

Psychology 405: Psychometric Theory

Scale Construction: an example

William Revelle

Department of Psychology
Northwestern University
Evanston, Illinois USA



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Outline

- 1 Steps towards scale construction
- 2 A demonstration
- 3 Preliminary steps
 - Data checking
- 4 Score the scales
- 5 Determining how many constructs are in a set of items
- 6 Scoring the alternative solutions
- 7 Show the items

Scale construction: A 10 steps program

- 1 Personality scales are not created in a theoretical vacuum. Perhaps the most important step in developing a new scale is a consideration of what is the construct of interest. What is it, what are manifestations of it, what is it not, and what should it not relate to.
- 2 Then, what is the population of interest? Are they young old, highly literate, or somewhat challenged by literacy. Write items suitable for the population of interest.
- 3 Give the items to the participants. Make sure that they are engaged in the task.

Scale construction: A 10 steps program (continued)

- 4 To analyze the data, it is necessary to enter the data into a machine readable form.
 - This is a source of error. Double check for data entry errors.
 - Double entry (two different people enter the data and then the two files are automatically compared) is recommended.
 - Even better is automatic data entry (but then you need to check and double check the program).
 - `my.data <- read.table(myfile)`
 - `my.data <- read.clipboard()`
- 5 Run basic descriptive statistics to do one more check for errors. Graphically check as well.
 - `describe()`
 - `pairs.panels()`
- 6 Form the variance/covariance matrix from the items and examine the dimensionality of the resulting space.

Scale construction: A 10 steps program (continued)

- 7 Apply various data reduction techniques (factor analysis, principal components analysis, cluster analysis).
 - fa
 - principal
 - iclust
- 8 Form composite scales of the selected items. Check these scales for various measures of internal consistency.
 - make.keys
 - score.items
- 9 Discriminant validity requires that the scales not correlate with other, unrelated traits.
- 10 Convergent validity requires that the scale do correlate with other, alternative measures of the same trait.

Basic item development

As a demonstration of scale construction and validation, consider the following problem. N self report items are given to a number of people. This inventory has is composed of subsets of items that measure believed to measure different traits. In addition, each subject is rated by a friend on those same traits. There are several questions we can ask of these data:

- 1 Do the items form reliable scales?
- 2 What are the correlations of these scales?
- 3 Do the scales correlate with the peer ratings?
- 4 Can we empirically find a better structure of the items?
- 5 Do these revised scales show greater independence, reliability, and validity?

Item writing

To show the procedures, 14 students in a personality research course spent several weeks learning about each of four personality dimensions. Each student then wrote five items to assess each of four constructs.

- 1 Need for Achievement
- 2 Anxiety
- 3 Sociability
- 4 Impulsivity

As a group they examined all of the items and formed the best 84 items into one questionnaire with 21 items believed to measure each of the constructs. They administered this questionnaire to approximately ten friends each whom they also rated on these four constructs. Thus, we have a data set of about 140 participants assessed on 88 items (the 84 self report items and the 4 peer ratings). These four sets of items can be seen as samples from four domains.

Initial data reading

The data, item labels, and scoring keys are saved on a web server. They may be accessed by the `read.table(file.name)` command. We then use the `dim` command to find out the dimensions of the data file as well as the `names` command to find out what the names are.

```
prq.data.name <- "http://personality-project.org/revelle/syllabi/301/prq.data"
prq.keys.name <- "http://personality-project.org/revelle/syllabi/301/prq.keys"
prq.labels.name <- "http://personality-project.org/revelle/syllabi/301/prq.labels"
prq.data <- read.table(prq.data.name,header=TRUE)
prq.keys <- read.table(prq.keys.name,header=TRUE)
prq.labels <- read.table(prq.labels.name,header=TRUE)
dim(prq.data)
names(prq.data)
#only 75 subjects!

[1] 75 91
> names(prq.data)
 [1] "Exp"          "Subject"      "NeedAch"      "Anxiety"      "Sociability"  "Impulsivity"
 [7] "Gender"       "q1"           "q2"           "q3"           "q4"           "q5"
[13] "q6"          "q7"           "q8"           "q9"           "q10"          "q11"
[19] "q12"         "q13"          "q14"          "q15"          "q16"          "q17"
[25] "q18"         "q19"          "q20"          "q21"          "q22"          "q23"
[31] "q24"         "q25"          "q26"          "q27"          "q28"          "q29"
[37] "q30"         "q31"          "q32"          "q33"          "q34"          "q35"
...
[85] "q78"         "q79"          "q80"          "q81"          "q82"          "q83"
[91] "q84"
```


Data checking

The first two variables are not particularly interesting, so we create a new data.frame without them. Then find the descriptive statistics of the data in order to make sure that the data were entered correctly.

```
> prq.items <- prq.data[, -c(1:2)]
> describe(prq.items)
```

	var	n	mean	sd	median	trimmed	mad	min	max	range	skew	kurtosis	se
NeedAch	1	75	6.39	1.92	7	6.48	1.48	2	10	8	-0.40	-0.64	0.22
Anxiety	2	75	5.24	2.28	5	5.21	2.97	1	10	9	0.09	-1.18	0.26
Sociability	3	75	6.15	2.13	7	6.31	1.48	1	9	8	-0.69	-0.60	0.25
Impulsivity	4	75	5.16	2.35	5	5.20	2.97	1	9	8	-0.13	-1.32	0.27
Gender	5	74	1.51	0.50	2	1.52	0.00	1	2	1	-0.05	-2.02	0.06
q1	6	75	4.27	1.15	4	4.34	1.48	1	6	5	-0.52	-0.08	0.13
q2	7	75	3.37	1.39	3	3.33	1.48	1	6	5	0.21	-0.73	0.16
...													
q82	87	75	3.84	1.46	4	3.92	1.48	1	6	5	-0.39	-0.74	0.17
q83	88	75	4.08	1.33	4	4.10	1.48	2	6	4	-0.35	-1.06	0.15
q84	89	75	3.89	1.33	4	3.92	1.48	1	6	5	-0.32	-0.80	0.15

Data checking

In doing this, we discovered (on the first pass through the data) that one of the variables had a range of 32 rather than the 6 that was appropriate. Correcting the data, we can start over again. Even with well meaning, careful data entry, mistakes will happen in data entry. It is recommended that data be entered twice and then compared using software that compares the two files line by line and entry by entry. In all cases, make sure to describe the data and check that the ranges are appropriate for the data.

Thus, the data were edited and the prior steps were done again until there were no incorrectly entered subjects. One error that makes data checking complicated is a blank field in Excel is read improperly. Using NA to specify not available is better. Note that the describe output shows that some variables do not have as many subjects as others.

Score the scales

- 1 Forming scale scores as linear sums (or averages) of the items is easy to do in R.
- 2 One technique (not recommended) is to do a series of recodings, creating new variables for each scale.
- 3 A simpler technique, using the `score.items` function from the *psych* package does this for all scales defined in a matrix of keys (the keys matrix).
- 4 This is essentially a matrix of -1, 0, and 1s where 0 means don't include the item in the scale, and a 1 means to include it. -1 means to reverse key the item.

A keys matrix

```
> prq.keys
```

	PNach	PAnx	PSoc	PImp	G	Nach	Anx	Soc	Imp
1	1	0	0	0	0	0	0	0	0
2	0	1	0	0	0	0	0	0	0
3	0	0	1	0	0	0	0	0	0
4	0	0	0	1	0	0	0	0	0
5	0	0	0	0	1	0	0	0	0
6	0	0	0	0	0	1	0	0	0
7	0	0	0	0	0	0	1	0	0
8	0	0	0	0	0	0	0	1	0
9	0	0	0	0	0	0	0	0	-1
...									
84	0	0	0	0	0	0	0	-1	0
85	0	0	0	0	0	0	0	0	1
86	0	0	0	0	0	1	0	0	0
87	0	0	0	0	0	0	1	0	0
88	0	0	0	0	0	0	0	1	0
89	0	0	0	0	0	0	0	0	1

Making a keys matrix

Although it is possible to make up a keys matrix in Excel and copy it into R, it is easier to use the `make.keys` function.

```
> my.keys <- make.keys(nvars=89,keys.list =list(Pnach=1,PAnx=2,PSoc=3,PImp=4,G=5,
      Nach=c(6,10,14,18),Anx=c(7,11,15,19),Soc=c(8,12,-16,20),Imp=c(-9,13,-17,21)))
> my.keys
```

	Pnach	PAnx	PSoc	PImp	G	Nach	Anx	Soc	Imp
[1,]	1	0	0	0	0	0	0	0	0
[2,]	0	1	0	0	0	0	0	0	0
[3,]	0	0	1	0	0	0	0	0	0
[4,]	0	0	0	1	0	0	0	0	0
[5,]	0	0	0	0	1	0	0	0	0
[6,]	0	0	0	0	0	1	0	0	0
[7,]	0	0	0	0	0	0	1	0	0
[8,]	0	0	0	0	0	0	0	1	0
[9,]	0	0	0	0	0	0	0	0	-1
[10,]	0	0	0	0	0	1	0	0	0
[11,]	0	0	0	0	0	0	1	0	0
[12,]	0	0	0	0	0	0	0	1	0
[13,]	0	0	0	0	0	0	0	0	1
[14,]	0	0	0	0	0	1	0	0	0
[15,]	0	0	0	0	0	0	1	0	0
[16,]	0	0	0	0	0	0	0	-1	0
[17,]	0	0	0	0	0	0	0	0	-1
[18,]	0	0	0	0	0	1	0	0	0
[19,]	0	0	0	0	0	0	1	0	0
[20,]	0	0	0	0	0	0	0	1	0
[21,]	0	0	0	0	0	0	0	0	1
...									

Score the items

We use the `score.items` function.

We first do this just for the items. The `item.scores` is a list of multiple values:

- 1 scores – the actual scores for each subject
- 2 missing – where there any missing values for any subject?
- 3 alpha – coefficient alpha for each scale
- 4 av.r – the average r within each scale
- 5 n.items – how many items in each scale?
- 6 item.cor – the correlation of each item with each scale
- 7 cor – the correlation matrix of the scales
- 8 corrected – the raw correlations of the scales (below the diagonal), the alpha reliabilities (on the diagonal), and the intercorrelations corrected for unreliability (above the diagonal).

Using `score.items`

```
> tem.scores <- score.items(prq.keys[,6:9],prq.items)
> print(item.scores$corrected)
> round(item.scores$corrected,2)
```

```
> print(item.scores$corrected)
      Nach      Anx      Soc      Imp
Nach 0.83928271 0.08193167 0.2755535 -0.22602236
Anx  0.06805717 0.82212135 -0.2545185 0.09498796
Soc  0.23857492 -0.21809815 0.8931604 0.44321814
Imp  -0.19259110 0.08010640 0.3895946 0.86509005
```

```
#rounding to 2 decimal places is nicer
```

```
      Nach  Anx  Soc  Imp
Nach 0.84 0.08 0.28 -0.23
Anx  0.07 0.82 -0.25 0.09
Soc  0.24 -0.22 0.89 0.44
Imp  -0.19 0.08 0.39 0.87
```

Show more of the output

```
> item.scores
```

```
Call: score.items(keys = prq.keys[, 6:9], items = prq.items)
```

```
(Unstandardized) Alpha:
```

```
      Nach Anx Soc Imp
alpha 0.84 0.82 0.89 0.87
```

```
Average item correlation:
```

```
      Nach Anx Soc Imp
average.r 0.2 0.18 0.28 0.23
```

```
Guttman 6* reliability:
```

```
      Nach Anx Soc Imp
Lambda.6 0.93 0.89 0.93 0.92
```

```
Scale intercorrelations corrected for attenuation
```

```
raw correlations below the diagonal, alpha on the diagonal
```

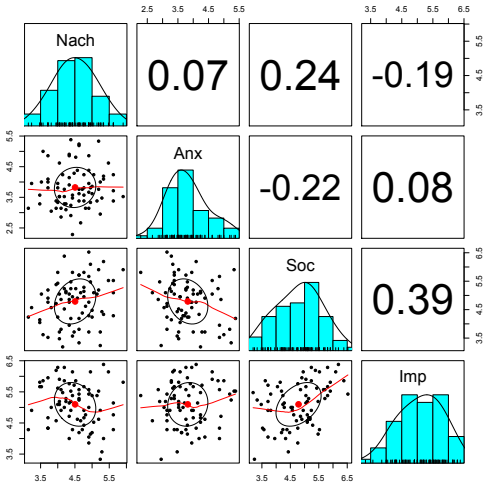
```
corrected correlations above the diagonal:
```

```
      Nach   Anx   Soc   Imp
Nach 0.839 0.082 0.28 -0.226
Anx  0.068 0.822 -0.25 0.095
Soc  0.239 -0.218 0.89 0.443
Imp  -0.193 0.080 0.39 0.865
```

```
In order to see the item by scale loadings and frequency counts of the data
print with the short option = FALSE
```

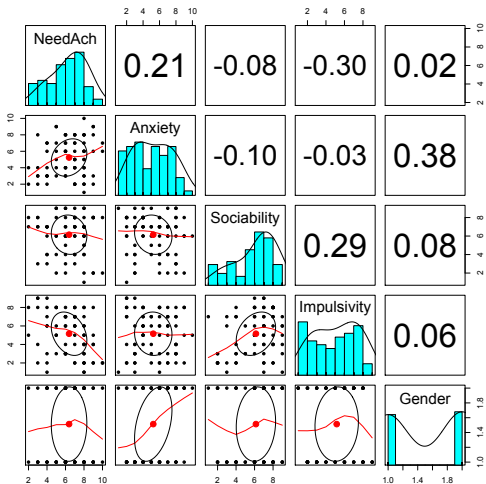

Display the four self report dimensions

`pairs.panels(item.scores$scores)` # note that scores are not shown in output



Show the peer rating structure

```
pairs.panels(prq.data[,3:7])
```



Score the peer ratings and the scales

```

mmtm <- score.items(prq.keys,prq.items)
> mmtm
Call: score.items(keys = prq.keys, items = prq.items)

(Unstandardized) Alpha:
      PNach PANx PSoc PImp G Nach Anx Soc Imp
alpha   1    1    1    1  1 0.84 0.82 0.89 0.87

Average item correlation:
      PNach PANx PSoc PImp G Nach Anx Soc Imp
average.r  NaN  NaN  NaN  NaN NaN  0.2 0.18 0.28 0.23

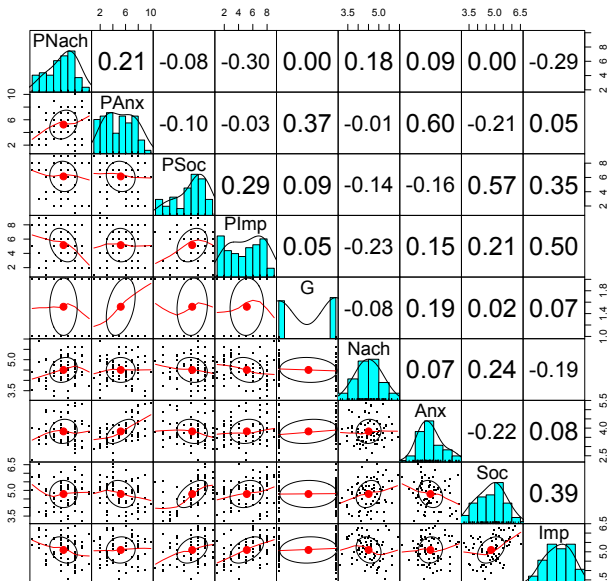
Guttman 6* reliability:
      PNach PANx PSoc PImp G Nach Anx Soc Imp
Lambda.6  0.27 0.19 0.22 0.18 4 0.93 0.89 0.93 0.92

Scale intercorrelations corrected for attenuation
raw correlations below the diagonal, alpha on the diagonal
corrected correlations above the diagonal:
      PNach  PANx  PSoc  PImp    G  Nach  Anx  Soc  Imp
PNach  1.0000  0.207 -0.077 -0.304 -0.0011  0.1993  0.098 -0.0041 -0.311
PANx  0.2068  1.000 -0.102 -0.030  0.3733 -0.0065  0.659 -0.2173  0.059
PSoc  -0.0767 -0.102  1.000  0.293  0.0919 -0.1555 -0.175  0.6014  0.372
PImp  -0.3041 -0.030  0.293  1.000  0.0545 -0.2502  0.161  0.2214  0.535
G      -0.0011  0.373  0.092  0.054  1.0000 -0.0865  0.210  0.0241  0.079
Nach   0.1826 -0.006 -0.142 -0.229 -0.0792  0.8393  0.082  0.2756 -0.226
Anx    0.0888  0.597 -0.159  0.146  0.1901  0.0681  0.822 -0.2545  0.095
Soc   -0.0039 -0.205  0.568  0.209  0.0228  0.2386 -0.218  0.8932  0.443
Imp   -0.2891  0.055  0.346  0.498  0.0736 -0.1926  0.080  0.3896  0.865

```

In order to see the item by scale loadings and frequency counts of the data

Show the MMTM matrix graphically



Factor Analysis

The items analysed were meant to represent four constructs. Given the previous analysis, they probably do. But what if we did not know how many separate dimensions were in the data? Is it possible to find out? Three alternative procedure address this question.

- 1 Principal components analysis
- 2 Factor analysis
- 3 Cluster analysis

All three of these procedures are attempting to approximate the $nvar * nvar$ correlation matrix R with a matrix of lesser rank, one that is $nvar * nf$. That is, can we find a Factor (Component or Cluster) such that

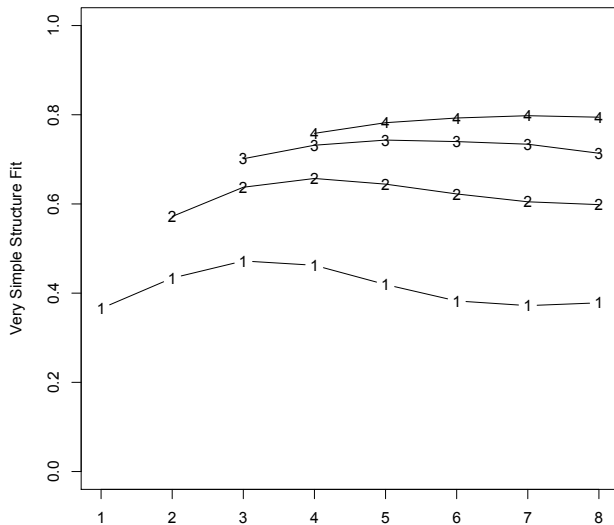
$$R \approx FF' + U^2 \quad (1)$$

Factor analysis of PRQ

- 1 More items than people makes the matrix not invertible
- 2 Can be solved by the fa function using minres option
- 3 How many factors to extract?
 - VSS(prq.items)
 - Use VSS 3 (complexity 1) or 4 (complexity 2)
 - Use MAPS 8
- 4 Theory says 4

VSS of prq

Very Simple Structure



Find a 4 factor as well as a 4 component solution

```
f4 <- fa(prq.items,4)
p4 <- principal(prq.items,4)
summary(f4)
```

```
> summary(f4)
```

Factor analysis with Call: fa(r = prq.items, nfactors = 4)

Test of the hypothesis that 4 factors are sufficient.

The degrees of freedom for the model is 3566 and the objective function was 523.91

The number of observations was 75 with Chi Square = 21393.03 with prob < 0

The root mean square of the residuals (RMSA) is 0.07

The df corrected root mean square of the residuals is 0.1

Tucker Lewis Index of factoring reliability = NaN

RMSEA index = 0.365 and the 90 % confidence intervals are 0.255 0.262

BIC = 5996.86

With factor correlations of

	MR1	MR2	MR4	MR3
MR1	1.00	0.12	0.17	-0.16
MR2	0.12	1.00	-0.03	-0.03
MR4	0.17	-0.03	1.00	0.02
MR3	-0.16	-0.03	0.02	1.00

Also try a cluster analysis

```
> ic <- ICLUST(prq.items,labels=strtrim(prq.labels[,2],20))  
> summary(ic)
```

```
ICLUST (Item Cluster Analysis)Call: ICLUST(r.mat = prq.items, labels = strtrim(prq.labels[, 2], 20))  
ICLUST
```

Purified Alpha:

C83	C82	C85	C81
0.92	0.89	0.87	0.54

Guttman Lambda6*

C83	C82	C85	C81
1	1	1	1

Original Beta:

C83	C82	C85	C81
0.58	0.47	0.48	0.40

Cluster size:

C83	C82	C85	C81
40	25	20	4

Purified scale intercorrelations

reliabilities on diagonal

correlations corrected for attenuation above diagonal:

	C83	C82	C85	C81
C83	0.920	-0.0312	0.1818	0.11
C82	-0.028	0.8922	-0.0036	0.18
C85	0.162	-0.0032	0.8679	0.20
C81	0.074	0.1280	0.1386	0.54

Compare the solutions

```
> factor.congruence(list(f4,p4,ic))
```

	MR1	MR2	MR4	MR3	PC1	PC2	PC4	PC3	C83	C82	C85	C81
MR1	1.00	0.04	0.11	-0.11	0.99	0.14	0.22	-0.16	-0.84	0.22	-0.39	0.30
MR2	0.04	1.00	-0.05	0.00	0.08	0.99	-0.10	0.01	0.02	0.98	0.05	0.25
MR4	0.11	-0.05	1.00	0.02	0.18	-0.03	0.99	0.06	-0.63	-0.06	0.17	-0.54
MR3	-0.11	0.00	0.02	1.00	-0.19	-0.05	-0.01	1.00	0.19	-0.08	0.92	0.45
PC1	0.99	0.08	0.18	-0.19	1.00	0.17	0.28	-0.24	-0.87	0.25	-0.44	0.22
PC2	0.14	0.99	-0.03	-0.05	0.17	1.00	-0.07	-0.04	-0.07	0.99	-0.02	0.25
PC4	0.22	-0.10	0.99	-0.01	0.28	-0.07	1.00	0.02	-0.70	-0.08	0.10	-0.52
PC3	-0.16	0.01	0.06	1.00	-0.24	-0.04	0.02	1.00	0.21	-0.07	0.95	0.41
C83	-0.84	0.02	-0.63	0.19	-0.87	-0.07	-0.70	0.21	1.00	-0.13	0.32	0.13
C82	0.22	0.98	-0.06	-0.08	0.25	0.99	-0.08	-0.07	-0.13	1.00	-0.09	0.28
C85	-0.39	0.05	0.17	0.92	-0.44	-0.02	0.10	0.95	0.32	-0.09	1.00	0.19
C81	0.30	0.25	-0.54	0.45	0.22	0.25	-0.52	0.41	0.13	0.28	0.19	1.00

Combine the factor scores with the empirical scores

```
> scores.df <- data.frame(f4$scores,item.scores$scores)  
> lowerCor(scores.df)
```

	X1	X2	X3	X4	Nach	Anx	Soc	Imp
X1	1.00							
X2	0.37	1.00						
X3	0.56	-0.05	1.00					
X4	-0.56	-0.20	-0.10	1.00				
Nach	0.33	0.95	-0.04	-0.13	1.00			
Anx	-0.25	0.04	0.18	0.89	0.07	1.00		
Soc	0.95	0.29	0.51	-0.50	0.24	-0.22	1.00	
Imp	0.46	-0.24	0.96	-0.12	-0.19	0.08	0.39	1.00

Compare original, factors and clusters

```
> fkeys <- factor2cluster(f4)
> ckeys <- cluster2keys(ic)
> all.keys <- cbind(prq.keys,fkeys,ckeys)
> all.scores <- score.items(all.keys,prq.items)
> lowerMat(all.scores$cor)
```

	PNach	PAnx	PSoc	PImp	G	Nach	Anx	Soc	Imp	MR1	MR2	MR4	MR3
PNach	1.00												
PAnx	0.21	1.00											
PSoc	-0.08	-0.10	1.00										
PImp	-0.30	-0.03	0.29	1.00									
G	0.00	0.37	0.09	0.05	1.00								
Nach	0.18	-0.01	-0.14	-0.23	-0.08	1.00							
Anx	0.09	0.60	-0.16	0.15	0.19	0.07	1.00						
Soc	0.00	-0.21	0.57	0.21	0.02	0.24	-0.22	1.00					
Imp	-0.29	0.05	0.35	0.50	0.07	-0.19	0.08	0.39	1.00				
MR1	-0.03	-0.24	0.65	0.25	-0.01	0.23	-0.22	0.98	0.37	1.00			
MR2	0.21	-0.02	-0.26	-0.33	-0.05	0.94	0.03	0.14	-0.30	0.12	1.00		
MR4	-0.39	0.11	0.21	0.56	0.07	-0.09	0.22	0.30	0.94	0.27	-0.18	1.00	
MR3	0.13	0.66	-0.18	0.07	0.21	-0.01	0.93	-0.30	-0.06	-0.28	-0.05	0.03	1.00
C83	0.26	0.14	-0.60	-0.49	-0.03	-0.04	0.07	-0.86	-0.77	-0.85	0.08	-0.71	0.19
C82	0.18	-0.04	-0.22	-0.28	-0.04	0.94	0.05	0.19	-0.24	0.16	0.99	-0.13	-0.04
C85	0.16	0.69	-0.19	0.08	0.23	0.02	0.97	-0.30	0.04	-0.31	0.00	0.17	0.94
C81	0.01	0.01	0.02	0.03	-0.10	0.24	0.22	0.05	-0.31	0.14	0.13	-0.28	0.36
	C83	C82	C85	C81									
C83	1.00												
C82	0.01	1.00											
C85	0.17	0.00	1.00										
C81	0.07	0.11	0.16	1.00									

ICLUST output

```
> print(ic,sort=TRUE,labels=prq.labels)
```

ICLUST (Item Cluster Analysis)

Call: ICLUST(r.mat = prq.items, labels = strtrim(prq.labels[, 2], 20))

Purified Alpha:

C83	C82	C85	C81
0.92	0.89	0.87	0.54

G6* reliability:

C83	C82	C85	C81
0.84	0.97	0.96	1.00

Original Beta:

C83	C82	C85	C81
0.58	0.47	0.48	0.40

Cluster size:

C83	C82	C85	C81
40	25	20	4

Cluster 1

Item by Cluster Structure matrix: Sorted by loading

item	content	cluster	C83	C82	C85	C81
q35	40 I have a large soci	1	-0.71		-0.31	
q83	88 I am a very sociable	1	-0.64	0.37		
q67	72 I am always willing	1	-0.64			
q23	28 I make friends easi	1	-0.63			
q51	56 People are more lik	1	0.63			
q11	16 I tend to avoid soc	1	0.63		0.32	
q16	21 I tend to make deci	1	-0.61			
q43	48 I am happier when I	1	-0.60	0.43		
q39	44 Id rather spend tim	1	-0.58			
q24	29 I often change my p	1	-0.58			-0.30
Sociability	3 Sociability	1	-0.57			
q84	89 I am an impulsive pe	1	-0.56			
q56	61 I often and activel	1	-0.55			
q80	85 I often say the fir	1	-0.53			
q59	64 I prefer large crow	1	-0.52			
q44	49 I often regret deci	1	-0.52			
q19	24 I am good at mainta	1	-0.52			
q3	8 I like to meet new	1	-0.51			
q31	36 I tend to talk a lo	1	-0.50		-0.34	
q76	81 I sometimes look ba	1	-0.50			
q8	13 I say things that I	1	-0.49			
q47	52 I enjoy being alone	1	0.48			
q52	57 I often get sidetra	1	-0.48			-0.34
q40	45 I act on sudden urg	1	-0.47			
Impulsivity	4 Impulsivity	1	-0.47			
q32	37 I indulge in my des	1	-0.46			
q69	74 I tend to procrasti	1	-0.46	-0.31		
q55	60 Ill spend time talk	1	-0.44			
q68	73 I always think befo	1	0.44			
q63	68 A good night for me	1	0.43		0.31	

Cluster 2

NeedAch	1	NeedAch	1	
q81	86	I believe that if so	2	0.74
q17	22	I have high standar	2	0.72
q33	38	I find myself needi	2	0.69
q4	9	I am thoughtful and	2	0.63
q41	46	I always make sure	2	0.63
q25	30	If I fail, I keep t	2	0.62
q13	18	I like to go the ex	2	0.61
q1	6	I love to seek out	2	0.60
q77	82	I always see projec	2	0.60
q61	66	I experience great	2	0.60
q49	54	The joy of success	2	0.59
q60	65	I stay on task unti	2	0.58
q45	50	I prefer challengin	2	0.55
q73	78	I set long term and	2	0.51
q12	17	I weigh all the opt	2	0.50
q57	62	I always reach the	2	0.49
q78	83	I tend to back away	2	-0.47
q37	42	I get bored if a ta	2	0.46
q58	63	I prefer to work in	2	0.46
q27	32	I tend to enjoy sma	2	0.41
q5	10	Personal satisfacti	2	0.35
q21	26	I am a perfectionis	2	0.35
q65	70	I tend to have trou	2	-0.32
q75	80	I work better when	2	
q29	34	I seek the enjoymen	2	

Cluster 3

q42	47	Even trivial proble	3		0.71	
Anxiety	2	Anxiety	3		0.68	
q6	11	I dont handle stress	3		0.68	
q50	55	Even in non stressf	3		0.64	0.34
q2	7	I get nervous very e	3		0.62	
q18	23	I rarely feel tense	3		-0.61	
q34	39	I have a hard time f	3		0.59	
q10	15	I am easily bothered	3		0.57	
q26	31	I often feel anxious	3		0.56	
q22	27	I feel stressed when	3		0.55	
q30	35	I often feel tense,	3		0.54	
q62	67	A small unpleasant	3		0.54	
q66	71	I worry about what	3		0.52	
q70	75	I bounce back quick	3	-0.33	0.35	-0.47
q74	79	I tend to dwell on	3		0.46	
q54	59	I feel tension in m	3		0.46	
q38	43	I often have unwanted	3		0.45	
q14	19	Measures of skill or	3		0.42	
Gender	5	Gender	3		0.31	

Cluster 4

q72	77 I always stick to p	4		0.72
q64	69 I dislike changing	4		0.69
q82	87 I am more emotional	4		0.63
q9	14 I am a good multi t	4	0.31	0.55

Show the items for the factors

```
> rownames(f4$loadings) <- strtrim(prq.labels[,2],20)
> print(f4,sort=TRUE)
```

Factor Analysis using method = minres

Call: fa(r = prq.items, nfactors = 4)

Standardized loadings (pattern matrix) based upon correlation matrix

	item	MR1	MR2	MR4	MR3	h2	u2
	I have a large soci	40	0.79	0.02	0.06	-0.13	0.696 0.30
	I like to meet new	8	0.78	0.11	-0.16	0.15	0.607 0.39
	I tend to avoid soc	16	-0.77	-0.08	0.02	0.09	0.644 0.36
	I am a very sociable	88	0.75	0.25	0.04	-0.13	0.725 0.28
	Id rather spend tim	44	0.74	-0.03	-0.01	0.14	0.531 0.47
	I make friends easi	28	0.70	0.14	0.10	-0.09	0.596 0.40
	Sociability	3	0.64	-0.32	0.04	-0.04	0.482 0.52
	I am happier when I	48	0.62	0.32	0.15	0.08	0.575 0.42
	People are more lik	56	-0.61	-0.09	-0.15	0.13	0.487 0.51
	I often and activel	61	0.57	0.01	0.18	0.23	0.400 0.60
	I can easily start	12	0.55	0.10	-0.10	0.00	0.319 0.68
	I prefer large crow	64	0.55	-0.20	0.06	-0.10	0.352 0.65
	I am always willing	72	0.54	0.08	0.27	-0.08	0.455 0.54
	I am good at mainta	24	0.54	0.23	0.15	-0.12	0.454 0.55
	When given the choi	84	-0.50	0.35	0.10	0.01	0.327 0.67
	I enjoy being alone	52	-0.47	0.01	-0.13	0.09	0.278 0.72
	I dont understand h	76	0.43	-0.28	0.02	-0.13	0.272 0.73
	I tend to lead the	20	0.41	-0.06	-0.07	-0.25	0.256 0.74
	A good night for me	68	-0.39	-0.06	-0.06	0.27	0.283 0.72
	I am a good multi t	14	0.31	0.28	-0.22	-0.03	0.225 0.78

Factor 2

I believe that if so	86	0.12	0.75	0.14	0.03	0.619	0.38
I have high standar	22	0.13	0.67	-0.16	0.14	0.516	0.48
I find myself needi	38	0.17	0.65	0.08	0.07	0.487	0.51
I like to go the ex	18	0.12	0.65	-0.10	0.04	0.457	0.54
I stay on task unti	65	0.03	0.61	-0.07	0.07	0.390	0.61
I always make sure	46	0.10	0.61	-0.04	-0.03	0.403	0.60
I always see projec	82	0.07	0.61	-0.07	0.13	0.404	0.60
I am thoughtful and	9	0.03	0.60	-0.36	-0.11	0.517	0.48
The joy of success	54	0.20	0.57	0.06	0.06	0.393	0.61
I set long term and	78	-0.19	0.56	0.17	-0.02	0.336	0.66
I love to seek out	6	0.23	0.56	0.08	-0.14	0.436	0.56
If I fail, I keep t	30	0.27	0.56	-0.06	-0.04	0.424	0.58
I experience great	66	0.04	0.56	-0.07	-0.05	0.327	0.67
I tend to back away	83	0.20	-0.52	-0.03	0.39	0.427	0.57
I prefer challengin	50	0.09	0.48	0.00	-0.12	0.268	0.73
I tend to enjoy sma	32	-0.30	0.46	0.16	0.02	0.278	0.72
I prefer to work in	63	-0.03	0.46	0.22	0.09	0.255	0.74
I tend to procrasti	74	0.29	-0.45	0.27	0.13	0.372	0.63
I always reach the	62	0.24	0.44	-0.15	-0.04	0.293	0.71
Personal satisfacti	10	-0.10	0.43	0.16	0.21	0.252	0.75
I tend to have trou	70	0.11	-0.43	0.10	0.33	0.306	0.69
I weigh all the opt	17	-0.16	0.42	-0.21	0.05	0.253	0.75
I get bored if a ta	42	0.09	0.38	0.22	-0.15	0.239	0.76
I am a perfectionis	26	-0.10	0.35	-0.10	0.28	0.222	0.78
I only work as hard	58	0.12	-0.29	0.20	0.18	0.172	0.83
I seek the enjoymen	34	0.15	0.25	0.13	-0.14	0.140	0.86
I work better when	80	0.15	-0.23	0.05	-0.03	0.077	0.92
When working on a n	41	-0.12	0.23	-0.23	-0.02	0.129	0.87

Factor 3

I act on sudden urg	45	-0.03	0.07	0.72	-0.13	0.528	0.47
I often change my p	29	0.12	0.05	0.72	0.00	0.559	0.44
I often get sidetra	57	0.03	-0.14	0.69	0.12	0.521	0.48
I say things that I	13	0.09	-0.16	0.61	0.08	0.426	0.57
I often have unwanted	43	-0.34	0.08	0.60	0.17	0.458	0.54
I am an impulsive pe	89	0.22	0.03	0.59	-0.08	0.452	0.55
I dislike planning	33	0.02	-0.17	0.57	0.09	0.379	0.62
I often regret deci	49	0.20	-0.14	0.54	0.28	0.457	0.54
I indulge in my des	37	0.16	0.14	0.54	0.11	0.373	0.63
I always stick to p	77	0.28	0.07	-0.53	0.24	0.352	0.65
I always think befo	73	-0.10	0.26	-0.51	0.05	0.360	0.64
I sometimes look ba	81	0.20	0.08	0.51	0.07	0.341	0.66
I tend to act on my	53	0.04	0.27	0.48	-0.19	0.349	0.65
I tend to make deci	21	0.33	0.14	0.48	-0.23	0.494	0.51
I plan my activitie	25	-0.05	0.35	-0.45	0.18	0.368	0.63
I often have diffic	51	-0.33	0.17	0.41	-0.01	0.239	0.76
Impulsivity	4	0.26	-0.31	0.39	0.14	0.355	0.65
I often say the fir	85	0.30	-0.21	0.36	-0.09	0.307	0.69
I feel tension in m	59	-0.26	0.21	0.33	0.24	0.252	0.75
NeedAch	1	-0.04	0.23	-0.25	0.11	0.135	0.86

Factor 4

I dont handle stress	11	-0.12	-0.11	0.03	0.70	0.556	0.44
Even trivial proble	47	-0.15	0.01	0.04	0.67	0.506	0.49
Even in non stressf	55	0.01	0.01	-0.01	0.66	0.437	0.56
I worry about what	71	0.23	-0.09	-0.08	0.64	0.421	0.58
I get nervous very e	7	-0.16	-0.18	-0.01	0.62	0.490	0.51
I am easily bothered	15	0.05	-0.04	0.01	0.60	0.361	0.64
A small unpleasant	67	0.02	0.20	-0.04	0.60	0.396	0.60
I feel stressed when	27	0.00	0.25	-0.19	0.59	0.431	0.57
Anxiety	2	-0.21	0.07	0.21	0.57	0.436	0.56
I have a hard time f	39	-0.08	0.24	-0.03	0.55	0.371	0.63
I often feel anxious	31	-0.12	0.25	0.26	0.52	0.404	0.60
I am more emotional	87	0.34	-0.10	-0.10	0.49	0.303	0.70
I often feel tense,	35	-0.10	-0.01	0.25	0.47	0.304	0.70
I dislike changing	69	0.12	-0.03	-0.44	0.46	0.375	0.63
I rarely feel tense	23	0.25	-0.28	-0.29	-0.45	0.415	0.58
I tend to talk a lo	36	0.33	0.14	0.27	-0.45	0.485	0.52
I tend to dwell on	79	0.17	0.29	0.14	0.44	0.314	0.69
I bounce back quick	75	0.23	0.30	0.16	-0.43	0.418	0.58
Ill spend time talk	60	0.35	-0.15	0.29	0.38	0.363	0.64
Measures of skill or	19	-0.22	-0.04	0.13	0.29	0.162	0.84
Gender	5	-0.07	0.00	0.16	0.19	0.069	0.93