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Factor structure of the WISC-V in four standardization age groups: Exploratory and hierarchical factor analyses with the 16 primary and secondary subtests

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Preliminary results were presented at the 2015 Annual Convention of the American Psychological Association, Toronto, Ontario, Canada.

Abstract

This study examined the factor structure of the Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) with four standardization sample age groups (6-8, 9-11, 12-14, 15-16 years) using exploratory factor analysis (EFA), multiple factor extraction criteria, and hierarchical EFA not included in the WISC-V Technical and Interpretation Manual. Factor extraction criteria suggested that one to four factors might be sufficient despite the publisher-promoted, five-factor solution. Forced extraction of five factors resulted in only one WISC-V subtest obtaining a salient pattern coefficient on the fifth factor in all four groups, rendering it inadequate. Evidence did not support the publisher's desire to split Perceptual Reasoning into separate Visual Spatial and Fluid Reasoning dimensions. Results indicated that most WISC-V subtests were properly associated with the four theoretically oriented first-order factors resembling the WISC-IV, the g factor accounted for large portions of total and common variance, and the four first-order group factors accounted for small portions of total and common variance. Results were consistent with EFA of the WISC-V total standardization sample.

KEYWORDS

exploratory factor analysis, factor extraction criteria, Schmid-Leiman higher-order analysis, structural validity, WISC-V

The Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V; Wechsler, 2014a) includes 16 intelligence-related subtests, five first-order factor index scores (Verbal Comprehension [VC], Visual Spatial [VS], Fluid Reasoning [FR], Working Memory [WM], and Processing Speed [PS]), and the hierarchically ordered Full Scale score (FSIQ). The Word Reasoning and Picture Completion subtests of the Wechsler Intelligence Scale for Children-Fourth Edition (WISC-IV; Wechsler, 2003) were removed and three new subtests were added. New subtests include Picture Span (adapted from

the Wechsler Preschool and Primary Scale of Intelligence-Fourth Edition [Wechsler, 2012]) to measure visual working memory and Visual Puzzles and Figure Weights (adapted from the Wechsler Adult Intelligence Scale-Fourth Edition [Wechsler, 2008]) to measure visual spatial and fluid reasoning, respectively. Separating the former Perceptual Reasoning factor into separate and distinct Visual Spatial and Fluid Reasoning factors was a major goal in developing and marketing the WISC-V.

The WISC-V includes seven "Primary" subtests (Similarities [SI], Vocabulary [VC], Block Design [BD], Matrix Reasoning [MR], Figure Weights [FW], Digit Span [DS], and Coding [CD]) that are used in producing the FSIQ; and three additional "Primary" subtests (Visual Puzzles [VP], Picture Span [PS], and Symbol Search [SS]) that are used in producing the five-factor index scores (two subtests each). There are six "Secondary" subtests (Information [IN], Comprehension [CO], Picture Concepts [PC], Arithmetic [AR], Letter–Number Sequencing [LN], and Cancellation [CN]) that are used for substitution in FSIQ estimation or in estimating newly created (Quantitative Reasoning, Auditory Working Memory, Nonverbal) and previously existing (General Ability, Cognitive Proficiency) Ancillary Index Scores. Like other recent editions of intelligence tests (e.g., WISC-IV, Stanford–Binet Intelligence Scales-Fifth Edition [SB5; Roid, 2003a], Kaufman Assessment Battery for Children-Second Edition [KABC-II; Kaufman & Kaufman, 2004], Reynolds Intellectual Assessment Scales [RIAS; Reynolds & Kamphaus, 2003a], Wide Range Intelligence Test [WRIT; Glutting, Adams, & Sheslow, 2000]), the WISC-V attempted to reflect conceptualizations of intellectual measurement articulated by Spearman (1927), Carroll, Cattell, and Horn (Carroll, 1993, 2003; Cattell & Horn, 1978; Horn, 1991; Horn & Blankson, 2012; Horn & Cattell, 1966), as well as other neuropsychological constructs.

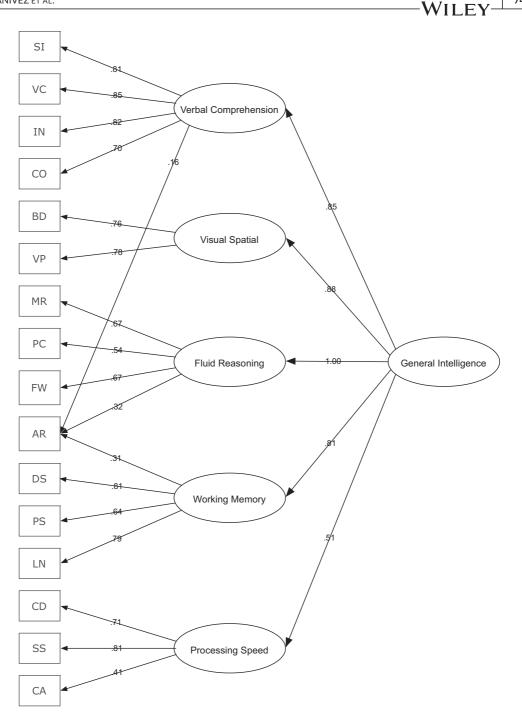
Evidence of WISC-V structural validity reported in the WISC-V *Technical and Interpretive Manual* was based exclusively on confirmatory factor analyses (CFA). A one-factor model served as the baseline and all other models were higher-order models with a general intelligence factor indirectly influencing subtests via full mediation through two through five first-order factors. Table 5.3 in the WISC-V *Technical and Interpretive Manual* illustrates all CFA models tested and Figure 5.10 (reproduced in modified form here as Figure 1) presents the standardized measurement model for the final publisher-preferred, five-factor, higher-order model for WISC-V primary and secondary subtests for the total standardization sample. This model included a higher-order general intelligence dimension with five first-order factors (VC, VS, FR, WM, PS) and the 16 subtest indicators were uniquely associated with one latent first-order factor except for Arithmetic, which was cross-loaded on VC, FR, and WM. This preferred measurement model, however, included a standardized path coefficient of 1.00 between the higher-order general intelligence factor and the FR factor, which indicates that FR may be empirically redundant. This final model was also reported to fit five different age groupings (6-7, 8-9, 10-11, 12-13, 14-16) equally well (Wechsler, 2014b) and a subsequent study by Chen, Zhang, Raiford, Zhu, and Weiss (2015) showed factorial invariance of this final model across gender.

CFA reported in the WISC-V *Technical and Interpretive Manual* contained numerous notable psychometric concerns (Beaujean, 2016; Canivez & Watkins, 2016; Canivez, Watkins, & Dombrowski, 2016, 2017a). Details regarding CFA methods are lacking, such as the absence of explanation for selecting weighted least squares (WLS) estimation rather than maximum likelihood (ML) estimation. Latent constructs (i.e., factors) have no natural scale of measurement, so specification by the analyst is necessary to achieve model identification. The choice of metric can affect unstandardized parameters and may "yield different conclusions regarding the statistical significance of freely estimated parameters" (Brown, 2015, p. 133). Kline (2011) noted that "use of an estimation method other than ML requires explicit justification" (p. 154). WLS is typically used with data that are categorical or nonnormally distributed and may not produce chi-values nor approximate fit indices equivalent to those produced by ML estimation (Yuan & Chan, 2005); neither of which pertains to WISC-V subtest scores (Chen et al., 2015). Thus, the use of WLS is perplexing, and a significant departure from the typical use of ML estimation in CFA of intelligence tests. Further, Beaujean (2016) replicated the WISC-V CFA results reported in Wechsler (2014b), deducing that an effects-coding method was modified and caused degrees of freedom to be understated, which has consequences for fit statistics that rely on degrees of freedom for their computation.

The complex CFA model adopted by the publisher (as a result of including Arithmetic subtest cross-loadings) is also problematic because it abandons the parsimony of simple structure (Thurstone, 1947). Further, the publisher's

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FIGURE 1 Higher-order measurement model with standardized coefficients (adapted from Figure 5.1 [Wechsler, 2014b]), for WISC-V standardization sample (N = 2,200) 16 Subtests. SI = Similarities, VC = Vocabulary, IN = Information, CO = Comprehension, BD = Block Design, VP = Visual Puzzles, MR = Matrix Reasoning, PC = Picture Concepts, FW = Figure Weights, AR = Arithmetic, DS = Digit Span, PS = Picture Span, LN = Letter-Number Sequencing, CD = Coding, SS = Symbol Search, CA = Cancellation. *Wechsler Intelligence Scale for Children, Fifth Edition (WISC-V)*. Copyright © 2014 NCS Pearson, Inc. Reproduced with permission. All rights reserved. *"Wechsler Intelligence Scale for Children"* and *"WISC"* are trademarks, in the US and/or other countries, of Pearson Education, Inc. or its affiliates(s).

preferred model produced a standardized path coefficient of 1.00 between the latent general intelligence (mislabeled in Figure 5.1 [Wechsler, 2014b] as "Full Scale") factor and the Fluid Reasoning factor; indicating *g* and FR were empirically redundant (Le, Schmidt, Harter, & Lauver, 2010). This constitutes a major threat to discriminant validity and indicates that the WISC-V has likely been overfactored (Frazier & Youngstrom, 2007).

Another issue concerning the WISC-V *Technical and Interpretive Manual* was the acknowledgment of the sensitivity of the chi-square test to trivial differences with large samples, but the subsequent use of chi-square difference tests of nested models to identify the preferred five-factor model (Wechsler, 2014b). The same sensitivity to large samples is true for chi-square difference tests (Millsap, 2007), suggesting that the model differences reported in the WISC-V *Technical and Interpretive Manual* might be statistically significant yet trivial. For example, Table 5.4 in Wechsler (2014b, p. 82) reveals that the difference between models 4a and 5a was statistically significant but those two models exhibited identical comparative fit index (CFI) and root mean square error of approximation (RMSEA) values. Likewise, the preferred five-factor higher-order model was significantly different from other five-factor models but all exhibited identical CFI and RMSEA values (e.g., .98 and .04, respectively). Cheung and Rensvold (2002) demonstrated, in the context of factorial invariance, that practical differences independent of sample size and model complexity could be identified by Δ CFI > .01; this condition was not met when moving from a four- to a five-factor solution.

Another criticism of WISC-V CFA reported in the WISC-V *Technical and Interpretive Manual* is that there was a failure to test rival bifactor measurement models against the higher-order measurement models. Bifactor models have several benefits over higher-order models (Canivez, 2016; Reise, 2012), have been found to fit data from other Wechsler scales (viz., Canivez, 2014a; Canivez, Watkins, Good, James, & James, 2017b; Gignac & Watkins, 2013; Lecerf & Canivez, 2017; Nelson, Canivez, & Watkins, 2013; Watkins, 2010; Watkins & Beaujean, 2014; Watkins, Canivez, James, James, James, & Good, 2013), and have been recommended for cognitive tests (Brunner, Nagy, & Wilhelm, 2012; Canivez, 2016; Cucina & Byle, 2017; Cucina & Howardson, 2017; Gignac, 2005, 2006; Morin, Arens, Tran, & Caci, 2016). A higher-order structural model posits general intelligence as a superordinate construct that is fully mediated by the lower-order factors and indirectly influences the subtest indicators. In contrast, the bifactor model hypothesizes general intelligence as a breadth factor with direct influence on subtests in addition to direct influence on subtests by group factors (Canivez, 2016; Gignac, 2008). The bifactor model appears to be more consistent with Spearman's (1927) conceptualization of intelligence and a more conceptually parsimonious explanation than the higher-order model (Canivez, 2016; Cucina & Howardson, 2017; Gignac, 2006). Further, the structure of intelligence described by Carroll (1993) is better represented by the bifactor model (Beaujean, 2015a; Cucina & Howardson, 2017).

Another significant problem is that the publisher did not provide decomposed variance estimates to disclose how much subtest variance is a result of the hierarchical *g* factor and how much is a result of the lower-order group factors. This makes it difficult for clinicians and researchers to judge the adequacy of the group factors (VC, VS, FR, WM, PS) based on how much unique variance the group factors capture when purged of the effects of general intelligence (Reise, Moore, & Haviland, 2010), although this could be computed by hand from the model. As noted by DeVellis (2017), relying on statistical fit alone "may obscure the fact that some statistically significant factors may account for uninterestingly small proportions of variance" (p. 199).

Also missing from the WISC-V *Technical and Interpretive Manual* are model-based reliability estimates (omega). It has long been argued that classical estimates of reliability are biased (Raykov, 1997). Model-based estimates, such as omega-hierarchical (ω_H) and omega-hierarchical subscales (ω_{HS}), have been recommended as superior metrics for determining construct-based reliability (Rodriguez, Reise, & Haviland, 2016; Watkins, 2017). These problems were highlighted in several reviews and critiques of Wechsler scales including the WAIS-IV, WPPSI-IV, and WISC-IV (Canivez, 2010, 2014b; Canivez & Kush, 2013); however, omega estimates are notably absent from the WISC-V *Technical and Interpretive Manual*.

Although Chen et al. (2015) used ML estimation in their WISC-V invariance study, their chosen model replicated the standardized path coefficient of 1.0 from the FSIQ to FR and cross-loading of Arithmetic on three first-order factors. Further, there was no consideration of rival bifactor models nor was there decomposition of subtest variance or estimation of latent factor reliabilities to understand the relative contributions of the higher-order versus first-order factors. Reynolds and Keith (2017) examined WISC-V invariance across standardization sample age groups, but the

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model examined for invariance was an oblique five-factor model rather than the bifactor or higher-order model, which thus ignored general intelligence and its unmodeled variance.

Reynolds and Keith (2017) also explored numerous post hoc modifications for first-order models with five factors and then for both higher-order and bifactor models with five group factors in an attempt to better understand WISC-V measurement. Whereas such explorations are possible, they may capitalize on chance and it could be argued that such exploratory interest might be better served by using EFA (Carroll, 1995) or exploratory structural equation modeling (Asparouhov & Muthén, 2009). Their final best fitting WISC-V higher-order model was different from the publisher-preferred model in that Arithmetic was given a direct loading from general intelligence and a "cross-loading" on Working Memory, but Reynolds and Keith also added correlated disturbance of Visual Spatial and Fluid Reasoning group factors yet the model still produced a standardized path coefficient of .97 from general intelligence to Fluid Reasoning. Further, decomposed variance estimates of their higher-order model showed that the WISC-V subtests primarily reflected general intelligence variance with small portions of variance unique to the group factors (except for the Processing Speed subtests). Their best WISC-V bifactor model also added a covariance estimate between Visual Spatial and Fluid Reasoning (.62), which appears necessary to salvage five group factors. Watkins, Dombrowski, and Canivez (2017) also tested a similar bifactor model with the Canadian WISC-V (WISC-V^{CDN}), but this bifactor model with five group factors and VS-FR covariance estimate was not superior to the bifactor model with four group factors.

A final criticism is that the WISC-V *Technical and Interpretive Manual* includes explicit preference for CFA over EFA methods rather than taking advantage of each method's unique strengths. EFA and CFA are complementary procedures, so greater confidence in the latent factor structure is achieved when EFA and CFA are in agreement (Gorsuch, 1983). Carroll (1995) and Reise (2012) both noted that EFA procedures are especially useful in suggesting possible models to be tested in CFA, and Carroll (1998) suggested that "CFA should derive its initial hypotheses from EFA results, rather than starting from scratch or from a priori hypotheses...[and] CFA analyses should be done to check my EFA analyses" (p. 8). The deletion of Word Reasoning and Picture Completion subtests; the addition of Visual Puzzles, Figure Weights, and Picture Span subtests; and the inclusion of new or revised items across all WISC-V subtests suggests that relationships among retained and new subtests might result in associations and latent structure unanticipated by *a priori* conceptualizations (Beaujean, 2015b; Strauss, Spreen, & Hunter, 2000).

Intelligence test factor structure research using EFA procedures have consistently produced serious and substantial challenges to the optimistic conclusions from CFA-based latent structures reported in test technical manuals. DIStefano and Dombrowski (2006) and Canivez (2008), using data from the SB5 (Roid, 2003a) standardization sample, obtained markedly different results for the SB5 than CFA results presented in the technical manual (Roid, 2003b) and concluded that the SB5 essentially measured one dimension (g). Three studies of the WISC-IV (Wechsler, 2003) and two studies of the Wechsler Adult Intelligence Scale-Fourth Edition (WAIS-IV; Wechsler, 2008) using EFA (Bodin, Pardini, Burns, & Stevens, 2009; Canivez & Watkins, 2010a, 2010b; Watkins, 2006; Watkins, Wilson, Kotz, Carbone, & Babula, 2006) indicated that most variance was associated with general intelligence (substantially lesser amounts at the factor level) and suggested that interpretation of both the WISC-IV and WAIS-IV should focus on the global FSIQ score because it accounts for most of the common variance and additional research showing FSIQ superiority in predictive validity with little to no meaningful incremental prediction by the factor index scores (Canivez, 2014a; Canivez, Watkins, James, James, & Good, 2014; Glutting, Watkins, Konold, & McDermott, 2006; Glutting, Youngstrom, Ward, Ward, & Hale, 1997; Nelson et al., 2013). The limited unique variance captured by the first-order factors is likely responsible for the poor incremental predictive validity of the WISC-IV and WAIS-IV factor index scores. EFA studies of other intelligence tests such as RIAS (Reynolds & Kamphaus, 2003a) have also indicated that fundamental measurement is primarily that of general intelligence (Dombrowski, Watkins, & Brogan, 2009; Nelson & Canivez, 2012; Nelson, Canivez, Lindstrom, & Hatt, 2007), which was by design its primary goal (Reynolds & Kamphaus, 2003b). Similar findings were obtained with a joint examination of the Wide Range Intelligence Test (WRIT; Glutting et al., 2000) and Wechsler Abbreviated Scale of Intelligence (WASI; Psychological Corporation, 1999) where most subtest variability was associated with a hierarchical general intelligence dimension and smaller portions of variance were apportioned to the first-order factors; supporting primary interpretations of the FSIQ and general intelligence test (Canivez, Konold, Collins, & Wilson, 2009).

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Independent assessment of the WISC-V using EFA with the total standardization sample (*n* = 2,200) was reported by Canivez et al. (2016) and no evidence was found for five factors. The intended separation of Visual Spatial and Fluid Reasoning dimensions was not supported as extracting five factors resulted in the fifth factor including only one subtest (Figure Weights) with a salient factor pattern coefficient, and Picture Concepts failed to saliently load on any factor. Extraction of four factors produced a structure very similar to the WISC-IV with Visual Spatial and Fluid Reasoning collapsing into one Perceptual Reasoning factor. Schmid and Leiman (SL, 1957) orthogonalization found the *g* factor accounted for large portions of total and common variance and provided little evidence for interpretation of the lower-ordered factors. The omega-hierarchical coefficient of the *g* factor was large while the omega-hierarchical subscale coefficients for the four lower-order factors were too low for confident interpretation, except perhaps for the Processing Speed factor. Canivez et al. (2017a) replicated WISC-V EFA results with CFA using maximum likelihood estimation, further challenging results in the WISC-IV, WAIS-IV), and other tests of intelligence, they were obtained with the entire standardization sample and it is possible that different structures might be observed within different age ranges; therefore, Canivez et al. recommended examination of WISC-V structure with different age groups using similar EFA procedures.

Following that recommendation, the present study investigated the factor structure of the WISC-V with four age groups (6–8, 9–11, 12–14, 15–16 years) from the WISC-V standardization sample using EFA followed by a Schmid-Leiman orthogonalization, the same procedures used by Canivez et al. (2016) when investigating the WISC-V total sample to allow for direct comparison of results. The EFA-based SL orthogonalization procedure produces an approximate bifactor solution that is a reparameterization of the higher-order structure and contains proportionality constraints (Yung, Thissen, & McLeod, 1999), but is the dominant exploratory approach to assessing bifactor structure (Reise, 2012). Also, the present study used identical EFA methods to Canivez et al. (2016), which allows for direct comparison of results of the more homogeneous age groups to the full standardization sample results but does not directly test the factorial invariance of the WISC-V across age/development. The primary research questions included (1) how many WISC-V factors should be extracted and retained in each age subgroup; (2) how are subtests associated with the latent factors; (3) was there evidence for the publisher's claim of five first-order factors; and (4) what proportion of variance was a result of general intelligence versus the first-order group ability factors following a Schmid–Leiman orthogonalization?

1 | METHOD

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1.1 | Participants

Participants were members of the WISC-V standardization sample and included a total of 2,200 individuals ranging in age from 6 to 16 years. Demographic characteristics are provided in detail in the WISC-V *Technical and Interpretive Manual* (Wechsler, 2014b). Stratified proportional sampling was used across variables of age, sex, race/ethnicity, parental education level, and geographic region in obtaining the standardization sample. Education level was a proxy for socioeconomic status where accurate information about income is often difficult to obtain. Examination of tables in the *Technical and Interpretive Manual* revealed a close match to the U.S. census across stratification variables.

1.2 | Instrument

The WISC-V is an individual test of general intelligence for children ages 6–16 years and originated with the first WISC (Wechsler, 1949). Consistent with Wechsler's definition of intelligence (i.e., "global capacity;" Wechsler, 1939, p. 229), the WISC-V includes numerous subtests that provide estimates of general intelligence but also are combined to measure group factors. WISC-V measurement of intelligence continues to include narrow ability subtests (16), broad group factors (5), and general intelligence.

Organization and subtest administration order of the WISC-V reflect a new four-level organization. The FSIQ is composed of seven primary subtests across the five domains (VC, VS, FR, WM, PS), but if one of the FSIQ subtests is invalid or missing, that subtest may be substituted by a secondary subtest from within the same domain. Only one substitution is allowed. The Primary Index Scale level is composed of 10 WISC-V subtests (primary subtests) and are used to estimate the five WISC-V factor index scores (VCI, VSI, FRI, WMI, PSI). No substitutions are allowed for the Primary Index Scales. Complementary subtests are not intelligence subtests and so were not included in the present analyses.

1.3 | Procedure

NCS Pearson denied without rationale the request for WISC-V standardization sample raw data to conduct these (and other) independent analyses. Absent raw data, WISC-V subtest scaled score correlation matrices for each age group (n = 200) in the standardization sample were obtained from the WISC-V *Technical and Interpretive Manual Supplement* (Wechsler, 2014c) and combined by averaging correlations through Fisher transformations. Four correlation matrices (16 primary and secondary intelligence subtests) were created to represent four broad age subgroups (ages 6–8 [n = 600], 9–11 [n = 600], 12–14 [n = 600], and 15–16 [n = 400] years). The sample size of single age groups (n = 200) would be too small for stable results (Goldberg & Velicer, 2006; Mundfrom & Shaw, 2005). In contrast, these four age groups should allow developmental differences to emerge while still providing robust factor recovery.

1.4 Analyses

Principal axis exploratory factor analyses (Fabrigar, Wegener, MacCallum, & Strahan, 1999) were used to analyze the combined WISC-V standardization sample correlation matrices from the four age groups using SPSS 21 for Macintosh OSX. Principal axis EFA was selected for comparison to Canivez et al. (2016) and because it "frequently outperformed ML in the recovery of relatively weak common factors" (Briggs & MacCallum, 2003, p. 49). Multiple criteria (Gorsuch, 1983) were examined to determine the number of factors to retain and included eigenvalues >1 (Kaiser, 1960), the scree test (Cattell, 1966), standard error of scree (SE_{scree} ; Zoski & Jurs, 1996), Horn's parallel analysis (HPA; Horn, 1965), and minimum average partials (MAP; Velicer, 1976). The scree test is a subjective criterion so the SE_{scree} as programmed by Watkins (2007) was used because it was reportedly the most accurate objective scree method (Nasser, Benson, & Wisenbaker, 2002).

HPA and MAP were included because they are considered more accurate and less likely to overfactor (Frazier & Youngstrom, 2007; Velicer, Eaton, & Fava, 2000; Zwick & Velicer, 1986), although in the presence of a strong general factor HPA tends to underfactor (Crawford et al., 2010). HPA indicates meaningful factors when eigenvalues from the WISC-V standardization sample data were larger than eigenvalues produced by random data containing the same number of participants and factors. Random data eigenvalues for HPA were produced using the Monte Carlo principal components analysis for the Parallel Analysis computer program (Watkins, 2000) with 100 replications to provide stable eigenvalue estimates. Retained factors were subjected to promax (oblique) rotation (k = 4; Gorsuch, 1983). Setting k to 4 produced greater hyperplane count compared to k = 2 with the present data. Salient factor pattern coefficients were defined as those \geq .30 (Child, 2006). Factor solutions were examined for interpretability and theoretical plausibility (Fabrigar et al., 1999) with the empirical requirement that each factor should be marked by two or more salient loadings and no salient cross-loadings (Gorsuch, 1983). Subtest g loadings (first unrotated factor coefficients) were evaluated based on Kaufman's (1994) criteria (\geq .70 = good, .50 – .69 = fair, <.50 = poor).

Cognitive ability subtest scores reflect combinations of both first-order and second-order factor variance and, because of this, Carroll (1993, 1995, 1997, 2003) argued that variance from the higher-order factor must be extracted first to residualize the lower-order factors, leaving them orthogonal to the higher-order factor. The Schmid and Leiman (1957) procedure has been recommended as the statistical method to accomplish this residualization (Carroll, 1993, 1995, 1997, 2003; Carretta & Ree, 2001; Gustafsson & Snow, 1997; McClain, 1996; Ree, Carretta, & Green, 2003; Thompson, 2004). It is a reparameterization of a higher-order model and an approximate bifactor solution (Reise,

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2012). Accordingly, first-order factors were orthogonalized by removing all variance associated with the secondorder dimension using the Schmid and Leiman (1957) procedure as programmed in the MacOrtho computer program (Watkins, 2004). This transforms "an oblique factor analysis solution containing a hierarchy of higher-order factors into an orthogonal solution which not only preserves the desired interpretation characteristics of the oblique solution, but also discloses the hierarchical structuring of the variables" (Schmid & Leiman, 1957, p. 53).

The Schmid–Leiman (SL) orthogonalization procedure may be constrained by proportionality (Yung et al., 1999) and may be problematic with nonzero cross-loadings (Reise, 2012). Reise also noted two additional and more recent alternative exploratory bifactor methods that do not include proportionality constraints: analytic bifactor (Jennrich & Bentler, 2011) and target bifactor (Reise, Moore, & Maydeu-Olivares, 2011). However, the present application of the SL orthogonalization procedure was selected for direct comparison to WISC-V results obtained by Canivez et al. (2016) with the total WISC-V standardization sample and comparisons to the numerous studies of SL application with Wechsler scales (Canivez & Watkins, 2010a; 2010b; Golay & Lecerf, 2011; Lecerf & Canivez, 2017; Watkins, 2006; Watkins et al., 2017) and with other intelligence tests (Canivez, 2008, 2011; Canivez & McGill, 2016; Canivez et al., 2009; Dombrowski, 2013, 2014a, 2014b; Dombrowski & Watkins, 2013; Dombrowski et al., 2009; Dombrowski, McGill, & Canivez, 2017b; Nelson & Canivez, 2012; Nelson et al., 2007; Strickland, Watkins, & Caterino, 2015). For convenience, this method is labeled the SL bifactor (Reise, 2012).

Omega-hierarchical and omega-hierarchical subscale coefficients (Reise, 2012; Rodriguez et al., 2016) were estimated as model-based reliability estimates of the latent factors (Gignac & Watkins, 2013). Chen, Hayes, Carver, Laurenceau, & Zhang (2012) noted that "for multidimensional constructs, the alpha coefficient is complexly determined, and McDonald's (1999) omega-hierarchical (ω_{H}) provides a better estimate for the composite score and thus should be used" (p. 228). These same problems are inherent with other internal consistency estimates such as split-half or KR-20. Omega-hierarchical (ω_{H}) is the model-based reliability estimate for the hierarchical general intelligence factor independent of the variance of group factors. Omega-hierarchical subscale (ω_{HS}) is the model-based reliability estimate of a group factor with all other group and general factors removed (Reise, 2012). Omega estimates (ω_{H} and ω_{HS}) may be obtained from EFA SL bifactor solutions and were produced using the Omega program (Watkins, 2013), which was based on the tutorial by Brunner et al. (2012) and the work of Zinbarg, Revelle, Yovel, and Li (2005) and Zinbarg, Yovel, Revelle, and McDonald (2006). Omega-hierarchical coefficients should at a minimum exceed .50, but .75 would be preferred (Reise, 2012; Reise, Bonifay, & Haviland, 2013).

2 | RESULTS

2.1 | Factor extraction criteria comparisons

Figures A1–A4 (Appendix A in online supplemental materials) show scree plots from HPA for the four age groups. Table 1 summarizes results from the multiple factor extraction criteria (eigenvalues > 1, scree test, standard error of scree, HPA, MAP, theory) for determining the number factors to extract and retain. As shown in Table 1, only the publisher recommended/theory justified extraction of five factors. All other criteria across the four age groups mostly recommended extraction of only one to three factors.

2.2 | Five-factor exploratory and hierarchical analyses

It has been suggested that it is better to overextract than underextract (Gorsuch, 1997; Wood, Tataryn, & Gorsuch, 1996) so EFA began with extracting five factors to examine subtest associations based on the publisher's suggested structure and to allow examination of the performance of smaller factors. Tables B1 through B8 (Appendix B in online supplemental materials) show exploratory factor analyses results (odd-numbered Tables B1-B7) and exploratory SL bifactor model results (even-numbered Tables B2-B8) for the four age groups. In each of the four age groups, extraction of five factors produced psychometrically inadequate results as the fifth factor included only one salient factor

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 TABLE 1
 Number of WISC-V factors suggested for extraction across five different criteria by age group

			WISC-V Age Grou	ps	
Extraction Criterion	6-8	9-11	12-14	15-16	6-16
Eigenvalue >1	3	3	2	3	2
Scree test (visually examined)	2	2	2	2	2
Standard error of scree (SE _{scree})	2	4	3	3	3
Horn's parallel analysis (HPA)	2	2	2	2	2
Minimum average partials (MAP)	1	1	1	2	1
Prior Wechsler structure/theory	4	4	4	4	4
Publisher (theory) proposed	5	5	5	5	5

pattern coefficient (Cancellation [ages 6–8], Arithmetic [ages 9–11], Picture Concepts [ages 12–14 and 15–16]) and factors cannot be defined by only one indicator (see odd-numbered Tables B1–B7 in online supplemental materials). Further, contrary to the publisher's desire to split the Perceptual Reasoning factor into separate Visual Spatial (Block Design, Visual Puzzles) and Fluid Reasoning (Matrix Reasoning, Figure Weights) factors, extraction of five factors still resulted in Block Design, Visual Puzzles, Matrix Reasoning, and Figure Weights having salient factor pattern loadings on the same (Perceptual Reasoning) factor. Exploratory SL bifactor model results (see even-numbered Tables B2–B8 in online supplemental materials) also show the dominance of the general intelligence factor for all subtests except Coding, Symbol Search, and Cancellation (Processing Speed subtests), known to be poor indicators of general intelligence.

2.3 | Four-factor exploratory and hierarchical analyses

2.3.1 | Ages 6-8 first-order EFA: Four-factor extraction

Table 2 shows results of four-factor extraction with promax rotation for the 6- to 8-year-olds. The *g* loadings ranged from .175 (Cancellation) to .746 (Information) and all were within the fair-to-good range (except Coding and Cancellation). Picture Concepts failed to exhibit salient pattern loadings on any group factor. Table 2 shows robust Verbal Comprehension (Similarities, Vocabulary, Information, Comprehension), Working Memory (Digit Span, Picture Span, Letter-Number Sequencing), Perceptual Reasoning (Block Design, Visual Puzzles, Matrix Reasoning, Figure Weights), and Processing Speed (Coding, Symbol Search, Cancellation) factors with theoretically consistent subtest associations. Picture Concepts, a fair indicator of general intelligence, was not adequately associated with any of the four group factors; although its highest pattern coefficient was on the Perceptual Reasoning factor. There were no subtests with salient cross-loadings. The moderate-to-high factor correlations shown in Table 2 (.372 to .710) imply a higher-order or hierarchical structure that required explication (Gorsuch, 1983) and the Schmid–Leiman procedure was applied to better understand variance apportionment among general and group factors.

2.3.2 Ages 6-8 SL bifactor analyses: Four first-order factors

Results for the Schmid and Leiman orthogonalization of the higher-order factor analysis are shown in Table 3. All subtests were properly associated (higher residual variance) with their theoretically proposed factor after removing *g* variance. The *g* factor accounted for 33.2% of the total variance and 66.4% of the common variance.

The general factor also accounted for between 2.3% (Cancellation) and 49.7% (Digit Span) of individual subtest variability. At the first-order level, VC accounted for an additional 4.6% of the total variance and 9.1% of the common variance, WM accounted for an additional 3.2% of the total variance and 6.5% of the common variance, PR accounted for an additional 3.3% of the total variance and 6.6% of the common variance, and PS accounted for an additional 5.7% of the total variance and 11.4% of the common variance. The general and group factors combined to measure 50.0% of the variance in WISC-V scores, resulting in 50.0% unique variance (combination of specific and error variance).

	General ^a	F1: Verbal Comprehension	mprehension	F2: Working Memory	g Memory	F3: Perceptual Reasoning	al Reasoning	F4: Processing Speed	ing Speed	
WISC-V subtest	S	Ρ	S	Р	S	٩	S	Р	S	h^2
Similarities	.721	.643	.759	.169	.620	001	.560	009	.318	.589
Vocabulary	.703	.885	.828	086	.530	.055	.577	095	.213	669.
Information	.746	.766	.806	005	.598	.004	.585	.109	.393	.660
Comprehension	.626	.571	.660	.150	.541	034	.475	.018	.293	.447
Block Design	.611	.026	.491	021	.483	.608	.675	.145	.406	.472
Visual Puzzles	.667	013	.548	.001	.521	.827	.795	055	.297	.635
Matrix Reasoning	.704	.035	.572	.253	.628	.487	.700	.042	.397	.536
Figure Weights	.550	.093	.473	.197	.487	.365	.545	042	.253	.328
Picture Concepts	.537	.098	.452	.190	.481	.281	.504	.062	.318	.297
Arithmetic	.680	.121	.565	.629	.718	023	.501	.035	.397	.523
Digit Span	.732	012	.576	.788	.796	.075	.570	064	.374	.639
Picture Span	.507	029	.392	.500	.540	.103	.413	015	.278	.297
Letter-Number Sequencing	.692	.051	.561	.788	.766	035	.503	065	.348	.592
Coding	.368	089	.197	.101	.361	078	.239	.721	707.	.510
Symbol Search	.517	.036	.353	.073	.471	005	.380	.722	.771	.602
Cancellation	.175	.058	.124	279	.083	.161	.189	.408	.354	.159
Eigenvalue		6.50		1.44		1.04		.93		
% Variance		37.73		5.94		3.23		2.99		
Promax-based factor correlations		F1: VC		F2: WM		F3: PR		F4: PS		
F1: Verbal Comprehension (VC)		I								
F2: Working Memory (WM)		.710		I						
F3: Perceptual Reasoning (PR)		.701		.674		I				
F4: Processing Speed (PS)		.372		.520		.431				

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TABLE 2 We chaler Intelligence Scale for Children-Fifth Edition (WISC-V) exploratory factor analysis: Four oblique factor solution for the standardization sample 6- to 8-year-olds

	Gen	General	F1: Verbal Comprehension	nprehension	F2: Working Memory	g Memory	F3: Perceptu	F3: Perceptual Reasoning	F4: Processing Speed	sing Speed		
WISC-V subtest	q	S ²	q	S ²	q	S ²	q	S ²	q	S ²	h²	u ²
Similarities	.665	.442	.374	.140	.082	.007	001	000	008	000	.589	.411
Vocabulary	.640	.410	.515	.265	042	.002	.032	.001	080	900.	.684	.316
Information	.680	.462	.445	.198	002	000	.002	000	.092	.008	669.	.331
Comprehension	.578	.334	.332	.110	.073	.005	020	000	.015	000	.450	.550
Block Design	.575	.331	.015	000	010	000	.353	.125	.123	.015	.471	.529
Visual Puzzles	.634	.402	008	000	000	000	.481	.231	047	.002	.636	.364
Matrix Reasoning	.668	.446	.020	000	.123	.015	.283	.080	.036	.001	.543	.457
Figure Weights	.522	.272	.054	.003	.096	600.	.212	.045	036	.001	.331	699.
Picture Concepts	.507	.257	.057	.003	.093	600.	.163	.027	.053	.003	.298	.702
Arithmetic	.647	.419	.070	.005	.307	.094	013	000	.030	.001	.519	.481
Digit Span	.705	.497	007	000	.384	.147	.044	.002	054	.003	.649	.351
Picture Span	.489	.239	017	000	.244	090.	090.	.004	013	000	.303	697.
Letter-Number Sequencing	666.	.444	.030	.001	.384	.147	020	000	055	.003	.595	.405
Coding	.335	.112	052	.003	.049	.002	045	.002	.611	.373	.493	.507
Symbol Search	.472	.223	.021	000	.036	.001	003	000	.612	.375	.599	.401
Cancellation	.151	.023	.034	.001	136	.018	.094	600.	.346	.120	.171	.829
Total variance		.332		.046		.032		.033		.057	.500	.500
Common variance		.664		.091		.065		.066		.114		
	$\omega_{\rm H} =$	$\omega_{\rm H}$ = .821	$\omega_{\rm HS} = .253$.253	$\omega_{HS} = .174$.174	$\omega_{HS} =$	=.165	$\omega_{\rm HS} = .478$: 478		

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TABLE 3 Sources of variance in the Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) for the standardization sample 6- to 8-year-olds (n = 600) according to an

Table 3 also shows $\omega_{\rm H}$ and $\omega_{\rm HS}$ that were estimated based on the SL results. The $\omega_{\rm H}$ coefficient for general intelligence (.821) was high and sufficient for scale interpretation; however, the $\omega_{\rm HS}$ coefficients for the four group factors (VC, WM, PR, PS) were considerably lower (.174–.478). Thus, for the four group factors, with the possible exception of PS, unit-weighted composite scores based on these indicators would likely possess too little true score variance for clinical interpretation (Reise, 2012; Reise et al., 2013) for the 6- to 8-year-old age group.

2.3.3 Ages 9–11 first-order EFA: Four-factor extraction

Table 4 shows results of four-factor extraction with promax rotation for 9- to 11-year-olds. The *g* loadings ranged from .226 (Cancellation) to .803 (Vocabulary) and all were within the fair-to-good range (except Coding, Symbol Search, Cancellation). Picture Concepts failed to exhibit salient pattern loadings on any group factor. Table 4 shows robust Verbal Comprehension (Similarities, Vocabulary, Information, Comprehension), Perceptual Reasoning (Block Design, Visual Puzzles, Matrix Reasoning, Figure Weights), Working Memory (Arithmetic, Digit Span, Picture Span, Letter-Number Sequencing), and Processing Speed (Coding, Symbol Search, Cancellation) factors with theoretically consistent subtest associations. Picture Concepts was again a fair indicator of general intelligence but was not adequately associated with any of the four group factors; although its highest pattern coefficient was on Perceptual Reasoning. There were no subtests with salient cross-loadings. The moderate-to-high factor correlations shown in Table 4 (.392 to .724) imply a higher-order or hierarchical structure that requires explication (Gorsuch, 1983) and the Schmid-Leiman procedure was applied to better understand variance apportionment among general and group factors.

2.3.4 Ages 9–11 SL bifactor analyses: Four first-order factors

Results for the Schmid and Leiman orthogonalization of the higher-order factor analysis are shown in Table 5. All subtests were properly associated (higher residual variance) with their theoretically proposed factor after removing *g* variance except Picture Concepts, which had equivalent residual loadings with Perceptual Reasoning and Verbal Comprehension. The *g* factor accounted for 33.6% of the total variance and 64.1% of the common variance.

The general factor also accounted for between 4.0% (Cancellation) and 52.4% (Vocabulary) of individual subtest variability. At the first-order level, VC accounted for an additional 5.4% of the total variance and 10.4% of the common variance, PR accounted for an additional 3.3% of the total variance and 6.4% of the common variance, WM accounted for an additional 3.6% of the total variance and 6.9% of the common variance, and PS accounted for an additional 6.4% of the total variance and 12.3% of the common variance. The general and group factors combined to measure 52.4% of the variance in WISC-V scores resulting in 47.6% unique variance (combination of specific and error variance).

Also presented in Table 5 are $\omega_{\rm H}$ and $\omega_{\rm HS}$ coefficients that were estimated based on the SL results. The $\omega_{\rm H}$ coefficient for general intelligence (.817) was high and sufficient for scale interpretation; however, the $\omega_{\rm HS}$ coefficients for the four group factors (VC, WM, PR, PS) were considerably lower (.064–.517). Thus, unit-weighted composite scores for the four group factors, with the possible exception of PS, would likely possess too little true-score variance for clinical interpretation (Reise, 2012; Reise et al., 2013) for the 9- to 11-year-old age group.

2.3.5 Ages 12–14 first-order EFA: Four-factor extraction

Table 6 shows results of four-factor extraction with promax rotation for 12- to 14-year-olds. The g loadings ranged from .252 (Cancellation) to .806 (Vocabulary) and all were within the fair-to-good range (except Coding, Symbol Search, Cancellation). Picture Concepts and Arithmetic had salient factor pattern coefficients on the Verbal Comprehension factor but no other factors. Table 6 shows robust Verbal Comprehension (Similarities, Vocabulary, Information, Comprehension), Working Memory (Digit Span, Picture Span, Letter–Number Sequencing), Perceptual Reasoning (Block Design, Visual Puzzles, Matrix Reasoning, Figure Weights), and Processing Speed (Coding, Symbol Search, Cancellation) factors with theoretically consistent subtest associations. Oddly, Picture Concepts and Arithmetic migrated away from their theoretically consistent factors to the Verbal Comprehension factor. No salient cross-loadings were observed. The moderate-to-high factor correlations presented in Table 6 (.399 to .732) imply a higher-order or hierarchical

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	General ^a	F1: Verbal Co	F1: Verbal Comprehension	F2: Perceptu	F2: Perceptual Reasoning	F3: Worki	F3: Working Memory	F4: Processing Speed	sing Speed	
WISC-V subtest	S	٩	S	Р	S	Р	S	Р	s	h^2
Similarities	.744	.805	.815	.034	.602	035	.539	.022	.336	.665
Vocabulary	.803	.872	.883	.043	.652	014	.588	027	.327	.781
Information	.749	.807	.823	.035	.608	.005	.555	032	.301	.678
Comprehension	.646	.668	.694	064	.495	.087	.510	.038	.313	.488
Block Design	.683	031	.541	.780	.770	053	.511	.110	.417	.603
Visual Puzzles	.653	.119	.568	.673	.718	024	.488	056	.276	.524
Matrix Reasoning	.532	021	.421	.475	.562	.114	.450	.056	.311	.326
Figure Weights	.627	.020	.522	.715	.710	.027	.488	084	.252	.510
Picture Concepts	.507	.239	.470	.262	.479	.031	.395	.052	.276	.265
Arithmetic	.712	.165	.608	.220	.630	.387	.677	090.	.408	.526
Digit Span	.647	080	.491	.084	.541	.802	.770	075	.315	.601
Picture Span	.562	.031	.446	059	.429	.666	.659	.025	.330	.436
Letter-Number Sequencing	.677	.100	.560	026	.532	.729	.766	028	.350	.591
Coding	.461	044	.302	023	.345	.124	.421	.714	.746	.563
Symbol Search	.455	.040	.317	017	.340	038	.362	.801	.792	.628
Cancellation	.226	.008	.151	.051	.185	119	.148	.474	.443	.203
Eigenvalue		6.61		1.52		1.07		.99		
% Variance		38.59		6.60		3.92		3.31		
Promax-Based Factor Correlations		F1: VC		F2: PR		F3: WM		F4: PS		
F1: Verbal Comprehension (VC)		I								
F2: Perceptual Reasoning (PR)		.724		I						
F3: Working Memory (WM)		.672		.682		I				
E4: Processing Speed (PS)		392		441		479		ı		

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TABLE 5	Sources of variance in the Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) for the standardization sample 9- to 11-year-olds (n = 600) according to an
exploratory	ry SL bifactor model with four first-order factors

WISC-V subtest b					-	D		-			
	S ²	q	S ²	q	S ²	q	S ²	q	S ²	h²	u ²
Similarities .670	.449	.463	.214	.018	000.	020	000	.019	000	.664	.336
Vocabulary .724	.524	.502	.252	.023	.001	008	000	023	.001	777.	.223
Information .677	.458	.464	.215	.018	000.	.003	000	027	.001	.675	.325
Comprehension .584	.341	.384	.147	034	.001	.049	.002	.032	.001	.493	.507
Block Design .652	.425	018	000	.409	.167	030	.001	.094	.009	.602	.398
Visual Puzzles .621	.386	.068	.005	.353	.125	013	000	048	.002	.517	.483
Matrix Reasoning .511	.261	012	000	.249	.062	.064	.004	.048	.002	.330	.670
Figure Weights .603	.364	.012	000	.375	.141	.015	000	071	.005	.510	.490
Picture Concepts .471	.222	.138	.019	.137	.019	.017	000	.044	.002	.262	.738
Arithmetic .674	.454	.095	.009	.115	.013	.218	.048	.051	.003	.527	.473
Digit Span .630	.397	046	.002	.044	.002	.451	.203	064	.004	608.	.392
Picture Span .539	.291	.018	000	031	.001	.374	.140	.021	000	.432	.568
Letter-Number Sequencing .648	.420	.058	.003	014	000.	.410	.168	024	.001	.592	.408
Coding .423	.179	025	.001	012	000	.070	.005	608.	.370	.554	.446
Symbol Search .408	.166	.023	.001	009	000.	021	000	.682	.465	.633	.367
Cancellation .201	.040	.005	.000	.027	.001	067	.004	.403	.162	.208	.792
Total Variance	.336		.054		.033		.036		.064	.524	.476
Common Variance	.641		.104		.064		.069		.123		
^Φ H	$\omega_{\rm H} = .817$	$\omega_{HS} = .280$:.280	$\omega_{HS} = .174$.174	$\omega_{\rm HS} = .207$	207	$\omega_{HS} = .$.517		

Note b = loading of subtest on factor, $S^2 = \text{variance explained}$, $h^2 = \text{communality}$, $u^2 = \text{uniqueness}$, $\omega_{\text{H}} = \text{omega-hierarchical}$, $\omega_{\text{HS}} = \text{omega-hierarchical subscale}$. Bold type shows coefficients and variance estimates consistent with the theoretically proposed factor. Italic type shows coefficients and variance estimates associated with an alternate factor (where cross-loading b was larger than for the theoretically assigned factor). For theoretical and practical reasons, Picture Concepts was assigned to Perceptual Reasoning for omega subscale estimation.

meter r.t. versus comportance r.t. versus comportance		- -						-	Ľ	-	
6 7		General	F1: Verbal Col	mprenension	FZ: Workin	ig Memory	F3: Perceptu	ial Keasoning	F4: Process	sing speed	
771 770 833 013 6.20 074 6.46 -0.02 344 806 919 879 -011 6.23 -034 6.24 -0.03 326 647 807 828 -015 550 -134 500 364 363 647 048 553 -024 512 512 513 514 514 514 514 514 514 514 514 514 514 515 514 515 514 515 514 515 514 515 514 515 514 515 514 515 514 515 <t< th=""><th>WISC-V Subtest</th><th>S</th><th>٩</th><th>S</th><th>Р</th><th>S</th><th>Р</th><th>S</th><th>Р</th><th>S</th><th>h²</th></t<>	WISC-V Subtest	S	٩	S	Р	S	Р	S	Р	S	h²
806 919 879 -0.01 6.23 -0.34 6.24 -0.16 328 771 807 828 -0.34 581 1.03 641 -0.06 363 667 794 746 0.16 550 -1.34 500 0.96 363 702 -0.01 593 -0.73 522 889 836 -0.03 343 702 -0.01 533 -0.73 522 889 836 -0.03 343 703 512 0.47 544 702 0.03 343 734 523 546 703 646 703 343 731 0.32 547 702 546 703 343 744 0.13 640 703 646 703 343 744 134 713 743 743 743 343 744 134 713 743 743 743	Similarities	.791	.770	.833	.013	.620	.074	.646	002	.344	697.
771 807 828 -054 581 103 641 -040 300 692 794 746 016 590 -134 500 996 33 667 048 563 -001 519 519 500 345 416 702 -001 593 -013 512 643 503 343 643 516 517 514 546 702 503 343 730 515 647 514 719 710 513 313 731 512 514 543 546 513 313 731 512 543 546 513 313 731 513 543 543 543 323 731 513 543 543 543 323 732 543 543 543 543 323 734 533 543 543 543	Vocabulary	.806	.919	.879	011	.623	034	.624	018	.328	.774
	Information	.771	.807	.828	054	.581	.103	.641	040	.300	.691
	Comprehension	.692	.794	.746	.016	.550	134	.500	.096	.363	.570
.702 -001 593 -073 522 889 536 -006 549 .653 .228 603 .154 566 .243 .627 .032 .349 .703 .347 .512 .047 .414 .199 .469 .037 .349 .716 .347 .512 .047 .414 .199 .469 .037 .349 .716 .329 .687 .282 .689 .510 .037 .315 .716 .136 .512 .204 .513 .646 .013 .315 .716 .136 .213 .840 .716 .312 .312 .716 .136 .213 .840 .312 .312 .312 .716 .136 .213 .213 .213 .312 .312 .716 .136 .213 .216 .213 .216 .312 .716 .131 .131 .131	Block Design	.667	.048	.563	021	.519	.656	.731	.124	.416	.547
	Visual Puzzles	.702	001	.593	073	.522	.889	.836	008	.340	.702
	Matrix Reasoning	.663	.228	.603	.154	.566	.343	.627	.032	.349	.451
	Figure Weights	.708	.169	.635	.206	.613	.464	.702	058	.315	.542
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Picture Concepts	.515	.347	.512	.047	.414	.199	.469	037	.211	.282
.718 022 .592 .892 .840 022 .545 042 .392 .629 .050 .514 .631 .688 .165 .538 .036 .335 .716 .136 .620 .783 .808 .128 .510 .024 .420 .716 .136 .307 .082 .421 .110 .299 .805 .802 .490 .053 .342 .035 .433 .132 .413 .666 .739 .151 .151 .174 .068 .213 .489 .489 .459 .152 .057 .185 .176 .178 .132 .413 .489 .479 .151 .151 .176 .178 .183 .413 .489 .479 .151 .151 .151 .188 .132 .413 .489 .479 .151 .176 .178 .178 .178 .413 .479 .479 .151 .178 .178 .183 .183 .1	Arithmetic	.737	.329	.687	.262	.658	.223	.646	.013	.373	.548
i629 -050 514 631 688 1165 538 -036 335 i716 i136 620 783 808 -128 510 024 420 i716 i136 620 783 808 -128 510 024 420 i490 -053 342 035 433 132 413 686 739 i520 057 185 -161 174 068 213 686 739 i520 153 -161 174 068 213 686 739 i521 153 -161 174 068 739 459 759 is1 174 068 312 174 068 311 759 is1 170 174 068 316 311 758 311 is1 175 1 175 175 175 175 175 is1 1	Digit Span	.718	022	.592	.892	.840	022	.545	042	.392	.708
.716 .136 .620 .783 .808 128 .510 .024 .420 .446 .007 .307 .082 .421 110 .299 .805 .802 .440 .007 .307 .082 .433 .132 .413 .805 .802 .440 .057 .185 .161 .174 .068 .739 .439 .252 .057 .185 .161 .174 .068 .213 .489 .459 .252 .057 .185 .161 .174 .068 .213 .489 .459 .254 .151 .151 .174 .98 .213 .87 .87 .151 .151 .18 .18 .18 .18 .81 .41 .151 .151 .18 .18 .18 .81 .18 .11 .151 .19 .19 .18 .18 .18 .11 .11 .11 .151 .19 .19 .18 .18 .18 .14	Picture Span	.629	050	.514	.631	.688	.165	.538	036	.335	.485
446 007 307 082 421 -110 299 805 802 802 802 802 803 490 -053 342 035 433 132 443 566 739 739 252 057 185 -161 174 068 213 489 739 736 151 151 151 362 311 87 459 736 151 151 362 312 131 147 736 151 151 151 188 311 157 736 151 151 168 157 161 161 73 1 157 157 157 151 151 732 1 1 1 1 1<1	Letter-Number Sequencing	.716	.136	.620	.783	.808	128	.510	.024	.420	.662
490 053 .342 .035 .433 .132 .413 .686 .739 .252 .057 .185 161 .174 .068 .213 .489 .459 .252 .057 .185 161 .174 .068 .213 .489 .459 .736 .151 .98 .98 .91 .87 .87 .736 .151 .56 .98 .312 .87 .87 .01 .43.50 .6.56 .98 .362 .311 .87 .01 .17 .98 .15P .98 .97 .91 .01 .1 .1 .1 .1 .1 .1 .01 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1 .1	Coding	.446	.007	.307	.082	.421	110	.299	.805	.802	.649
.252 .057 .185 161 .174 .068 .213 .489 .459 7.36 .151 .98 .98 .87 .87 7.36 .151 .98 .3.1 .87 8.1 .151 .98 .87 .87 9.1 .151 .56 .3.62 .3.1 9.1 .1.51 .3.62 .3.1 .3.1 9.1 .1.51 .3.62 .3.1 .3.1 9.1 .1.51 .1.51 .1.51 .51 9.1 .1.51 .1.51 .1.51 .1.51 9.1 .1.52 .1.51 .1.51 .1.51 9.1 .1.52 .1.51 .1.51 .1.51 9.2 .57 .50 .43 .1.51 .1.51	Symbol Search	.490	053	.342	.035	.433	.132	.413	.686	.739	.558
7.36 1.51 .98 A3.50 6.56 3.62 A3.50 5.56 3.62 Ons F1:VC F2:WM F3:PR) - - - .725 - - - .732 .674 - - .399 .507 .433 -	Cancellation	.252	.057	.185	161	.174	.068	.213	.489	.459	.222
43.50 6.56 3.62 ons F1:VC F2:WM F3:PR) - - - .725 - - - .732 .674 - - .399 .507 .433	Eigenvalue		7.36		1.51		.98		.87		
ons F1:VC F2:WM F3:PR)	% Variance		43.50		6.56		3.62		3.11		
)	Promax-based factor correlations		F1:VC		F2: WM		F3: PR		F4: PS		
.725 - .732 .674 - .399 .507 .433	F1: Verbal Comprehension (VC)		I								
.732	F2: Working Memory (WM)		.725		I						
.399 .507 .433	F3: Perceptual Reasoning (PR)		.732		.674		I				
	F4: Processing Speed (PS)		.399		.507		.433		ı		

TABLE 6 Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) exploratory factor analysis; Four oblique factor solution for the standardization sample 12- to 14-year-

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structure that required explication (Gorsuch, 1983) and the Schmid-Leiman procedure was applied to better understand variance apportionment among general and group factors.

2.3.6 Ages 12–14 SL bifactor analyses: Four first-order factors

Results for the Schmid and Leiman orthogonalization of the higher-order factor analysis are shown in Table 7. All subtests were properly associated (higher residual variance) with their theoretically proposed factor after removing *g* variance except Picture Concepts and Arithmetic, which had somewhat higher residual loadings with the Verbal Comprehension factor. The *g* factor accounted for 38.3% of the total variance and 67.4% of the common variance.

The general factor also accounted for between 5.1% (Cancellation) and 53.6% (Vocabulary) of individual subtest variability. At the first-order level, VC accounted for an additional 5.4% of the total variance and 9.5% of the common variance, WM accounted for an additional 3.3% of the total variance and 5.8% of the common variance, PR accounted for an additional 3.5% of the total variance and 6.2% of the common variance, and PS accounted for an additional 6.3% of the total variance and 11.1% of the common variance. The general and group factors combined to measure 56.7% of the variance in WISC-V scores resulting in 43.3% unique variance (combination of specific and error variance).

Table 7 also shows $\omega_{\rm H}$ and $\omega_{\rm HS}$ coefficients that were estimated based on the SL results. Because of subtest migration of Picture Concepts and Arithmetic on Verbal Comprehension, omega-hierarchical and omega-subscale coefficients were estimated with Picture Concepts and Arithmetic loadings on Verbal Comprehension as well as with their theoretically consistent loadings on Perceptual Reasoning and Working Memory, respectively. The $\omega_{\rm H}$ coefficient for general intelligence (.847, .842) was high and sufficient for scale interpretation; however, the $\omega_{\rm HS}$ coefficients for the four group factors (VC, WM, PR, PS) were considerably lower (.149–.503, .173–.503). Thus, unit-weighted composite scores for the four group factors based on these indicators, with the possible exception of PS, likely possess too little true score variance for clinical interpretation (Reise, 2012; Reise et al., 2013) for 12- to 14-year-olds.

2.3.7 Ages 15–16 first-order EFA: Four-factor extraction

Table 8 shows the results of four-factor extraction with promax rotation. The g loadings ranged from .243 (Cancellation) to .813 (Vocabulary) and all were within the fair-to-good range (except Coding, Symbol Search, and Cancellation). Picture Concepts had a salient pattern coefficient on the Verbal Comprehension factor. Arithmetic failed to exhibit salient pattern loadings on any group factor but had split loadings on Verbal Comprehension (.299), Working Memory (.291), and Perceptual Reasoning (.291), that would be salient considering a confidence interval. Figure Weights had a secondary cross-loading with Verbal Comprehension. Table 8 shows robust Verbal Comprehension (Similarities, Vocabulary, Information, Comprehension), Working Memory (Digit Span, Picture Span, Letter–Number Sequencing), Perceptual Reasoning (Block Design, Visual Puzzles, Matrix Reasoning, Figure Weights), and Processing Speed (Coding, Symbol Search, Cancellation) factors with theoretically consistent subtest associations. Picture Concepts again migrated away from its theoretically related factor to the Verbal Comprehension factor. The moderate-to-high factor correlations shown in Table 6 (.323 to .754) imply a higher-order or hierarchical structure that required explication (Gorsuch, 1983) and the Schmid–Leiman procedure was applied to better understand variance apportionment among general and group factors.

2.3.8 Ages 15–16 SL bifactor analyses: Four first-order factors

Results for the Schmid and Leiman orthogonalization of the higher-order factor analysis are shown in Table 9. All subtests were properly associated (higher residual variance) with their theoretically proposed factor after removing *g* variance except Picture Concepts, which had higher residual loading on the Verbal Comprehension factor. The *g* factor accounted for 37.5% of the total variance and 66.7% of the common variance.

The general factor also accounted for between 5.2% (Cancellation) and 56.9% (Arithmetic) of individual subtest variability. At the first-order level, VC accounted for an additional 5.1% of the total variance and 9.1% of the common variance, WM accounted for an additional 4.2% of the total variance and 7.4% of the common variance, PS accounted for an additional 6.8% of the total variance and 12.1% of the common variance, and PR accounted for an additional 2.6%

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	D	General	F 1: Verbal Col	F1: Verbal Comprehension	F2: Working Memory	g Memory	F3: Perceptu	F3: Perceptual Reasoning	F4: Processing Speed	sing Speed		
WISC-V subtest	q	S ²	q	S ²	q	S ²	q	S ²	q	S ²	h ²	u ²
Similarities	.724	.524	.408	.166	.007	000	.042	.002	002	000	.692	.308
Vocabulary	.732	.536	.487	.237	006	000	019	000	015	000	.774	.226
Information	.702	.493	.428	.183	028	.001	.058	.003	034	.001	.681	.319
Comprehension	.628	.394	.421	.177	.008	000	076	900.	.081	.007	.584	.416
Block Design	.628	.394	.025	.001	011	000	.372	.138	.105	.011	.545	.455
Visual Puzzles	.664	.441	001	000.	038	.001	.504	.254	007	000	696.	.304
Matrix Reasoning	.625	.391	.121	.015	.079	900.	.195	.038	.027	.001	.450	.550
Figure Weights	.671	.450	060.	.008	.106	.011	.263	.069	049	.002	.541	.459
Picture Concepts	.479	.229	.184	.034	.024	.001	.113	.013	031	.001	.278	.722
Arithmetic	.694	.482	.174	.030	.135	.018	.127	.016	.011	000	.546	.454
Digit Span	.705	.497	012	000.	.459	.211	012	000.	036	.001	.709	.291
Picture Span	.615	.378	027	.001	.325	.106	.094	600.	031	.001	.494	.506
Letter-Number Sequencing	.694	.482	.072	.005	.403	.162	073	.005	.020	000	.655	.345
Coding	.412	.170	.004	000.	.042	.002	062	.004	.683	.466	.642	.358
Symbol Search	.457	.209	028	.001	.018	000.	.075	900.	.582	.339	.554	.446
Cancellation	.225	.051	030.	.001	083	.007	.039	.002	.415	.172	.232	.768
Total Variance		.383		.054		.033		.035		.063	.567	.433
Common Variance		.674		.095		.058		.062		.111		
PC with PR, AR with WM	ω _H =	$\omega_{\rm H} = .847$	$\omega_{\rm HS} = .252$.252	$\omega_{HS} = .163$.163	$\omega_{HS} = .149$:.149	$\omega_{HS} = .503$.503		
PC and AR with VC	$\omega_{\rm H} =$	$\omega_{\rm H}$ = .842	$\omega_{HS} = .195$.195	$\omega_{\rm HS} = .213$.213	$\omega_{\rm HS} = .173$	173	$\omega_{\rm HS} = .503$.503		

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TABLE 7 Sources of variance in the Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) for the standardization sample 12- to 14-year-olds (*n* = 600) according to an

larger than for the theoretically assigned factor).

	General ^a	F1: Verbal Comprehension	mprehension	F2: Working Memory	ig Memory	F3: Proces	F3: Processing Speed	F4: Perceptual Reasoning	al Reasoning	
WISC-V Subtest	S	Ρ	S	Р	S	Р	S	Р	S	h ²
Similarities	.767	.956	.845	000	.582	.051	.288	169	.574	.726
Vocabulary	.813	.755	.856	.046	.629	081	.236	.126	.692	.743
Information	.749	.767	.804	055	.546	054	.224	.123	.642	.654
Comprehension	.680	.698	.716	010	.513	.068	.292	.004	.553	.517
Block Design	.728	.032	.614	106	.515	.135	.449	.813	.824	696.
Visual Puzzles	.709	.148	.636	055	.515	004	.321	.701	.774	.607
Matrix Reasoning	.640	.171	.577	.194	.552	009	.280	.361	.615	.425
Figure Weights	.680	.303	.653	.076	.537	118	.200	.443	.672	.513
Picture Concepts	.529	.313	.510	.121	.443	.018	.229	.141	.465	.283
Arithmetic	.799	.299	.732	.291	.707	.026	.367	.291	.722	.645
Digit Span	.713	074	.567	.770	.812	.052	.392	.108	.589	.667
Picture Span	.614	.088	.517	.625	.686	.077	.342	048	.468	.478
Letter-Number Sequencing	.673	.055	.568	.899	.825	053	.277	137	.482	.692
Coding	.387	006	.245	.035	.324	.739	.745	016	.318	.555
Symbol Search	.451	.078	.313	036	.348	.786	.801	600.	.380	.645
Cancellation	.243	100	.141	.048	.210	.396	.428	.105	.230	.190
Eigenvalue		7.24		1.61		1.03		.82		
% Variance		42.78		7.20		4.14		2.36		
Promax-based factor correlations		F1: VC		F2: WM		F3: PS		F4: PR		
F1: Verbal Comprehension (VC)		ı								
F2: Working Memory (WM)		.705		I						
F3: Processing Speed (PS)		.323		.412		I				
F4: Perceptual Reasoning (PR)		.754		.667		.427		I		

nality. Salient pattern coefficients are shown in bold (pattern coefficient \geq .30).

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TABLE 8 Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) exploratory factor analysis: Four oblique factor solution for the standardization sample 15- to 16-year-

WISC-V subtest			F1: Verbal Comprehension	prehension	F2: Working Memory	g Memory	F3: Processing Speed	ing Speed	F4: Perceptu	F4: Perceptual Reasoning		
	q	S ²	q	S ²	q	S ²	q	S ²	q	S ²	h²	u ²
Similarities .	.688	.473	.507	.257	000	000	.045	.002	085	.007	.740	.260
Vocabulary	.749 .5	.561	.400	.160	.027	.001	072	.005	.063	.004	.731	.269
Information .	.687 .4	.472	.407	.166	032	.001	048	.002	.061	.004	.645	.355
Comprehension .	.619 .3	.383	.370	.137	006	000	090.	.004	.002	000	.524	.476
Block Design	.707	500	.017	000	062	.004	.120	.014	.406	.165	.683	.317
Visual Puzzles	.686	.471	.078	900.	032	.001	004	000	.350	.123	600.	.400
Matrix Reasoning	.611 .3	.373	.091	.008	.113	.013	008	000	.180	.032	.427	.573
Figure Weights	.648 .4	.420	.161	.026	.044	.002	105	.011	.222	.049	.508	.492
Picture Concepts	.494 .2	.244	.166	.028	.071	.005	.016	000	.070	.005	.282	.718
Arithmetic .	.754 .5	.569	.158	.025	.170	.029	.023	.001	.145	.021	.644	.356
Digit Span	.680	.462	039	.002	.449	.202	.046	.002	.054	.003	.671	.329
Picture Span	.576 .3	.332	.047	.002	.364	.132	.068	.005	024	.001	.472	.528
Letter-Number Sequencing	.634 .4	402	.029	.001	.524	.275	047	.002	069	.005	.684	.316
Coding .	.349 .1	.122	003	000	.020	000	.656	.430	008	000	.553	.447
Symbol Search	.406 .1	.165	.041	.002	021	000	.698	.487	.004	000	.654	.346
Cancellation .	.227 .0	.052	053	.003	.028	.001	.352	.124	.052	.003	.182	.818
Total Variance	Ċ.	.375		.051		.042		.068		.026	.562	.438
Common Variance	Ŷ.	.667		.091		.074		.121		.047		
PC with PR	$\omega_{\rm H}$ = .844	-	$\omega_{HS} = .241$	241	$\omega_{\rm HS}$ = .209	209	$\omega_{\rm HS} = .530$.530	e SHω	$\omega_{\rm HS} = .108$		
PC with VC	$\omega_{\rm H}$ = .841		$\omega_{\rm HS} = .214$	214	$\omega_{\rm HS}$ = .209	209	$\omega_{\rm HS} = .530$.530	$\omega_{\rm HS} = .131$.131		

TABLE 9 Sources of variance in the Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) for the standardization sample 15- to 16-year-olds (*n* = 400) according to an

and variance estimates consistent with the theoretically proposed factor. Italic type shows coefficients and variance estimates associated with an alternate factor (where cross-loading b was larger than for the theoretically assigned factor).

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of the total variance and 4.7% of the common variance. The general and group factors combined to measure 56.2% of the variance in WISC-V scores resulting in 43.8% unique variance (combination of specific and error variance).

Also shown in Table 9 are $\omega_{\rm H}$ and $\omega_{\rm HS}$ coefficients that were estimated based on the SL results. Because of subtest migration of Picture Concepts on Verbal Comprehension, omega-hierarchical and omega-subscale coefficients were estimated with Picture Concepts loading on Verbal Comprehension as well as with its theoretically consistent loading on Perceptual Reasoning. The $\omega_{\rm H}$ coefficient for general intelligence (.844, .841) was high and sufficient for scale interpretation; however, the $\omega_{\rm HS}$ coefficients for the four group factors (VC, WM, PR, PS) were considerably lower (.108–.530, .131–.530). Thus, for the four group factors, with the possible exception of PS, unit-weighted composite scores based on these indicators would likely possess too little true score variance for clinical interpretation (Reise, 2012; Reise et al., 2013) for the 15- to 16-year-old age group.

2.4 | One-, two-, and three-factor extraction

Examination of results when extracting fewer than four factors paralleling those of Canivez et al. (2016) resulted in structures that were not consistent with previous versions of the WISC nor other Wechsler scales. One-, two-, and three-factor models fused theoretically meaningful constructs indicative of underextraction and were judged unsatisfactory (Gorsuch, 1983; Wood et al., 1996).

3 DISCUSSION

The WISC-V *Technical and Interpretive Manual* claimed support for a five first-order and one higher-order (g) factor model for the 16 primary and secondary subtests. Structural validity support was based exclusively on CFA as no EFA results were included. Also absent were decomposed variance estimates (or any variance estimates) for the higherorder and lower-order factors and model-based reliability ($\omega_{\rm H}$ and $\omega_{\rm HS}$) estimates that would provide users of the WISC-V information necessary for judging the psychometric fitness of provided scores (Canivez, 2010, 2014a; Canivez & Kush, 2013; Rodriguez et al., 2016). Given the absence of these necessary analyses and summary statistics, the present study used EFA and hierarchical EFA methods to assess the WISC-V structure to examine CFA and EFA convergence or divergence among four age groups in the WISC-V standardization sample.

Consistent with the findings from Canivez et al. (2016), who investigated the WISC-V structure with the total standardization sample, the present study also indicated there was no EFA evidence to support a five-factor representation of the WISC-V within any of the four age groups examined (see Figures A1–A4 in Appendix A and Tables B1 through B8 in Appendix B in the online supplemental materials). Forced extraction of five factors resulted in the fifth factor having only one subtest with a salient factor pattern loading and is inadequate (Preacher & MacCallum, 2003).

Also consistent with Canivez et al. (2016) was general support for most subtests' association with a four-factor model that was similar to the WISC-IV. In each of the four age groups, the Verbal Comprehension subtests (Similarities, Vocabulary, Information, Comprehension), Working Memory subtests (Digit Span, Picture Span, Letter–Number Sequencing), Perceptual Reasoning subtests (Block Design, Visual Puzzles, Matrix Reasoning, Figure Weights), and Processing Speed subtests (Coding, Symbol Search, Cancellation) were consistently associated with the theoretical constructs previously posited (i.e., WISC-IV, WAIS-IV, WPPSI-IV) despite changes in subtest content. The subtests thought to represent separate Visual Spatial (Block Design and Visual Puzzles) and Fluid Reasoning (Matrix Reasoning and Figure Weights) factors merged together in all four age groups and appear to represent the former Perceptual Reasoning factor present in the WISC-IV and WAIS-IV. It appears that FW and MR are weaker indicators of Perceptual Reasoning than are BD and VP, but they clearly did not produce a separate Fluid Reasoning factor. These results, as with those from Canivez et al. (2016), fail to support the publisher's creation of separate Visual Spatial and Fluid Reasoning factors and standardized factor index scores that represent them. Other evidence of problems with specifying separate Visual Spatial and Fluid Reasoning factors is present in the redundant loading of FR on general intelligence

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reported in CFA in the WISC-V *Technical and Interpretive Manual* and shown in Figure 1, as well as in Chen et al. (2015). Thus, it appears that the WISC-V has been overfactored as represented in the WISC-V *Technical and Interpretive Manual*.

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Following transformation with the Schmid and Leiman (1957) procedure, the WISC-V g factor accounted for 5 to 6 times more variance than any single group factor and approximately twice the variance of all four group factors combined in all four age groups. To further show the general dominance of subtest measurement of general intelligence, Figure 2 shows the portions of subtest variance apportioned to the general intelligence dimension and the portions of subtest variance apportioned to the four WISC-V group factors. With the exception of the CD, SS, and CN subtests; most common subtest variance was that associated with general intelligence in each of the four age groups and that what is primarily measured is general intelligence, not the group factors.

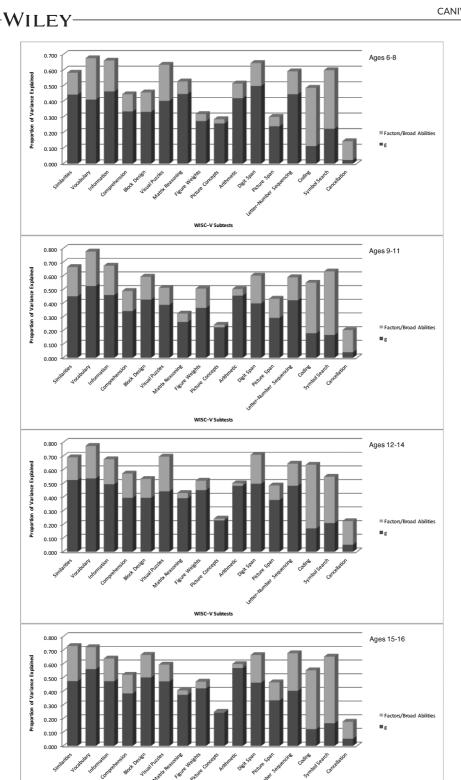
Also, the $\omega_{\rm H}$ coefficients for the g factor in all four age groups (.817–.847) were high and indicated large portions of true score variance attributable to unit-weighted scores based on all subtests. The $\omega_{\rm HS}$ coefficients for the four group factors in all four age groups were considerably lower (range of .131 to .280 for the VC, PR, and WM factors), falling far below the minimum threshold of .50 suggested by Reise (2012) and Reise et al. (2013) for confident clinical interpretation. That is, they captured too little unique true score variance once g variance was removed. The $\omega_{\rm HS}$ coefficients for the PS factor in all four age groups ranged from .478 to .530 and approached or met the minimum standard for possible interpretation. These results appear to support Carroll's model but not Cattell–Horn, as pointed out by Cucina and Howardson (2017).

Arithmetic was associated with Working Memory for the 6–8 and 9- to 11-year-old age groups, but migrated to Verbal Comprehension for the 12- to 14-year-old age group and was not saliently associated with any group factor in the 15- to 16-year-old age group (its variance spread evenly between VC, PR, and WM). Numerous problems with Arithmetic as a subtest in Wechsler scales have been described (Canivez & Kush, 2013; Canivez et al., 2016; Watkins & Ravert, 2013). As suggested previously (Canivez & Kush, 2013; Canivez et al., 2015; Watkins & Ravert, 2013) it is likely time for Arithmetic to be removed as an indicator of Working Memory.

As observed by Canivez et al. (2016), Picture Concepts failed to demonstrate salient loadings on any factors in the 6–8 and 9- to 11-year-old age groups and when it did saliently load on a factor it was on a theoretically inconsistent one (VC). This may be the reason the publisher does not include Picture Concepts in any regularly calculated factor-based scores (PC is only used to replace a Fluid Reasoning subtest in calculating the FSIQ because of spoiling either Matrix Reasoning or Figure Weights). Given its failure to saliently load on any latent factor, its inclusion as a substitute for Matrix Reasoning or Figure Weights for estimating the FSIQ from a Fluid Reasoning area may be questionable.

The superiority of general intelligence observed in all four age groups is identical to that found by Canivez et al. (2016) with the total WISC-V standardization sample and similar to other studies of Wechsler scales using both EFA and CFA methods (Bodin et al., 2009; Canivez, 2014a; Canivez & Watkins, 2010a, 2010b; Canivez et al., 2017a; Dombrowski, Canivez, & Watkins, 2018; Gignac & Watkins, 2013; Lecerf & Canivez, 2017; McGill & Canivez, 2016, 2017; Nelson et al., 2013; Watkins, 2006; 2010; Watkins & Beaujean, 2014; Watkins et al., 2006, 2013, 2017) and other intelligence tests (Canivez, 2008; Canivez & McGill, 2016; Canivez et al., 2009; Cucina & Howardson, 2017; DiStefano & Dombrowski, 2006; Dombrowski, 2013, 2014a, 2014b; Dombrowski & Watkins, 2013; Dombrowski et al., 2009; Dombrowski, Golay, McGill & Canivez, 2018a; Dombrowski, McGill, & Canivez, 2017a, 2017b, 2018b; Dombrowski, McGill, Canivez & Peterson, 2018c; Nelson & Canivez, 2012; Nelson et al., 2007). These results are also consistent with the broader professional literature on the importance and dominance of general intelligence (Deary, 2013; Jensen, 1998; Lubinski, 2000; Ree et al., 2003).

As would be predicted by Frazier and Youngstrom (2007), too little true score variance was associated with the four WISC-V group factors, with the possible exception of PS, to warrant confident clinical interpretation (Reise, 2012; Reise et al., 2013). Gustafsson (1984) noted that, "individual differences in cognitive performance can be understood in terms of several sources of variance, some of which are broad and some of which are narrow" (p. 67) and Gorsuch (1983) explained that, "in science, the concern is with generalizing as far as possible and as accurately as possible. Only when the broad and not so broad generalities do not apply to a given solution does one move to the narrowest, most specific level of generality" (p. 249). Most of the WISC-V variance was contributed by a broad general factor so the WISC-V general factor is "of definite interest" (Gorsuch, 1983, p. 253) but the "lower order factors may be of little



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FIGURE 2 Sources of variance for the 16 WISCV primary and secondary subtests for the four age groups based on Schmid and Leiman (1957) orthogonalization of higher-order extraction with four first-order factors (VC, PR, WM, PS) based on Tables 3, 5, 7, and 9.

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interest" (Wolff & Preising, 2005, p. 50). As pointed out by Cucina and Howardson (2017), such evidence supports the three-stratum theory proposed by Carroll (1993, 2003) but not the structure advanced by Cattell-Horn, which ostensibly is a two-stratum model (no g factor).

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Given the absence of important information from the WISC-V *Technical and Interpretive Manual* as described in the present study as well as results from Canivez et al. (2016) and Canivez et al. (2017a), researchers and clinicians using the WISC-V must rely on the extant literature to adequately evaluate which WISC-V scores have sufficient reliability and validity for interpretation and use. Numerous studies have published results at odds with those provided in test technical manuals (cf. Canivez, 2008; Canivez & McGill, 2016; Canivez & Watkins, 2010a, 2010b; DiStefano & Dombrowski, 2006; Dombrowski, 2013, 2014a, 2014b; Dombrowski et al., 2009; Dombrowski, McGill, & Canivez, 2017a,b; McGill & Canivez, 2007; Watkins, 2006), but such information should have been included in those technical manuals in the first place.

Researchers and clinicians must rely on more than the test technical manuals to use test scores appropriately as they bear "the ultimate responsibility for appropriate test use and interpretation" (American Educational Research Association, American Psychological Association, & National Council on Measurement in Education, 2014, p. 141). The present results, in addition to those of Canivez et al. (2016, 2017a), will assist users of the WISC-V to "know what their tests can do and act accordingly (Weiner, 1989, p. 829).

3.1 | Limitations

Correlations provided in the Technical and Interpretive Manual Supplement (Wechsler, 2014c) were analyzed because NCS Pearson declined to provide the WISC-V standardization sample raw data. Analytical methods such as exploratory structural equation modeling (ESEM; Asparouhov & Muthén, 2009) might be a viable alternative to traditional EFA, but ESEM requires participant raw data, which were unavailable. Thus, the correlations from the technical manual were used but are rounded to two decimals and therefore could be less precise than correlations produced from raw data. However, greater precision would not be warranted by the sample size of each age group (Bedeian, Sturman, & Streiner, 2009) and it is unlikely that the present results were substantially impacted by two-digit precision (Carroll, 1993). Another limitation is that the present study, while informative, may provide results that differ from those that might be produced by a CFA bifactor model. Reise (2012) indicated that the EFA-based SL procedure produces an approximate bifactor solution that is a reparameterization of the higher-order structure and contains proportionality constraints (Yung et al., 1999), but the SL procedure is the dominant exploratory approach to assessing bifactor structure in EFA. Use of CFA bifactor modeling as well as examination of factor invariance across these four age groups will further test the latent structure of the WISC-V and the present results will facilitate plausible CFA models to test invariance examination (Brown, 2015; Carroll, 1998). Such analyses would extend those of Reynolds and Keith (2017) by examining invariance of the bifactor structure with four group factors rather than only the first-order subtest alignment. Finally, these results may not extend to populations not well represented in the WISC-V normative sample. For example, profoundly gifted individuals may exhibit meaningful cognitive patterns that do not emerge among standardization samples (Robertson, Smeets, Lubinski, & Benbow, 2010).

4 | CONCLUSIONS

Results from this study provide important considerations for clinical interpretation of scores from the WISC-V. The results of analyses across the four age groups support interpretation of the general intelligence estimate (FSIQ). Lowerorder (index scores) are generally not supported for interpretation with the possible exception of the PSI. Independent analyses of the WISC-V failed to support the test publisher's posited five-factor structure. Because there was no evidence for separate Visual Spatial and Fluid Reasoning factors in any of these four age groups or the full standardization sample (Canivez et al., 2016, 2017a), the publisher should consider producing revised norms tables for a fourfactor model where the former Perceptual Reasoning factor is estimated in place of separate Visual Spatial and Fluid

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Reasoning factors. The overfactoring of the WISC-V in the WISC-V *Technical and Interpretive Manual* and factor index scores for VS and FR will likely result in misinterpretation and errors in clinical decision making (Beaujean, 2015b; Dombrowski, 2015). As shown in the present study as well as with the full standardization sample (Canivez et al., 2016; 2017a; Dombrowski, Canivez, Watkins, & Beaujean, 2015), primary interpretation of the WISC-V should be at the FSIQ level and consideration of other score interpretations must be made in light of the extremely small portions of true score variance uniquely captured by the factor index scores.

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SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section at the end of the article.

How to cite this article: Canivez GL, Dombrowski SC, Watkins MW. Factor structure of the WISC-V in four standardization age groups: Exploratory and hierarchical factor analyses with the 16 primary and secondary subtests. *Psychol Schs*. 2018;55:741–769. https://doi.org/10.1002/pits.22138

Appendix A

Supplemental Figures

Figures A1–A4 are parallel analysis scree plots for the four WISC-V age groups (6–8, 9–11, 12–14, 15–16) from exploratory factor analyses of standardization sample data.

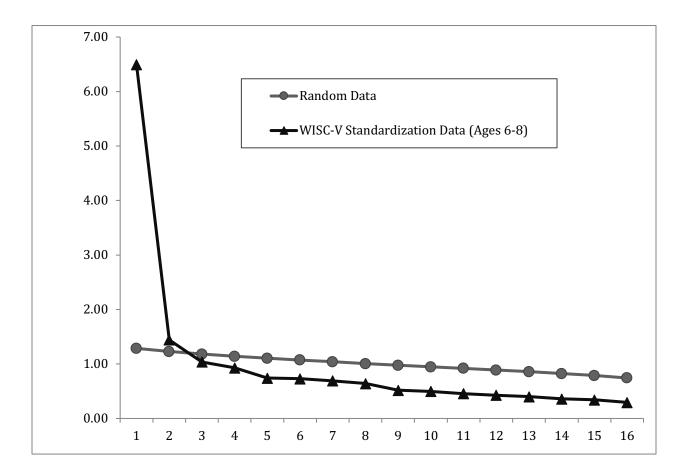


Figure A1. Scree plots for Horn's parallel analysis for WISC-V standardization sample ages 6-8 (N = 600).

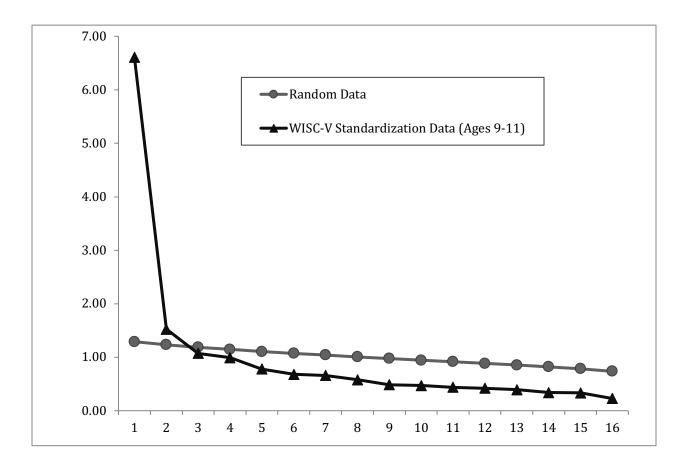


Figure A2. Scree plots for Horn's parallel analysis for WISC-V standardization sample ages 9-11 (N = 600).

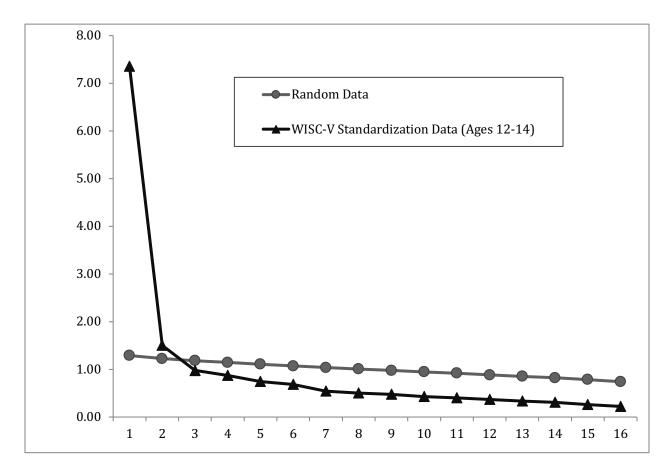


Figure A3. Scree plots for Horn's parallel analysis for WISC-V standardization sample ages 12-14 (N = 600).

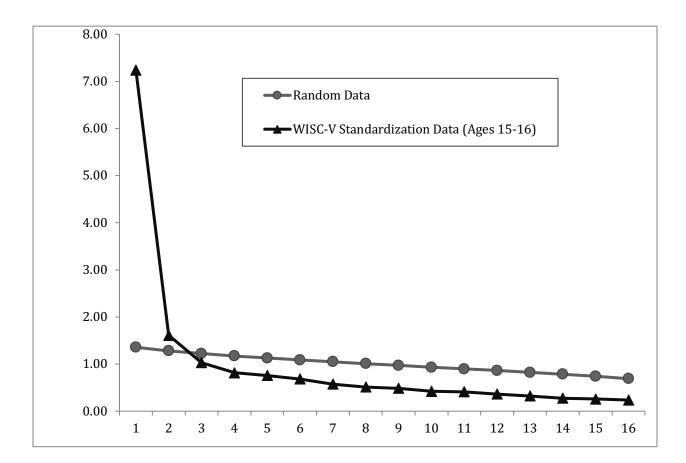


Figure A4. Scree plots for Horn's parallel analysis for WISC-V standardization sample ages 15-16 (N = 400).

Appendix B

Supplemental Tables

Tables B1–B8 are exploratory factor analyses results (odd numbered tables) extracting five factors and subsequent Schmid and Leiman (1957) exploratory Schmid-Leiman (SL) bifactor models (orthogonalized higher-order factor models) with five first–order factors (even numbered tables) for the four WISC–V age groups (6–8, 9–11, 12–14, 15–16) from standardization sample correlation matrices (Wechsler, 2014c).

F1: Verba		F1: Verbal	rbal	F2: Working	rking	F3: Perceptual	ceptual	F4: Processing	essing	F5:		
	General ¹	Comprehension	ension	Memory	ory	Reasoning	ning	Speed	pa	Inadequate	luate	
WISC-V Subtest	S	Р	S	P	S	P	S	Р	S	Р	S	h^2
SI	.723	999.	.772	.033	.569	.023	.587	043	.403	.185	.494	.621
VC	.703	.913	.826	034	.523	.049	.581	109	.254	106	.260	.709
IN	.745	.765	.804	021	.563	.019	909.	.108	.445	023	.398	.654
CO	.626	.592	.660	.182	.537	046	.488	.043	.328	092	.271	.455
BD	.610	.028	.494	042	.455	.642	.675	.138	.428	071	.293	.467
VP	.667	006	.546	.024	.518	.875	.786	077	.317	134	.218	.643
MR	.706	.017	.579	.087	.577	.577	.730	.002	.470	.181	.487	.570
FW	.550	960.	.477	760.	.457	.411	.557	068	.307	.104	.335	.333
PC	.537	.110	.454	.208	.479	.290	.512	060.	.336	107	.221	.304
AR	.679	.136	.576	.459	.677	600 [.]	.540	.055	.478	.201	.510	.523
DS	.744	024	.577	.896	.865	.033	.592	005	.413	077	.330	.753
PS	.508	035	.402	.317	.501	.151	.445	032	.355	.267	.443	.319
LN	069.	.084	.570	.673	.753	022	.536	017	.415	.102	.427	.580
CD	.367	089	.210	.067	.307	075	.277	.821	.686	146	.303	.500
SS	.517	.032	.366	.017	.406	.003	.421	.823	.774	123	.396	609.
CA	.177	.070	.123	121	060.	.133	.187	.521	.304	429	078	.215
Eigenvalue		.9	6.50	1.44	4	1.04)4	0.	0.93	0.	0.74	
% Variance		37.84	84	5.9	8	3.4	13	3.	3.05	1.	1.29	
Factor Correlations	rol	F1: VC	'C	F2: WM	/M	F3: PR	PR	F4: PS	Sc	F5	2	
Verbal Comprehension (VC)	nension (VC)	Ι										
Working M	Working Memory (WM)	.687	2	Ι								
Perceptual Reasoning (PR)	asoning (PR)	.723	~	.684	4	Ι						
Processin	Processing Speed (PS)	.459	•	.511	1	.531	1	Ι				
	F5	.466	5	.454	4	.446	9	.602	2	Ι		
<i>Note.</i> ¹ Factor structure coefficients from first unrotated factor (g-loadings) are correlations between the subtest and the general factor. WISC–V Subtest: SI = Similarities, VC = Vocabulary, IN = Information, CO = Comprehension, BD = Block Design, VP = Visual Puzzles, MR = Matri Reasoning, FW = Figure Weights, PC = Picture Concepts, AR = Arithmetic, DS = Digit Span, PS = Picture Span, LN = Letter–Number	ture coefficient ilarities, $VC = ^{1}$ Figure Weights.	s from first u Vocabulary, , PC = Pictur	Inrotated factor () IN = Information e Concepts, AR	actor (g-loa mation, C(s, AR = A1	adings) ar) = Comp rithmetic,	loadings) are correlations betwee CO = Comprehension, BD = Bloc Arithmetic, DS = Digit Span, PS	ons betwe BD = Bl it Span, P	en the subtest and the ge ock Design, VP = Visua S = Picture Span, LN = 1	est and the $V_1 = V_1$ Span, LN	e general factor. W sual Puzzles, MR = Letter–Number	actor. WIS s, MR = N Vumber	VISC-V = Matrix
Sequencing, CD = Coding, SS = Symbol Search, CA = Cancellation. $S =$ Structure Coefficient, $P =$ Pattern C Salient pattern coefficients presented in bold (pattern coefficient $\geq .30$). The eigenvalue for factor six was .73	Coding, SS = S fficients presen	Symbol Searc ted in bold (J	ch, CA = C pattern coe	= Cancellation. coefficient $\ge .3$	n. $S = Str. 30$). The	= Structure Coefficient, P The eigenvalue for factor	efficient, life for facto	$p = Pattern Coefficient, h^2 = r six was .73.$	Coefficie 3.	nt, $h^2 = Co$	Communality	<u>.</u> .
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Table B1 Wechsler Intelligence Scale for Children–Fifth Edition (WISC–V) Exploratory Factor Analysis: Five Obliaue Factor Solution for the

000) According to an Exploratory SL	to an Exp	<u>ioratory 2</u>	P		Model With Five First		Urder Factors	tors	F					
WISC-V	General	eral	FI: VC	VC	F2: WM	٧M	F3: PR	PR	F4:	PS	F5	2		
Subtest	p	S^2	p	S^2	p	S^2	p	S^2	p	S^2	p	S^2	h^2	u^2
SI	699.	.448	395	.156	.020	000^{-1}	.013	000	032	.001	.145	.021	.626	.374
VC	.610	.372	.541	.293	020	000.	.028	.001	081	.007	083	.007	.679	.321
N	.674	.454	.453	.205	013	000 ⁻	.011	000 [.]	.080	900.	018	000 [.]	.666	.334
CO	.557	.310	.351	.123	.109	.012	026	.001	.032	.001	072	.005	.452	.548
BD	.569	.324	.017	000 [.]	025	.001	.360	.130	.102	.010	056	.003	.468	.532
VP	.603	.364	004	000	.014	000 ⁻	.491	.241	057	.003	105	.011	.619	.381
MR	.675	.456	.010	000	.052	.003	.324	.105	.001	000	.142	.020	.584	.416
FW	.514	.264	.057	.003	.058	.003	.231	.053	050	.003	.081	.007	.333	.667
PC	.490	.240	.065	.004	.124	.015	.163	.027	.067	.004	084	.007	.298	.702
AR	.648	.420	.081	.007	.274	.075	.005	000 ⁻	.041	.002	.157	.025	.528	.472
DS	.676	.457	014	000	.535	.286	.019	000	004	000	060	.004	.747	.253
PS	.496	.246	021	000	.189	.036	.085	.007	024	.001	.209	.044	.334	.666
LN	.642	.412	.050	.003	.402	.162	012	000	013	000 ⁻	.080	900.	.583	.417
CD	.382	.146	053	.003	.040	.002	042	.002	.607	.380	114	.013	.534	.466
SS	.520	.270	.019	000	.010	000 [.]	.002	000	608.	.370	096	600.	.650	.350
CA	.153	.023	.041	.002	072	.005	.075	.006	.385	.148	336	.113	.297	.703
Total S ²		.325		.050		.038		.036		.058		.018	.525	.475
Common S^2		.620		.095		.072		.068		.110		.035		
<i>Note</i> . WISC–V Subtests: SI = Similarities,	Subtests:	SI = Sim	nilarities,	VC =	Vocabulary, IN = Information, CO	v, IN =	Informati	on, CO =	= Compre	Comprehension, BD	-	= Block Design,	ign, VP =	: Visual
Puzzles, MR = Matrix Reasoning, FW = Figur	Matrix Re	asoning, l	FW = Fig	gure Wei	e Weights, PC = Picture Concepts, AR = Arithmetic, DS = Digit Span, PS = Picture Span, LN	= Picture	S Concept	s, $AR = I$	Arithmeti	c, $DS = I$	Digit Span	, $PS = Pi$	cture Spa	n, LN =
Letter-Number Sequencing, CD = Coding, SS	Sequencin	ng, CD =	Coding,	τ	= Symbol Search, CA		= Cancellation.		TSC-V F	WISC-V Factors: VO	$^{r}C = Verb$	al Compi	C = Verbal Comprehension, WM	, WM =
Working Memory, $VS = Visual Spatial, PS$	ity, VS =	- Visual S	patial, P	S = Pro	= Processing Speed, FR = Fluid Reasoning.	peed, Fl	R = Fluid	l Reasoni	ng. b = .	loading	$b =$ loading of subtest on factor, $S^2 =$ variance	t on fact	or, $S^2 = 1$	/ariance
explained, h^{-} = communality, u^{-} = uniqueness. Bold type indicates coefficients and variance estimates consistent with the theoretically expressed factor (where cross loading h was larged variance).	Commun Italic tur	nality, <i>u</i> ⁻ se indicate	npinu =	eness. B viants and	old type	Indicate	s coeffici	ents and stad with	Variance	estimate	es consist	ent with	the theo.	cetically s larger
than for the theoretically assigned factor). Given the inadequacy of a five-factor solution Omega coefficients were not estimated for the five-	retically	assigned 1	factor). C	Jiven the	s inadequa	cy of a	five-factc	or solution	n Omega	coefficie	nts were 1	not estima	ated for th	ne five-
factor model.														

F1: Verbal		F1: Verbal	rbal	F2: Perceptual	eptual	F3: Working	orking	F4: Processing	essing	F5		
	General ¹	Comprehension	ension	Reasoning	ning	Memory	ory	Speed	pç	Inadequate	luate	
WISC-V Subtest	S	Ρ	S	P	S	P	S	Р	S	Р	S	h^2
SI	.743	.794	.814	.027	595.	049	.517	.027	.321	.047	.417	.665
VC	.802	.864	.883	.048	.646	002	.572	033	.312	004	.423	.780
IN	.749	.795	.822	.008	.595	033	.527	018	.285	.103	.466	.684
CO	.646	.671	697.	041	.493	.109	.505	.023	.303	051	.301	.494
BD	.682	025	.539	.767	.772	026	.498	860.	.405	006	.353	.603
VP	.655	.122	.568	.712	.733	.028	.482	086	.264	098	.302	.552
MR	.532	024	.417	.443	.555	.088	.432	.062	.299	.094	.345	.325
FW	.628	.005	.516	.661	.702	033	.451	063	.234	.178	.474	.522
PC	.508	.251	.473	.296	.487	.083	399	.026	.270	109	.194	.281
AR	.724	.115	.602	.115	.611	.236	.643	.117	.388	.437	.674	.631
DS	.644	071	.487	.086	.529	.715	.749	069	.299	.125	.453	.582
PS	.567	.034	.447	019	.429	.761	.703	025	.324	121	.234	.505
LN	.676	.095	.555	035	.518	.636	.744	016	.332	.166	.486	.583
CD	.459	045	.302	017	.347	.118	.429	869.	.736	.042	.171	.552
SS	.456	.046	.320	.007	.350	005	.380	.792	.801	055	.086	.645
CA	.226	004	.149	600.	.180	176	.136	.506	.446	.146	.143	.224
Eigenvalue		9.9	61	1.52	2	1	.07	0.	0.99	0.	0.78	
% Variance		38.68	68	9.9	4	3.	3.98	3.	3.38	1.	1.25	
Factor Correlations		F1: VC	'C	F2: PR	PR	F3: WM	ΝM	F4: PS	S	F5		
Verbal Comprehension (VC)	ension (VC)	I										
Perceptual Reasoning (PR)	soning (PR)	.713	~	Ι								
Working Memory (WM)	mory (WM)	.648	~	.648	18	I						
Processing	Processing Speed (PS)	.377	2	.430	30	.475	5	Ι				
	F5	.474	+	.481	31	.46	5	.150	0	Ι		
<i>Note.</i> ¹ Factor structure coefficients from first unrotated factor (g-loadin Subtests: SI = Similarities, VC = Vocabulary, IN = Information, CO = (Reasoning FW = Figure Weights PC = Picture Concents AR = Arithm	ure coefficient larities, VC = ^v ioure Weights	s from first u Vocabulary, PC = Pichur	Inrotated factor () IN = Information in Concents AR	actor (g-lo: mation, C(s AR = A1	loadings) are corre CO = Comprehens Arithmetic DS = 1	e correlatio rehension, DS = Digi	lations betwee ion, BD = Blo Digit Snan PS	en the subtest and the g ock Design, VP = Visua S = Picture Span I N =	est and the $VP = V_1$.	(g-loadings) are correlations between the subtest and the general factor. WISC-V n, CO = Comprehension, BD = Block Design, VP = Visual Puzzles, MR = Matri: = Arithmetic DS = Diort Snan PS = Picture Snan LN = Letter-Number	actor. WIS s, MR = N Jumber	VISC-V = Matrix
Sequencing, CD = Coding, SS = Symbol Search	Coding, $SS = S$	Symbol Searc	h, CA = C	= Cancellation.	n. $S = Str$	= Structure Coefficient, P	efficient, J	$P = Pattern Coefficient, h^2 =$	Coefficie	nt, $h^2 = Co$	Communality	
Salient pattern coefficients presented in bold (pattern coefficient \geq .30). The eigenvalue for factor six was .68	ficients presen	ted in bold (J	pattern coe	efficient \geq .	.30). The (eigenvalue	for facto	r six was .6	œ.			

 Table B3

 Wechsler Intelligence Scale for Children–Fifth Edition (WISC–V) Exploratory Factor Analysis: Five Oblique Factor Solution for the

$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	Sources of Variance in the Wechsler Intelligence Scale for Children-Fifth Edition (WISC-V) for the Standardization Sample 9–11 Year Olds $(N = 600)$ According to an Exploratory SL Bifactor Model with Five First-Order Factors	iance in th ording to a	ie Wechsi in Explor	ler Intellig atory SL E	tence Sc Sifactor	ale for Chi Model with	lldren–F h Five F	'ijth Editic ⁷ irst–Orde	nn (WISC r Factors	-V) for th s	ie Standa	rdization	Sample 5	9–11 Year	Olds
b S^2 b S^2 b S^2 b S^2 b S^2 h S^2	WISC-V	Gen	eral	F1: 1	VC	F2:]	PR	F3: `	WM		PS	H	5		
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Subtest	q	S^2	q	S^2	p	S^2	q	S^2	q	S^2	q	S^2	h^2	u^2
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	SI	.685	.469	.451	.203	.014	000 ⁻	029	.001	.024		.039	.002	.676	.324
675 456 451 203 004 000 -016 000 086 007 667 598 358 381 145 -022 000 064 004 000 -043 002 500 566 507 567 510 000 -043 002 510 510 510 510 500 567 501 000 007 500 547 500 547 520 510 510 510 510 510 510 510 510 510 510 510 510 510 520 510 520 510 500 510 520 510 500 510 510 520 510 500 510 510 520 510 510 520 510 520 510 520 510 510 510 520 520 510 <t< td=""><td>VC</td><td>.720</td><td>.518</td><td>.490</td><td>.240</td><td>.025</td><td>.001</td><td>001</td><td>000[.]</td><td>029</td><td></td><td>003</td><td>000[.]</td><td>.760</td><td>.240</td></t<>	VC	.720	.518	.490	.240	.025	.001	001	000 [.]	029		003	000 [.]	.760	.240
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	N	.675	.456	.451	.203	.004	000 [.]	019	000.	016		.086	.007	.667	.333
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	CO	.598	.358	.381	.145	022	000 ⁻	.064	.004	.020		043	.002	.510	.490
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	BD	.688	.473	014	000 ⁻	.407	.166	015	000	.086		005	000	.647	.353
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	VP	.600	.360	690.	.005	.378	.143	.016	000	075		082	.007	.520	.480
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	MR	.531	.282	014	000.	.235	.055	.051	.003	.054		079.	900.	.349	.651
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	FW	.581	.338	.003	000.	.351	.123	019	000	055		.149	.022	.486	.514
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	PC	.487	.237	.142	.020	.157	.025	.048	.002	.023		091	.008	.293	.707
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	AR	.720	.518	.065	.004	.061	.004	.138	.019	.102		.367	.135	069.	.310
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	DS	.605	.366	040	.002	.046	.002	.418	.175	060		.105	.011	.559	.441
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	PS	.543	.295	.019	000 ⁻	010	000 ⁻	.444	.197	022		102	.010	.503	.497
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	LN	.642	.412	.054	.003	019	000.	.371	.138	014		.139	.019	.573	.427
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	CD	.658	.433	026	.001	-000	000.	.069	.005	.610	.372	.035	.001	.812	.188
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	SS	.680	.462	.026	.001	.004	000.	003	000	.692	.479	046	.002	.944	.056
$n S^2 \qquad$	CA	.369	.136	002	000 ⁻	.005	000.	029	.001	.442	.195	.123	.015	.357	.643
.654 .089 .056 .059 .116	Total S^2		.382		.052		.032		.035		.068		.016	.584	.416
	Common S^2		.654		080.		.056		.059		.116		.027		
	Puzzles, MR =	- Matrix Ro	easoning,	FW = Fig	gure We	ights, PC =	= Pictur	e Concept.	s, $AR = 1$	Arithmeti	c, $DS = I$	Digit Spar	$1, \mathbf{PS} = \mathbf{P}_{1}$	icture Spa	n, LN =
Puzzles, MR = Matrix Reasoning, FW = Figure Weights, PC = Picture Concepts, AR = Arithmetic, DS = Digit Span, PS = Picture Span, LN =	Letter-Numbe	r Sequenci	ing, CD =	= Coding,		/mbol Sear	ch, CA		ation. W	75	actors: V	$^{r}C = Vert$	al Comp	rehension	, WM =
re Weights, PC = Picture Concepts, AR = \$ = Symbol Search, CA = Cancellation. V	Working Men	10ry, VS =	= Visual	Spatial, P		ocessing S ₁	peed, F	R = Fluid	Reason	ing. $b = \frac{1}{2}$	loading	of subtes	st on fact	tor, $S^2 = \frac{1}{2}$	/ariance
	explained, n ⁻	= commu r Italic tyr	nality, <i>u</i> ne indice	tes coeffic	eness. I Vients ar	sold type	indicaté	tes accoris	ents and ated with	variance an altern	estimation estimation	es consisi	tent with	l the theo	retically s larger
Puzzles, MR = Matrix Reasoning, FW = Figure Weights, PC = Picture Concepts, AR = Arithmetic, DS = Digit Span, PS = Picture Span, LN = Letter–Number Sequencing, CD = Coding, SS = Symbol Search, CA = Cancellation. WISC–V Factors: VC = Verbal Comprehension, WM = Working Memory, VS = Visual Spatial, PS = Processing Speed, FR = Fluid Reasoning. $b = $ loading of subtest on factor, S^2 = variance explained, h^2 = communality, u^2 = uniqueness. Bold type indicates coefficients and variance estimates consistent with the theoretically memory of factor Italic type indicates coefficients and variance estimates factor (where cross-loading h was larger by the proceeding h and type indicates secondated with an alternate factor (where cross-loading h was larger by the processing and variance estimates consistent with the theoretically become factor Italic type indicates coefficients and with an alternate factor (where cross-loading h was larger by the processing become estimates consistent with the theoretically become factor Italic type indicates coefficients and with an alternate factor (where cross-loading h was larger by the indicates coefficients and with an alternate factor (where cross-loading h was larger by the indicates coefficients and with an alternate factor (where cross-loading h was larger by the indicates coefficients and with an alternate factor (where cross-loading h was larger by the indicates coefficients and with an alternate factor (where cross-loading h was larger by the larger by the indicates second at the bar cross-loading h was larger by the indicates coefficients and with an alternate factor (where cross-loading h was larger by the larger by the indicates coefficients and with an alternate factor (where cross-loading h was larger by the coefficients and was larger by the l	than for the th	eoretically	assigned	factor). (Given th	te inadequé	acy of a	five-facto	or solutio	n Omega	coefficie	ints were	not estim	nated for t	he five-
eights, PC = Picture Concepts, AR = Arithmetic, DS = Digit Span, PS = Picture Sp lymbol Search, CA = Cancellation. WISC–V Factors: VC = Verbal Comprehension rocessing Speed, FR = Fluid Reasoning. b = loading of subtest on factor, S^2 = Bold type indicates coefficients and variance estimates consistent with the theo ind variance estimates associated with an alternate factor (where cross–loading b w the inadequacy of a five–factor solution Omega coefficients were not estimated for	factor model.														
Puzzles, MR = Matrix Reasoning, FW = Figure Weights, PC = Picture Concepts, AR = Arithmetic, DS = Digit Span, PS = Picture Span, LN = Letter–Number Sequencing, CD = Coding, SS = Symbol Search, CA = Cancellation. WISC–V Factors: VC = Verbal Comprehension, WM = Working Memory, VS = Visual Spatial, PS = Processing Speed, FR = Fluid Reasoning. $b = \text{loading of subtest on factor}$, S^2 = variance explained, h^2 = communality, u^2 = uniqueness. Bold type indicates coefficients and variance estimates consistent with the theoretically proposed factor. Italic type indicates coefficients and variance estimates consistent with the theoretically proposed factor. Italic type indicates coefficients and variance estimates factor (where cross–loading <i>b</i> was larger than for the theoretically assigned factor). Given the inadequacy of a five–factor solution Omega coefficients were not estimated for the five-factor model.															

Table B4

		E1. V/outhol	امط	ED. Working	-lin a	E2. Dorominal	loutant	E4. Droo		DC		
	General ¹	Comprehension	ension	Memory	ğınıxı JIV	Reasoning	ning	Speed	guneer bt	Inadequate	Juate	
WISC-V Subtest	S	Р	S	Р	S	Р	S	Р	S	Р	S	h^2
SI	062.	.750	.831	.020	.626	.081	.655	005	.340	.016	.490	.695
VC	.805	305	.879	005	.629	029	.634	021	.325	.012	.500	.774
IN	.770	808.	.831	051	.586	.120	.652	046	.295	020	.465	.698
CO	.691	.790	.747	.022	.555	126	.512	.093	.360	004	.407	.570
BD	699.	.055	.565	028	.523	.731	.750	.110	.413	095	.336	.578
VP	.700	031	.588	062	.527	.876	.825	010	.336	.037	.451	.685
MR	.663	.146	.591	.160	.569	.307	.623	.041	.349	.152	.483	.456
FW	707.	.157	.633	.213	.617	.480	.707	064	.311	015	.414	.544
PC	.556	.020	.506	014	.417	020	.465	.002	.210	.904	899.	.808
AR	.736	.319	.686	.269	.661	.230	.652	600.	.370	004	.428	.548
DS	.717	011	.591	606 .	.843	017	.551	046	.389	048	.357	.714
PS	.628	089	.506	.638	.688	.145	.537	031	.334	.064	.386	.486
LN	.714	.141	.619	.789	.807	127	.517	.024	.417	019	.373	.660
CD	.445	.005	.306	.079	.420	117	.304	.813	808.	.017	.190	.659
SS	.488	048	.342	.036	.433	.139	.418	.677	.735	011	.215	.553
CA	.251	.059	.186	161	.175	.073	.216	.485	.458	003	.108	.220
Eigenvalue		7.36	36	1	.51	0.	0.98	0.	0.87	0.	0.75	
% Variance		43.	65	6.4	6.66	3.	3.74	3.	3.13	3.1	11	
Factor Correlations		F1: VC	'C	F2: WM	M'	F3: VS	NS	F4: PS	Sc	F5		
Verbal Comprehension (VC	ension (VC)	Ι										
Working Memory (WM)	mory (WM)	.727	2	Ι								
Perceptual Reasoning (PR)	soning (PR)	.739	~	.685	5	I						
Processing	Processing Speed (PS)	.395		.504	4	.436	9	Ι				
	F5	.564		.474	4	.529	6	.239	6	Ι		
<i>Note.</i> ¹ Factor structure coefficients from first unrotated factor (g-loadings) are correlations between the subtest and the general factor. WISC–V Subtests: SI = Similarities, VC = Vocabulary, IN = Information, CO = Comprehension, BD = Block Design, VP = Visual Puzzles, MR = Matrix Reasoning, FW = Figure Weights, PC = Picture Concepts, AR = Arithmetic, DS = Digit Span, PS = Picture Span, LN = Letter–Number	ure coefficients larities, $VC = V$ igure Weights,	s from first u /ocabulary,] PC = Pictur	Inrotated factor (g- IN = Information, e Concepts, AR =	actor (g-loi mation, C(s, AR = AI	adings) ar) = Comp ithmetic,	loadings) are correlations betwee CO = Comprehension, BD = Blo Arithmetic, DS = Digit Span, PS	ons betwe , BD = Bl it Span, Pt	en the subtest and the geock Design, VP = Visua S = Picture Span, LN = 1	est and the , $VP = V_{1i}$ Span, LN	e general factor. W sual Puzzles, MR = Letter-Number	actor. WIS s, MR = N Vumber	C–V Iatrix
Sequencing, CD = Coding, SS = Symbol Search, CA = Cancellation. $S = Structure Coefficient$, $P = Pattern C Salient pattern coefficients presented in bold (pattern coefficient \geq .30). The eigenvalue for factor six was .69$	Coding, SS = S ficients present	ymbol Searc ed in bold (J	ch, CA = C pattern coe	= Cancellation. coefficient $\ge .3$	n. $S = Str$ 30). The	S = Structure Coefficient, .). The eigenvalue for facto	efficient, a for facto	P = Patternr six was .6	= Pattern Coefficient, h^2 = six was .69.		Communality	

Table B5 Wechsler Intelligence Scale for Children–Fifth Edition (WISC–V) Exploratory Factor Analysis: Five Obliaue Factor Solution for the

Table B6						
Sources of Var	ources of Variance in the Wechsler In	ttel	ale for Children-F	ifth Edition (WISC-	V) for the Standardiz	ligence Scale for Children–Fifth Edition (WISC–V) for the Standardization Sample 12–14 Year Olds
(N = 600) Acco	ording to an Explor	atory SL Bifactor M	Model with Five Fi	Model with Five First-Order Factors		
WISC-V	General	$F1 \cdot VC$	F2. WM	F3· PR	F4· PS	FS

(N = 600) According to an Exploratory SL Bi	rding to c	un Explor	atory SL B		Model wit	h Five F	factor Model with Five First-Order Factors	Factor	S				5 - - 1	
WISC-V	Genera	eral	F1: V(VC	F2: WM	ΜV	F3:	PR	F4: PS	PS	F	5		
Subtest	q	S^2	q	S^2	q	S^2	q	S^2	q	S^2	q	S^2	h^2	u^2
SI	.746	.557	.367	.135	.011	000^{-1}	.043	.002	004	000^{-1}	.013	000 [.]	.693	.307
VC	.757	.573	.443	.196	003	000.	016	000 [.]	018	000 [.]	.010	000.	.770	.230
N	.729	.531	.396	.157	028	.001	.064	.004	040	.002	016	000 [.]	.695	.305
CO	.645	.416	.387	.150	.012	000.	067	.004	.080	.006	003	000 [.]	.577	.423
BD	.641	.411	.027	.001	015	000.	.391	.153	.095	600.	076	.006	.580	.420
VP	.678	.460	015	000 ⁻	034	.001	.468	.219	-000	000.	.030	.001	.681	.319
MR	.633	.401	.072	.005	.087	.008	.164	.027	.035	.001	.122	.015	.456	.544
FW	.680	.462	.077	900.	.116	.013	.257	.066	055	.003	012	000.	.551	.449
PC	.531	.282	.010	000 ⁻	008	000.	011	000	.002	000 ⁻	.724	.524	.806	.194
AR	.700	.490	.156	.024	.147	.022	.123	.015	.008	000.	003	000.	.551	.449
DS	.687	.472	005	000 ⁻	.495	.245	009	000	040	.002	038	.001	.720	.280
PS	.603	.364	044	.002	.348	.121	.078	900.	027	.001	.051	.003	.496	.504
LN	.678	.460	690.	.005	.430	.185	068	.005	.021	000.	015	000 [.]	.655	.345
CD	.392	.154	.002	000	.043	.002	063	.004	.702	.493	.014	000.	.652	.348
SS	.440	.194	024	.001	.020	000.	.074	.005	.585	.342	-000	000.	.542	.458
CA	.221	.049	.029	.001	088	.008	.039	.002	.419	.176	002	000.	.235	.765
Total S^2		.392		.043		.038		.032		.065		.034	.604	.396
Common S^2		.649		.071		.063		.053		.107		.057		
Note. WISC-V Subtests: SI = Similarities,	Subtests	: SI = Si	milarities	VC = 1	Vocabulary	y, IN =	Information	on, CO	= Compre	Comprehension,	BD = BI	Block Design	gn, VP =	- Visual
Puzzles, MR = Matrix Reasoning, FW = Fig	Matrix R	easoning.	FW = Fi		tre Weights, PC =	= Picture (Concepts,	, AR =	Arithmetic, DS =	c, DS = I	Digit Span, PS =	, $PS = Pi$	Picture Span,	n, LN =
Letter–Number Sequencing, CD = Coding, S	Sequenci	ing, CD =	= Coding,	= S	Symbol Search,	ch, CA	= Cancellation.	· .	WISC-V F	Factors: V	C = Verb	al Compi	Verbal Comprehension,	WM =
Working Memory, VS = Visual Spatial, PS	ory, VS =	= Visual	Spatial, H	<u></u>	Processing S	Speed, FR	X = Fluid	Fluid Reasoning.	ing. b	= loading	of subtes	subtest on factor,	<i>S</i> ² =	variance
explained, $h^2 = \text{communality}$, $u^2 = \text{uniqueness}$.	- commu	nality, <i>u</i> '	f = uniqu	ness.	Bold type	indicate	indicates coefficients and	ents and	va	estimate		ent with	consistent with the theoretically	retically
proposed factor. Italic type indicates coeffici	. Italic ty	pe indica	ttes coeffi	ents	d variance	estimat	and variance estimates associated with	ited with	n an altern:	ate facto	alternate factor (where cross-loading b was larger	ross-loa	ding $b w_{3}$	العامة العامة الم رو
factor model for the theoretically assigned factor). U	oretically	assignec	I factor).	ven	e inadequa	cy of a	tive-facto	r solutic	the inadequacy of a five-factor solution Omega coefficients were not estimated for the five-	coefficie	ints were 1	not estim	ated for t	le live-

factor model.

F1: Ver		F1: Verbal	rbal	F2: Working	rking	F3: Perceptual	ceptual	F4: Processing	essing	F5		
	General ¹	Comprehension	ension	Memory	ory	Reasoning	ning	Speed	pa	Inadequate	luate	
WISC-V Subtest	S	Р	S	P	S	Р	S	Р	S	Р	S	h^2
SI	.775	1.022	.886	.018	589.	100	.603	.027	.262	140	.401	.807
VC	.811	.649	.841	.054	.629	.179	.716	084	.199	.075	.582	.735
IN	.749	.637	.786	073	.542	.117	.663	035	.185	.208	.607	.659
CO	679.	.622	.709	002	.514	.053	.578	.063	.263	.053	.452	.511
BD	.729	014	.591	091	.513	606 .	.832	.109	.426	078	.448	.712
VP	.708	.085	.611	051	.511	.726	.772	008	.292	.034	.509	.601
MR	.641	.147	.566	.220	.555	.451	.629	039	.257	104	.377	.440
FW	629.	.221	.630	077	.533	.470	.680	118	.170	.063	.513	.512
PC	.537	.183	.487	.044	.435	030	.468	.091	.200	.468	.589	.389
AR	.800	.194	.704	.265	.703	.251	.731	.050	.332	.217	.634	.660
DS	.715	076	.553	.814	.825	.171	.601	.019	.370	111	.390	969.
PS	.613	.070	.504	.612	.684	057	.477	079.	.320	.059	.400	.476
LN	.671	.036	.553	.876	.818	149	.493	047	.251	.068	.458	.683
CD	.388	.039	.247	.049	.332	.016	.328	.725	.753	068	.074	.572
SS	.449	.074	.306	030	.354	.017	.390	.764	.786	.047	.172	.627
CA	.244	158	.118	.002	.206	.001	.226	.447	.434	.259	.220	.233
Eigenvalue		7.	7.24	1.	1.61	1	.03	0.	0.82	0.	0.76	
% Variance		42.88	88	7.	25	4	4.17	2.	2.53	1.	.38	
Factor Correlations		F1: VC	/C	F2: WN	MV	F3: VS	VS	F4: PS	S	F5		
Verbal Comprehension (VC	ension (VC)	I										
Working Memory (WM	mory (WM)	.688	8	I								
Perceptual Reasoning (PR	tsoning (PR)	.751	_	.67	74	I						
Processing	Processing Speed (PS)	.278	8	.388	88	.674	4	I				
	F5	.578	8	.53	34	.38	8	.114	4	I		
Note. ¹ Factor structure coefficients from first unrotated factor (g-loadin Subtests: SI = Similarities, VC = Vocabulary, IN = Information, CO = Desconing EW - Figure Weighte, DC - Distribute Concerts, AD - Arithm	ure coefficient larities, VC =	ts from first u Vocabulary, DC – Distant	Inrotated factor (IN = Information Concerts AD	actor (g-lo mation, C(loadings) are corre CO = Comprehens	e correlation prehension	lations betweel ion, BD = Bloc		the subtest and the g k Design, VP = Visua - Dicture Scon, I N -	e general factor. W sual Puzzles, MR	actor. WIS s, MR = N	SC-V Matrix
Sequencing, CD = Coding, SS = Symbol Search	Coding, SS = 9	Symbol Searc	, CA	at	S	= Structure Coefficient	efficient,		Coefficie	nt, $h^2 = Co$	Communality	
Salient pattern coefficients presented in bold (pattern coefficient \geq .30). The eigenvalue for factor six was .68	ficients preser	ted in bold (pattern coe	efficient ≥	.30). The	eigenvalue	e for facto	r six was .6	8.		,	
•	•	,	_)						

Table B7 Wechsler Intelligence Scale for Children–Fifth Edition (WISC–V) Exploratory Factor Analysis: Five Obliaue Factor Solution for the

F3: PR F4: PS $F3$ $F3$ $F4$ $F3$ $F4$ $F5$	neral F1: VC F2: WM F3: PK F4: PS S^2 b S^2 b S^2 b S^2 .471 .536 .287 .011 .000 044 .002 .025 .001 .471 .536 .287 .011 .000 044 .002 .025 .001 -		1 ²
test b S^2 b <th< th=""><th>S^2 b S^2 b S^2 b S^2 b S^2 b S^2 b S^2 .471 .536 .287 .011 .000044 .002 .025 .001 - 503 348 171 032 001 078 006 078 006</th><th></th><th>n^2</th></th<>	S^2 b S^2 b S^2 b S^2 b S^2 b S^2 b S^2 .471 .536 .287 .011 .000044 .002 .025 .001 - 503 348 171 032 001 078 006 078 006		n^2
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$.471 .536 .287 .011 .000044 .002 .025 .001 - 503 348 171 032 001 078 006 078 006		и
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	503 348 131 033 001 078 006 078 006	.011	
$\begin{array}{l l l l l l l l l l l l l l l l l l l $	000, 0/0,- 000, 0/0, 100, 200, 121, 0+C. <i>CC</i> .	.003	•
$(630 \ 397)$ $334 \ 112$ $.001 \ 000$ $.058 \ 0.03$ $.040$ $.722 \ 521$ $.008 \ 000$ $.054 \ 0.03$ $.397 \ 158$ $.101 \ 0.10$ $.059 \ 0.05$ $.703 \ 494$ $.046 \ 0.02$ $.003 \ 0.01$ $.317 \ 100$ $.007 \ 0.00$ $.059 \ 0.05$ $.703 \ 494$ $.079 \ 0.06$ $.131 \ 0.17$ $.197 \ 0.39$ $.007 \ 0.00$ $.026 \ 0.01$ $.668 \ 446$ $.119 \ 0.14$ $.046 \ 0.02$ $.001 \ 0.17$ $.197 \ 0.39$ $.001 \ 0.02$ $.506 \ 256 \ 0.98 \ 0.10$ $.014 \ 0.02$ $.025 \ 0.01$ $.013 \ 0.01$ $.013 \ 0.01$ $.078 \ 0.01$ $.7764 \ 584 \ .104 \ 0.11$ $.014 \ 0.02$ $.025 \ 0.01$ $.013 \ 0.02$ $.013 \ 0.01$ $.013 \ 0.02$ $.041 \ 0.02$ $.764 \ 584 \ .104 \ 0.02$ $.041 \ .012$ $.013 \ 0.02$ $.013 \ 0.01$ $.013 \ 0.01$ $.013 \ 0.01$ $.013 \ 0.01$ $.013 \ 0.01$ $.013 \ 0.02$ $.041 \ .002$ $.041 \ .023 \ .031$ $.764 \ .202 \ .011 \ .000 \ .000 \ .001 \ .000 \ .001 \ $.501 .342 .117043 .002 .051 .003032 .001	.024	•
$\begin{array}{cccccccccccccccccccccccccccccccccccc$.397 .334 .112 001 .000 .023 .001 .058 .003	.002	•
$\begin{array}{cccccccccccccccccccccccccccccccccccc$.521008 .000054 .003 .397 .158 .101 .010		.304
$\begin{array}{cccccccccccccccccccccccccccccccccccc$.494 .046 .002030 .001 .317 .100 007 .000	.001	
668 446 119 014 046 002 205 042 -109 012 047 506 256 098 010 026 001 -013 000 084 007 351 764 584 $.104$ 011 $.158$ 025 $.110$ 012 $.046$ $.007$ $.361$ $.764$ $.584$ $.1041$ $.002$ $.485$ $.235$ $.012$ $.046$ $.007$ $.163$ $.677$ $.458$ $.205$ $.110$ $.012$ $.006$ $.018$ $.000$ $.084$ $.007$ $.033$ $.568$ $.323$ $.019$ $.000$ $.522$ $.272$ $.065$ $.004$ $.003$ $.063$ $.044$ $.003$ $.063$ $.044$ $.003$ $.044$ $.003$ $.063$ $.044$ $.063$ $.044$ $.003$ $.063$ $.044$ $.063$ $.044$ $.063$ $.044$ $.063$ $.044$ $.063$ $.064$ $.061$ $.000$ $.061$ <	.388 .079 .006 .131 .017 .197 .039 036 .001	.006	
506 256 $.098$ $.010$ $.026$ $.001$ $.013$ $.000$ $.084$ $.007$ $.351$ $.764$ $.584$ $.104$ $.011$ $.158$ $.025$ $.110$ $.012$ $.046$ $.002$ $.163$ $.163$ $.677$ $.458$ 041 $.002$ $.485$ $.235$ $.075$ $.006$ $.018$ $.000$ $.063$ $.568$ $.323$ $.001$ $.365$ $.133$ 025 $.001$ $.003$ $.044$ $.084$ $.5627$ $.393$ $.001$ $.365$ $.133$ 025 $.001$ $.003$ $.044$ $.083$ $.516$ $.100$ $.029$ $.001$ $.002$ $.001$ $.002$ $.044$ $.051$ $.316$ $.100$ $.022$ $.001$ $.007$ $.000$ $.071$ $.062$ $.051$ $.316$ $.100$ $.022$ $.001$ $.007$ $.000$ $.051$ $.051$ $.051$ $.051$ $.051$ $.051$ $.051$ $.051$ <td>.446 .119 .014 .046 .002 .205 .042 -.109 .012</td> <td>.002</td> <td></td>	.446 .119 .014 .046 .002 .205 .042 - .109 .012	.002	
$\begin{array}{cccccccccccccccccccccccccccccccccccc$.256 .098 .010 .026 .001013 .000 .084 .007	.123	•
$\begin{array}{cccccccccccccccccccccccccccccccccccc$.584 .104 .011 .158 .025 .110 .012 .046 .002	.027	
.568 .323 .038 .001 .365 .133 .025 .001 .073 .005 .044 .627 .393 .019 .000 .522 .272 .065 .004 .044 .005 .051 .627 .393 .019 .000 .522 .272 .265 .004 044 .002 .051 .316 .100 .021 .000 .029 .001 .007 .000 .051 .051 .373 .139 .040 .002 .018 .000 .007 .000 .057 .051 .051 .209 .044 .001 .000 .000 .000 .011 .195 .195 .382 .043 .043 .023 .073 .073 .073 .073	7 .458041 .002 .485 .235 .075 .006 .018 .000	.007	
.627 .393 .019 .000 .522 .272 .265 .004 .044 .002 .051 . .316 .100 .021 .000 .029 .001 .007 .000 .671 .450 .051 . .373 .139 .040 .002 .018 .000 .007 .000 .057 .051 . .209 .044 085 .007 .000 .000 .000 .035 . .209 .044 .085 .001 .000 .000 .000 .035 . .382 .043 .043 .023 .073 . .073 .	.323 .038 .001 .365 .133 025 .001 .073 .005	.002	•
.316 .100 .021 .000 .029 .001 .007 .000 .671 .450 051 . .373 .139 .040 .002 018 .000 .007 .000 .707 .500 .035 . .209 .044 085 .001 .000 .000 .000 .035 . . .195 . .209 .044 085 .007 .000 .000 .000 .035 . . .171 .195 . .382 .043 .023 .023 .023 .073 . .073 . .	393 .019 .000 .522 .272 065 .004044 .002	.003	·
.373 .139 .040 .002 018 .000 .007 .000 .035 .035 .035 .209 .044 085 .007 .000 .000 .000 .035 .073	.100 .021 .000 .029 .001 .007 .000 .671 .450 -	.003	·
.209 .044085 .007 .001 .000 .000 .000 .414 .171 .195 . .382 .043 .043 .043 .023 .073 .	.139 .040 .002018 .000 .007 .000 .707 .500	.001	.358
.382 .043 .043 .023 .073	.044085 .007 .001 .000 .000 .000 .414 .171	•	
	.043 .043 .023	.016 .581	.419
.040 .126	.075 .075 .040	.027	

Table B8