PSİKOLOJİ ALANINDA R KULLANIMI

Why R? Turkey 2021 17/04/2021 Eren Halil ÖZBERK





İTİM

HACETTEPE ÜNİVERSİTESİ



NEYİM









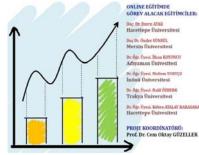


R ile Veri Analizi ve Psikometri Uygulamaları

Burcu ATAR - Kübra ATALAY KABASAKAL Elif Bengi ÜNSAL ÖZBERK - Eren Halil ÖZBERK Nermin KIBRISLIOĞLU UYSAL











EPOD2016 Kongresi - R Yaz Giriş Çalıştayı

y Follow



Serhat Kalkınma Ajansı -Uygulamalı R Yazılımı Veri

y Follow







rather surprising that systematic studies of human abilities were not undertaken until the and half of the last century... An accurate method was available for measuring the umference of the earth 2,000 years before the first systematic measures of human ability be developed

Jum Nunnally, Psychometric Theory (1967)

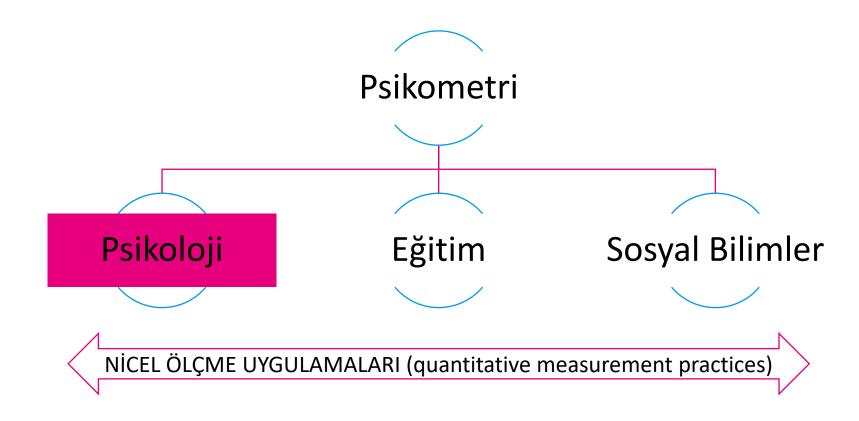
chometry, it is hardly necessary to say, means the art of imposing measurement and nber upon operations of the mind ..."

Francis Galton, "Psychometric Experiments" (1879)



KOMETRI NEDIR?

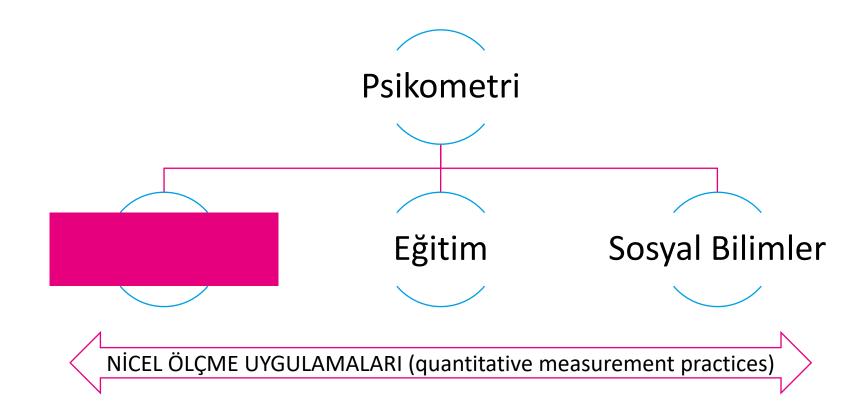






KOMETRI NEDIR?







KOMETRI NEDIR?

TURKEY

Ölçme araçlarının geliştirilmesi

Gözlenen değişkenler (Zeka testi maddeleri) ile teorik yapıların (Zeka) ilişkilendirilmesi

Yeteneğin belirlenmesi



NGİ YÖNTEMLER



Psikolojik Ölçekleme (Psychological Scaling)

Stevens (1951) Ölçme objelere ve olayları numaralandırma anlamına gelir.

Thurstone, bu numaralandırmayı standartlaştırmak için «scale» kavramı üzerinde duruyor.

Psikometrik Ölçme (Psychological measurement-test theory)

Gerçek Puanlar Teorisi (True Score Theory) – Güvenirlik Geçerlik

Madde Tepki Kuramı (Item Response Theory) –Rasch, Parameter based LM, GRM,PCM,NRM

DIF, DMF Değişen madde fonksiyonu

Faktör Analizi

Temel Bileşenler Analizi

Kovaryans Yapıları

Örtük Sınıf/Profil Analizleri

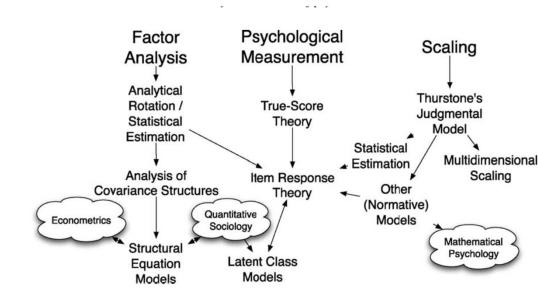
Yapısal Eşitlik Modeli

Psikolojik İstatistikler

Survey /Anket

Betimsel

Veri Görselleştirme



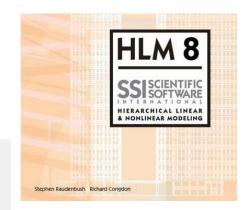
Jones, L. V., & Thissen, D. (2006). 1 A History and Overview of Psychometrics. Handbook of statistics, 26, 2





About























Xcalibre 4.2

IRT Item parameter calibration

Copyright © 2014 - Assessment Systems Corporation



TAP : Test Analysis Program v.19.1.4







Output kullanımı

Model Summary

| | П | | | Adjusted R | Std. Error of | Change Statistics | | | | | | |
|-----|----|---------------------------|------|--------------|-----------------|-------------------|--------------|---|---------------|------|--|--|
| Mod | el | R R Square Square the Est | | the Estimate | R Square Change | F Change | F Change df1 | | Sig. F Change | | | |
| 1 | | .497ª | .247 | .232 | 20.362 | .247 | 15.767 | 1 | 48 | .000 | | |
| 2 | | .614 ^b | .377 | .351 | 18.719 | .130 | 9.796 | 1 | 47 | .003 | | |
| 3 | | .680° | .462 | .427 | 17.587 | .085 | 7.248 | 1 | 46 | .010 | | |
| 4 | | .691 | .477 | .431 | 17.521 | .016 | 1.344 | 1 | 45 | .252 | | |
| 5 | | .695 ^e | .483 | .424 | 17.631 | .005 | .441 | 1 | 44 | .510 | | |

Predictors: (Constant), I have good labor conditions

Predictors: (Constant), I have good labor conditions, My work is interesting

- c. Predictors: (Constant), I have good labor conditions, My work is interesting, My workplace is good
- d. Predictors: (Constant), I have good labor conditions, My work is interesting, My workplace is good, I have nice colleagues
- e. Predictors: (Constant), I have good labor conditions, My work is interesting, My workplace is good, I have nice colleagues, I have a nice supervisor

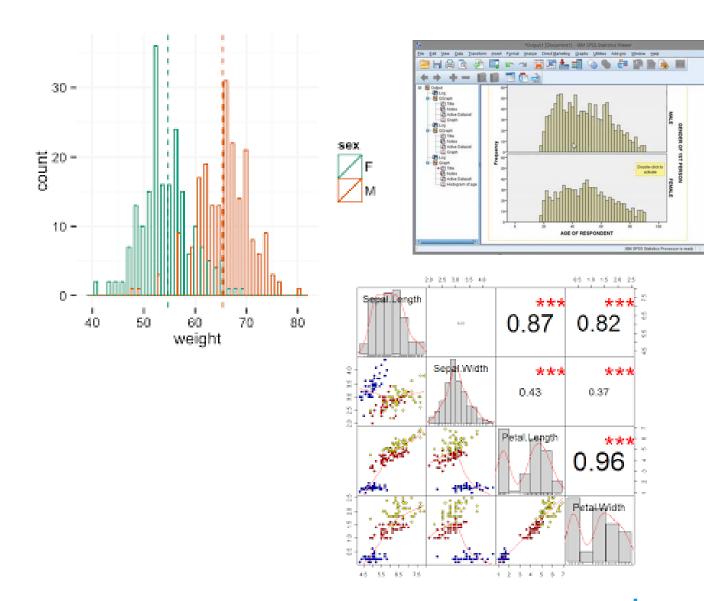
Correlations

| | | Wechsel IQ Test Score | Depression Test Score | Anxiety Test Score | Social Functioning Test Score | Ge Beir |
|-----------------------------|------------------------|--------------------------|--------------------------|-----------------------|----------------------------------|------------|
| Wechsel IQ Test Score | Pearson Correlation | 1 | .121 | .152 | 151 | |
| | Sig. (2-tailed) | | .194 | .110 | .104 | |
| | N | 122 | 116 | 112 | 117 | |
| Depression Test Score | Pearson Correlation | .121 | 1 | .282** | 328 ^{**} | |
| | Sig. (2-tailed) | .194 | | .003 | .000 | |
| | N | 116 | 122 | 112 | 116 | |
| Anxiety Test Score | Pearson Correlation | .152 | .282** | 1 | 534** | |
| | Sig. (2-tailed) | .110 | .003 | | .000 | |
| | N | 112 | 112 | 117 | 111 | |
| Social Functioning | Pearson Correlation | 151 | 328** | 534** | 1 | |
| Test Score | Sig. (2-tailed) | .104 | .000 | .000 | | |
| | N | 117 | 116 | 111 | 122 | |
| Generall Well Being Test | Pearson Correlation | 080 | 801** | 296** | .333** | |
| Score | Sig. (2-tailed) | .393 | .000 | .001 | .000 | |
| | N | 117 | 117 | 113 | 117 | |

^{**.} Correlation is significant at the 0.01 level (2-tailed). © 2017 www.spss-tutorials.com

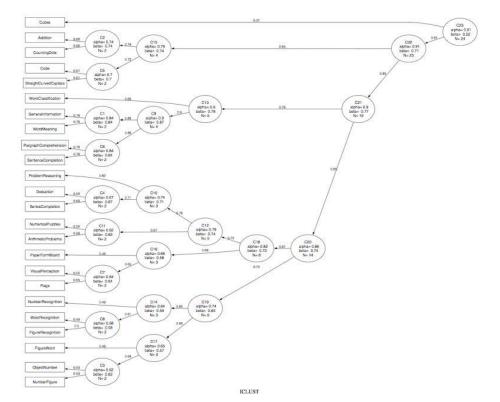


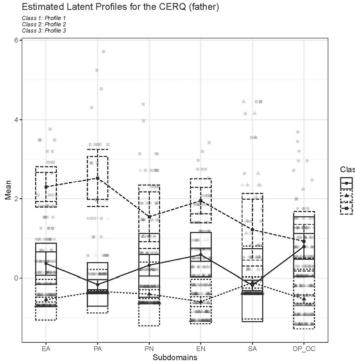
Output kullanımı Veri görselleştirme





Output kullanımı Veri görselleştirme Psikometrik analizler

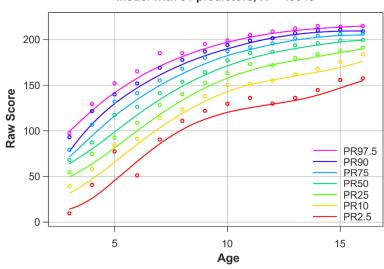




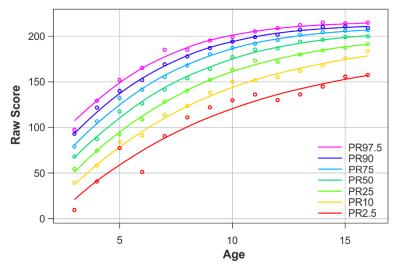


Output kullanımı Veri görselleştirme Psikometrik analizler Yaş Normları

Observed and Modeled Percentile Curves Model with 31 predictors, R² = .9949



Observed and Modeled Percentile Curves Model with 7 predictors, R² = .9929







Output kullanımı
Veri görselleştirme
Psikometrik analizler
Yaş Normları
Denklemlerin Kontrolü

The equation for the threeparameter model is:

$$P(\theta) = c + (1-c)\frac{1}{1+e^{-a(\theta-b)}}$$

where: b is the difficulty parameter
a is the discrimination parameter
c is the guessing parameter and
è is the ability level



```
ers\ozber\Downloads\Read R Parscale.r - Notepad++
7) Düzenle Ara Görünüm Kodlama Diller Ayarlar Araçlar Makrolar Çalıştır Eklentiler Pencereler
H 🖺 🖺 🖟 🔚 X 🐚 🖍 🗩 C I 🏔 🐆 🔍 🤏 🖳 🔜 📑 1 👺 🐷 💹 🔎 🚞 👁 🗨 🗉 🕩 🖼
R Parscale.r 🔣
# Notice the function returns whatever variable is at the end
# Skipper tells how many lines to skip at the top of a file
readParscale <- function(file and path, skipper){
  #Line lengths 77, 20, 20
   IRTPars <- read.fwf(file and path,</pre>
      width=list(c(7,-13,7,-3,7,-2,8,-3,7,-4,7,-2,7)), c(-2,8,-2,8,-2,8), c(-1)), skip=skipper,
      col.names=c("ItemID", "slope", "slopeSE",
                   "loc", "locSE", "guess", "guessSE", "loc1", "loc2", "loc3"), as.is=T)
IRTPars
readParscale ("RN.par", 9)
# An improvement to automatically determine the number of skip lines
readParscale <- function(file and path){
  FindSkip <- read.fwf(file and path, 8, as.is=TRUE)
  FindSkip <- FindSkip == "GROUP 01"
  FindSkip <- order (FindSkip, decreasing=TRUE) [1]
  #Line lengths 77, 20, 20
   IRTPars <- read.fwf (file and path,
      width=list(c(7,-13,7,-3,7,-2,8,-3,7,-4,7,-2,7)), c(-2,8,-2,8,-2,8), c(-1)), skip=FindSkip,
      col.names=c("ItemID", "slope", "slopeSE",
                   "loc", "locSE", "quess", "quessSE", "loc1", "loc2", "loc3"), as.is=T)
IRTPars
```





KOLOJİNİN ÇALIŞMA ALANLARI (PSİKOLOJİ ÖZELİNDE)



Örgütsel

Davranış Bilimleri

Nöropsikoloji

Klinik

Sosyal







International Journal of Assessment Tools in Education

2021, Vol. 8, No. 2, 296-309

https://doi.org/10.21449/ijate.728362

Published at https://ijate.net/

https://dergipark.org.tr/en/pub/ijate

Research Article

Examining the Dimensionality and Monotonicity of an Attitude Dataset based on the Item Response Theory Models

Seval Kula Kartal ^{1,*}, Ezgi Mor Dirlik ²









International Journal of Assessment Tools in Education

2020, Vol. 7, No. 1, 18-29

https://dx.doi.org/10.21449/ijate.629584

Published at http://www.ijate.net

http://dergipark.org.tr/en/pub/ijate

Research Article

Use of Item Response Theory to Validate Cyberbullying Sensibility Scale for University Students

Osman Tolga Arıcak 📵¹, Akif Avcu 📵²,*, Feyza Topçu 📵¹, Merve Gülçin Tutlu 📵¹





Psychological Methods 2003, Vol. 8, No. 2, 164–184 Copyright 2003 by the American Psychological Association, Inc. 1082-989X/03/\$12.00 DOI: 10.1037/1082-989X.8.2.164

How Many IRT Parameters Does It Take to Model Psychopathology Items?

Steven P. Reise University of California, Los Angeles Niels G. Waller Vanderbilt University

The authors compared the fit of the 2- and 3-parameter logistic models (2PLM; 3PLM) on 15 unidimensional factor scales derived from the Minnesota Multiphasic Personality Inventory—Adolescent item pool. Log-likelihood chi-square deviance tests indicated that a 3PLM provided an improved fit. However, residual statistics indicated that the difference in fit between the 2 models was negligible. An unexpected finding was that from 10% to 30% of the items had substantial lower asymptote parameters ($c \ge .10$) when the scales were scored in the pathology or nonpathology directions. The authors argue that the large lower asymptote parameters



TURKEY

Article

A Demonstration of Mokken Scale Analysis Methods Applied to Cognitive Test Validation Using the Egyptian WAIS-IV

Journal of Psychoeducational Assessment

|-|4|
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Gomaa S. M. Abdelhamid^{1,2}, Juana Gómez-Benito¹, Ahmed T. M. Abdeltawwab², Mostafa H. S. Abu Bakr², and Amina M. Kazem³

Abstract

The fourth edition of the Wechsler Adult Intelligence Scale (WAIS-IV) has been used extensively for assessing adult intelligence. This study uses Mokken scale analysis to investigate the psychometric proprieties of WAIS-IV subtests adapted for the Egyptian population in a sample of 250 adults between 18 and 25 years of age. The monotone homogeneity model and the double monotonicity model were consistent with the subtest data. The items of all subtests except Matrix Reasoning, Information, Similarities, and Vocabulary formed a unidimensional scale. The WAIS-IV subtests have discriminatory and invariantly ordered items, although some items violated the invariant item ordering and scalability criteria. Therefore, the WAIS-IV subtests—with the exception of some items—are hierarchical scales that allow items to be





Application of the Double Monotonicity Model to Polytomous Items

Scalability of the Beck Depression Items on Subjects with Eating Disorders

Teresa Rivas, Rosa Bersabé, and Carmen Berrocal

Facultad de Psicología, Universidad de Málaga, Spain

Abstract. This paper investigates the item scalability of the Beck Depression Inventory (BDI) in 252 subjects; 126 with and 126 without eating disorders. To do so, an order was established regarding the BDI items according to the clinical characteristics of the subjects with eating disorders. The nonparametric Item Response Theory (NIRT) model was applied to evaluate Monotone Homogeneity and Double Monotonicity of items, as well as the reliability of the scale in both groups. The results show that the order of the items is satisfied in the group with eating disorders, but not in the control group. Therefore, the results obtained allow the ordering of depression scores of subjects with eating disorders according to their clinical characteristics. This order is not valid for the depression scores of subjects who did not have eating disorders. It should be noted that the application of the Double Monotonicity model to polytomous items provides new and relevant information when compared to the data provided by the Classical Test Model. In addition, it is very useful for other items and subjects having certain characteristics.





Article

Exploratory Mokken Scale Analysis as a Dimensionality Assessment Tool: Why Scalability Does Not Imply Unidimensionality

Applied Psychological Measurement 36(6) 516–539
© The Author(s) 2012
Reprints and permission: sagepub.com/journalsPermissions.nav
DOI: 10.1177/0146621612451050
http://apm.sagepub.com

Iris A. M. Smits¹, Marieke E. Timmerman¹, and Rob R. Meijer¹





Personality and Individual Differences 91 (2016) 89-97



Contents lists available at ScienceDirect

Personality and Individual Differences

journal homepage: www.elsevier.com/locate/paid



Psychometric qualities of the Thought Suppression Inventory-Revised in different age groups



Kevin van Schie a,b,*,1, Sabine Wanmaker b, Iris Yocarini b, Samantha Bouwmeester b



^a Clinical Psychology, Utrecht University, PO Box 80140, NL-3508 TC Utrecht, The Netherlands

^b Institute of Psychology, Erasmus University Rotterdam, PO Box 1738, 3000 DR Rotterdam, The Netherlands

KETLER



psych: Procedures for Psychological, Psychometric, and Personality
Research

A general-purpose toolbox for personality, psychometric theory and experimental psychology. Functions are primarily for multivariate analysis and scale construction using factor analysis, principal component analysis, cluster analysis and reliability analysis, although others provide basic descriptive statistics. Item Response Theory is done using factor analysis of tetrachoric and polychoric correlations. Functions for analyzing data at multiple levels include within and between group statistics, including correlations and factor analysis. Functions for simulating and testing particular item and test structures are included. Several functions serve as a useful front end for structural equation modeling. Graphical displays of path diagrams, factor analysis and structural equation models are created using basic graphics.

NetworkToolbox: Methods and Measures for Brain, Cognitive, and Psychometric Network Analysis

Implements network analysis and graph theory measures used in neuroscience, cognitive science, and psychology. Aims to provide researchers with state-of-the-art methods and measures for estimating and analyzing brain, cognitive, and psychometric networks.



KETLER



psychometric: Applied Psychometric Theory

Contains functions useful for correlation theory, meta-analysis (validity-generalization), reliability, item analysis, inter-rater reliability, and classical utility

lavaan: Latent Variable Analysis

Fit a variety of latent variable models, including confirmatory factor analysis, structural equation modeling and latent growth curve models.

sem: Structural Equation Models

Functions for fitting general linear structural equation models (with observed and latent variables) using the RAM approach, and for fitting structural equations in observed-variable models by two-stage least squares.

cNORM: Continuous Norming

Conventional methods for producing standard scores in psychometrics or biometrics are often plagued with "jumps" or "gaps" (i.e., discontinuities) in norm tables and low confidence for assessing extreme scores.





McArdle and Prescott (1992, p.90) verisi

IN information

CO comprehension

SI similarities

VO vocabulary

PC picture completion

BD block design

PA picture arrangement

OA object assembly

Arithmetic, Memory Span for Digits, ve Digit Symbol Substitution dışarıda tutulmuştur.



Verbal



AcArdle and Prescott (1992, p.90) verisi

IN information

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PA picture arrangement

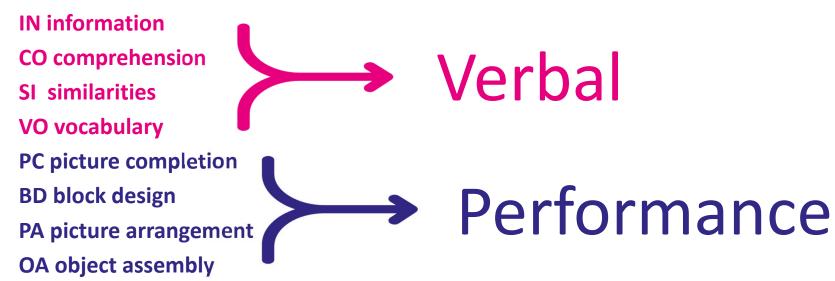
OA object assembly

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AcArdle and Prescott (1992, p.90) verisi

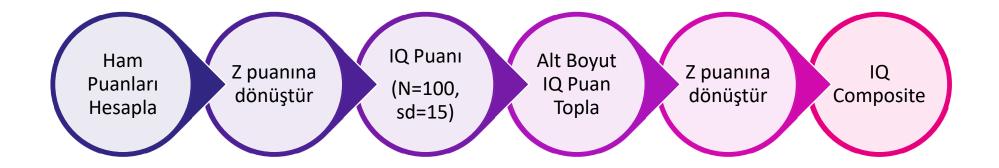


Arithmetic, Memory Span for Digits, ve Digit Symbol Substitution dışarıda tutulmuştur.

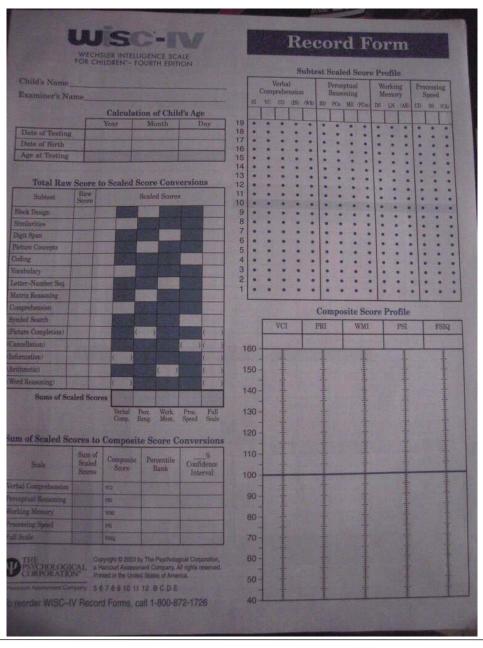




VAIS Zeka Puanı Hesaplama Yöntemi,











| - | _ | IAB | LE C | F SC | | w sc | | φυιν | ALEN' | TS* | | | | | 2000000 | MARY | Scaled | |
|-----------------|-------------|---------------|--------------------|----------------|------------------|----------------|----------------|-------------------------|---------------------|------------------------|------------------------|---------------------|---------------------|-----------------|--|-------------|-------------|-----------------|
| Scaled Score | | | | | KA | w sc | OKE | _ | | | | | Score Score | | | | | |
| | | sion | | | | | _ | | | | nbly | Scaled Score | 2000 | format | 2000 | | | |
| | Information | Comprehension | ofic | Hies | pan | lary | Digit Symbol | Picture Completion | Block Design | Picfure Arrangement | Object Assembly | Pale | | | hension | 100 | | |
| | | ошрг | Arithmetic | Similarities | Digit Span | Vocabulary | git Sy | ture | ck D | range | ject | Scc | | rithme | | | | |
| | 100 | | 4 | 1000 | | 1000 | 200 | 20 | iii | E A | Ö | - | | imilarit | 1320 | | | |
| 19 | 29 28 | 27-28 | | 26 25 | 17 | 78-80 76-77 | 87-90 | | | 36 | 44 | 19 | | Digit Span | | | the same | |
| 17 | 27 | 25 | 18 | 24 | | 74-75 | 79-82 | 2 | 48 | 35 | 43 | 18 | V | ocabul | SSA CONTRACTOR | | | |
| 16 | 26 25 | 24 23 | 17 | 23 22 | 16 | 71-73 67-70 | 76-78 | | 47 | 34 | 42 | 16 | | | 7/2//7/20 | bal Score | | |
| 14 | 23-24 | 22 | 15 | 21 | 14 | 63-66 | | | 44-45 | | 40 | 14 | | ligit Sy | And the second s | | | |
| 13 | 21-22 | 21 20 | 14 | 19-20 17-18 | 13 | 59-62 54-58 | | | | 30-31 | 38-39 36-37 | | | \$1970 CO | Completion | | | |
| 11 | 17-18 | 19 | 12 | 15-16 13-14 | 12 | 47-53 40-46 | | | | 26-27 23-25 | 34-35 | 11 10 | BI | lock D | esign | | | |
| 9 | 13-14 | 15-16 | 10 | 11-12 | 10 | 32-39 | 47-5 | 1 12-13 | 28-30 | 20-22 | 28-30 | 9 | | | Arrangement | | | The same of |
| 8 7 | 9-10 | 14 | 7-8 | 9-10 7-8 | 9 | 26-31 | 35-4 | | 25-27 | 18-19 | 25-27 | 8 7 | | | Assembly | | | |
| 6 | 7-8 | 10-11 | 6 | 5-6 | 8 | 18-21 | 29-3 | 4 6-7 | 17-20 | 12-14 | 19-21 | 6 | | Perform | | nce Score | | |
| 5 4 | 5-6 | 8-9 6-7 | 5 4 | 4 3 | 7 | 14-17 | 23-2 | | 13-16 | | 15-18 | 5 4 | | | | | - | |
| 3 | 3 | 5 | 3 | 2 | | 10 | 15-1 | 7 3 | 6-9 | 7 | 8-10 | 3 | | | | | | |
| 1 | 2 | 4 3 | 2 | 1 | 6 4-5 | 9 8 | 13-14 | 4 2 | 3-5 | 5 | 5-7 3-4 | 2 | PERFORMANCE SCOREIÇ | | | | _ | |
| 0 | 0 | 0-2 | 0 | 0 | 0-3 | 0-7 | 0-1 | | | | | | | | FULL SCALE S | | | - |
| linici ke ir | to accou | wish to | draw a eliabili | "psycho | graph' he sub | on the | above the I | e table n lower reli | nay do abilities | of diffe | onnecting erences b | the sul etween t | subtest | scores. | es. The Interpretation | of any such | profile, ho | wever, shou |
| | 1. | INFO | RMA | TION | | SCO | RE | | | | | | S | SCORE I or 0 | | | | SCORE For 0 |
| 1. | Flag | | | | | | | II. Height | | | | | | | 21. Senators | | | |
| 2. | Ball | | | | | | | 12. Italy | | | | | | | 22. Genesis | | | |
| 3. | Month | s | | | | | | 13. Clothes | | | | | | | 23. Temperatu | re | | |
| 4. | Therm | omete | r | | | | | 14. Washington | | | | | | | 24. Iliad | | | here. |
| 5. | Rubbe | г | | | | | | 15. Hamlet | | | | | | -0.1 | 25. Blood vessels | | | - |
| 6. | Preside | ents | | | | | | 16. Vatican | | | | | | | 26. Koran | | | |
| 7. | Longfe | llow | | | | | | 17. Paris | | | | | | | 27. Faust | | | |
| | | | | | | | | 18. Egypt | | | | | | | 28. Ethnology | | | |
| THE CONTRACTOR | | | | | | 19. Yeast | | | | | | | 29. Apocrypha | | | 19 1 | | |
| 9. | | | | | | | | 20. Po | pulatio | on | | | | | | | | Service Service |





Table 1
Conversion Table for Calculating Performance IQ When Symbol Search is Substituted for Coding on the WISC-III

| 0 | | B | Confidence | e intervals | 6 6 | | Decree 1 | Confidence intervals | | |
|----------------------|-----|--------------------|------------|-------------|----------------------|-----|--------------------|----------------------|--------|--|
| Sum of scaled scores | IQ | Percentile rank | 90% | 95% | Sum of scaled scores | IQ | Percentile rank | 90% | 95% | |
| 5 | 44 | | 42-56 | 41-57 | 51 | 102 | 55 | 95-109 | 94-11 | |
| 6 | 46 | < 0.1 | 44-58 | 43-59 | 52 | 103 | 58 | 95-109 | 95-11 | |
| 7 | 47 | < 0.1 | 45-58 | 44-60 | 53 | 104 | 61 | 97-110 | 96-11 | |
| 8 | 48 | < 0.1 | 46-59 | 45-61 | 54 | 105 | 66 | 98-111 | 96-11 | |
| 9 | 49 | < 0.1 | 47-60 | 45-62 | 55 | 107 | 68 | 100-113 | 98-11 | |
| 10 | 51 | 0.1 | 49-62 | 47-63 | 56 | 108 | 70 | 101-114 | 99-11 | |
| 11 | 52 | 0.1 | 49-63 | 48-65 | 57 | 109 | 75 | 101-115 | 100-11 | |
| 12 | 53 | 0.1 | 50-64 | 49-65 | 58 | 110 | 77 | 102-116 | 101-11 | |
| 13 | 54 | 0.1 | 51-65 | 50-66 | 59 | 112 | 79 | 104-118 | 103-11 | |
| 14 | 56 | 0.2 | 53-67 | 52-68 | 60 | 113 | 81 | 105-119 | 104-12 | |
| 15 | 57 | 0.3 | 54-68 | 53-69 | 61 | 114 | 84 | 106-119 | 105-12 | |
| 16 | 58 | 0.3 | 55-68 | 54-70 | 62 | 115 | 86 | 107-120 | 106-12 | |
| 17 | 59 | 0.4 | 56-69 | 55-71 | 63 | 116 | 87 | 108-121 | 106-12 | |
| 18 | 61 | 1 | 58-71 | 56-72 | 64 | 118 | 90 | 110-123 | 108-12 | |
| 19 | 62 | 1 | 59-72 | 57-73 | 65 | 119 | 91 | 110-124 | 109-12 | |
| 20 | 63 | 1 | 60-73 | 58-74 | 66 | 120 | 92 | 111-125 | 110-12 | |
| 21 | 64 | 1 | 60-74 | 59-75 | 67 | 121 | 94 | 113-127 | 112-12 | |
| 22 | 66 | 1 | 62-76 | 61-77 | 68 | 123 | 95 | 114-127 | 112-12 | |
| 23 | 67 | 2 | 63-77 | 62-78 | 69 | 124 | 96 | 115-128 | 114-13 | |
| 24 | 68 | 2 2 | 64-78 | 63-79 | 70 | 125 | 96 | 116-129 | 115-13 | |
| 25 | 69 | 2 | 65-78 | 64-80 | 71 | 126 | 97 | 117-130 | 115-13 | |
| 26 | 71 | 3 | 67-80 | 66-82 | 72 | 128 | 98 | 119-132 | 117-13 | |
| 27 | 72 | 3 | 68-81 | 66-83 | 73 | 129 | 98 | 119-133 | 118-13 | |
| 28 | 73 | 4 | 69-82 | 67-83 | 74 | 130 | 98 | 120-134 | 119-13 | |
| 29 | 74 | 4 | 70-83 | 68-84 | 75 | 131 | 99 | 121-135 | 120-13 | |
| 30 | 76 | 5 | 72-85 | 70-86 | 76 | 133 | 99 | 123-137 | 122-13 | |
| 31 | 77 | 6 | 72-86 | 71-87 | 77 | 134 | 99 | 124-137 | 123-13 | |
| 32 | 78 | 7 | 73-87 | 72-88 | 78 | 135 | 99 | 125-138 | 124-14 | |
| 33 | 79 | 8 | 74-88 | 73-89 | 79 | 136 | 99.5 | 126-139 | 124-14 | |
| 34 | 80 | 9 | 75-89 | 74-90 | 80 | 138 | 99.6 | 128-141 | 126-14 | |
| 35 | 82 | 10 | 77-90 | 76-92 | 81 | 139 | 99.7 | 129-142 | 127-14 | |
| 36 | 82 | 12 | 78-91 | 77-93 | 82 | 140 | 99.7 | 129-143 | 128-14 | |
| 37 | 83 | 13 | 79-92 | 78-94 | 83 | 141 | 99.8 | 130-144 | 129-14 | |
| 38 | 84 | 14 | 80-93 | 78-94 | 84 | 143 | 99.9 | 132-146 | 131-14 | |
| 39 | 85 | 18 | 82-95 | 80-96 | 85 | 144 | 99.9 | 133-147 | 132-14 | |
| 40 | 87 | 19 | 82-96 | 81-97 | 86 | 145 | 99.9 | 134-148 | 133-14 | |
| 41 | 89 | 23 | 83-97 | 82-98 | 87 | 146 | 99.9 | 135-148 | 134-15 | |
| 42 | 90 | 25 | 84-98 | 83-99 | 88 | 147 | > 99.9 | 136-149 | 134-15 | |
| 43 | 92 | 27 | 86-100 | 85-101 | 89 | 149 | > 99.9 | 138-151 | 136-1 | |
| 44 | 93 | 32 | 87-100 | 86-102 | 90 | 150 | > 99.9 | 139-152 | 137-1 | |
| 45 | 94 | 34 | 88-101 | 87-103 | 91 | 151 | > 99.9 | 139-153 | 138-1 | |
| 46 | 95 | 37 | 89-102 | 87-103 | 92 | 152 | > 99.9 | 140-154 | 139-1 | |
| 47 | 97 | 39 | 91-104 | 88-104 | 93 | 154 | > 99.9 | 142-156 | 141-1 | |
| 48 | 98 | 45 | 91-105 | 90-106 | 94 | 155 | > 99.9 | 143-157 | 142-1 | |
| 49 | 99 | 47 | 92-106 | 91-107 | 95 | 156 | > 99.9 | 144-159 | 143-1 | |
| 50 | 100 | 50 | 93-107 | 92-108 | | | - 7717 | | | |

Note. WISC-III = Wechsler Intelligence Scale for Children—III.





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