

PASS Theory of Intelligence and Academic Achievement: A Meta-Analytic Review

by

Nithya Naveenkumar

A thesis submitted in partial fulfillment of the requirements for the degree of

Master of Education

in

Special Education

Department of Educational Psychology

University of Alberta

© Nithya Naveenkumar, 2020

## Abstract

Although Planning, Attention, Simultaneous and Successive (PASS) processing theory of intelligence has been argued to offer an alternative look at intelligence and PASS processes – operationalized with the Cognitive Assessment System – have been used in several studies, it remains unclear how well the PASS processes relate to academic achievement. Thus, this study aimed to estimate the size of their relation by conducting a meta-analysis. To select the studies for the meta-analysis, I conducted an electronic database search (e.g., ERIC, PubMed, PsycINFO), an ancestral search, and also reviewed book chapters, dissertations, and interpretive and technical manuals. A total of 62 studies, involving 13,356 participants, met the inclusionary criteria. A random-effects model analysis of data from 62 studies with 93 independent samples revealed a moderate-to-strong relation between PASS processes and reading,  $r = .409$ , 95% CI = [.363, .454]), and mathematics,  $r = .461$ , CI = [.405, .517]. Moderator analyses further showed that (1) PASS processes were more strongly related with reading and mathematics in English than in other languages, (2) Simultaneous processing was more strongly related to mathematics accuracy and problem solving than mathematics fluency, (3) Simultaneous processing was more strongly related to problem solving than Attention, and (4) Planning was more strongly related to mathematics fluency than Simultaneous processing. Age, grade level, and sample characteristics did not moderate the PASS-reading/mathematics relation. Taken together, these findings suggest

that PASS cognitive processes are significant correlates of academic achievement, but their relation may be affected by the language in which the study is conducted and the type of mathematics outcome.

## **Preface**

This dissertation is submitted for the degree of Master of Education in Special Education at the University of Alberta. The research reported herein was conducted under the supervision of Dr. George K. Georgiou, Department of Educational Psychology, University of Alberta. The literature review, data analysis, and discussion are the original work of Nithya Naveenkumar.

## Acknowledgements

I gratefully acknowledge the guidance and immense support I received from my supervisor Dr. George K. Georgiou. I am forever indebted to him for seeing my ability and believing that I could certainly obtain knowledge and skills required to conduct this study. I feel proud of the core competencies and skills I was able to develop under his supervision. I also want to express my profound gratitude to Dr. Guo Kan (Beijing Normal University), whose statistical expertise was invaluable during the analysis and interpretation of the data.

I would also like to extend my profound appreciation to my committee members, Dr. Alison McInnes and Dr. Lynn McQuarrie for their academic and professional perspectives as well as feedback on this thesis. Especially, I thank my reading research lab mates Ana Paula Alves Vieira for her active participation as a second coder in my study, Kristy Dunn, Dalia Martinez, and Sandra Romero Carreón for the voice of support and friendship.

Finally, and most importantly, I feel extremely grateful to my husband Naveenkumar, father Ganesan, and my children Moabin and Rudhvin. Your continuous encouragement and unfailing support during my years of study and throughout the process of research made me to accomplish this thesis.

## Table of Contents

	Page
<b>Chapter 1: Introduction .....</b>	<b>1</b>
<b>Chapter 2: Literature Review .....</b>	<b>4</b>
PASS Theory of Intelligence .....	4
CAS: Operationalization of PASS Theory.....	5
PASS Theory of Intelligence and Reading .....	7
PASS Theory of Intelligence and Mathematics.....	9
The Present Study .....	10
<b>Chapter 3: Method .....</b>	<b>14</b>
Data Collection.....	14
Operational Criteria for Inclusion and Elimination of Studies.....	16
Coding Procedures.....	18
Moderator Variables.....	19
Statistical Analysis.....	21
Publication Bias.....	23
<b>Chapter 4: Results .....</b>	<b>24</b>
Study Features.....	24
Meta-Analytic Results.....	25
Results of the Moderator Analyses.....	29
Publication Bias.....	35
<b>Chapter 5: Discussion .....</b>	<b>39</b>
<b>References .....</b>	<b>45</b>

### List of Tables

		Page
Table 1:	Overall Meta-Analytic Results.....	25
Table 2:	Moderator Analyses for Reading: Categorical Moderator Variables (PASS and Outcome Subtype) .....	30
Table 3:	Moderator Analyses for Math: Categorical Moderator Variables (PASS and Outcome Subtype) .....	31
Table 4:	Moderator Analyses for Reading: Categorical Moderator Variables.....	33
Table 5:	Moderator Analyses for Math: Categorical Moderator Variables.....	34
Table 6:	Publication Bias Analyses.....	36

### List of Figures

		Page
Figure 1:	Flow Diagram for the Search and Inclusion on Studies.....	15
Figure 2:	Forest Plot: Strength of the Correlations between PASS and Reading.....	27
Figure 3:	Forest Plot: Strength of the Correlations between PASS and Math.....	28
Figure 4:	Funnel Plots for Reading and Funnel Plots with Imputed Samples for Reading.....	37
Figure 5:	Funnel Plots for Math and Funnel Plots with Imputed Samples for Math.....	38



## Appendices

	Page
Appendix A: Studies on PASS and Reading Outcomes.....	63
Appendix B: Studies on PASS and Math Outcomes.....	71

## Chapter 1: Introduction

Reading and mathematics are certainly some of the most important skills children are asked to master in their early school life. Performance in these academic skills serves as a vital indicator of an individual's future education and occupational choices (e.g., Ek, Sovio, Remes, & Järvelin, 2005; Flashman, 2012; Lubinski, Benbow, & Kell, 2014; Rana & Mahmood, 2010). Consequently, knowing what influences performance in these two academic skills is of utmost importance not only because they determine an individual's future, but also because they can adversely affect society's future as well as a country's economic growth (e.g., Coulombe, Tremblay, & Marchant, 2004).

Intelligence has been found to be one of the factors affecting reading and mathematics performance (e.g., Barton, Dielman, & Cattell, 1972; Deary, Strand, Smith, & Fernandes, 2007; Mayes, Calhoun, Bixler, & Zimmerman, 2009; Soares, Lemos, Primi, & Almeida, 2015). In general, individuals with higher IQ outperform others with lower IQ in reading and mathematics. Meta-analyses have also been conducted to determine the size of the relation between intelligence and academic achievement (e.g., Peng, Wang, Wang, & Lin, 2019; Roth et al., 2015). Roth et al. (2015), for example, estimated the average correlation between IQ (operationalized with different IQ measures) and school grades to be .44.

Although several studies have established a close connection between IQ and academic achievement, some researchers have argued that traditional IQ tests (e.g., Wechsler Intelligence

Scale for Children) and academic achievement tests (measures include vocabulary, general information, arithmetic) often have very similar content (Naglieri, 2008; Naglieri & Borenstein, 2003). This leaves open the possibility that performance in these IQ measures is influenced by what a child already knows (or what s/he has learned at school) and not by how s/he thinks (which should be the target of intelligence testing) (e.g., Das, 2002; Gardner, 1993; Naglieri & Otero, 2018). Researchers have also claimed that the most popular psychometric batteries of IQ lack a theory base that relates intelligence measures to achievement (Das, Naglieri, & Kirby, 1994; Fagan, 2000; Sternberg, 1988).

To address these issues and to expand the scope of abilities measured, Das et al. (1994) proposed a neurocognitive theory of intelligence called PASS (for Planning, Attention, Simultaneous, and Successive processing). The PASS theory of intelligence provides a comprehensive way of conceptualizing and measuring human cognitive competencies through the Cognitive Assessment System (CAS; Naglieri & Das, 1997; CAS2; Naglieri, Das, & Goldstein, 2014). Researchers have found that all four PASS processes - measured by CAS - are correlated with academic performance (see Kroesbergen, Van Luit, Naglieri, Taddei, & Franchi, 2010; Naglieri, Rojahn, Matto, & Aquilino, 2005; Naglieri, Delauder, Goldstein, & Schwebech, 2006; Naglieri & Rojahn, 2004; Rosário, 2007). Although several studies have examined the relation of CAS measures with academic achievement, we are still lacking a quantitative synthesis of this line of research. Thus, the purpose of this meta-analysis was to estimate the size

of the relation between PASS processes and academic achievement as well as the role of different moderators in their relation (e.g., the type of reading and mathematics outcome, the age of participants, the sample characteristics, and the language in which the study was conducted).

This thesis is expected to make two important contributions to the literature: First, it provides an integrative review of the existing research on the relation between PASS processes and academic achievement that cannot be provided by a single study. Second, given that PASS theory has been studied for more than 20 years, this is the first meta-analysis of correlational studies examining the association of PASS cognitive processes with reading/mathematics performance.

## Chapter 2: Literature review

### PASS Theory of Intelligence

Based on the neuropsychological work of Luria (1966, 1973), PASS theory of intelligence asserts that human cognition is organized into three functional units that support four cognitive processes (Planning, Attention, Simultaneous, and Successive) (e.g., Das et al., 1994; Naglieri & Das, 1990). According to Das et al. (1994), the first functional unit, Attention-Arousal is responsible for maintaining adequate level of arousal that facilitates focus of attention on a specific direction and on a particular stimulus. Difficulties in this first functional unit may result in difficulties with information coding (Simultaneous and Successive processing), as well as difficulties in controlling, organizing, and monitoring behavior (Planning).

The second functional unit is involved in the acquisition, storage and retrieval of information from the external environment through Simultaneous and Successive processes. Simultaneous processing involves integrating stimuli into groups or the recognition that a number of items share a common characteristic, for example, comprehending a sentence/paragraph. In turn, Successive processing is required for organizing separate items in a sequence, for example, remembering a sequence of words.

The third functional unit is responsible for selection and application of strategies, self-monitoring performance, evaluation and modification of approach if necessary. The abilities and behaviors involved in the Planning construct closely resembles executive functioning (EF) skills

(Das & Misra, 2005). All the three functional units are interrelated but have specific functions that are influenced by an individual's knowledge base.

### **Operationalization of PASS Theory**

The PASS theory has been operationalized by Cognitive Assessment System (CAS; Naglieri & Das, 1997; Naglieri et al., 2014) that measures the PASS constructs in children from ages 5 through 17. The PASS constructs are comprised of four scales that are defined by three subtests in each scale. There are two forms of CAS battery – standard and basic. In the standard battery, each PASS scale consists of three subtests whereas the basic battery is composed of two subtests in each scale. All four PASS scales and the respective subtests are described below.

***Planning.*** Planning is assessed with three subtests: Matching Numbers, Planned Codes, and Planned Connections. Matching Numbers requires the examinee to rapidly underline two identical numbers in each row. Planned Codes requires an individual to fill out empty boxes, as quickly and accurately as possible based on the given key on the top of each page (i.e., A=XO, B=XX, C=OX, D=OO). In Planned Connections, the examinee is asked to connect a series of boxes with numbers or letters in a sequential order. Some of the items in this subtest include a sequence of numbers only, though others include both numbers and letters sequence to connect in an alternating manner (e.g., 1-A-2-B-3-C).

***Attention.*** Attention is assessed with three measures: Expressive Attention, Number Detection, and Receptive Attention. In Expressive Attention, the examinee is asked to name

stimuli on three pages within a specific time limit – i) to read colour names such as “Blue”, “Yellow”, “Green” and “Red” that are printed in black ink, ii) to name the colors of a series of rectangles, and iii) to name the colour of the ink in which the colour names are printed (e.g., Blue printed in red ink to be read as “Red”). Number Detection requires an individual to locate the target number as quickly as possible in the same font (i.e., the numbers 1, 2, and 3 printed in an open font) among the distractors in a varied font type. In Receptive Attention, an individual is asked to rapidly underline pairs of letters that are similar in appearance (e.g., TT or t t but not N t) and then identify pairs of letters that share a similar lexical perspective (e.g., t and T but not t and N).

***Simultaneous processing.*** Simultaneous processing is measured with three tasks:

Nonverbal Matrices, Verbal-Spatial Relations, and Figure Memory. Nonverbal Matrices requires an individual to choose a missing piece from six possible answers that completes correctly a matrix. Verbal-Spatial Relations consists of six drawings (organized in a certain spatial manner) and a printed question (e.g., *Which picture shows a circle to the left of a cross under a triangle above a square?*). The examiner reads out the question and then asks the participant to select the correct response from six options. In Figure Memory, the examiner shows two- or three-dimensional geometric figures to the subject for 5 secs and then takes away the picture. Next, the examiner provides a response page and asks the subject to identify the original design that is enclosed in a larger complex geometric figure.

***Successive processing.*** Successive processing is assessed with three measures: Word Series, Sentence Questions, and Visual Digit Span. Word Series requires the subject to repeat a series of single-syllable, high frequency words (e.g., book, car, cow, dog) in the same order s/he had heard them by the examiner. In Sentence Questions, the examiner read out loud a series of sentences (i.e., “The blue is yellowing”) and then asks questions related to the sentences read (e.g., “Who is yellowing?”). Finally, in Visual Digit Span, the examiner shows a series of numbers from a stimulus book and then asks the subject to recall the numbers in the same order.

### **PASS Processes and Reading**

To date several studies have shown that PASS processes are related to reading performance (see e.g., Das, Snart, & Mulcahy, 1982; Joseph, McCachran, & Naglieri, 2003; Kirby & Robinson, 1987; Naglieri, & Rojahn, 2004). For example, Naglieri and Rojahn (2004) studied the association between PASS processes and Broad Reading (Letter-Word Identification and Passage Comprehension) with the CAS standardization sample. The results indicated that the correlations between the four cognitive processes and Broad Reading ranged between .43 and .55 (the highest being between Simultaneous processing and Broad Reading).

Researchers have also proposed theoretical links between PASS processes and reading performance (see Das et al., 1994; Das & Misra, 2015; Naglieri & Otero, 2018; Papadopoulos, Parrila, & Kirby, 2015). Das et al. (1994), for example, proposed that Successive processing contributes to word reading through the effects of phonological recoding (sounding out) and



Simultaneous processing contributes to word reading through the effects of orthographic knowledge (the ability to form, store, and access orthographic representations). Planning and Attention have also been viewed as critical for reading comprehension. To succeed in reading comprehension, individuals need to develop a plan on how to approach a passage, actively revise their plan as they read a passage, and inhibit irrelevant information in order to develop a coherent text representation.

Findings of previous studies with typically-developing children (e.g., Das, Georgiou, & Janzen, 2008; Kendeou, Papadopoulos, & Spanoudis, 2015; Naglieri & Rojahn, 2004; Papadopoulos, 2001; Wang, Georgiou, & Das, 2012) have confirmed these predictions. For example, Papadopoulos (2001) found that the effect of Successive processing on reading accuracy (Word Identification and Word attack) in Greek-speaking Grade 1 children was mediated by phonological awareness. Likewise, these predictions were confirmed in the studies that involved children with reading difficulties (e.g., Das, Janzen, & Georgiou, 2007; Joseph et al., 2003; Wang, Georgiou, Das, & Li, 2012).

On the basis of these relations, Das (1999) also developed the PASS Reading Enhancement Program (PREP) that trains children on Simultaneous and Successive processing. Intervention studies showed that PREP training in children with reading difficulties led to a significant improvement in reading (Mahapatra, Das, Stack-Cutler, & Parrila, 2010; Papadopoulos, Charalambous, Kanari, & Loizou, 2004). Mahapatra et al. (2010), for example,

examined the effects of PREP on word reading and comprehension in two groups of 14 fourth grade poor readers. The intervention group received PREP, while the control group received regular classroom instruction. Grade level comparisons showed that the intervention group showed a significant increase in word reading equal to 3.5 grades and in reading comprehension equal to 3.2 grades.

### **PASS Processes and Mathematics**

Researchers have shown that PASS processes are also related to mathematics performance (e.g., Iseman & Naglieri, 2011; Kirby & Ashman, 1984; Kroesbergen et al., 2010; Naglieri & Rojahn, 2004). For example, Naglieri and Rojahn (2004) found that the correlations between the four cognitive processes and Broad Math (Calculation and Applied Problems) ranged between .45 and .58 (the highest being between Simultaneous processing and Broad Math).

Planning is important for mathematics because individuals must make decisions on how to solve a math problem and monitor their own performance. Attention is involved in selectively attending to the components of a problem and for suppressing irrelevant information.

Simultaneous processing is relevant for tasks that consist of different interrelated elements that must be integrated into a whole, as in solving an equation with multiple operations (e.g.,  $(3 + 5) \times (4 + 4) / 2 = ?$ ) or in areas of mathematics that involve interpretation of spatial information (e.g.,

geometry). Finally, Successive processing is relevant when information has to be processed in a certain order, as in counting.

Findings of previous studies with typically-developing children (e.g., Georgiou, Manolitsis, & Tziraki, 2015; Naglieri & Das, 1987; Naglieri & Rojahn, 2004; Kroesbergen et al., 2010) as well as children with mathematics difficulties (e.g., Cai, Li, & Deng, 2013; Iglesias-Sarmiento & Deaño, 2011; Kroesbergen, Van Luit, & Naglieri, 2003) have confirmed these predictions. For example, Cai et al. (2013) found that children with Math Learning Difficulties (MLD; those who scored at the bottom 20% in a standardized math test and in three math class tests) demonstrated fairly lower scores on all four PASS processes than good math performers (those who scored at the top 20% in a standardized math test and in three math class tests).

Intervention studies have also shown that cognitive strategy instruction (especially Planning) significantly improved children's math performance (Hald, 1999; Iseman & Naglieri, 2011; Naglieri & Gottling, 1997; Naglieri & Johnson, 2000). For example, Naglieri and Johnson (2000) found positive effects of Planning intervention in 19 children with learning disabilities and mild developmental disabilities (ages 12 and 14). The children with cognitive weaknesses in Planning showed substantial improvement in math accuracy (with a large effect size of 1.4), whereas the children with other cognitive weaknesses (Attention, Simultaneous, and Successive) and no cognitive weaknesses displayed comparably smaller effect sizes that ranged from -.2 to .5.

## The Present Study

The purpose of this meta-analysis was to examine the strength of the relation between PASS processes and reading/mathematics performance. We aimed to answer the following five research questions:

- (1) What is the size of the relation between PASS processes and reading/mathematics performance? Based on the findings of previous studies (e.g., Naglieri & Rojahn, 2004), we hypothesized that PASS processes would be strongly related to both reading and mathematics performance.
- (2) Does the PASS–reading/mathematics relation vary as a function of the type of reading/mathematics outcome (e.g., reading accuracy vs. reading fluency vs. reading comprehension)? Because Planning and Attention are operationalized with measures that involve response times, we hypothesized that Planning and Attention would be more strongly related to reading/math fluency than reading/math accuracy. Furthermore, taking into consideration that Simultaneous processing requires comprehension of word relations and spatial orientation, we hypothesized that Simultaneous processing would be more strongly related to reading comprehension/problem solving than reading/math accuracy/fluency.
- (3) Does the PASS–reading/mathematics relation vary as a function of age/grade level? Because the focus of reading/mathematics instruction changes across time (i.e.,

children focus more on decoding/calculations in early grades and on reading comprehension/problem solving in upper grades), the role of PASS processes should also change across time. Based on the findings of previous studies that showed that Successive processing predicts reading in early grades because of its connection to decoding (e.g., Papadopoulos, 2001), we hypothesized that Successive processing would be more strongly related to children in primary schools than in middle/high schools. Because the findings of previous studies that displayed Planning processing predicts reading in upper grades because of its connection to comprehension (e.g., Kendeou et al., 2015), we hypothesized that Planning would be more strongly related to children in middle/high schools and university students.

- (4) Does the PASS–reading/mathematics relation vary as a function of language in which the studies were conducted? To our knowledge, only one study has directly compared the contribution of PASS processes to mathematics across languages (Italian vs. Dutch) and reported no significant differences across languages (Kroesbergen et al., 2010). Because of the dearth of studies that compared the PASS–reading/mathematics relation across languages, we did not formulate a specific hypothesis.
- (5) Does the PASS–reading/mathematics relation vary as a function of sample characteristics (e.g., typically–developing children vs. children with learning disabilities vs. gifted children)? To our knowledge, only one study directly compared the relations of PASS processes in groups of different ability levels and reported

stronger correlations between Simultaneous and Successive processing with problem solving in the group of children with math disabilities than in the group of typically-developing children (Iglesias-Sarmiento, Deaño, Alfonso, & Conde, 2017). Because of the dearth of studies that compared the PASS–reading/mathematics relation with various ability groups, we did not formulate a specific hypothesis

## Method

### Data Collection

The inclusion, search, and coding procedures are detailed in Figure 1. To identify the studies for the meta-analysis, we first searched in electronic databases (i.e., ERIC, PubMed, Medline, PsycINFO, ProQuest Educational, Scopus, and Google Scholar) for publications between January 1997 (the year CAS was published) and March 2019. The following descriptors were used in our search: Set 1 *PASS theory\**, *PASS cognitive processes\**, *planning\**, *attention\**, *simultaneous processing\**, *successive processing\**, *cognitive assessment system\**, *CAS\**, combined with Set 2 - *reading ability\**, *reading achievement\**, *reading skills\**, *reading accuracy\**, *reading fluency\**, *reading comprehension\**, *character recognition\**, *oral reading\**, *decoding\**, *word recognition\**, and/or Set 3 - *math ability\**, *math performance\**, *arithmetic\**, *math achievement\**, *calculation fluency\**, *problem solving\**, *math skills\**, *numeracy skills\**. Within each set the OR command was used and between sets the AND command was used. Second, we searched for additional papers in e-books, interpretive and technical manuals (Naglieri & Das, 1997; Naglieri et al., 2014; Naglieri & Otero, 2017), and the reference lists of the studies identified through the initial database search. Finally, we contacted all authors who published studies on PASS processes and asked for any unpublished data.

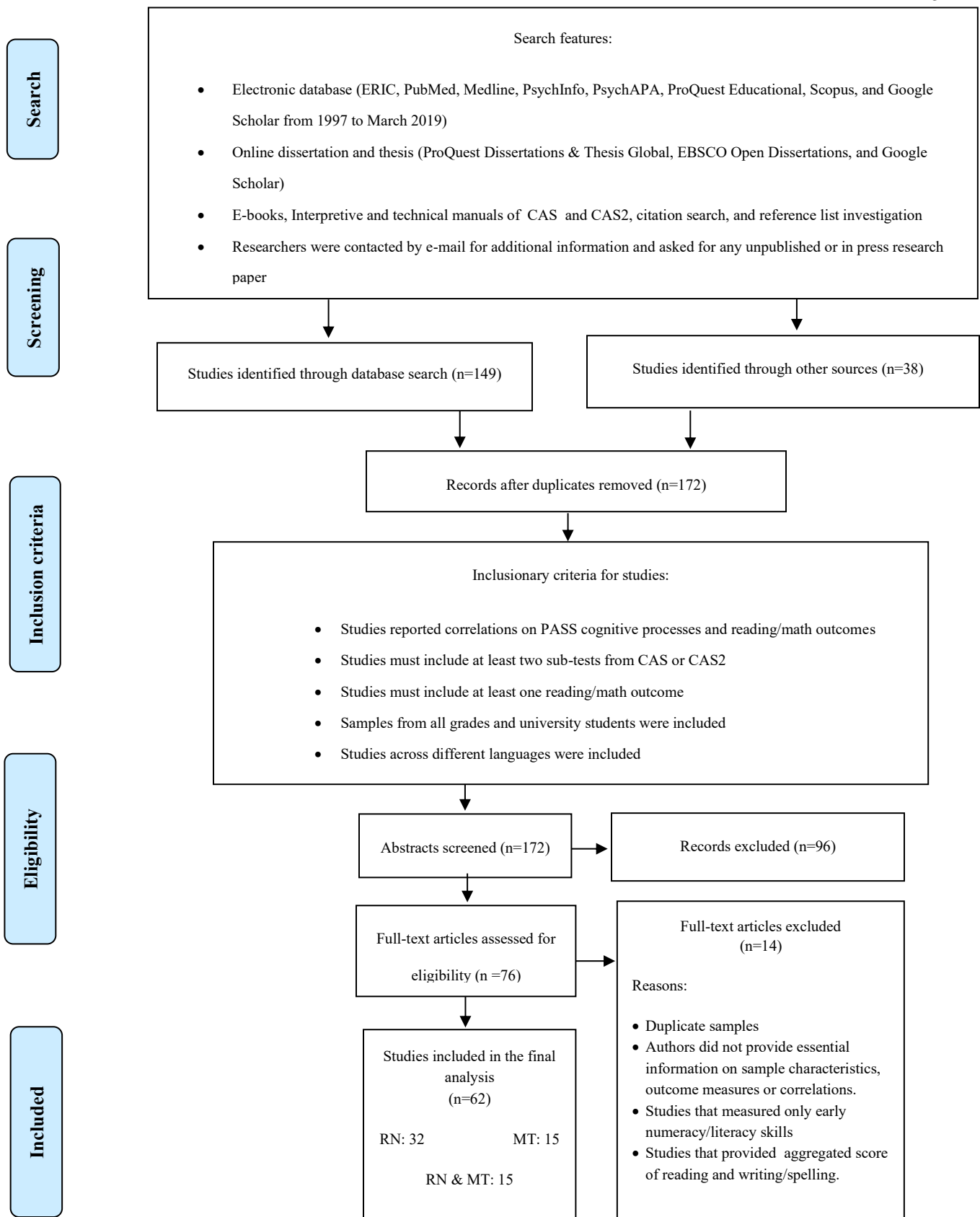


Figure 1. Flow diagram for the search and inclusion on studies. RN = Reading; MT = Mathematics



## **Operational Criteria for Inclusion and Elimination of Studies**

For the target constructs included in this study, we first established operational criteria to determine the indicators of each construct. In regard to PASS processes, studies were considered if they had assessed one of the PASS processes (e.g., Planning) with at least two tasks from either edition of the CAS (Naglieri & Das, 1997; Naglieri et al., 2014) or more than one PASS process (e.g., Planning and Attention) with at least one task (e.g., Planned Codes to operationalize Planning and Expressive Attention to operationalize Attention)<sup>1</sup>. Studies that included only one measure of CAS (most often the Nonverbal Matrices as an indicator of nonverbal IQ) were excluded.

In regard to reading, we considered four types of outcomes (reading accuracy, reading fluency, reading comprehension, and Broad Reading). To be considered a measure of reading accuracy the task should require individuals to read aloud words or nonwords without any time limits. A task was considered a measure of reading fluency if it required individuals to read as many words, nonwords, or sentences as quickly and as accurately as possible within a specified time limit. Text reading speed was also considered a measure of reading fluency. To be considered a measure of reading comprehension, the task should require individuals to answer questions about a story they read or provide the missing word to complete accurately the meaning of a sentence. Finally, we included studies that reported correlations between PASS processes

and Broad Reading (a cluster score derived from combining scores in reading accuracy, fluency, and comprehension).

In regard to mathematics, we considered four types of outcomes (math accuracy, math fluency, problem solving, and Broad Math). To be considered a measure of mathematics accuracy the task should require individuals to solve calculations under untimed conditions. To be considered a measure of mathematics fluency, the task should require individuals to solve as many arithmetic problems as possible within a given time limit. Problem solving tasks included either mathematical reasoning or applied math problems. Finally, we considered studies that reported correlations between PASS processes and Broad Math (a cluster score derived from combining scores in calculation, math fluency, and applied problems).

We further applied the following three exclusionary criteria:

- 1) To avoid including the same data from more than one study, we selected the study that was published earlier and excluded the later studies.
- 2) If a dissertation was also published as an article, we only considered the article.
- 3) The studies that examined the relation between PASS and academic achievement before 1997 were excluded.

After these criteria were applied, we identified 62 studies with 93 unique samples and the size of the samples ranged from 20 to 1691 participants. Of the 62 studies, one was published in Chinese and two in Portuguese.

### **Coding Procedures**

To begin the coding process, we first created a coding spreadsheet and the following components of the selected studies were coded: a) mean age of the participants at the time of assessment, b) grade level, c) language in which the study was conducted, d) sample characteristics (e.g., reading: unselected, good readers, or poor readers; mathematics: unselected, good mathematicians, or poor mathematicians), e) type of reading outcome (accuracy/fluency/comprehension/Broad Reading), and f) type of math outcome (accuracy/fluency/problem solving/Broad Math).

Second, we coded all the effect sizes for each of the target constructs. Several studies reported more than one measure to examine the association between PASS processes and reading/math. To have one effect size per construct, we established a set of rules. For PASS processes, if more than one subtest (e.g., Matching Numbers, Planned Codes, Planned Connections) was used to measure a PASS process (i.e., Planning), an arithmetic mean of  $r$  values was coded. For reading accuracy, reading fluency, and reading comprehension tasks, the multiple effect sizes for each construct (e.g., Word Identification and Word Attack as indicators

of reading accuracy) were aggregated using an arithmetic mean. Similarly, for math accuracy, math fluency, and problem-solving tasks, when there was more than one effect size for each construct, the average  $r$  was coded. Third, for longitudinal studies, the data from the first measurement of reading and/or math ability was coded.

Finally, to ensure accuracy of coding, all 62 studies were coded individually by myself and a graduate student who received training in the coding procedures. The data were recorded into two coding spreadsheets (one for the reading studies and one for the math studies; see Appendix A and B) and the interrater agreement was calculated. The consensus rate varied between 95% and 98%. Differences in coding were due to inadequate information provided in some studies about their participants and measures. The discrepancy between the coders was resolved by revisiting the studies and after having a discussion between the two raters.

### **Moderator Variables**

In each study, we coded five moderators that could explain the variation between studies. Studies that reported effect sizes from a pooled sample of poor and good readers/math performers were not coded.

**Age.** The participants' age was coded in three ways: a) If both the age range and mean age were reported and the age range was not larger than one year, the mean age was coded (e.g., the age 9.14 years was coded, when the age ranged between 9-10 years); b) If the age range alone

was reported and it was not larger than one year, the median value was coded (e.g., when the age range was 9-10 years, 9.5 years was coded as age); and c) The mean age was coded, when the mean age was solely reported. The studies that reported the age range (i.e., larger than one year) without providing information on the mean age were excluded from the moderator analyses.

**Grade level.** Grade was coded as a moderator variable. The studies that had samples from different grades and reported the effect sizes separately for each grade were included in the moderator analyses. Studies were excluded if they assessed children from different grades and reported results from the pooled sample.

**Sample characteristics.** Reading performance was coded to differentiate the ability level of the samples. The studies that had unselected samples of children were coded as “unselected”. The sample that consisted of gifted children, good comprehenders, and skilled readers were coded as “good readers”. The participants described as reading below grade level, poor comprehenders, and less-skilled readers were coded as “poor readers”. Two of the studies with a sample of children with Attention Deficit Hyperactive Disorder (ADHD), emotional or behavioral problems and one study with children with mild developmental disorders were eliminated from the moderator analyses.

Math performance was also coded to differentiate the ability level of the samples. The studies that had unselected samples of children were coded as “unselected”. The samples that

consisted of gifted children were coded as “good performers”. The participants described as having below grade level math performance were coded as “poor performers”. Three studies that included children with Math Learning Disabilities (MLD) in their sample and reported combined scores were excluded. Although, in one study, the correlations were reported separately for the participants with MLD, neither was included.

**Language.** The majority of the studies were conducted in English and we coded them as “English”. The studies in Greek, Portuguese, Dutch, Italian, and Spanish were coded as “other European languages”, and the studies in Chinese, Oriya, Arabic, and Malay were coded as “non-European languages”. Finally, four studies with English Language Learners were coded as “ELL”.

**Task type.** The reading outcome tasks were classified into accuracy, fluency, and comprehension. Likewise, math tasks were categorized into accuracy, fluency, and problem solving. Thirteen studies that reported correlations between PASS processes and Broad Reading, and 10 studies that provided correlations between PASS processes and Broad Math were excluded from the moderator analysis.

### **Statistical Analysis**

The metafor package for the R statistical program (Viechtbauer, 2010) was used for the analyses. The effect sizes for all the studies were displayed by the Pearson’s  $r$  correlation

coefficient. When a correlation between PASS Full Scale and reading/mathematics outcome was available, it was used before the mean of  $r$  values of other subtests of PASS. For both reading and mathematics outcomes we estimated the overall weighted average effect using a random-effects model (Borenstein, Hedges, Higgins, & Rothstein, 2009) instead of a fixed-effects model, because it rests on the assumption that variation between studies can be systematic and not only due to random error. Whether or not the overall effect size differed from zero was tested with a  $z$  test. The 95% CI was also calculated for each overall effect size to provide more information about the correlation.

To examine whether variation in the  $r$  value between studies was significant, the  $Q$  test of homogeneity was used (Hedges & Olkin, 2014). A significant value on this test indicates a reliable variability between the effect sizes in the sample of studies.  $I^2$  was used to determine the magnitude of the heterogeneity.  $I^2$  is the proportion of total variation between effect sizes that is caused by real heterogeneity rather than chance. Values around 25% are typically considered 'low', values around 50% 'moderate', and values around 75% 'high' (Higgins et al., 2003).

Moderator variables were also explored as potential sources of additional variance in the effect sizes. Linear models were used to predict the study's outcome from the moderator variables, both for the continuous (i.e., age) and categorical (i.e., grade level, task type, sample characteristics, language) moderators. For a continuous moderator, a regression coefficient was estimated, and a  $z$  test was used to determine the significance. The degree of differences between

the subsets of studies was tested with a  $Q$  test and by comparing the correlation magnitude with CIs between the study subsets.

### **Publication Bias**

To test for publication bias, we first computed Rosenthal's Fail-Safe N and we also conducted the Rank Correlation and Egger's Regression tests. These tests examine the relationship between the size of the effects from each study and the associated standard error. Furthermore, funnel plots were created to assess whether the studies were distributed asymmetrically around the mean effect size, which may also indicate the presence of publication bias (Borenstein et al., 2009). In the funnel plot, the sample size is plotted on the  $y$  axis and the effect size on the  $x$  axis. In the absence of retrieval bias, this plot should form an inverted funnel. In the presence of bias, the funnel should be asymmetric. Finally, the "trim and fill" method for random-effects models (Duval & Tweedie, 2000) was used in order to examine the impact of possible missing studies. The "trim and fill" method imputes values to make the funnel plot symmetrical and calculate an estimated overall effect size on this basis.



## Results

### Study Features

Of the 62 studies that were included in our final analysis, 15 reported results on both reading and math outcomes, 32 reported results on only reading, and 15 on only math. There were 13,356 participants represented, with sample sizes ranging from 20 to 1,691. The mean age reported in the studies ranged from 4.91 to 22.26 years, and the grade level ranged from kindergarten to adults.

### Meta-Analytic Results

The random-effects model demonstrated that the overall mean correlations between PASS and both reading, and math outcomes were significant (see Table 1). For reading, the mean effect size across the 66 effects from 47 studies was  $r = .409$  ( $z = 17.666$ ,  $p < .0001$ , 95% CI = [.363, .454]; see also Figure 2 for the forest plot), indicating a large effect size. The mean effect size for studies that reported correlations between PASS Full Scale and reading was even larger,  $r = .605$  ( $z = 21.236$ ,  $p < .0001$ , 95% CI = [.549, .661]).

Table 1

*Overall Meta-Analytic Results*

Outcomes	$k$	$n$	$r$	S.E.	Z value	$p$ value	95% CI	Heterogeneity		
								$I^2$ (%)	$Q$	$p$ value
Reading <sup>a</sup>	66	11230	.409	.023	17.666	< .0001	[.363, .454]	90.31	688.335	< .0001
	20	5902	.605	.029	21.236	< .0001	[.549, .661]	90.73	102.267	< .0001
Math <sup>a</sup>	48	8621	.461	.029	16.110	< .0001	[.405, .517]	93.09	452.096	< .0001
	22	6063	.615	.022	28.041	< .0001	[.572, .658]	82.90	68.527	< .0001

*Note:* <sup>a</sup>. The second row under reading or mathematics refers to the estimates obtained when the PASS Full Scale was used.  $k$  = number of correlations;  $n$  = total sample size;  $r$  = estimated correlation size (Pearson's  $r$ ) in random-effects model;  $I^2$  = the proportion of total variation caused by real heterogeneity;  $Q$  = Hedge's  $Q$  test of homogeneity

The overall effect size for math (estimated from 48 effects and 30 studies) was also large ( $r = .461$ ;  $z = 16.110$ ,  $p < .0001$ , 95% CI = [ .405, .517]; see also Figure 3 for the forest plot).

Again, the mean effect size for studies that reported correlations between PASS Full Scale and mathematics was larger,  $r = .615$  ( $z = 28.041$ ,  $p < .0001$ , 95% CI = [ .572, .658]. The

heterogeneity analysis further showed that the variation between studies was significant and large for both reading ( $Q = 688.335$ ,  $I^2 = 90.31\%$ ,  $p < .0001$ ) and mathematics ( $Q = 452.096$ ,  $I^2 = 93.09\%$ ,  $p < .0001$ ).

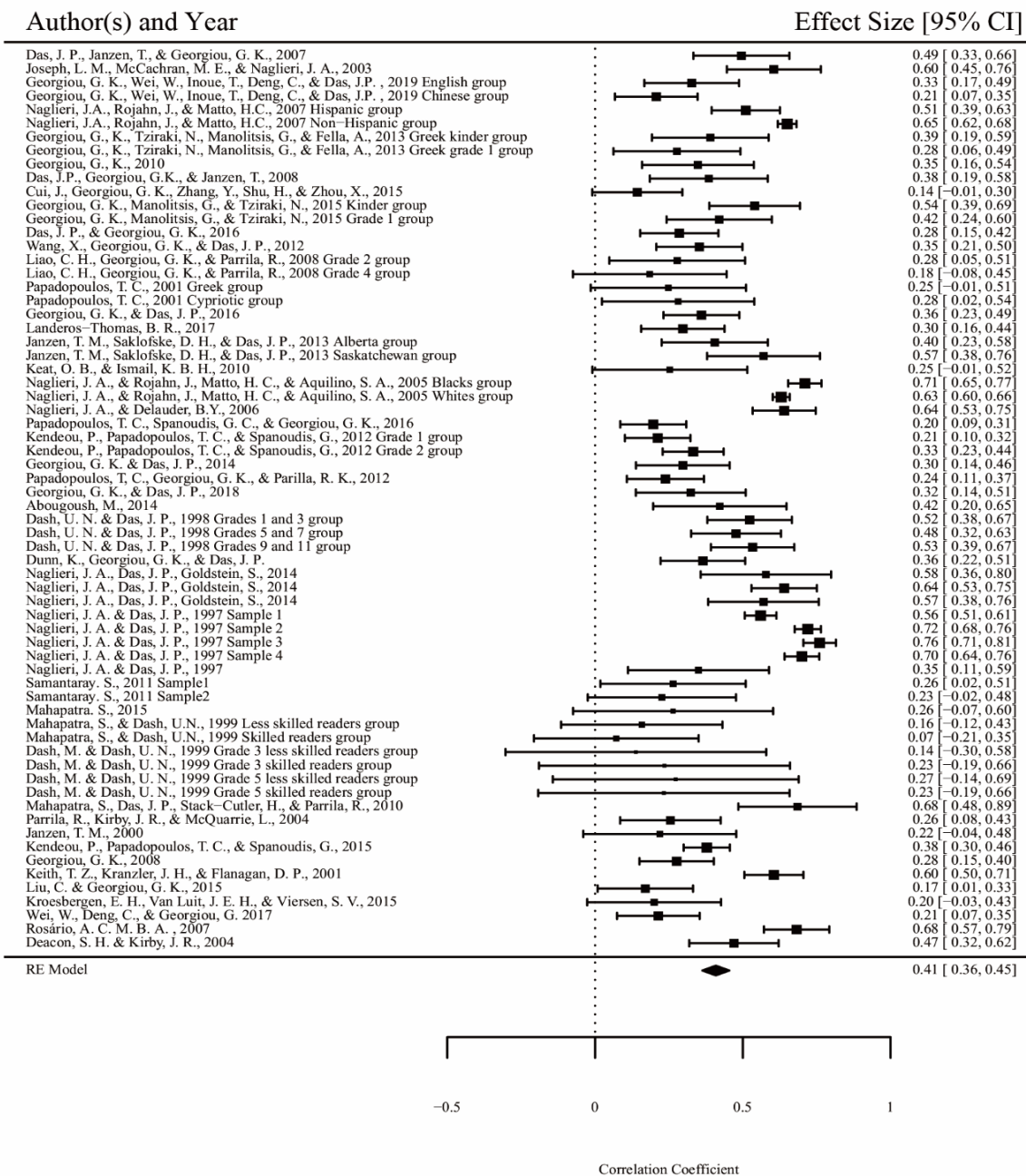


Figure 2. Forest plot: Strength of the correlations between PASS and reading

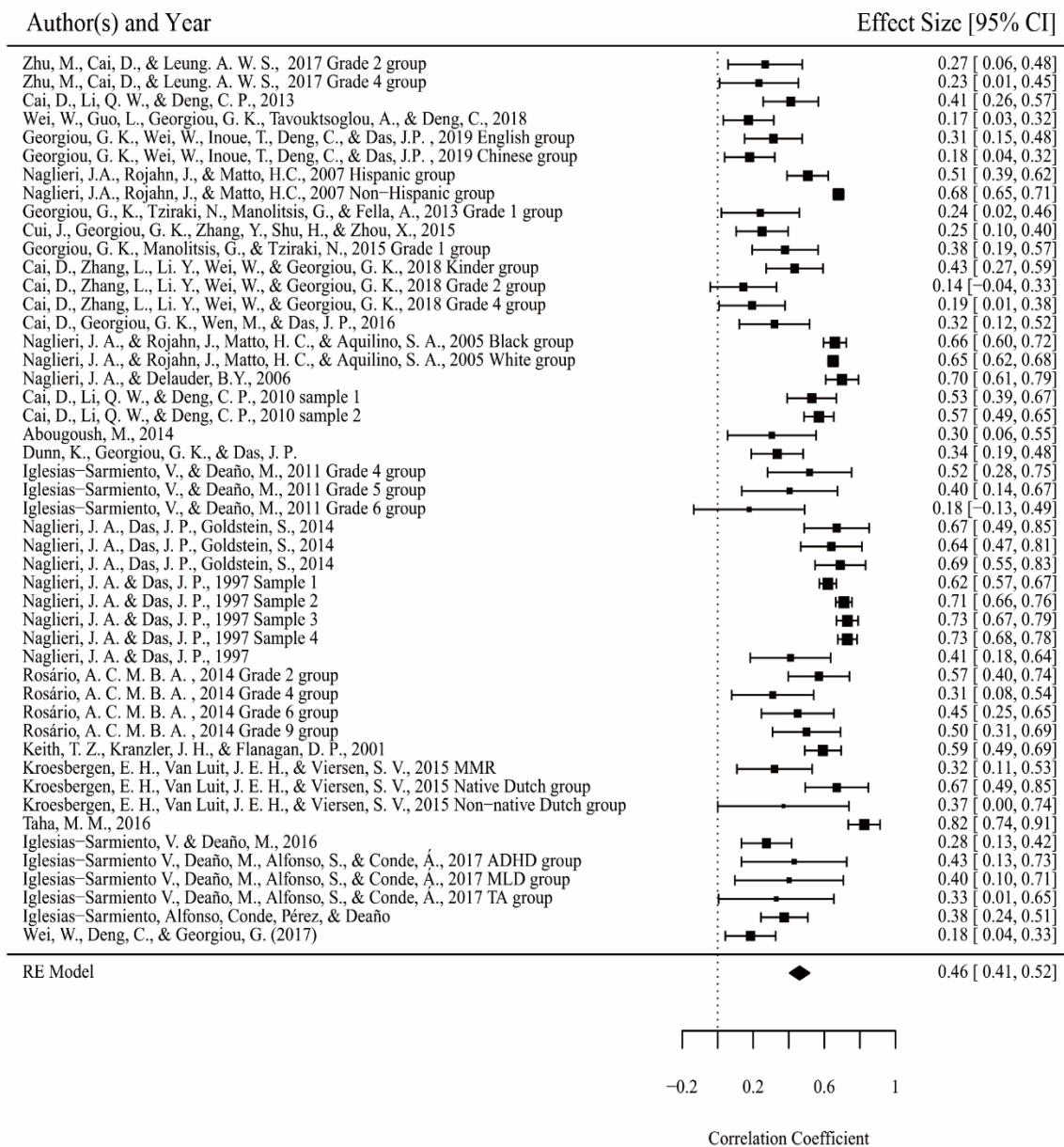


Figure 3. Forest plot: Strength of the correlations between PASS and math

## Results of the Moderator Analyses

First, we examined if the type of reading/mathematics outcome moderates the PASS–reading/mathematics relations. The results are presented in Tables 2 and 3. When considering differences among PASS processes and outcome subtypes, the correlations were stable across PASS or outcome subtypes for reading. However, the correlations varied significantly for mathematics. First, Simultaneous processing produced significantly stronger correlations with math accuracy ( $.416 > .179$ ,  $z = 3.3523$ ,  $p = .0008$ ) and problem solving ( $.478 > .179$ ,  $z = 4.2783$ ,  $p < .0001$ ) than math fluency. Second, Planning correlated more strongly with math fluency than Simultaneous processing ( $.421 > .179$ ,  $z = 2.6455$ ,  $p = .0082$ ). Finally, Simultaneous processing correlated more strongly with problem solving than Attention ( $.478 > .342$ ,  $z = 2.1169$ ,  $p = .0343$ ).

Table 2

*Moderator Analyses for Reading: Categorical Moderator Variables (PASS and Outcome Subtype)*

Moderator variable	Number of correlations ( <i>k</i> )	Correlation ( <i>r</i> )	<i>p</i> value	95% CI	Difference in <i>r</i> (highest-lowest category)	Significance test of difference ( <i>Q</i> test)	<i>p</i> value
Planning					.032	.2849	.8672
Accuracy	33	.347	< .0001	[.286, .409]			
Fluency	11	.315	< .0001	[.214, .416]			
Comprehension	26	.330	< .0001	[.261, .399]			
Attention					.032	.5042	.7772
Accuracy	25	.292	< .0001	[.223, .361]			
Fluency	13	.322	< .0001	[.229, .415]			
Comprehension	19	.324	< .0001	[.248, .400]			
Simultaneous					.105	2.4599	.2923
Accuracy	34	.355	< .0001	[.295, .414]			
Fluency	12	.310	< .0001	[.210, .409]			
Comprehension	21	.415	< .0001	[.343, .487]			
Successive					.033	.5938	.7431
Accuracy	29	.368	< .0001	[.304, .433]			
Fluency	15	.353	< .0001	[.267, .439]			
Comprehension	20	.386	< .0001	[.311, .460]			

*Note:* *k* = number of correlations; *r* = correlation size (Pearson's *r*) for subsets of studies belonging to different categories of the moderator variable.

Table 3

*Moderator Analyses for Math: Categorical Moderator Variables (PASS and Outcome Subtype)*

Moderator variable	Number of correlations ( <i>k</i> )	Correlation ( <i>r</i> )	<i>p</i> value	95% CI	Difference in <i>r</i> (highest-lowest category)	Significance test of difference ( <i>Q</i> test)	<i>p</i> value
Planning					.061	.6714	.7148
Accuracy	13	.409	< .0001	[ .313, .504]			
Fluency	6	.421	< .0001	[ .277, .564]			
Problem solving	12	.470	< .0001	[ .368, .571]			
Attention					.028	.1532	.9262
Accuracy	16	.349	< .0001	[ .262, .436]			
Fluency	10	.321	< .0001	[ .206, .435]			
Problem solving	16	.342	< .0001	[ .253, .433]			
Simultaneous					.299	18.4090	.0001
Accuracy	15	.416	< .0001	[ .327, .505]			
Fluency	12	.179	.0009	[ .073, .285]			
Problem solving	17	.478	< .0001	[ .392, .564]			
Successive					.144	4.1787	.1238
Accuracy	12	.320	< .0001	[ .219, .422]			
Fluency	8	.250	.0002	[ .120, .379]			
Problem solving	12	.394	< .0001	[ .290, .498]			

*Note:* *k* = number of correlations; *r* = correlation size (Pearson's *r*) for subsets of studies belonging to different categories of the moderator variable



Next, we examined the role of language, grade level, age, and sample characteristics in the PASS–reading/mathematics relation. As shown in Table 4, language was a significant moderator of the PASS–reading relation. Studies with English–speaking participants produced significantly larger correlations than studies in which the participants spoke other European or non-European languages ( $p < .001$ ). Grade level, reading level, and mean age ( $\beta = .0007, p = .9003, k = 43$ ) did not reliably explain variation in the correlations. Language was also a significant moderator in the PASS–mathematics relation: studies with English–speaking participants produced significantly larger correlations than studies in other languages (see Table 5). The difference between other European and non-European languages was also significant. The correlations between PASS and mathematics were stable across different grade levels, math level range, and mean age ( $\beta = .0136, p = .2452, k = 30$ ).

Table 4

*Moderator Analyses for Reading: Categorical Moderator Variables*

Moderator variable	Number of correlations ( <i>k</i> )	Correlation ( <i>r</i> )	<i>p</i> value	95% CI	Difference in <i>r</i> (highest-lowest category)	Significance test of difference ( <i>Q</i> test)	<i>p</i> value
Language					.198	21.2367	< .0001
English	31	.503	< .0001	[.447, .559]			
Other European	12	.316	< .0001	[.224, .408]			
Non-European	19	.305	< .0001	[.224, .386]			
ELL	4	.390	< .0001	[.202, .579]			
Grade					.224	2.0580	.5605
Kindergarten	4	.309	< .0001	[.153, .464]			
G1 to G6	42	.365	< .0001	[.314, .416]			
G7 to G12	1	.533	.0005	[.235, .831]			
Adults	4	.317	< .0001	[.165, .468]			
Sample characteristics					.119	2.3237	.3129
Unselected	50	.416	< .0001	[.366, .466]			
Poor readers	5	.326	.0010	[.131, .520]			
Good readers	6	.297	.0007	[.125, .470]			

*Note:* *k* = number of correlations; *r* = correlation size (Pearson's *r*) for subsets of studies belonging to different categories of the moderator variable.

Table 5

*Moderator Analyses for Math: Categorical Moderator Variables*

Moderator variable	Number of correlations ( <i>k</i> )	Correlation ( <i>r</i> )	<i>p</i> value	95% CI	Difference in <i>r</i> (highest-lowest category)	Significance test of difference ( <i>Q</i> test)	<i>p</i> value
Language					.250	22.0784	< .0001
English	17	.601	< .0001	[ .528, .675]			
Other European	17	.403	< .0001	[ .315, .490]			
Non-European	14	.351	< .0001	[ .265, .436]			
Grade					.176	2.3173	.3139
Kindergarten	3	.405	< .0001	[ .259, .552]			
G1 to G6	25	.324	< .0001	[ .267, .381]			
G7 to G12	1	.500	.0005	[ .219, .782]			
Sample characteristics					.106	.7245	.3947
Unselected	28	.456	< .0001	[ .384, .528]			
Good performers	3	.350	.0034	[ .116, .584]			

*Note:* *k* = number of correlations; *r* = correlation size (Pearson's *r*) for subsets of studies belonging to different categories of the moderator variable.

## Publication Bias

The results of the Fail-Safe N analysis suggested that the estimated effect sizes were reasonably stable. More than 60,000 additional participants would be needed to achieve a null  $p$  value for each outcome ( $N = 81,128$  for reading,  $N = 66,470$  for math). The results of the Egger's Regression Test suggested the presence of publication bias in both the reading ( $z = -5.3765, p < .0001$ ) and the mathematics ( $z = -4.9684, p < .0001$ ) model (see Table 6). As suggested by the Rank Correlation Test, Kendall's tau for reading was not significant and tau for mathematics was significant ( $\text{tau} = -.2287, p = .0218$ ). Subsequently, the "trim and fill" analyses were performed. The funnel plot indicated that studies were missing to the right of the mean (i.e., studies with effect sizes over the overall mean) (see Figures 4 and 5). Thus, the true effect size may be somewhat higher for reading (corrected effect size = .470) and mathematics (corrected effect size = .566) than what has been reported in the initial analyses.

Table 6

*Publication Bias Analyses*

Outcomes	Fail-Safe N	Egger's Method		Rank Correlation Test		Trim and Fill Procedure	
		<i>z</i>	<i>p</i>	Kendall's tau	<i>p</i>	Imputed	Corrected effect sizes
Reading	81128	-5.3765	< .0001	- .1170	.1671	17	.470
Math	66470	-4.9684	< .0001	-. 2287	.0218	16	.566

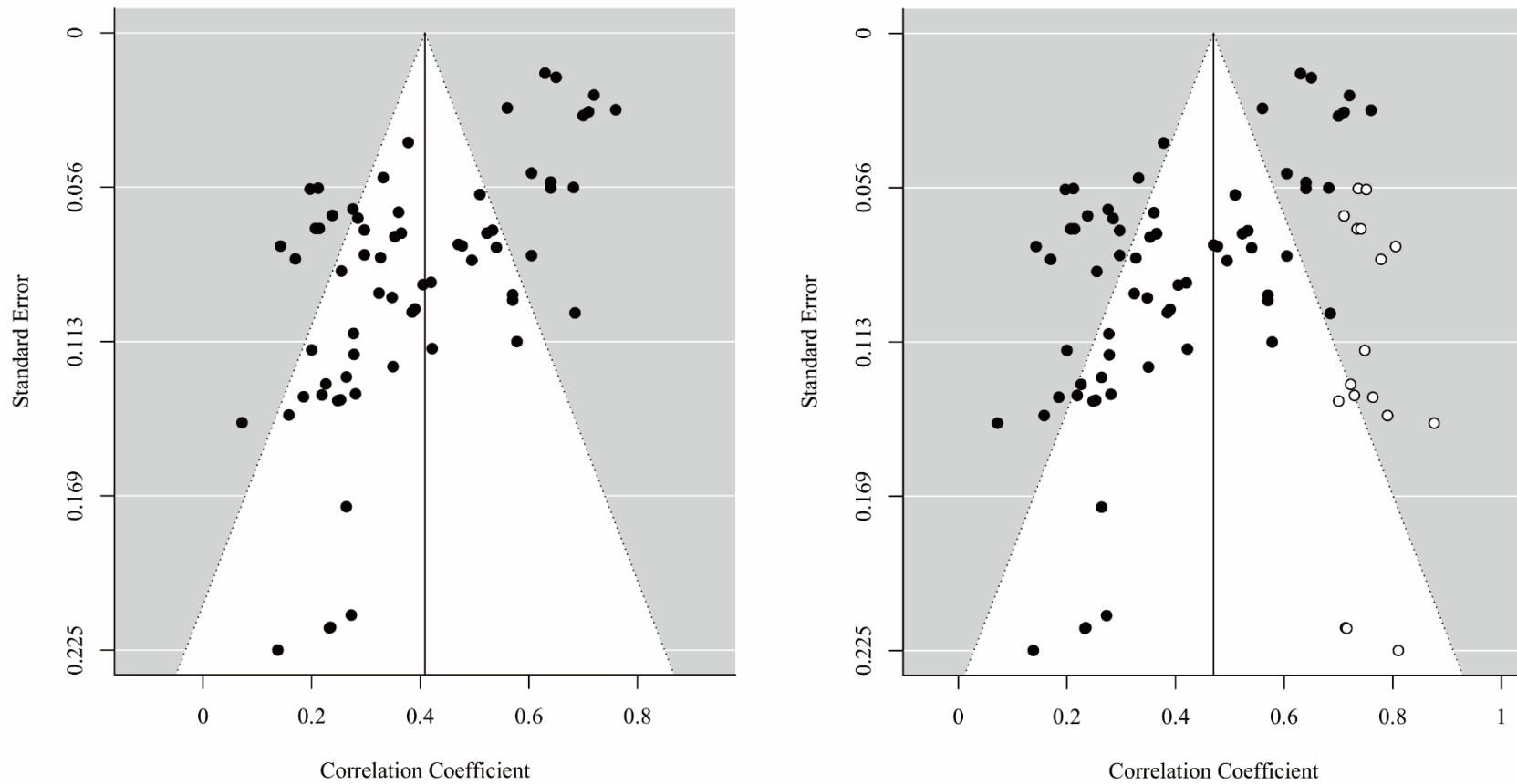


Figure 4. Funnel plots for reading (left) and funnel plots with imputed samples for reading (right).

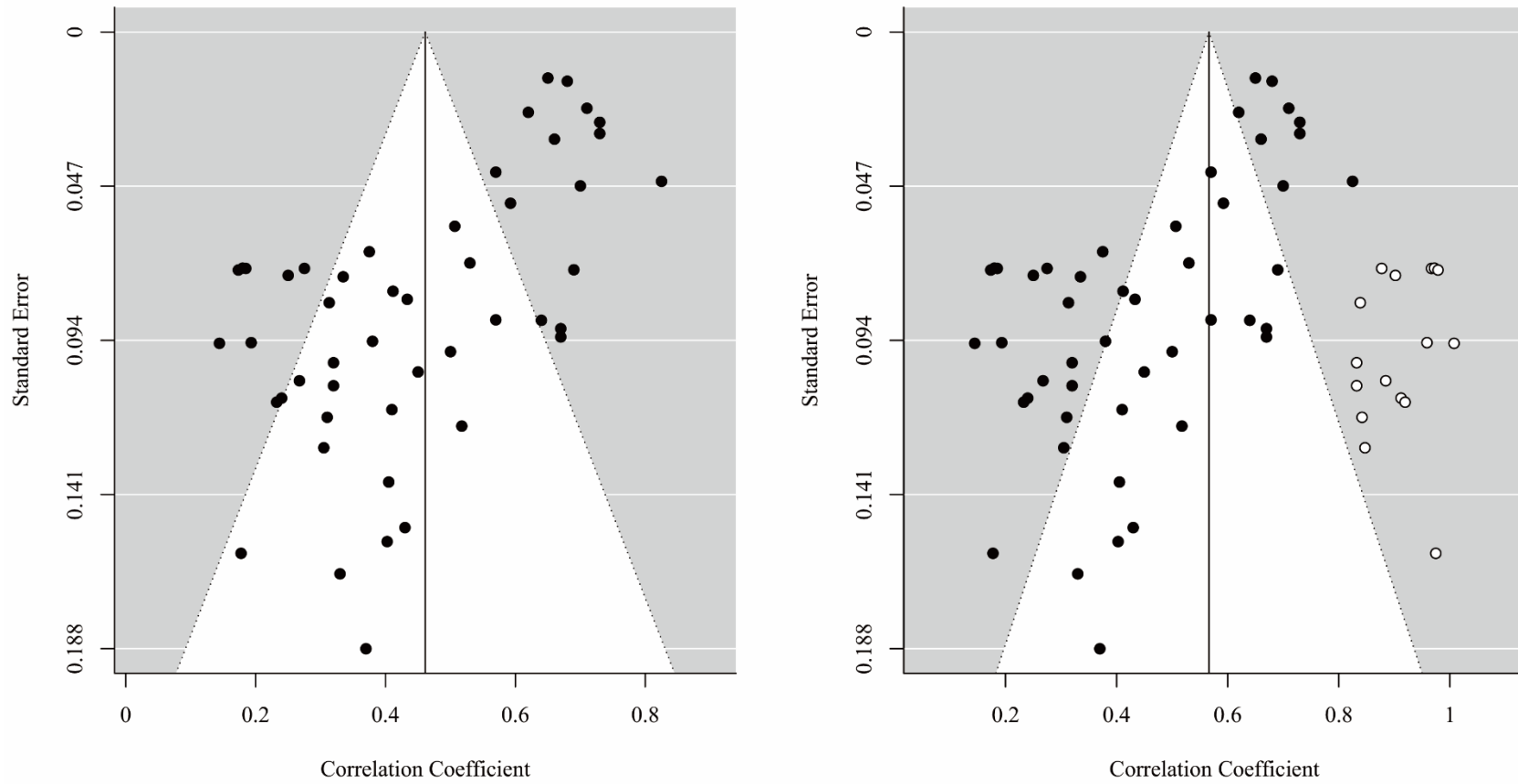


Figure 5. Funnel plots for math (left) and funnel plots with imputed samples for math (right)

## Discussion

The purpose of this meta-analysis was to estimate the size of the relation between PASS processes and reading/mathematics performance and if different factors (type of reading/mathematics outcome, age/grade level, language, and sample characteristics) moderate their relation. When any correlation by any PASS process was included in the statistical analysis, we found significant relations between the PASS processes and reading or mathematics (the average mean correlation was .409 and .461, respectively). These correlations are similar to those reported in previous meta-analyses on the relation between intelligence and academic achievement (e.g., Peng et al., 2019; Roth et al., 2015; Swanson, Trainin, Necochea, & Hammill, 2003). They are also as strong as those reported in previous meta-analyses for key predictors of reading (e.g., phonological awareness, rapid naming; see Melby-Lervåg, Lyster, & Hulme, 2012; Ruan, Georgiou, Song, Li, & Shu, 2018; Swanson et al., 2003) and mathematics (e.g., approximate number system, working memory; see Chen & Li, 2014; Peng, Namkung, Barnes, & Sun, 2016; Schneider et al., 2017).

One could argue though that PASS theory is not adequately represented by these correlations as they were calculated by considering the correlations of individual PASS processes with reading/mathematics and not by considering the correlations generated by the combination of PASS processes (indexed by PASS Full Scale score). Indeed, when we repeated the analyses with the PASS Full Scale that takes into account the scores across all four sub-processes, the



correlations were significantly larger ( $r = .605$  for reading and  $r = .615$  for mathematics; see Table 1). Although we do not directly compare these correlations to the ones generated by other IQ tests (obviously this is beyond the scope of this meta-analysis; however, see Naglieri et al., 2006; Naglieri & Otero, 2018), to our knowledge, none of the previous meta-analyses examining the relation between intelligence and academic achievement (e.g., Peng et al., 2019; Roth et al., 2015; Zaboski, Kranzler, & Gage, 2018) have produced equally strong correlations. This is remarkable if we consider that comparing PASS correlations with academic achievement to, for example, Wechsler or Woodcock ability tests to academic achievement puts PASS at a relative disadvantage because the measures included in CAS do not contain subtests with considerable knowledge requirements such as Vocabulary and Arithmetic, and even more subtle tasks in a scale like Fluid Reasoning, which also demands some knowledge.

However, we also found great heterogeneity in the correlations with reading and mathematics performance. Language explained some of this heterogeneity. Larger correlations with reading/mathematics performance were reported in English than in other languages. An explanation might be that CAS was originally developed in English and the adaptations that followed in other languages did not produce the desirable outcome. We acknowledge that language constraints may be partly responsible for that. For example, in Chinese, there is no present continuous tense and items like “The blue is yellowing the green” in the Sentence Repetition task do not have a direct translation. Unfortunately, many of the studies conducted in

these other languages (particularly those conducted in India) failed to provide information on the psychometric properties of the CAS tasks (e.g., Dash & Das, 1998; Mahapatra & Dash, 1999; Samantaray, 2011) and, as a result, we do not know how well the CAS measures behaved. Notice also that these studies are associated with the highest standard error (see Figure 2).

Our results further showed significant differences in the relations of the four PASS processes with mathematics performance. In line with our expectation, Planning correlated more strongly with math fluency than Simultaneous processing and, in turn, Simultaneous processing correlated more strongly with problem solving than Attention. Math proficiency comprises computing and solving word problems (see Das & Misra, 2015, for a math proficiency model). Whereas computing is dependent on planning and executive functions, word problems that involve logical–grammatical relations rely more on Simultaneous processing. An alternative explanation may relate to the nature of the tasks. Because the Planning measures were all speeded, this may have inflated their relation with math fluency as opposed to Simultaneous processing tasks that did not have any speed requirements. Interestingly, no differences between the PASS processes and the reading outcomes were found. This reinforces the findings of previous studies in different languages (e.g., Kendeou et al., 2015; Naglieri & Rojahn, 2004) suggesting that all PASS processes are important in reading.

Age, grade level, and sample characteristics did not moderate the PASS–reading/math relations either. We interpret this to be evidence of domain general processes that are best

described as cognitive universals. These are represented in the broad functional organization of the brain as proposed by Luria (1966, 1973). The present meta-analysis, based as it is on 62 empirical studies, supports the idea that PASS cognitive functions provide the foundation for the development of specific skills associated with reading and mathematics.

Some limitations of our study are worth mentioning. First, we acknowledge that some of the categories in the moderator analyses did not have many studies. For example, when examining the role of grade level in the PASS–reading relation, we only had one study in the 7–12 grade range, four studies in kindergarten, and four studies in adults. This may have inflated the standard error and reduced our chances to find significant differences. Second, we chose to examine the relations of PASS processes after the publication of CAS in 1997. We acknowledge that some studies with tasks that were subsequently included in CAS were published before 1997 (e.g., Das, Snart, & Mulcahy, 1982; Kirby & Das, 1977; Leong, Cheng, & Das, 1985). Third, our study showed no significant differences in the role of PASS processes in reading across the four groups we created in our meta-analysis. This finding is based on correlations obtained from studies conducted in different single languages and not from cross–linguistic studies that also control for other confounding variables (e.g., family’s socioeconomic background). Indeed, our meta-analysis has shown that very little cross–linguistic research on PASS processes has been done (see Kroesbergen et al., 2010; Naglieri et al., 2007, for exceptions). Fourth, we acknowledge that we examined here the relations of the CAS measures with academic

achievement, not the more broadly defined PASS processes. Clearly, the CAS was designed with PASS in mind, but the CAS tests are not the only measures of PASS processes. Fifth, we did not control for the role of instruction in the relations between PASS and reading/mathematics. Different forms of instruction may alter the cognitive processes brought to bear on particular tasks; for example, some education systems may employ arithmetic drills more than others, perhaps increasing calculation fluency and perhaps reducing the correlation with generic processing abilities. Finally, because the number of studies within each academic domain was relatively small, we could not further test for the effects of different interaction terms.

To conclude, the present meta-analysis adds to a growing body of research examining the role of intelligence in academic achievement (e.g., Peng et al., 2019; Roth et al., 2015) suggesting that there are significant benefits if we conceptualize intelligence as a constellation of cognitive processes that are linked to the functional organization of the brain. First, these cognitive processes (operationalized here with CAS) can produce correlations that are stronger than those derived from popular IQ batteries (e.g., WISC) that include tasks (e.g., Arithmetic, Vocabulary) whose content is often confounded by school learning. Second, these processes have direct implications for instruction and intervention programming. For example, cognitive strategy instruction based on PASS processes has been found to improve children's math calculation (Iseman & Naglieri, 2011) and PREP has been found to improve children's decoding (Papadopoulos et al., 2004) and reading comprehension (Mahapatra et al., 2010). However, this

meta-analysis has also revealed areas in which more research is needed. This includes studies on PASS and academic achievement across languages as well as studies with specific student populations such as poor or good readers/mathematicians.

## References

- \* An asterisk precedes the studies that were included in the meta-analysis.
- \* Abougoush, M. (2014). *PASS theory of intelligence and giftedness*. Master's Thesis, University of Alberta, Edmonton, Alberta.
- Barton, K., Dielman, T. E., & Cattell, R. B. (1972). Personality and IQ measures as predictors of school achievement. *Journal of Educational Psychology*, *63*, 398-404. doi: 10.1037/h003357
- Borenstein, M., Hedges, L. V., Higgins, J. P. T., & Rothstein, H. R. (2009). *Introduction to meta-analysis*. Chichester, UK: Wiley.
- \* Cai, D., Georgiou, G. K., Wen, M., & Das, J. P. (2016). The role of planning in different mathematical skills. *Journal of Cognitive Psychology*, *28*, 234-241. doi: 10.1080/20445911.2015.1103742
- \* Cai, D., Li, Q. W., & Deng, C. P. (2010). The PASS feature of grade 3–8 students' mathematics learning. *Psychological Science (Chinese)*, *33*, 274–277.
- \* Cai, D., Li, Q. W., & Deng, C. P. (2013). Cognitive processing characteristics of 6th and 8th grade Chinese students with mathematics learning disability: Relationship among working memory, PASS processes, and processing speed. *Learning and Individual Differences*, *27*, 120-127. doi:10.1016/j.lindif.2013.07.008

\* Cai, D., Zhang, L., Li, Y., Wei, W., & Georgiou, G. K. (2018). The role of approximate number system in different mathematics skills across grades. *Frontiers in Psychology, 9*, 1-10.

doi:10.3389/fpsyg.2018.01733

Carlson, J. S., & Das, J. P. (1997). A process approach to remediating word decoding deficiencies in Chapter 1 children. *Learning Disability Quarterly, 20*, 93-102. doi:10.2307/1511217

Chen, Q., & Li, J. (2014). Association between individual differences in non-symbolic number acuity and math performance: a meta-analysis. *Acta Psychologica, 148*, 163–172. doi:

10.1016/j.actpsy.2014.01.016

Coulombe, S., Tremblay, J.-F., & Marchant, S. (2004). *International adult literacy survey: Literacy scores, human capital, and growth across 14 OECD countries*. Ottawa, Ontario: Ministry of Industry.

\* Cui, J., Georgiou, G. K., Zhang, Y., Shu, H., & Zhou, X. (2015). Unpublished data.

Das, J. P. (2002). A better look at intelligence. *Current Directions in Psychology, 11*, 28-32.

doi:0.1111/1467-8721.00162

\* Das, J. P., & Georgiou, G. K. (2016). Levels of planning predict different reading comprehension outcomes. *Learning and Individual Differences, 48*, 24-28. doi:

10.1016/j.lindif.2016.04.004

\* Das, J. P., Georgiou, G. K., & Janzen, T. (2008). Influence of distal and proximal cognitive processes on word reading. *Reading Psychology, 29*, 366-393. doi:

10.1080/02702710802153412

\* Das, J. P., Janzen, T., & Georgiou, G. K. (2007). Correlates of Canadian native children's reading performance: From cognitive styles to cognitive processes. *Journal of School Psychology, 45*, 589-602. doi:10.1016/j.jsp.2007.06.004

Das, J. P., & Misra, S. B. (2015). *Cognitive Planning and Executive Functions*. New Delhi: SAGE.

Das, J. P., Naglieri, J. A., & Kirby, J. R. (1994). *Assessment of cognitive processes: The PASS theory of intelligence*. Boston: Allyn & Bacon.

Das, J. P., Snart, F., & Mulcahy, R. F. (1982). Reading disability and its relation to information integration. In J. P. Das, R. F. Mulcahy, & A. E. Wall (Eds.), *Theory and research in learning disabilities* (pp. 85-110). New York: Plenum.

\* Dash, U. N., & Das, J. P. (1998). Developmental norms for the PASS (Planning-Attention-Simultaneous-Successive) processes: Oriya Adaptation. *Psychology and Developing Societies, 10*, 189-213. doi:10.1177/097133369801000205

\* Dash, M., & Dash, U. N. (1999). Information processing correlates of reading. In U.N. Dash, & U. Jain (Eds.), *Perspectives on Psychology and Social Development* (pp. 304-315). New Delhi, India: Concept Publishing Company.



- \* Deacon, S. H., & Kirby, J. R. (2004). Morphological awareness: Just “more phonological”? The roles of morphological and phonological awareness in reading development. *Applied Psycholinguistics*, *25*, 223–238. doi:10.1017.S0124716404001117
- Deary, I. J., Strand, S., Smith, P., & Fernandes, C. (2007). Intelligence and educational achievement. *Intelligence*, *35*, 13-21. doi:10.1016/j.intell.2006.02.001
- Deng, C., & Georgiou, G. K. (2015). Establishing measurement invariance of Cognitive Assessment System across cultures. In T. C. Papadopoulos, R. Parrila, & J. R. Kirby (Eds.), *Cognition, intelligence, and achievement* (pp. 137-148). New York: Elsevier.
- \* Dunn, K., Georgiou, G. K., & Das, J. P. (in press). The relationship of cognitive processes with reading and mathematics achievement in intellectually gifted children. *Roeper Review*.
- Duval, S., & Tweedie, R. (2000). Trim and fill: A simple funnel-plot–based method of testing and adjusting for publication bias in meta-analysis. *Biometrics*, *56*, 455–463. doi:10.1111/j.0006-341X.2000.00455.x
- Ek, E., Sovio, U., Remes, J., & Järvelin, M. (2005). Social predictors of unsuccessful entrance into the labour market: A socialization process perspective. *Journal of Vocational Behavior*, *66*, 471-486. <https://doi.org/10.1016/j.jvb.2004.02.002>
- Fagan, J. R. (2000). A theory of intelligence as processing: Implications for society. *Psychology, Public Policy, and Law*, *6*, 168–179.

Flashman, J. (2012). Academic achievement and its impact on friend dynamics. *Sociology of Education*, 85, 61-80. <http://dx.doi.org/10.1177%2F0038040711417014>

Gardner, H. (1993). *Frames of mind: The theory of multiple intelligences* (2nd ed.). New York: Basic Books.

\* Georgiou, G. K. (2008). *Why is rapid naming speed related to reading? Examining different theoretical accounts*. Unpublished doctoral dissertation, University of Alberta, Edmonton, Alberta, Canada.

\* Georgiou, G. K. (2010). PASS cognitive processes: Can they explain the RAN-reading relationship. *Psychological Science*, 33, 1291-1298.

\* Georgiou, G. K., & Das, J. P. (2014). Reading comprehension in university students: relevance of PASS theory of intelligence. *Journal of Research in Reading*, 37, S101-S115.  
doi:10.1111/j.1467-9817.2012.01542.x

\* Georgiou, G. K., & Das, J. P. (2016). What components of executive functions contribute to normal and impaired reading comprehension in young adults? *Research in Developmental Disabilities*, 49-50, 118-128. doi:10.1016/j.ridd.2015.12.001

\* Georgiou, G. K., & Das, J. P. (2018). Direct and indirect effects of executive functions on reading comprehension in young adults. *Journal of Research in Reading*, 41, 243-258.

- \* Georgiou, G. K., Manolitsis, G., & Tziraki, N. (2015). Is Intelligence relevant in reading "μανα" and calculating "3+5"? In T. C. Papadopoulos, R. K. Parrila, & J. R. Kirby (Eds.), *Cognition, intelligence, and achievement* (pp. 225-240). San Diego, CA: Academic Press.
- \* Georgiou, G. K., Tziraki, N., Manolitsis, G., & Fella, A. (2013). Is rapid automatized naming related to reading and mathematics for the same reason(s)? A follow-up study from kindergarten to Grade 1. *Journal of Experimental Child Psychology*, *115*, 481–496.  
doi:10.1016/j.jecp.2013.01.004
- \* Georgiou, G. K., Wei, W., Inoue, T., Deng, C., & Das, J. P. (2019). Cultural influences on the relation between executive functions and academic achievement. *Reading & Writing: An interdisciplinary Journal*.
- Hald, M. E. (1999). *A PASS cognitive processes intervention study in mathematics*. (Unpublished doctoral dissertation). University of Northern Colorado, Colorado.
- Hedges, L. V., & Olkin, I. (2014). *Statistical method for meta-analysis*. New York, NY: Academic Press.
- Higgins, J. P., Thompson, S. G., Deeks, J. J., & Altman, D. G. (2003). Measuring inconsistency in meta-analyses. *BMJ*, *327*(7414), 557–560.
- \* Iglesias-Sarmiento, V., Alfonso, S., Conde, Á., Pérez, L., & Deaño, M. (2019). *Mathematical learning disabilities vs. high achievement: an analysis of arithmetical cognition in elementary school*. Manuscript submitted for publication.

- \* Iglesias-Sarmiento, V., & Deaño, M. (2011). Cognitive processing and mathematical achievement: A study with schoolchildren between fourth and sixth grade of primary education. *Journal of Learning Disabilities, 44*, 570-583. doi:0.1177/0022219411400749
- \* Iglesias-Sarmiento, V., & Deaño, M. (2016). Arithmetical difficulties and low arithmetic achievement: Analysis of the underlying cognitive functioning. *The Spanish Journal of Psychology, 19*(e36), 1-14. doi:10.1017/sjp.2016.40
- \* Iglesias-Sarmiento, V., Deaño, M., Alfonso, S., & Conde, Á. (2017). Mathematical learning disabilities and attention deficit and/or hyperactivity disorder: A study of the cognitive processes involved in arithmetic problem solving. *Research in Developmental Disabilities, 61*, 44-54. doi:10.1016/j.ridd.2016.12.012
- Iseman, S. J., & Naglieri, J. A. (2011). A cognitive strategy instruction to improve math calculation for children with ADHD and LD: A randomized controlled study. *Journal of Learning Disabilities, 44*, 184–195. doi:10.1177/0022219410391190
- \* Janzen, T. M. (2000). *Assessment and remediation using the PASS theory with Canadian natives*. Doctoral thesis, University of Alberta, Edmonton, Alberta, Canada.
- \* Janzen, T. M., Saklofske, D. H., & Das, J. P. (2013). Cognitive and reading profiles of two samples of Canadian first nations children: Comparing two models for identifying reading disability. *Canadian Journal of School Psychology, 28*, 323-344. doi: 10.1177/0829573513507419

- \* Joseph, L. M., McCachran, M. E., & Naglieri, J. A. (2003). PASS cognitive processes, phonological processes, and basic reading performance for a sample of referred primary-grade children. *Journal of Research in Reading, 26*, 304–314. doi:10.1111/1467-9817.00206
- \* Keat, O. B., & Ismail, K. B. H. (2010). The PASS cognitive functions of children with reading difficulties: a Malaysian study. *Procedia Social and Behavioral Sciences, 5*, 2182-2193. doi:10.1016/j.sbspro.2010.07.434
- \* Keith, T. Z., Kranzler, J. H., & Flanagan, D. P. (2001). What does the Cognitive Assessment System (CAS) measure? Joint confirmatory factor analysis of the CAS and the Woodcock-Johnson tests of cognitive ability (3rd Edition). *School Psychology Review, 30*, 89-119.
- \* Kendeou, P., Papadopoulos, T. C., & Spanoudis, G. (2012). Processing demands of reading comprehension tests in young readers. *Learning and Instruction, 22*, 354-367. doi: 10.1016/j.learninstruc.2012.02.001
- \* Kendeou, P., Papadopoulos, T. C., & Spanoudis, G. (2015). Reading comprehension and PASS theory. In T. C. Papadopoulos, R. K. Parrila, & J. R. Kirby (Eds.), *Cognition, intelligence, and achievement* (pp. 117-136). New York, NY: Elsevier.
- Kirby, J. R., & Ashman, A. F. (1984). Planning skills and mathematics achievement: Implications regarding learning disability. *Journal of Psychoeducational Assessment, 2*, 9-22. doi:10.1177/073428298400200102.

Kirby, J. R., & Das, J. P. (1977). Reading achievement, IQ, and simultaneous-successive processing. *Journal of Educational Psychology, 69*, 564-570. doi:10.1037/0022-0663.69.5.565

Kirby, J. R., & Robinson, G. L. W. (1987). Simultaneous and successive processing in reading disabled children. *Journal of Learning Disabilities, 20*, 243-252. doi:10.1177/002221948702000409

Kroesbergen, E. H., Van Luit, J. E. H., & Naglieri, J. A. (2003). Mathematical learning differences and PASS cognitive processes. *Journal of Learning Disabilities, 36*, 574-582. doi:10.1177/00222194030360060801

Kroesbergen, E. H., Van Luit, J. E. H., Naglieri, J. A., Taddei, S., & Franchi, E. (2010). PASS processes and early mathematics skills in Dutch and Italian kindergarteners. *Journal of Psychoeducational Assessment, 28*, 585-593. doi:10.1177/0734282909356054

\* Kroesbergen, E. H., Van Luit, J. E. H., & Viersen, S. V. (2015). PASS theory and special educational needs: A European perspective. In T. C. Papadopoulos, R. K. Parrila, & J. R. Kirby (Eds.), *Cognition, intelligence, and achievement* (pp. 245-265). New York, NY: Elsevier.

\* Landeros-Thomas, B. R. (2017). *The relationship between American Indian student's cognitive processing and their reading skills*. Doctoral Dissertation, Grant Canyon University, Arizona, USA.

Leong, C. K., Cheng, S. C., & Das, J. P. (1985). Simultaneous–successive syntheses and planning in Chinese readers. *International Journal of Psychology, 20*, 19-31. doi: 10.1002/j.1464-066X.1985.tb00012.x

\* Liao, C.-H., Georgiou, G., & Parrila, R. (2008). Rapid naming speed and Chinese character recognition. *Reading and Writing, 21*, 231-253. doi:10.1007/s11145-007-9071-0

\* Liu, C., & Georgiou, G. K. (2015). Unpublished data.

Lubinski, D., Benbow, C. P., & Kell, H. J. (2014). Life paths and accomplishments of mathematically precocious males and females four decades later. *Psychological Science, 25*, 2217–2232. doi:10.1177/0956797614551371

Luria, A. R. (1966). *Human brain and psychological processes*. New York, NY: Harper and Row.

Luria, A. R. (1973). *The working brain*. New York: Basic Books.

\* Mahapatra, S. (2015). Reading difficulties in children: The role of language and cognitive processes. *Journal of Humanities and Social Science, 20*, 10-18. doi:10.9790/0837-20241018

\* Mahapatra, S., Das, J. P., Stack-Cutler, H., & Parrila, R. (2010). Remediating reading comprehension difficulties: A cognitive processing approach. *Reading Psychology, 31*, 428-453. doi:10.1080/02702710903054915

- \* Mahapatra, S., & Dash, U. N. (1999). Reading achievement in relation to PASS processes. In U.N. Dash, & U. Jain (Eds.), *Perspectives on Psychology and Social Development* (pp. 282-303). New Delhi, India: Concept Publishing Company.
- Mayes, S. D., Calhoun, S. L., Bixler, E. O., & Zimmerman, D. N. (2009). IQ and neuropsychological predictors of academic achievement. *Learning and Individual Differences, 19*, 238-241. doi:10.1016/j.lindif.2008.09.001
- Melby-Lervåg, M., Lyster, S.-A. H., & Hulme, C. (2012). Phonological skills and their role in learning to read: A meta-analytic review. *Psychological Bulletin, 138*, 322-352. doi:doi.org/10.1037/a0026744
- Naglieri, J. A. (2008). Traditional IQ of misconceptions and its relationship to minority representation in gifted programs. In J. L. VanTassel-Baska (Ed.), *The critical issues in equity and excellence in gifted education series. Alternative assessments with gifted and talented students* (p. 67–88). Prufrock Press.
- Naglieri, J. A., & Bornstein, B. T. (2003). Intelligence and achievement: Just how correlated are they? *Journal of Psychoeducational Assessment, 21*, 244–260. doi:10.1177/073428290302100302
- Naglieri, J. A., & Das, J. P. (1987). Construct and criterion related validity of planning, simultaneous and successive cognitive processing tasks. *Journal of Psychoeducational Assessment, 4*, 353–363. doi:10.1177/073428298700500405



\* Naglieri, J. A., & Das, J. P. (1997). *Das-Naglieri Cognitive Assessment System*. Itasca, IL: Riverside.

\* Naglieri, J. A., Das, J. P., & Goldstein, S. (2014). *Cognitive Assessment System (2<sup>nd</sup> ed.)*. Austin, TX: Pro-Ed.

\* Naglieri, J. A., DeLauder, B. Y., Goldstein, S., & Schwebech, A. (2006). WISC III and CAS: Which correlates higher with achievement for a clinical sample. *School Psychology Quarterly*, 21, 62-76. doi: 10.1521/scpq.2006.21.1.62

Naglieri, J. A., & Gottling, S. H. (1997). Mathematics instruction and PASS cognitive processes: An intervention study. *Journal of Learning Disabilities*, 30, 513-520. doi: 10.1177/002221949703000507

Naglieri, J. A., & Johnson, D. (2000). Effectiveness of a cognitive strategy intervention in improving arithmetic computation based on the PASS theory. *Journal of Learning Disabilities*, 33, 591-597. doi: 10.1177/002221940003300607

Naglieri, J. A., & Otero, T. M. (2017). *Essentials of CAS 2 Assessment*. Hoboken, NJ: Wiley.

Naglieri, J. A., & Otero, T. M. (2018). Redefining intelligence with the Planning, Attention, Simultaneous and Successive theory of neurocognitive processes. In D. P. Flanagan & E. M., McDonough (Eds.), *Contemporary intellectual assessment: Theories, tests, and issues* (4<sup>th</sup> ed) (pp. 195-218). New York, NY: Guilford Press.

Naglieri, J. A., Otero, T., DeLauder, B., & Matto, H. (2007). Bilingual Hispanic children's performance on the English and Spanish versions of the Cognitive Assessment System.

*School Psychology Quarterly*, 22, 432-448. doi: 10.1037/1045-3830.22.3.432

Naglieri, J. A., & Rojahn, J. (2004). Construct validity of the PASS theory and CAS: Correlations with achievement. *Journal of Educational Psychology*, 96, 174-181. doi: 10.1037/0022-

0663.96.1.174

\* Naglieri, J. A., Rojahn, J., & Matto, H. C. (2007). Hispanic and non-Hispanic children's performance on PASS cognitive processes and achievement. *Intelligence*, 35, 568-579.

doi:10.1016/j.intell.2006.11.001

\* Naglieri, J. A., Rojahn, J., Matto, H. C., & Aquilino, S. A. (2005). Black-White differences in cognitive processing: A study of the Planning, Attention, Simultaneous, and Successive theory of intelligence. *Journal of Psychoeducational Assessment*, 23, 146-160.

doi:10.1177/073428290502300204

Naglieri, J. A., Taddei, S., & Williams, K. M. (2013). Multigroup confirmatory factor analysis of U.S. and Italian children's performance on the PASS theory of intelligence as measured by the Cognitive Assessment System. *Psychological Assessment*, 25, 157-166.

doi:10.1037/a0029828

\* Papadopoulos, T. C. (2001). Phonological and cognitive correlates of word-reading acquisition under two different instructional approaches in Greek. *European Journal of Psychology of Education, 14*, 549-568. doi:10.1007/BF03173197

Papadopoulos, T. C., Charalambous, A., Kanari, A., & Loizou, M. (2004). Kindergarten intervention for dyslexia: The PREP remediation in Greek. *European Journal of Psychology of Education, 19*, 79–105.

\* Papadopoulos, T. C., Georgiou, G. K., & Parrila, R. (2012). Low-level deficits in beat perception: Neither necessary nor sufficient for explaining developmental dyslexia in a consistent orthography. *Research in Developmental Disabilities, 33*, 1841-1856. doi: 10.1016/j.ridd.2012.04.009

Papadopoulos, T. C., Kendeou, P., & Shiakalli, M. (2014). Reading comprehension tests and poor comprehenders: Do different processing demands mean different profiles? *Topics in Cognitive Psychology, 114*, 725-753.

Papadopoulos, T. C., Parrila, R. K., & Kirby J. R. (Eds.). (2015). *Cognition, intelligence, and achievement: A tribute to J. P. Das*. San Diego, CA: Elsevier

\* Papadopoulos, T. C., Spanoudis, G., & Georgiou, G. K. (2016). How is RAN is related to reading fluency? A comprehensive examination of the prominent theoretical accounts. *Frontiers in Psychology, 7*, 1-15. doi:10.3389/fpsyg.2016.01217

- \* Parrila, R., Kirby, J. R., & McQuarrie, L. (2004). Articulation rate, naming speed, verbal short memory and phonological awareness: Longitudinal predictors of early reading development. *Scientific Studies of Reading, 8*, 3-26. doi:10.1207/s1532799xssr0801\_2
- Peng, P., Namkung, J., Barnes, M., & Sun, C. (2016). A meta-analysis of mathematics and working memory: Moderating effects of working memory domain, type of mathematics skill, and sample characteristics. *Journal of Educational Psychology, 108*, 455-473. doi:10.1037/edu0000079
- Peng, P., Wang, T., Wang, C. C., & Lin, X. (2019). A meta-analysis on the relation between fluid intelligence and reading/mathematics: Effects of tasks, age, and social economic status. *Psychological Bulletin, 145*, 189-236. doi:10.1037/bul0000182
- Rana, R., & Mahmood, N. (2010). The relationship between test anxiety and academic achievement. *Bulletin of Education and Research, 32*, 63-74.
- \* Rosário, A. C. M. B. A. (2007). *Assessment of cognitive processes in reading: Exploratory study from PASS theory with students from Grades 2, 4 and 6*. Master's Thesis, Universidade de Évora, Portugal.
- \*Rosário, A. C. M. B. A. (2014). *Cognitive assessment system: Contributions to their validation with elementary school students from the municipality of Évora*. Doctoral Thesis, Universidade de Évora, Portugal.

Roth, B., Becker, N., Romeyke, S., Schäfer, Domic, F., & Spinath, F. M. (2015). Intelligence and school grades: A meta-analysis. *Intelligence*, *53*, 118-137.

doi:10.1016/j.intell.2015.09.002

Ruan, Y., Georgiou, G. K., Song, S., Li, Y., & Shu, H. (2018). Does writing system influence the association between phonological awareness, morphological awareness and reading? A meta-analysis. *Journal of Educational Psychology*, *110*, 180-202. doi:10.1037/edu0000216

\* Samantaray, S. (2011). *Identifying concurrent predictors for reading difficulties among bilingual children*. Master's Thesis, University of London, England.

Schneider, M., Beeres, K., Coban, L., Merz, S., Schmidt, S., Stricker, J., & De Smedt, B. (2017).

Associations of non-symbolic and symbolic numerical magnitude processing with mathematical competence: A meta-analysis. *Developmental Science*, *20*, e12372.

doi:10.1111/desc.12372

Soares, D. L. Lemos, G. C., Primi, R., & Almeida, L. S. (2015). The relationship between intelligence and academic achievement throughout middle school: The role of students' prior academic performance. *Learning and Individual Differences*, *41*, 73-78. doi:10.

1016/j.kindif.2015.02.00

Sternberg, R. (1988). Applying cognitive theory to the testing and teaching of intelligence.

*Applied Cognitive Psychology*, *2*, 231-255. doi:10.1002/acp.2350020402

Swanson, H. L., Trainin, G., Necochea, D. M., & Hammill, D. D. (2003). Rapid naming, phonological awareness, and reading: A meta-analysis of the correlation evidence. *Review of Educational Research, 73*, 407-440. doi:10.3102/00346543073004407

\* Taha, M. M. (2016). Structural model of the relationships among cognitive processes, visual motor integration, and academic achievement in students with mild intellectual disability (MID). *Insights into Learning Disabilities, 13*, 135-150.

Viechtbauer, W. (2010). Conducting Meta-Analyses in R with the metafor Package. *Journal of Statistical Software, 36*, 1-48. doi: 10.18637/jss.v036.i03

\* Wang, X., Georgiou, G. K., & Das, J. P. (2012). Examining the effects of PASS cognitive processes on Chinese reading accuracy and fluency. *Learning and Individual Differences, 22*, 139–143. doi:10.1016/j.lindif.2011.11.006

Wang, X., Georgiou, G. K., Das, J. P., & Li, Q. (2012). Cognitive processing skills and developmental dyslexia in Chinese. *Journal of Learning Disabilities, 45*, 526-537. doi: 10.1177/0022219411402693

\* Wei, W., Deng, C., & Georgiou, G. K. (2017, February). *The role of PASS cognitive processes in Chinese word reading and reading comprehension: A 4-year longitudinal study*. Paper presented at the fifth annual meeting of the Association for Reading and Writing in Asia, Hong Kong, China.

- \* Wei, W., Guo, L., Georgiou, G. K., Tavouktsoglou, A., & Deng, C. (2018). Different subcomponents of executive functioning predict different growth parameters in mathematics: Evidence from a 4-year longitudinal study with Chinese children. *Frontiers in Psychology, 9*, 1-10. doi:10.3389/fpsyg.2018.01037
- Zaboski, B. A., Kranzler, J. H., & Gage, N. A. (2018). Meta-analysis of the relationship between academic achievement and broad abilities of the Cattell-Horn-Carroll theory. *Journal of School Psychology, 71*, 42–56. doi:10.1016/j.jsp. 2018.10.001
- \* Zhu, M., Cai, D., & Leung, A. W. S. (2017). Number line estimation predicts mathematical skills: Difference in grades 2 and 4. *Frontiers in Psychology, 8*, 1-8. doi: 10.3389/fpsyg.2017.01576

Appendix A (continued)

## Appendix A

## Studies on PASS and Reading outcomes

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
Abougoush, 2014	English	Good	G1 to G6			52	Pl				0.58
	English	Good	G1 to G6			52	Att				0.23
	English	Good	G1 to G6			52	Sim				0.43
	English	Good	G1 to G6			52	Suc				0.45
Cui et al., 2015	Non-Euro	Unselct	Kind		5.11	160	Sim	0.19			
	Non-Euro	Unselct	Kind		5.11	160	Suc	0.10			
Das & Georgiou, 2016	English	Unselct	Adults		22.26		Pl		0.25		
Das et al., 2008	English	Unselct	G1 to G6		9.97	71	Pl	0.42			
	English	Unselct	G1 to G6		9.97	71	Att	0.08			
	English	Unselct	G1 to G6		9.97	71	Sim	0.25			
	English	Unselct	G1 to G6		9.97	71	Suc	0.26			
Das et al., 2007	English	Unselct	G1 to G6		9.5	84	Pl	0.26			
	English	Unselct	G1 to G6		9.5	84	Att	0.26			
	English	Unselct	G1 to G6		9.5	84	Sim	0.28			
	English	Unselct	G1 to G6		9.5	84	Suc	0.45			
Dash & Das, 1998	Non-Euro	Unselct	G1 to G6	Grade 1 and 3		100	Pl	0.57			
	Non-Euro	Unselct	G1 to G6	Grade 1 and 3		100	Sim	0.50			
	Non-Euro	Unselct	G1 to G6 and G7 to G12	Grade 5 and 7		100	Pl	0.53			
	Non-Euro	Unselct	G1 to G6 and G7 to G12	Grade 5 and 7		100	Sim	0.45			
	Non-Euro	Unselct	G7 to G12	Grade 9 and 11		100	Pl	0.46			
	Non-Euro	Unselct	G7 to G12	Grade 9 and 11		100	Sim	0.57			
Dash & Dash, 1999	Non-Euro	Poor	G1 to G6	Grade 3 less skilled		20	Pl	0.15		0.13	

(Appendices continue)



## Appendix A (continued)

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
	Non-Euro	Good	G1 to G6	Grade 3 skilled		20	Pl	0.32		0.15	
	Non-Euro	Poor	G1 to G6	Grade 5 less skilled		20	Pl	0.21		0.34	
	Non-Euro	Good	G1 to G6	Grade 5 skilled		20	Pl	0.18		0.29	
Deacon & Kirby, 2004	English	Unselect	G1 to G6		7.38	103	Sim	0.44		0.50	
Dunn et al., in press	English	Good	G1 to G6		10.62	142	Pl				0.40
	English	Good	G1 to G6		10.62	142	Att				0.33
	English	Good	G1 to G6		10.62	142	Sim				0.32
	English	Good	G1 to G6		10.62	142	Suc				0.41
Georgiou, 2008	Other Euro	Unselect	G1 to G6		9.77	208	Att		0.26	0.32	
	Other Euro	Unselect	G1 to G6		9.77	208	Suc		0.21	0.35	
Georgiou, 2010	English	Unselect	G1 to G6		9.47	84	Pl	0.34			
	English	Unselect	G1 to G6		9.47	84	Att	0.32			
	English	Unselect	G1 to G6		9.47	84	Sim	0.31			
	English	Unselect	G1 to G6		9.47	84	Suc	0.42			
Georgiou & Das, 2014	English	Unselect	Adults		22.07	128	Pl		0.22	0.25	
	English	Unselect	Adults		22.07	128	Att		0.27	0.22	
	English	Unselect	Adults		22.07	128	Sim		0.27	0.50	
	English	Unselect	Adults		22.07	128	Suc		0.37	0.40	
Georgiou & Das, 2016	English	Unselect	Adults		21.83	178	Pl			0.36	
Georgiou & Das, 2018	English	Unselect	Adults		21.82	90	Pl		0.22	0.39	
	English	Unselect	Adults		21.82	90	Att		0.42	0.22	
Georgiou et al., 2015	Other Euro	Unselect	Kind	Kinder	5.42	83	Pl	0.67			
	Other Euro	Unselect	Kind	Kinder	5.42	83	Att	0.50			
	Other Euro	Unselect	Kind	Kinder	5.42	83	Sim	0.46			
	Other Euro	Unselect	Kind	Kinder	5.42	83	Suc	0.53			
	Other Euro	Unselect	G1 to G6	Grade 1	6.83	83	Pl		0.50		
	Other Euro	Unselect	G1 to G6	Grade 1	6.83	83	Att		0.40		

(Appendices continue)

## Appendix A (continued)

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
Georgiou et al., 2013	Other Euro	Unselct	G1 to G6	Grade 1	6.83	83	Sim		0.33		
	Other Euro	Unselct	G1 to G6	Grade 1	6.83	83	Suc		0.45		
	Other Euro	Unselct	Kind	Kinder	5.38	72	Att	0.14			
	Other Euro	Unselct	Kind	Kinder	5.38	72	Sim	0.49			
	Other Euro	Unselct	Kind	Kinder	5.38	72	Suc	0.54			
	Other Euro	Unselct	G1 to G6	Grade 1	6.92	72	Att		0.08		
	Other Euro	Unselct	G1 to G6	Grade 1	6.92	72	Sim		0.32		
Georgiou et al., 2019	Other Euro	Unselct	G1 to G6	Grade 1	6.92	72	Suc		0.43		
	English	Unselct	G1 to G6	English	6.41	120	Pl	0.47		0.45	
	English	Unselct	G1 to G6	English	6.41	120	Att	0.27		0.38	
	English	Unselct	G1 to G6	English	6.41	120	Sim	0.19		0.20	
	Non-Euro	Unselct	G1 to G6	Chinese	7.15	181	Pl	0.21		0.14	
	Non-Euro	Unselct	G1 to G6	Chinese	7.15	181	Att	0.20		0.34	
	Non-Euro	Unselct	G1 to G6	Chinese	7.15	181	Sim	0.15		0.20	
Janzen, 2000	English	Unselct	G1 to G6		9	53	Pl	0.30		0.15	
	English	Unselct	G1 to G6		9	53	Att	0.32		0.44	
	English	Unselct	G1 to G6		9	53	Sim	0.13		-0.01	
	English	Unselct	G1 to G6		9	53	Suc	0.30		0.29	
Janzen et al., 2013	English	Unselct	G1 to G6	Alberta	9.5	84	Pl	0.12			
	English	Unselct	G1 to G6	Alberta	9.5	84	Att	0.19			
	English	Unselct	G1 to G6	Alberta	9.5	84	Sim	0.14			
	English	Unselct	G1 to G6	Alberta	9.5	84	Suc	0.38			
	English	Unselct	G1 to G6	Saskatchewan	9.4	49	Pl	0.39			
	English	Unselct	G1 to G6	Saskatchewan	9.4	49	Att	0.23			
	English	Unselct	G1 to G6	Saskatchewan	9.4	49	Sim	0.15			
	English	Unselct	G1 to G6	Saskatchewan	9.4	49	Suc	0.45			
Joseph et al., 2003	English	Poor	G1 to G6		8.4	62	Pl	0.47			
	English	Poor	G1 to G6		8.4	62	Att	0.37			
	English	Poor	G1 to G6		8.4	62	Sim	0.51			

(Appendices continue)

## Appendix A (continued)

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
Keat & Ismail, 2010	English	Poor	G1 to G6		8.4	62	Suc	0.41			
	ELL	Poor	G1 to G6			50	Pl	0.35		-0.14	0.27
	ELL	Poor	G1 to G6			50	Att	0.25		-0.30	0.16
	ELL	Poor	G1 to G6			50	Sim	0.28		0.12	0.32
Keith et al., 2001	ELL	Poor	G1 to G6			50	Suc	0.29		-0.10	0.26
	English	Unselct	G1 to G6		9.81	155	Pl			0.64	
	English	Unselct	G1 to G6		9.81	155	Att			0.50	
	English	Unselct	G1 to G6		9.81	155	Sim			0.75	
Kendeou et al., 2012	English	Unselct	G1 to G6		9.81	155	Suc			0.53	
	Other Euro	Unselct	G1 to G6	Grade 1	6.6	286	Sim		0.14		
	Other Euro	Unselct	G1 to G6	Grade 1	6.6	286	Suc		0.28		
	Other Euro	Unselct	G1 to G6	Grade 2	7.7	286	Sim			0.30	
Kendeou et al., 2015	Other Euro	Unselct	G1 to G6	Grade 2	7.7	286	Suc			0.37	
	Other Euro	Unselct				462	Pl			0.52	
	Other Euro	Unselct				462	Att			0.32	
	Other Euro	Unselct				462	Sim			0.31	
Kroesbergen et al., 2015	Other Euro	Unselct				462	Suc			0.36	
	Other Euro					70	Pl				0.09
	Other Euro					70	Att				0.06
	Other Euro					70	Sim				0.02
Landeros-Thomas, 2017	Other Euro					70	Suc				0.27
	English	Unselct	G1 to G6		9.8	162	Pl	0.09	0.23	0.30	0.23
	English	Unselct	G1 to G6		9.8	162	Att	0.12	0.20	0.22	0.28
	English	Unselct	G1 to G6		9.8	162	Sim	0.22	0.27	0.31	0.25
Liao et al., 2008	English	Unselct	G1 to G6		9.8	162	Suc	0.54	0.55	0.58	0.61
	Non-Euro	Unselct	G1 to G6	Grade 2	8	63	Sim	0.36	0.26		
	Non-Euro	Unselct	G1 to G6	Grade 2	8	63	Suc	0.19	0.25		
	Non-Euro	Unselct	G1 to G6	Grade 4	10.01	54	Sim	0.29	0.14		
	Non-Euro	Unselct	G1 to G6	Grade 4	10.01	54	Suc	0.11	0.16		

(Appendices continue)

## Appendix A (continued)

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
Liu & Georgiou, 2015	Non-Euro	Unselct	Kind		4.91	140	Pl	0.24			
	Non-Euro	Unselct	Kind		4.91	140	Att	0.01			
	Non-Euro	Unselct	Kind		4.91	140	Sim	0.21			
	Non-Euro	Unselct	Kind		4.91	140	Suc	0.22			
Mahapatra, 2015	ELL	Poor and good	G1 to G6			30	Pl	0.21		0.14	
	ELL	Poor and good	G1 to G6			30	Att	0.24		0.23	
	ELL	Poor and good	G1 to G6			30	Sim	0.56		0.70	
	ELL	Poor and good	G1 to G6			30	Suc	0.01		0.02	
Mahapatra et al., 2010	ELL	Poor and good	G1 to G6		9.4	28	Sim	0.62		0.75	
Mahapatra & Dash, 1999	Non-Euro	Poor	G1 to G6	Less skilled	10	50	Pl	0.11		0.12	
	Non-Euro	Poor	G1 to G6	Less skilled	10	50	Sim	0.09		0.07	
	Non-Euro	Poor	G1 to G6	Less skilled	10	50	Suc	0.46		0.27	
	Non-Euro	Good	G1 to G6	Skilled	10	50	Pl	0.08		0.11	
	Non-Euro	Good	G1 to G6	Skilled	10	50	Sim	-0.06		0.06	
	Non-Euro	Good	G1 to G6	Skilled	10	50	Suc	0.10		0.06	
Naglieri & Das, 1997	English	Unselct		5- 7 yrs		630	Pl	0.43		0.37	0.41
	English	Unselct		5- 7 yrs		630	Att	0.36		0.33	0.41
	English	Unselct		5- 7 yrs		630	Sim	0.41		0.36	0.48
	English	Unselct		5- 7 yrs		630	Suc	0.37		0.36	0.36
	English	Unselct		8- 10 yrs		454	Pl	0.37		0.45	0.44
	English	Unselct		8- 10 yrs		454	Att	0.39		0.40	0.42
	English	Unselct		8- 10 yrs		454	Sim	0.59		0.65	0.67
	English	Unselct		8- 10 yrs		454	Suc	0.54		0.55	0.57
	English	Unselct		11- 13 yrs		228	Pl	0.57		0.53	0.63
	English	Unselct		11- 13 yrs		228	Att	0.48		0.48	0.55
	English	Unselct		11- 13 yrs		228	Sim	0.64		0.64	0.64

(Appendices continue)

## Appendix A (continued)

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
	English	Unselct		11- 13 yrs		228	Suc	0.57		0.64	0.68
	English	Unselct		14- 17 yrs		288	Pl	0.54		0.52	0.46
	English	Unselct		14- 17 yrs		288	Att	0.44		0.40	0.49
	English	Unselct		14- 17 yrs		288	Sim	0.52		0.63	0.64
	English	Unselct		14- 17 yrs		288	Suc	0.57		0.62	0.59
Naglieri & Das, 1997	English	Unselct		Gifted	13.4	53	Pl			0.09	
	English	Unselct		Gifted	13.4	53	Att			0.17	
	English	Unselct		Gifted	13.4	53	Sim			0.36	
	English	Unselct		Gifted	13.4	53	Suc			0.34	
Naglieri et al., 2014	English			Sample 1		36	Pl				0.51
	English			Sample 1		36	Att				0.51
	English			Sample 1		36	Sim				0.52
	English			Sample 1		36	Suc				0.69
Naglieri et al., 2014	English	Unselct		Sample 2		110	Pl		0.50		
	English	Unselct		Sample 2		110	Att		0.53		
	English	Unselct		Sample 2		110	Sim		0.49		
	English	Unselct		Sample 2		110	Suc		0.44		
Naglieri et al., 2014	English	Unselct		Sample 3		51	Pl		0.34		
	English	Unselct		Sample 3		51	Att		0.35		
	English	Unselct		Sample 3		51	Sim		0.46		
	English	Unselct		Sample 3		51	Suc		0.54		
Naglieri et al., 2006	English					119	Pl				0.48
	English					119	Att				0.36
	English					119	Sim				0.51
	English					119	Suc				0.50
Naglieri et al., 2007	English	Unselct		Hispanics	9.66	159	FS	0.51		0.43	0.51
	English	Unselct		Non-Hispanics	9.85	1274	FS	0.63		0.63	0.65
Naglieri et al., 2005	English	Unselct		Blacks		298	FS	0.70		0.68	0.71

(Appendices continue)

## Appendix A (continued)

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
Papadopoulos, 2001	English	Unselet		Whites		1691	FS	0.60		0.60	0.63
	Other Euro	Unselet	G1 to G6	Greek	6.43	50	Pl	0.28			
	Other Euro	Unselet	G1 to G6	Greek	6.43	50	Att	0.25			
	Other Euro	Unselet	G1 to G6	Greek	6.43	50	Sim	0.31			
	Other Euro	Unselet	G1 to G6	Greek	6.43	50	Suc	0.19			
	Other Euro	Unselet	G1 to G6	Cypriot	6.3	50	Pl	0.16			
	Other Euro	Unselet	G1 to G6	Cypriot	6.3	50	Att	0.19			
	Other Euro	Unselet	G1 to G6	Cypriot	6.3	50	Sim	0.31			
Papadopoulos et al., 2012	Other Euro	Unselet	G1 to G6		9.77	202	Att		0.26		
	Other Euro	Unselet	G1 to G6		9.77	202	Suc		0.21		
Papadopoulos et al., 2016	Other Euro	Unselet	G1 to G6		6.6	286	Pl		0.24		
	Other Euro	Unselet	G1 to G6		6.6	286	Att		0.13		
	Other Euro	Unselet	G1 to G6		6.6	286	Suc		0.25		
Parrila et al., 2004	Other Euro	Unselet	Kind and G1 to G6			117	Suc	0.25		0.27	
Rosário, 2007	Other Euro	Unselet	G1 to G6		9.11	91	Pl	0.61	0.57	0.69	
	Other Euro	Unselet	G1 to G6		9.11	91	Att	0.62	0.56	0.72	
	Other Euro	Unselet	G1 to G6		9.11	91	Sim	0.53	0.45	0.60	
	Other Euro	Unselet	G1 to G6		9.11	91	Suc	0.36	0.37	0.50	
Samantaray, 2011	ELL	Unselet	G1 to G6		9.3	56	Pl	-0.10	0.18	-0.01	
	ELL	Unselet	G1 to G6		9.3	56	Att	0.43	0.32	0.31	
	ELL	Unselet	G1 to G6		9.3	56	Sim	0.30	0.22	0.33	
	ELL	Unselet	G1 to G6		9.3	56	Suc	0.42	0.40	0.37	
	Non-Euro	Unselet	G1 to G6		9.3	56	Pl	0.11		0.13	
	Non-Euro	Unselet	G1 to G6		9.3	56	Att	0.19		0.15	
	Non-Euro	Unselet	G1 to G6		9.3	56	Sim	0.24		0.25	
	Non-Euro	Unselet	G1 to G6		9.3	56	Suc	0.36		0.38	
Wang et al., 2012	Non-Euro	Unselet	G1 to G6		10	140	Pl	0.16	0.18		

(Appendices continue)

## Appendix A (continued)

Study	Language	Reading ability level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Reading accuracy	Reading fluency	Reading comprehension	Broad Reading
Wei et al., 2017	Non-Euro	Unselct	G1 to G6		10	140	Att	0.38	0.39		
	Non-Euro	Unselct	G1 to G6		10	140	Sim	0.48	0.33		
	Non-Euro	Unselct	G1 to G6		10	140	Suc	0.48	0.42		
	Non-Euro	Unselct	G1 to G6		7.17	180	Pl	0.31		0.16	
	Non-Euro	Unselct	G1 to G6		7.17	180	Att	0.12		0.15	
	Non-Euro	Unselct	G1 to G6		7.17	180	Sim	0.17		0.28	
	Non-Euro	Unselct	G1 to G6		7.17	180	Suc	0.26		0.26	

*Note.* Non-Euro = Non-European language; Other Euro= Other European language; ELL = English language learners; Unselct = Unselected; Kind = Kindergarten; Pl = Planning;

Att = Attention; Sim = Simultaneous; Suc = Successive; FS = Full Scale.

(Appendices continue)

Appendix B (continued)

## Appendix B

## Studies on PASS and Math Outcomes

Study	Language	Math performance level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Math accuracy	Math fluency	Problem-solving	Broad Math
Abougoush, 2014	English	Good	G1 to G6			52	PI				0.37
	English	Good	G1 to G6			52	Att				0.13
	English	Good	G1 to G6			52	Sim				0.48
	English	Good	G1 to G6			52	Suc				0.24
Cai et al., 2016	Non-Euro	Unselct	G1 to G6		7.89	80	PI		0.41	0.43	
	Non-Euro	Unselct	G1 to G6		7.89	80	Sim		0.02	0.32	
Cai et al., 2010	Non-Euro	Unselct	Kind	Kinder	5.8	105	PI	0.40			
	Non-Euro	Unselct	Kind	Kinder	5.8	105	Att	0.42			
	Non-Euro	Unselct	Kind	Kinder	5.8	105	Sim	0.46			
	Non-Euro	Unselct	Kind	Kinder	5.8	105	Suc	0.45			
	Non-Euro	Unselct	G1 to G6 and G7 to G12	Grade 3,4,5,6,7,8 group	11.6	250	PI				0.43
	Non-Euro	Unselct	G1 to G6 and G7 to G12	Grade 3,4,5,6,7,8 group	11.6	250	Att				0.41
	Non-Euro	Unselct	G1 to G6 and G7 to G12	Grade 3,4,5,6,7,8 group	11.6	250	Sim				0.51
	Non-Euro	Unselct	G1 to G6 and G7 to G12	Grade 3,4,5,6,7,8 group	11.6	250	Suc				0.39
Cai et al., 2013	Non-Euro	Good and MLD	G1 to G6 and G7 to G12		11.97	111	PI				0.44
	Non-Euro	Good and MLD	G1 to G6 and G7 to G12		11.97	111	Att				0.38
	Non-Euro	Good and MLD	G1 to G6 and G7 to G12		11.97	111	Sim				0.46

(Appendices continue)



## Appendix B (continued)

Study	Language	Math performance level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Math accuracy	Math fluency	Problem-solving	Broad Math
Cai et al., 2018	Non-Euro	Good and MLD	G1 to G6 and G7 to G12		11.97	111	Suc				0.36
	Non-Euro	Unselct	Kind	Kinder	5.54	100	Att	0.37		0.33	
	Non-Euro	Unselct	Kind	Kinder	5.54	100	Sim	0.63		0.68	
	Non-Euro	Unselct	Kind	Kinder	5.54	100	Suc	0.22		0.22	
	Non-Euro	Unselct	G1 to G6	Grade 2	7.68	107	Att	0.12	0.12	0.27	
	Non-Euro	Unselct	G1 to G6	Grade 2	7.68	107	Sim	0.10	0.01	0.26	
	Non-Euro	Unselct	G1 to G6	Grade 2	7.68	107	Suc	-0.08	0.21	0.29	
	Non-Euro	Unselct	G1 to G6	Grade 4	9.65	104	Att	0.27	0.24	0.21	
	Non-Euro	Unselct	G1 to G6	Grade 4	9.65	104	Sim	0.26	0.05	0.38	
Cui et al., 2015	Non-Euro	Unselct	G1 to G6	Grade 4	9.65	104	Suc	0.15	0.09	0.09	
	Non-Euro	Unselct	Kind		5.11	160	Sim		0.26		
	Non-Euro	Unselct	Kind		5.11	160	Suc		0.24		
Dunn et al., 2019	English	Good	G1 to G6		10.62	142	Pl				0.40
	English	Good	G1 to G6		10.62	142	Att				0.30
	English	Good	G1 to G6		10.62	142	Sim				0.42
	English	Good	G1 to G6		10.62	142	Suc				0.22
Georgiou et al., 2015	Other Euro	Unselct	G1 to G6	Grade 1	6.83	83	Pl		0.46		
	Other Euro	Unselct	G1 to G6	Grade 1	6.83	83	Att		0.36		
	Other Euro	Unselct	G1 to G6	Grade 1	6.83	83	Sim		0.34		
	Other Euro	Unselct	G1 to G6	Grade 1	6.83	83	Suc		0.36		
Georgiou et al., 2013	Other Euro	Unselct	G1 to G6	Grade 1	6.92	72	Att		0		
	Other Euro	Unselct	G1 to G6	Grade 1	6.92	72	Sim		0.40		
	Other Euro	Unselct	G1 to G6	Grade 1	6.92	72	Suc		0.32		
Georgiou et al., 2019	English	Unselct	G1 to G6	English	6.41	120	Pl	0.33		0.45	
	English	Unselct	G1 to G6	English	6.41	120	Att	0.26		0.31	
	English	Unselct	G1 to G6	English	6.41	120	Sim	0.23		0.30	
	Non-Euro	Unselct	G1 to G6	Chinese	7.15	181	Pl	0.06		0.22	
	Non-Euro	Unselct	G1 to G6	Chinese	7.15	181	Att	0.05		0.25	

(Appendices continue)

## Appendix B (continued)

Study	Language	Math performance level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Math accuracy	Math fluency	Problem-solving	Broad Math	
Iglesias-Sarmiento et al., 2019	Non-Euro	Unselct	G1 to G6	Chinese	7.15	181	Sim	0.12		0.38		
	Other Euro	MLD, good and poor	G1 to G6		10.6	165	Pl	0.29				
	Other Euro	MLD, good and poor	G1 to G6		10.6	165	Att	0.22				
	Other Euro	MLD, good and poor	G1 to G6		10.6	165	Sim	0.57				
Iglesias-Sarmiento & Deaño, 2011	Other Euro	MLD, good and poor	G1 to G6	Grade 4		38	Pl				0.29	
	Other Euro	MLD, good and poor	G1 to G6	Grade 4		38	Att				0.55	
	Other Euro	MLD, good and poor	G1 to G6	Grade 4		38	Sim				0.78	
	Other Euro	MLD, good and poor	G1 to G6	Grade 4		38	Suc				0.45	
	Other Euro	MLD, good and poor	G1 to G6	Grade 5		38	Pl				0.35	
	Other Euro	MLD, good and poor	G1 to G6	Grade 5		38	Att				0.29	
	Other Euro	MLD, good and poor	G1 to G6	Grade 5		38	Sim				0.49	
	Other Euro	MLD, good and poor	G1 to G6	Grade 5		38	Suc				0.49	
	Other Euro	MLD, good and poor	G1 to G6	Grade 6		38	Pl				0.01	
	Other Euro	MLD, good and poor	G1 to G6	Grade 6		38	Att				-0.21	
	Other Euro	MLD, good and poor	G1 to G6	Grade 6		38	Sim				0.57	
	Other Euro	MLD, good and poor	G1 to G6	Grade 6		38	Suc				0.34	
	Iglesias-Sarmiento & Deaño, 2016	Other Euro	MLD, good and poor	G1 to G6			165	Pl	0.18			

(Appendices continue)

## Appendix B (continued)

Study	Language	Math performance level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Math accuracy	Math fluency	Problem-solving	Broad Math
	Other Euro	MLD, good and poor	G1 to G6			165	Att	0.23			
	Other Euro	MLD, good and poor	G1 to G6			165	Sim	0.42			
	Other Euro	MLD, good and poor	G1 to G6			165	Suc	0.27			
Iglesias-Sarmiento et al., 2017	Other Euro	ADHD	G1 to G6	ADHD	10.5	30	PI			0.71	
	Other Euro	ADHD	G1 to G6	ADHD	10.5	30	Att			0.69	
	Other Euro	ADHD	G1 to G6	ADHD	10.5	30	Sim			0.13	
	Other Euro	ADHD	G1 to G6	ADHD	10.5	30	Suc			0.19	
	Other Euro	MLD	G1 to G6	MLD	10.6	30	PI			0.32	
	Other Euro	MLD	G1 to G6	MLD	10.6	30	Att			0.20	
	Other Euro	MLD	G1 to G6	MLD	10.6	30	Sim			0.58	
	Other Euro	MLD	G1 to G6	MLD	10.6	30	Suc			0.51	
	Other Euro	Good	G1 to G6	TA	10.9	30	PI			0.41	
	Other Euro	Good	G1 to G6	TA	10.9	30	Att			0.28	
	Other Euro	Good	G1 to G6	TA	10.9	30	Sim			0.37	
	Other Euro	Good	G1 to G6	TA	10.9	30	Suc			0.26	
Keith et al., 2001	English	Unselct	G1 to G6		9.81	155	PI			0.68	
	English	Unselct	G1 to G6		9.81	155	Att			0.44	
	English	Unselct	G1 to G6		9.81	155	Sim			0.77	
	English	Unselct	G1 to G6		9.81	155	Suc			0.48	
Kroesbergen et al., 2015	Other Euro					70	PI				0.23
	Other Euro					70	Att				0.26
	Other Euro					70	Sim				0.20
	Other Euro					70	Suc				0.06
Kroesbergen, et al., 2015	Other Euro			Dutch		38	PI				0.64
	Other Euro			Dutch		38	Att				0.42
	Other Euro			Dutch		38	Sim				0.54

(Appendices continue)

## Appendix B (continued)

Study	Language	Math performance level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Math accuracy	Math fluency	Problem-solving	Broad Math
				Dutch		38	Suc				0.33
				Non-native Dutch		22	Pl				0.23
				Non-native Dutch		22	Att				0.14
				Non-native Dutch		22	Sim				0.43
				Non-native Dutch		22	Suc				0.21
Naglieri & Das, 1997	English	Unselct		5 -7 yrs		630	Pl	0.47		0.44	0.53
	English	Unselct		5 -7 yrs		630	Att	0.39		0.47	0.47
	English	Unselct		5 -7 yrs		630	Sim	0.33		0.60	0.53
	English	Unselct		5 -7 yrs		630	Suc	0.28		0.48	0.44
	English	Unselct		8 -10 yrs		454	Pl	0.53		0.51	0.57
	English	Unselct		8 -10 yrs		454	Att	0.40		0.40	0.44
	English	Unselct		8 -10 yrs		454	Sim	0.50		0.62	0.61
	English	Unselct		8 -10 yrs		454	Suc	0.42		0.49	0.49
	English	Unselct		11 -13 yrs		228	Pl	0.58		0.61	0.60
	English	Unselct		11 -13 yrs		228	Att	0.46		0.49	0.48
	English	Unselct		11 -13 yrs		228	Sim	0.58		0.66	0.65
	English	Unselct		11 -13 yrs		228	Suc	0.52		0.57	0.58
	English	Unselct		14 -17 yrs		288	Pl	0.59		0.53	0.59
	English	Unselct		14 -17 yrs		288	Att	0.46		0.48	0.48
	English	Unselct		14 -17 yrs		288	Sim	0.61		0.67	0.68
	English	Unselct		14 -17 yrs		288	Suc	0.52		0.53	0.58
Naglieri & Das, 1997	English	Good			13.4	53	Pl	0.35			
	English	Good			13.4	53	Att	0.28			
	English	Good			13.4	53	Sim	0.43			
	English	Good			13.4	53	Suc	0.13			
Naglieri et al., 2014	English			Sample 1		36	Pl				0.63

(Appendices continue)

## Appendix B (continued)

Study	Language	Math performance level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Math accuracy	Math fluency	Problem-solving	Broad Math
	English			Sample 1		36	Att				0.40
	English			Sample 1		36	Sim				0.61
	English			Sample 1		36	Suc				0.66
Naglieri et al., 2014	English			Sample 2		46	PI		0.64		
	English			Sample 2		46	Att		0.51		
	English			Sample 2		46	Sim		0.49		
	English			Sample 2		46	Suc		0.26		
Naglieri et al., 2014	English			Sample 3		53	PI	0.51			
	English			Sample 3		53	Att	0.49			
	English			Sample 3		53	Sim	0.65			
	English			Sample 3		53	Suc	0.37			
Naglieri et al., 2006	English					119	PI		0.50		0.51
	English					119	Att		0.39		0.39
	English					119	Sim		0.58		0.63
	English					119	Suc		0.45		0.51
Naglieri et al., 2007	English	Unselct		Hispanic	9.66	158	FS	0.40		0.62	
	English	Unselct		Non-Hispanics	9.85	1284	FS	0.69		0.65	
Naglieri et al., 2005	English	Unselct		Blacks		298	FS	0.69		0.60	0.66
	English	Unselct		Whites		1691	FS	0.65		0.64	0.65
Rosário, 2014	Other Euro	Unselct	G1 to G6	Grade 2	7.25	60	PI				0.34
	Other Euro	Unselct	G1 to G6	Grade 2	7.25	60	Att				0.43
	Other Euro	Unselct	G1 to G6	Grade 2	7.25	60	Sim				0.53
	Other Euro	Unselct	G1 to G6	Grade 2	7.25	60	Suc				0.39
	Other Euro	Unselct	G1 to G6	Grade 4	9.14	60	PI				-0.04
	Other Euro	Unselct	G1 to G6	Grade 4	9.14	60	Att				0.27
	Other Euro	Unselct	G1 to G6	Grade 4	9.14	60	Sim				0.47
	Other Euro	Unselct	G1 to G6	Grade 4	9.14	60	Suc				0.15
	Other Euro	Unselct	G1 to G6	Grade 6	11.09	60	PI				0.22

(Appendices continue)

## Appendix B (continued)

Study	Language	Math performance level	Grade level	Subgroup	Mean age	Sample Size	Types of PASS processes	Math accuracy	Math fluency	Problem-solving	Broad Math
	Other Euro	Unselct	G1 to G6	Grade 6	11.09	60	Att				0.26
	Other Euro	Unselct	G1 to G6	Grade 6	11.09	60	Sim				0.44
	Other Euro	Unselct	G1 to G6	Grade 6	11.09	60	Suc				0.32
	Other Euro	Unselct	G7 to G12	Grade 9	14.27	60	Pl				0.44
	Other Euro	Unselct	G7 to G12	Grade 9	14.27	60	Att				0.36
	Other Euro	Unselct	G7 to G12	Grade 9	14.27	60	Sim				0.52
	Other Euro	Unselct	G7 to G12	Grade 9	14.27	60	Suc				0.31
Taha, 2016	Non-Euro				13.52	50	Pl	0.81			
	Non-Euro				13.52	50	Att	0.84			
Wei et al., 2017	Non-Euro		G1 to G6		7.17	180	Pl		0.39	0.21	
	Non-Euro		G1 to G6		7.17	180	Att		0.12	0.03	
	Non-Euro		G1 to G6		7.17	180	Sim		0.03	0.39	
	Non-Euro		G1 to G6		7.17	180	Suc		0.07	0.24	
Wei et al., 2018	Non-Euro	Unselct	G1 to G6		8.16	179	Pl	0.09	0.14		
	Non-Euro	Unselct	G1 to G6		8.16	179	Att	0.19	0.33		
	Non-Euro	Unselct	G1 to G6		8.16	179	Sim	0.23	0.06		
Zhu et al., 2017	Non-Euro	Unselct	G1 to G6	Grade 2	7.72	77	Att		0.50	0.30	
	Non-Euro	Unselct	G1 to G6	Grade 2	7.72	77	Sim		0.01	0.26	
	Non-Euro	Unselct	G1 to G6	Grade 4	9.69	71	Att		0.39	0.16	
	Non-Euro	Unselct	G1 to G6	Grade 4	9.69	71	Sim		0.08	0.30	

*Note.* Non-Euro = Non-European language; Other Euro= Other European language; Unselct = Unselected; Kind = Kindergarten; ADHD = Attention Deficit Hyperactive Disorder; MLD = Math Learning Disabilities; TA = Typical Achievers; Pl = Planning; Att = Attention; Sim = Simultaneous; Suc = Successive; FS = Full Scale.