Package 'psychTools'

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Description Support functions, data sets, and vignettes for the 'psych' package. Contains several of the biggest data sets for the 'psych' package as well as one vignette. A few helper functions for file manipulation are included as well. For more information, see the https://personality-project.org/r/ web page.
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ability

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16 ability items scored as correct or incorrect.

Description

16 multiple choice ability items 1525 subjects taken from the Synthetic Aperture Personality Assessment (SAPA) web based personality assessment project are saved as iqitems. Those data are shown as examples of how to score multiple choice tests and analyses of response alternatives. When scored correct or incorrect, the data are useful for demonstrations of tetrachoric based factor analysis irt.fa and finding tetrachoric correlations.

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Usage

```
data(iqitems)
```

Format

A data frame with 1525 observations on the following 16 variables. The number following the name is the item number from SAPA.

```
reason.4 Basic reasoning questions
reason.16 Basic reasoning question
reason.17 Basic reasoning question
reason.19 Basic reasoning question
letter.7 In the following alphanumeric series, what letter comes next?
letter.33 In the following alphanumeric series, what letter comes next?
letter.34 In the following alphanumeric series, what letter comes next
letter.58 In the following alphanumeric series, what letter comes next?
matrix.45 A matrix reasoning task
matrix.46 A matrix reasoning task
matrix.47 A matrix reasoning task
matrix.55 A matrix reasoning task
rotate.3 Spatial Rotation of type 1.2
rotate.4 Spatial Rotation of type 1.1
```

Details

rotate.8 Spatial Rotation of type 2.3

16 items were sampled from 80 items given as part of the SAPA (https://www.sapa-project.org/) project (Revelle, Wilt and Rosenthal, 2009; Condon and Revelle, 2014) to develop online measures of ability. These 16 items reflect four lower order factors (verbal reasoning, letter series, matrix reasoning, and spatial rotations. These lower level factors all share a higher level factor ('g'). This data set may be used to demonstrate item response functions, tetrachoric correlations, or irt.fa as well as omega estimates of of reliability and hierarchical structure.

In addition, the data set is a good example of doing item analysis to examine the empirical response probabilities of each item alternative as a function of the underlying latent trait. When doing this, it appears that two of the matrix reasoning problems do not have monotonically increasing trace lines for the probability correct. At moderately high ability (theta = 1) there is a decrease in the probability correct from theta = 0 and theta = 2.

Source

The example data set is taken from the Synthetic Aperture Personality Assessment personality and ability test at https://www.sapa-project.org/. The data were collected with David Condon from 8/08/12 to 8/31/12.

Similar data are available from the International Cognitive Ability Resource at https://www.icar-project.org/.

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References

Condon, David and Revelle, William, (2014) The International Cognitive Ability Resource: Development and initial validation of a public-domain measure. Intelligence, 43, 52-64.

Revelle, William, Dworak, Elizabeth M. and Condon, David (2020) Cognitive ability in everyday life: the utility of open-source measures. Current Directions in Psychological Science, 29, (4) 358-363. Open access at doi:10.1177/0963721420922178.

Dworak, Elizabeth M., Revelle, William, Doebler, Philip and Condon, David (2021) Using the International Cognitive Ability Resource as an open source tool to explore individual differences in cognitive ability. Personality and Individual Differences, 169. Open access at doi:10.1016/j.paid.2020.109906. Revelle, William, Wilt, Joshua, and Rosenthal, Allen (2010) Personality and Cognition: The Personality-Cognition Link. In Gruszka, Alexandra and Matthews, Gerald and Szymura, Blazej (Eds.) Handbook of Individual Differences in Cognition: Attention, Memory and Executive Control, Springer.

Examples

```
data(ability)
cs<- psych::cs
keys <- list(ICAR16=colnames(ability), reasoning = cs(reason.4, reason.16, reason.17, reason.19),</pre>
  letters=cs(letter.7, letter.33,letter.34,letter.58),
  matrix=cs(matrix.45,matrix.46,matrix.47,matrix.55),
  rotate=cs(rotate.3,rotate.4,rotate.6,rotate.8))
  psych::scoreOverlap(keys,ability)
    #this next step takes a few seconds to run and demonstrates IRT approaches
     ability.irt <- psych::irt.fa(ability)</pre>
     ability.scores <- psych::scoreIrt(ability.irt,ability)</pre>
     ability.sub.scores <- psych::scoreIrt.2pl(keys,ability) #demonstrate irt scoring
#It is sometimes asked how to handle missing data when finding scores
#this next example compares 3 ways of scoring ability items from icar
#Just sum the items
#Sum the means for the items
#IRT score the items
total <- rowSums(ability, na.rm=TRUE)</pre>
 means <- rowMeans(ability, na.rm=TRUE)</pre>
irt <- psych::scoreIrt(items=ability)[1]</pre>
 df <- data.frame(total, means,irt)</pre>
 psych:: pairs.panels(df)
```

affect

Two data sets of affect and arousal scores as a function of personality and movie conditions

affect 5

Description

A recurring question in the study of affect is the proper dimensionality and the relationship to various personality dimensions. Here is a data set taken from two studies of mood and arousal using movies to induce affective states.

Usage

data(affect)

Details

These are data from two studies conducted in the Personality, Motivation and Cognition Laboratory at Northwestern University. Both studies used a similar methodology:

Collection of pretest data using 5 scales from the Eysenck Personality Inventory and items taken from the Motivational State Questionnaire (see msq. In addition, state and trait anxiety measures were given. In the "maps" study, the Beck Depression Inventory was given also.

Then subjects were randomly assigned to one of four movie conditions: 1: Frontline. A documentary about the liberation of the Bergen-Belsen concentration camp. 2: Halloween. A horror film. 3: National Geographic, a nature film about the Serengeti plain. 4: Parenthood. A comedy. Each film clip was shown for 9 minutes. Following this the MSQ was given again.

Data from the MSQ were scored for Energetic and Tense Arousal (EA and TA) as well as Positive and Negative Affect (PA and NA).

Study flat had 170 participants, study maps had 160.

These studies are described in more detail in various publications from the PMC lab. In particular, Revelle and Anderson, 1997 and Rafaeli and Revelle (2006). An analysis of these data has also appeared in Smillie et al. (2012).

For a much more complete data set involving film, caffeine, and time of day manipulations, see the msqR data set.

Source

Data collected at the Personality, Motivation, and Cognition Laboratory, Northwestern University.

References

Revelle, William and Anderson, Kristen Joan (1997) Personality, motivation and cognitive performance: Final report to the Army Research Institute on contract MDA 903-93-K-0008

Rafaeli, Eshkol and Revelle, William (2006), A premature consensus: Are happiness and sadness truly opposite affects? Motivation and Emotion, 30, 1, 1-12.

Smillie, Luke D. and Cooper, Andrew and Wilt, Joshua and Revelle, William (2012) Do Extraverts Get More Bang for the Buck? Refining the Affective-Reactivity Hypothesis of Extraversion. Journal of Personality and Social Psychology, 103 (2), 206-326.

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Examples

Athenstaedt

Gender Role Self Concept data from Athenstaedt (2003)

Description

Athenstaedt (2003) examined Gender Role Self-Concept. She reports two independent dimensions of Male and Female behaviors. While there are large gender/sex differences on both of these dimensions, the two represent independent factorsl Eagly and Revelle (2022) have used these data to explore the power of aggregation when examining sex differences. This data set is also useful to show various graphical display procedures.

Usage

```
data("Athenstaedt")
```

Format

A data frame with 576 observations on the following 117 variables.

```
STUDIE a numeric vector
```

gender Male =1, Female= 2

V1 - V74 self report items (see Athenstaedt.dictionary)

V1 Gender (Male = 1, Female = 2)

V2 To pay attention to ones appearance in the office

V3 Offer fire to somebody

V4 Paint an Apartment

V5 Mow the Lawn

V6 Make the Bed

V7 Hold the Door Open for your Partner

V8 Do the Dishes

V9 Do Extreme Sports

V10 Tinker with the Car

V11 Talk about Sports

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- V12 Assemble Prefabricated Furniture
- V13 Drive a Car in a Risky Way
- V14 Listen Attentively to Others
- V15 Tell your Partner about Problems at Work
- V16 Play on a Computer
- V17 Set the Table
- V18 Watch ones Weight
- V19 Care for a Partner if he/she is Ill
- V20 Play Chess
- V21 Meet with friends at a Regulars Table
- V22 Watch Soap Operas
- V23 Take a Friends Arm
- V24 Wrap Presents Beautifully
- V25 In case of Vacation with Partner Packing the Luggage for Both
- V26 To admit own Occupational Weekness
- V27 Work Overtime
- V28 Openly Show Vulnerability
- V29 Babysit
- V30 Change Fuses
- V31 Clean a Drain
- V32 Take Care of Somebody
- V33 Do Repair Work
- V34 Change Light Bulbs
- V35 Wash the Car
- V36 Ride a Motorcycle
- V37 Cook Meat on the Grill
- V38 Thump Carpets
- V39 Dust the Furniture
- V40 Buy Electric Appliances
- V41 Go Dancing
- V42 Go for a Walk through Town
- V43 Go to the Ballet
- V44 Hug a Friend
- V45 Do Handiwork (e.g. Knitting)
- V46 Change Bed Sheets
- V47 Sew on a Button
- V48 Do Aerobics

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- V49 Watch Sports on Television
- V50 Talk about Problems
- V51 Play Parlor Games
- V52 Talk about Politics
- V53 Take Care of Flowers
- V54 Make Coffee in the Office
- V55 Shovel Snow
- V56 Read non-Fiction Books
- V57 Organize Company Parties
- **V58** Do Home Improvement Jobs
- V59 Plead for the Socially Disadvantaged
- V60 Buy a Present for a Colleague
- V61 To Talk with Colleagues about Family Matters
- V62 Make Jam
- V63 Frquently Ask Colleagues Questions
- V64 Decorate the Office with Flowers
- V65 Pick up the Dinner Bill
- V66 Shop for the Family
- V67 Have Problem using Technical Devices
- V68 Care for Family Besides a Job
- V69 Watch Action Movies
- V70 Cook
- V71 Help your Partner Put on His or Her Coat
- V72 Wash Windows
- V73 Do the Ironing
- V74 Do the Laundry
- V75 Put on Make-up
- V76 Femininity Scale
- V77 Masculinity Scale
- V78 Femininity Scale
- V79 Masculinity Scale
- V80 Pooled Scale

MMINUS1 - MPLUS see the original Athenstaedt paper

FBEHAV a numeric vector

MBEHAV a numeric vector

Femininity a numeric vector

Masculinity a numeric vector

MF a numeric vector

bfi 9

Details

Ursala Athenstaedt (2003) reported several analyses of items and scales measuring Gender Role Self-Concept. Eagly and Revelle (2022) have used these data in an analysis of the power of aggregation. Here are the original items as well as the three scales Eagly and Revelle (2022). The accompanying Athenstaedt.dictionary may be used to see the items.

See the GERAS data set for a related example.

Source

Ursala Athenstaedt, personal communication, 2022, provided a SPSS sav file with the original data from which the complete cases in this set were selected.

References

Ursula Athenstaedt (2003) On the Content and Structure of the Gender Role Self-Concept: Including Gender-Stereotypical Behaviors in Addition to Traits. Psychology of Women Quarterly, 27, 309-318. doi: 10.1111/1471-6402.00111.

Alice Eagly and William Revelle (2022) Understanding the Magnitude of Psychological Differences Between Women and Men Requires Seeing the Forest and the Trees. Perspectives in Psychological Science doi:10.1177/17456916211046006.

Examples

```
data(Athenstaedt)
psych::scatterHist(Femininity ~ Masculinity + gender, data =Athenstaedt,
cex.point=.4,smooth=FALSE, correl=FALSE,d.arrow=TRUE,col=c("red","blue"),
    lwd=4, cex.main=1.5,main="Scatter Plot and Density",cex.axis=2)

psych::cohen.d(Athenstaedt[2:76], group="gender", dictionary=Athenstaedt.dictionary)
#show the top 5 items for each scale
select <- c(psych::selectFromKeys(Athenstaedt.keys$MF10),"gender")
psych::corPlot(Athenstaedt[,select], main="F and M items from Athenstaedt")</pre>
```

bfi

25 Personality items representing 5 factors

Description

25 personality self report items taken from the International Personality Item Pool (ipip.ori.org) were included as part of the Synthetic Aperture Personality Assessment (SAPA) web based personality assessment project. The data from 2800 subjects are included here as a demonstration set for scale construction, factor analysis, and Item Response Theory analysis. Three additional demographic variables (sex, education, and age) are also included.

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Usage

```
data(bfi)
data(bfi.dictionary)
```

Format

A data frame with 2800 observations on the following 28 variables. (The q numbers are the SAPA item numbers).

- A1 Am indifferent to the feelings of others. (q_146)
- A2 Inquire about others' well-being. (q_1162)
- A3 Know how to comfort others. (q_1206)
- A4 Love children. (q_1364)
- A5 Make people feel at ease. (q_1419)
- C1 Am exacting in my work. (q_124)
- C2 Continue until everything is perfect. (q_530)
- C3 Do things according to a plan. (q_619)
- C4 Do things in a half-way manner. (q_626)
- C5 Waste my time. (q_1949)
- E1 Don't talk a lot. (q_712)
- E2 Find it difficult to approach others. (q_901)
- E3 Know how to captivate people. (q_1205)
- E4 Make friends easily. (q_1410)
- E5 Take charge. (q_1768)
- N1 Get angry easily. (q_952)
- N2 Get irritated easily. (q_974)
- N3 Have frequent mood swings. (q_1099
- N4 Often feel blue. (q_1479)
- N5 Panic easily. (q_1505)
- 01 Am full of ideas. (q_128)
- 02 Avoid difficult reading material.(q_316)
- 03 Carry the conversation to a higher level. (q_492)
- 04 Spend time reflecting on things. (q_1738)
- 05 Will not probe deeply into a subject. (q_1964)

```
gender Males = 1, Females = 2
```

education 1 = HS, 2 = finished HS, 3 = some college, 4 = college graduate 5 = graduate degree age age in years

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Details

The first 25 items are organized by five putative factors: Agreeableness, Conscientiousness, Extraversion, Neuroticism, and Openness. The scoring key is created using make.keys, the scores are found using score.items.

These five factors are a useful example of using irt.fa to do Item Response Theory based latent factor analysis of the polychoric correlation matrix. The endorsement plots for each item, as well as the item information functions reveal that the items differ in their quality.

The item data were collected using a 6 point response scale: 1 Very Inaccurate 2 Moderately Inaccurate 3 Slightly Inaccurate 4 Slightly Accurate 5 Moderately Accurate 6 Very Accurate

as part of the Synthetic Apeture Personality Assessment (SAPA https://www.sapa-project.org/) project. To see an example of the data collection technique, visit https://www.SAPA-project.org/ or the International Cognitive Ability Resource at https://icar-project.org. The items given were sampled from the International Personality Item Pool of Lewis Goldberg using the sampling technique of SAPA. This is a sample data set taken from the much larger SAPA data bank.

Note

The bfi data set and items should not be confused with the BFI (Big Five Inventory) of Oliver John and colleagues (John, O. P., Donahue, E. M., & Kentle, R. L. (1991). The Big Five Inventory–Versions 4a and 54. Berkeley, CA: University of California, Berkeley, Institute of Personality and Social Research.)

Source

The items are from the ipip (Goldberg, 1999). The data are from the SAPA project (Revelle, Wilt and Rosenthal, 2010), collected Spring, 2010 (https://www.sapa-project.org/).

References

Goldberg, L.R. (1999) A broad-bandwidth, public domain, personality inventory measuring the lower-level facets of several five-factor models. In Mervielde, I. and Deary, I. and De Fruyt, F. and Ostendorf, F. (eds) Personality psychology in Europe. 7. Tilburg University Press. Tilburg, The Netherlands.

Revelle, W., Wilt, J., and Rosenthal, A. (2010) Individual Differences in Cognition: New Methods for examining the Personality-Cognition Link In Gruszka, A. and Matthews, G. and Szymura, B. (Eds.) Handbook of Individual Differences in Cognition: Attention, Memory and Executive Control, Springer.

Revelle, W, Condon, D.M., Wilt, J., French, J.A., Brown, A., and Elleman, L.G. (2016) Web and phone based data collection using planned missing designs. In Fielding, N.G., Lee, R.M. and Blank, G. (Eds). SAGE Handbook of Online Research Methods (2nd Ed), Sage Publications.

See Also

bi.bars to show the data by age and gender, irt.fa for item factor analysis applying the irt model.

Examples

```
data(bfi)
psych::describe(bfi)
# create the bfi.keys (actually already saved in the data file)
bfi.keys <-
    list(agree=c("-A1","A2","A3","A4","A5"),conscientious=c("C1","C2","C3","-C4","-C5"),
extraversion=c("-E1","-E2","E3","E4","E5"),neuroticism=c("N1","N2","N3","N4","N5"),
openness = c("01","-02","03","04","-05"))

scores <- psych::scoreItems(bfi.keys,bfi,min=1,max=6) #specify the minimum and maximum values
scores
#show the use of the keys.lookup with a dictionary
psych::keys.lookup(bfi.keys,bfi.dictionary[,1:4])</pre>
```

BFI.adjectives.dictionary

Dictionary for the 100 Big Five Adjectives

Description

Lew Goldberg organized 100 adjectives to measure 5 factors of personality (The Big5). 500 hundred participants were given these adjectives along with other personality measures. This dictionary allows for easy item labeling of the results. ~

Usage

```
data("BFI.adjectives.dictionary")
```

Format

A data frame with 100 observations on the following 2 variables.

numer a character vector of the item label

Item a character vector of the actual adjectives

Details

Keying information for the 100 adjectives:

Source

Data collected at the Personality, Motivation, and Cognition Laboratory, Northwestern University.

References

Lewis R. Goldberg,(1992) The development of markers for the Big-Five factor structure, Psychological Assessment, 4 (1) 26-42.

big5.100.adjectives

See Also

big5.100.adjectives for examples of the data. msqR for 3896 participants with scores on five scales of the EPI. affect for an example of the use of some of these adjectives in a mood manipulation study.

Examples

big5.100.adjectives

100 adjectives describing the "big 5" for 502 subjects

Description

Lew Goldberg organized 100 adjectives to measure 5 factors of personality (The Big5). 500 hundred participants were given these adjectives along with other personality measures in the Personality, Motivation and Cognition (PMC) lab. This data set is for demonstrations of factor and cluster analysis.

Usage

```
data("big5.100.adjectives")
```

Format

A data frame with 554 observations on the following 102 variables.

```
study a character vector
id a numeric vector
V1 numeric vector (see big5.adjectives.dictionary)
V100 A numeric vector. (see big5.adjectives.dictionary)
bfi.adjectives.keys a key list
```

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Details

Procedure. The data were collected over nine years in the Personality, Motivation and Cognition laboratory at Northwestern, as part of a series of studies examining the effects of personality and situational factors on motivational state and subsequent cognitive performance. In each of 38 studies, prior to any manipulation of motivational state, participants signed a consent form and in some studies, consumed 0 or 4mg/kg of caffeine. In caffeine studies, they waited 30 minutes and then filled out the MSQ as well as other personality trait measures (e.g. the Big 5 adjectives)

Source

Data collected at the Personality, Motivation, and Cognition Laboratory, Northwestern University.

References

Lewis R. Goldberg,(1992) The development of markers for the Big-Five factor structure, Psychological Assessment, 4 (1) 26-42.

Revelle, W. and Anderson, K.J. (1998) Personality, motivation and cognitive performance: Final report to the Army Research Institute on contract MDA 903-93-K-0008. (https://www.personality-project.org/revelle/publications/ra.ari.98.pdf).

Examples

```
data(big5.100.adjectives)
five.scores <- psych::scoreItems(big5.adjectives.keys,big5.100.adjectives)
summary(five.scores)</pre>
```

blant

A 29 x 29 matrix that produces weird factor analytic results

Description

Normally, min.res factor analysis and maximum likelihood produce very similar results. This data set (from Alexandra Blant) does not. Warnings are given for the min.res solution, the pa solution, but not the old.min nor the mle solution. Included as a test case for the factor analysis function.

Usage

```
data("blant")
```

Format

```
The format is: num [1:29, 1:29] 1 0.77 0.813 0.68 0.717 ... - attr(*, "dimnames")=List of 2 ..$ : NULL ..$ : chr [1:29] "V1" "V2" "V3" "V4" ...
```

blot 15

Details

This data matrix was sent by Alexandra Blant as an example of a problem with the minres solution in the fa function. The default solution, using fm="minres" issues a warning that the solution has improper factor score weights. This is not the case for the fm="old.min" and fm="mle" options, but is for fm="pa", fm="ols".

The residuals are indeed smaller for fm="minres" than for fm="old.min" or fm="mle".

"old.min" attempts to find the minimum residual but uses the gradient for mle. This was the approach until version 1.7.5 but was changed (see the help page for fa) following extensive communication with Hao Wu.

The problem with this matrix is probably that it is almost singular, with some smcs approaching 1 and the smallest three eigenvalues of .006, .004 and .001.

This problem matrix was provided by Alexandra Blant.

Source

Alexandra Blant, personal communication

Examples

```
data(blant)
#compare
f5 <- psych::fa(blant,5,rotate="none") #the default minres
f5.old <- psych::fa(blant,5, fm="old.min",rotate="none") #old version of minres
f5.mle <- psych::fa(blant,5,fm="mle",rotate= "none") #maximum likelihood
#compare solutions
psych::factor.congruence(list(f5,f5.old,f5.mle))
#compare sums of squared residuals
sum(residuals(f5,diag=FALSE)^2,na.rm=TRUE) # 1.355489
sum(residuals(f5.old,diag=FALSE)^2,na.rm=TRUE) # 1.539757
sum(residuals(f5.mle,diag=FALSE)^2,na.rm=TRUE) # 2.402092
#but, when we divide the squared residuals by the original (squared) correlations, we find
#a different ordering of fit
f5$fit
        # 0.9748177
f5.old$fit # 0.9752774
f5.mle$fit # 0.9603324
```

blot

Bond's Logical Operations Test – BLOT

Description

35 items for 150 subjects from Bond's Logical Operations Test. A good example of Item Response Theory analysis using the Rasch model. One parameter (Rasch) analysis and two parameter IRT analyses produce somewhat different results.

16 blot

Usage

```
data(blot)
```

Format

A data frame with 150 observations on 35 variables. The BLOT was developed as a paper and pencil test for children to measure Logical Thinking as discussed by Piaget and Inhelder.

Details

Bond and Fox apply Rasch modeling to a variety of data sets. This one, Bond's Logical Operations Test, is used as an example of Rasch modeling for dichotomous items. In their text (p 56), Bond and Fox report the results using WINSTEPS. Those results are consistent (up to a scaling parameter) with those found by the rasch function in the ltm package. The WINSTEPS seem to produce difficulty estimates with a mean item difficulty of 0, whereas rasch from ltm has a mean difficulty of -1.52. In addition, rasch seems to reverse the signs of the difficulty estimates when reporting the coefficients and is effectively reporting "easiness".

However, when using a two parameter model, one of the items (V12) behaves very differently.

This data set is useful when comparing 1PL, 2PL and 2PN IRT models.

Source

The data are taken (with kind permission from Trevor Bond) from the webpage https://www.winsteps.com/BF3/bondfox3.htm and read using read.fwf.

References

T.G. Bond. BLOT:Bond's Logical Operations Test. Townsville, Australia: James Cook University. (Original work published 1976), 1995.

T. Bond and C. Fox. (2007) Applying the Rasch model: Fundamental measurement in the human sciences. Lawrence Erlbaum, Mahwah, NJ, US, 2 edition.

See Also

See also the irt.fa and associated plot functions.

Examples

```
data(blot)

#ltm is not required by psychTools, but if available, may be run to show a Rasch model

#do the same thing with functions in psych
blot.fa <- psych::irt.fa(blot)  # a 2PN model
plot(blot.fa)</pre>
```

burt 17

burt

11 emotional variables from Burt (1915)

Description

Cyril Burt reported an early factor analysis with a circumplex structure of 11 emotional variables in 1915. 8 of these were subsequently used by Harman in his text on factor analysis. Unfortunately, it seems as if Burt made a mistake for the matrix is not positive definite. With one change from .87 to .81 the matrix is positive definite.

Usage

data(burt)

Format

A correlation matrix based upon 172 "normal school age children aged 9-12".

Sociality Sociality

Sorrow Sorrow

Tenderness Tenderness

Joy Joy

Wonder Wonder

Elation Elation

Disgust Disgust

Anger Anger

Sex Sex

Fear Fear

Subjection Subjection

Details

The Burt data set is interesting for several reasons. It seems to be an early example of the organization of emotions into an affective circumplex, a subset of it has been used for factor analysis examples (see Harman.Burt, and it is an example of how typos affect data. The original data matrix has one negative eigenvalue. With the replacement of the correlation between Sorrow and Tenderness from .87 to .81, the matrix is positive definite.

Alternatively, using cor. smooth, the matrix can be made positive definite as well, although cor.smooth makes more (but smaller) changes.

Source

(retrieved from the web at https://www.biodiversitylibrary.org/item/95822#790) Following a suggestion by Jan DeLeeuw.

18 cities

References

Burt, C.General and Specific Factors underlying the Primary Emotions. Reports of the British Association for the Advancement of Science, 85th meeting, held in Manchester, September 7-11, 1915. London, John Murray, 1916, p. 694-696 (retrieved from the web at https://www.biodiversitylibrary.org/item/95822#790)

See Also

Harman.Burt in the Harman dataset and cor.smooth

Examples

```
data(burt)
eigen(burt)$values #one is negative!
burt.new <- burt
burt.new[2,3] <- burt.new[3,2] <- .81
eigen(burt.new)$values #all are positive
bs <- psych::cor.smooth(burt)
round(burt.new - bs,3)</pre>
```

cities

Distances between 11 US cities

Description

Airline distances between 11 US cities may be used as an example for multidimensional scaling or cluster analysis.

Usage

```
data(cities)
```

Format

A data frame with 11 observations on the following 11 variables.

ATL Atlana, Georgia

BOS Boston, Massachusetts

ORD Chicago, Illinois

DCA Washington, District of Columbia

DEN Denver, Colorado

LAX Los Angeles, California

MIA Miami, Florida

JFK New York, New York

SEA Seattle, Washington

SFO San Francisco, California

MSY New Orleans, Lousianna

cubits 19

Details

An 11 x11 matrix of distances between major US airports. This is a useful demonstration of multiple dimensional scaling.

city.location is a dataframe of longitude and latitude for those cities.

Note that the 2 dimensional MDS solution does not perfectly capture the data from these city distances. Boston, New York and Washington, D.C. are located slightly too far west, and Seattle and LA are slightly too far south.

Source

```
https://www.timeanddate.com/worldclock/distance.html
```

Examples

```
data(cities)
city.location[,1] <- -city.location[,1] #included in the cities data set</pre>
plot(city.location, xlab="Dimension 1", ylab="Dimension 2",
   main ="Multidimensional scaling of US cities")
#do the mds
city.loc <- cmdscale(cities, k=2) #ask for a 2 dimensional solution round(city.loc,0)
city.loc <- -city.loc #flip the axes
city.loc <- psych::rescale(city.loc,apply(city.location,2,mean),apply(city.location,2,sd))</pre>
points(city.loc,type="n") #add the date point to the map
text(city.loc,labels=names(cities))
## Not run:
               #we need the maps package to be available
#an overlay map can be added if the package maps is available
if(require(maps)) {
  map("usa",add=TRUE)
## End(Not run)
```

cubits

Galton's example of the relationship between height and 'cubit' or forearm length

Description

Francis Galton introduced the 'co-relation' in 1888 with a paper discussing how to measure the relationship between two variables. His primary example was the relationship between height and forearm length. The data table (cubits) is taken from Galton (1888). Unfortunately, there seem to be some errors in the original data table in that the marginal totals do not match the table.

The data frame, heights, is converted from this table.

20 cubits

Usage

```
data(cubits)
```

Format

A data frame with 9 observations on the following 8 variables.

```
16.5 Cubit length < 16.5

16.75 16.5 <= Cubit length < 17.0

17.25 17.0 <= Cubit length < 17.5

17.75 17.5 <= Cubit length < 18.0

18.25 18.0 <= Cubit length < 18.5

18.75 18.5 <= Cubit length < 19.0

19.25 19.0 <= Cubit length < 19.5

19.75 19.5 <= Cubit length
```

Details

Sir Francis Galton (1888) published the first demonstration of the correlation coefficient. The regression (or reversion to mediocrity) of the height to the length of the left forearm (a cubit) was found to .8. There seem to be some errors in the table as published in that the row sums do not agree with the actual row sums. These data are used to create a matrix using table2matrix for demonstrations of analysis and displays of the data.

Source

```
Galton (1888)
```

References

Galton, Francis (1888) Co-relations and their measurement. Proceedings of the Royal Society. London Series, 45, 135-145,

See Also

```
table2matrix, table2df, ellipses, heights, peas,galton
```

Examples

cushny 21

cushny	A data set from Cushny and Peebles (1905) on the effect of three drugs
cusiniy	
	on hours of sleep, used by Student (1908)

Description

The classic data set used by Gossett (publishing as Student) for the introduction of the t-test. The design was a within subjects study with hours of sleep in a control condition compared to those in 3 drug conditions. Drug1 was 06mg of L Hscyamine, Drug 2L and Drug2R were said to be .6 mg of Left and Right isomers of Hyoscine. As discussed by Zabell (2008) these were not optical isomers. The detal1, delta2L and delta2R are changes from the baseline control.

Usage

```
data(cushny)
```

Format

A data frame with 10 observations on the following 7 variables.

```
Control Hours of sleep in a control condition
```

drug1 Hours of sleep in Drug condition 1

drug2L Hours of sleep in Drug condition 2

drug2R Hours of sleep in Drug condition 3 (an isomer of the drug in condition 2

delta1 Change from control, drug 1

delta2L Change from control, drug 2L

delta2R Change from control, drug 2R

Details

The original analysis by Student is used as an example for the t-test function, both as a paired t-test and a two group t-test. The data are also useful for a repeated measures analysis of variance.

Source

Cushny, A.R. and Peebles, A.R. (1905) The action of optical isomers: II hyoscines. The Journal of Physiology 32, 501-510.

Student (1908) The probable error of the mean. Biometrika, 6 (1), 1-25.

References

See also the data set sleep and the examples for the t.test

S. L. Zabell. On Student's 1908 Article "The Probable Error of a Mean" Journal of the American Statistical Association, Vol. 103, No. 481 (Mar., 2008), pp. 1-20

22 df2latex

Examples

df2latex

Convert a data frame, correlation matrix, or factor analysis output to a LaTeX table

Description

A set of handy helper functions to convert data frames or matrices to LaTeX tables. Although Sweave is the preferred means of converting R output to LaTeX, it is sometimes useful to go directly from a data.frame or matrix to a LaTeX table. cor2latex will find the correlations and then create a lower (or upper) triangular matrix for latex output. fa2latex will create the latex commands for showing the loadings and factor intercorrelations. As the default option, tables are prepared in an approximation of APA format.

Usage

```
df2latex(x,digits=2,rowlabels=TRUE,apa=TRUE,short.names=TRUE,font.size="scriptsize",
      big.mark=NULL,drop.na=TRUE, heading="A table from the psych package in R",
   caption="df2latex",label="default", char=FALSE,
  stars=FALSE, silent=FALSE, file=NULL, append=FALSE, cut=0, big=0, abbrev=NULL, long=FALSE)
cor2latex(x,use = "pairwise", method="pearson", adjust="holm",stars=FALSE,
       digits=2,rowlabels=TRUE,lower=TRUE,apa=TRUE,short.names=TRUE,
   font.size ="scriptsize", heading="A correlation table from the psych package in R.",
    caption="cor2latex",label="default",silent=FALSE,file=NULL,append=FALSE,cut=0,big=0)
fa2latex(f,digits=2,rowlabels=TRUE,apa=TRUE,short.names=FALSE,cumvar=FALSE,
       cut=0,big=.3,alpha=.05,font.size ="scriptsize",long=FALSE,
       heading="A factor analysis table from the psych package in R",
       caption="fa2latex",label="default",silent=FALSE,file=NULL,append=FALSE)
omega2latex(f,digits=2,rowlabels=TRUE,apa=TRUE,short.names=FALSE,cumvar=FALSE,cut=.2,
        big=.3, font.size ="scriptsize",
        heading="An omega analysis table from the psych package in R",
     caption="omega2latex",label="default",silent=FALSE,file=NULL,append=FALSE)
irt2latex(f,digits=2,rowlabels=TRUE,apa=TRUE,short.names=FALSE,
       font.size ="scriptsize", heading="An IRT factor analysis table from R",
       caption="fa2latex",label="default",silent=FALSE,file=NULL,append=FALSE)
ICC2latex(icc, digits=2, rowlabels=TRUE, apa=TRUE, ci=TRUE,
   font.size ="scriptsize",big.mark=NULL, drop.na=TRUE,
    heading="A table from the psych package in R",
  caption="ICC2latex",label="default",char=FALSE,silent=FALSE,file=NULL,append=FALSE)
```

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Arguments

Х	A data frame or matrix to convert to LaTeX. If non-square, then correlations will be found prior to printing in cor2latex
digits	Round the output to digits of accuracy. NULL for formatting character data
abbrev	How many characters should be used in column names –defaults to digits + 3
rowlabels	If TRUE, use the row names from the matrix or data.frame
short.names	Name the columns with abbreviated rownames to save space
ара	If TRUE formats table in APA style
cumvar	For factor analyses, should we show the cumulative variance accounted for?
font.size	e.g., "scriptsize", "tiny" or anyother acceptable LaTeX font size.
heading	The label appearing at the top of the table
caption	The table caption
lower	in cor2latex, just show the lower triangular matrix
f	The object returned from a factor analysis using fa or irt.fa.
label	The label for the table
big.mark	Comma separate numbers large numbers (big.mark=",")
drop.na	Do not print NA values
method	When finding correlations, which method should be used (pearson)
use	use="pairwise" is the default when finding correlations in cor2latex
adjust	If showing probabilities, which adjustment should be used (holm)
stars	Should probability 'magic astericks' be displayed in cor2latex (FALSE)
char	char=TRUE allows printing tables with character information, but does not allow for putting in commas into numbers
cut	In omega2latex, df2latex and fa2latex, do not print abs(values) < cut
big	In fa2latex and df2latex boldface those abs(values) > big
alpha	If fa has returned confidence intervals, then what values of loadings should be boldfaced?
icc	Either the output of an ICC, or the data to be analyzed.
ci	Should confidence intervals of the ICC be displayed
silent	If TRUE, do not print any output, just return silently - useful if using Sweave
file	If specified, write the output to this file
append	If file is specified, then should we append (append=TRUE) or just write to the file
long	if TRUE, then do long tables. (requires the longtables package in latex)

Value

A LaTeX table. Note that if showing "stars" for correlations, then one needs to use the siunitx package in LaTex. The entire LaTeX output is also returned invisibly. If using Sweave to create tables, then the silent option should be set to TRUE and the returned object saved as a file. See the last example.

Finally, some users have asked for the ability to convert these output tables into HTML. This may be done using the tth package.

24 dfOrder

Author(s)

William Revelle with suggestions from Jason French and David Condon and Davide Morselli

See Also

The many LaTeX conversion routines in Hmisc.

To convert these LaTex objects to HTML, you should install the tth package.

Consider the last example

Examples

```
df2latex(psych::Thurstone,rowlabels=FALSE,apa=FALSE,short.names=FALSE,
        caption="Thurstone Correlation matrix")
df2latex(psych::Thurstone,heading="Thurstone Correlation matrix in APA style")
df2latex(psych::describe(psych::sat.act)[2:10],short.names=FALSE)
cor2latex(psych::Thurstone)
cor2latex(psych::sat.act,short.names=FALSE)
fa2latex(psych::fa(psych::Thurstone,3),heading="Factor analysis from R in quasi APA style")
#To convert these latex tables to HTML
#f3.lat <- fa2latex(psych::fa(psych::Thurstone,3),</pre>
    heading="Factor analysis from R in quasi APA style")
#library(tth)
#f3.ht <- tth(f3.lat)
#print(as.data.frame(f3.ht),row.names=FALSE)
###
#If using Sweave to create a LateX table as a separate file then set silent=TRUE
#e.g.,
#LaTex preamble
#....
#<<pre>print=FALSE,echo=FALSE>>=
#f3 <- fa(Thurstone,3)
#fa2latex(f3,silent=TRUE,file='testoutput.tex')
#@
#\input{testoutput.tex}
```

dfOrder 25

Description

Although order will order a vector, and it is possible to order several columns of a data.frame by specifying each column individually in the call to order, dfOrder will order a dataframe or matrix by as many columns as desired. The default is to sort by columns in lexicographic order. If the object is a correlation matrix, then the selected columns are sorted by the (abs) max value across the columns (similar to fa.lookup in psych). If object is a correlation matrix, rows and columns are sorted.

Usage

```
dfOrder(object, columns,absolute=FALSE,ascending=TRUE)
```

Arguments

object The data.frame or matrix to be sorted

columns Column numbers or names to use for sorting. If positive, then they will be sorted

in increasing order. If negative, then in decreasing order

absolute If TRUE, then sort the absolute values ascending By default, order from smallest to largest.

Details

This is just a simple helper function to reorder data.frames and correlation matrices. Originally developed to organize IRT output from the ltm package. It is a basic add on to the order function.

(Completely rewritten for version 1.8.1. and then again for 2.2.1 to allow sorting correlation matrices by numeric values.)

Value

The original data frame is now in sorted order. If the input is a correlation matrix, the output is sorted by rows and columns.

Author(s)

William Revelle

See Also

Other useful file manipulation functions include read.file to read in data from a file or read.clipboard from the clipboard, fileScan, filesList, filesInfo, and fileCreate

dfOrder code is used in the test.irt function to combine ltm and sim.irt output.

Examples

```
#create a data frame and then sort it in lexicographic order set.seed(42) x <- matrix(sample(1:4,64,replace=TRUE),ncol=4) dfOrder(x) # sort by all columns
```

26 eminence

```
dfOrder(x,c(1,4)) #sort by the first and 4th column
x.df <- data.frame(x)
dfOrder(x.df,c(1,-2)) #sort by the first in increasing order,
    #the second in decreasing order

#now show sorting correlation matrices
r <- cor(sat.act,use="pairwise")
r.ord <- dfOrder(r,columns=c("education","ACT"),ascending=FALSE)
psych::corPlot(r.ord)</pre>
```

eminence

Eminence of 69 American Psychologists

Description

Marco Del Giudice criticized an earlier study by Simonton for using partial regression weights to estimate the importance of various predictors of rated eminence. This is a nice example of the (mis)interpretation of beta weights of highly correlated predictors.

Usage

```
data("eminence")
```

Format

A data frame with 69 observations on the following 9 variables.

```
name a character vector
reputation Log of rated reputation
birth.year Year of birth
first.year Year of first cited publicatin
last.year Year of last cited publication
works Log of number of publications
citations Log of number of citations
composite A composite index of publications
h The 'h' index of citations
```

Details

Simonton (1997, 2014) discusses various estimates of eminence among 69 psychologists born between 1842 and 1912 and reports that the regression weights are small and interprets this as meaning number of publications and citations are not very important. Del Giudice (2020) points out that citations and the number of publications are highly collinear and thus while their independent contributions are small, their joint effect is quite large (R=.69). These data are given here as an example of multiple correlation and partial correlation

epi 27

Source

Del Giudice (2020) links to a web page with the data.

References

Marco Del Giudice (2020). How Well Do Bibliometric Indicators Correlate With Scientific Eminence? A Comment on Simonton (2016). Perspective in Psychological Science, 15, 202-203.

Simonton, D. K. (1992). Leaders of American psychology, 1879-1967: Career development, creative output, and professional achievement. Journal of Personality and Social Psychology, 62, 5-17.

Simonton, D. K. (2016). Giving credit where credit is due: Why it's so hard to do in psychological science. Perspectives on Psychological Science, 11, 888-892.

Examples

```
data(eminence)
psych::lowerCor(eminence)
cs <- psych::cs
psych::partial.r(eminence, x= cs(reputation, works, citations),y=cs(birth.year))
psych::setCor(reputation ~ works + h + first.year,data=eminence)</pre>
```

epi

Eysenck Personality Inventory (EPI) data for 3570 participants

Description

The EPI is and has been a very frequently administered personality test with 57 measuring two broad dimensions, Extraversion-Introversion and Stability-Neuroticism, with an additional Lie scale. Developed by Eysenck and Eysenck, 1964. Eventually replaced with the EPQ which measures three broad dimensions. This data set represents 3570 observations collected in the early 1990s at the Personality, Motivation and Cognition lab at Northwestern. An additional data set (epiR) has test and retest information for 474 participants. The data are included here as demonstration of scale construction and test-retest reliability.

Usage

```
data(epi)
data(epi.dictionary)
data(epiR)
```

Format

A data frame with 3570 observations on the following 57 variables.

id The identification number within the study

time First (group testing) or 2nd time (before a lab experiment) for the epiR data set.

study Four lab based studies and their pretest data

28 epi

- V1 a numeric vector
- V2 a numeric vector
- V3 a numeric vector
- V4 a numeric vector
- V5 a numeric vector
- V6 a numeric vector
- V7 a numeric vector
- V8 a numeric vector
- V9 a numeric vector
- V10 a numeric vector
- V11 a numeric vector
- V12 a numeric vector
- V13 a numeric vector
- V14 a numeric vector
- V15 a numeric vector
- V16 a numeric vector
- V17 a numeric vector
- V18 a numeric vector
- V19 a numeric vector
- V20 a numeric vector
- V21 a numeric vector
- V22 a numeric vector
- V23 a numeric vector
- V24 a numeric vector
- V25 a numeric vector
- V26 a numeric vector
- V27 a numeric vector
- V28 a numeric vector
- V29 a numeric vector
- V30 a numeric vector
- V31 a numeric vector
- V32 a numeric vector
- V33 a numeric vector
- V34 a numeric vector
- V35 a numeric vector
- V36 a numeric vector
- V37 a numeric vector

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V38 a numeric vector V39 a numeric vector V40 a numeric vector V41 a numeric vector V42 a numeric vector V43 a numeric vector V44 a numeric vector V45 a numeric vector V46 a numeric vector V47 a numeric vector V48 a numeric vector V49 a numeric vector V50 a numeric vector V51 a numeric vector V52 a numeric vector V53 a numeric vector V54 a numeric vector V55 a numeric vector

V56 a numeric vector V57 a numeric vector

Details

The original data were collected in a group testing framework for screening participants for subsequent studies. The participants were enrolled in an introductory psychology class between Fall, 1991 and Spring, 1995.

The actual items may be found in the epi.dictionary.

The structure of the E scale has been shown by Rocklin and Revelle (1981) to have two subcomponents, Impulsivity and Sociability. These were subsequently used by Revelle, Humphreys, Simon and Gilliland (1980) to examine the relationship between personality, caffeine induced arousal, and cognitive performance.

The epiR data include the original group testing data and matched data for 474 participants collected several weeks later. This is useful for showing that internal consistency estimates (e.g. alpha or omega) can be low even though the test is stable across time. For more demonstrations of the distinction between immediate internal consistency and delayed test-retest reliability see the msqR and sai data sets and testRetest.

Source

Data from the PMC laboratory at Northwestern.

30 epi.bfi

References

Eysenck, H.J. and Eysenck, S. B.G. (1968). Manual for the Eysenck Personality Inventory. Educational and Industrial Testing Service, San Diego, CA.

Revelle, W. and Humphreys, M. S. and Simon, L. and Gilliland, K. (1980) Interactive effect of personality, time of day, and caffeine: A test of the arousal model, Journal of Experimental Psychology General, 109, 1, 1-31,

Examples

```
data(epi)
epi.keys <- list(E = c("V1", "V3", "V8", "V10", "V13", "V17", "V22", "V25", "V27", "V39",
    "V44", "V46", "V49", "V53", "V56", "-V5", "-V15", "-V20", "-V29", "-V32", "-V34","-V37",
    ""-V41", "-V51"),
N = c( "V2", "V4", "V7", "V9", "V11", "V14", "V16", "V19", "V21", "V23", "V26", "V28",
    "V31", "V33", "V35", "V38", "V40","V43", "V45", "V47", "V50", "V52","V55", "V57"),
L = c("V6", "V24", "V36", "-V12", "-V18", "-V30", "-V42", "-V48", "-V54"),
Imp = c( "V1", "V3", "V8", "V10", "V13", "V22", "V39", "-V5", "-V41"),
Soc = c( "V17", "V25", "V27", "V44", "V46", "V53", "-V11", "-V15", "-V20",
    "-V29", "-V32", "-V37", "-V51")
)
scores <- psych::scoreItems(epi.keys,epi)

psych::keys.lookup(epi.keys[1:3],epi.dictionary) #show the items and keying information
#a variety of demonstrations (not run) of test retest reliability versus alpha versus omega

E <- psych::selectFromKeys(epi.keys$E)
#look at the testRetest help file for more examples</pre>
```

epi.bfi

13 personality scales from the Eysenck Personality Inventory and Big 5 inventory

Description

A small data set of 5 scales from the Eysenck Personality Inventory, 5 from a Big 5 inventory, a Beck Depression Inventory, and State and Trait Anxiety measures. Used for demonstrations of correlations, regressions, graphic displays.

Usage

```
data(epi.bfi)
```

epi.bfi 31

Format

```
epiE EPI Extraversion

epiS EPI Sociability (a subset of Extraversion items

epiImp EPI Impulsivity (a subset of Extraversion items

epilie EPI Lie scale

epiNeur EPI neuroticism

bfagree Big 5 inventory (from the IPIP) measure of Agreeableness

bfcon Big 5 Conscientiousness

bfext Big 5 Extraversion

bfneur Big 5 Neuroticism

bfopen Big 5 Openness

bdi Beck Depression scale

traitanx Trait Anxiety
```

A data frame with 231 observations on the following 13 variables.

Details

stateanx State Anxiety

Self report personality scales tend to measure the "Giant 2" of Extraversion and Neuroticism or the "Big 5" of Extraversion, Neuroticism, Agreeableness, Conscientiousness, and Openness. Here is a small data set from Northwestern University undergraduates with scores on the Eysenck Personality Inventory (EPI) and a Big 5 inventory taken from the International Personality Item Pool.

Source

Data were collected at the Personality, Motivation, and Cognition Lab (PMCLab) at Northwestern by William Revelle)

References

https://personality-project.org/pmc.html

Examples

```
data(epi.bfi)
psych::pairs.panels(epi.bfi[,1:5])
psych::describe(epi.bfi)
```

32 galton

galton

Galton's Mid parent child height data

Description

Two of the earliest examples of the correlation coefficient were Francis Galton's data sets on the relationship between mid parent and child height and the similarity of parent generation peas with child peas. This is the data set for the Galton height.

Usage

```
data(galton)
```

Format

A data frame with 928 observations on the following 2 variables.

```
parent Mid Parent heights (in inches)
child Child Height
```

Details

Female heights were adjusted by 1.08 to compensate for sex differences. (This was done in the original data set)

Source

This is just the galton data set from UsingR, slightly rearranged.

References

Stigler, S. M. (1999). Statistics on the Table: The History of Statistical Concepts and Methods. Harvard University Press. Galton, F. (1886). Regression towards mediocrity in hereditary stature. Journal of the Anthropological Institute of Great Britain and Ireland, 15:246-263. Galton, F. (1869). Hereditary Genius: An Inquiry into its Laws and Consequences. London: Macmillan.

Wachsmuth, A.W., Wilkinson L., Dallal G.E. (2003). Galton's bend: A previously undiscovered nonlinearity in Galton's family stature regression data. The American Statistician, 57, 190-192.

See Also

The other Galton data sets: heights, peas, cubits

GERAS 33

Examples

GERAS

Data from Gruber et al, 2020, Study 2: Gender Related Attributes Survey

Description

Gruber et al. (2020) report on the psychometric properties of a multifaceted Gender Related Attributes Survey. Here are the data from their 3 domains (Personality, Cognition and Activities and Interests from their study 2. Eagly and Revelle (2022) include these data in their review of the power of aggregation. The data are included here as demonstrations of the cohen.d and scatterHist functions in the psych package and may be used to show the power of aggregation.

Usage

```
data("GERAS")
#These other objects are included in the file
# data("GERAS.scales")
# data("GERAS.dictionary")
# data("GERAS.items")
# data("GERAS.keys")
```

Format

A data frame with 471 observations on the following 51 variables (selected from the original 93) The code numbers are item numbers from the bigger set.

```
V15 reckless
V22 willing to take risks
V11 courageous
V6 a adventurous
V19 dominant
V14 controlling
V20 boastful
```

34 GERAS

- V21 rational
- V23 analytical
- V9 pragmatic
- V44 to find an address for the first time
- V45 to find a way again
- V46 to understand equations
- V50 to follow directions
- V51 to understand equations
- V53 day-to-day calculations
- V48 to write a computer program
- V69 paintball
- V73 driving go-cart
- V71 drinking beer
- V68 watching action movies
- V75 playing cards (poker)
- V72 watching sports on TV
- V67 doing certain sports (e.g. soccer, ...)
- V74 Gym (weightlifting)
- V27 warm-hearted
- V28 loving
- V29 caring
- V26 compassionate
- V32 delicate
- V30 tender
- V24 familiy-oriented
- V40 anxious
- V39 thin-skinned
- V41 careful
- V55 to explain foreign words
- V58 to find the right words to express certain content
- V59 synonyms for a word in order to avoid repetitions
- V60 to phrase a text
- V54 remembering events from your own life
- V63 to notice small changes
- V57 to remember names and faces
- V89 shopping
- V92 gossiping

GERAS 35

```
V81 watching a romantic movie
V80 talking on the phone with a friend
V90 yoga
V83 rhythmic gymnastics
V84 going for a walk
V86 dancing
gender gender (M=1 F=2)
```

Details

These 50 items (+ gender) may be formed into scales using the GERAS.keys The first 10 items are Male Personality, the next 10 are Female Personality, then 7 and 7 M and F Cognition, then 8 and 8 M and F Activity items. The Pers, Cog and Act scales are formed from the M-F scales for the three domains. M and F are the composites of the Male and then the Female scales. MF.all is the composite of the M - F scales. See the GERAS.keys object for scoring directions.

```
"M.pers" "F.pers" "M.cog" "F.cog" "M.act" "F.act" "Pers" "Cog" "Act" "M" "F" "MF.all" "gender" See the Athenstaedt data set for a related data set.
```

Source

Study 2 data downloaded from the Open Science Framework https://osf.io/42jhr/ Used by kind permission of Freya M. Gruber, Tullia Ortner, and Belinda A. Pletzer.

References

Alice H. Eagly and William Revelle (2022), Understanding the Magnitude of Psychological Differences Between Women and Men Requires Seeing the Forest and the Tree. Perspectives in Psychological Science doi:10.1177/17456916211046006

Gruber, Freya M. and Distlberger, Eva and Scherndl, Thomas and Ortner, Tuulia M. and Pletzer, Belinda (2020) Psychometric properties of the multifaceted Gender-Related Attributes Survey (GERAS) European Journal of Psychological Assessment, 36, (4) 612-623.

Examples

```
data(GERAS)
GERAS.keys #show the keys
#show the items from the dictionary
psych::lookupFromKeys(GERAS.keys, GERAS.dictionary[,4,drop=FALSE])

#now, use the GERAS.scales to show a scatterHist plot showing univariate d and bivariate
# Mahalanobis D.

psych::scatterHist(F ~ M + gender, data=GERAS.scales, cex.point=.3,smooth=FALSE,
xlab="Masculine Scale",ylab="Feminine Scale",correl=FALSE,
d.arrow=TRUE,col=c("red","blue"), bg=c("red","blue"), lwd=4, title="Combined M and F scales",cex.cor=2,cex.arrow=1.25, cex.main=2)
```

36 globalWarm

globalWarm

7 attitude items about Global Warming policy from Erik Nisbet

Description

Erik Nisbet reported the relationship between emotions, ideology, and party affiliation as predictors of attitudes towards government action on climate change. The data were used by Hayes (2013) in a discussion of regression. They are available as the glbwarm data set in the processR package. They are copied here for examples of mediation.

Usage

```
data("globalWarm")
```

Format

A data frame with 815 observations on the following 7 variables.

```
govact a numeric
posemot a numeric
negemot a numeric
ideology a numeric
age a numeric
sex a numeric
partyid a numeric
```

Details

More details to follow

Source

The raw data are available from the processR package (Keon-Woong Moon, 2020). as the glbwarm data set. The data set is used by Hayes in several examples. Used here by kind permission of Erik Nisbet.

References

```
Nisbet, E. (?)
```

Hayes, Andrew F. (2013) Introduction to mediation, moderation, and conditional process analysis: A regression-based approach. Guilford Press.

heights 37

Examples

```
data(globalWarm)
psych::lowerCor(globalWarm)
```

heights

A data frame of the Galton (1888) height and cubit data set.

Description

Francis Galton introduced the 'co-relation' in 1888 with a paper discussing how to measure the relationship between two variables. His primary example was the relationship between height and forearm length. The data table (cubits) is taken from Galton (1888). Unfortunately, there seem to be some errors in the original data table in that the marginal totals do not match the table.

The data frame, heights, is converted from this table using table2df.

Usage

data(heights)

Format

A data frame with 348 observations on the following 2 variables.

```
height Height in inches
cubit Forearm length in inches
```

Details

Sir Francis Galton (1888) published the first demonstration of the correlation coefficient. The regression (or reversion to mediocrity) of the height to the length of the left forearm (a cubit) was found to .8. The original table cubits is taken from Galton (1888). There seem to be some errors in the table as published in that the row sums do not agree with the actual row sums. These data are used to create a matrix using table2matrix for demonstrations of analysis and displays of the data.

Source

Galton (1888)

References

Galton, Francis (1888) Co-relations and their measurement. Proceedings of the Royal Society. London Series, 45, 135-145,

See Also

```
table2matrix, table2df, cubits, ellipses, galton
```

Examples

```
data(heights)
psych::ellipses(heights,n=1,main="Galton's co-relation data set")
```

holzinger.swineford

The raw and transformed data from Holzinger and Swineford, 1939

Description

A classic data set in psychometrics is that from Holzinger and Swineford (1939). A 4 and 5 factor solution to 24 of these variables problem is presented by Harman (1976), and 9 of these are used by the lavaan package. The two data sets were supplied by Keith Widaman.

Usage

Format

A data frame with 301 observations on the following 33 variables. Longer descriptions taken from Thompson, (1998).

```
case a numeric vector
school School Pasteur or Grant-White
grade Grade (7 or 8)
female male = 1, female = 2
ageyr age in years
mo months over year
agemo Age in months
t01_visperc Visual perception test from Spearman VPT Part I
t02_cubes Cubes, Simplification of Brighams Spatial Relations Test
t03_frmbord Paper formboard-Shapes that can be combined to form a target
t04_lozenges Lozenges from Thorndike-Shapes flipped over then identify target
t05_geninfo General Information Verbal Test
t06_paracomp Paragraph Comprehension Test
t07_sentcomp Sentence Completion Test
t08_wordclas Word clasification-Which word not belong in set
t09_wordmean Word Meaning Test
```

```
t10_addition Speeded addition test
t11_code Speeded codetest-Transform shapes into alpha with code
t12_countdot Speeded counting of dots in shap
t13_sccaps Speeded discrimation of straight and curved caps
t14_wordrecg Memory of Target Words
t15_numbrecg Memory of Target Numbers
t16_figrrecg Memory of Target Shapes
t17_objnumb Memory of object-Number association targets
t18_numbfig Memory of number-Object association targets
t19_figword Memory of figure-Word association target
t20_deduction Deductive Math Ability
t21_numbpuzz Math number puzzles
t22_probreas Math word problem reasoning
t23_series Completion of a Math Number Series
t24_woody Woody-McCall mixed math fundamentals test
t25_frmbord2 Revision of t3-Paper form board
t26_flags Flags-possible substitute for t4 lozenges
```

Details

The following commentary was provided by Keith Widaman:

"The Holzinger and Swineford (1939) data have been used as a model data set by many investigators. For example, Harman (1976) used the "24 Psychological Variables" example prominently in his authoritative text on multiple factor analysis, and the data presented under this rubric consisted of 24 of the variables from the Grant-White school (N = 145). Meredith (1964a, 1964b) used several variables from the Holzinger and Swineford study in his work on factorial invariance under selection. Joreskog (1971) based his work on multiple-group confirmatory factor analysis using the Holzinger and Swineford data, subsetting the data into four groups.

Rosseel, who developed the 'lavaan' package for R, included 9 of the manifest variables from Holzinger and Swineford (1939) as a "resident" data set when one downloads the 'lavaan' package. Several background variables are included in this "resident" data set in addition to 9 of the psychological tests (which are named x1 - x9 in the data set). When analyzing these data, I found the distributions of the variables (means, SDs) did not match the sample statistics from the original article. For example, in the "resident" data set in 'lavaan', scores on all manifest variables ranged between 0 and 10, sample means varied between 3 and 6, and sample SDs varied between 1.0 and 1.5. In the original data set, scores ranges were rather different across tests, with some variables having scores that ranged between 0 and 20, but other manifest variables having scores ranging from 50 to over 300 - with obvious attendant differences in sample means and SDs.

After a bit of snooping (i.e., data analysis), I discovered that the 9 variables in the "resident" data set in 'lavaan' had been rescored through ratio transformations. The ratio transformations involved dividing the raw score for each person on a given test by a particular constant for that test that transformed scores on the test to have the desired range.

I decided to perform transformations of all 26 variables so that two data sets could be available to interested researchers:"

holzinger.raw are the raws scores on all variables from Holzinger & Swineford (1939)

holzinger.swineford are rescaled scores on all variables from Holzinger & Swineford.

holzinger.dictionary is a list of the variable names in short and long form.

... Widaman continues:

"As several persons have noted, Harman (1976) used data only from the Grant-White school (N = 145) for his 24 Psychological Variables data set. In doing so, Harman replaced t03_frmbord and t04_lozenges with t25_frmbord2 and t26_flags, because the latter two tests were experimental tests that were designed to be more appropriate for this age level. This substitution is fine, as long as one analyzes data from only the Grant- White school. If one wishes to perform multiple-group analyses and uses school as a grouping variable (as Meredith, 1964a, 1964b, and Joreskog, 1971, did), then tests 25 and 26 should not be used."

"As have others, Gorsuch (1983) mentioned that analyses based on the raw data reported by Holzinger and Swineford (1939) will not produce statistics (means, SDs, correlations) that match precisely the values reported by Holzinger and Swineford or Harman (1976). Following Gorsuch, I have assumed that the raw data are correct. Applying factor analytic techniques to the raw data from the Grant-White school and to the summary data reported by Harman (1976) will produce slightly different results, but results that differ in only minor, unimportant details."

These data are interesting not just for the historical completeness of having the orinal data, but also as an example of suppressor variables. Age and grade are positively correlated, and scores are higher in the 8th grade than in the 7th grade. But age (particularly in months) is negatively correlated with many of the cognitive tasks, and when grade and age are both entered into regression, this negative correlation is enhanced. That is, although increasing grade increases cognitive perforamnce, younger children in both grades do better than the older children.

Note

As discussed by Widaman, the descriptive values reported in Harman (1967) (p 124) do not quite match the descriptive statistics in holzinger.raw. Further note that the correlation matrix and factor loadings are trivially different from the Harman.24 factor loadings in the GPA rotation package.

The purpose behind presenting both the raw and transformed data is to show that the fit statistics from factor analysis are identical for these two data sets.

The variables v1 ... v9 in the lavaan package correspond to tests 1, 2, 4, 6, 7, 9, 10, 12 and 13.

Source

Keith Widaman (2019, personal communication). Original data from Holzinger and Swineford (1939).

References

Gorsuch, R. L. (1983). Factor analysis (2nd ed.). Hillsdale, NJ: Erlbaum.

Harman, Harry Horace (1967), Modern factor analysis. Chicago, University of Chicago Press.

Holzinger, K. J., & Swineford, F. (1939). A study in factor analysis: The stability of a bi-factor solution. Supplementary Educational Monographs, no. 48. Chicago: University of Chicago, Department of Education.

Joreskog, K. G. (1971). Simultaneous factor analysis in several populations. Psychometrika, 36, 409-426.

Meredith, W. (1964a). Notes on factorial invariance. Psychometrika, 29, 177-185.

Meredith, W. (1964b). Rotation to achieve factorial invariance. Psychometrika, 29, 177-206.

Meredith, W. (1977). On weighted Procrustes and hyperplane fitting in factor analytic rotation. Psychometrika, 42, 491-522.

Thompson, Bruce. Five Methodology Errors in Educational Research: The Pantheon of Statistical Significance and Other Faux Pas. Paper presented at the Annual Meeting of the American Educational Research Association (San Diego, CA, April 13-17,1998)

See Also

psych::Holzinger

Examples

```
data(holzinger.raw)
psych::describe(holzinger.raw)
data(holzinger.dictionary)
holzinger.dictionary #to see the longer names for these data (taken from Thompson)
#Compare these to the lavaan correlation matrix
psych::lowerCor(holzinger.swineford[ 7+ c(1, 2, 4, 6, 7, 9, 10, 12, 13)])
psych::setCor(t01_visperc + t05_geninfo + t08_wordclas ~ grade + agemo,data = holzinger.raw)
psych::setCor( t06_paracomp ~ grade + agemo, data=holzinger.swineford)
psych::mediate(t06_paracomp ~ grade + (agemo),data = holzinger.raw,std=TRUE)
#show the omega structure of the 24 variables
om4 <- psych::omega(holzinger.swineford[8:31],4)</pre>
psych::omega.diagram(om4,sl=FALSE,main="26 variables from Holzinger-Swineford")
#these data also show an interesting suppression effect
psych::lowerCor(holzinger.swineford[c(3,7,12:14)])
psych::setCor( t06_paracomp ~ grade + agemo, data=holzinger.swineford)
#or show as a mediation effect
mod <- psych::mediate(t06_paracomp ~ grade + (agemo),data = holzinger.raw,std=TRUE,n.iter=50)</pre>
summary(mod)
#now, show a plot of these effets
plot(t07_sentcomp ~ agemo, col=c("red","blue")[holzinger.swineford$grade -6],
 pch=26-holzinger.swineford$grade,data=holzinger.swineford,
  ylab="Sentence Comprehension", xlab="Age in Months",
  main="Sentence Comprehension varies by age and grade")
by(holzinger.swineford,holzinger.swineford$grade -6,function(x) abline(
     lm(t07_sentcomp ~ agemo,data=x),lty=c("dashed","solid")[x$grade-6]))
```

42 income

```
text(190,3.3,"grade = 8")
text(190,2,"grade = 7")
```

income

US family income from US census 2008

Description

US census data on family income from 2008

Usage

```
data(income)
```

Format

A data frame with 44 observations on the following 4 variables.

value lower boundary of the income group count Number of families within that income group mean Mean of the category

prop proportion of families

Details

The distribution of income is a nice example of a log normal distribution. It is also an interesting example of the power of graphics. It is quite clear when graphing the data that income statistics are bunched to the nearest 5K. That is, there is a clear sawtooth pattern in the data.

The all.income set is interpolates intervening values for 100-150K, 150-200K and 200-250K

Source

US Census: Table HINC-06. Income Distribution to \$250,000 or More for Households: 2008 https://www.census.gov/hhes/www/cpstables/032009/hhinc/new06_000.htm

Examples

iqitems 43

iqitems

16 multiple choice IQ items

Description

16 multiple choice ability items taken from the Synthetic Aperture Personality Assessment (SAPA) web based personality assessment project. The data from 1525 subjects are included here as a demonstration set for scoring multiple choice inventories and doing basic item statistics. For more information on the development of an open source measure of cognitive ability, consult the readings available at the https://personality-project.org/.

Usage

```
data(iqitems)
```

Format

A data frame with 1525 observations on the following 16 variables. The number following the name is the item number from SAPA.

```
reason.4 Basic reasoning questions
```

reason. 16 Basic reasoning question

reason.17 Basic reasoning question

reason.19 Basic reasoning question

letter.7 In the following alphanumeric series, what letter comes next?

letter.33 In the following alphanumeric series, what letter comes next?

letter.34 In the following alphanumeric series, what letter comes next

letter. 58 In the following alphanumeric series, what letter comes next?

matrix.45 A matrix reasoning task

matrix.46 A matrix reasoning task

matrix.47 A matrix reasoning task

matrix.55 A matrix reasoning task

rotate.3 Spatial Rotation of type 1.2

rotate.4 Spatial Rotation of type 1.2

rotate.6 Spatial Rotation of type 1.1

rotate.8 Spatial Rotation of type 2.3

44 iqitems

Details

16 items were sampled from 80 items given as part of the SAPA (https://www.sapa-project.org/) project (Revelle, Wilt and Rosenthal, 2009; Condon and Revelle, 2014) to develop online measures of ability. These 16 items reflect four lower order factors (verbal reasoning, letter series, matrix reasoning, and spatial rotations. These lower level factors all share a higher level factor ('g'). Similar data are available from the International Cognitive Abiity Resource at https://www.icar-project.org/.

This data set and the associated data set (ability based upon scoring these multiple choice items and converting them to correct/incorrect may be used to demonstrate item response functions, tetrachoric correlations, or irt.fa as well as omega estimates of of reliability and hierarchical structure.

In addition, the data set is a good example of doing item analysis to examine the empirical response probabilities of each item alternative as a function of the underlying latent trait. When doing this, it appears that two of the matrix reasoning problems do not have monotonically increasing trace lines for the probability correct. At moderately high ability (theta = 1) there is a decrease in the probability correct from theta = 0 and theta = 2.

Source

The example data set is taken from the Synthetic Aperture Personality Assessment personality and ability test at https://www.sapa-project.org/. The data were collected with David Condon from 8/08/12 to 8/31/12.

References

Condon, David and Revelle, William, (2014) The International Cognitive Ability Resource: Development and initial validation of a public-domain measure. Intelligence, 43, 52-64.

Revelle, William, Dworak, Elizabeth M. and Condon, David (2020) Cognitive ability in everyday life: the utility of open-source measures. Current Directions in Psychological Science, 29, (4) 358-363. Open access at doi:10.1177/0963721420922178.

Dworak, Elizabeth M., Revelle, William, Doebler, Philip and Condon, David (2021) Using the International Cognitive Ability Resource as an open source tool to explore individual differences in cognitive ability. Personality and Individual Differences, 169. Open access at doi:10.1016/j.paid.2020.109906.

Revelle, W., Wilt, J., and Rosenthal, A. (2010) Individual Differences in Cognition: New Methods for examining the Personality-Cognition Link In Gruszka, A. and Matthews, G. and Szymura, B. (Eds.) Handbook of Individual Differences in Cognition: Attention, Memory and Executive Control, Springer.

Revelle, W, Condon, D.M., Wilt, J., French, J.A., Brown, A., and Elleman, L.G. (2016) Web and phone based data collection using planned missing designs. In Fielding, N.G., Lee, R.M. and Blank, G. (Eds). SAGE Handbook of Online Research Methods (2nd Ed), Sage Publications.

Examples

```
data(iqitems)
iq.keys <- c(4,4,4, 6, 6,3,4,4, 5,2,2,4, 3,2,6,7)
```

msq

75 mood items from the Motivational State Questionnaire for 3896 participants

Description

Emotions may be described either as discrete emotions or in dimensional terms. The Motivational State Questionnaire (MSQ) was developed to study emotions in laboratory and field settings. The data can be well described in terms of a two dimensional solution of energy vs tiredness and tension versus calmness. Additional items include what time of day the data were collected and a few personality questionnaire scores.

Usage

data(msq)

Format

A data frame with 3896 observations on the following 92 variables.

```
active a numeric vector
afraid a numeric vector
alert a numeric vector
angry a numeric vector
anxious a numeric vector
aroused a numeric vector
ashamed a numeric vector
astonished a numeric vector
at.ease a numeric vector
at.rest a numeric vector
```

attentive a numeric vector blue a numeric vector bored a numeric vector calm a numeric vector cheerful a numeric vector clutched.up a numeric vector confident a numeric vector content a numeric vector delighted a numeric vector depressed a numeric vector determined a numeric vector distressed a numeric vector drowsy a numeric vector dull a numeric vector elated a numeric vector energetic a numeric vector enthusiastic a numeric vector excited a numeric vector fearful a numeric vector frustrated a numeric vector full.of.pep a numeric vector gloomy a numeric vector grouchy a numeric vector guilty a numeric vector happy a numeric vector hostile a numeric vector idle a numeric vector inactive a numeric vector inspired a numeric vector intense a numeric vector interested a numeric vector irritable a numeric vector jittery a numeric vector lively a numeric vector lonely a numeric vector nervous a numeric vector placid a numeric vector

pleased a numeric vector proud a numeric vector quiescent a numeric vector quiet a numeric vector relaxed a numeric vector sad a numeric vector satisfied a numeric vector scared a numeric vector serene a numeric vector sleepy a numeric vector sluggish a numeric vector sociable a numeric vector sorry a numeric vector still a numeric vector strong a numeric vector surprised a numeric vector tense a numeric vector tired a numeric vector tranquil a numeric vector unhappy a numeric vector upset a numeric vector vigorous a numeric vector wakeful a numeric vector warmhearted a numeric vector wide.awake a numeric vector alone a numeric vector kindly a numeric vector scornful a numeric vector EA Thayer's Energetic Arousal Scale TA Thayer's Tense Arousal Scale PA Positive Affect scale NegAff Negative Affect scale Extraversion Extraversion from the Eysenck Personality Inventory Neuroticism Neuroticism from the Eysenck Personality Inventory Lie Lie from the EPI Sociability The sociability subset of the Extraversion Scale Impulsivity The impulsivity subset of the Extraversions Scale

MSQ_Time Time of day the data were collected

MSQ_Round Rounded time of day

TOD a numeric vector

T0D24 a numeric vector

ID subject ID

condition What was the experimental condition after the msq was given

scale a factor with levels msq r original or revised msq

exper Which study were the data collected: a factor with levels AGES BING BORN CART CITY COPE EMIT FAST Fern FILM FLAT Gray imps item knob MAPS mite pat-1 pat-2 PATS post RAFT Rim.1 Rim.2 rob-1 rob-2 ROG1 ROG2 SALT sam-1 sam-2 SAVE/PATS sett swam swam-2 TIME VALE-1 VALE-2 VIEW

Details

The Motivational States Questionnaire (MSQ) is composed of 72 items, which represent the full affective space (Revelle & Anderson, 1998). The MSQ consists of 20 items taken from the Activation-Deactivation Adjective Check List (Thayer, 1986), 18 from the Positive and Negative Affect Schedule (PANAS, Watson, Clark, & Tellegen, 1988) along with the items used by Larsen and Diener (1992). The response format was a four-point scale that corresponds to Russell and Carroll's (1999) "ambiguous—likely-unipolar format" and that asks the respondents to indicate their current standing ("at this moment") with the following rating scale:

Not at all A little Moderately Very much

The original version of the MSQ included 70 items. Intermediate analyses (done with 1840 subjects) demonstrated a concentration of items in some sections of the two dimensional space, and a paucity of items in others. To begin correcting this, 3 items from redundantly measured sections (alone, kindly, scornful) were removed, and 5 new ones (anxious, cheerful, idle, inactive, and tranquil) were added. Thus, the correlation matrix is missing the correlations between items anxious, cheerful, idle, inactive, and tranquil with alone, kindly, and scornful.

Procedure. The data were collected over nine years, as part of a series of studies examining the effects of personality and situational factors on motivational state and subsequent cognitive performance. In each of 38 studies, prior to any manipulation of motivational state, participants signed a consent form and filled out the MSQ. (The procedures of the individual studies are irrelevant to this data set and could not affect the responses to the MSQ, since this instrument was completed before any further instructions or tasks). Some MSQ post test (after manipulations) is available in affect.

The EA and TA scales are from Thayer, the PA and NA scales are from Watson et al. (1988). Scales and items:

Energetic Arousal: active, energetic, vigorous, wakeful, wide.awake, full.of.pep, lively, -sleepy, -tired, - drowsy (ADACL)

Tense Arousal: Intense, Jittery, fearful, tense, clutched up, -quiet, -still, - placid, - calm, -at rest (ADACL)

Positive Affect: active, alert, attentive, determined, enthusiastic, excited, inspired, interested, proud, strong (PANAS)

Negative Affect: afraid, ashamed, distressed, guilty, hostile, irritable, jittery, nervous, scared, upset (PANAS)

The PA and NA scales can in turn can be thought of as having subscales: (See the PANAS-X) Fear: afraid, scared, nervous, jittery (not included frightened, shaky) Hostility: angry, hostile, irritable, (not included: scornful, disgusted, loathing guilt: ashamed, guilty, (not included: blameworthy, angry at self, disgusted with self, dissatisfied with self) sadness: alone, blue, lonely, sad, (not included: downhearted) joviality: cheerful, delighted, energetic, enthusiastic, excited, happy, lively, (not included: joyful) self-assurance: proud, strong, confident, (not included: bold, daring, fearless) attentiveness: alert, attentive, determined (not included: concentrating)

The next set of circumplex scales were taken (I think) from Larsen and Diener (1992). High activation: active, aroused, surprised, intense, astonished Activated PA: elated, excited, enthusiastic, lively Unactivated NA: calm, serene, relaxed, at rest, content, at ease PA: happy, warmhearted, pleased, cheerful, delighted Low Activation: quiet, inactive, idle, still, tranquil Unactivated PA: dull, bored, sluggish, tired, drowsy NA: sad, blue, unhappy, gloomy, grouchy Activated NA: jittery, anxious, nervous, fearful, distressed.

Keys for these separate scales are shown in the examples.

In addition to the MSQ, there are 5 scales from the Eysenck Personality Inventory (Extraversion, Impulsivity, Sociability, Neuroticism, Lie). The Imp and Soc are subsets of the total extraversion scale.

Source

Data collected at the Personality, Motivation, and Cognition Laboratory, Northwestern University.

References

Larsen, R. J., & Diener, E. (1992). Promises and problems with the circumplex model of emotion. In M. S. Clark (Ed.), Review of personality and social psychology, No. 13. Emotion (pp. 25-59). Thousand Oaks, CA, US: Sage Publications, Inc.

Rafaeli, Eshkol and Revelle, William (2006), A premature consensus: Are happiness and sadness truly opposite affects? Motivation and Emotion, 30, 1, 1-12.

Revelle, W. and Anderson, K.J. (1998) Personality, motivation and cognitive performance: Final report to the Army Research Institute on contract MDA 903-93-K-0008. (https://www.personality-project.org/revelle/publications/ra.ari.98.pdf).

Thayer, R.E. (1989) The biopsychology of mood and arousal. Oxford University Press. New York, NY.

Watson, D., Clark, L.A. and Tellegen, A. (1988) Development and validation of brief measures of positive and negative affect: The PANAS scales. Journal of Personality and Social Psychology, 54(6):1063-1070.

See Also

msqR for a larger data set with repeated measures for 3032 participants measured at least once, 2753 measured twice, 446 three times and 181 four times. affect for an example of the use of some of these adjectives in a mood manipulation study.

make.keys, scoreItems and scoreOverlap for instructions on how to score multiple scales with and without item overlap. Also see fa and fa.extension for instructions on how to do factor analyses or factor extension.

Examples

```
data(msq)
   #in in the interests of time
#basic descriptive statistics
psych::describe(msq)
#score them for 20 short scales -- note that these have item overlap
#The first 2 are from Thayer
#The next 2 are classic positive and negative affect
#The next 9 are circumplex scales
#the last 7 are msg estimates of PANASX scales (missing some items)
kevs.list <- list(</pre>
EA = c("active", "energetic", "vigorous", "wakeful", "wide.awake", "full.of.pep",
       "lively", "-sleepy", "-tired", "-drowsy"),
NAf =c("jittery", "nervous", "scared", "afraid", "guilty", "ashamed", "distressed", "upset", "hostile", "irritable"),
HAct = c("active", "aroused", "surprised", "intense", "astonished"),
aPA = c("elated", "excited", "enthusiastic", "lively"),
uNA = c("calm", "serene", "relaxed", "at.rest", "content", "at.ease"),
pa = c("happy", "warmhearted", "pleased", "cheerful", "delighted" ),
LAct = c("quiet", "inactive", "idle", "still", "tranquil"),
uPA =c( "dull", "bored", "sluggish", "tired", "drowsy"),
naf = c( "sad", "blue", "unhappy", "gloomy", "grouchy"),
aNA = c("jittery", "anxious", "nervous", "fearful", "distressed"), Fear = <math>c("afraid", "scared", "nervous", "jittery"),
Hostility = c("angry" , "hostile", "irritable", "scornful" ),
Guilt = c("guilty", "ashamed"),
Sadness = c("sad", "blue", "lonely", "alone"),
Joviality =c("happy","delighted", "cheerful", "excited", "enthusiastic", "lively", "energetic"),
Self.Assurance=c( "proud", "strong" , "confident" , "-fearful" ),
Attentiveness = c("alert" , "determined" , "attentive" )
#, acquiscence = c("sleepy" , "wakeful" , "relaxed","tense")
#dropped because it has a negative alpha and throws warnings
   )
msq.scores <- psych::scoreItems(keys.list,msq)</pre>
#show a circumplex structure for the non-overlapping items
fcirc <- psych::fa(msq.scores$scores[,5:12],2)</pre>
psych::fa.plot(fcirc,labels=colnames(msq.scores$scores)[5:12])
#now, find the correlations corrected for item overlap
msq.overlap <- psych::scoreOverlap(keys.list,msq)</pre>
#a warning is thrown by smc because of some NAs in the matrix
```

msqR

75 mood items from the Motivational State Questionnaire for 3032 unique participants

Description

Emotions may be described either as discrete emotions or in dimensional terms. The Motivational State Questionnaire (MSQ) was developed to study emotions in laboratory and field settings. The data can be well described in terms of a two dimensional solution of energy vs tiredness and tension versus calmness. Alternatively, this space can be organized by the two dimensions of Positive Affect and Negative Affect. Additional items include what time of day the data were collected and a few personality questionnaire scores. 3032 unique participants took the MSQ at least once, 2753 at least twice, 446 three times, and 181 four times. The 3032 participants also took the sai state anxiety inventory at the same time. Some studies manipulated arousal by caffeine, others manipulations included affect inducing movies.

Usage

```
data("msqR")
```

Format

A data frame with 6411 observations on the following 88 variables.

```
active a numeric vector
afraid a numeric vector
alert a numeric vector
```

alone a numeric vector angry a numeric vector aroused a numeric vector ashamed a numeric vector astonished a numeric vector at.ease a numeric vector at.rest a numeric vector attentive a numeric vector blue a numeric vector bored a numeric vector calm a numeric vector clutched.up a numeric vector confident a numeric vector content a numeric vector delighted a numeric vector depressed a numeric vector determined a numeric vector distressed a numeric vector drowsy a numeric vector dull a numeric vector elated a numeric vector energetic a numeric vector enthusiastic a numeric vector excited a numeric vector fearful a numeric vector frustrated a numeric vector full.of.pep a numeric vector gloomy a numeric vector grouchy a numeric vector guilty a numeric vector happy a numeric vector hostile a numeric vector inspired a numeric vector intense a numeric vector interested a numeric vector irritable a numeric vector jittery a numeric vector

lively a numeric vector lonely a numeric vector nervous a numeric vector placid a numeric vector pleased a numeric vector proud a numeric vector quiescent a numeric vector quiet a numeric vector relaxed a numeric vector sad a numeric vector satisfied a numeric vector scared a numeric vector serene a numeric vector sleepy a numeric vector sluggish a numeric vector sociable a numeric vector sorry a numeric vector still a numeric vector strong a numeric vector surprised a numeric vector tense a numeric vector tired a numeric vector unhappy a numeric vector upset a numeric vector vigorous a numeric vector wakeful a numeric vector warmhearted a numeric vector wide.awake a numeric vector anxious a numeric vector cheerful a numeric vector idle a numeric vector inactive a numeric vector tranquil a numeric vector kindly a numeric vector scornful a numeric vector Extraversion Extraversion from the EPI Neuroticism Neuroticism from the EPI

Lie Lie from the EPI

Sociability Sociability from the EPI

Impulsivity Impulsivity from the EPI

gender 1= male, 2 = female (coded on presumed x chromosome). Slowly being added to the data set.

TOD Time of day that the study was run

drug 1 if given placebo, 2 if given caffeine

film 1-4 if given a film: 1=Frontline, 2= Halloween, 3=Serengeti, 4 = Parenthood

time Measurement occasion (1 and 2 are same session, 3 and 4 are the same, but a later session)

id a numeric vector

form msq versus msqR

study a character vector of the experiment name

Details

The Motivational States Questionnaire (MSQ) is composed of 75 items, which represent the full affective space (Revelle & Anderson, 1998). The MSQ consists of 20 items taken from the Activation-Deactivation Adjective Check List (Thayer, 1986), 18 from the Positive and Negative Affect Schedule (PANAS, Watson, Clark, & Tellegen, 1988) along with the affective circumplex items used by Larsen and Diener (1992). The response format was a four-point scale that corresponds to Russell and Carroll's (1999) "ambiguous—likely-unipolar format" and that asks the respondents to indicate their current standing ("at this moment") with the following rating scale:

Not at all A little Moderately Very much

The original version of the MSQ included 70 items. Intermediate analyses (done with 1840 subjects) demonstrated a concentration of items in some sections of the two dimensional space, and a paucity of items in others. To begin correcting this, 3 items from redundantly measured sections (alone, kindly, scornful) were removed, and 5 new ones (anxious, cheerful, idle, inactive, and tranquil) were added. Thus, the correlation matrix is missing the correlations between items anxious, cheerful, idle, inactive, and tranquil with alone, kindly, and scornful.

2605 individuals took Form 1 version, 3806 the Form 2 version. 3032 people (1218 form 1, 1814 form 2) took the MSQ at least once. 2086 at least twice, 1112 three times, and 181 four times.

To see the relative frequencies by time and form, see the first example.

Procedure. The data were collected over nine years in the Personality, Motivation and Cognition laboratory at Northwestern, as part of a series of studies examining the effects of personality and situational factors on motivational state and subsequent cognitive performance. In each of 38 studies, prior to any manipulation of motivational state, participants signed a consent form and in some studies, consumed 0 or 4mg/kg of caffeine. In caffeine studies, they waited 30 minutes and then filled out the MSQ. (Normally, the procedures of the individual studies are irrelevant to this data set and could not affect the responses to the MSQ at time 1, since this instrument was completed before any further instructions or tasks. However, caffeine does have an effect.) The MSQ post test following a movie manipulation) is available in affect as well as here.

The XRAY study crossed four movie conditions with caffeine. The first MSQ measures are showing the effects of the movies and caffeine, but after an additional 30 minutes, the second MSQ seems

to mainly show the caffeine effects. The movies were 9 minute clips from 1) a BBC documentary on British troops arriving at the Bergen-Belsen concentration camp (sad); 2) an early scene from Halloween in which the heroine runs around shutting doors and windows (terror); 3) a documentary about lions on the Serengeti plain, and 4) the "birthday party" scene from Parenthood.

The FLAT study measured affect before, immediately after, and then after 30 minutes following a movie manipulation. See the affect data set.

To see which studies used which conditions, see the second and third examples.

The EA and TA scales are from Thayer, the PA and NA scales are from Watson et al. (1988). Scales and items:

Energetic Arousal: active, energetic, vigorous, wakeful, wide.awake, full.of.pep, lively, -sleepy, -tired, - drowsy (ADACL)

Tense Arousal: Intense, Jittery, fearful, tense, clutched up, -quiet, -still, - placid, - calm, -at rest (ADACL)

Positive Affect: active, alert, attentive, determined, enthusiastic, excited, inspired, interested, proud, strong (PANAS)

Negative Affect: afraid, ashamed, distressed, guilty, hostile, irritable, jittery, nervous, scared, upset (PANAS)

The PA and NA scales can in turn can be thought of as having subscales: (See the PANAS-X) Fear: afraid, scared, nervous, jittery (not included frightened, shaky) Hostility: angry, hostile, irritable, (not included: scornful, disgusted, loathing guilt: ashamed, guilty, (not included: blameworthy, angry at self, disgusted with self, dissatisfied with self) sadness: alone, blue, lonely, sad, (not included: downhearted) joviality: cheerful, delighted, energetic, enthusiastic, excited, happy, lively, (not included: joyful) self-assurance: proud, strong, confident, (not included: bold, daring, fearless) attentiveness: alert, attentive, determined (not included: concentrating)

The next set of circumplex scales were taken from Larsen and Diener (1992). High activation: active, aroused, surprised, intense, astonished Activated PA: elated, excited, enthusiastic, lively Unactivated NA: calm, serene, relaxed, at rest, content, at ease PA: happy, warmhearted, pleased, cheerful, delighted Low Activation: quiet, inactive, idle, still, tranquil Unactivated PA: dull, bored, sluggish, tired, drowsy NA: sad, blue, unhappy, gloomy, grouchy Activated NA: jittery, anxious, nervous, fearful, distressed.

Keys for these separate scales are shown in the examples.

In addition to the MSQ, there are 5 scales from the Eysenck Personality Inventory (Extraversion, Impulsivity, Sociability, Neuroticism, Lie). The Imp and Soc are subsets of the the total extraversion scale based upon a reanalysis of the EPI by Rocklin and Revelle (1983). This information is in the msq data set as well.

Note

In December, 2018 the caffeine, film and personality conditions were added. In the process of doing so, it was discovered that the EMIT data had been incorrectly entered. This has been fixed.

Source

Data collected at the Personality, Motivation, and Cognition Laboratory, Northwestern University.

References

Larsen, R. J., & Diener, E. (1992). Promises and problems with the circumplex model of emotion. In M. S. Clark (Ed.), Review of personality and social psychology, No. 13. Emotion (pp. 25-59). Thousand Oaks, CA, US: Sage Publications, Inc.

Rafaeli, Eshkol and Revelle, William (2006), A premature consensus: Are happiness and sadness truly opposite affects? Motivation and Emotion, 30, 1, 1-12.

Revelle, W. and Anderson, K.J. (1998) Personality, motivation and cognitive performance: Final report to the Army Research Institute on contract MDA 903-93-K-0008. (https://www.personality-project.org/revelle/publications/ra.ari.98.pdf).

Smillie, Luke D. and Cooper, Andrew and Wilt, Joshua and Revelle, William (2012) Do Extraverts Get More Bang for the Buck? Refining the Affective-Reactivity Hypothesis of Extraversion. Journal of Personality and Social Psychology, 103 (2), 206-326.

Thayer, R.E. (1989) The biopsychology of mood and arousal. Oxford University Press. New York, NY.

Watson, D., Clark, L.A. and Tellegen, A. (1988) Development and validation of brief measures of positive and negative affect: The PANAS scales. Journal of Personality and Social Psychology, 54(6):1063-1070.

See Also

msq for 3896 participants with scores on five scales of the EPI. affect for an example of the use of some of these adjectives in a mood manipulation study.

make.keys, scoreItems and scoreOverlap for instructions on how to score multiple scales with and without item overlap. Also see fa and fa.extension for instructions on how to do factor analyses or factor extension.

Given the temporal ordering of the sai data and the msqR data, these data are useful for demonstrations of testRetest reliability. See the examples in testRetest for how to combine the sai tai and msqR datasets.

Examples

```
PA =c("active", "excited", "strong", "inspired", "determined", "attentive",
          "interested", "enthusiastic", "proud", "alert"),
NAf =c("jittery", "nervous", "scared", "afraid", "guilty", "ashamed", "distressed",
         "upset", "hostile", "irritable"),
HAct = c("active", "aroused", "surprised", "intense", "astonished"),
aPA = c("elated", "excited", "enthusiastic", "lively"),
uNA = c("calm", "serene", "relaxed", "at.rest", "content", "at.ease"),
pa = c("happy", "warmhearted", "pleased", "cheerful", "delighted" ),
LAct = c("quiet", "inactive", "idle", "still", "tranquil"),
uPA =c( "dull", "bored", "sluggish", "tired", "drowsy"),
naf = c( "sad", "blue", "unhappy", "gloomy", "grouchy"),
aNA = c("jittery", "anxious", "nervous", "fearful", "distressed"), Fear = c("afraid", "scared", "nervous", "jittery"),
Hostility = c("angry" , "hostile", "irritable", "scornful" ),
Guilt = c("guilty", "ashamed"),
Sadness = c("sad", "blue", "lonely", "alone"),
Joviality =c("happy","delighted", "cheerful", "excited", "enthusiastic", "lively", "energetic"),
Self.Assurance=c( "proud", "strong" , "confident" , "-fearful" ),
Attentiveness = c("alert" , "determined" , "attentive" ))
#acquiscence = c("sleepy" , "wakeful" , "relaxed","tense"))
#Yik Russell and Steiger list the following items
Yik.keys <- list(</pre>
pleasure =psych::cs(happy,content,satisfied, pleased),
act.pleasure =psych::cs(proud,enthusiastic,euphoric),
pleasant.activation = psych::cs(energetic,full.of.pep,excited,wakeful,attentive,
   wide.awake,active,alert,vigorous),
activation = psych::cs(aroused, hyperactivated, intense),
unpleasant.act = psych::cs(anxious, frenzied, jittery, nervous),
activated.displeasure =psych::cs(scared,upset,shaky,fearful,clutched.up,tense,
    ashamed, guilty, agitated, hostile),
displeaure =psych::cs(troubled,miserable,unhappy,dissatisfied),
Ueactivated.Displeasure = psych::cs(sad,down,gloomy,blue,melancholy),
Unpleasant.Deactivation = psych::cs(droopy,drowsy,dull,bored,sluggish,tired),
Deactivation =psych::cs( quiet,still),
pleasant.deactivation = psych::cs(placid,relaxed,tranquil, at.rest,calm),
deactived.pleasure =psych::cs( serene, soothed, peaceful, at.ease, secure)
)
#of these 60 items, 46 appear in the msqR
Yik.msq.keys <- list(</pre>
Pleasure =psych::cs(happy,content,satisfied, pleased),
Activated.Pleasure =psych::cs(proud,enthusiastic),
Pleasant.Activation = psych::cs(energetic,full.of.pep,excited,wakeful,attentive,
    wide.awake, active, alert, vigorous),
Activation = psych::cs(aroused,intense),
Unpleasant.Activation = psych::cs(anxious, jittery, nervous),
Activated.Displeasure =psych::cs(scared,upset,fearful,
          clutched.up,tense,ashamed,guilty,hostile),
Displeasure = psych::cs(unhappy),
Deactivated.Displeasure = psych::cs(sad,gloomy,blue),
Unpleasant.Deactivation = psych::cs(drowsy,dull,bored,sluggish,tired),
Deactivation =psych::cs( quiet,still),
```

```
Pleasant.Deactivation = psych::cs(placid,relaxed,tranquil, at.rest,calm),
Deactivated.Pleasure =psych::cs( serene,at.ease)
yik.scores <- psych::scoreItems(Yik.msq.keys,msqR)</pre>
yik <- yik.scores$scores</pre>
f2.yik <- psych::fa(yik,2) #factor the yik scores
psych::fa.plot(f2.yik,labels=colnames(yik),title="Yik-Russell-Steiger circumplex",cex=.8,
      pos=(c(1,1,2,1,1,1,3,1,4,1,2,4)))
msq.scores <- psych::scoreItems(keys.list,msqR)</pre>
#show a circumplex structure for the non-overlapping items
fcirc <- psych::fa(msq.scores$scores[,5:12],2)</pre>
psych::fa.plot(fcirc,labels=colnames(msq.scores$scores)[5:12])
#now, find the correlations corrected for item overlap
msq.overlap <- psych::scoreOverlap(keys.list,msqR)</pre>
f2 <- psych::fa(msq.overlap$cor,2)</pre>
psych::fa.plot(f2,labels=colnames(msq.overlap$cor),
          title="2 dimensions of affect, corrected for overlap")
#extend this solution to EA/TA NA/PA space
fe <- psych::fa.extension(cor(msq.scores$scores[,5:12],msq.scores$scores[,1:4]),fcirc)</pre>
psych::fa.diagram(fcirc,fe=fe,main="Extending the circumplex structure to EA/TA and PA/NA")
#show the 2 dimensional structure
f2 \leftarrow psych::fa(msqR[1:72],2)
psych::fa.plot(f2,labels=colnames(msqR)[1:72],title="2 dimensions of affect at the item level")
#sort them by polar coordinates
round(psych::polar(f2),2)
#the msqR and sai data sets have 10 overlapping items which can be used for
#testRetest analysis. We need to specify the keys, and then choose the appropriate
sai.msq.keys <- list(pos =c( "at.ease" , "calm" , "confident", "content", "relaxed"),</pre>
  neg = c("anxious", "jittery", "nervous", "tense", "upset"),
anx = c("anxious", "jittery", "nervous", "tense", "upset", "-at.ease", "-calm",
  "-confident", "-content", "-relaxed"))
select <- psych::selectFromKeys(sai.msq.keys$anx)</pre>
#The following is useful for examining test retest reliabilities
msq.control <- subset(msqR,is.element( msqR$study , c("Cart", "Fast", "SHED", "SHOP")))</pre>
msq.film <- subset(msqR,(is.element( msqR$study , c("FIAT", "FILM","FLAT","MIXX","XRAY"))
    & (msqR$time < 3) ))
msq.film[((msq.film$study == "FLAT") & (msq.film$time ==3)) ,] <- NA</pre>
msq.drug <- subset(msqR,(is.element( msqR$study , c("AGES","SALT", "VALE", "XRAY")))</pre>
   %(msqR$time < 3))
msq.day <- subset(msqR,is.element( msqR$study , c("SAM", "RIM")))</pre>
```

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NEO correlation matrix from the NEO_PI_R manual

neo

Description

The NEO.PI.R is a widely used personality test to assess 5 broad factors (Neuroticism, Extraversion, Openness, Agreeableness and Conscientiousness) with six facet scales for each factor. The correlation matrix of the facets is reported in the NEO.PI.R manual for 1000 subjects.

Usage

data(neo)

Format

A data frame of a 30 x 30 correlation matrix with the following 30 variables.

- N1 Anxiety
- N2 AngryHostility
- N3 Depression
- N4 Self-Consciousness
- N5 Impulsiveness
- N6 Vulnerability
- E1 Warmth
- E2 Gregariousness
- E3 Assertiveness
- E4 Activity
- E5 Excitement-Seeking
- E6 PositiveEmotions
- O1 Fantasy
- **O2** Aesthetics
- O3 Feelings
- O4 Ideas
- O5 Actions
- O6 Values
- A1 Trust
- A2 Straightforwardness
- A3 Altruism

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- A4 Compliance
- A5 Modesty
- A6 Tender-Mindedness
- C1 Competence
- C2 Order
- C3 Dutifulness
- C4 AchievementStriving
- C5 Self-Discipline
- C6 Deliberation

Details

The past thirty years of personality research has led to a general consensus on the identification of major dimensions of personality. Variously known as the "Big 5" or the "Five Factor Model", the general solution represents 5 broad domains of personal and interpersonal experience. Neuroticism and Extraversion are thought to reflect sensitivity to negative and positive cues from the environment and the tendency to withdraw or approach. Openness is sometimes labeled as Intellect and reflects an interest in new ideas and experiences. Agreeableness and Conscientiousness reflect tendencies to get along with others and to want to get ahead.

The factor structure of the NEO suggests five correlated factors as well as two higher level factors. The NEO was constructed with 6 "facets" for each of the five broad factors.

For a contrasting structure, examine the items of the link{spi} data set (Condon, 2017).

Source

Costa, Paul T. and McCrae, Robert R. (1992) (NEO PI-R) professional manual. Psychological Assessment Resources, Inc. Odessa, FL. (with permission of the author and the publisher)

References

Condon, D. (2017) The SAPA Personality Inventory: An empirically-derived, hierarchically-organized self-report personality assessment model

Digman, John M. (1990) Personality structure: Emergence of the five-factor model. Annual Review of Psychology. 41, 417-440.

John M. Digman (1997) Higher-order factors of the Big Five. Journal of Personality and Social Psychology, 73, 1246-1256.

McCrae, Robert R. and Costa, Paul T., Jr. (1999) A Five-Factor theory of personality. In Pervin, Lawrence A. and John, Oliver P. (eds) Handbook of personality: Theory and research (2nd ed.) 139-153. Guilford Press, New York. N.Y.

Revelle, William (1995), Personality processes, Annual Review of Psychology, 46, 295-328.

Joshua Wilt and William Revelle (2009) Extraversion and Emotional Reactivity. In Mark Leary and Rick H. Hoyle (eds). Handbook of Individual Differences in Social Behavior. Guilford Press, New York, N.Y.

Joshua Wil and William Revelle (2016) Extraversion. In Thomas Widiger (ed) The Oxford Handbook of the Five Factor Model. Oxford University Press.

peas 61

Examples

```
data(neo)
n5 <- psych::fa(neo,5)
neo.keys <- psych::make.keys(30,list(N=c(1:6),E=c(7:12),0=c(13:18),A=c(19:24),C=c(25:30)))
n5p <- psych::target.rot(n5,neo.keys) #show a targeted rotation for simple structure
n5p</pre>
```

peas

Galton's Peas

Description

Francis Galton introduced the correlation coefficient with an analysis of the similarities of the parent and child generation of 700 sweet peas.

Usage

```
data(peas)
```

Format

A data frame with 700 observations on the following 2 variables.

parent The mean diameter of the mother pea for 700 peas child The mean diameter of the daughter pea for 700 sweet peas

Details

Galton's introduction of the correlation coefficient was perhaps the most important contribution to the study of individual differences. This data set allows a graphical analysis of the data set. There are two different graphic examples. One shows the regression lines for both relationships, the other finds the correlation as well.

Source

Stanton, Jeffrey M. (2001) Galton, Pearson, and the Peas: A brief history of linear regression for statistics intstructors, Journal of Statistics Education, 9. (retrieved from the web from https://www.amstat.org/publications/jse reproduces the table from Galton, 1894, Table 2.

The data were generated from this table.

References

Galton, Francis (1877) Typical laws of heredity. paper presented to the weekly evening meeting of the Royal Institution, London. Volume VIII (66) is the first reference to this data set. The data appear in

Galton, Francis (1894) Natural Inheritance (5th Edition), New York: MacMillan).

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See Also

The other Galton data sets: heights, galton, cubits

Examples

```
data(peas)
psych::pairs.panels(peas,lm=TRUE,xlim=c(14,22),ylim=c(14,22),main="Galton's Peas")
psych::describe(peas)
psych::pairs.panels(peas,main="Galton's Peas")
```

Pollack

Pollack et al (2012) correlation matrix for mediation example

Description

A correlation matrix taken from Pollack (2012) with 9 variables. Primarily used as an example for setCor and mediation.

Usage

```
data("Pollack")
```

Format

A correlation matrix based upon 262 participants.

```
sex Male = 1, Female = 0, 62% male
age mean =33
tenure length of employent, mean = 5.9 years
self.efficacy self ratings
competence self rating of competence
social.ties Contact with business-related social ties
economic.stress mean of two items on economic stress
depression 6 items from MAACL measuring depression
withdrawal Withdrawal intentions in domain of entrepreneurship
```

Details

This is the correlation matrix from Pollack et al. (2012) p 797. The raw data are available from the processR package (Keon-Woong Moon, 2020). The data set is used by Hayes in example p 179 in example 3.

Source

Pollack et al. 2012

psychTools 63

References

Pollack, Jeffrey M. and Vanepps, Eric M. and Hayes, Andrew F. (2012). The moderating role of social ties on entrepreneurs' depressed affect and withdrawal intentions in response to economic stress, Journal of Organizational Behavior 33 (6) 789-810.

Hayes, Andrew F. (2013) Introduction to mediation, moderation, and conditional process analysis: A regression-based approach. Guilford Press.

Examples

psych::lowerMat(Pollack)

psychTools: datasets and utility functions to accompany the psych package

Description

PsychTools includes the larger data sets used by the psych package and also includes a few general utility functions such as the read.file and read.clipboard functions. The data sets ara made available for demonstrations of a variety of psychometric functions.

Details

See the various helpfiles listed in the index or as links from here. Also see the main functions in the psych package 00.psych-package.

Data sets from the SAPA/ICAR project:

ability 16 ICAR ability items scored as correct or incorrect for 1525 participants.

igitems multiple choice IQ items (raw responses)

Two data sets of affect and arousal scores as a function of personality and movie conditions
25 Personality items representing 5 factors from the SAPA project for 2800 participants

bfi.dictionary Dictionary of the bfi

epi Eysenck Personality Inventory (EPI) data for 3570 participants

epi.dictionary The items for the epi

epi.bfi 13 personality scales from the Eysenck Personality Inventory and Big 5 inventory

epiR 474 participants took the epi twice

msq 75 mood items from the Motivational State Questionnaire for 3896 participants
75 mood items from the Motivational State Questionnaire for 3032 unique participants
tai Trait Anxiety data from the PMC lab matching the sai sample. 3032 unique subjects
sai State Anxiety data from the PMC lab over multiple occasions. 3032 unique subjects.

sai.dictionary items used in the sai

spi 4000 cases from the SAPA Personality Inventory including an item dictionary and scoring keys.

spi.dictionary spi.keys The items for the spi Scoring keys for the spi

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Historically interesting data sets

burt 11 emotional variables from Burt (1915) galton Galtons Mid parent child height data

A data.frame of the Galton (1888) height and cubit data set heights

cubits Galtons example of the relationship between height and cubit or forearm length

peas Galtons Peas

cushny

26 cognitive variables + 7 demographic variables for 301 cases from Holzinger and Swineford. holzinger.swineford

The data set from Cushny and Peebles (1905) on the effect of three drugs on hours of sleep, used by

Miscellaneous example data sets

blant A 29 x 29 matrix that produces weird factor analytic results

Bonds Logical Operations Test - BLOT blot

cities Distances between 11 US cities and their geograpical location city.location

US family income from US census 2008 income US family income from US census 2008 all.income

neo NEO correlation matrix from the NEO_PI_R manual

The Schutz correlation matrix example from Shapiro and ten Berge Schutz

The Spengler and Damian correlation matrix example from Spengler, Damian and Roberts (2018) Spengler

Another correlation matrix from Spengler, Damian and Roberts (2018) Damian

usaf A correlation of 17 body size (anthropometric) measures from the US Air Force. Adapted from the Anthropometric

Paired comparison of preferences for 9 vegetables (scaling example) veg

Functions to convert various objects to latex

fa2latex Convert a data frame, correlation matrix, or factor analysis output to a LaTeX table df2latex Convert a data frame, correlation matrix, or factor analysis output to a LaTeX table

Convert an ICC analyssis output to a LaTeX table ICC2latex irt2latex Convert an irt analysis output to a LaTeX table Convert a correlation matrix output to a LaTeX table cor2latex

Convert a data frame, correlation matrix, or factor analysis output to a LaTeX table omega2latex

File manipulation functions

fileCreate Create a file

Show the first few lines of multitple files fileScan filesInfo Show the information for all files in a directory filesList Show the names of all files in a directory

df0rder Sorts a data frame

File input/output functions

read.clipboard Shortcuts for reading from the clipboard or a file

```
read.clipboard.csv
read.clipboard.fwf
read.clipboard.lower
read.clipboard.tab
read.clipboard.upper
read.file Read a file according to its suffix
read.file.csv
read.https
write.file Write data to a file
write.file.csv
```

Examples

```
psych::describe(ability)
```

read.file

Shortcuts for reading from the clipboard or a file

Description

Input from a variety of sources may be read. data.frames may be read from files with suffixes of .txt, .text, .TXT, .dat, .DATA,.data, .csv, .rds, rda, .xpt, XPT, or .sav (i.e., data from SPSS sav files may be read as can files saved by SAS using the .xpt option). Data exported by JMP or EXCEL in the csv format are also able to be read. Fixed Width Files saved in .txt mode may be read if the widths parameter is specified. Files saved with writeRDS have suffixes of .rds or Rds, and are read using readRDS. Files associated with objects with suffixes .rda and .Rda are loaded (following a security prompt). The default values for read.spss are adjusted for more standard input from SPSS files. Input from the clipboard is easy but a bit obscure, particularly for Mac users. read.clipboard and its variations are just an easier way to do so. Data may be copied to the clipboard from Excel spreadsheets, csv files, or fixed width formatted files and then into a data.frame. Data may also be read from lower (or upper) triangular matrices and filled out to square matrices. Writing text files may be done using write.file which will prompt for a file name (if not given) and then write or save to that file depending upon the suffix (text, txt, or csv will call write.table, R, or r will dput, rda, Rda will save, Rds,rds will saveRDS).

Usage

```
read.file(file=NULL,header=TRUE,use.value.labels=FALSE,to.data.frame=TRUE,sep=",",
quote="\"", widths=NULL,f=NULL, filetype=NULL,...)
#for .txt, .text, TXT, .csv, .sav, .xpt, XPT, R, r, Rds, .rds, or .rda,
# .Rda, .RData, .Rdata, .dat and .DAT files

read.clipboard(header = TRUE, ...) #assumes headers and tab or space delimited
read.clipboard.csv(header=TRUE,sep=',',...) #assumes headers and comma delimited
read.clipboard.tab(header=TRUE,sep='\t',...) #assumes headers and tab delimited
#read in a matrix given the lower off diagonal
```

```
read.clipboard.lower(diag=TRUE,names=FALSE,...)
read.clipboard.upper(diag=TRUE,names=FALSE,...)

#read in data using a fixed format width (see read.fwf for instructions)
read.clipboard.fwf(header=FALSE,widths=rep(1,10),...)

read.https(filename,header=TRUE)
read.file.csv(file=NULL,header=TRUE,f=NULL,...)

#For output:
write.file(x,file=NULL,row.names=FALSE,f=NULL,...)
write.file.csv(x,file=NULL,row.names=FALSE,f=NULL,...)
```

Arguments

	header	Does the first row have variable labels (generally assumed to be TRUE).
	sep	What is the designated separater between data fields? For typical csv files, this will be a comma, but if commas designate decimals, then a; can be used to designate different records.
	quote	Specified to
	diag	for upper or lower triangular matrices, is the diagonal specified or not
	names	for read.clipboard.lower or upper, are colnames in the the first column
	widths	how wide are the columns in fixed width input. The default is to read 10 columns of size $1.$
	filename	Name or address of remote https file to read.
		Other parameters to pass to read
	f	A file name to read from or write to. If omitted, file.choose is called to dynamically get the file name.
	file	A file name to read from or write to. (same as f, but perhaps more intuitive) If omitted and if f is omitted,then file.choose is called to dynamically get the file name.
	x	The data frame or matrix to write to f
	row.names	Should the output file include the rownames? By default, no.
	to.data.frame	Should the spss input be converted to a data frame?
use.value.labels		
		Should the SPSS input values be converted to numeric?
	filetype	If specified the reading will use this term rather than the suffix.

Details

A typical session of R might involve data stored in text files, generated online, etc. Although it is easy to just read from a file (particularly if using read.file, an alternative is to use one's local system to copy from the file to the clipboard and then read from the clipboard read.clipboard. This

is very convenient (and somewhat more intuitive to the naive user). This is particularly convenient when copying from a text book or article and just moving a section of text into R.) However, copying from a file and then reading the clipboard is hard to automate in a script. Thus, coderead.file will read from a file.

The read.file function combines the file.choose and either read.table, read.fwf, read.spss or read.xport(from foreign) or load or readRDS commands. By examining the file suffix, it chooses the appropriate way to read the file. For more complicated file structures, see the foreign package. For even more complicated file structures, see the rio or haven packages.

Note that read.file assumes by default that the first row has column labels (header =TRUE). If this is not true, then make sure to specify header = FALSE. If the file is fixed width, the assumption is that it does not have a header field. In the unlikely case that a fwf file does have a header, then you probably should try fn <- file.choose() and then my.data <- read.fwf(fn,header=TRUE,widths= widths)

Further note: If the file is a .Rda, .rda, etc. file, the read.file command will return the name and location of the file. It will prompt the user to load this file. In this case, it is necessary to either assign the output (the file name) to an object that has a different name than any of the objects in the file, or to call read.file() without any specification. Notice that loading an .Rda file can overwrite existing objects. Thus the warning and the need to do the second step.

If the file has no suffix the default action is to quit with a warning. However, if the filetype is specified, it will use that type in the reading (e.g. filetype="txt" will read as text file, even if there is no suffix.)

If the file is specified and has a prefix of http:// https:// it will be downloaded and then read.

Currently supported input formats are

.sav	SPSS.sav files
.csv	A comma separated file (e.g. from Excel or Qualtrics)
.txt	A typical text file
.TXT	A typical text file
.text	A typical text file
.data	A data file
.dat	A data file
.rds	A R data file
.Rds	A R data file (created by a write)
.Rda	A R data structure (created using save)
.rda	A R data structure (created using save)
.RData	A R data structure (created using save)
.rdata	A R data structure (created using save)
.R	A R data structure created using dput
.r	A R data structure created using dput
.xpt	A SAS data file in xport format
.XPT	A SAS data file in XPORT format

Some data files have an extra ' in the data (e.g. the NYT covid data base). These files can be read specifying quote ""

The foreign function read.spss is used to read SPSS .sav files using the most common options.

Just as read.spss issues various warnings, so does read.file. In general, these can be ignored. For more detailed information about using read.spss, see the help pages in the foreign package.

If you have a file written by JMP, you must first export to a csv or text file.

The write.file function combines the file.choose and either write.table or saveRDS. By examining the file suffix, it chooses the appropriate way to write. For more complicated file structures, see the foreign package, or the save function in R Base. If no suffix is added, it will write as a .txt file.write.file.csv will write in csv format to an arbitrary file name.

Currently supported output formats are

```
.csv
        A comma separated file (e.g. for reading into Excel)
.txt
        A typical text file
        A typical text file
.text
        A R data file
.rds
       A R data file (created by a write)
.Rds
.Rda A R data structure (created using save)
.rda
        A R data structure (created using save)
.R
        A R data structure created using dput
        A R data structure created using dput
.r
```

Many Excel based files specify missing values as a blank field. When reading from the clipboard, using read.clipboard.tab will change these blank fields to NA.

Sometimes missing values are specified as "." or "999", or some other values. These can be converted by the read.file command specifying what values are missing (e.g., na ="."). See the example for the reading from the remote mtcars.csv file.

read.clipboard was based upon a suggestion by Ken Knoblauch to the R-help listserve.

If the input file that was copied into the clipboard was an Excel file with blanks for missing data, then read.clipboard.tab() will correctly replace the blanks with NAs. Similarly for a csv file with blank entries, read.clipboard.csv will replace empty fields with NA.

read.clipboard.lower and read.clipboard.upper are adapted from John Fox's read.moments function in the sem package. They will read a lower (or upper) triangular matrix from the clipboard and return a full, symmetric matrix for use by factanal, fa, ICLUST, pca. omega, etc. If the diagonal is false, it will be replaced by 1.0s. These two function were added to allow easy reading of examples from various texts and manuscripts with just triangular output.

Many articles will report lower triangular matrices with variable labels in the first column. read.clipboard.lower will handle this case. Names must be in the first column if names=TRUE is specified.

Other articles will report upper triangular matrices with variable labels in the first row. read.clipboard.upper will handle this. Note that labels in the first column will not work for read.clipboard.upper. The names, if present, must be in the first row.

Consider the following lower triangular matrix. To read it, copy it to the clipboard and read.clipboard.lower(names=TRUE)

A1 1.00 A2 -0.34 1.00 A3 -0.27 0.49 1.00 A4 -0.15 0.34 0.36 1.00

A5 -0.18 0.39 0.50 0.31 1.00 C1 0.03 0.09 0.10 0.09 0.12 1.00

However, if the data are strung out e.g.,

-.34 -.27 -.15 -.18 .03 .49 .34 .39 .09 .36 .50 .10

Then one needs to read it using the read.clipboard.upper(names=FALSE,diag=FALSE) option. read.clipboard.fwf will read fixed format files from the clipboard. It includes a patch to read.fwf which will not read from the clipboard or from remote file. See read.fwf for documentation of how to specify the widths.

Value

the contents of the file to be read or of the clipboard.

Author(s)

William Revelle

Examples

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```
filesList() #search the system for a particular file and then list all the files in that directory
fileCreate() #search for a particular directory and create a file there.
write.file(Thurstone) #open the search window, choose a location and name the output file,
# write the data file (e.g., Thurstone ) to the file chosen

#the example data set from read.delim in the readr package to read a remote csv file
my.data <-read.file(
"https://github.com/tidyverse/readr/raw/master/inst/extdata/mtcars.csv",
na=".") #the na option is used for an example, but is not needed for these data

#These functions read from the local clipboard and thus are interactive
my.data <- read.clipboard() #space delimited columns
my.data <- read.clipboard.csv() # , delimited columns
my.data <- read.clipboard.tab() #typical input if copied from a spreadsheet
my.data <- read.clipboard.lower()

my.matrix <- read.clipboard.lower()
}</pre>
```

sai

State Anxiety data from the PMC lab over multiple occasions.

Description

State Anxiety was measured two-three times in 11 studies at the Personality-Motivation-Cognition laboratory. Here are item responses for 11 studies (9 repeated twice, 2 repeated three times). In all studies, the first occasion was before a manipulation. In some studies, caffeine, or movies or incentives were then given to some of the participants before the second and third STAI was given. In addition, Trait measures are available and included in the tai data set (3032 subjects).

Usage

```
data(sai)
data(tai)
data(sai.dictionary)
```

Format

A data frame with 3032 unique observations on the following 23 variables.

```
id a numeric vector study a factor with levels ages cart fast film flat home pat rob salt shedshop xray time 1=First, 2 = Second, 3=third administration

TOD TOD (time of day 1= 8:50-9:30 am,2 = 1=3 pm, 3= 7:-8pm drug (placebo (0) vs. caffeine (1))
```

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```
film film (1=Frontline (concentration camp), 2 = Halloween 3= National Geographic (control), 4-
     Parenthood (humor)
anxious anxious
at.ease at ease
calm calm
comfortable comfortable
confident confident
content content
high.strung high.strung
jittery jittery
joyful joyful
nervous nervous
pleasant pleasant
rattled over-excited and rattled
regretful regretful
relaxed relaxed
rested rested
secure secure
tense tense
upset upset
worried worried
worrying worrying
```

Details

The standard experimental study at the Personality, Motivation and Cognition (PMC) laboratory (Revelle and Anderson, 1997) was to administer a number of personality trait and state measures (e.g. the epi, msq, msqR and sai) to participants before some experimental manipulation of arousal/effort/anxiety. Following the manipulation (with a 30 minute delay if giving caffeine/placebo), some performance task was given, followed once again by measures of state arousal/effort/anxiety.

Here are the item level data on the sai (state anxiety) and the tai (trait anxiety). Scores on these scales may be found using the scoring keys. The affect data set includes pre and post scores for two studies (flat and maps) which manipulated state by using four types of movies.

In addition to being useful for studies of motivational state, these studies provide examples of testretest and alternate form reliabilities. Given that 10 items overlap with the msqR data, they also allow for a comparison of immediate duplication of items with 30 minute delays.

Studies CART, FAST, SHED, RAFT, and SHOP were either control groups, or did not experimentally vary arousal/effort/anxiety.

AGES, CITY, EMIT, RIM, SALT, and XRAY were caffeine manipulations between time 1 and 2 (RIM and VALE were repeated day 1 and day 2)

72 sai

FIAT, FLAT, MAPS, MIXX, and THRU were 1 day studies with film manipulation between time 1 and time 2.

SAM1 and SAM2 were the first and second day of a two day study. The STAI was given once per day. MSQ not MSQR was given.

VALE and PAT were two day studies with the STAI given pre and post on both days

RIM was a two day study with the STAI and MSQ given once per day.

Usually, time of day 1 = 8:50-9am am, and 2 = 7:30 pm, however, in rob, with paid subjects, the times were 0530 and 22:30.

Source

Data collected at the Personality, Motivation, and Cognition Laboratory, Northwestern University, between 1991 and 1999.

References

Charles D. Spielberger and Richard L. Gorsuch and R. E. Lushene, (1970) Manual for the State-Trait Anxiety Inventory.

Revelle, William and Anderson, Kristen Joan (1997) Personality, motivation and cognitive performance: Final report to the Army Research Institute on contract MDA 903-93-K-0008

Rafaeli, Eshkol and Revelle, William (2006), A premature consensus: Are happiness and sadness truly opposite affects? Motivation and Emotion, 30, 1, 1-12.

Smillie, Luke D. and Cooper, Andrew and Wilt, Joshua and Revelle, William (2012) Do Extraverts Get More Bang for the Buck? Refining the Affective-Reactivity Hypothesis of Extraversion. Journal of Personality and Social Psychology, 103 (2), 206-326.

Examples

```
data(sai)
table(sai$study,sai$time) #show the counts for repeated measures
#Here are the keys to score the sai total score, positive and negative items
sai.keys <- list(sai = c("tense","regretful", "upset", "worrying", "anxious", "nervous",</pre>
"jittery" , "high.strung", "worried" , "rattled","-calm",
"-secure","-at.ease","-rested","-comfortable", "-confident","-relaxed", "-content",
"-joyful", "-pleasant" ) ,
sai.p = c("calm","at.ease","rested","comfortable", "confident", "secure", "relaxed",
       "content", "joyful", "pleasant"),
sai.n = c( "tense" , "anxious", "nervous" , "jittery" , "rattled",
                                                                          "high.strung",
         "upset", "worrying","worried","regretful" )
tai.keys <- list(tai=c("-pleasant" ,"nervous" , "not.satisfied", "wish.happy",</pre>
   "failure","-rested", "-calm", "difficulties", "worry", "-happy",
   "disturbing.thoughts","lack.self.confidence",
  "-secure", "decisive" , "inadequate", "-content", "thoughts.bother", "disappointments" , "-steady" , "tension" ),  
   tai.pos = c("pleasant", "-wish.happy", "rested", "calm", "happy" , "secure",
   "content", "steady" ),
```

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```
tai.neg = c("nervous", "not.satisfied", "failure", "difficulties", "worry",
   "disturbing.thoughts" , "lack.self.confidence", "decisive", "inadequate" ,
    "thoughts.bother", "disappointments", "tension" )
#using the is.element function instead of the %in% function
#just get the control subjects
control <- subset(sai,is.element(sai$study,c("Cart", "Fast", "SHED", "RAFT", "SHOP")) )</pre>
#pre and post drug studies
drug <- subset(sai,is.element(sai$study, c("AGES","CITY","EMIT","SALT","VALE","XRAY")))</pre>
#pre and post film studies
film <- subset(sai,is.element(sai$study, c("FIAT", "FLAT", "MAPS", "MIXX") ))</pre>
#this next set allows us to score those sai items that overlap with the msq item sets
msq.items <- c("anxious", "at.ease" ,"calm", "confident","content", "jittery",</pre>
"nervous", "relaxed", "tense", "upset") #these overlap with the msq
sai.msq.keys <- list(pos =c( "at.ease" , "calm" , "confident", "content", "relaxed"),</pre>
 neg = c("anxious", "jittery", "nervous", "tense" , "upset"),
 anx = c("anxious", "jittery", "nervous", "tense", "upset", "-at.ease" , "-calm" ,
  "-confident", "-content", "-relaxed"))
sai.not.msq.keys <- list(pos=c( "secure","rested","comfortable","joyful", "pleasant"),</pre>
   neg=c("regretful","worrying", "high.strung","worried", "rattled" ),
   anx = c("regretful", "worrying", "high.strung", "worried", "rattled",
                                                                              "-secure",
   "-rested", "-comfortable", "-joyful", "-pleasant" ))
sai.alternate.forms <- list( pos1 =c( "at.ease","calm","confident","content","relaxed"),</pre>
 neg1 = c("anxious", "jittery", "nervous", "tense", "upset"),
 \verb"anx1 = c("anxious", "jittery", "nervous", "tense", "upset", "-at.ease", "-calm",
      "-confident", "-content", "-relaxed"),
 pos2=c( "secure","rested","comfortable" ,"joyful" , "pleasant" ),
 neg2=c("regretful","worrying", "high.strung","worried", "rattled" ),
 anx2 = c("regretful","worrying", "high.strung","worried", "rattled", "-secure",
    "-rested", "-comfortable", "-joyful", "-pleasant" ))
sai.repeated <- c("AGES","Cart","Fast","FIAT","FILM","FLAT","HOME","PAT","RIM","SALT",
    "SAM", "SHED", "SHOP", "VALE", "XRAY")
sai12 <- subset(sai,is.element(sai$study, sai.repeated)) #the subset with repeated measures
#Choose those studies with repeated measures by :
sai.control <- subset(sai,is.element(sai$study, c("Cart", "Fast", "SHED", "SHOP")))</pre>
sai.film <- subset(sai,is.element(sai$study, c("FIAT","FLAT") ) )</pre>
sai.drug <- subset(sai,is.element(sai$study, c("AGES", "SALT", "VALE", "XRAY")))</pre>
sai.day <- subset(sai,is.element(sai$study, c("SAM", "RIM")))</pre>
```

74 sat.act

Description

Self reported scores on the SAT Verbal, SAT Quantitative and ACT were collected as part of the Synthetic Aperture Personality Assessment (SAPA) web based personality assessment project. Age, gender, and education are also reported. The data from 700 subjects are included here as a demonstration set for correlation and analysis.

Usage

```
data(sat.act)
```

Format

A data frame with 700 observations on the following 6 variables.

```
gender males = 1, females = 2
education self reported education 1 = high school ... 5 = graduate work
age age
ACT ACT composite scores may range from 1 - 36. National norms have a mean of 20.
SATV SAT Verbal scores may range from 200 - 800.
SATQ SAT Quantitative scores may range from 200 - 800
```

Details

hese items were collected as part of the SAPA project (https://www.sapa-project.org/)to develop online measures of ability (Revelle, Wilt and Rosenthal, 2009). The score means are higher than national norms suggesting both self selection for people taking on line personality and ability tests and a self reporting bias in scores.

See also the iq.items data set.

Source

```
https://personality-project.org/
```

References

Revelle, William, Wilt, Joshua, and Rosenthal, Allen (2009) Personality and Cognition: The Personality-Cognition Link. In Gruszka, Alexandra and Matthews, Gerald and Szymura, Blazej (Eds.) Handbook of Individual Differences in Cognition: Attention, Memory and Executive Control, Springer.

Examples

```
data(sat.act)
psych::describe(sat.act)
psych::pairs.panels(sat.act)
```

Schutz 75

Schutz

The Schutz correlation matrix example from Shapiro and ten Berge

Description

Shapiro and ten Berge use the Schutz correlation matrix as an example for Minimum Rank Factor Analysis. The Schutz data set is also a nice example of how normal minres or maximum likelihood will lead to a Heywood case, but minrank factoring will not.

Usage

```
data("Schutz")
```

Format

The format is: num [1:9, 1:9] 1 0.8 0.28 0.29 0.41 0.38 0.44 0.4 0.41 0.8 ... - attr(*, "dimnames")=List of 2 ..\$:1] "Word meaning" "Odd Words" "Boots" "Hatchets"\$: chr [1:9] "V1" "V2" "V3" "V4" ...

Details

These are 9 cognitive variables of importance mainly because they are used as an example by Shapiro and ten Berge for their paper on Minimum Rank Factor Analysis.

The solution from the fa function with the fm='minrank' option is very close (but not exactly equal) to their solution.

This example is used to show problems with different methods of factoring. Of the various factoring methods, fm = "minres", "uls", or "mle" produce a Heywood case. Minrank, alpha, and pa do not.

See the blant data set for another example of differences across methods.

Source

Richard E. Schutz, (1958) Factorial Validity of the Holzinger-Crowdeer Uni-factor tests. Educational and Psychological Measurement, 48, 873-875.

References

Alexander Shapiro and Jos M.F. ten Berge (2002) Statistical inference of minimum rank factor analysis. Psychometrika, 67. 70-94

Examples

```
data(Schutz)
psych::corPlot(Schutz,numbers=TRUE,upper=FALSE)

f4min <- psych::fa(Schutz,4,fm="minrank") #for an example of minimum rank factor Analysis
#compare to
f4 <- psych::fa(Schutz,4,fm="mle") #for the maximum likelihood solution which has a Heywood case</pre>
```

76 Spengler

Spengler

Project Talent data set from Marion Spengler and Rodica Damian

Description

Project Talent gave 440,000 US high school students a number of personality and ability tests. Of these, the data fror 346,000 were available for followup. Subsequent followups were collected 11 and 50 years later. Marion Spengler and her colleagues Rodica Damian, and Brent Roberts reported on the stability and change across 50 years of personality and ability. Here is the correlation matrix of 25 of their variables (Spengler) as well as a slightly different set of 19 variables (Damian). This is a nice example of mediation and regression from a correlation matrix.

Usage

```
data("Damian")
```

Format

A 25 x 25 correlation matrix of demographic, personality, and ability variables, based upon 346,660 participants.

Race/Ethnicity 1 = other, 2 = white/caucasian

Sex 1=Male, 2=Female

Age Cohort =9th grade, 10th grade, 11th grade, 12th grade

Parental Parental SES based upon 9 questions of home value, family income, etc.

IQ Standardized composite of Verbal, Spatial and Mathematical

Sociability etc. 10 scales based upon prior work by Damian and Roberts

Maturity A higher order factor from the prior 10 scales

Extraversion The second higher order factor

Interest Self reported interest in school

Reading Self report reading skills

Writing Self report writing skills

Responsible Self reported responsibility scale

Ed.11 Education level at 11 year followup

Educ. 50 Education level at 50 year followup

OccPres.11 Occupational Prestige at 11 year followup

OccPres. 50 Occupational Prestige at 50 year followup

Income.11 Income at 11 year followup

Income. 50 Income at 50 year followup

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Details

Data from Project Talent was collected in 1960 on a representative sample of American high school students. Subsequent follow up 11 and 50 years later are reported by Spengler et al (2018) and others.

Source

Marion Spengler, supplementary material to Damian et al. and Spengler et al.

References

Rodica Ioana Damian and Marion Spengler and Andreea Sutu and Brent W. Roberts, 2019, Sixteen going on sixty-six: A longitudinal study of personality stability and change across 50 years Journal of Personality and Social Psychology, 117, (3) 274-695.

Marian Spengler and Rodica Ioana Damian and Brent W. Roberts (2018), How you behave in school predicts life success above and beyond family background, broad traits, and cognitive ability Journal of Personality and Social Psychology, 114 (4) 600-636

Examples

spi

A sample from the SAPA Personality Inventory including an item dictionary and scoring keys.

Description

The SPI (SAPA Personality Inventory) is a set of 135 items primarily selected from International Personality Item Pool (ipip.ori.org). This is an example data set collected using SAPA procedures the sapa-project.org web site. This data set includes 10 demographic variables as well. The data set with 4000 observations on 145 variables may be used for examples in scale construction and validation, as well as empirical scale construction to predict multiple criteria.

Usage

```
data("spi")
data(spi.dictionary)
data(spi.keys)
```

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Format

A data frame with 4000 observations on the following 145 variables. (The q numbers are the SAPA item numbers).

```
age Age in years from 11 -90
```

sex Reported biological sex (coded by X chromosones => 1=Male, 2 = Female)

health Self rated health 1-5: poor, fair, good, very good, excellent

p1edu Parent 1 education

p2edu Parent 2 education

education Respondents education: less than 12, HS grad, current univ, some univ, associate degree, college degree, in grad/prof, grad/prof degree

wellness Self rated "wellnes" 1-2

exer Frequency of exercise: very rarely, < 1/month, < 1/wk, 1 or 2 times/week, 3-5/wk, > 5 times/week

smoke never, not last year, < 1/month, <1/week, 1-3 days/week, most days, up to 5 x /day, up to 20 x /day, > 20x/day

ER Emergency room visits none, 1x, 2x, 3 or more times

q_253 see the spi.dictionary for these items (q_253

q_1328 see the dictionary for all items q_1328)

Details

Using the data contributed by about 125,000 visitors to the https://www.SAPA-project.org/website, David Condon has developed a hierarchical framework for assessing personality at two levels. The higher level has the familiar five factors that have been studied extensively in personality research since the 1980s – Conscientiousness, Agreeableness, Neuroticism, Openness, and Extraversion. The lower level has 27 factors that are considerably more narrow. These were derived based on administrations of about 700 public-domain IPIP items to 3 large samples. Condon describes these scales as being "empirically-derived" because relatively little theory was used to select the number of factors in the hierarchy and the items in the scale for each factor (to be clear, he means relatively little personality theory though he relied on quite a lot of sampling and statistical theory). You can read all about the procedures used to develop this framework in his book/manual. If you would like to reproduce these analyses, you can download the data files from Dataverse (links are also provided in the manual) and compile this script in R (he used knitR). Instructions are provided in the Preface to the manual.

The content of the spi items may be seen by examining the spi.dictionary. Included in the dictionary are the item_id number from the SAPA project, the wording of the item, the source of the item, which Big 5 scale the item marks, and which "Little 27" scale the item marks.

This small subset of the data is provided for demonstration purposes.

Source

https://sapa-project.org/research/SPI/SPIdevelopment.pdf.

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References

Condon, D. (2017) The SAPA Personality Inventory: An empirically-derived, hierarchically-organized self-report personality assessment model (https://psyarxiv.com/sc4p9/)

An analysis using the spi data set and various tools from the psych package may be found at

Revelle, Dworak and Condon, (2021) Exploring the persone: the power of the item in understanding personality structure. Personality and Individual Differences, 169, 1. Doi: 10.1016/j.paid.2020.109905.

Examples

```
data(spi)
data(spi.dictionary)
psych::bestScales(spi, criteria="health",dictionary=spi.dictionary)

sc <- psych::scoreVeryFast(spi.keys,spi) #much faster scoring for just scores
sc <- psych::scoreOverlap(spi.keys,spi) #gives the alpha reliabilities and various stats
    #these are corrected for overlap
psych::corPlot(sc$corrected,numbers=TRUE,cex=.4,xlas=2,min.length=6,
    main="Structure of SPI (Corrected for overlap) disattenuated r above the diagonal)")</pre>
```

usaf

17 anthropometric measures from the USAF showing a general factor

Description

The correlation matrix of 17 anthropometric measures from the United States Air Force survey of 2420 airmen. The data are taken from the Anthropometry package and included here as a demonstration of a hierarchical factor structure suitable for analysis by the omega or omegaSem.

Usage

```
data("USAF")
```

Format

```
The format is: num [1:17, 1:17] 1 0.1148 -0.0309 -0.028 -0.0908 ... - attr(*, "dimnames")=List of 2 ..$: chr [1:17] "age" "weight" "grip" "height" ... ..$: chr [1:17] "age" "weight" "grip" "height" ...
```

Details

The original data were collected by the USAF and reported in Churchill et al, 1977. They are included as a data file of 2420 participants and 202 variables (the first being an id) in the Anthropometry package. The list of variable names may be found in Churchill et al, on pages 96-99.

The three (correlated) factor structure shows a clear height, bulk, and head size structure with an overall general factor (g) which may be interpreted as body size.

The variables included (and their variable numbers in Antropometry) are:

age V1

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weight	V2
grip strength	V12
height (stature)	V13
leg length	V26
knee height	V37
upper arm	V42
thumb tip reach	V47
in sleeve	V49
chest breadth	V52
hip breadth	V55
waist circumference	V71
thigh circumference	V97
scye circumference	V103
head circumference	V141
bitragion coronal	V145
head length	V150
glabella to wall	V181
external canthus to wall	V183

Note that these numbers are equivalant to the numbers in Churchill et al. The numbers in Anthropometry are these + 1.

Source

Guillermo Vinue, Anthropometry: An R Package for Analysis of Anthropometric Data, Journal of Statistical Software, (2017), 77, 6.

References

Edmund Churchill, Thomas Churchill, Paul Kikta (1977) The AMRL anthropmetric data bank library, volumes I-V. (Technical report AMRL-TR-77-1) https://apps.dtic.mil/dtic/tr/fulltext/u2/a047314.pdf

Guillermo Vinue, Anthropometry: An R Package for Analysis of Anthropometric Data, Journal of Statistical Software, (2017), 77, 6.

Examples

```
data(USAF)
psych::corPlot(USAF,xlas=3)
psych::omega(USAF[c(4:8,10:19),c(4:8,10:19)]) #just the size variables
```

Utility

Useful utility functions for file/directory exploration and manipulation.

Utility 81

Description

Wrappers for dirname, file.choose, readLines. file.create, file.path to be called directly for listing directories, creating files, showing the files in a directory, and listing the content of files in a directory. fileCreate gives the functionality of file.choose(new=TRUE). filesList combines file.choose, dirname, and list.files to show the files in a directory, fileScan extends this and then returns the first few lines of each readable file

Usage

```
fileScan(f = NULL, nlines = 3, max = NULL, from = 1, filter = NULL)
filesList(f=NULL)
filesInfo(f=NULL,max=NULL)
fileCreate(newName="new.file")
```

Arguments

f File path to use as base path (will use file.choose() if missing. If f is a directory,

will list the files in that directory, if f is a file, will find the directory for that file

and then list all of those files.)

nlines How many lines to display

max maximum number of files to display

from First file (number) to display

filter Just display files with "filter" in the name

newName The name of the file to be created.

Details

Just a collection of simple wrappers to powerful core R functions. Allows the user more direct control of what directory to list, to create a file, or to display the content of files. The functions called include file.choose, file.path, file.info,file.create, dirname, and dir.exists. All of these are very powerful functions, but not easy to call interactively.

fileCreate will ask to locate a file using file.choose, set the directory to that location, and then prompt to create a file with the new.name. This is a workaround for file.choose(new=TRUE) which only works for Macs not using R.studio.

filesInfo will interactively search for a file and then list the information (size, date, ownership) of all the files in that directory.

filesList will interactively search for a file and then list all the files in same directory.

Note

Work arounds for core-R functions for interactive file manipulation

Author(s)

William Revelle

82 vegetables

See Also

read.file to read in data from a file or read.clipboard from the clipboard. dfOrder to sort data.frames.

Examples

```
if(interactive()) {
#all of these require interactive input and thus are not given as examples
fileCreate("my.new.file.txt")
filesList()  #show the items in the directory where a file is displayed
fileScan() #show the content of the files in a directory
#or, if you have a file in mind
  f <- file.choose()  #go find it
filesList(f)
fileScan(f)
}</pre>
```

vegetables

Paired comparison of preferences for 9 vegetables

Description

A classic data set for demonstrating Thurstonian scaling is the preference matrix of 9 vegetables from Guilford (1954). Used by Guiford, Nunnally, and Nunally and Bernstein, this data set allows for examples of basic scaling techniques.

Usage

```
data(vegetables)
```

Format

A data frame with 9 choices on the following 9 vegetables. The values reflect the perecentage of times where the column entry was preferred over the row entry.

```
Turn Turnips
Cab Cabbage
Beet Beets
Asp Asparagus
Car Carrots
Spin Spinach
S.Beans String Beans
Peas Peas
Corn Corn
```

zola 83

Details

Louis L. Thurstone was a pioneer in psychometric theory and measurement of attitudes, interests, and abilities. Among his many contributions was a systematic analysis of the process of comparative judgment (thurstone, 1927). He considered the case of asking subjects to successively compare pairs of objects. If the same subject does this repeatedly, or if subjects act as random replicates of each other, their judgments can be thought of as sampled from a normal distribution of underlying (latent) scale scores for each object, Thurstone proposed that the comparison between the value of two objects could be represented as representing the differences of the average value for each object compared to the standard deviation of the differences between objects. The basic model is that each item has a normal distribution of response strength and that choice represents the stronger of the two response strengths. A justification for the normality assumption is that each decision represents the sum of many independent inputs and thus, through the central limit theorem, is normally distributed.

Thurstone considered five different sets of assumptions about the equality and independence of the variances for each item (Thurston, 1927). Torgerson expanded this analysis slightly by considering three classes of data collection (with individuals, between individuals and mixes of within and between) crossed with three sets of assumptions (equal covariance of decision process, equal correlations and small differences in variance, equal variances).

This vegetable data set is used by Guilford and by Nunnally to demonstrate Thurstonian scaling.

Source

Guilford, J.P. (1954) Psychometric Methods. McGraw-Hill, New York.

References

Nunnally, J. C. (1967). Psychometric theory., McGraw-Hill, New York.

Revelle, W. An introduction to psychometric theory with applications in R. (in preparation), Springer. $\label{eq:springer} \texttt{https://personality-project.org/r/book/}$

See Also

thurstone

Examples

data(vegetables)
psych::thurstone(veg)

zola

Correlation matrix of 135 self report and 30 peer report personality items

84 zola

Description

Zola et al., (2021) reported the validity of self report personality items from the SAPA personality inventory (SPI) (Condon, 2018) in terms of 30 peer reports on 8 dimensions. Here are the polychoric correlations of these items. spi items were collected using SAPA procedures for 158,631 participants (mean n/item = 18,180), 908 of whom received peer ratings.

Usage

```
data("zola")
```

Format

```
The format is: num [1:165, 1:165] 1 -0.242 0.282 0.65 0.223 ... - attr(*, "dimnames")=List of 2 ..$ : chr [1:165] "q_253" "q_4296" "q_1855" "q_90" ... ..$ : chr [1:165] "q_253" "q_4296" "q_1855" "q_90" ...
```

Details

The polychoric correlation matrix of the spi and peer report data. To see the item labels, use the lookupFromKeys.

This data set is a nice example of a multi-trait, multi-method correlation matrix. (see the scoring example). Five dimensions of self report show high correlations with the corresonding peer report scales.

Source

```
A. Zola, D.M. Condon, and W. Revelle, (2021)
```

References

A. Zola, D.M. Condon, and W. Revelle, (2021) The Convergence of Self and Informant Reports in a Large Online Sample, Collabra: Psychology, 7, 1. doi: 10.1525/collabra.25983

Examples

```
data(zola)
psych::lookupFromKeys(zola.keys,zola.dictionary)
scores <- psych::scoreOverlap(zola.keys[c(1:5,33:37)],zola) #MTMM of Big 5
scores</pre>
```

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