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The relation between executive functioning, reaction time, naming speed, and single word reading in children with typical development and language impairments

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Background. Few investigations have examined the relationship between a comprehensive range of executive functioning (EF) abilities and reading.

Aims. Our investigation identified components of EF that independently predicted single word reading, and determined whether their predictive role remained when additional variables were included in the regression analyses. This provided information about the EF processes that are related to reading, and the unity and diversity of EF.

Sample. This study consisted of 160 children: 88 were typically developing with no language difficulties; 72 had language impairments.

Method. The assessments involved decoding, 10 measures of EF, reaction time, naming speed, non-verbal and verbal age-equivalent scores.

Results and conclusions. In the first regression analysis, which only concerned the EF variables, the following verbal forms of EF had significant relationships with decoding: working memory, fluency, planning, and inhibition. Further regression analyses included additional predictor variables: reaction time, naming speed, and age-equivalent scores. These analyses indicated that most of the EF variables continued to predict decoding even when entered with competitor variables. Furthermore, after the entry of EF variables, there were no group differences in decoding (typical vs. language difficulties). We discuss the contribution of EF and other variables to reading abilities.

Executive functioning (EF) concerns higher order thinking and usually is considered a complex, fractionated skill (Baddeley, 2007; Miyake & Friedman, 2012) with a number of subcomponents (but see Duncan, Eme, Williams, Johnson, & Freer, 1996). These subcomponents include planning/problem-solving, mental flexibility, inhibition, executive-loaded working memory (ELWM), contextual memory, and fluency (Pennington & Ozonoff, 1996; van der Ven, Kroesbergen, Boom, & Leseman, 2012). Many of these abilities might be involved in decoding (Christopher et al., 2012), with the possibility of being non-phonological predictors of reading. Identifying predictors of decoding helps in understanding the component abilities that contribute to reading, potentially informing future interventions.

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Previous research has indicated significantly lower EF abilities in children with literacy impairments compared to typically developing children, suggesting that EF and reading abilities are associated. For example, Booth, Boyle, and Kelly (2010) in a meta-analysis noted that several areas of EF, including working memory, were impaired. Statistical associations of various sizes have been identified between literacy and updating/working memory in typical children and those with dyslexia (Arrington, Kulesz, Francis, Fletcher, & Barnes, 2014; Berninger et al., 2010; Booth, Boyle, & Kelly, 2014; Bull, Espy, & Wiebe, 2008; Jerman & Swanson, 2005; van der Sluis, de Jong, & van der Leij, 2007), although Savage, Lavers, and Pillay (2007) have raised questions about whether working memory makes a unique contribution to literacy (see also Christopher et al., 2012). In addition, although some investigators report relationships between inhibition and reading (Altemeier, Abbott, & Berninger, 2008; Blair & Razza, 2007; Bull et al., 2008), others have failed to identify significant relationships (Christopher et al., 2012; van der Sluis et al., 2007).

Little is known about the independent associations between comprehensive sets of EF abilities and decoding that take us beyond bivariate relationships to identify which EF abilities are independently related to decoding. Furthermore, addressing these issues is relevant to Miyake and Friedman’s (2012) suggestions about unity and diversity, that is, EF variables are related to one another, but not so closely as to represent the same construct. Consequently, the first research question concerned identifying the EF abilities that were significant independent predictors of decoding. The EF abilities that were assessed corresponded closely to those identified by Pennington and Ozonoff (1996): ELWM, fluency, planning/problem-solving, inhibition, and switching. The EF assessments involved both verbal and non-verbal versions, following Booth et al.’s (2010) recommendation for the inclusion of non-verbal tasks to assess EF when, as in our sample, language difficulties are present. Multiple regression analyses examined the independent contribution of each EF variable to decoding.

A second research question concerned whether EF variables still predicted decoding when other relevant variables were entered into the regression. It is rare for such ‘covariates’ to be included in studies of EF and this can provides important information about the shared variance between EF and other abilities that are related to decoding. If an EF variable does not ‘survive’ the competition and becomes non-significant, this suggests the EF variable shares variance with the competitor. The competitor variables were simple reaction time, naming speed, non-verbal and verbal age-equivalent scores.

Reaction time was included in the regression analysis because reading may be dependent on the ‘rapid execution of the underlying process’ (Kail, Hall, & Caskey, 1999, p. 312; Miller, Kail, Leonard, & Tomblin, 2001; Wolf & Bowers, 1999). Furthermore, Christopher et al. (2012) reported that processing speed, as assessed by several methods, was a significant predictor of reading. We assessed one of the most basic forms of information processing, motor reaction time to a visual stimulus, labelling this variable reaction time (RT), to emphasize that there was little or no cognitive processing involved. Thus, the analyses concerned whether there was an independent or a shared contribution of RT and EF to the prediction of decoding.

A more complex speed of processing ability is naming speed. Since the development of the Rapid Automatized Naming (RAN) task by Denckla and Rudel (1974), a significant body of research has been concerned with describing the strength of its relationship with literacy and trying to understand the reasons for these relationships (Kirby, Georgiou, Martinussen, & Parrila, 2010; Norton & Wolf, 2012). Naming speed reflects the time taken to name 30-50 items presented on a single page (e.g., digits, letters, object, and colours).
Many investigations have reported that naming speed predicts literacy abilities (Kirby et al., 2010; but see Christopher et al., 2012), and it appears that the pauses between naming items are the critical predictive aspects of the task (Georgiou, Parrila, & Kirby, 2006; Lervåg & Hulme, 2009). Letter and number RAN tasks usually have the strongest correlations with literacy (Kirby et al., 2010; Messer & Dockrell, 2011), possibly because these tasks involve reading-related stimuli. To minimize potential confounds, our naming speed task involved pictures and colours. Naming speed involves complex information processing which may have similarities to some EF processes. For example, it has been suggested that naming speed involves working memory processes and this accounts for its relationship with reading (Arnell, Joanisse, Klein, Busseri, & Tannock, 2009). Furthermore, naming speed tasks involve inhibition of the target names as these become prepotent stimuli. As a result, it was expected that including naming speed in the regressions would result in reduced associations between at least some EF variables and decoding.

Given that the EF assessments involved non-verbal and verbal tasks, non-verbal age-equivalent scores (NVAE) and verbal age-equivalent scores (VAE) were included in the regressions to control for general age-related abilities that might underlie EF performance. Age-equivalent scores were chosen, rather than chronological age, because chronological age would not have provided a valid indicator of the verbal abilities of children with language impairments. Because decoding involves language-related abilities, it was expected that NVAE would have only a small effect on the relation between EF and decoding. In contrast, it was thought that there could be a greater effect of VAE as this might be involved in both EF and decoding abilities.

Our analyses were based on a sample that included both typically developing children and children with language impairments, many of whom are likely to have reading impairments (Catts, Fey, Tomblin, & Zhang, 2002; McArthur, Hogben, Edwards, Heath, & Mengler, 2000; but see Catts, Adlof, Hogan, & Ellis-Weismer, 2005). The inclusion of the latter group increases the variance in the data and assists the detection of associations (Christopher et al., 2012; Moll, Loff, & Snowling, 2013). Although the sample was not a representative one, the children corresponded to the ability range seen in most UK classrooms. The typical group consisted of children with comparable chronological ages to the language-impaired sample and those with similar language ages. Thus, a strategy of including virtually all of the children who were assessed was adopted to obtain a sample with increased variance and a large number of participants.

To answer our research questions, two sets of multiple regression analyses were conducted. The first addressed the question of which EF variables were significant predictors of decoding. The significant predictors were then included in the second regression analysis, which addressed the question of whether EF variables remained significant predictors of decoding when additional competitor variables were included (RT, naming speed, NVAE and VAE).

**Method**

**Participants**

A total of 161 participants were recruited from schools within Greater London (both mainstream classes and specialist language units/classes) and, very occasionally, via contact with parents/guardians. Every child who completed all relevant assessments was included, except one with intellectual disabilities (BAS-II T-score of 20).
The typical group (n = 88; 28 females; mean age 117 months, SD = 28) had scaled scores of 8 or higher on four subscales from the Clinical Evaluation of Language Fundamentals-4-UK (CELF-4-UK, Semel, Wiig, & Secord, 2006; mean = 10 and SD = 3; Recalling Sentences; Formulated Sentences; Word Classes-Receptive; and Word Classes-Expressive), and all had non-verbal abilities in the average range (T-scores of 40 or greater on BAS-II). There also were 72 children with language impairments. Thirty-one of these children had less severe language impairments (7 females; mean age 126 months, SD = 27) with scores of 7 or below on 1 or 2 subscales of the CELF-4-UK. Twenty-two of these 31 children had T-scores in the typical range on the non-verbal subscale of the British Abilities Scales-II (BAS-II; >39; Elliot, Smith, & McCullough, 1996; mean = 50 and SD = 10). There were 41 children with more severe language impairments and who met the criteria for specific language impairment (13 females; mean age 138 months, SD = 16) who had scores below 8 on at least three of four subscales of the CELF-4-UK and had non-verbal abilities in the average range (T-scores of 40 or greater on BAS-II).

The mean standardized score for decoding based on the Test of Word Reading Efficiency (TOWRE; Torgesen, Wagner, & Rashotte, 1999) was 98.6 (SD = 20.2) and 20% of the sample had standardized scores below 86. The mean TOWRE score of the typical group was 108.0 (SD = 14.5), that of the group with less severe language impairments was 91.3 (SD = 24.9), and that of the more language-impaired group was 84.1 (SD = 15.6). For further details about group abilities, see Henry, Messer and Nash (2012). No participants had diagnoses of hearing impairments, intellectual disability, or other developmental disorders according to the school records and teacher reports (e.g., ADHD, ASD).

This project had ethical approval from the relevant University Ethics Committee. Written informed consent was obtained from parents/guardians (telephone permission occasionally) and from children. Testing involved 3–8 sessions, taking about 3.5 hr in total, usually conducted at school, but occasionally at the participant’s home. The tests were presented in the same order for all children, except when the assessor thought a different order would help the motivation and engagement of a child.

**Tools and materials**

Assessment of language, verbal and non-verbal age equivalence and decoding

The standardized assessments consisted of four subtests from the CELF-4-UK. Also administered were four subtests from the British Ability Scales (BAS-II, Elliot et al., 1996), namely Word Definitions, Verbal Similarities, Matrices and Quantitative Reasoning; verbal age-equivalent scores of scaled abilities in months (VAE) were based on the first two assessments, and non-verbal age-equivalent scores of scaled abilities in months (NVAE) were based on the second two assessments. Decoding was assessed by the ‘A’ version of the TOWRE containing both real word decoding (sight word efficiency) and non-word decoding (phonemic word efficiency).

Reaction time

The Cambridge Neuropsychological Test Automated Battery Motor Screening Test (CANTAB; Cambridge Cognition, 2006; Gau & Shang, 2010) was used to obtain simple reaction time. The children were told that a series of crosses would appear in different locations on the computer, and their task was to touch them as soon as possible. There
were three practice trials and ten test trials. The mean latency of responses was automatically recorded.

**Naming speed**
This was a subtest from the CELF-4-UK. Item 1 comprised of 36 circles coloured yellow, red, green, or blue, and the task was to name the colours as quickly as possible. Item 2 comprised of 36 shapes (triangles, squares, stars, and circles) with no colours, and the task was to name these as quickly as possible. No repetitions of test items were allowed and there was no discontinue rule. The time taken and errors were recorded. The variable used in analyses was the average time taken to name the two sets of stimuli.

**EF Assessments**
Where possible, pairs of *simple* standardized tasks that assessed the verbal and non-verbal executive skills in each domain were selected, and new tasks were developed for inhibition.

**Executive-loaded working memory.** The verbal task was Listening Recall (Working Memory Test Battery for Children, WMTB-C, Pickering & Gathercole, 2001). The experimenter read out a series of short sentences and the child judged whether each was true or not. Then, the child was asked to remember the final word from each sentence in correct serial order. Trials commenced with list lengths of one item and proceeded to longer lists, with six trials per list length, until 4/6 trials were incorrect. Scores were based on total trials correct, as recommended by Ferguson, Bowey, and Tilley (2002). Test–retest reliabilities are .38–.83 for the relevant ages (Pickering & Gathercole, 2001).

The non-verbal ELWM task was the odd-one-out test (Henry, 2001). The experimenter displayed three cards depicting simple nonsense diagrams (horizontally orientated on 20 × 4 cm cards). The child pointed to the ‘odd-one-out’. The spatial location of each odd-one-out card was then recalled via a set of response sheets (20 × 30 cm) depicting the relevant number of ‘empty’ cards. Trials commenced with lists of one item and proceeded to longer lists, with three trials per list length, until 2/3 trials were incorrect. Scores were based on total trials correct. The span version of this task has a reliability of .80 (Henry, 2001).

**Fluency.** The Verbal Fluency scale of the Delis–Kaplan Executive Functioning System (D-KEFS, Delis, Kaplan, & Kramer, 2001) involves generating as many words as possible in one minute. The ‘category fluency’ task required the participant to generate items in two semantic categories, ‘animals’ and ‘boy’s names’. The ‘letter fluency’ component of this task was omitted from the analyses as performance on this task contains elements of decoding (de Jong, 2011).

Non-verbal fluency (Design Fluency, D-KEFS) involved a response booklet containing patterns of dots in boxes. The participant had one minute to draw as many different designs as possible, each in a different box. The participant was asked to connect the dots using four straight lines (no lines could be drawn in isolation). Condition 1 contained only filled dots; condition 2 contained filled and empty dots and the child connected only empty dots. Design fluency was the average raw score from these two conditions. Test–
retest reliabilities: letter (.67), category (.70), filled dots (.66), and empty dots (.43; Delis et al., 2001).

Planning. The Sorting Test (D-KEFS) was used to assess planning. Participants sorted sets of six cards into two groups of three, in as many ways as they could. For the verbal task, there were three correct sorts (e.g., transport/animals; things that fly/things that move along the ground); for the ‘perceptual’ task, there were five correct sorts (e.g., small/large; straight/curved edges). The total numbers of correct verbal/perceptual sorts were used as the measures of verbal/non-verbal planning (test–retest reliability .49, Delis et al., 2001).

Inhibition. A new test was developed, ‘Verbal Inhibition, Motor Inhibition’ (VIMI). There were two types of response: either to copy the Experimenter, or to inhibit copying and produce a different response. This task was based on inhibition measures such as Luria’s hand game. For Part A of the verbal task, in Block 1, the experimenter said either ‘doll’ or ‘car’ and the participant was asked to repeat the same word. Next, in block 2, the child was asked to say the opposite word: ‘If I say doll, you say car’. This was followed by a second ‘copy’ block and a second ‘inhibit’ block. Each of the four blocks consisted of 20 trials. This entire sequence of copy/inhibit blocks was repeated in Part B, with new stimuli (‘bus’ and ‘drum’).

The non-verbal motor task followed the same format, but words were replaced with hand actions. For Part A, the stimuli were a pointed finger versus a fist; for Part B, the stimuli were a flat horizontal hand versus a flat vertical hand. The combined number of errors made across Parts A and B on each task was used as the measure of inhibition. Cronbach’s alpha, based on total error scores from Parts A and B was .72 for the verbal task and .92 for the non-verbal task.

Switching. The Trail Making Test (D-KEFS) was used to assess verbal switching. Children joined small circles containing letters and numbers alternately, in sequence (1-A-2-B-3-C through to 16-P), this required cognitive flexibility on a sequencing task. Two control assessments were also used, number sequencing (connecting the numbers 1–16) and letter sequencing (connecting the letters A–P). ‘Switching cost’ was calculated as follows. We first calculated how many numbers and letters, respectively, per second were connected and took an average of these times (average ‘single task’ letter/number connection speed). We then calculated how many items were connected per second in the letter/number switching condition. To obtain ‘switching cost’, the items per second scores for letter number switching were subtracted from the average items per second single task letter/number scores. Scores were expressed as an ‘items per second’ switching cost. Test–retest reliabilities reported for the component tasks of the Trail Making Test are as follows: number sequencing (.77), letter sequencing (.57), and letter/number switching (.20; Delis et al., 2001).

The non-verbal switching test, Intra/Extra Dimensional Shift, was from a computer presented battery (CANTAB). This involved rule acquisition and reversal. Two simple coloured stimuli were presented on a computer screen, and by touching one, children learnt from feedback which was ‘correct’. The children followed this rule in a number of trials. Later, the second dimension, an irrelevant white line was introduced adjacent to the
coloured shape, but subsequently overlaying it. At this stage, children had to maintain the same response to solve the problem. In later trials, the complex stimuli changed and the children had to switch attention to the previously irrelevant dimension, to give the ‘correct’ response. Total error scores were used as the dependent measure (test–retest reliability .40, Cambridge Cognition, 2006).

**Data screening**

Table 1 gives details of the children’s performance, and further details can be found in Henry et al. (2012). Previous analyses on this sample have shown that after controlling for age and non-verbal IQ, there were significant differences between the group with more severe language impairment and the typical group in verbal and non-verbal ELWM, verbal and non-verbal fluency, non-verbal inhibition, and non-verbal planning. Furthermore, the children in the two language-impaired subgroups showed a similar pattern of EF performance.

Bivariate correlations were calculated between all the variables used in the analyses (Table 1). The intercorrelations between the EF measures were mostly moderate and significant, and the lowest correlations were between verbal switching and the other EF variables. This suggested that collinearity was unlikely to be a problem with these data.

For all variables, checks were made on skewness and kurtosis. In several instances, transformations provided more suitable distributions: log transformations to simple reaction time, RAN and non-verbal inhibition; and square root transformations to design fluency and verbal inhibition. Some outliers were identified; therefore, further statistical checks were carried out in relation to the regression analyses (Durbin–Watson, tolerance/VIF statistics, Cook’s/Mahalanobis distances, and standardized DFbetas). These checks identified one potentially influential case according to Mahalanobis distance for the first regression analysis, and this participant was removed to give a sample of 159 children (Field, 2009).

**Results**

*The prediction of decoding from executive functioning*

The 10 EF variables were simultaneously entered into a hierarchical multiple regression analysis with the raw TOWRE score as the dependent variable. There was a significant change in the $R^2$ value (.51; $p < .001$) reflecting the entry of the EF variables. Four verbal forms of EF (ELWM, fluency, planning, and inhibition) were significant independent predictors of decoding (Table 2).

*The prediction of decoding from EF, processing speed, and age equivalence*

Further regression analyses were conducted to investigate the prediction of decoding from EF and competitor variables consisting of RT, naming speed, NVAE and VAE. The four EF variables that were significant predictors in the previous analysis were entered at Step 1, and the $R^2$ change was similar to that in the previous analysis, indicating that these EF variables accounted for most of the shared variance between EF and decoding (Table 3).

Non-verbal age-equivalent scores and VAE were entered next (i.e., at steps 2 and 3: see Table 3). This provided information about the way age equivalence affected the
| Variable                  | Mean   | SD     | 95% CI   | 1   | 2     | 3     | 4     | 5     | 6     | 7     | 8     | 9     | 10    | 11    | 12    | 13    | 14    | 15    |
|---------------------------|--------|--------|----------|-----|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| TOWRE (raw)               | 45.52  | 15.29  | 43.1, 47.9 | .53 | .47   | .61   | .53   | .39   | .32   | -.31  | -.30  | -.06  | -.16  | -.09  | -.66  | .57   | .64   |       |       |
| Verbal ELWM               | 12.94  | 4.05   | 12.3, 13.6  | .45 | .58   | .52   | .31   | .39   | .20   | -.37  | .03   | -.28  | -.20  | -.61  | .61   | .65   |       |       |       |
| Non-verbal ELWM          | 9.32   | 3.24   | 8.8, 9.8    | .48 | .50   | .24   | .36   | -.28  | -.31  | -.03  | -.28  | -.16  | -.52  | .55   | .47   |       |       |       |       |
| Verbal Fluency            | 31.90  | 7.82   | 30.7, 33.1  | .66 | .36   | .42   | -.18  | -.23  | .01   | -.29  | -.28  | -.64  | .61   | .66   |       |       |       |       |       |
| Non-verbal Fluency       | 7.58   | 2.93   | 3.7, 3.9    | .43 | .38   | -.18  | -.25  | .03   | -.26  | -.21  | -.63  | .66   | .60   |       |       |       |       |       |       |
| Verbal Planning           | 2.56   | 1.15   | 2.4, 2.7    | .01 | -.07  | -.16  | -.00  | -.05  | -.11  | -.40  | .31   | .37   |       |       |       |       |       |       |       |
| Non-verbal Planning      | 4.67   | 2.25   | 4.3, 5.0    | -.18 | -.20  | .01   | -.32  | -.08  | -.32  | .49   | .49   |       |       |       |       |       |       |       |       |
| Verbal Inhibition         | 6.92   | 5.11   | 2.3, 2.6    | .45 | .02   | .17   | .03   | .33   | -.22  | -.22  | -.20  |       |       |       |       |       |       |       |       |
| Non-verbal Inhibition    | 19.99  | 11.09  | 1.2, 1.3    | -.12 | .22   | .14   | .41   | -.30  | -.25  |       |       |       |       |       |       |       |       |       |       |
| Verbal Switching          | 0.08   | 0.09   | 0.7, 1.0    | .10 | -.15  | -.03  | -.03  | .03   | .03   |       |       |       |       |       |       |       |       |       |       |
| Non-verbal Switching     | 28.90  | 11.97  | 27.0, 30.8  | .07 | .22   | -.38  | -.34  |       |       |       |       |       |       |       |       |       |       |       |       |
| Simple Reaction Time     | 2.93   | 0.12   | 2.9, 2.9    | .22 | -.13  | -.12  |       |       |       |       |       |       |       |       |       |       |       |       |       |
| Naming Speed             | 1.45   | 0.13   | 1.4, 1.4    | -.68 | -.61  |       |       |       |       |       |       |       |       |       |       |       |       |       |       |
| BAS-II NVAE              | 141.08 | 38.07  | 135.1, 147.0| .78 |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |
| BAS-II VAE               | 130.14 | 31.45  | 135.1, 147.0|       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |       |

Note. *Correlations where $p < .05$ are in bold.
standardized beta-coefficients of the EF variables. At Step 2, NVAE was a significant
predictor of decoding, but it only had an appreciable effect on the beta-coefficient for
ELWM, which was reduced to a marginally significant predictor. At Step 3, VAE was a
significant predictor, and its entry made ELWM and NVAE non-significant predictors.
Thus, verbal fluency, verbal planning, and verbal inhibition all survived competition from
age equivalence variables, and the analysis indicated that VAE shared variance with ELWM
and NVAE. The entry of RT at Step 4 did not result in a significant \( R^2 \) change. In contrast,
naming speed when entered at Step 5 was a significant and important predictor, and the
beta-coefficients of some other variables were reduced, most notably the EF variables.

Next, the regression was repeated entering the information processing variables
before the age equivalence variables, to better understand the relationships between the
information processing variables and EF (Table 4). RT when entered at Step 2 did not
result in a significant \( R^2 \) change, nor did its entry appreciably reduce the standardized beta-
coefficients of the EF variables. However, naming speed was a significant predictor at Step
3, and its entry resulted in ELWM becoming a non-significant predictor, and the beta-
coefficients for the other EF variables were reduced in size. As would be expected from
the previous analysis, the entry of NVAE at Step 4 had very little effect on the size of the
beta-coefficients of the EF variables. The entry of VAE at Step 5 further reduced the beta-
coefficients of the EF variables: Verbal fluency and verbal inhibition remained as
significant predictors, with verbal planning becoming a non-significant predictor. The
entry of VAE also resulted in NVAE becoming a non-significant predictor of decoding. A
further regression analysis was conducted to check whether, in the absence of the verbal
EF variables, the five non-verbal EF variables predicted decoding, none of the variables
were significant predictors.

In both competitor analyses, ELWM was not a significant predictor in the presence of
naming speed and VAE. ELWM is highly related to IQ and general ability (Friedman,
Miyake, Corley, Young, & DeFries, 2006; Gathercole & Pickering, 2000). Further analysis
indicated that there was a significant correlation between verbal ELWM and NVAE and
also with VAE, \( r(159) = .61, p < .001 \); \( r(159) = .65, p < .001 \). There also was a significant

<table>
<thead>
<tr>
<th>Variables</th>
<th>Standardized beta-coefficients</th>
<th>Unstandardized beta-coefficients</th>
<th>Standard error</th>
<th>95% CI</th>
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<td>( R^2 ) Change at Step 1</td>
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<td>1.8</td>
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<td>0.1</td>
<td>−0.0, 0.3</td>
</tr>
</tbody>
</table>

*\( p < .05 \); ***\( p < .001 \).
correlation between verbal ELWM and naming speed, $r(159) = -.61, p < .001$. This suggests that ELWM became a non-significant predictor because of shared variance with variables assessing general ability and naming speed.

Additional regression analyses were conducted to investigate the effect of group (typical vs. any language impairments). Group was dummy-coded and entered either at Step 1 or after the four EF variables (i.e., Step 5). When group was entered at Step 1, it was a

<table>
<thead>
<tr>
<th>Variables</th>
<th>Step 1 (EF)</th>
<th>Step 2 (EF + NVAE)</th>
<th>Step 3 (EF + NVAE + VAE)</th>
<th>Step 4 (previous variables + RT)</th>
<th>Step 5 (previous variables + naming speed)</th>
</tr>
</thead>
<tbody>
<tr>
<td>$R^2$ Change</td>
<td>.48***</td>
<td>.02*</td>
<td>.03**</td>
<td>.00</td>
<td>.04***</td>
</tr>
<tr>
<td>Verbal ELWM</td>
<td>.21**</td>
<td>.14†</td>
<td>.09</td>
<td>.09</td>
<td>.03</td>
</tr>
<tr>
<td>(0.8 to 1.3)</td>
<td>(0.0 to 1.0)</td>
<td>(0.3 to 0.9)</td>
<td>(0.2 to 0.9)</td>
<td>(0.5 to 0.7)</td>
<td></td>
</tr>
<tr>
<td>Verbal Fluency</td>
<td>.39***</td>
<td>.32***</td>
<td>.26**</td>
<td>.28**</td>
<td>.21*</td>
</tr>
<tr>
<td>(0.8 to 1.1)</td>
<td>(0.3 to 0.9)</td>
<td>(0.2 to 0.8)</td>
<td>(0.2 to 0.9)</td>
<td>(0.1 to 0.7)</td>
<td></td>
</tr>
<tr>
<td>Verbal Planning</td>
<td>.17**</td>
<td>.16*</td>
<td>.14*</td>
<td>.14*</td>
<td>.10</td>
</tr>
<tr>
<td>(2.3 to 3.9)</td>
<td>(0.5 to 3.8)</td>
<td>(0.2 to 3.5)</td>
<td>(0.3 to 3.5)</td>
<td>(0.3 to 2.9)</td>
<td></td>
</tr>
<tr>
<td>Verbal Inhibition</td>
<td>-.19**</td>
<td>-.17**</td>
<td>-.17**</td>
<td>-.17**</td>
<td>-.11*</td>
</tr>
<tr>
<td>(2.8 to 1.0)</td>
<td>(2.5 to 0.9)</td>
<td>(2.5 to 0.9)</td>
<td>(2.5 to 0.9)</td>
<td>(2.5 to 0.9)</td>
<td></td>
</tr>
<tr>
<td>NVAE</td>
<td>.20*</td>
<td>.06</td>
<td>.05</td>
<td>.06</td>
<td>.06</td>
</tr>
<tr>
<td>(0.1 to 0.1)</td>
<td>(0.0, 0.0)</td>
<td>(0.0, 0.0)</td>
<td>(0.0, 0.0)</td>
<td>(0.0, 0.0)</td>
<td>(0.0, 0.0)</td>
</tr>
<tr>
<td>VAE</td>
<td>.28**</td>
<td>.27**</td>
<td>.28**</td>
<td>(0.1, 0.5)</td>
<td>(0.1, 0.1)</td>
</tr>
<tr>
<td>RT</td>
<td>.07</td>
<td>.08</td>
<td>(8.2, 7.5, 10.3, 7.2, -6.6 to 23.1)</td>
<td>-4.0 to 24.5)</td>
<td></td>
</tr>
<tr>
<td>Naming Speed</td>
<td>- .32***</td>
<td>- .32***</td>
<td>- .32***</td>
<td>(38.0, 10.2, -58.1 to -17.9)</td>
<td></td>
</tr>
</tbody>
</table>

Note. Step 1 and Step 5 are included to help with the inspection of the data. Data about unstandardized beta, standard error, lower and upper 95% confidence intervals are given in parentheses.

*p < .05; **p < .01; ***p < .001; †p < .01.
significant predictor of decoding, indicating there were differences in the decoding abilities of the two groups (standardized beta-coefficient = .26; \( p < .001; SE = 0.3; 95\% \text{ CI} [0.1, 1.2] \)). However, when group was entered at Step 5, it was no longer a significant predictor, indicating that the four EF variables had removed the variance responsible for the significant difference at Step 1 (standardized beta-coefficient = .05; \( SE = 1.9; 95\% \text{ CI} [-2.3, 5.2] \)).

**Table 4.** Standardized beta-coefficients from the regression analysis of the prediction of decoding from executive functioning (EF) variables (step 1), simple processing speed (step 2), naming speed (step 3), and non-verbal age-equivalent score (step 4) and verbal age-equivalent score (step 5)

<table>
<thead>
<tr>
<th>Variables</th>
<th>Step 1 (EF)</th>
<th>Step 2 (EF + RT)</th>
<th>Step 3 (EF + RT + Naming Speed)</th>
<th>Step 4 (previous variables + NVAE)</th>
<th>Step 5 (previous variables + VAE)</th>
</tr>
</thead>
<tbody>
<tr>
<td>R² Change</td>
<td>.48***</td>
<td>.01</td>
<td>.05***</td>
<td>.00</td>
<td>.03***</td>
</tr>
<tr>
<td>Verbal ELWM</td>
<td>.21**</td>
<td>.22*</td>
<td>.11</td>
<td>.09</td>
<td>.03</td>
</tr>
<tr>
<td></td>
<td>(0.8, 0.3, 0.3 to 1.3)</td>
<td>(0.8, 0.3, 0.3 to 1.4)</td>
<td>(0.4, 0.3, 0.1 to 1.0)</td>
<td>(0.3, 0.3, 0.2 to 0.9)</td>
<td>(0.1, 0.3, 0.5 to 0.7)</td>
</tr>
<tr>
<td>Verbal Fluency</td>
<td>.39***</td>
<td>.42***</td>
<td>.29**</td>
<td>.21*</td>
<td>.21*</td>
</tr>
<tr>
<td></td>
<td>(0.8, 0.1, 0.5 to 1.1)</td>
<td>(0.8, 0.1, 0.5 to 1.1)</td>
<td>(0.6, 0.2, 0.3 to 0.9)</td>
<td>(0.5, 0.2, 0.2 to 0.8)</td>
<td>(0.4, 0.2, 0.1 to 0.7)</td>
</tr>
<tr>
<td>Verbal Planning</td>
<td>.17**</td>
<td>.17**</td>
<td>.12*</td>
<td>.10</td>
<td>.10</td>
</tr>
<tr>
<td></td>
<td>(2.3, 0.9, 0.6 to 3.9)</td>
<td>(2.3, 0.8, 0.6 to 3.9)</td>
<td>(1.6, 0.8, 0.0 to 3.2)</td>
<td>(1.5, 0.9, 0.0 to 3.2)</td>
<td>(1.3, 0.8, 0.3 to 2.9)</td>
</tr>
<tr>
<td>Verbal Inhibition</td>
<td>-.19***</td>
<td>-.18**</td>
<td>-.12*</td>
<td>-.11*</td>
<td>-.11*</td>
</tr>
<tr>
<td></td>
<td>(-2.8, 0.9, -4.5 to -1.0)</td>
<td>(-2.7, 0.9, -4.5 to -1.0)</td>
<td>(12.9, 7.3, -1.6 to 27.4)</td>
<td>(-1.7, 0.9, -3.5 to -0.0)</td>
<td>(-1.7, 0.9, -3.4 to -0.0)</td>
</tr>
<tr>
<td>RT</td>
<td>.09</td>
<td>.10</td>
<td>.08</td>
<td>.08</td>
<td>.08</td>
</tr>
<tr>
<td></td>
<td>(11.5, 7.7, -3.8 to 26.7)</td>
<td>(12.9, 7.3, -3.8 to 26.7)</td>
<td>(12.1, 7.4, -2.4 to 26.7)</td>
<td>(10.3, 7.2, -4.0 to 24.5)</td>
<td>(10.3, 7.2, -4.0 to 24.5)</td>
</tr>
<tr>
<td>Naming Speed</td>
<td>-.31***</td>
<td>-.32**</td>
<td>-.32***</td>
<td>-.32***</td>
<td>-.32***</td>
</tr>
<tr>
<td></td>
<td>(-41.1, 9.8, -60.4 to -21.8)</td>
<td>(-37.1, 10.4, -60.4 to -21.8)</td>
<td>(-38.0, 10.2, -57.7 to -16.6)</td>
<td>(-38.0, 10.2, -58.1 to -17.9)</td>
<td>(-38.0, 10.2, -58.1 to -17.9)</td>
</tr>
<tr>
<td>NVAE</td>
<td>.09</td>
<td>-.06</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.0, 0.0, 0.0, 0.0, -0.0 to 0.1)</td>
<td>(0.0, 0.0, 0.0, 0.0, -0.1 to 0.1)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>VAE</td>
<td>.28**</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.1, 0.0, 0.5 to 0.2)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Note. Data about unstandardized beta, standard error, lower and upper 95% confidence intervals are given in parentheses.

*p < .05; **p < .01; ***p < .001.
To check whether there might be group differences in the prediction of decoding, separate bivariate correlations were conducted between the verbal EF variables and decoding. There was only one instance when the difference between the correlations ($r$) of the three groups was $>.3$, and this involved verbal fluency, and the specific language impairment group had a lower non-significant correlation ($r = .27$) with decoding than the other two groups.

**Discussion**

The purpose of this study was to investigate the role of different forms of EF ability in predicting decoding. Relationships between EF and decoding were first examined alone and then in the presence of relevant ‘competitor’ variables. In the first regression analysis that did not include competitor variables, four verbal EF tasks (ELWM, fluency, planning, and inhibition) were significant predictors of decoding. In the regression analysis that included competitor variables (VAE, NVAE, RT, and naming speed), two of the EF variables, namely verbal fluency and verbal inhibition, remained significant predictors of decoding, with VAE and naming speed also making significant contributions to the regression model. The only EF domain that was not a significant predictor of decoding was switching: This variable has failed to be a significant predictor of important cognitive processes in other investigations (Henry *et al.*, 2012).

Thus, the findings provide evidence that verbal components of EF contribute independently to decoding ability, even when important cognitive variables have been controlled. Further evidence of the importance of the contribution of verbal EF to decoding comes from the finding that when we included ‘group’ (typical vs. language impairment) in the analyses, it only was significantly related to decoding when it was entered at Step 1. Group was no longer a significant predictor of decoding when entered after the four EF variables, indicating that removing the variance associated with EF variables resulted in there no longer being a significant different between the groups in decoding. This suggests that EF abilities could be responsible for the differences in decoding found at Step 1.

Our findings concerning EF variables and decoding are relevant to current discussions of ‘unity and diversity’ in relation to the structure of EF (Miyake & Friedman, 2012). In our analyses, the bivariate correlations showed that most of the EF variables were significantly correlated with one another, indicating unity. Nevertheless, the regression analyses showed that EF variables had significant, independent relations with another variable (decoding), indicating diversity, and providing additional support for the unity and diversity model.

Of the EF variables, verbal fluency was the best predictor of decoding (taking into consideration beta values and the second regression analysis with competitor variables). This may be because both verbal fluency and decoding involve lexical search and retrieval processes of representations in long-term memory (Lervåg & Hulme, 2009; Seidenberg, 2007). Support for this suggestion was provided by the large decrement in the beta-coefficient for verbal fluency when naming speed (i.e., lexical retrieval) was entered into the regression, suggesting shared variance between verbal fluency and naming speed. However, despite this decrement, verbal fluency remained a significant independent predictor of decoding. This could be because an important component of fluency was the organization of lexical searches, whereas the naming task involved only the retrieval of
identified items. Hence, the ‘EF’ aspects of verbal fluency, which are relevant to decoding, appear to involve both independent and shared processes with naming tasks.

Our analyses also indicated that verbal inhibition was a key predictor of decoding; furthermore, it shared relatively little variance with the other competitor variables. Inhibition processes, especially response inhibition, may contribute to decoding abilities because they ensure a written word is fully processed rather than guessed ‘too early’ on the basis of either incomplete information and/or a tendency to produce a pre-potent response (Arrington et al., 2014; Booth et al., 2014; Diamond, 2013). In contrast to our findings, Christopher et al. (2012) failed to detect significant relations between inhibition and decoding. This discrepancy may be due to methodological differences. Christopher et al. used several tasks involving ‘the ability to remove outdated information and ignore irrelevant extraneous information to help maintain current goals’ (i.e., attentional inhibition), whereas our EF measures assessed the inhibition of pre-potent responses. Thus, the current findings suggest that the inhibition of pre-potent responses may be a better predictor of decoding than attentional inhibition, and this conclusion is consistent with several studies (Arrington et al., 2014; Bull et al., 2008; van der Sluis et al., 2007).

Verbal planning was significantly related to decoding in the first regression analysis. This ability is likely to be required when identifying the appropriate strategy for tackling orthographically or phonologically complex items (Farrington-Flint, Canobi, Wood, & Faulkner, 2010). However, verbal planning was no longer a significant predictor in the presence of naming speed (Table 3), indicating that verbal planning shared variance with serial naming, possibly because both tasks involve making decisions about the appropriateness of planned responses. Interestingly, verbal ELWM and verbal fluency also shared variance with naming speed, suggesting that naming speed involves several EF abilities. Such findings merit future research, as the overlapping abilities shared between EF tasks and other cognitive variables are not well understood.

The findings in relation to verbal ELWM were more complex than for the other EF variables. The first regression analysis indicated that verbal ELWM was a significant predictor of decoding. This is consistent with previous findings (Arrington et al., 2014; Booth et al., 2010, 2014; Christopher et al., 2012) and could be because of the need to temporarily store information, while simultaneously processing other information when reading. However, ELWM did not remain a significant predictor after the entry of VAE or naming speed (Tables 3 and 4). One explanation for this is that ELWM and IQ are closely related (Coloma, Rebolloa, Palaciosoa, Juan-Espinosa, & Kyllonen, 2004), although these abilities may involve independent predictive components (Alloway & Alloway, 2010). This explanation is consistent with the high correlations between verbal ELWM and both age equivalence variables. Furthermore, Arnell et al. (2009) have argued that ELWM processes may be partly responsible for significant relationships between RAN and reading, and our findings are consistent with this claim. Shared variance between ELWM, naming speed, and age-equivalent scores provides an explanation of some of the discrepancies in previous investigations. For example, although ELWM is often poor in children with dyslexia and SLI (Booth et al., 2010), and significant relationships have been reported between ELWM and reading, questions have been raised about the relations between ELWM and reading (Savage, Cornish, Manly, & Hollis, 2006). Our findings indicate that the identification of such relations is likely to be affected by the control variables that are included in the analyses, given the shared variance between verbal ELWM, naming speed, and VMA.

The competitor analyses were also relevant to our general understanding of whether the prediction of decoding was a result of shared variance between verbal ability and
verbal EF (Booth et al., 2014; Pimperton & Nation, 2014). Verbal fluency and verbal inhibition continued to be significant predictors of decoding, despite competition from VAE (and NVAE). Furthermore, verbal planning was a significant predictor of decoding after inclusion of VAE and NVAE (but was not in the presence of naming speed). These findings indicate that after taking account of verbal ability, three verbal EF tasks continued significantly to predict decoding. We suggest that this is because these tasks assess higher-level abilities involved in the processing of complex verbal material and that as a result, they continue to predict decoding because these higher-level abilities were statistically independent of children’s verbal ability (largely based on vocabulary and grammar).

Another feature of the findings concerned the prediction of decoding from information processing variables. Contrary to some predictions (Kail et al., 1999), our simple RT variable was not a significant predictor of decoding, nor did it share variance with EF variables or even with naming speed. In contrast, naming speed was an important predictor of decoding. Wolf and Bowers (1999), and more recently Norton and Wolf (2012), have argued that naming speed tasks contain many of the specific cognitive operations involved in reading, with serial processing of visual items and the retrieval of phonological information from long-term memory likely to be particularly salient (Kirby et al., 2010; but see Christopher et al., 2012). Our findings support the argument of Wolf et al., and, as already discussed, it is possible to see Naming speed as a type of complex EF task.

Lastly, a potential limitation of the study is that the dependent variable (TOWRE) required decoding to take place in a fixed time-period, possibly favouring the detection of relations between speed of information processing and decoding. However, it also is the case that the TOWRE has high and significant associations with other non-timed measures of reading, so it is unlikely that the speed of processing element of the TOWRE is the only reason for these associations. In addition, the inclusion of processing speed variables in the competitor analyses controlled for such effects. A better understanding of the relationships between variables could be obtained from structural equation modelling; however, the sample size and number of variables meant this analysis was not feasible. There also needs to be caution about interpretation of group characteristics as the selection criteria for children in the language-impaired group relied on school reports of comorbid disabilities, and it is possible that some children had other disabilities that were not identified.

To summarize, regression analyses identified several verbal forms of EF (ELWM, fluency, inhibition, planning) that were significant independent predictors of decoding. Further analyses, which included competitor variables, indicated that some verbal EF variables continued to be significant predictors of decoding (fluency, inhibition). Thus, EF abilities appear to share cognitive abilities with decoding and/or contribute to the development of decoding abilities. These findings provide a new understanding of the role of non-phonological cognitive processes in decoding. They also contribute important information of practical significance for classroom teaching, by revealing the way that distinct components of verbal EF, particularly inhibition and fluency, provide independent contributions to an important educational outcome.

References


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