The Predictive Validity of IELTS Scores: A Meta-Analysis

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Abstract and Keywords

Abstract

Thousands of education institutions worldwide rely on IELTS scores as criteria for accepting international students whose first language is not English. Individual studies have found varying degrees of strength of correlations and conflicting results between IELTS scores and academic success. These conflicting results were examined through a meta-analysis, while also investigating multiple moderating variables: research funding bias, individual skill scores, level, field, & country of study, presence of additional English courses, and GPA timepoint.

Results from 18 studies show an approaching-small effect size of $r = .227$ for the relationship between IELTS scores and post-secondary GPA. The majority of macro skills (listening, writing, and speaking) do not reach the small effect size, however reading approaches it with an effect size of $r = .215$. Most moderator analyses were inconclusive owing to the small amount of studies, but potential differences from individual studies are examined and discussed.

144 words

Keywords:
Summary for Lay Audience

Increases in international student enrolment has led to an increased need to examine the language tests used to ensure they have the necessary English skills to succeed in university. One of the most popular of those tests is the *International English Language Testing System*, or IELTS. The present study is a meta-analysis, or synthesis of large amounts of already published research, into how IELTS scores used for admission relate to actual student scores once in post-secondary education. A total of 18 different studies were aggregated and examined to look for any evidence of publication bias from the IELTS organization, as well as any stronger links for specific English skills (listening, reading, writing, and speaking). Additional examination is made into if the relationships are stronger for: different majors, different levels of study (graduate or undergraduate), different countries, presence of extra English courses, and results at different times.

Results show that the IELTS test has a relatively small predictive effect for GPA in post-secondary programs. No evidence of bias was found in publishing source, and overall scores were much more predictive than any one individual skill such as listening or writing, although reading shows a slight relationship. Due to limited sample sizes, few conclusions can be drawn from the presence of any extraneous variables.
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Table of Contents

Abstract and Keywords........................................................................................................ ii
Summary for Lay Audience .................................................................................................. iii
Acknowledgements............................................................................................................ iv
List of Tables ...................................................................................................................... vii
List of Figures ..................................................................................................................... viii
List of Appendices ............................................................................................................ ix
List of Abbreviations and Symbols.................................................................................. x

Chapter 1 - Introduction .................................................................................................... 1

Chapter 2 – Literature Review .......................................................................................... 3
  2.1 IELTS ...................................................................................................................... 3
  2.2 Academic Success ................................................................................................... 5
  2.3 Validity ................................................................................................................... 6
  2.4 Meta-Analysis ......................................................................................................... 8
  2.5 Previous Meta-Analyses on English Test Validity ..................................................... 9

Chapter 3 – Integrated Article ........................................................................................... 11
  3.1 Introduction .............................................................................................................. 11
    3.1.1 Research Questions .......................................................................................... 15
  3.2 Literature Review ...................................................................................................... 16
    3.2.1 IELTS .............................................................................................................. 16
    3.2.2 Meta-Analysis .................................................................................................. 17
    3.2.3 Conducting a Meta-Analysis ........................................................................... 17
  3.3 Method ..................................................................................................................... 20
    3.3.1 Defining the Research Domain ....................................................................... 20
    3.3.2 Conducting the Literature Search .................................................................. 21
    3.3.3 Coding ............................................................................................................ 23
    3.3.4 Analysis ............................................................................................................ 25
  3.4 Results ...................................................................................................................... 26
    3.4.1 Descriptive Statistics ...................................................................................... 28
    3.4.2 Research Question 1: Overall Effect ............................................................... 30
    3.4.3 Research Question 2: Funding Bias ................................................................. 32
    3.4.4 Research Question 3: Subscores ..................................................................... 33
List of Tables

Table 1: Literature Search Results .............................................................................................................. 22
Table 2: Analysis Type Moderator Analysis ................................................................................................. 27
Table 3: Descriptive Statistics ......................................................................................................................... 29
Table 4: IELTS Publication Status ................................................................................................................. 32
Table 5: Individual Macro Skill Analysis ....................................................................................................... 33
Table 6: Moderator Analysis – Field of Study ............................................................................................... 34
Table 7: Moderator Analysis – Level of Study ............................................................................................... 35
Table 8: Moderator Analysis – Country of Study ............................................................................................ 35
Table 9: Moderator Analysis – Country of Study without Quebec ................................................................. 36
Table 10: Moderator Analysis – Top-Up ......................................................................................................... 37
Table 11: Moderator Analysis – Timepoint ..................................................................................................... 37
List of Figures

Figure 1. Publications by year ........................................................................................................ 30
Figure 2. Forest plot.......................................................................................................................... 31
Figure 3. Funnel plot.......................................................................................................................... 47
List of Appendices

Appendix A: Email Template

........................................................................................................64

Appendix B: Included Studies

...........................................................................................................65
# List of Abbreviations and Symbols

- **CMA** – Comprehensive Meta Analysis
- **EAP** – English for Academic Purposes
- **EFL** – English as a Foreign Language
- **ESL** – English as a Second Language
- **GPA** – Grade Point Average
- **IELTS** – International English Language Testing System
- **L1** – First Language
- **L2** – Second Language
- **TOEFL** – Test of English as a Foreign Language
- **\( \rho \)** – Spearman’s Rho
- **\( r \)** – Pearson’s Product-Moment Correlation
- **\( \beta \)** – Beta Coefficient
- **\( p \)** – Probability Value
- **\( Z \)** – Standard Score
- **\( I^2 \)** – Degree of Inconsistency
Chapter 1 - Introduction

Recent years have seen a large rise in the number of international students worldwide, with the number of international students globally doubling from 2000-2013 (Choudaha, 2017). Many of these students want to study in the Western world and will be studying in English, which may not be their native language. As universities recruit more and more international students, they need some way to measure their English language ability in order to ensure they will be able to succeed in an academic setting. One of the most popular English language tests used for university admissions around the world is the International English Language Testing System or IELTS (IELTS, 2017b). Thousands of universities rely on IELTS scores as one of the key factors in determining whether to accept a new student or not. The question then arises: is the IELTS test capable of predicting academic success?

Academic success can be a vague term and mean many different things to many people. For the purposes of the present study and the need for numerical analysis, Grade Point Average, or GPA will suffice. Methods of calculating GPA differ from institution to institution, but generally all institutions offer some form of numerical or ordinal score to represent how well a student is performing in their academic classes. By looking at correlations between IELTS admission scores and subsequent GPAs, a measure of predictive ability can be found.

The following thesis outlines a meta-analysis examining that very construct. A widespread literature search was conducted, hundreds of results were examined in detail, 28 studies were coded and 18 of them were included in the final analysis. These studies were examined for a variety of moderating factors such as IELTS funding status, study location, student major, level of study, enrolment in a “top-up” program, and timepoint of GPA measure.
The results from this meta-analysis may be of use to many different stakeholders. They may help to validate the test itself for the IELTS organization, they may help validate the use of IELTS scores for admission decisions by post-secondary institutions, and they may even demonstrate the value of the test to the test-takers themselves.
Chapter 2 – Literature Review

2.1 IELTS

The IELTS test was first introduced in 1989 and has risen to become one of the top English tests worldwide (Davies, 2008). It is used for admission to universities, migration to commonwealth countries, and even by a variety of employers. There have been several revisions throughout the past decades, but the overall format has stayed the same since 1995 (IELTS, 2014). There are over 1000 testing centers around the world and the test is administered approximately 40 times a year. It is owned and managed by three separate language groups: the British Council IDP, IELTS Australia, and Cambridge English Language Assessment.

The IELTS test is scored according to a band system with levels ranging from 1 to 9 by half-band increments (IELTS, 2014). Level 1 is labelled as a “non user” wherein the participant has no ability to communicate in English, whereas level 9 is considered an “expert user” with full command of the language, along with fluency and complete comprehension. The test-taker is given a band score in each of the four components (listening, reading, writing, and speaking), and the four scores are averaged and rounded to the nearest half-band to give an overall score. In 2017 the average overall score of all test-takers worldwide was roughly a band 6 for the academic test and 6.5 for the general training test (IELTS, 2018a). IELTS as an organization recommends that different academic institutions choose their own cut-off bands for entrance, but overall suggests a 6.5-7.0 minimum for academic courses and 6.0-6.5 for training programs.

The IELTS test is offered in two different versions each containing four components (IELTS, 2014). Test takers can choose between the academic or general training test, depending on their desired purpose. Generally, the academic test is used for most university entrances
(though some institutions do accept the general training test) while the general training test is used for immigration or evidence of English proficiency. The academic test is much more common overall, with 78.10% of test-takers in 2017 choosing to take the academic version (IELTS, 2018a). Each test contains the same four components: listening, reading, writing, and speaking. The listening and speaking sections of the test are the same for the general and academic tests, while the writing and reading sections differ.

The listening section contains a total of 40 questions and takes 40 minutes. The test-takers hear four different recordings of various scenarios ranging from informal conversations to academic lectures. Each recording is played only once, and the process takes approximately thirty minutes, after which the test-takers are given 10 minutes to transfer their answers to an answer sheet. The questions come in a range of formats from multiple choice and matching to table completion and short answers (IELTS, 2017a).

The reading section also contains 40 questions, and students are given 60 minutes to answer all the questions. There are three sections in total, with the total word count ranging from 2150 to 2750 words across all three (IELTS, 2017a). The academic version of the reading section is made up of three longer readings. They are authentic materials on academic topics that are meant to be understandable by a lay-person and do not require specialized knowledge. The general training version consists of a larger amount of shorter texts that are more general in nature. They are meant to represent everyday English needs, and can consist of advertisements, newspaper articles, schedules, magazine articles, etc. For both versions of the test, students must answer a variety of questions such as multiple choice, matching headings, sentence completion, and true/false. One noticeable difference from other traditional test types is that true/false
questions contain a third option of “not given” if there is not enough relevant information to answer the question.

The writing section of the test takes 60 minutes and requires the test-taker to write approximately 400 words across two different tasks. The first task in the academic version of the test requires the test-taker to describe or explain a graphic such as a chart, table, or graph in 150 words. The second task is to write a 250-word academic style essay in response to a prompt. In the general training version, the first task is to write a 150-word letter to someone either explaining a situation or requesting more information. The second task is to write a 250-word essay in response to a prompt, but the topic is generally more personal and less academic than in the academic format of the test.

The speaking section of the test is the same across academic or general training versions and consists of 11-14 minutes of speaking face-to-face with an examiner. The speaking test is made up of three parts in total. First the test-taker must answer some general questions about themselves and familiar topics such as background or hobbies over the course of four-five minutes. In the second section, the test-takers are given a topic and 60 seconds to prepare themselves, after which they must speak uninterrupted about that topic for two minutes. The examiner then transitions into the third section by asking the test-taker follow-up questions on the topic, which leads to a more formal discussion of the topic. The final section should last roughly four-five minutes as the test-taker expands on their ideas.

2.2 Academic Success

There are many measures used to quantify the construct of academic success. Some may choose to define it as a simple binary measure of completing a program or not, while other more
specific measures of individual course grades may also be examined. For the current study any form of data that can be correlated with scores will be included.

One of the most common measures used is grade point average or GPA (Abunawas, 2014). This can be examined at the semester, year, or overall program level. GPA has been shown to have a high internal reliability, with previous year’s GPA being one of the most effective predictors of current year’s GPA (Bacon & Bean, 2006). Similarly there have been some findings that high school GPA can be a better predictor of post-secondary GPA than many standardized tests (Zahner, Ramsaran, & Steedle, 2012). GPA has been found to also be one of the strongest predictors of academic retention (Millea, Wills, Elder, & Molina, 2018), which is obviously a prerequisite for academic success.

GPAs are not the perfect measure, and there are issues to making use of them as a metric. GPA standards can vary between courses, institutions, and countries (Bacon & Bean, 2006). While they do reflect on success in the course, they may have little external validity outside of the course context. Using simple course grades or GPAs without the necessary context of the course can lead to overgeneralization and chances for errors (Brown, Plonsky, & Teimouri, 2018). For the context of this current study, they are a relied upon measure, but their imperfections and the inherent dangers of using them must be recognized.

2.3 Validity

There have been many studies examining predictive validity of various entrance exams. Kuncel and Hezlett (2007) conducted a synthesis of meta-analyses on standardized tests used for university acceptance at the graduate level. They found that overall standardized tests predict most measures of academic success better than undergraduate GPA, but that a combination of
the two is the strongest. They argue heavily for the overall strengths of standardized testing and the predictive validity they offer. While language tests were not included in their synthesis, most other major standardized tests (SAT, GRE, etc.) had positive correlations with academic success.

One important area of note when examining entrance test validity is that of survivor bias. Given that only students with acceptable scores are offered entrance to the university, there is no available information about how test-takers with scores below the acceptable threshold would perform in the academic context. This has been referred to as a gatekeeping effect, and must be recognized as a form of selection or survivor bias (Woodrow, 2006). For the context of this meta-analysis, it is highly likely that there is little-to-no data for students with band scores below 5.0, as entry cut-off criteria create a floor effect. This results in a non-normal distribution of scores which makes the sample not truly representative of the overall population score distribution.

The IELTS organization checks their own internal validity of the objective tests used throughout the year. While the writing and speaking scores are marked by professionally trained graders, both the reading and listening are objectively scored out of 40 points each. For 2017 the listening sections had an overall Chronbach’s alpha of 0.91, the general training reading was 0.92, and academic reading was 0.90 (IELTS, 2018b). A Chronbach’s alpha of .90 or above is generally considered excellent (Gliem & Gliem, 2003), though some statisticians have raised concerns that a value that high implies redundancy in test items and signals that there should be fewer items (Tavakol & Dennick, 2011). Given that each item it supposed to be a study of English proficiency and not general knowledge, redundancy is not a large concern.
Independent researchers have also examined the validity from different viewpoints. Hyatt (2013) surveyed test-takers in the UK and found that 88% of them felt that the test was an accurate indicator of the necessary academic English proficiency necessary for higher education. Similar views from students and test-takers were found in a previous study, but the staff did not have the same opinions. Coleman, Starfield, and Hagan (2003) found that academic and administrative staff did not agree as much as the students did that IELTS entry scores used by universities were an accurate measure of the necessary abilities to succeed in university. This disparity between student and staff opinion is an interesting one and adds further value to the need for more research into the true predictive validity of IELTS scores.

2.4 Meta-Analysis

The concept of using meta-analytic techniques to aggregate information from multiple studies has existed for over a century, with the first use being credited to a medical study from 1904 (O’Rourke, 2007; Pearson, 1904). The actual usage of the term *meta-analysis* did not come about until Glass’s seminal paper where he applied the same techniques to educational research and coined the term *meta-analysis*, which he succinctly explained as the “analysis of analyses” (Glass, 1976, p. 3). He argued that educational research was expanding at a rapid rate and needed more summaries, especially given the wide variety of contextual differences and conflicting results across studies. Meta-analyses caught on rather quickly and became increasingly popular, by 2005 over 2000 meta-analyses were being published each year in the medical field alone (Sutton & Higgins, 2008). This has since increased to over 9000 medical meta-analyses on PubMed in just the calendar year of 2014 (Ioannidis, 2016). While the domain of second language acquisition and applied linguistics is notably smaller than the medical
domain, there has also been a steady increase in the amount of meta-analyses being conducted in the field in recent years (Oswald & Plonsky, 2010).

Berman and Parker (2002) argue that a meta-analysis should have at least one of two goals: to summarize the data and/or to explain the variability between studies. The current study aims to meet both those goals. There have been a significant number of studies that have examined IELTS predictive power for academic success, many of which have found differing results (Neumann, Padden, & McDonough, 2018). Summarizing the data alone will help to determine the true value of IELTS as a predictor of academic success. Ideally, by examining the moderating variables the conflicting results between studies can also be examined.

2.5 Previous Meta-Analyses on English Test Validity

While there has not yet been a meta-analysis on the predictive validity of IELTS, there have been two separate meta-analyses conducted that examine the links between another English language test, the Test of English as a Foreign Language (TOEFL), and academic success. Coincidentally enough both are PhD dissertations, and a brief overview of each follows.

Wongtrirat (2010) meta-analysed 22 studies from 1987-2009 that looked at TOEFL scores and academic achievement of international students in the USA specifically. An overall weighted average Z score of .181 was found for TOEFL scores and GPA, and .173 for TOEFL scores and course completion, which the author classifies as small predictive validity. No significant difference was found between the undergraduate and graduate levels for either GPA or course completion TOEFL correlates.

The second meta-analysis examining TOEFL scores and academic success took a more broad approach (Abunawas, 2014). A total of 47 effect sizes from 40 articles were included and
an overall effect size of .21 was found, which is considered small. Abunawas also chose to look at the moderating effect of studies conducted within the USA versus elsewhere in the world and found significant differences with studies outside of the USA reporting stronger effects.
Chapter 3 – Integrated Article

3.1 Introduction

As our society becomes more and more globalized, so too does our education system and the diversity of our students. An international education is highly prestigious and sought after in many parts of the world, leading to many students traveling abroad for their post-secondary studies (Altbach, 2015). English is the global lingua franca, and so too is it the language of higher education (Arkoudis, Baik, & Richardson, 2012). For students wanting to study abroad, English skills are a necessity, and to the institutions that they enter, having an effective way to measure those English skills is vital. One of the most popular and common methods of testing English language abilities is through the International English Language Testing System, or IELTS. It is used worldwide for admissions into English speaking universities, with over three millions participants taking the test each year (IELTS, 2017b). With the widespread prevalence of the test for post-secondary admissions, it begs the question: how reliable is IELTS at predicting academic success?

IELTS as a test of English language ability has existed since 1989 (Davies, 2008). Research into various aspects of IELTS has existed for almost as long, with a large variety of focus on different aspects of the test itself, preparatory programs, validity, etc. Much of this research is published by the IELTS organization, who also offer funding to research pertaining to IELTS (IELTS, n.d.). While this resource has helped to inspire and support research into the test itself, there is an argument to be made for the danger of bias and conflicts of interest in relation to the research they fund and publish. Evidence of funding bias, wherein studies sponsored by industry find more beneficial results for that industry, has been found in a variety of studies ranging from cell phone safety (Huss, Egger, Hug, Huwiler-Müntener, & Röösli, 2007) to cost
effectiveness studies (Bell et al., 2006) to a wide range of pharmaceutical efficacy studies (Lundh, Lexchin, Mintzes, Schroll, & Bero, 2017). Even among review articles such as syntheses and meta-analyses evidence of stronger results from authors with ties to industry have been found (Barnes & Bero, 1998). Given that the IELTS organization has a vested interest in positive results, another question must be asked: Do studies funded or supported by the IELTS organization show stronger results than independent studies?

The last few decades have seen numerous studies examining the predictive validity of IELTS, published both by the IELTS organization and peer-reviewed journals (Neumann et al., 2018). There have been mixed results between the studies, with many finding non-significant effect sizes or weak-moderate correlations between both overall IELTS scores and academic achievement, as well as individual skill bands and academic achievement (e.g., Arrigoni & Clark, 2015; Kerstjens & Nery, 2000; Oliver, Vanderford, & Grote, 2012). One very recent and comprehensive study found one of the highest correlations to date, \( r = .509 \) (Müller & Daller, 2019). These discrepancies in the literature are troublesome and need further examination. One method of examining and attempting to explain such varying results is through a meta-analysis. The purpose of a meta-analysis is to explain the variability among studies and summarize the data (Berman & Parker, 2002).

The concept of a meta-analysis in education research was first introduced over 40 years ago (Glass, 1976). Meta-analyses have been gaining popularity and in recent years are becoming more and more prevalent within the field of second language acquisition (Plonsky & Brown, 2015). There has been a recent push in the second language research community for more meta-analyses (Plonsky & Oswald, 2012). By applying statistical analyses to corpus of data within the field, patterns and trends can emerge to help shed greater light on concepts and constructs as a
whole. Meta-analyses are beneficial for determining overall means of interventions on populations, variability across studies, and the effects (and existence of) various moderator variables (Field & Gillett, 2010).

While the overall scores are the most commonly examined measure, the individual scores in the different macro skills are also of importance. IELTS gives the test-takers an individual score for listening, reading, writing, and speaking, as well as an overall composite score. Certain macro skills may be more important than others in academic contexts. For example, Kerstjens and Nery (2000) found significant correlations of $r = .262$ for the reading score and grade point average (GPA) and $r = .204$ for the writing score and GPA among post-secondary students, but no significant results for listening or speaking. However Woodrow (2006) found correlations of $r = .39$ for speaking, $r = .33$ for writing, and $r = .35$ for listening when compared to GPA, while reading scores did not have significant results. Woodrow attributes this difference to be potentially caused by differences in the curriculum and courses covered. Woodrow’s study also focused on education postgraduate students, who have been shown to require a high level of language proficiency (Faez & Karas, 2017). Nevertheless, the sub-scores offered for each individual macro skill are important and must be examined.

As Woodrow (2006) claimed, not all academic programs are created equally, nor will they be equally demanding on a student’s language abilities. Some programs may require more use of high-level reading and writing abilities, while others may focus more on numbers and formulae. Similarly, graduate level study is much more likely to be intensive than the undergraduate level. Building on these assumptions, we can also surmise that a student pursuing a PhD in journalism would face a much higher language burden than an undergraduate engineering degree. Due to these differences, it would be improper to treat all programs and
study levels as equivalent, as there will most likely be strong differences between them. Therefore, both the discipline studied and the level of study should be examined for their moderating effects on any results for predictive validity.

In addition to variances across fields and levels of study, there also may be variances across individual institutions. Measuring individual institutional differences would be near impossible at a large scale, but generalizations can be made. Abunawas (2014) found that GPA and TOEFL scores were more strongly correlated in studies done outside of the USA. This raises some interesting implications with regards to rigour or English necessity in programs outside of the “Anglosphere”, or English-speaking world. Many universities in non-English speaking countries offer programs in English and require proof of English abilities (IELTS, 2017b). Given that IELTS is used for post-secondary entrance worldwide, it is worth examining the differences that can come from the context of English in the country of study.

While IELTS score cut-off points exist at every institution, many universities are still looking for ways to recruit and accept more international students (Sá & Sabzalieva, 2018). One type of program that is becoming more common is the “top-up” program, sometimes also referred to as additional EAP, conditional entry, foundation courses, etc. These programs allow universities to still recruit students with lower English abilities and give the students a chance to study and live in the culture before beginning their formal education experience in their programs of study. It benefits the students through conditional admission instead of an outright rejection, and it benefits the institutions by giving them more academically prepared students, who also happen to be paying extra tuition before beginning their program. Schoepp (2018) examined the GPAs of students that participated in such programs and compared them to the GPAs of international students who had the requisite IELTS scores for traditional entry. He
found that the students that took part in the top-up programs had significantly lower GPAs than those who didn’t. He even uses this as evidence to state that “This is a powerful demonstration that IELTS scores have a predictive validity towards academic success.” (Schoepp, 2018, p. 281). Similarly, Johnson & Tweedie (2017) found that initial IELTS and TOEFL scores were better predictors of GPA than grades received from top-up programs. Given the prevalence of these top-up programs being offered worldwide, it is worth exploring their relationship with predictive validity and GPA.

Post-secondary enrolment is an ongoing process, lasting multiple semesters or years, and yet many studies focus solely on first semester results (Woodrow, 2006). To truly know if students are prepared for a lengthy program, their performance must be evaluated at multiple time points. Student retention is one of the key factors universities are examining in improving their graduation rates (Millea et al., 2018). Therefore, it is of vital importance to examine the power that IELTS scores have in predicting success at different timepoints throughout a student’s academic journey.

3.1.1 Research Questions

The following research questions were developed based on the previous literature and goals of this study:

1. How valid are IELTS scores at predicting academic success, as measured by GPA?
2. Do studies that were published by the IELTS organization itself report stronger results than those published by third-parties?
3. To what degree are each of the four macro skills (listening, speaking, reading, and writing) associated with academic success?
4. What moderating effects exist across:
   a. Field of study
   b. Level of study (graduate vs undergraduate)
   c. Role of English (native or foreign language) in the country of study
   d. Completion of a “top-up” program or entry through IELTS scores alone
   e. Different time periods in their program

3.2 Literature Review

3.2.1 IELTS

The IELTS test was first introduced in 1989, as an updated version of the previous ELTS (English Language Testing Service), which was built upon the original EPTB (English Proficiency Test Battery) from the 1960’s (Davies, 2008). It has quickly become one of the most popular English tests worldwide, with over three million people taking the test in 2016 (IELTS, 2017b). It consists of four sections: listening, speaking, reading, and writing. Test-takers are scored on a band ranging from a “non user” of 1 to an “expert user” of 9 for each of the individual skills, as well as an overall composite score. Thousands of post-secondary institutions worldwide use IELTS for proof of English competency for speakers of other languages.

Validation of English tests has been taking place for well over 50 years (Gue & Holdaway, 1973). The IELTS test is no exception, it is continually validated, both internally and externally. IELTS publishes annual data on the validity of their measures, for example both their listening and reading tests in 2017 had Chronbach’s alpha of 0.9 or higher, which is considered extremely high (IELTS, 2018b). Surveys of test-takers find that the vast majority of them believe that IELTS scores are an accurate indicator of their English abilities (Hyatt, 2013).
3.2.2 Meta-Analysis

Meta-analyses have been a cornerstone of educational research for over 40 years (Glass, 1976). They are a synthesis of already existing research on a topic, which allows for summarizing existing data, as well as explaining variability that exists between studies (Berman and Parker, 2002). While the current study is a novel one, there have been two prior meta-analyses examining standardized English tests and academic success, both of which looked at TOEFL scores (Abunawas, 2014; Wongtrirat, 2010). Both of these PhD dissertations found overall small effects between TOEFL scores for admission and subsequent GPAs in relevant post-secondary institutions.

3.2.3 Conducting a Meta-Analysis

When conducting a meta-analysis, the literature review is one of the most important steps. The databases that are searched must be carefully chosen so as to not miss any relevant studies. In’nami and Koizumi (2010) found that LLBA was the only database to include all 24 applied linguistics journals. ERIC and MLA International Bibliography covered 22 of the 24, Linguistics Abstract included 20, and Scopus included 18. The other databases they examined all covered significantly fewer journals, as of 2010 at least. They offer the following advice: “a combination of either ERIC and Linguistics Abstracts or MLA and Linguistics Abstracts must be used, and extra attention must be paid in conducting manual searches of journals that are missing from these databases.” (In’nami & Koizumi, 2010, p. 178).

In a more recent study, Plonsky and Brown (2015) examined which databases were most commonly used in L2 Meta-analyses. They found that ERIC was the most popular by far, with
81% of L2 meta-analyses making use of it. This was followed by LLBA (49%), PsycINFO (41%), ProQuest Dissertations and Theses (39%), Web of Science (14%), and Google Scholar (10%). There were other databases included in some meta-analyses, but they were included in less than 10% of examined meta-analyses overall.

Another database that is of particular use is the IELTS Research Reports database, offered by the IELTS organization itself and available at https://www.ielts.org/teaching-and-research/research-reports. These reports are of studies related to IELTS and supported by the organization. The database itself contains over 100 separate reports, many of which focus on different forms of validity. All studies found through this database were coded as such and examined for potential impacts of bias in more detail.

Another search method that was utilized was the backwards search or snowball technique (Wohlin, 2014). This consists of checking the references of studies to find other similar studies that they have cited that may be of use. A modified version known as forward snowballing was also conducted, where papers that cite the original study are examined.

Once the literature search is completed, the researcher must then begin coding the studies. Plonsky and Oswald (2015) advise, “The meta-analyst should be prepared to pivot, revise, and repilot the coding sheet before and even during the coding process” (p. 111). The initial design for a coding sheet should include the necessary identification variables such as author(s), year, publication type, journal impact factor, study design, setting, etc. For the present study, as many variables about the participants as possible were included, such as age, education level, first and target languages, context, etc. The most important points of focus were on the scores themselves: the IELTS scores, success measures, and subsequent correlations.
Once coding is complete, the data must be analyzed. There are multiple programs specifically designed for meta-analyses, each with their own strengths and weaknesses. A 2007 study compared six different programs designed for conducting meta-analyses and found that CMA, alongside another program named MIX, had the highest usability, with the authors also stating: “CMA was generally most versatile, in particular in options for analysis of various types of data.” (Bax, Yu, Ikeda, & Moons, 2007, p. 7).

During analysis, the author must make multiple important decisions, one of the most important is which model to use. The two most common modern models used for meta-analyses are the fixed-effects (FE) and random-effects (RE) models, which have some important differences to consider when designing a meta-analysis (Plonsky & Oswald, 2015). Schmidt, Oh, and Hayes (2009) summarize the main difference as FE models assume that population parameters are equivalent across studies, while RE models assume that there are differences in population values used. They re-analysed a series of FE meta-analyses using RE techniques and found an average underestimation of confidence intervals by 52% across the FE meta-analyses, as well as the 95% confidence intervals in FE meta-analyses to actually be 56% confidence intervals on average. They recommend the use of RE model meta-analyses in virtually all situations, as do Plonsky and Oswald (2015), therefore a random-effect model was used for this study.

Once the data has been analyzed, it must be interpreted. Traditionally researchers have used Cohen’s suggested initial cut-off values of $r = .1$ for small, $r = .3$ for medium, and $r = .5$ for large effect sizes in correlational research (Cohen, 1988). However, these values are not meant
to be treated as equivalent across all domains and special considerations must be taken. Plonsky and Oswald (2014) analyzed 175 studies across 20 correlational meta-analyses and found that the original cut-off points do not match well within the domain of second language research. They advise instead to use the benchmarks of $r = .25$ for small, $r = .40$ for medium, and $r = .60$ for large effect sizes.

A vital part of any meta-analysis is the test for heterogeneity, which examines whether all studies are truly evaluating the same effect. If effect sizes are to be combined, they must be comparable, and two common methods to measure their similarity are Cochran’s $Q$ and the more recent $I^2$ (Higgins, Thompson, Deeks, & Altman, 2003). Cochran’s $Q$ is used to determine if there is heterogeneity or not, while $I^2$ shows the extent of any heterogeneity. Huedo-Medina, Sánchez-Meca, Marín-Martínez, and Botella (2006) conducted a Monte Carlo simulation to compare the effectiveness of the two methods at measuring heterogeneity and found they were roughly equally effective, but that $I^2$ provides more context in a much more concise manner, as $I^2$ values are meant to directly signify the percentage of heterogeneity present (for example, an $I^2$ value of 25.63 would indicate 25.63% heterogeneity across studies).

3.3 Method

3.3.1 Defining the Research Domain

As previously mentioned, the focus of this meta-analysis is on IELTS scores and their relation to academic success. Academic success can be defined in a number of different ways such as first year GPA, overall GPA, course completion, etc. Measures of GPA at different points in time are the most common measures of academic success found in the literature (Abunawas, 2014; Zahner et al., 2012).
In an attempt to be as inclusive as possible, a broad level of inclusion criteria and minimal exclusion criteria were applied

Included studies must:

1. Be of quantitative design, or include a quantitative component
2. Examine the relationship between IELTS raw scores and some measure of academic success at English-medium universities
3. Be available in English
4. Be published after 1995, or make use of IELTS scores from after the revisions in 1995

The inclusion criteria were left very broad and general on purpose. The study sought to take somewhat of an exploratory approach and begin with a large amount of included studies. Berman and Parker (2002) recommend inclusion criteria include the type of study, subject characteristics, treatment modalities, and outcome measures. These inclusion criteria mostly meet those, as subject characteristics are not a concern for the current study.

Studies were excluded that:

1. Made use of equivalent unverifiable categories to IELTS scores (for example reporting CEFR scores or general categories such as “intermediate” instead of actual IELTS bands)
2. Re-used datasets from previously published and included studies

3.3.2 Conducting the Literature Search

A variety of different techniques previously mentioned were used to find and identify appropriate literature for inclusion. While the primary method of finding appropriate literature to include is from a traditional database search, multiple other methods were also included.
The results of the database literature search can be seen in Table 1.

Table 1

*Literature Search Results*

<table>
<thead>
<tr>
<th>Database</th>
<th>Keyword(s)</th>
<th>Results found</th>
</tr>
</thead>
<tbody>
<tr>
<td>IELTS Research Reports</td>
<td><em>Manual Review</em></td>
<td>101</td>
</tr>
<tr>
<td>ERIC</td>
<td>IELTS</td>
<td>143</td>
</tr>
<tr>
<td>LLBA</td>
<td>IELTS</td>
<td>175</td>
</tr>
<tr>
<td>Proquest Dissertations and Theses</td>
<td>IELTS</td>
<td>71</td>
</tr>
<tr>
<td>PsycInfo</td>
<td>IELTS</td>
<td>69</td>
</tr>
<tr>
<td>Scopus</td>
<td>IELTS GPA</td>
<td>6</td>
</tr>
<tr>
<td><strong>Individual Journals</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Higher Education Research &amp; Development</td>
<td>IELTS</td>
<td>31</td>
</tr>
<tr>
<td>Higher Education Quarterly</td>
<td>IELTS</td>
<td>5</td>
</tr>
<tr>
<td>Studies in Higher Education</td>
<td>IELTS</td>
<td>14</td>
</tr>
<tr>
<td>Assessment in Education: Principles, Policy &amp; Practice</td>
<td>IELTS</td>
<td>11</td>
</tr>
<tr>
<td>Language Testing</td>
<td>GPA</td>
<td>15</td>
</tr>
<tr>
<td>University of Melbourne Theses Repository</td>
<td>IELTS</td>
<td>59</td>
</tr>
<tr>
<td><strong>Multi-Databases</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Google Scholar</td>
<td>IELTS GPA</td>
<td>2280</td>
</tr>
<tr>
<td>Western Libraries</td>
<td>IELTS GPA</td>
<td>285</td>
</tr>
</tbody>
</table>

*Note.* A manual review of each individual IELTS Research Report was conducted, no search terms were required.

Given that each of the databases returned less than 200 results each, every individual article was reviewed by title and abstract. Due to the use of the IELTS acronym in the medical field, many of the results on Google Scholar were deemed irrelevant, even with the addition of the search term “GPA”. The first 10 pages of results for “IELTS GPA” were examined, after which point the articles relation to the topic dropped significantly. Across all of the databases...
and catalogues mentioned in Table 1, a total of 3265 studies were found, not accounting for overlap.

Based on a review of titles and abstracts a total of 42 studies were originally identified for inclusion. Throughout the coding process 14 of these studies were removed for a variety of reasons. Despite delays in publishing, some still used data that was from prior to the 1995 revision (e.g., Ferguson & White, 1998; Huong, 2001), some had both GPA and IELTS scores but did not report any correlations (e.g., Kiany, 1998; Lloyd-Jones, Neame, & Medaney, 2011; Tweedie & Chu, 2017), some used practice tests or unofficial IELTS measures (e.g., Daller & Phelan, 2013; Khemakhem, 2016), some used other measures than GPA that were less reliable such as observation scores of language use (e.g., Bayliss & Ingram, 2006; Ingram & Bayliss, 2007; Paul, 2007), some had mixed groups of IELTS and TOEFL combined for one correlation (e.g., Floyd, 2015), and some did not include enough detail, even after email correspondence (e.g., Assylbekov, 2013).

3.3.3 Coding

The coding process consisted of constructing a coding sheet in Microsoft Excel and then entering the relevant data from the studies into it. Early versions of the coding sheet were shown to two different colleagues familiar with the meta-analysis process, and minor revisions were made. A wide and rich variety of data was captured, including other details that could be of interest for further analysis such as first language of test-takers, age-range, comparison to other language tests, etc. All of this data was categorized and entered into the Excel sheet.

As this meta-analysis was conducted for a graduate thesis, the coding was entirely conducted by a sole coder, the author. While this is not ideal methodology for a meta-analysis,
the context of the research required it. For best results, multiple coders should code each study (Schmidt & Hunter, 2015). Nonetheless, three colleagues were approached and asked to assist with double coding. Raters 1 and 2 have both published multiple meta-analytic studies in various journals as well as have presented their findings at international conferences, they are both quite familiar with meta-analyses. Rater 3 was not as familiar with meta-analytic research and was therefore given a smaller amount of studies to code.

The three raters were trained on the coding book within Microsoft Excel, shown multiple examples of other coded studies, and then left to do their coding without the bias of the researcher present. A total of 14 studies were coded by the additional raters, and their results were compared cell by cell with the author’s original coding sheet and results, excluding bibliographical data. Rater 1 had 108/111 or 97.3% matching cells, Rater 2 had 63/66 or 95.45%, and Rater 3 had 56/56 or 100% matching, for a cumulative total of 227/233 matching cells or 97.42% overall interrater reliability. When discrepancies arose, the raters discussed them and reached a consensus. This resulted in 50% of the studies being coded by multiple raters, as well as external validation of the coding sheet itself.

A total of 28 studies remained after removing inappropriate studies, and all were coded fully. During this process two more studies that initially seemed appropriate were removed. Arrigoni and Clark (2015) was a very promising study with many effect sizes to examine, however they were comparing IELTS scores to what seem to be EAP courses. The authors do argue that the courses offer more than just English language skills but given that these courses were English skill based in nature, and do not seem to be a requirement for native-speaking students, it did not seem appropriate to include this study. Additionally, Breeze and Miller (2011) made use of only listening scores, of which they administered the (presumably practice)
tests on their own. Their report is published by IELTS, so an argument could be made for the potential validity of their scores but given that they were not fully official IELTS test scores, the study was removed.

Three of the studies (Feast, 2002; Phakiti, 2008; Thorpe, Snell, Davey-Evans, & Talman, 2017) made use of regression analysis as opposed to correlations, and thus reported beta coefficients (β) instead of r values. In order to analyse these results, the data needed to be converted. Peterson and Brown (2005) offer the following formula to convert beta coefficients into r values:

\[ r = \beta + .05\lambda \]

Where λ is equal to 1 if the β value is non-negative and 0 if the β value is negative. They do warn that this formula works best with β values of ±.50, which is valid for two of the three studies, the third study (Thorpe et al., 2017) reported beta coefficients much higher, and thus a valid r value was not calculable. Therefore, this study was removed from analysis.

The remaining 25 studies were inputted into the Comprehensive Meta Analysis software (version 3.3.070) and analysed. A total of 3475 students from 31 subgroups across 24 of the studies reported on overall effect sizes; one study, Humphreys et al. (2012), did not report an overall effect size, but just effect sizes for each of the four macro skills. This study was used for examination of the individual macro skills, but not included in the other analyses. Including the 51 students from their study brings the total included students up to 3526 over 32 subgroups from 25 studies.

3.3.4 Analysis
The meta-analytic software *Comprehensive Meta Analysis*, or CMA (available at [https://www.meta-analysis.com/](https://www.meta-analysis.com/)) was used for analysis of the data. It is a powerful tool that allows for great control over the data as well as assistance with the meta-analysis process itself. It allows a more user-friendly approach than using *R* or *SPSS*, as it is designed solely for meta-analyses.

### 3.4 Results

An overall effect size of $r = .299^{* *}$, $p = .000$ was found, 95% confidence interval of [.207, .386], and $Z = 6.155$. However, it must be noted that this is a composite of Pearson’s correlations, Spearman’s rho, regression beta coefficients, and a singular t-test. While both Pearson’s correlation and Spearman’s rho report correlations, they are distinctly different statistics and must be acknowledged as such.

There are formulae to convert the two measures between each other. Lajeunesse (2013) recommends the following formula to convert Spearman’s rho ($\rho$) to Pearson’s $r$ (if the sample size is under 90, otherwise he equates them):

$$r = 2\sin \left( \frac{\pi \rho}{6} \right)$$

And de Winter et al. (2016) make use of the following conversion for Pearson’s $r$ to Spearman’s rho:

$$\rho = \frac{6}{\pi} \arcsin \left( \frac{r}{2} \right)$$

However, one of the coded studies did report both measures for the same correlation. Breeze & Miller (2011) reported an overall Pearson’s $r$ of .344 and a Spearman’s rho of .408.
When using those values in the above formulae, the results do not match their reported values. This is most likely due to the non-normal distribution of the data, as there is the previously mentioned floor effect of IELTS admissions cut-off scores. Therefore, these formulae were deemed unreliable for this meta-analysis, and conversions were not made.

To determine if the two correlations offered significantly different results, a moderator analysis was conducted between the analysis types. For the sake of posterity, the regression and t-test data were also included in the analysis as well, the results of which can be seen in Table 2.

Table 2

<table>
<thead>
<tr>
<th>Analysis type</th>
<th>k</th>
<th>$r$</th>
<th>95% CL</th>
<th>Z</th>
<th>$p$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pearson’s r</td>
<td>17</td>
<td>.239**</td>
<td>[.165, .310]</td>
<td>6.200</td>
<td>.000</td>
</tr>
<tr>
<td>Spearman’s rho</td>
<td>4</td>
<td>.586**</td>
<td>[.326, .764]</td>
<td>3.948</td>
<td>.000</td>
</tr>
<tr>
<td>Regression</td>
<td>2</td>
<td>.371**</td>
<td>[.237, .491]</td>
<td>5.159</td>
<td>.000</td>
</tr>
<tr>
<td>t-test</td>
<td>1</td>
<td>.140*</td>
<td>[.030, .246]</td>
<td>2.498</td>
<td>.012</td>
</tr>
</tbody>
</table>

*Note. $Q_{between}= 13.378, df= 23, p=.004*  

Given that $Q_{between}$ reaches significance at the $p<.01$ level, these are significant differences, therefore Pearson’s $r$ alone was chosen to focus on. While removing the Spearman’s rho results does eliminate a significant amount of studies, four out of 25, it would be inappropriate to treat the two different correlations as equivalent. Conducting secondary analysis of each measure for both correlations was examined but given that only four studies in total used Spearman’s rho, the sample and resulting power would be too low.
Additionally, the two regression studies appear to be significantly distinct from the Pearson’s $r$ results. This could be due to the complicated nature of beta coefficients and the many other variables that can affect them within their relevant models (Peterson & Brown, 2005). Due to the significant difference and dangers of including them, they were also removed from the study. Finally, the single t-test was also removed in order to focus solely on studies that reported a Pearson’s correlation. While removing so many studies does cause issues of comprehensiveness, it raises the strength of the results by focusing solely on one analysis and variable type.

One final note must be made that all studies except for one (Humphreys et al., 2012) reported a correlation using the overall composite score, as well as occasional sub-scores for individual macro skills. Humphreys et al. (2012) reported only the sub-scores with no composite score correlation. The composite score is similar to an average, and arguments could be made for averaging the other scores to impute the data (Lajeunesse, 2013). However, when trialling averaging imputed data with other studies, the imputed data did not match exactly with the overall correlation. This could be due to rounding with overall composite scores, as all IELTS bands are rounded up or down to the nearest half-band at the very most, which would provide less accuracy. To err on the side of caution and not include inaccurate data, the overall scores were not imputed. Therefore, results from this study are only included when individual skills are examined for research question 3 and were not part of any of the overall calculations or other moderator analyses.

3.4.1 Descriptive Statistics
A total of 18 studies were included in the final meta-analysis, and the descriptive statistics are outlined in Table 3.

Table 3

**Descriptive Statistics**

<table>
<thead>
<tr>
<th>Authors</th>
<th>Year</th>
<th>Publication</th>
<th>N</th>
<th>Country of Study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cotton &amp; Conrow</td>
<td>1998</td>
<td>IELTS Report</td>
<td>26</td>
<td>Australia</td>
</tr>
<tr>
<td>Hill, Storch, &amp; Lynch</td>
<td>1999</td>
<td>IELTS Report</td>
<td>35</td>
<td>Australia</td>
</tr>
<tr>
<td>Kerstjens &amp; Nery</td>
<td>2000</td>
<td>IELTS Report</td>
<td>113</td>
<td>Australia</td>
</tr>
<tr>
<td>Dooey &amp; Oliver</td>
<td>2002</td>
<td>Journal</td>
<td>65</td>
<td>Australia</td>
</tr>
<tr>
<td>Woodrow</td>
<td>2006</td>
<td>Journal</td>
<td>62</td>
<td>Australia</td>
</tr>
<tr>
<td>Yen &amp; Kuzma</td>
<td>2009</td>
<td>Journal</td>
<td>61</td>
<td>UK</td>
</tr>
<tr>
<td>Avdi</td>
<td>2011</td>
<td>Journal</td>
<td>40</td>
<td>Australia</td>
</tr>
<tr>
<td>Humphreys, Haugh, Fenton-Smith, Lobo,</td>
<td>2012</td>
<td>IELTS Report</td>
<td>51</td>
<td>Australia</td>
</tr>
<tr>
<td>Oliver, Vanderford, &amp; Grote</td>
<td>2012</td>
<td>Journal</td>
<td>376</td>
<td>Australia</td>
</tr>
<tr>
<td>Arcuino</td>
<td>2013</td>
<td>PhD Thesis</td>
<td>29</td>
<td>USA</td>
</tr>
<tr>
<td>Garinger &amp; Schoepp</td>
<td>2013</td>
<td>Journal</td>
<td>181</td>
<td>UAE</td>
</tr>
<tr>
<td>Riazi</td>
<td>2013</td>
<td>Journal</td>
<td>60</td>
<td>Australia</td>
</tr>
<tr>
<td>Yixin &amp; Daller</td>
<td>2014</td>
<td>Paper</td>
<td>57</td>
<td>UK</td>
</tr>
<tr>
<td>Schoepp &amp; Garinger</td>
<td>2016</td>
<td>Journal</td>
<td>241</td>
<td>UAE</td>
</tr>
<tr>
<td>Johnson &amp; Tweedie</td>
<td>2017</td>
<td>Chapter</td>
<td>33</td>
<td>Qatar (Canada)</td>
</tr>
<tr>
<td>Neumann, Padden, &amp; McDonough</td>
<td>2018</td>
<td>Journal</td>
<td>54</td>
<td>Quebec</td>
</tr>
<tr>
<td>Schoepp</td>
<td>2018</td>
<td>Journal</td>
<td>953</td>
<td>UAE</td>
</tr>
<tr>
<td>Müller &amp; Daller</td>
<td>2019</td>
<td>Journal</td>
<td>46</td>
<td>Australia</td>
</tr>
</tbody>
</table>

Overall there were 11 journal articles, four IELTS research reports, one PhD thesis, one conference paper, and one book chapter. The studies were conducted across multiple continents, with the majority taking place in Australia. There were four authors with multiple works included: Rhonda Oliver, Michael Daller, Dawn Garinger, and Kevin Schoepp, but all of their datasets appear to be unique. Participant pool sizes ranged widely from 26 – 953 students per study, and the overall range of research covers more than 20 years; from 1998-2019. There has
been an overall slight upward trend in number of publications over time. This can be seen in Figure 1.

![Publications by Year](image)

**Figure 1.** Publications by year

### 3.4.2 Research Question 1: Overall Effect

An overall effect size of $r = .227$ was found, 95% confidence intervals of [.162, .290], $Z=6.677$, $p = .000$, from 29 effect sizes across 17 studies with 2432 students total. This is not quite a small effect size, as it does not reach the updated .25 cut-off point (Plonsky & Oswald, 2014). The 95% confidence interval does pass it by a significant margin, so it is safe to say that this effect size is approaching small, or nearly small. The overall results and effect sizes of each individual study can be seen in the forest plot in Figure 2.

Regarding heterogeneity, a Q-value of 45.353 was calculated, $p=.004$. This leads to an $I^2$ value of 49.286, which is a moderate level of heterogeneity (Higgins et al., 2003), and discussed further in the discussion section.
Of the 17 included studies, one (Cotton & Conrow, 1998) is bordering on being a negative outlier; the upper bound of its 95% confidence level is equal to the lower bound of the overall effect size 95% confidence level at a value of .162. Upon closer examination at four decimal places, it is slightly higher (.1624 for the individual study, .1616 for overall), so the confidence intervals do intersect, albeit to an almost negligible degree. Viechtbauer and Cheung (2010) warn against simply deleting outliers and suggest looking deeper into the numbers. Upon a more in-depth review of the article (Cotton & Conrow, 1998), one issue immediately stands out. This study has an overall relatively small sample size of 26, which makes it the smallest study included in the meta-analysis. The authors acknowledge this as a potential issue multiple
times throughout the paper, and repeatedly advise that their results are based upon such a small
group. They draw attention to a few individual students who were outliers within the study itself
and heavily affected the overall data, but were included nonetheless. Additionally, the authors
note that theirs is the first predictive validity study of IELTS done after the 1995 revision, and
perhaps there was still some fine-tuning going on with the early scores.

While the weighting calculations take the small sample size into consideration, an
additional analysis was run excluding this study to determine if there was a significant effect.
With it removed, the overall r value raises to \( r = .236 \) with a Z value of 7.192. While this is an
increase, it is not one of overly significant proportions, therefore the study was included for all
relevant analyses.

3.4.3 Research Question 2: Funding Bias

To answer the second research question, if there was a publishing bias from the IELTS
organization itself, a moderator analysis was conducted between those studies published by
IELTS and those published elsewhere. For this analysis only, studies themselves were used as
the unit of analysis. For all other analyses individual subgroups were examined, but this analysis
is focused on the meta-level study variable itself, so the subgroups within the studies were
averaged for an overall study measure. The results can be seen in Table 4.

Table 4

<table>
<thead>
<tr>
<th>Publisher</th>
<th>( k )</th>
<th>( r )</th>
<th>95% CL</th>
<th>( Z )</th>
<th>( p )</th>
</tr>
</thead>
<tbody>
<tr>
<td>IELTS</td>
<td>3</td>
<td>.137</td>
<td>[-.277, .508]</td>
<td>0.639</td>
<td>.523</td>
</tr>
<tr>
<td>Elsewhere/Unpublished</td>
<td>14</td>
<td>.249**</td>
<td>[.180, .316]</td>
<td>6.896</td>
<td>.000</td>
</tr>
</tbody>
</table>
Overall the small number of studies and lack of statistical significance results in few conclusions that can be reliably drawn from this data. While there is no evidence of higher results from IELTS Research Reports, there is also no evidence of a lack of bias.

### 3.4.4 Research Question 3: Subscores

To examine the relationship between individual skills IELTS scores and GPA, four separate analyses were conducted examining each outcome. Most of the studies included both overall or composite scores alongside individual skill scores, but four of the studies (Arcuino, 2013; Hill, Storch, & Lynch, 1999; Müller & Daller, 2019; Yixin & Daller, 2014) only included the composite scores, so they were not included in the results in Table 5.

<table>
<thead>
<tr>
<th>Skill</th>
<th>k (subgroups)</th>
<th>r</th>
<th>95% CL</th>
<th>Z</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Listening</td>
<td>17</td>
<td>.157**</td>
<td>[.078, .233]</td>
<td>3.887</td>
<td>.000</td>
</tr>
<tr>
<td>Speaking</td>
<td>17</td>
<td>.086*</td>
<td>[-.005, .167]</td>
<td>2.076</td>
<td>.038</td>
</tr>
<tr>
<td>Reading</td>
<td>17</td>
<td>.215**</td>
<td>[.127, .300]</td>
<td>4.723</td>
<td>.000</td>
</tr>
<tr>
<td>Writing</td>
<td>17</td>
<td>.153**</td>
<td>[.089, .215]</td>
<td>4.659</td>
<td>.000</td>
</tr>
</tbody>
</table>

While each of the results are significant, none of them cross the $r = .250$ threshold necessary to be considered a small effect size. The reading $r = .215$ is the strongest of the subskills, and possible reasons behind this will be discussed in the discussion section.

### 3.4.5 Research Question 4A: Field of Study
To examine the first moderator, studies were coded based on the program the students were enrolled in. Many studies made use of overall mixed samples and could not be broken down for individual analysis. Those that did were classified into broad categories and the results of the moderator analysis can be seen in Table 6.

Table 6

*Moderator Analysis – Field of Study*

<table>
<thead>
<tr>
<th>Major</th>
<th>k (subgroups)</th>
<th>r</th>
<th>95% CL</th>
<th>Z</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Business</td>
<td>5</td>
<td>.226*</td>
<td>[-.034, .404]</td>
<td>2.273</td>
<td>.023</td>
</tr>
<tr>
<td>Healthcare</td>
<td>3</td>
<td>.336*</td>
<td>[.071, .557]</td>
<td>2.458</td>
<td>.014</td>
</tr>
<tr>
<td>Social</td>
<td>1</td>
<td>.400**</td>
<td>[.167, .591]</td>
<td>3.254</td>
<td>.001</td>
</tr>
<tr>
<td>STEM</td>
<td>2</td>
<td>.009</td>
<td>[-.344, .359]</td>
<td>0.046</td>
<td>.964</td>
</tr>
</tbody>
</table>

*Note. Q_{between} = 3.796, df = 10, p = .284*

Given the relatively small amount of studies in each category, no strong definitive conclusions can be drawn. Studies using business students seem to be almost perfectly representative of the overall sample however, with business majors have a value of \( r = .226 \), while the overall result was \( r = .227 \).

3.4.6 Research Question 4B: Level of Study

To examine the predictive validity of IELTS at different level of education, studies were coded for study level of the relevant student population. Categories were initially quite broad but then collapsed down to simply graduate or undergraduate level. Any program that required prior post-secondary education was classified as graduate level, such as teacher’s college programs...
requiring a prior bachelor’s degree. A number of studies made use of mixed samples containing both undergraduate and graduate students, they were not included in this moderator analysis, the results of which can be seen in Table 7.

Table 7

*Moderator Analysis – Level of Study*

<table>
<thead>
<tr>
<th>Level of Study</th>
<th>k (subgroups)</th>
<th>r</th>
<th>95% CL</th>
<th>Z</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Undergraduate</td>
<td>14</td>
<td>.202**</td>
<td>[.127, .275]</td>
<td>5.206</td>
<td>.000</td>
</tr>
<tr>
<td>Graduate</td>
<td>5</td>
<td>.271**</td>
<td>[.127, .404]</td>
<td>3.617</td>
<td>.000</td>
</tr>
</tbody>
</table>

*Note. Qbetween= 0.711, df= 18, p=.399*

While the values themselves seem to show a slightly higher correlation for graduate study as opposed to undergraduate, the Qbetween value is quite low and was not significant, so it cannot be reliably assumed to be a moderating variable.

### 3.4.7 Research Question 4C: Study Location

Study location was coded as another moderator, dividing studies into completed in an English-speaking country or not. For this coding, Australia, the USA, and the UK were considered English-speaking countries, while the UAE and Qatar were considered non-English speaking. Given that Quebec’s official language is solely French, it was treated as non-English speaking for initial coding, the results can be seen in Table 8.

Table 8

*Moderator Analysis – Country of Study*

<table>
<thead>
<tr>
<th>Country</th>
<th>k (subgroups)</th>
<th>r</th>
<th>95% CL</th>
<th>Z</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>English</td>
<td>16</td>
<td>.226**</td>
<td>[.124, .323]</td>
<td>4.293</td>
<td>.000</td>
</tr>
</tbody>
</table>
While there does not appear to be a significant difference, the English speaking correlation is virtually the same as the overall correlation, \( r = .226 \) and \( r = .227 \) respectively. Therefore the studies conducted in English countries may be seen as more representative of the true predictive validity than those outside.

Analysis was also conducted with the Quebec study removed, as it is in Canada and a unique language context in its own right. That analysis can be seen in Table 9.

Table 9

*Moderator Analysis – Country of Study without Quebec*

<table>
<thead>
<tr>
<th>Country</th>
<th>( k ) (subgroups)</th>
<th>( r )</th>
<th>95% CL</th>
<th>( Z )</th>
<th>( p )</th>
</tr>
</thead>
<tbody>
<tr>
<td>English</td>
<td>16</td>
<td>.226**</td>
<td>[.124, .323</td>
<td>4.293</td>
<td>.000</td>
</tr>
<tr>
<td>Non-English</td>
<td>6</td>
<td>.192**</td>
<td>[.121, .260</td>
<td>5.249</td>
<td>.000</td>
</tr>
</tbody>
</table>

*Note. Q*\(_\text{between} = 0.306, df = 21, p = .580*

Overall the \( Q_{\text{between}} \) value does not change significantly, and even loses power. The Quebec study does not seem to have a significant effect on changing the results of this analysis.

### 3.4.8 Research Question 4D: Top-up Programs

To examine the effectiveness of “top-up”, foundation, preparatory, preliminary, etc. programs, studies were coded as to whether the students had to take one of those programs before being granted access to formal post-secondary courses. Many studies used mixed samples.
or did not make mention of any such necessary courses, so only a few studies could be examined for moderator analysis, results of which can be seen in Table 10.

Table 10

Moderator Analysis – Top-Up

<table>
<thead>
<tr>
<th>Entry</th>
<th>k (subgroups)</th>
<th>r</th>
<th>95% CL</th>
<th>Z</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Direct</td>
<td>3</td>
<td>.252**</td>
<td>[.182, .320]</td>
<td>6.858</td>
<td>.000</td>
</tr>
<tr>
<td>EAP Top-up</td>
<td>2</td>
<td>.135**</td>
<td>[.057, .211]</td>
<td>3.393</td>
<td>.001</td>
</tr>
</tbody>
</table>

Note. $Q_{between} = 4.90^*$, $df = 4$, $p = .027$

This result is the sole moderator analysis to reach statistical significance, with the caveat that it really only includes 5 total effect sizes. While these results initially seem promising, the small sample makes it unable to be reliably confirmed. This is examined more in depth in the discussion section.

3.4.9 Research Question 4E: Timepoint

The final moderator analysis looked at different time points and the relevant predictive validities. Two of the studies, (Dooey & Oliver, 2002; Yen & Kuzma, 2009) were longitudinal in nature and had results for multiple time points. For all other analyses, these results were pooled so as to not be treated as independent samples. However, for this specific analysis those time points were treated as independent, which can create issues of validity as the groups are comprised of the same students and initial IELTS scores, and not truly independent. This was the most efficient way include the longitudinal studies for comparative analysis however.

Table 11

Moderator Analysis – Timepoint
While there appears to be a slight overall downwards trend with correlational power over time, the results do not reach necessary significance to confirm these findings. Given the extraneous variables introduced over time, it is logical to assume that the correlational power would be lower as time goes on, and this is examined further in the discussion section.

3.5 Discussion

3.5.1 Research Question 1: Overall Effect

While an approaching-small effect size might seem surprisingly low, the overall context is important to consider. English skills are only one small part of the necessary skills needed for academic success; many native speaking students fail to achieve academic success. The IELTS test may test English abilities, but it is unable to truly test other variables such as perseverance, focus, desire to study, etc. Even the English skills that it tests may not be enough, Sedgwick and Garner (2017) found that many necessary skills for successful communication are not tested by IELTS, such as the use of socio-pragmatic competence, which could be necessary for any direct communication the students must have with instructors or other students.

When compared to the prior meta-analyses conducted on TOEFL’s predictive validity, the overall results are quite similar. Wongtrirat (2010) had found an effect size of .181, and

<table>
<thead>
<tr>
<th>Timepoint</th>
<th>k (subgroups)</th>
<th>$r$</th>
<th>95% CL</th>
<th>$Z$</th>
<th>$p$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Semester 1</td>
<td>9</td>
<td>.245**</td>
<td>[.085, .392]</td>
<td>2.972</td>
<td>.003</td>
</tr>
<tr>
<td>Semester 2</td>
<td>4</td>
<td>.149</td>
<td>[.034, .322]</td>
<td>1.594</td>
<td>.111</td>
</tr>
<tr>
<td>First Year</td>
<td>10</td>
<td>.232**</td>
<td>[.142, .319]</td>
<td>4.974</td>
<td>.000</td>
</tr>
<tr>
<td>Overall GPA</td>
<td>5</td>
<td>.193*</td>
<td>[.024, .352]</td>
<td>2.234</td>
<td>.026</td>
</tr>
</tbody>
</table>

Note. $Q_{between} = 0.880, df = 27, p = .830$
Abunawas (2014) found an effect size of .21. Both authors refer to their effect sizes as small, using Cohen’s original effect size suggestions (Cohen, 1988). While the present study makes use of more updated and context relevant effect size cut-offs, it would also fit within the original definition for a small effect size.

These results are interesting when compared to the original studies contained within the meta-analysis. Four of the analysed studies compared IELTS and TOEFL’s predictive validity. Hill et al. (1999) found that both scores predicted GPA, with IELTS being a “moderately strong” predictor and TOEFL “relatively weak”. Woodrow (2006) found IELTS scores were a significant predictor of academic success, but did not find significance for TOEFL scores. Arcuino (2013) found that both scores were significant predictors, but that there was not a significant difference between their predictive validity. And most recently, Johnson & Tweedie (2017) found that both were valid predictors and roughly equitable in their strengths. Given these results, the overall effect size of the present study being similar or slightly higher than the previous TOEFL meta-analyses fits well with the past research.

The large difference of the removed Spearman’s studies must also be acknowledged. The results from those four studies provided an effect size of $r=.586$, which is vastly higher than the overall findings, and approaching a large effect size (Plonsky & Oswald, 2014). Given the small size of studies, cautious interpretation is necessary. For example, Erfani & Mardan (2017) reported a rho of $r=.794$, which is an abnormally high correlation to see. If this study had been included in the overall meta-analysis, it certainly would be an outlier, which was quite clear on early versions of the funnel plot. Further research into the relationships found using Spearman’s rho is a necessity.
3.5.2 Research Question 2: Funding Bias

There seems to be very little evidence of any bias from IELTS publishing themselves, in either direction. The Q between statistic was not significant, and the reported effect size of the IELTS publications is actually lower than the 3rd party studies, but those results are not significant due to low size and high p values.

One possible explanation may be related to publication year. Of the four IELTS publications, only three were included in the overall analysis as Humphreys et al. (2012) did not report composite scores. The three studies that were included were the first three studies in the meta-analysis chronologically, being published in 1998, 1999, and 2000. In the 21 years since 1998, there has been continual minor revisions and validations done to IELTS, perhaps making the scores more accurate. It is quite possible that scores reported now are more accurate than scores from 20 years ago; some of the strongest results were published in the last few years (e.g., Erfani & Mardan, 2017; Johnson & Tweedie, 2017; Müller & Daller, 2019).

3.5.3 Research Question 3: Subscores

No individual skill seems to be as strong at predicting academic success as the overall composite score. Reading comes the closest, with its 95% confidence interval crossing the threshold into a small effect size, with its upper bound of \( r = .300 \) crossing the recommended small threshold or \( r = .25 \) (Plonsky & Oswald, 2014). Given the nature of first year courses being heavily exam-based (Cuseo, 2007), and the prevalence of using first semester or year grades, reading being the strongest correlate is a logical conclusion. The skills necessary to do well on the IELTS reading test would be similar to the multiple-choice format exams that large introductory courses rely upon.
Across individual studies there is an extremely wide range of variance regarding individual skills. For example Neumann et al. (2018) found reading to have the strongest correlation, even stronger than overall scores, while Woodrow (2006) found reading to be the lowest of all the sub-scores. Some studies (e.g., Avdi, 2011; Cotton & Conrow, 1998) reported negative correlations for speaking sub-scores even, but it must be noted that they did not include p values, so it is unknown if these correlations reached statistical significance.

Lack of significance was a common problem in reporting correlations. A few of the studies simply did not include it at all (e.g., Avdi, 2011; Cotton & Conrow, 1998; Hill et al., 1999), which is a problem for quality of reporting (Plonsky, 2014). Other studies simply had trouble finding significant results. Dooey and Oliver (2002) reported 40 total correlations: all five scores from four different groups measured at two different time points, and only found four total significant values in all 40 of those scores, all of which happened to be reading correlations. A lack of real significant evidence was a common problem among many of the studies (e.g., Arcuino, 2013; Garinger & Schoepp, 2013; Humphreys et al., 2012; Kerstjens & Nery, 2000; Oliver et al., 2012).

**3.5.4 Research Question 4A: Field of Study**

Given the overall small samples for comparison, the lack of significant difference is not surprising. Many of the samples were simply too heterogenous for true comparison, which is unfortunate for this specific analysis.

Only one of the final studies examined and compared correlations by field of study. Dooey and Oliver (2002) compared effect sizes across three different majors: business, science, and engineering. The vast majority of effects they examined did not reach a significant p value,
except for the business group’s reading score correlation, which had a value of $r = .396$ at the first semester, $p < .05$. However, this is only a single study with a sample size of 30 students in the business cohort, so no definitive conclusions can be drawn from it unfortunately.

This could be an interesting avenue to explore for future research, as the information should be as readily available as student GPAs from university databases. Different programs require different cut-off scores, but there seems to be little evidence of validation for that distinction as of now.

3.5.5 Research Question 4B: Level of Study

Similar to the previous research question, there was a lack of proper studies to find definitive results when examining level of study. The graduate level results do cross into the small effect size while undergraduate studies do not, but the $Q_{between}$ is not significant enough to be certain they are significantly different.

Three different studies did examine the levels of study as separate groups with mixed results. Kerstjens and Nery (2000) found no significant correlations for a mixed group, but an undergraduate only group had a correlation of $r = .285, p < .05$ for their reading scores. Oliver et al. (2012) found an overall correlation of $r = .275, p=.000$ for a graduate group and no significant results for their undergraduate group. Additionally, their graduate group had significant correlations for listening, speaking, and reading sub-scores, while the undergraduate group only reached significance in reading. Finally, Johnson and Tweedie (2017) found an overall correlation of $r = .288, p=.039$ for their undergraduate group and no significant correlation for their graduate group. Clearly there are quite a few mixed results within these comparisons.

3.5.6 Research Question 4C: Study Location
Due to this analysis classifying countries as simply English-speaking or not, much subtly and nuance between any countries was lost. This was a necessity of the coding unfortunately, due to the amount of studies included. The vast majority of studies were conducted in Australia, and only four were conducted outside of the Anglosphere. It should be noted that that all the non-native studies were in the Middle East, results could differ in Asia, eastern Europe, sub-Saharan Africa, or South America.

Abunawas (2014) found that studies conducted outside of the USA had significantly stronger correlations between TOEFL scores and GPA than those conducted in the USA. This is slightly at odds with the results in the present study, but not a straightforward comparison to make. Conflating the USA with the entire Anglosphere would not be prudent, nor would simply focusing on the USA alone for the present study as there was only a single study published within the USA.

Further examination of the differences in English needs at universities outside of the Anglosphere compared to those within it are warranted.

**3.5.7 Research Question 4D: Top-up Programs**

While the overall sample pool was lower for this analysis, the $Q_{between}$ value was significant, showing that students who take “top-up” programs have less predictive IELTS scores than those who are granted direct entry. This result should not be surprising, as the inclusion of a “top-up” program significantly changes the student’s English abilities and introduces a strong variable into the correlation. Given that these students are most likely arriving to the school earlier for the program, studying more English, and spending more time in the culture, they are being exposed to many more sources of English than those who do not take part in the programs.
These additional factors would alter a student’s English abilities and make their IELTS scores less reliable for representing their true English skills after completing a top-up course.

Schoepp (2018) specifically looked at this comparison and found similar results, $r = .256$ for students with direct entry and $r = .159$ for students who completed top-up programs. He additionally compared their raw GPAs and found that students who completed top-up programs had significantly lower GPAs than direct entry students. Thorpe et al. (2017) found similar results regarding both lower GPAs and weaker relationships between IELTS scores and GPAs in their regression analysis. This does raise some concerns for the effectiveness of such programs. They do benefit both the student and institution, but are they truly preparing the students and getting them to an adequate English skill level for their studies?

**3.5.8 Research Question 4E: Timepoint**

This moderator was one of the most difficult to analyze due to differing standards across institutions. Some studies used “one year” to mean two semesters, some meant three semesters, and for some it was the entire length of the program. These inconsistencies made the coding difficult and many conservative decisions were made that might have hindered the overall analysis.

However, as previously mentioned, there were two studies that were more longitudinal in nature and directly looked at performance in multiple semesters. Dooey and Oliver (2002) examined first and second semester scores among international students. Of the 40 different correlations they examined, only reading correlations were significant. For their overall group the student’s IELTS score and GPA correlations went from $r = .273, p < .05$ for semester one to $r = .340, p < .01$ for semester two. This is a fair size increase, though the only one of significance
from their study. On the other hand, Yen and Kuzma (2009) found higher and more significant correlations for overall scores, listening, reading, and writing at semester one than semester two, some to a very substantial degree such as semester one writing sub-score of $r = .41$, $p < .01$, and semester $r = .02$ with no significance. There are clearly mixed results regarding predictive ability at different time points, and that warrants further examination.

3.5.9 Other Discussion

Overall there was a significant moderate amount of heterogeneity, $Q = 45.353$, $p = .004$, $I^2 = 49.286$. This is not surprising given how different the samples and studies were. While IELTS is highly standardized and should have homogenous scores across samples, GPA is extremely heterogenous. Each institution has their own standards for how to set it, and even individual instructors have their own criteria and limits. A high level of variance between courses, schools, and overall countries is to be expected. The language requirements necessary also differ to a large degree, which was unfortunately not fully captured by the moderator analysis in the present study. More large-scale exploratory studies are needed to help discern what moderates the relationship between IELTS scores and GPA.

One area of concern is the reporting standards of data and methodology in studies (Plonsky, 2014). As was previously mentioned, there are two distinct version of the IELTS test, general and academic. Of the original 28 coded studies, only four made explicit mention of using the academic version. Given that the academic test is designed for university entrance, and four times as common as the general test, it can be assumed that most studies should have been using academic scores. However, this important detail was still lacking from the vast majority of studies examined.
Unfortunately, there was a lack of available data for more moderator analyses as well. It would be particularly interesting to compare the predictive power of IELTS across first languages or regions of the world to see if it was more predictive for certain student populations. Perhaps some future research will examine that in a cross-cultural study. Similarly, age and gender data were rarely available to examine any effects they might have.

3.5.10 Publication Bias

A constant risk in scientific literature is the absence of results that do not achieve significance. This phenomenon has been termed “publication bias” and can severely impact a meta-analysis (Oswald & Plonsky, 2010). One of the most common ways to detect publication bias is through a funnel plot, as can be seen in Figure 3. A funnel plot compares the standard error to the Z scores based on sample and effect size. An asymmetrical funnel plot is generally considered to signal publication bias, but it can also reflect selective reporting, poor design, heterogeneity, fraud, or even just pure chance (Sterne et al., 2011). Fortunately, that does not seem to be the case for the present study, as Figure 3 appears to be quite symmetrical.
Figure 3. Funnel plot

3.5.11 Limitations

By having such broad inclusion criteria, there is a very real possibility that poor quality studies and results were included and influenced the overall results. The term “garbage in, garbage out” has been coined for this potential issue (Eysenck, 1984). However, determining the rigour of individual studies can be near impossible given the limited access to data overall. As Schmidt and Hunter (2015) state “there are no perfect studies” (p. 19). Moreover, excluding certain studies due to perceived weaknesses could be seen as a sign of bias and taint the results. Oswald and Plonsky (2010) recommend erring on the side of comprehensiveness and including as much as possible. By having a large amount of studies, any outliers or effects of low-quality studies were hopefully minimized. To account for any effects of any individual study, the “One study removed” analysis in CMA was used, which re-analyses the dataset excluding each study individually to examine any outlying effects. Overall results show that there was not a significant effect from any one study; the most extreme cases were $r = .215$ when removing Müller and Daller
(2019) and \( r = .236 \) when removing Kerstjens and Nery (2000). This shows that even removing the most extreme variance of studies does not shift the overall results to a significant degree, and therefore any one poor quality study would have had minimal impact on the overall findings.

Interrater reliability is a core part of meta-analysis (Schmidt & Hunter, 2015). Ideally the entire codebook would have been completed by multiple raters, but this was not an option due to constraints of the project. By having half of the studies dual-coded there is a level of interrater reliability, but it could certainly be better.

Another limitation that must be noted is the inclusion of only English studies. While this meta-analysis is focused on a test of English ability, there is the potential for studies to have been published in other languages and thus have been excluded from the current study. This can have significant effects: Grégoire, Derderian, and Le Lorier (1995) analyzed 36 different meta-analyses that had used exclusively English sources and found that one of those meta-analyses would have had significantly different conclusions if it had included a non-English source. They term this reliance on solely English literature the Tower of Babel bias, and it must certainly be acknowledged as a limitation. While the modern academic climate does push for publication in English, and there is a perceived higher quality to English publications (Curry & Lillis, 2015), it cannot be assumed that all quality research will be published in English, and quite frankly it would be strong case of linguistic imperialism to imply as much. Thus, this is another form of bias that must be considered, as conducting a meta-analysis across all the literature available in every language is simply not feasible.

3.6 Conclusion
An overall approaching-small effect size between IELTS scores and post-secondary GPA was found. There does not appear to be any evidence of bias from the IELTS organization in publishing stronger results, if anything their results were weaker than third party findings. Overall composite scores seem to be a stronger predictor than any one individual skill. Of the individual macro skills, reading has the highest predictive power and speaking the lowest. Most moderator results were inconclusive or lacked sample sizes to offer definitive results.

This study is not without its shortcomings. Conservative choices were made whenever possible and this resulted in data being removed that is arguably valid. This leaves many opportunities for future research, especially in regard to much of the moderator analysis.

The coding sheet and raw data collected for this meta-analysis will be made available publicly online after publication of this study. Given both the notion of transparency and the rise of open-access academic information, sharing the data upon which this thesis is based is the logical choice to make. Finally, this will also allow future researchers to examine the data to build upon or perhaps spot any errors or inconsistencies.

3.6.1 Disclosure Statement

The author has experience teaching IELTS preparatory materials at unaffiliated private language institutions but has never been directly employed or funded by the IELTS organization or any of its partners. There is no declared conflict of interest.
Chapter 4 – Conclusion

4.1 Review

This meta-analysis examined the predictive validity of IELTS scores on academic success, measured as GPA. An approaching-small effect size of $r = .227$ was found for the overall relationship between the two variables. Additional analyses show that there is unlikely any bias from IELTS funding of results, that no single subscore is as strong a predictor of success as the overall score, and that few conclusions can be drawn about differences among levels, field, country of study, top-up courses, or timepoint.

4.2 Implications

Given the large amount of heterogeneity and inconclusive moderator analyses, widespread interpretations must be made cautiously. At the very least, it can be stated that IELTS scores do have an approaching-small relationship to post-secondary GPA, with overall scores being the strongest predictor.

4.2.1 For the IELTS organization

There seems to be no clear evidence of publication bias coming from the organization itself, which helps with the credibility of their IELTS Research Reports series. While not being highly predictive, the test itself does have some predictive power for post-secondary study, which helps to validate its use in admissions. Given that the results are similar to prior studies of TOEFL, it is possible to believe we are approaching overall predictive validity of standardized English tests as a whole, however it is also possible that IELTS and TOEFL capture different
aspects of English and/or academic skills, and they are actually predicting separate portions of student GPA. Further study and comparisons are needed.

4.2.2 For Students/Test-Takers

Previous research has shown that IELTS scores are considered a valid representation of English skills (Hyatt, 2013). While they may reflect English abilities, they only reflect a small portion of academic potential. Receiving a low score on the IELTS test does not mean a student will do poorly in their post-secondary classes, nor does getting a high IELTS score guarantee a student will do well in those classes. Students and test-takers should be aware of the many other factors necessary to do well in academics, and not get too caught up on their IELTS score alone.

4.2.3 For Admissions Offices and Institutions

The results of this study show that IELTS scores are at the minimum just as valid as TOEFL scores for predicting academic success, which may help for decisions being made about which test scores to accept. However, it must be noted that it is still a weak correlation. English skills alone are only part of the big picture however, and a small one at that it seems. Admissions requirements are still encouraged to focus on previous GPA, as they are generally a stronger predictor than standardized tests (Zahner et al., 2012).

4.3 Future Research

While this meta-analysis sought to be inclusive, a large number of valid studies were removed to keep the coding process consistent and conservative. A separate analysis of just the studies making use of Spearman’s rho or regression analysis could be valuable, as those studies had some of the strongest results.
Despite the strengths of a meta-analyses, they are not equivalent to a large scale study (LeLorier, Grégoire, Benhaddad, Lapierre, & Derderian, 1997). For best practices, a true large-scale study examining the relationship between IELTS scores and GPA should be conducted. Many of the included studies had woefully small sample sizes, which could be improved upon. Given the increasing amounts of international students, as well as digitized admissions files and GPA, the data should be plentifully available for researchers, assuming they receive ethics approval. Some of the moderator analyses could be examined in more detail as well, as program and level of study would be just as easily available.
References


62

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Appendix A: Email Template

Subject: Clarification on Article XXX

Hello XXXX

My name is Tomlin Gagen and I am a graduate student at Western University currently working on a meta-analysis of IELTS predictive validity for my thesis. I have identified your study (XXXX) as a possible inclusion, but there is a bit more information I need to include it. If it’s not too much trouble, would you happen to have (XXXXX) (ex: The SD for your mean result, etc.). That information would be very helpful in improving the power and comprehensiveness of my results.

Thank you very much for your time, I look forward to your response. If you have any questions at all, I’d be happy to answer them

-Tomlin Gagen
Appendix B: Included Studies

Arcuino, C. L. T. (2013). The relationship between the test of English as a foreign language (TOEFL), the international English language testing system (IELTS) scores and academic success of international master’s students (Ph.D., Colorado State University). Retrieved from http://search.proquest.com/docview/1413309058/abstract/A4CE9F4F300140E8PQ/1


https://doi.org/10.1111/hequ.12163


Curriculum Vitae

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