode of traditional academic education. The delivery system for higher education has been a classroom setting with a professor giving a lecture and students listening and writing notes. Interaction between the professor and student has been viewed as an essential learning element within this arrangement (O’Malley and McCraw, 1999), often referred to as the “sage on the stage.”
Technological improvements such as printing machines, postal services, telephone, radio, television, and more recently the Internet, have been a driving force yielding new delivery methods and platforms. These new learning methods used to deliver distance education (DE) are proliferating exponentially in various learning programs, and leading some experts to predict that the “residential based model,” in the form of students attending classes at prearranged times and locations, will disappear in the near future (Blustain, Goldstein, and Lozier, 1999; Drucker, 1997, as cited in O’Malley, 1999). Although an expensive option today, video conferencing may create a virtual feeling that we are “back in the classroom.” Some forms of DE has progressed in concept and practice from an “anywhere,” to an “anytime,” to an “any pace” delivery method.

**Adjusting to Distance Education**

Academic and training communities have been continuously examining, assessing, criticizing, hallowing, and demonizing these new delivery methods as they appear. Without doubt, DE is of the highest relevance and importance to educators, students, and all other stakeholders. It is changing the physical face (i.e., massive buildings) of academic establishments. Students can now learn from the comfort of their homes or offices with no need to commute to campuses. Cutting-edge data are easily accessible on compact discs (CDs), portable personal computers (PCs), and have taken the place of instantly obsolete books. Online classrooms and libraries are replacing traditional campus facilities. Rather than requiring students to travel to a specific physical classroom or library, the Internet has facilitated the delivery of (nearly) unlimited learning resources to students.

Another facet of this change is evident in the increased accessibility of DE curricula and expert training and educational staff available at convenient venues for businesses and professional organizations. The need to train and develop employees on all levels has coincided with advances in new educational options. Organizations are continuously weighing the merits of in-house training versus sending candidates off-site to observe and train at other facilities. With the guidance of outside academic institutions specializing in DE training and development programs, human resource managers are implementing in-house DE programs. Access to courses, coaching, rotational assignments, and professional programs such as the American Management Association seminars, and university-sponsored Executive MBA programs, are now commonplace (Mondy, Noe, and Premeaux, 1999; Dessler, 1997; Westwood, 2001).

**The Problem, Question, and Purpose**

Distance education has existed for more than a century, although it has yet to be universally accepted relative to current and well-practiced face-to-face (F2F) programs provided by traditional brick and mortar institutions.

Some of the most frequently asked questions regarding DE, pertain to the quality of instruction and learning as compared to that experienced and achieved by students enrolled in F2F programs. Questions include: the cost of attendance compared to F2F institution; the needs of the “characteristic or average” DE students (e.g., DE students are generally older, employed, pay their own tuition, and computer skilled), as compared to “characteristic or average” students studying in traditional F2F environments; and finally, a comparison of factors affecting instructional efficacy and student learning in both situations. Many studies have researched various factors pertaining to DE, but as yet there is no comprehensive answer to these questions. Indeed, educational research often produces contradictory results. Differences among studies in
treatments, settings, measurement instruments, and research methods, make research findings difficult to compare and may even obscure existing trends (Bangert-Drowns and Rudner, 1991).

The purpose of this study was to research and provide an answer to the question of “quality” of DE programs: “Is there a difference in the quality of learning outcomes of DE programs compared to traditional F2F programs?” In order to examine the issue in its totality, a meta-analytical approach was adopted to synthesize various research studies in this field, and to examine differences between the two methods of delivery.

Distance Education and Quality Assessment

Distance Education Background

Distance education’s origins may be traced to nineteenth century in England and continental Europe when colleges used postal services for providing education by means of correspondence (Phipps and Merisotis, 1999; Ponzurick, Russo, and Logar, 2000; Sherry, 1996; Wernet, Olliges, and Delicath, 2000).

The term “distance education” has been used to describe the process of providing education where the instructor is distant (geographically separated) from the student (Gallagher and McCormick, 1999), or any instructional arrangement in which the teacher and learner are geographically separated to an extent that requires communication through media such as print or some other form of technology (Moore and Thompson, 1997, as cited in Spooner, Jordan, Algozzine, and Spooner, 1999; Perraton, 1988; Keegan, 1986; Garrison and Shale, 1987, as cited in Sherry, 1996).

Distance Education - Types and Technologies

Continued growth of DE has much to do with the advent of radio, television, and other media, which allowed for learning at a distance. This growth accelerated significantly during the 1990s with the use of computer-mediated learning technologies, e.g., two-way interactive video; two-way audio and Web-based asynchronous communication; and online or offline Internet Web-based instruction (Phipps and Merisotis, 1999; Ponzurick et al., 2000; Sherry, 1996; Wernet et al., 2000; Setaro, 2000). Advancements in increasingly flexible technology have enabled the Web’s visual, interactive nature to transform the traditional campus classroom-instructor system into a variety of different and innovative forms of instructional dissemination and to decentralized locations (Hall, 2002; Ponzurick et al., 2000).

Distance Education – Research Studies

A substantial body of research on DE, conducted between 1952 and 1992, showed that DE outcomes were not that different from those achieved in traditional classrooms (DeSantis, 2002). In their review of DE programs, Phipps and Merisotis” (1999) reported:

With few exceptions, the bulk of these writings suggest that the learning outcomes of students using technology at a distance are similar to the learning outcomes of students who participate in conventional classroom instruction. The attitudes and satisfaction of students using distance education also are characterized as generally positive. Most of these studies conclude that,
regardless of the technology used, distance education courses compare favorably with classroom-based instruction and enjoy high student satisfaction.

Russell (2002) also examined numerous studies and similarly reported further support of the “no significant difference” phenomenon.

On the other hand, numerous research studies have presented a different picture and therefore conflict with the conclusions cited above, creating a mixed and confusing situation (Dellana, Collins, and West, 2000).

**Distance Education – Quality / Effectiveness Factors**

A major concern about DE continues to be its quality compared to traditional classroom education. This concern has spurred extensive research into the factors that affect the quality of these programs. In many cases, “broad” measures of the effectiveness of DE have been examined (as shown in Figure 1).

**Figure 1.** Quality / Effectiveness Factors

![Quality/Effectiveness Factors](image)

Although student achievement is one common measure of a DE program’s success, it is recommended that program evaluators collect and report additional data to give the most exhaustive description possible.

Additional models of DE effectiveness that have been suggested include the “Flashlight Project” and Aptitude-Treatment Interaction (ATI) studies (Holmberg, 1989; Keegan, 1996; Thorpe, 1988, as cited in Lockee, Burton, and Cross, 1999); and assessment of cognitive and motivational characteristics of learners enrolled in Web-based instruction (Newlin, Wang, and Kosarzycki, 1998, as cited in Lockee et al., 1999). Research on DE effectiveness has focused on four domains: 1) student attitude and satisfaction regarding delivery of coursework; 2) interactions of students and faculty during delivery of coursework; 3) student outcomes in DE coursework; and 4) faculty satisfaction with delivery and coursework (Gallagher and McCormick, 1999). Additionally, Spooner et al. (1999) have analyzed many studies based on such comparative factors as: 1) cognitive factors, namely amount of learning, academic performance, achievement, and examination and assignment grades; and 2) other factors namely, student satisfaction, comfort, convenience, and communication with instructor, interaction and collaboration between students, independence, and perceptions of effectiveness.
Recent meta-analysis studies have focused on specific characteristics in DE: student satisfaction (Allen, Bourhis, Burrell, and Mabry, 2002); instructional features affecting learner achievement (Machtmes and Asher, 2000); and education technologies in K-12 learning (Cavanaugh, 2001).

Objective vs. Subjective Measurements

Measurement of such factors may be divided into two methods:

1. **Objective Measurements** of the academic performance factors. Operationalization of these measurements is based on course grades, tests, and exams. Scores are presented in quantitative measurements: points, percents, or letter grades that can be transformed to numeric values by a common transformation table. (Although course grades may carry some assessor subjectivity, they may be regarded as “more objective” than other factors.)

2. **Subjective Measurements** of the attitudes, satisfaction, and evaluation of instruction factors. Most researchers operationalize these variables using surveys and administering questionnaires with Likert-type scales, making comparability of results across different studies problematic.

Research Question and Hypothesis

Research Question

The literature search findings on one hand, and this study’s intent and purpose on the other, coincide in the direction of providing a comprehensive response to the following research question (RQ) focusing on the objective dimension of effectiveness pertaining to the quality of DE programs, in general, for the 1990-2002 period, and for the final academic performance dependant variable, in particular:

Is there a difference in the final academic performance of students enrolled in DE programs as compared to those enrolled in traditional F2F programs?

Hypothesis

The hypotheses was derived from all the aforementioned:

- H null: There is no difference in the final academic performance between students enrolled in DE programs and those enrolled in traditional F2F programs.

- H alt: The final academic performance grades of students enrolled in DE programs are higher than those enrolled in traditional F2F programs.

Methodology

The Meta-Analysis Concept

To synthesize the various studies, the statistical technique called “meta-analysis” has been implemented in this study. As there is more than one method of performing a meta-analysis and
calculating individual and aggregate effect-size, it is important to explain briefly the concept and rationale behind this technique in general, and how, in particular, it is implemented in this study.

- Comparison in terms of a standard effect-size can be conceptualized as a ‘standardized difference.” In the simplest form, effect-size, as denoted by the symbol “d,” is the mean difference between groups in standard score form – i.e., the ratio of the difference between the means to the standard deviation (Yu, 2001).

- “A meta-analysis on a given research topic is directed toward the quantitative integration of findings from various studies. Each study serves as the unit of analysis; the findings between studies are compared by transforming the results to a common metric called an effect-size (ES)” (Lemura, Von Duvillard, and Mookerjee, 2000; Cook, Heath, and Thompson, 2000; Becker, 1998).

- The approach provides “a clearer, more parsimonious review than previous qualitative discussions. The results are quantitative so a researcher gets a sense of the possible impact of a procedure against all published studies, rather than an illustrative few selected by a reviewer” (Heberlein and Baumgartner, 1978, p. 448).

- Bangert-Drowns and Rudner (1991) explained: Meta-analysis is a collection of systematic techniques for resolving apparent contradictions in research findings. Meta analysts translate results from different studies to a common metric and statistically explore relations between study characteristics and finding. Glass et al. (1981) argued that literature review should be as systematic as primary research and should interpret the results of individual studies in the context of distributions of findings, partially determined by study characteristics and partially random.

**Meta-Analytic Approaches**

With respect to meta-analysis, there are different approaches to the procedures, computations, and interpretation of results. Three main approaches (Bangert-Drowns and Rudner, 1991) are explained as follows:

1. **Classic or Glassian Meta-analysis:** Glass’ early meta-analyses set the pattern for conventional meta-analysis: define questions to be examined, collect studies, code study features and outcomes, and analyze relations between study features and outcomes. Additional features include: a) “classic” meta-analysis applies liberal inclusion criteria; b) The unit of analysis is the study finding. A single study can report many comparisons between groups and subgroups on different criteria, with effect sizes calculated for each comparison; c) Meta-analysts using this approach may average effects from different dependent variables, even when these measure different constructs. When submitted to critical re-analysis, Glassian meta-analysis has proven quite robust.

2. **Study Effect Meta-analysis:** This approach modifies the Glassian form in two ways: a) Inclusion rules are more selective. Studies with serious methodological flaws are excluded; b) Since the study is the unit of analysis, one effect size is computed for each study.
3. **Psychometric Meta-analysis:** Hunter and Schmidt’s (1990) approach to meta-analysis combined some of the best features of other approaches. All studies related to a given topic are gathered, regardless of quality. The distribution of effect sizes is corrected for sampling error, measurement error, range restriction, and other systematic artifacts.

### The Chosen Approach

The meta-analytic research type and approach chosen and implemented for this study was a combination of the ‘Study-Effect” and “Psychometric” methods. The parameters for this study are as follows: 1) Inclusion rules were more selective. Studies with serious methodological flaws were excluded. 2) The study is the unit of analysis. One effect size was computed for each study. 3) Effect sizes are of separate and independent studies. 4) Hunter and Schmidt’s (1990) corrections for sampling error, measurement error, range restriction, and other systematic artifacts were applied to the distribution of effect sizes. 5) Effect sizes were examined within each stratum and *across* all of the studies/strata.

### Procedures

Glass, McGraw, and Smith (1981) suggest procedures for conducting a meta-analysis that require a reviewer to complete the following steps: conduct a literature research to collect studies; code characteristics of studies; calculate effect sizes as common measures of study outcomes; and search for relationships between study features and study outcomes. These broad requirements and the methodological steps followed in this study are explained below.

1. **Defining the Domain of Research:** The “independent variable” is the method/mode of delivery. In our case, we have the: 1), DE mode, and 2) the F2F mode. The “dependent variable of choice” is final academic performance (final grade of course studies).

2. **Criteria for Including Studies in the Review were as follows:**
   - **Criterion 1:** The time period covered in the review: 1990–2002.
   - **Criterion 2:** Published/unpublished studies: Both types were included.
   - **Criterion 3:** The quality of a study. Only studies showing no severe methodological flaws were included. Because the materials gathered were from a wide scope of researchers and studies, a thorough ‘sensitivity analysis’ was consequently performed to assess the robustness of combined estimates to different assumptions and inclusion criteria.
   - **Criterion 4:** Control group – Each primary study had a control or comparison group.
   - **Criterion 5:** Sufficient quantitative data – The results in these studies all provided sufficient quantitative data (sample size, mean and standard deviation) from which effect sizes were calculated.

3. **Determining the Type of Effect Size to Use:** Different statistical methods exist for combining the data, but there is no single correct method (Egger, Smith, and Phillips, 1997). Two popular approaches are those of Glass (of which the basic formula for \( d \) is
“the Mean of control group minus the Mean of treatment group, divided by the Standard deviation of the control group” (Glass et al., 1981); and Hunter and Schmidt (1990) who suggested a “pooled within-group standard deviation” and correcting the effect size for measurement error. Based on the statistical methods described in Buchan (2000), the estimation of the effect size in this study was calculated by using:

1. “$g$” – Modified Glass statistic with pooled sample standard deviation, presented with an ‘exact’ 95 percent confidence interval as in the formula:

\[
g = \frac{\mu_e - \mu_c}{\sigma_{pooled}}
\]

4. “$d$” – The unbiased estimator (Hedges and Olkin, 1985), presented with an “approximate” 95 percent confidence interval (Where: sample size $N = e + c$, $J(m)$ is the correction factor).

\[
d = g\left[J(N - 2)\right]
\]

5. By convention, if the subtraction of the means ($M$) of the two groups (experimental and control) yields a positive difference if it is in the direction of improvement or in the predicted direction, and a negative difference if in the direction of deterioration or opposite to the predicted direction.

6. Searching for Relevant Studies: The search for published and unpublished materials was carried out using several different approaches:

   o **Computer Search.** All searches were for studies published principally in English. Studies in French, Spanish, and Hebrew were translated only if they showed relevancy. The electronic search engines Pro-Quest, Google, NorthernLight, and WiseNut were used to retrieve as many published articles as possible. The electronic library and inter-library data banks of Touro College and Tel Aviv University were consulted as well.

   o **Compilations, Reference Lists, and Authors.** Two robust compilations of published study materials: “What’s the Difference?” A review of contemporary research on the effectiveness of distance education in higher education (Phipps and Merisotis, 1999), and “The No Significant Difference Phenomenon” (Russell, 2002), proved invaluable in addendum to the reference lists and bibliographies of the studies collected through the online searches. In most cases, data were incomplete or not clear, requiring a personal follow-up letter to the authors.

   o **Yield.** Over 1,600 papers were reviewed, but initial criteria for screening for study content and timeframe (criterions 1 and 2), reduced the number to 259.

7. **Selecting the Final Set of Studies:** Results from the studies were entered into a database using Microsoft Excel, and every study was assigned a unique identification number. From the 259 studies aggregated within the database: 86 (33 percent) qualified; 105 (41
percent) did not meet criteria 3, 4, and 5; and the search for the full text versions of 68 (26 percent) studies was not successful.

8. **Data Extraction and Coding:** All studies were reviewed for relevant information and noteworthy characteristics that might be related to the effect size pertinent to this study. Data on variables of interest were extracted, recorded, and appended to the database and coded for the following characteristics: factors in research design, list of sample characteristics, and type of dependent variable. In addition to effect sizes derived for the dependent variable of final academic performance, data extraction and analysis of qualifying studies produced effect sizes for five other variables: 1) academic performance – sub tests; 2) satisfaction; 3) gain; 4) instruction evaluation; and 5) other factors. The sample size for each aforementioned category was small \( n < 30 \); therefore, it was not robust enough for inclusion in this statistical analysis.

9. **Determining The Mean and Confidence Intervals of Effect Sizes across Studies:**

   - **Individual Effect Sizes.** Individual results were expressed in a standardized format to allow for comparison between studies. A statistical computing software program, *Stats Direct LTD* (2002), was used to input and calculate individual and overall effect sizes. The statistical methods employed in this program were conservative and conformed to the current consensus in statistical literature.

   - **Overall Effect Size.** Once all effect sizes of the individual studies were recorded, the overall pooled mean effect size estimate \( d^+ \) was calculated using direct weights defined as the inverse of the variance of \( d \) for each study/stratum. An approximate confidence interval for \( d^+ \) is given with a Chi Square statistic with the probability of this pooled effect size being equal to zero (Hedges and Olkin, 1985).

   - **The Fixed Effect Method.** Egger et al. (1997) argued that the results from small studies are subject more to the play of chance and should, therefore, be given less weight. The statistical techniques or models to do this (differing in the way the variability of the results between the studies is treated) are the “fixed effects” model and the “random effects” model. Neither model can be said to be correct; both are subject to continuing disagreement and debate among statisticians (Fleiss and Gross, 1991; Sahai and Kurshid, 1996; DerSimonian and Laird, 1986). This research was implemented following the more conservative fixed effects method (Buchan, 2000).

10. **Homogeneity and Bias Analysis:** Because diverse studies and data based on different methods of calculation were synthesized, each with its own method of calculation, for the results to be accepted it was necessary to examine the robustness of the findings in light of different assumptions using a Homogeneity and Bias analysis.

   - **Homogeneity.** Individual trial results will show chance variation, therefore, it was necessary to explore whether the differences were larger than those that were expected. One of the main concerns in conducting meta-analysis is that there would be a publication bias arising when trials with statistically significant results are more likely to be published and cited, and are therefore preferentially published in English language journals (Jüni, Holenstein, Sterne, Bartlett, and Egger, 2001). Consequently, the plots of the trials’ variability or sample size...
against effect size are usually skewed and asymmetrical in the presence of publication bias and other biases (Sterne and Egger, 2001), and are more likely to affect small trials. This leads to the use of plots of sample size against effect estimate. Bias is likely to cause asymmetry in such plots, making the examination of the left-right symmetry of the plot required to detect bias (StatsDirect, 2002; Buchan, 2000).

- **Fail Safe N.** Since only published studies were analyzed, there was a “file drawer problem.” In other words, how many studies that did not find significant effects had not been published? If those studies in the file drawer had been published, then effect sizes for those treatments would be smaller. Therefore, the Fail Safe N is the number of non-significant studies (NFS) that would be necessary to reduce the effect size to a non-significant value. Based on Orwin’s (1983) formula, and Lipsey’s (1990) standards, an NFS was calculated for each meta-analysis iteration.

11. **Presenting the Results:** An effect size ($d^+$) calculated from a very large sample is likely to be more accurate, than one calculated from a small sample. This margin for error can be quantified using the idea of a confidence interval, which provides the same information that is usually contained in a significance test (i.e., a 95 percent confidence interval is equivalent to a 5 percent significance level). Because meta-analysis results are better understood when displayed graphically, the effect sizes with their 95 percent confidence intervals (CI) are presented using a Forest Plot (Egger et al., 1997), or by presenting the results in a histogram of the “g” effect size distribution.

12. **Testing the Hypothesis:** An approximate confidence interval for the effect size ($d^+$) is given with a Chi-square statistic including the probability that effect size is equal to zero. Following Hedges and Olkin (1985), the null hypothesis is rejected if the probability for $d^+$ being equal to zero is smaller than 0.01.

13. **Qualitative Interpretation of Effect Size ($d^+$):** Interpreting the results of a meta-analysis requires an understanding of the standards employed that allow for meaningful interpretation of effect sizes. Based on the findings of 102 selected mean effect size estimates from 186 meta-analyses of 6,700 studies involving 800,000 subjects, Lipsey (1990) categorized effect sizes into three groups. These groups and their range of effect sizes were: Small 0.00 to 0.32; Medium 0.33 to 0.55, and; Large 0.56 and higher. Lipsey’s (1990) categorizations were used in this study.

**Research Results**

Of the 259 studies aggregated within the database, 86 met all relevant criteria and therefore were included in this study. The data extraction and analysis from these works produced 86 calculated effect sizes, which yielded the final academic performance factor. These 86 effect sizes were the “basis” for the meta-analysis iterations conducted to answer the study’s research question:

**Research Question**

The central research question for this study was: “Is there a difference in the Final Academic Performance of students enrolled in DE programs compared to those enrolled in traditional F2F programs?”
Results

The Results of This Study Were as Follows:

1. **Sample:** Eighty-six effect sizes were calculated for final academic performance. There was a “clear distinction” between positive and negative results: 66 percent were positive (DE out-performed F2F). Aggregating all of the comparative studies provides for a large student subject body of over 15,300 students, divided almost equally between the two groups: DE (7,270) and F2F (8,076).

2. **Effect Sizes** $g$ and $d$: The breakdown of all 86 effect sizes, in terms of $d+$ standards, negative and positive, provides roughly the following findings: a one-third negative, a one-third small, and a one-third medium and large result (see Table 1 for detailed distribution).

<table>
<thead>
<tr>
<th>Standard</th>
<th>Overall N</th>
<th>Overall %</th>
</tr>
</thead>
<tbody>
<tr>
<td>Large: $d+ &gt; 0.56$</td>
<td>16</td>
<td>19</td>
</tr>
<tr>
<td>Medium: $0.33 &lt; d+ \leq 0.55$</td>
<td>11</td>
<td>13</td>
</tr>
<tr>
<td>Small: $d+ &lt; 0.32$</td>
<td>30</td>
<td>35</td>
</tr>
<tr>
<td>Negative: $d+ &lt; 0$</td>
<td>29</td>
<td>33</td>
</tr>
<tr>
<td>Total</td>
<td>86</td>
<td>100</td>
</tr>
</tbody>
</table>

3. **Bias Indicators Numbers and Plot:** From regression of normalized effect vs. precision: Intercept (0 if unbiased) = -1.908 (approximate 95 percent); CI = -3.012 to -0.804; $P = 0.0009$. From Kendall's test on standardized effect vs. variance: $\tau = 0.056$; $P = 0.452$. An examination of the left-right symmetry of the plot as depicted in Figure 2 denotes that there is, in fact, a small sample bias.

**Figure 2.** Bias Assessment Plot
4. **Fail Safe N**: Based on Orwin’s (1983) formula for calculating the $N$ fail safe number, when $dc = 0.01$ is selected as our criterion value, about 3,062 additional unreported studies averaging a “null” result are needed (existing somewhere), to “nullify” the average of 0.366. The Fail Safe $N$ for the overall study is large. It was unlikely that there were that many well constructed studies sitting in file drawers.

5. **Pooled Estimate of Effect Size $d^+$**: Computation of the pooled estimate of effect size $d^+$ yielded the final result of 0.366, with a 95 percent confidence interval of 0.33 to 0.40. This finding corresponded with Lipsey’s (1990) medium effect size definition. The Chi Square of 397 ($P < 0.0001$) is most significant. In Figure 3, the solid contour line is the control (F2F) group distribution; the dotted contour line is the experimental (DE) distribution; and $SD_x$ is the average effect size in standard deviation units. An effect size of 0.366 indicates that the mean of the treated group is at the 65th percentile of the untreated group (See Figure 3). Figure 4 depicts further detailed data relative to the distribution of the effect sizes ($g$) thought of as the average percentile standing of the average treated (or experimental) participant compared to the average untreated (or control) participant.

**Figure 3.** Percentile of F2F Group Distribution

![Figure 3](image)

**Figure 4.** Distribution of Effect Sizes ($g$)

![Figure 4](image)
6. **Hypothesis:** Based on these findings, providing an overall effect size of 0.366 from 86 studies, with a significant Chi-square of 397.55, we are accepting the alternative hypothesis that: The Final Academic Performance grades of students enrolled in DE programs are higher than those enrolled in traditional F2F programs.

**Discussion and Implications**

**Summary**

To facilitate review and discussion, all previous findings for the research question were aggregated and consolidated within Table 2. Of the 86 qualifying studies, over two-thirds were positive, denoting that DE out-performed F2F. The overall effect size calculated was positive in direction, and, by Lipsey’s (1990) standards, the overall effect size ($d^+ = 0.366$) was of a medium magnitude.

The overall effect size of 0.366 calculated from 86 studies and testing for Chi Square of 397 brought us to the conclusion that our null hypothesis defined as: “there is no difference between DE and F2F instruction for the final academic performance factor,” should be rejected in favor of our alternative hypothesis.

**Table 2. Summary of Study Results**

<table>
<thead>
<tr>
<th>Literature / Study Search Results</th>
<th>Studies (%)</th>
<th>Remarks</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total</td>
<td>259</td>
<td>100%</td>
</tr>
<tr>
<td>Studies – that met criteria</td>
<td>86</td>
<td>33%</td>
</tr>
<tr>
<td>Studies – That do not meet criteria</td>
<td>105</td>
<td>41%</td>
</tr>
<tr>
<td>Studies – Full text not found</td>
<td>68</td>
<td>26%</td>
</tr>
</tbody>
</table>

**Discussion**

Based on 86 studies and using learning outcome data from over 15,000 participating students, the results of the meta-analysis show a strong positive trend indicating that DE is an effective form of instruction. This analysis demonstrates that students engaged in DE academically outperform their F2F counterparts. We have been focusing all along on the question: “Is DE suitable for all students?” The results of this study may raise the inverse question: “Is F2F suitable for all students?” and may begin a paradigm shift in the way postsecondary education is pedagogically conceptualized.

Examining and reducing bias is of major concern to any researcher implementing meta-analysis procedures. Homogeneity and bias analysis have been the focus of many statistical discussions, and are worthy and deserving of specific attention in this discussion.
Homogeneity

Since all individual trials will show chance variation in their results, we need to explore whether the differences presenting in this study are larger than those caused by chance alone. If a series of independent studies provide a common (homogenous) estimate of the population effect size, then it is more likely that the various studies are testing the same hypothesis. If these estimates are heterogeneous, then the question of whether each study is testing the same hypothesis arises.

Heterogeneity provides a warning that it may not be appropriate to combine and synthesize all the study results in one meta-analysis (Wolf, 1986). Debate among statisticians continues as to how to minimize those studies affecting the homogeneity of the research. Hedges (1982) and Hunter, Schmidt, and Jackson (1982) suggest that it is inappropriate to include them in one meta-analysis. Harris and Rosenthal (1985) argue that heterogeneity is analogous to individual differences among subjects within single studies, and is common whenever many studies by different investigators using different methods are examined. Hedges (Becker and Hedges, 1984) admitted that: “It is not necessarily inadvisable to draw inferences from heterogeneous effects.”

The bottom line is that the more refined the research, and the more acute the coding of the individual study’s characteristics, and the less likely that heterogeneous studies will be accepted. This study has adopted and followed a conservative approach throughout all of the meta-analysis procedures. Therefore, even though the final results produce a significant $Q$ (non-combinability) statistic, it is strongly contended that the well defined scope of the study and the narrow and acute extraction of specific data pertaining to a defined dependent variable, override the need for the reduction of supposedly suspicious, outlier studies.

Bias Detection and Analysis

Publication bias arises when trials with statistically significant results are more likely to be published and cited, and are preferentially published in English language journals (Jüni et al., 2001). In this study a systematic approach – the usage and presentation of Bias Plots (where lateral asymmetry indicates bias), and the calculation of Fail Safe $N$ numbers, was taken to identify the possibility of publication bias.

Bias Plots

Examination of the bias indicators and plots for our research question shows that we have a small sample bias for the overall meta-analysis. Assessing this bias from the “half-full glass” perspective, had the studies that were analyzed been more robust in their student sample sizes, the bias would have been reduced and we would be even more confident in our current positive findings and results.

Fail Safe $N$ (NFS)

The “file drawer problem” (Rosenthal, 1979) refers to the question: How many studies that did not find significant effects have not been published? In other words, if those studies in the file drawer had been published, then the effect sizes for those treatments would be smaller. Therefore, the Fail Safe $N$ is the number of non-significant studies that would be necessary to reduce the effect size to a non-significant value.
In this study the NFS numbers were calculated based on Orwin’s (1983) formula, providing large numbers for our research questions sample sizes. When \( dc = 0.01 \) was taken as our criterion value, it was found that about 3,062 additional unreported studies averaging a “null” result would be needed, to nullify our overall \( d^+ \) average. It is unlikely that there were that many well constructed studies sitting in file drawers.

Distance education literature and research have revolved around three main issues: 1) differences in study formats; 2) variety of technologies; and 3) quality of instruction. Unfortunately, the plethora of information and the diversity of its presentation have obscured some significant findings. First, the differences among studies in terms of treatment, setting, measurement instruments, and research methods have made findings difficult to compare. These differences are so extensive that they obscure trends with an overwhelming amount of information (Bangert-Drowns and Rudner, 1991). Second, an emphasis on the technical aspects of overcoming the distance factor in DE has distracted the discussion from the efficacy of the education itself. The methods of overcoming the geographic distance, which separates instructor from student, vary widely (Gallagher and McCormick, 1999). Methods take the form of computer-mediated learning, two-way interactive video, Web-based asynchronous communication, and on and offline Internet Web-based instruction (Ponzurick et al., 2000; Phipps and Merisotis, 1999). Third, studies of individual DE courses generally conclude that the learning outcomes are not that different from those found in traditional F2F classroom formats (DeSantis, 2002; Phipps and Merisotis, 1999; Russell, 2002).

The meta-analysis approach, methodology, and findings of this study, specifically address these three issues. Since this study provided a positive result for the final academic performance dependent variable, we can deduce that DE not only is comparable to traditional instruction, but also, when subject to our criteria, can outperform traditional instruction. It is important to note that this deduction is further supported by the preliminary results of an ongoing meta-analytic study comparing DE to traditional classroom instruction (Bernard, Lou, Abrami, Wozney, and Borokhovski, et al., 2003). Also, this study did not differentiate between the educational delivery methods of time (synchronous and asynchronous) and place (same and different) dimensions, as categorized by O’Malley and McCraw (1999), or the various technological and telecommunication delivery systems, but rather remained with the general dichotomy of DE courses vs. traditional courses.

Most education researchers accept that if it could be shown that making a small and inexpensive change would raise academic achievement by an effect size of as little as 0.1, then this could be a significant improvement (Coe, 2000). Even a 0.50 standard deviation improvement in achievement scores is considered to be a conventional measure of practical significance (Rossi and Wright, 1977). A one-third (0.33sd), but at times as small as one-fourth (0.25sd), standard deviation improvement will also be considered to be educationally significant (Tallmadge, 1977). Therefore, the overall effect size \( d^+ = 0.366 \) found in this study, defined by Lipsey’s (1990) standard as a sound “medium,” denotes that average students in the DE group would find themselves in the 65th percentile of the traditional control group: an educational incremental achievement, at least.

**Implications**

The purpose of this work was to research and provide an answer to the “quality question” of distance-learning programs. Specifically: Is there a difference in the quality of DE programs relative to traditional F2F programs? Providing an answer to this question would serve the multifaceted fields of education, teaching, training, and learning. Such information would be a
supporting construct to the decision makers in academic, business, and professional organizations contemplating the various options for training and development.

These findings provide a definite and positive answer to the DE versus F2F conundrum. This study serves a variety of functions: a) it may serve future researchers in this field, in that this meta-analysis synthesizes a wide body of academic comparative studies of DE vs. F2F; b) with respect to quality of DE programs, it provides concrete findings based on a robust compilation of research studies across various subjects of learning and across all levels of higher education as to the quality of DE Programs; and c) by adopting a meta-analytic approach, it provides a comprehensive answer.

If we argued that DE has yet to be compared to F2F programs provided by the traditional brick and mortar institutions, then the implication of this study’s results should be to accept DE as a respectable and feasible option for education. The findings have implications for changes in each of the following entities:

**Academic Institutions and Policy Makers:** Educational institutions (mostly in the higher levels of academia) have adopted these new DE methods as the educational delivery systems of the future. We have seen the insertion of DE delivery methods into learning programs in a variety of formats: optional, complementary, interchangeable, or full and completely on-line programs. It is doubtful, however, that we will see materialization of some experts’ predictions of the disappearance of the residential based model in the near future (Blustain et al., 1999; Drucker, 1997 as cited in O’Malley and McCraw, 1999). Rather, these institutions should be encouraged to invest in DE technology, implement DE learning programs, and transform some existing F2F courses to DE so that a wider range of students, regardless of age or geographic location, may enroll in quality learning programs. The realistic possibility of hybrid programs, providing the student with the best of both worlds, may likewise be enhanced.

**Organizations:** This study may encourage organizations to enhance their in-house professional and managerial training and development programs by combining the power of the Internet and other telecommunication options to create the “information society” characterized by the emergence of the “knowledge worker” (Drucker, 1989). The case for what Dessler (1997) and Mondy, Noe, and Premeaux (1999) refer to as in-house training and development programs that are provided by numerous colleges and universities, can be strengthened by the supporting evidence of this study.

**Students:** When offered the option of DE as a means of pursuing their academic quest, students can now choose solely on the merits and quality of the program offered, without fear that DE may hinder their academic performance outcomes. For those prepared to commit themselves to this form of study, need not wonder if the DE option of learning anywhere, anytime and at any-pace, is any good. This study may provide positive reassurance, subject to their serious assessment of their specific field of study.
References


