

An introduction to R: a short course Part II

Multivariate analysis

The zelig website

ZELIG: EVERYONE'S STATISTICAL SOFTWARE

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Version:3.4-5 ([Download](#)) ([What's new](#))



Zelig is named after a Woody Allen movie about a man who had the strange ability to become the physical and psychological reflection of anyone he met and thus to fit perfectly in any situation.

Zelig is a single, easy-to-use program that can estimate, help interpret, and present the results of a large range of statistical methods. It literally *is* "everyone's statistical software" because Zelig uses (R) code from many researchers. We also hope it will *become* "everyone's statistical software" for applications, and we have designed it so that anyone can use it or add their methods to it. Zelig comes with detailed, self-contained documentation that minimizes startup costs for Zelig and R, automates graphics and summaries for all models, and, with only three simple commands required, generally makes the power of R accessible for all users. Zelig also works well for teaching, and is designed so that scholars can use the same program with students that they use for their research.

Zelig adds considerable infrastructure to improve the use of existing methods. It generalizes the program [Clarify](#) (for Stata), which translates hard-to-interpret coefficients into quantities of interest; combines multiply imputed data sets (such as output from [Amelia](#)) to deal with missing data; automates bootstrapping for all models; uses sophisticated nonparametric matching commands which improve parametric procedures (via [MatchIt](#)); allows one-line commands to run analyses in all designated strata; automates the creation of replication data files so that you (or, if you wish, anyone else) can replicate the results of your analyses (hence satisfying the [replication standard](#)); makes it easy to evaluate counterfactuals (via [WhatIf](#)); and allows conditional population and superpopulation inferences. Zelig includes many specific methods, based on likelihood, frequentist, Bayesian, robust Bayesian, and nonparametric theories of inference.

- [Documentation](#) and [Installation](#)
- [Installation](#), [Frequently Asked Questions](#)
- Please send **All** questions, bugs and requests: Zelig Mailing List, [\[Un\]Subscribe](#), or [Browse/Search Archives](#)
- A paper that describes the advances underlying Zelig software: Kosuke Imai, Gary King, and Olivia Lau. "**Toward A Common Framework for Statistical Analysis and Development**" *Journal of Computational and Graphical Statistics*, Vol. 17, No. 4 (December), pp. 892-913 (**Abstract: [HTML](#) | Paper: [PDF](#)**)
- [Slides used to introduce Zelig](#)
- [Be notified of Zelig changes](#)

<http://gking.harvard.edu/zelig/>

Multivariate analysis

- I. Dimension reduction through factor analysis, principal components analysis, cluster analysis, and multidimensional scaling
- II. Multiple measures of reliability
- III. Practical use of R for scoring inventories

Dimension Reduction

- I. The problem: How best to summarize and think about many variables some of which are moderately correlated
- II. The solutions: rank reduction through FA, PCA, CA, MDS
- III. Examples will be tests and then items

The Thurstone data set

```
> data(bifactor)
> colnames(Thurstone) <- c
("Sentences", "Vocab", "S.comp", "F.letter", "4.letter", "Suffix",
 "Series", "Pedi", "letters")

> round(Thurstone, 2)
```

	Sentences	Vocab	S.comp	F.letter	4.letter	Suffix	Series	Pedi	letters
Sentences	1.00	0.83	0.78	0.44	0.43	0.45	0.45	0.54	0.38
Vocabulary	0.83	1.00	0.78	0.49	0.46	0.49	0.43	0.54	0.36
Sent.Completion	0.78	0.78	1.00	0.46	0.42	0.44	0.40	0.53	0.36
First.Letters	0.44	0.49	0.46	1.00	0.67	0.59	0.38	0.35	0.42
4.Letter.Words	0.43	0.46	0.42	0.67	1.00	0.54	0.40	0.37	0.45
Suffixes	0.45	0.49	0.44	0.59	0.54	1.00	0.29	0.32	0.32
Letter.Series	0.45	0.43	0.40	0.38	0.40	0.29	1.00	0.56	0.60
Pedigrees	0.54	0.54	0.53	0.35	0.37	0.32	0.56	1.00	0.45
Letter.Group	0.38	0.36	0.36	0.42	0.45	0.32	0.60	0.45	1.00

How many dimensions

I. Chi Square test

II. Scree plot

III. Parallel analysis of random data

IV. Minimum Average Partial correlation

V. Very Simple Structure

VI. Do not use eigen value > 1 rule!

```

> factanal(covmat=Thurstone,factors=3,n.obs =213)
Call:
factanal(factors = 3, covmat = Thurstone, n.obs = 213)
Uniquenesses:
Sentences          Vocab          S.comp    F.letter    4.letter    Suffix      Series
Pedi  letters
    0.175    0.165    0.268    0.268    0.372    0.504    0.282
0.496    0.473
Loadings:
          Factor1 Factor2 Factor3
Sentences    0.834  0.244  0.264
Vocabulary   0.827  0.318  0.223
Sent.Completion 0.775  0.284  0.227
First.Letters 0.228  0.792  0.230
4.Letter.Words 0.213  0.706  0.291
Suffixes     0.314  0.616  0.134
Letter.Series 0.232  0.179  0.795
Pedigrees    0.446  0.166  0.527
Letter.Group 0.154  0.311  0.638
          Factor1 Factor2 Factor3
SS loadings    2.454  1.902  1.642
Proportion Var  0.273  0.211  0.182
Cumulative Var  0.273  0.484  0.666
Test of the hypothesis that 3 factors are sufficient.
The chi square statistic is 2.82 on 12 degrees of freedom.
The p-value is 0.997

```

MLE factor analysis: factanal

Call:

```
factanal(factors = 2, covmat = Thurstone, n.obs = 213)
```

Uniquenesses:

Sentences	Vocab	S.comp	F.letter	4.letter	Suffix	Series
Pedi	letters					
0.168	0.178	0.269	0.322	0.344	0.537	0.680
0.612	0.677					

Loadings:

	Factor1	Factor2
Sentences	0.866	0.287
Vocabulary	0.839	0.343
Sent.Completion	0.795	0.314
First.Letters	0.255	0.783
4.Letter.Words	0.235	0.775
Suffixes	0.317	0.602
Letter.Series	0.372	0.426
Pedigrees	0.524	0.336
Letter.Group	0.269	0.501

	Factor1	Factor2
SS loadings	2.793	2.420
Proportion Var	0.310	0.269
Cumulative Var	0.310	0.579

Test of the hypothesis that 2 factors are sufficient.

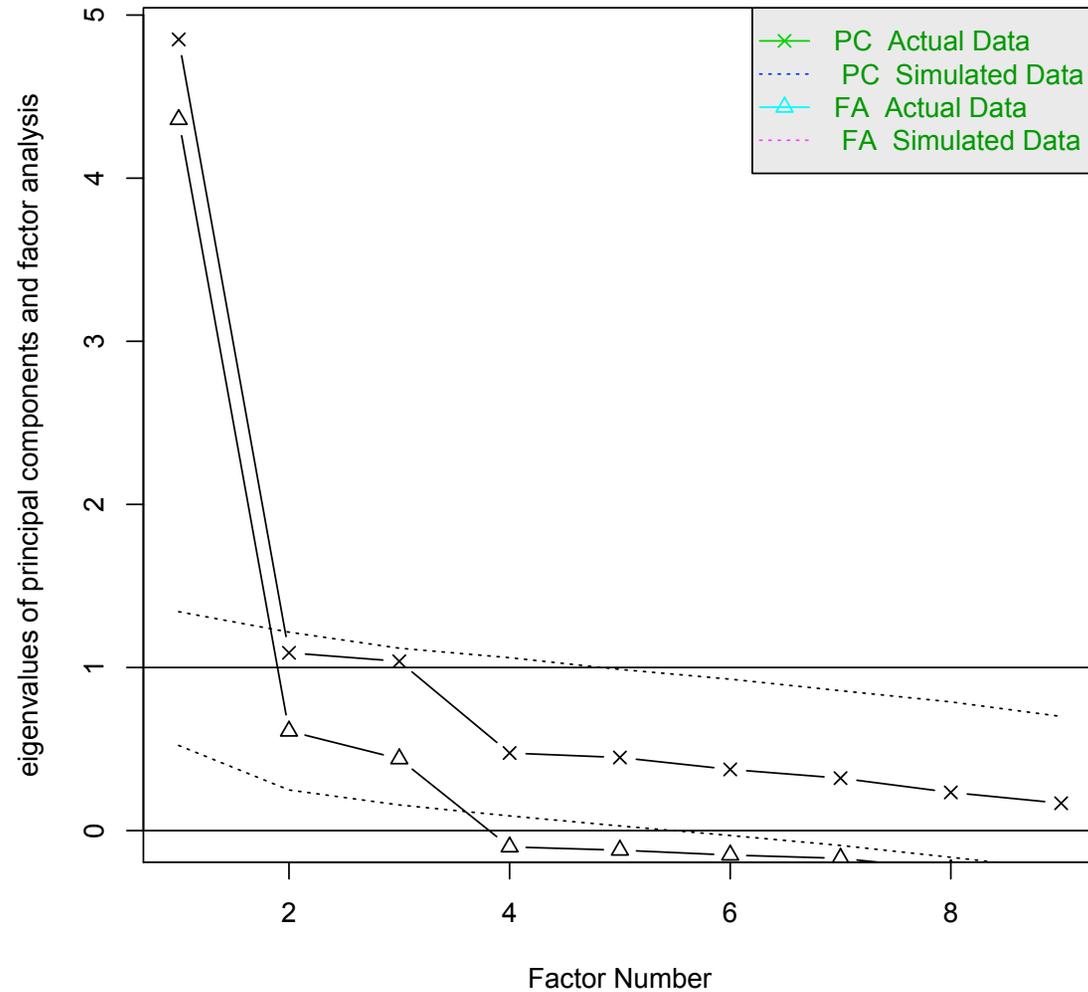
The chi square statistic is 82.84 on 19 degrees of freedom.

The p-value is 5.99e-10

chi square
rejects a 2
factor
solution

Parallel analysis

Parallel Analysis Scree Plots



Very Simple Structure and MAP

```
> vss <- VSS(Thurstone,n.obs=213,SMC=FALSE)
> vss
```

Very Simple Structure

```
Call: VSS(x = Thurstone, n.obs = 213, SMC = FALSE)
```

```
VSS complexity 1 achieves a maximum of 0.88 with 1 factors
```

```
VSS complexity 2 achieves a maximum of 0.92 with 2 factors
```

```
The Velicer MAP criterion achieves a minimum of 1 with 3 factors
```

Velicer MAP

```
[1] 0.07 0.07 0.07 0.11 0.20 0.31 0.59 1.00
```

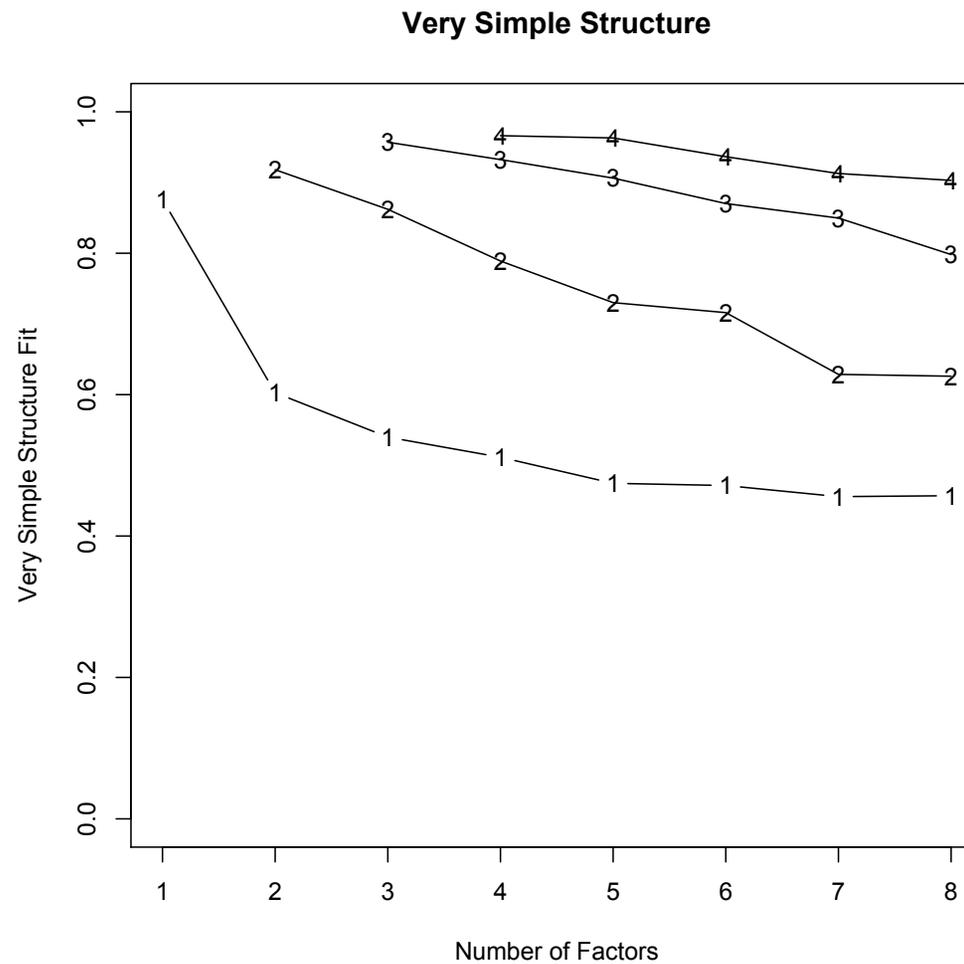
Very Simple Structure Complexity 1

```
[1] 0.88 0.60 0.54 0.51 0.47 0.47 0.46 0.46
```

Very Simple Structure Complexity 2

```
[1] 0.00 0.92 0.86 0.79 0.73 0.72 0.63 0.63
```

Very Simple Structure



Principal Axis FA

```
> pa3 <- factor.pa(Thurstone,nfactors=3,n.obs=213)
> pa3
```

	V	PA1	PA2	PA3
Sentences	1	0.83		
Vocab	2	0.83	0.32	
S.comp	3	0.78		
F.letter	4		0.79	
4.letter	5		0.71	
Suffix	6	0.31	0.62	
Series	7			0.79
Pedi	8	0.45		0.53
letters	9		0.31	0.64

	PA1	PA2	PA3
SS loadings	2.46	1.91	1.64
Proportion Var	0.27	0.21	0.18
Cumulative Var	0.27	0.49	0.67

Test of the hypothesis that 3 factors are sufficient.

The degrees of freedom for the model is 12 and the fit was 0.01
The number of observations was 213 with Chi Square = 2.97 with
prob < 1

Factor analysis options

I. factanal (MLE) is part of core R

II. fa is part of psych

A. minres (ols) (default)

B. weighted least squares

C. generalized weighted least squares

D. maximum likelihood

E. principal axis

Principal Components

```
> pc3
```

	V	PC1	PC2	PC3
Sentences	1	0.863		
Vocabulary	2	0.854	0.31	
Sent.Completion	3	0.849		
First.Letters	4		0.82	
4.Letter.Words	5		0.79	0.301
Suffixes	6	0.314	0.77	
Letter.Series	7			0.834
Pedigrees	8	0.534		0.613
Letter.Group	9		0.31	0.805

	PC1	PC2	PC3
SS loadings	2.73	2.25	1.99
Proportion Var	0.30	0.25	0.22
Cumulative Var	0.30	0.55	0.78

Test of the hypothesis that 3 factors are sufficient.

The degrees of freedom for the model is 12 and the fit was 0.62
The number of observations was 213 with Chi Square = 127.9 with
prob < 1.6e-21

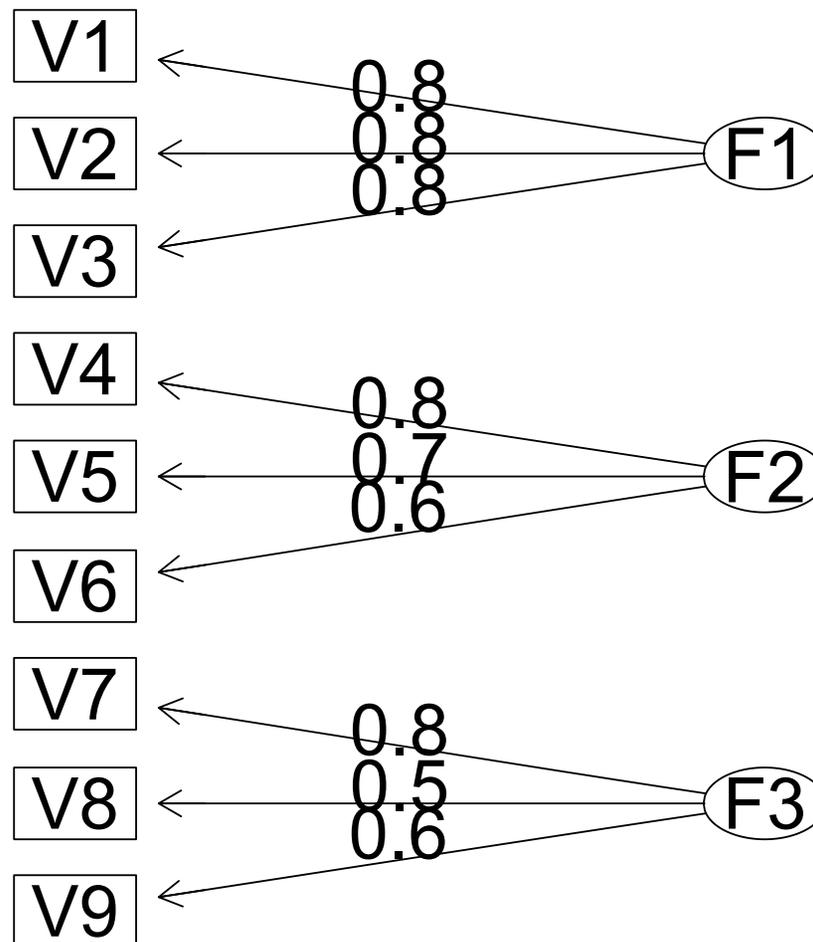
Comparing solutions: factor congruence

```
> factor.congruence(list(f3,pa3,pc3))
```

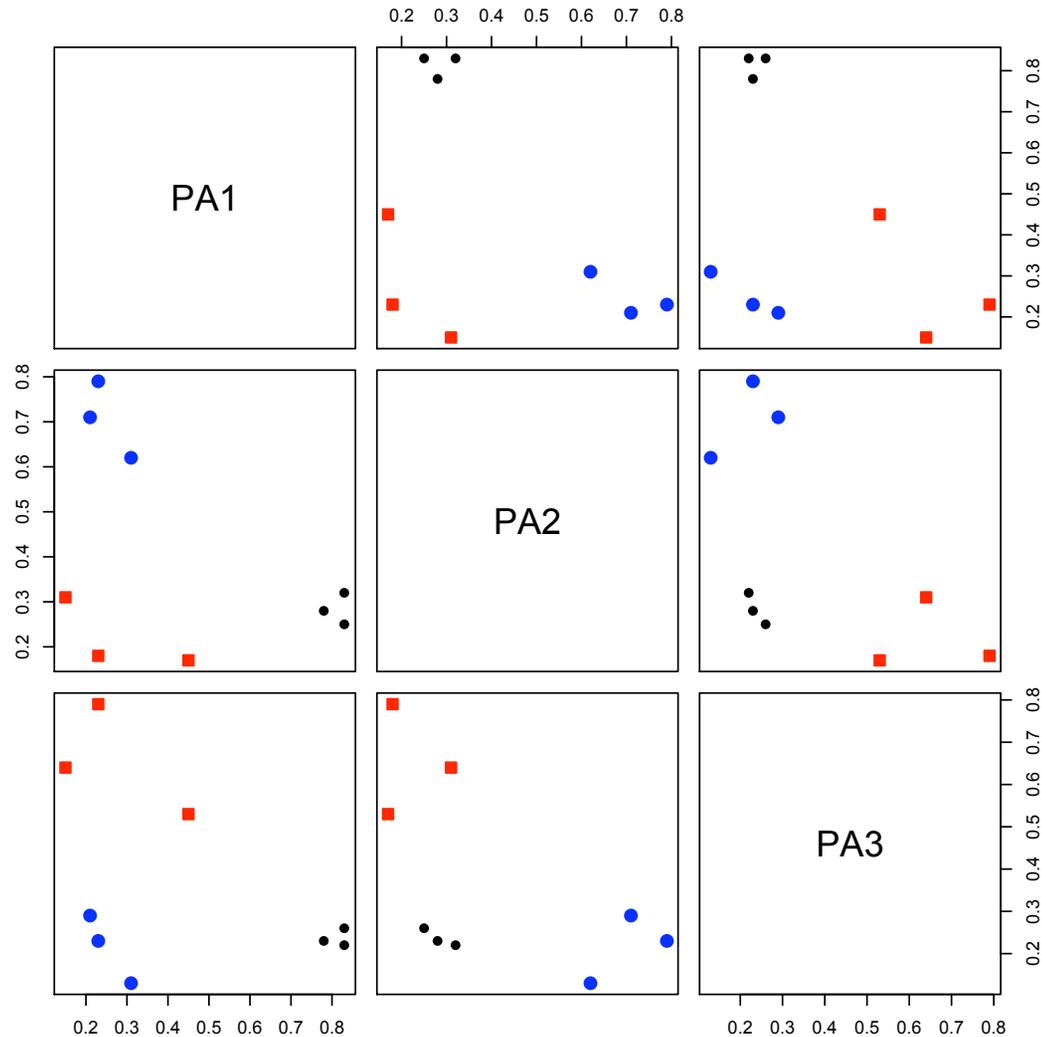
	Factor1	Factor2	Factor3	PA1	PA2	PA3	PC1	PC2	PC3
Factor1	1.00	0.64	0.62	1.00	0.64	0.62	1.00	0.59	0.55
Factor2	0.64	1.00	0.62	0.63	1.00	0.62	0.61	0.99	0.57
Factor3	0.62	0.62	1.00	0.62	0.62	1.00	0.61	0.56	0.99
PA1	1.00	0.63	0.62	1.00	0.64	0.62	1.00	0.59	0.55
PA2	0.64	1.00	0.62	0.64	1.00	0.62	0.61	0.99	0.57
PA3	0.62	0.62	1.00	0.62	0.62	1.00	0.61	0.56	0.99
PC1	1.00	0.61	0.61	1.00	0.61	0.61	1.00	0.56	0.54
PC2	0.59	0.99	0.56	0.59	0.99	0.56	0.56	1.00	0.50
PC3	0.55	0.57	0.99	0.55	0.57	0.99	0.54	0.50	1.00

A misleading graph

Factor Analysis



Plot the loadings: shows some cross loadings



plot(pa3)

Rotations and transformations

I. Orthogonal rotations

A. Varimax, Quartimax

II. Oblique transformations

A. Promax, Oblimin, Quartimin,
biquartimin, targeted, cluster, ...

```

> fa3o <- fa(Thurstone,3,rotate="oblimin")
Loading required package: GPArotation
> fa3o
Factor Analysis using method = minres
Call: fa(r = Thurstone, nfactors = 3, rotate = "oblimin")
      item  MR1  MR2  MR3  h2  u2
Sentences      1  0.90          0.82 0.18
Vocabulary     2  0.89          0.84 0.16
Sent.Completion 3  0.84          0.74 0.26
First.Letters  4          0.85  0.73 0.27
4.Letter.Words 5          0.75  0.63 0.37
Suffixes        6          0.63  0.50 0.50
Letter.Series   7          0.84  0.72 0.28
Pedigrees      8  0.38          0.47 0.50 0.50
Letter.Group    9          0.63  0.52 0.48

      MR1  MR2  MR3
SS loadings  2.64 1.87 1.49
Proportion Var 0.29 0.21 0.17
Cumulative Var 0.29 0.50 0.67
With factor correlations of
      MR1  MR2  MR3
MR1  1.00 0.59 0.53
MR2  0.59 1.00 0.52
MR3  0.53 0.52 1.00
Test of the hypothesis that 3 factors are sufficient.
The degrees of freedom for the model is 12 and the objective function

```

Oblique solution Oblimin

```
> print(fa3o,cut=0)
Factor Analysis using method = minres
Call: fa(r = Thurstone, nfactors = 3, rotate = "oblimin")
```

	item	MR1	MR2	MR3	h2	u2
Sentences	1	0.90	-0.04	0.04	0.82	0.18
Vocabulary	2	0.89	0.06	-0.03	0.84	0.16
Sent.Completion	3	0.84	0.04	0.00	0.74	0.26
First.Letters	4	0.00	0.85	0.00	0.73	0.27
4.Letter.Words	5	-0.02	0.75	0.10	0.63	0.37
Suffixes	6	0.18	0.63	-0.08	0.50	0.50
Letter.Series	7	0.03	-0.01	0.84	0.72	0.28
Pedigrees	8	0.38	-0.05	0.47	0.50	0.50
Letter.Group	9	-0.06	0.21	0.63	0.52	0.48

	MR1	MR2	MR3
SS loadings	2.64	1.87	1.49
Proportion Var	0.29	0.21	0.17
Cumulative Var	0.29	0.50	0.67

With factor correlations of

	MR1	MR2	MR3
MR1	1.00	0.59	0.53
MR2	0.59	1.00	0.52
MR3	0.53	0.52	1.00

Show all values!

```

> pa3p <- Promax(pa3)
> pa3p

```

	V	PA1	PA2	PA3
Sentences	1	0.911		
Vocab	2	0.904		
S.comp	3	0.848		
F.letter	4		0.869	
4.letter	5		0.759	
Suffix	6		0.650	
Series	7			0.8865
Pedi	8	0.350		0.4969
letters	9			0.6800

	PA1	PA2	PA3
SS loadings	2.54	1.80	1.52
Proportion Var	0.28	0.20	0.17
Cumulative Var	0.28	0.48	0.65

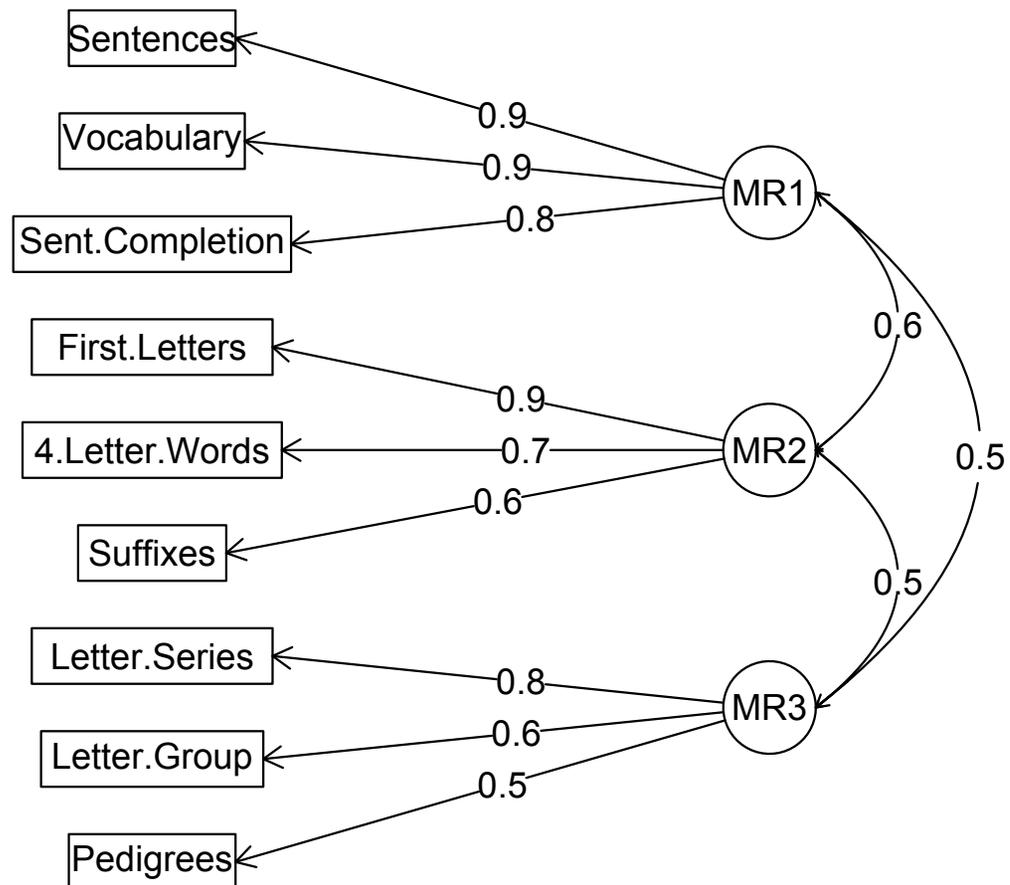
With factor correlations of

	PA1	PA2	PA3
PA1	1.00	0.61	0.61
PA2	0.61	1.00	0.58
PA3	0.61	0.58	1.00

Oblique: Promax

A more accurate graphic

3 oblique factors from Thurstone

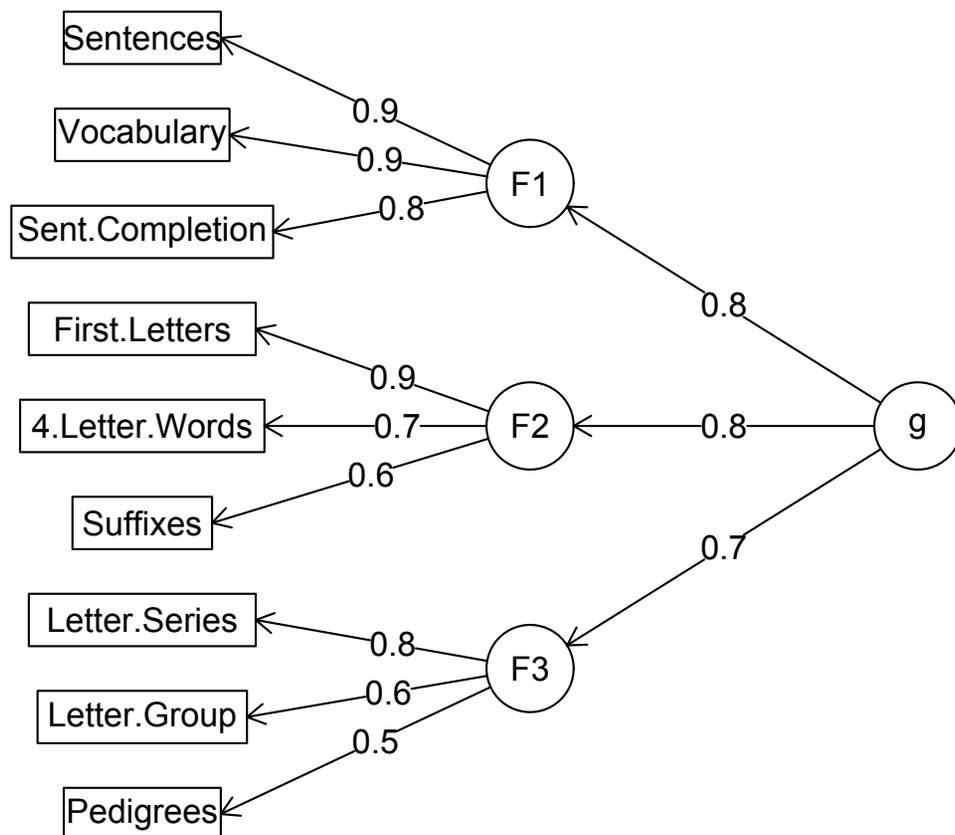
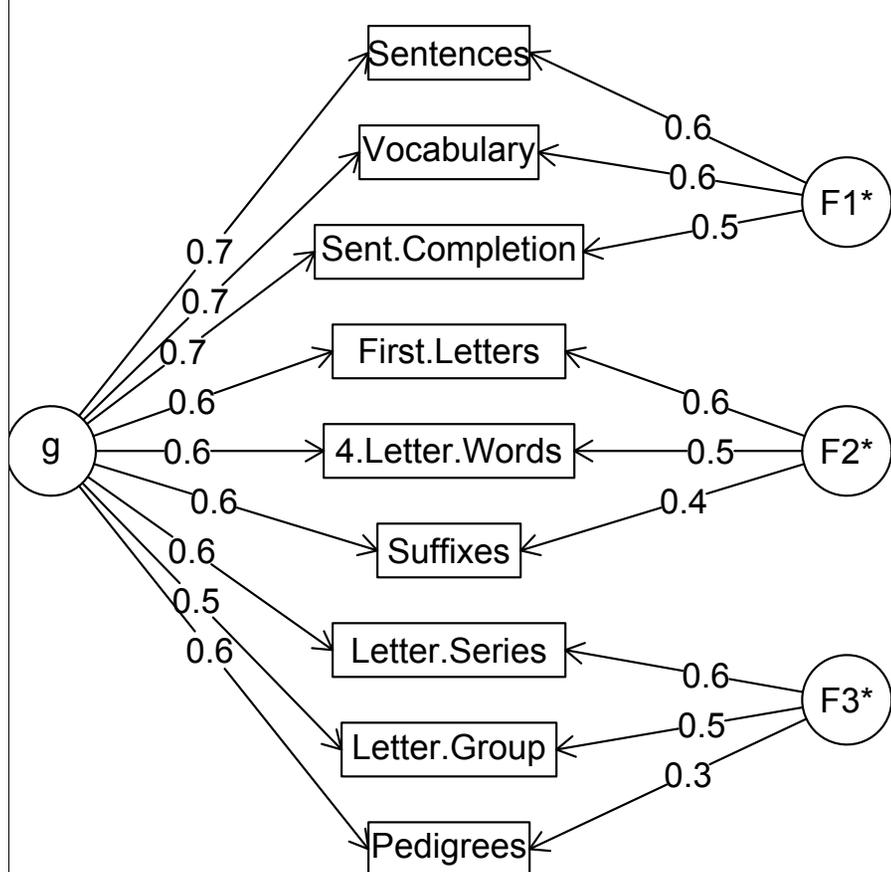


fa.diagram(fa3o)

Omega with Schmid Leiman Transformation

Hierarchical (multilevel) Structure

Hierarchical solutions



```
> oms1 <-omega(Thurstone,title="Bifactor")
```

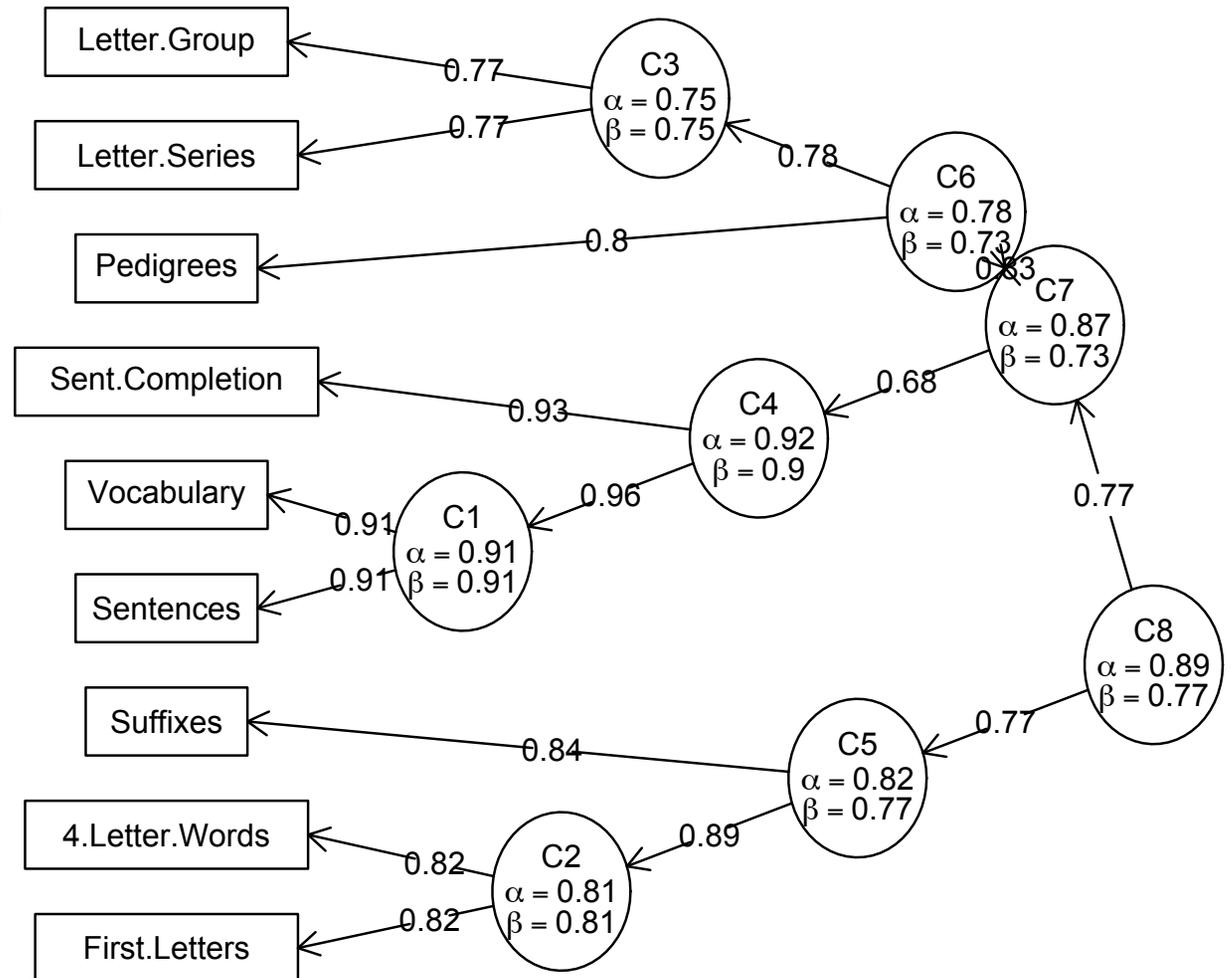
```
> oms1 <-omega(Thurstone,sl=FALSE,title="Hierarchical")
```

Hierarchical Clustering

- I. Find the similarity matrix (correlations)
- II. Find the most similar pair of items/tests
- III. Combine them and repeat II and III until
some criterion (alpha, beta) fails to increase

Hierarchical cluster analysis

ICLUST of Thurstone's 9 variables



Multidimensional scaling

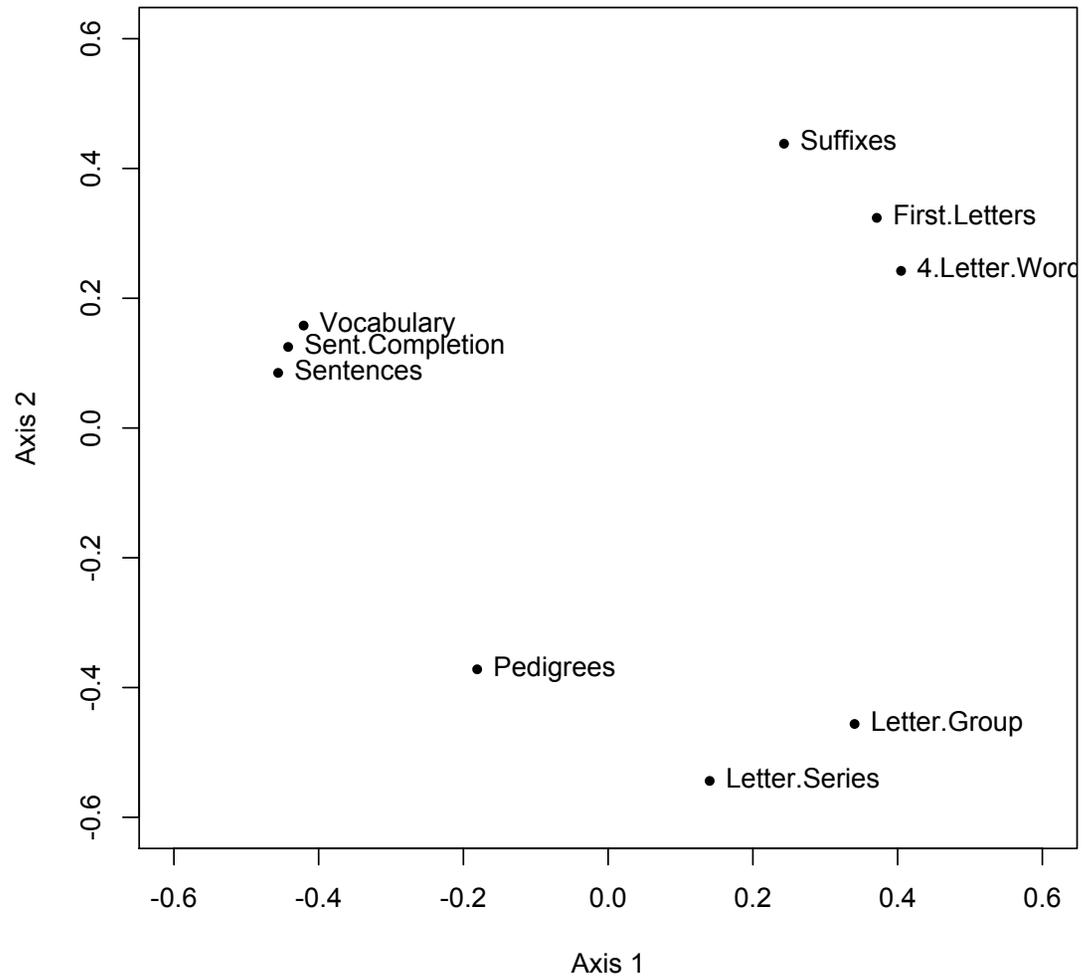
- I. Convert correlations to distances
- II. Multidimensional scaling will remove a general factor since it considers relative ranking of distances

MDS

```
> Thurs.dist <- sqrt(2*(1-Thurstone))  
> mdsT <- cmdscale(Thurs.dist,2)  
> round(mdsT,2)
```

	[,1]	[,2]
Sentences	-0.46	0.08
Vocabulary	-0.42	0.16
Sent.Completion	-0.44	0.12
First.Letters	0.37	0.32
4.Letter.Words	0.40	0.24
Suffixes	0.24	0.44
Letter.Series	0.14	-0.54
Pedigrees	-0.18	-0.37
Letter.Group	0.34	-0.46

MDS plot



```
position <- rep(4,9)
plot(mdsT,xlim=c(-.6,.6),ylim=c(-.6,.6),ylab="Axis 2",xlab="Axis 1")
text(mdsT,rownames(mdsT),pos=position)
```

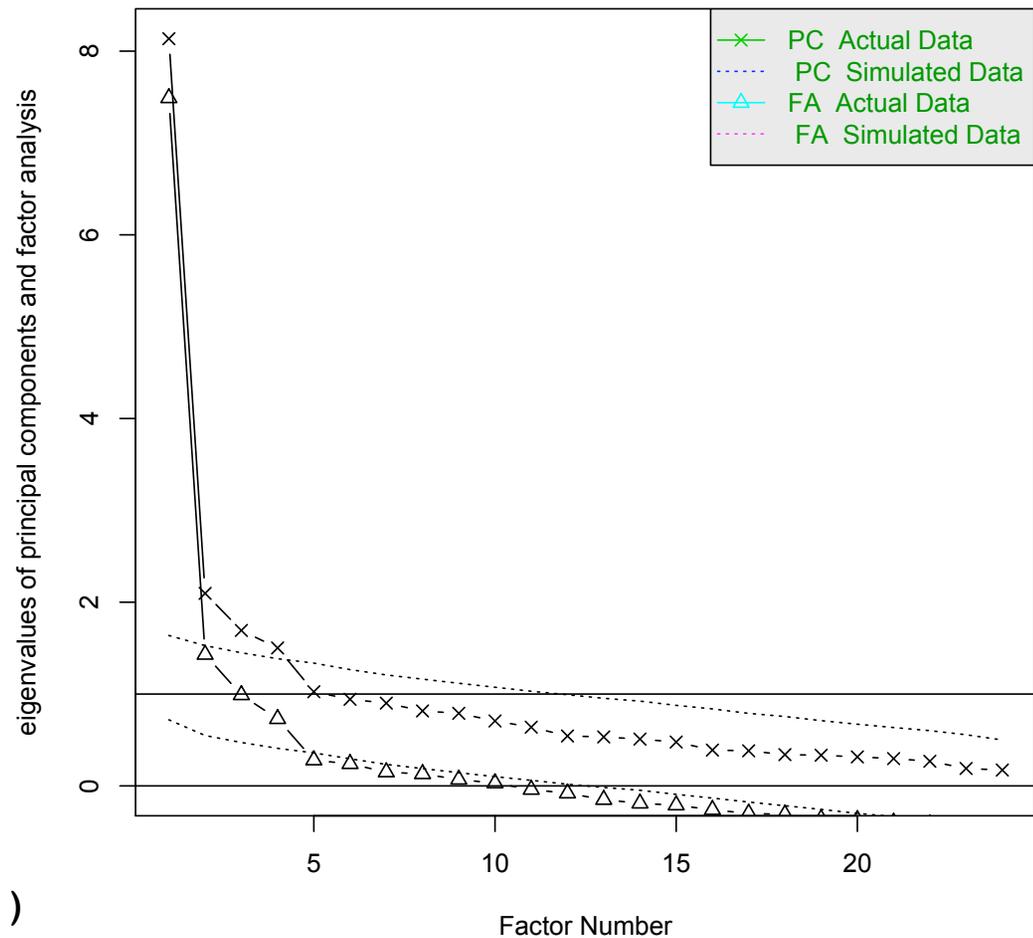
A more complex example

I. Holzinger-Harman 24 mental ability tests

II. Compare FA, CA, MDS

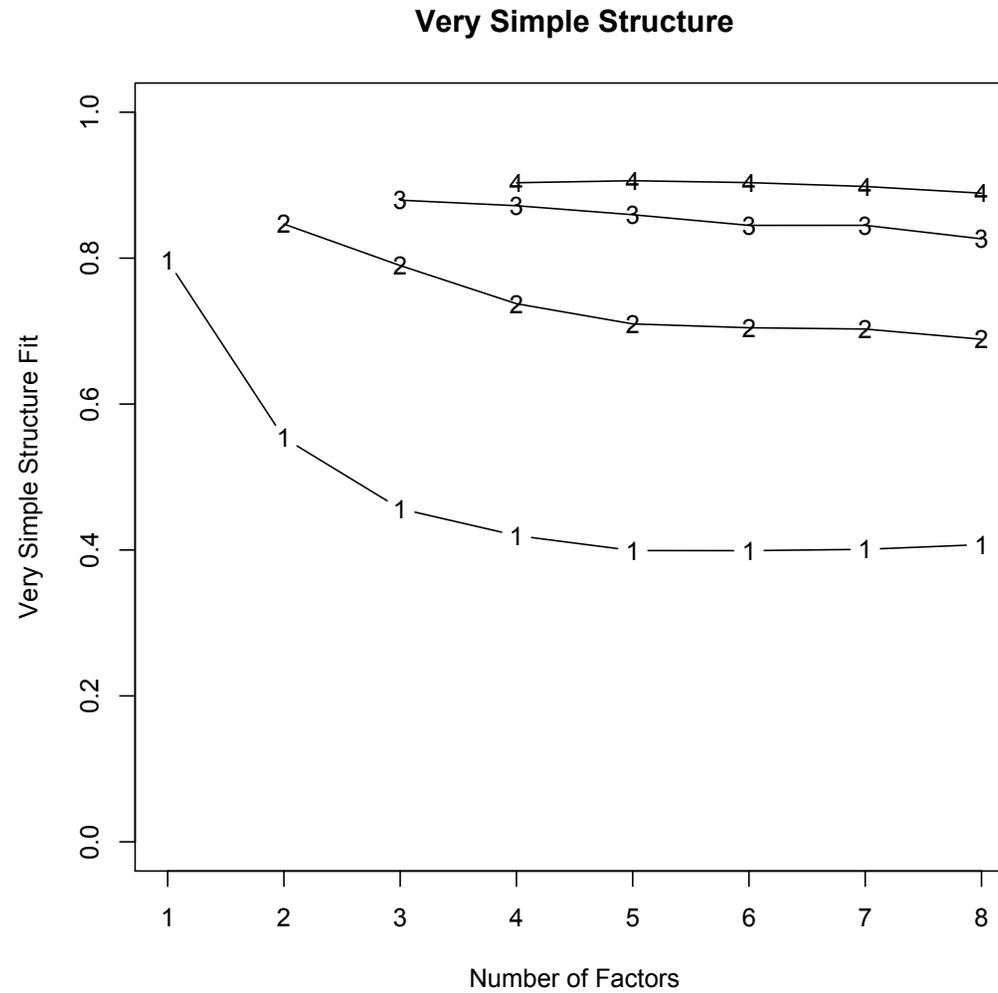
Parallel analysis

Parallel Analysis Scree Plots



```
> hh <- Harman74.cor$cov  
> fa.parallel(hh,n.obs=228)
```

VSS says 1 big factor



```
> vss <- VSS(hh, n.obs=228)
```

MAP suggests 4

```
> vss
```

```
Very Simple Structure
```

```
Call: VSS(x = hh, n.obs = 228)
```

```
VSS complexity 1 achieves a maximum of 0.8 with 1 factors
```

```
VSS complexity 2 achieves a maximum of 0.85 with 2 factors
```

```
The Velicer MAP criterion achieves a minimum of 0.03 with 4 factors
```

```
Velicer MAP
```

```
[1] 0.02 0.02 0.02 0.02 0.02 0.02 0.03 0.03
```

```
Very Simple Structure Complexity 1
```

```
[1] 0.80 0.55 0.46 0.42 0.40 0.40 0.40 0.41
```

```
Very Simple Structure Complexity 2
```

```
[1] 0.00 0.85 0.79 0.74 0.71 0.70 0.70 0.69
```

Chi square says > 8

```
> vss$vss.stats[,1:3]
  dof      chisq      prob
1 252 1012.9000 9.259262e-92
2 229  693.9911 2.649458e-48
3 207  485.1213 2.232607e-24
4 186  370.8825 2.174718e-14
5 166  307.7273 1.479600e-10
6 147  266.8324 5.628353e-09
7 129  226.6602 2.396296e-07
8 112  182.6407 2.832707e-05
>
```

```
> fa4o <- factor.pa(m, 4, n.obs=220, rotate="oblimin")
```

```
> fa4o
```

	V	PA1	PA3	PA2	PA4
VisualPerception	1		0.69		
Cubes	2		0.46		
PaperFormBoard	3		0.54		
Flags	4		0.52		
GeneralInformation	5	0.76			
ParagraphComprehension	6	0.80			
SentenceCompletion	7	0.87			
WordClassification	8	0.56			
WordMeaning	9	0.86			
Addition	10			0.86	
Code	11			0.49	0.30
CountingDots	12			0.70	
StraightCurvedCapitals	13		0.42	0.47	
WordRecognition	14				0.58
NumberRecognition	15				0.55
FigureRecognition	16		0.33		0.52
ObjectNumber	17				0.59
NumberFigure	18				0.43
FigureWord	19				0.32
Deduction	20	0.33	0.31		
NumericalPuzzles	21		0.37	0.33	
ProblemReasoning	22	0.31	0.30		
SeriesCompletion	23	0.30	0.44		
ArithmeticProblems	24			0.41	

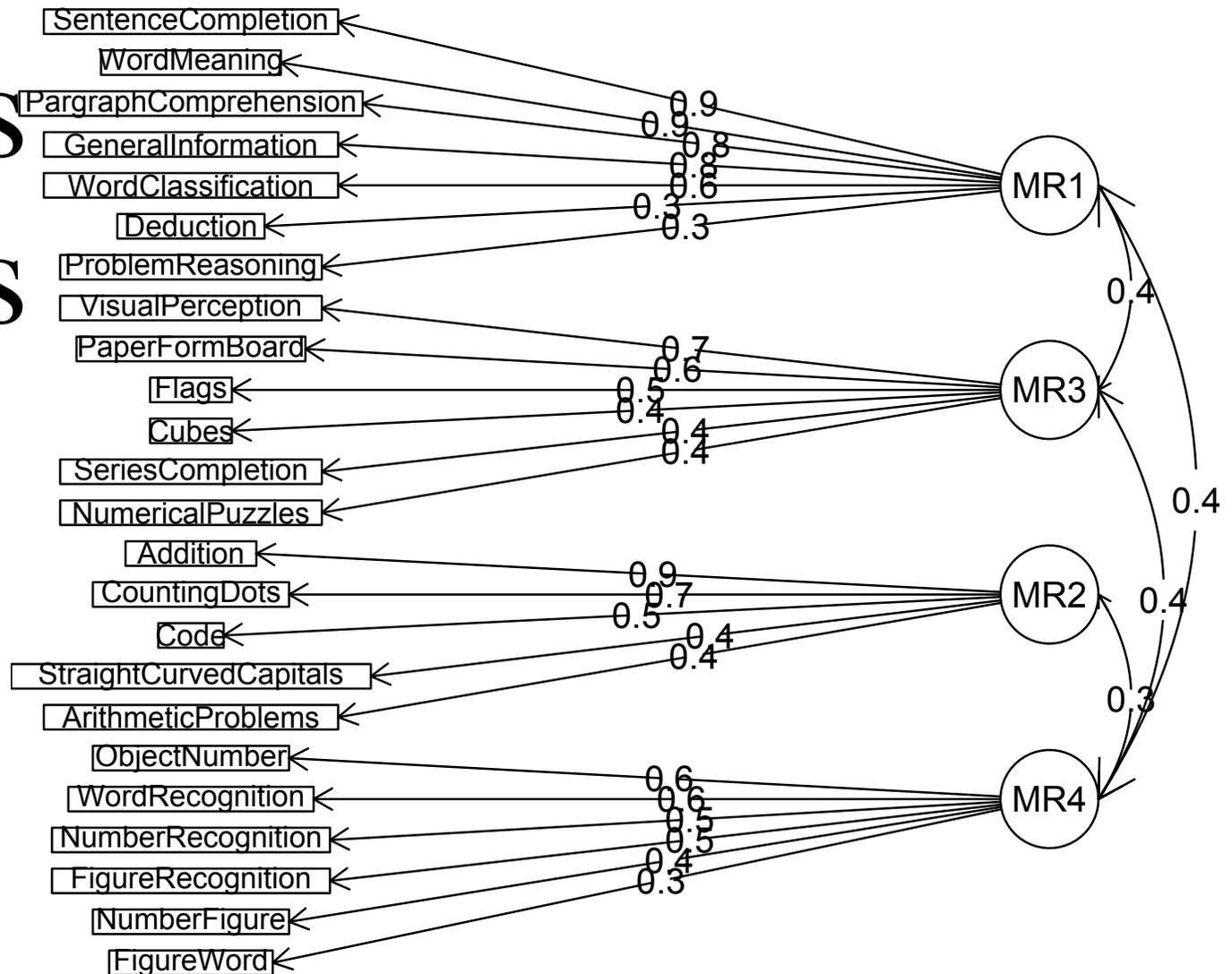
With factor correlations

	PA1	PA3	PA2	PA4
PA1	1.00	0.41	0.30	0.40
PA3	0.41	1.00	0.27	0.38
PA2	0.30	0.27	1.00	0.32

4 oblique factors

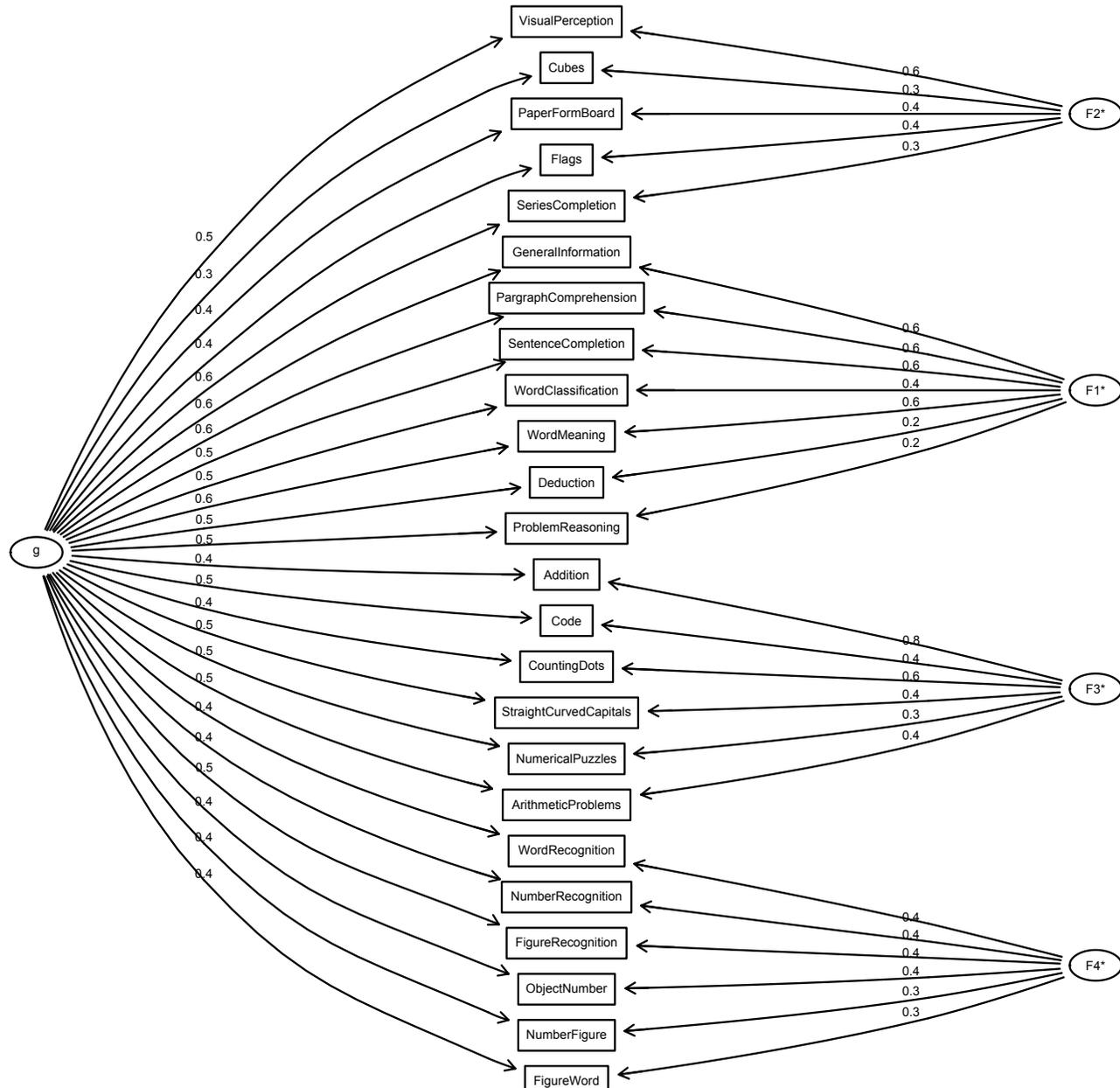
4 oblique factors of the Holzinger-Harman problem

4
minres
factors

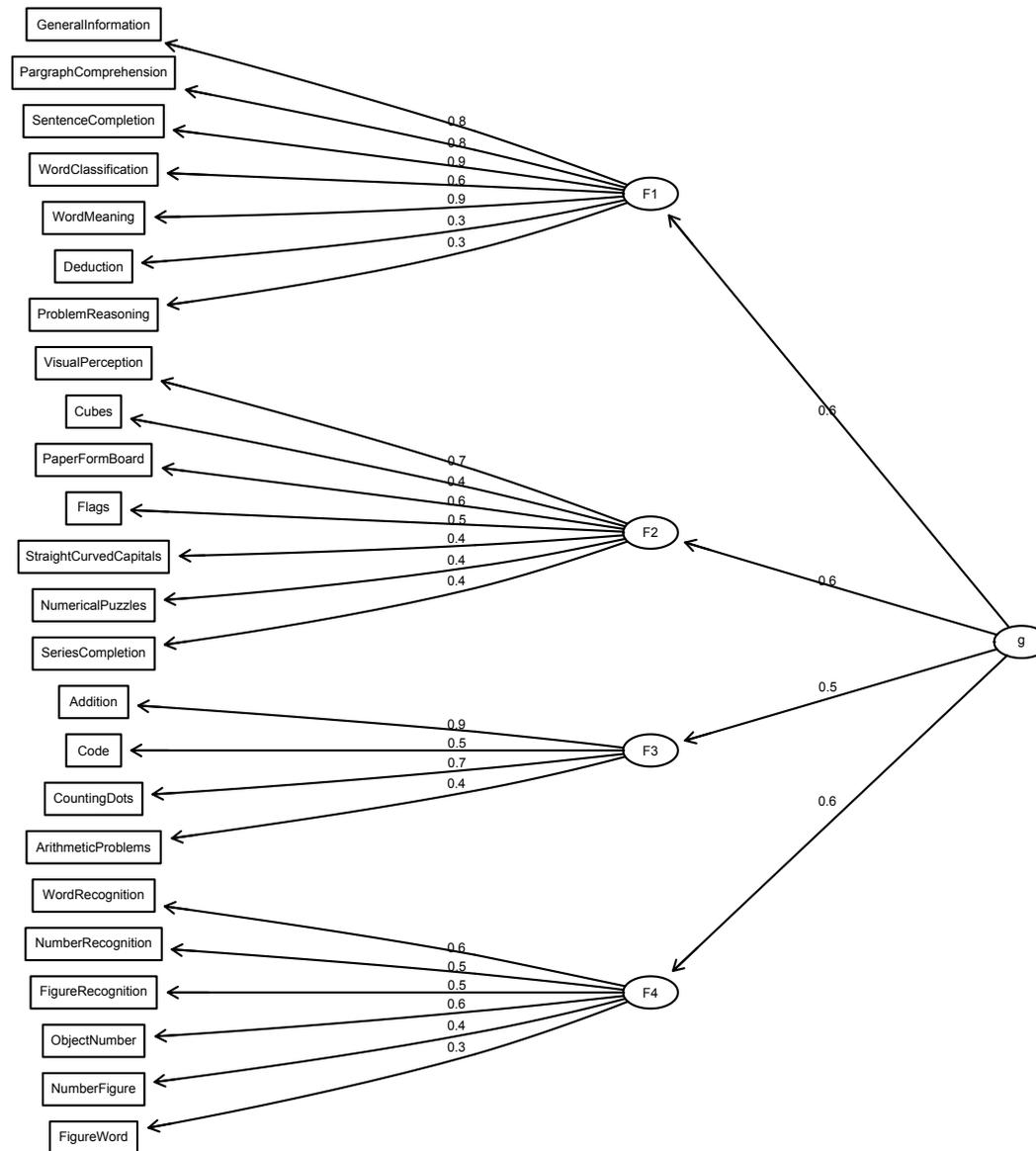


Holzinger Harman problem -- Bifactor

```
om4 <-omega(hh,4,title="Holzinger Harman problem -- Bifactor")
```



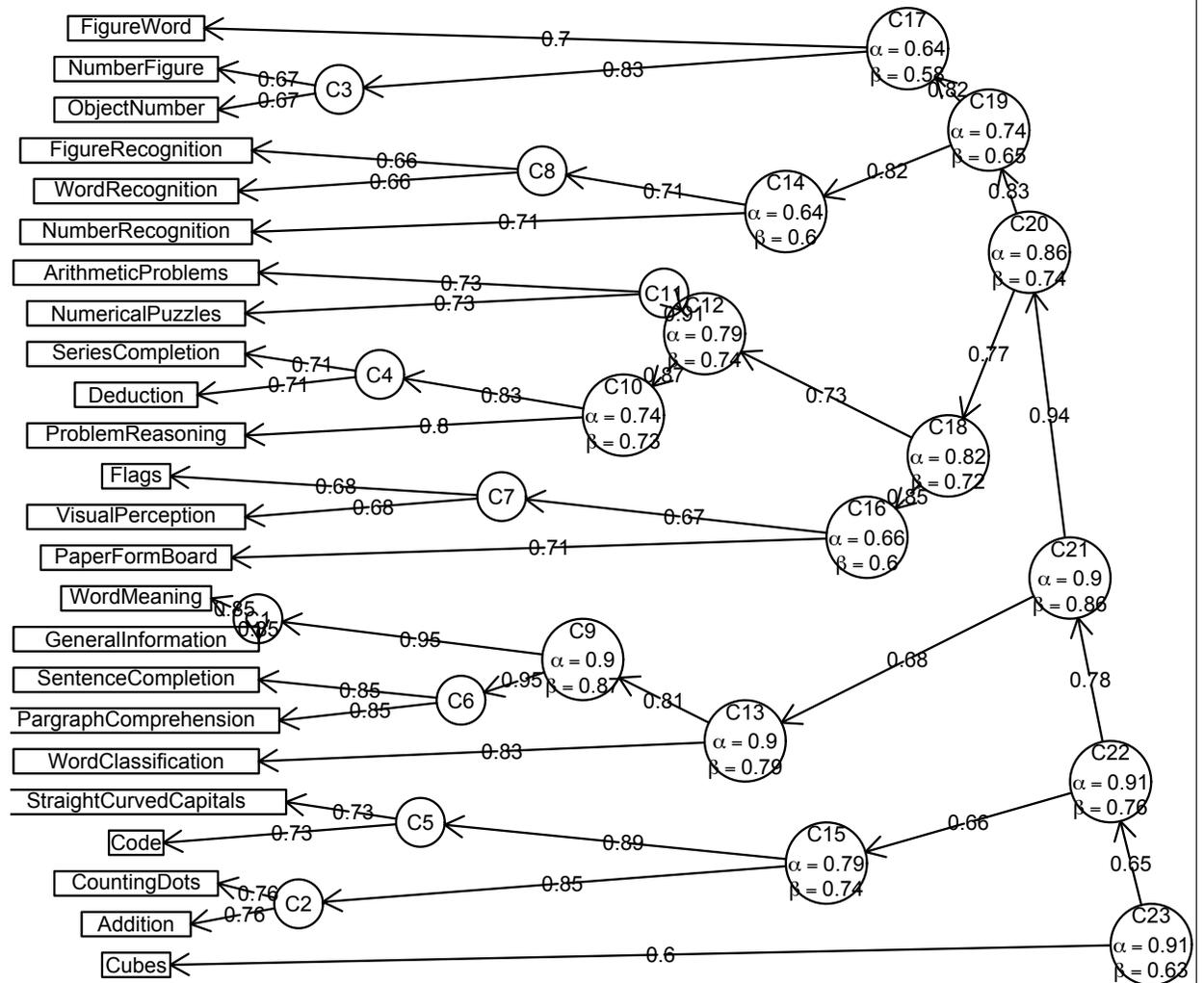
Holzinger Harman problem -- Hierarchical



```
> om4 <-omega(hh,4,title="Holzinger Harman problem -- Hierarchical",sl=FALSE)
```

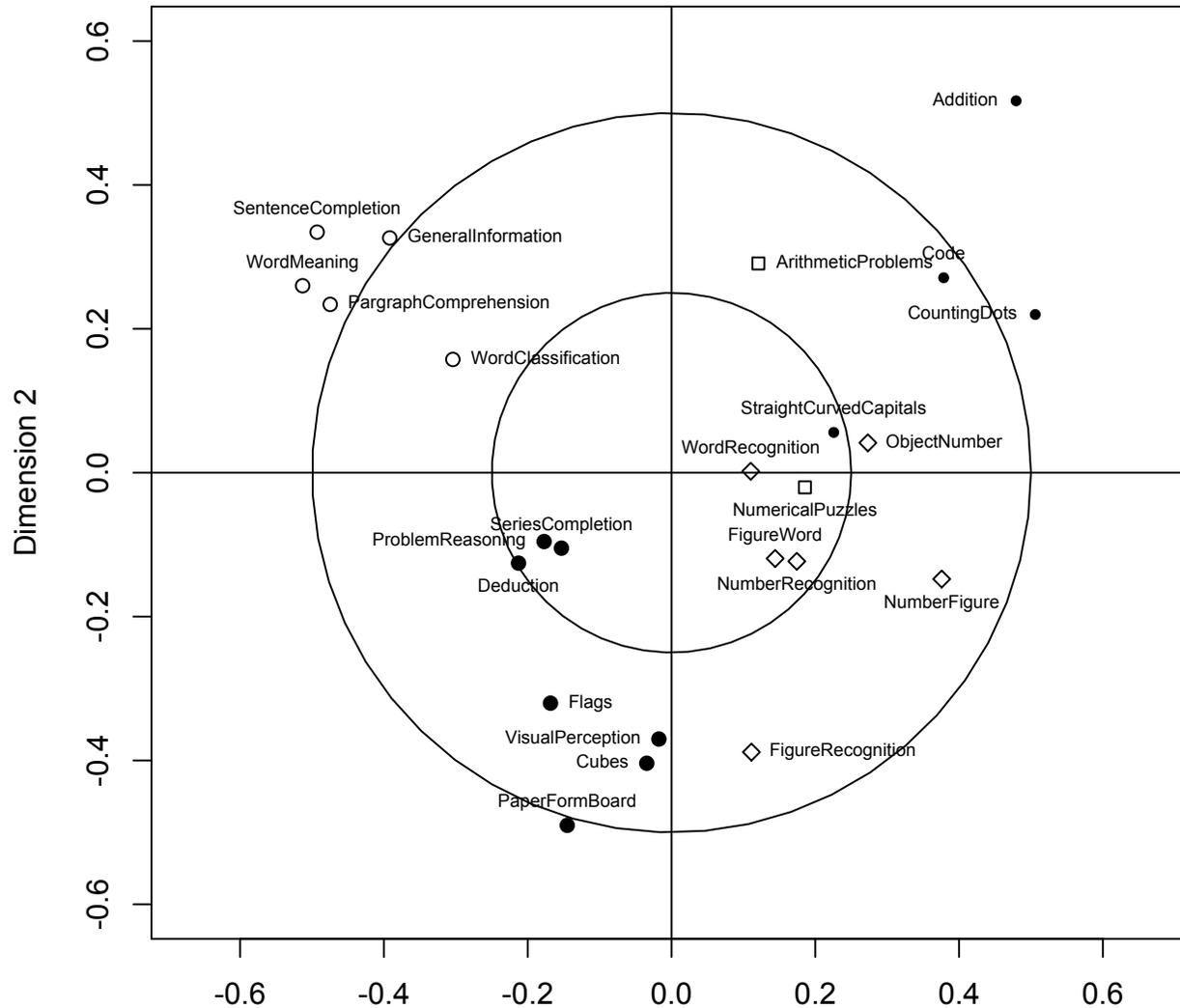

Hierarchical cluster analysis

ICLUST of Holzinger-Harman problem



MDS of HH problem

Multidimensional Scaling of 24 ability tests



code for MDS plot

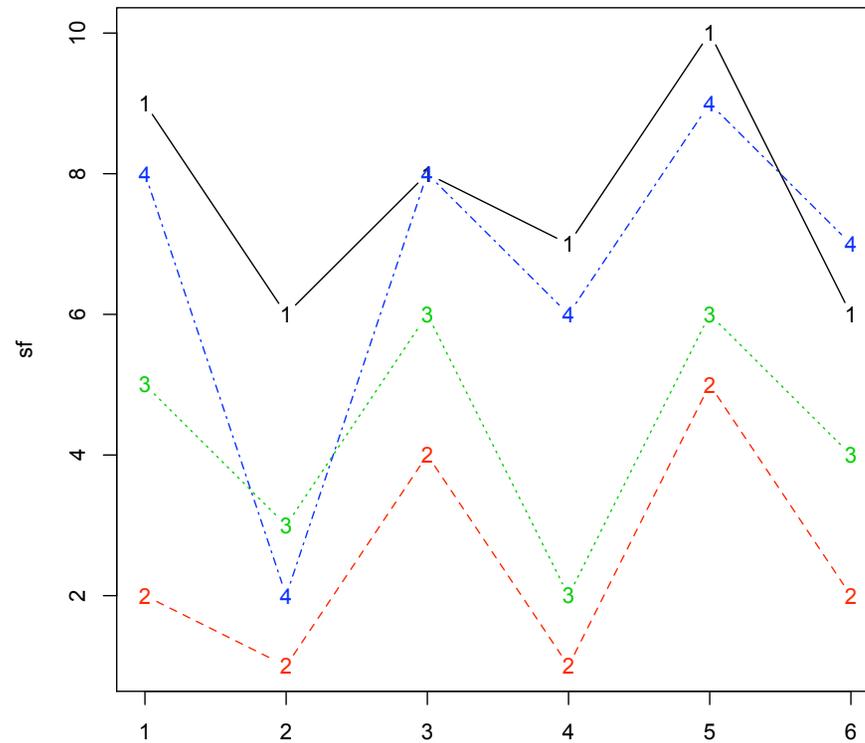
```
> dis24 <- sqrt(2*(1-Harman74.cor$cov))
> mds24 <- cmdscale(dis24,2)
> plot.char <- c( 19, 19, 19, 19,    21, 21,  21, 21,    21,
20, 20, 20,
+    20, 23, 23, 23,    23, 23, 23, 19,    22, 19, 19, 22 )
> plot(mds24,xlim=c(-.6,.6),ylim=c(-.6,.6),xlab="Dimension
1",ylab="Dimension 2",asp=1,pch=plot.char)
> position <- c(2,2,3,4,  4,4,3,4,  3,2,3,2,  3,3,1,4,
4,1,3,1,  1,2,3,4)
> text(mds24,rownames(mds24),cex=.6,pos=position)
> abline(v=0,h=0)
> title("Multidimensional Scaling of 24 ability tests")
> #draw circles at .25 and .50 units away from the center
> segments = 51
> angles <- (0:segments) * 2 * pi/segments
> unit.circle <- cbind(cos(angles), sin(angles))
> lines(unit.circle*.25)
> lines(unit.circle*.5)
```

Reliability of scales and raters

- I. First consider the reliability of raters using the IntraClass Correlation (see Shrout and Fleiss for the definitive discussion)
- II. Types of reliability of ratings (1 or n per target)
 - A. each target rated by a different judge, judges are random
 - B. random sample of k judges rate targets
 - C. Fixed set of k judges give ratings

4 judges rate 6 subjects

```
> sf
      J1 J2 J3 J4
S1    9  2  5  8
S2    6  1  3  2
S3    8  4  6  8
S4    7  1  2  6
S5   10  5  6  9
S6    6  2  4  7
```



Simple correlations (these will remove means for raters)

```
> round(cor(sf), 2)
      J1    J2    J3    J4
J1 1.00 0.75 0.73 0.75
J2 0.75 1.00 0.89 0.73
J3 0.73 0.89 1.00 0.72
J4 0.75 0.73 0.72 1.00
```

ICC

> ICC(sf)

	type	ICC	F	df1	df2	p	lower bound	upper bound
Single_raters_absolute	ICC1	0.17	1.79	5	18	0.16	-0.13	0.72
Single_random_raters	ICC2	0.29	11.03	5	15	0.00	0.02	0.76
Single_fixed_raters	ICC3	0.71	11.03	5	15	0.00	0.34	0.95
Average_raters_absolute	ICC1k	0.44	1.79	5	18	0.16	-0.88	0.91
Average_random_raters	ICC2k	0.62	11.03	5	15	0.00	0.07	0.93
Average_fixed_raters	ICC3k	0.91	11.03	5	15	0.00	0.68	0.99

> alpha(sf)

Alpha of raters

Reliability analysis

Call: alpha(x = sf)

raw_alpha	std.alpha	G6(smc)	average_r	mean	sd
0.91	0.93	0.92	0.76	21	6.7

Reliability if an item is dropped:

	raw_alpha	std.alpha	G6(smc)	average_r	Item statistics					
					n	r	r.cor	mean	sd	
J1	0.88	0.91	0.89	0.78	J1	6	0.89	0.83	7.7	1.6
J2	0.87	0.89	0.85	0.73	J2	6	0.93	0.92	2.5	1.6
J3	0.87	0.90	0.85	0.74	J3	6	0.92	0.91	4.3	1.6
J4	0.92	0.92	0.90	0.79	J4	6	0.88	0.82	6.7	2.5

Reliability of a single scale

```
> round(cor(bfi[,1:10],use="pairwise"),2)
```

	A1	A2	A3	A4	A5	C1	C2	C3	C4	C5
A1	1.00	-0.30	-0.23	-0.12	-0.19	-0.03	-0.06	0.01	0.19	0.08
A2	-0.30	1.00	0.39	0.24	0.41	0.06	0.06	0.22	-0.15	-0.12
A3	-0.23	0.39	1.00	0.27	0.45	0.07	0.12	0.16	-0.14	-0.12
A4	-0.12	0.24	0.27	1.00	0.22	0.07	0.17	0.08	-0.15	-0.17
A5	-0.19	0.41	0.45	0.22	1.00	0.10	0.06	0.20	-0.14	-0.10
C1	-0.03	0.06	0.07	0.07	0.10	1.00	0.44	0.41	-0.39	-0.23
C2	-0.06	0.06	0.12	0.17	0.06	0.44	1.00	0.35	-0.36	-0.24
C3	0.01	0.22	0.16	0.08	0.20	0.41	0.35	1.00	-0.37	-0.36
C4	0.19	-0.15	-0.14	-0.15	-0.14	-0.39	-0.36	-0.37	1.00	0.53
C5	0.08	-0.12	-0.12	-0.17	-0.10	-0.23	-0.24	-0.36	0.53	1.00

Mindless reliability

```
> alpha(bfi[1:10])
```

```
Reliability analysis
```

```
Call: alpha(x = bfi[1:10])
```

```
raw_alpha std.alpha G6(smc) average_r mean sd
      0.19      0.25      0.44      0.032  40 4.7
```

```
Reliability if an item is dropped:
```

```
Item statistics
```

	raw_alpha	std.alpha	G6(smc)	average_r		n	r	r.cor	mean	sd
A1	0.290	0.354	0.51	0.057	A1	1000	0.099	-0.21	2.3	1.3
A2	0.082	0.133	0.35	0.017	A2	994	0.508	0.47	4.8	1.1
A3	0.045	0.101	0.33	0.012	A3	989	0.552	0.54	4.6	1.2
A4	0.108	0.173	0.40	0.023	A4	993	0.448	0.31	4.8	1.4
A5	0.042	0.093	0.32	0.011	A5	988	0.563	0.55	4.6	1.2
C1	0.133	0.190	0.38	0.025	C1	997	0.421	0.34	4.4	1.2
C2	0.123	0.182	0.38	0.024	C2	997	0.434	0.35	4.2	1.3
C3	0.105	0.154	0.36	0.020	C3	995	0.478	0.44	4.3	1.3
C4	0.336	0.392	0.50	0.067	C4	986	0.005	-0.24	2.6	1.4
C5	0.329	0.364	0.49	0.060	C5	997	0.077	-0.17	3.5	1.5

somewhat better reliability

```
> keys <- make.keys(10,list(all=c(-1,2:8,-9,-10)))  
> alpha(bfi[1:10],keys)
```

Reliability analysis

Call: alpha(x = bfi[1:10], keys = keys)

```
raw_alpha std.alpha G6(smc) average_r mean sd  
0.72      0.72      0.75      0.21    40 4.7
```

Reliability if an item is dropped:

Item statistics

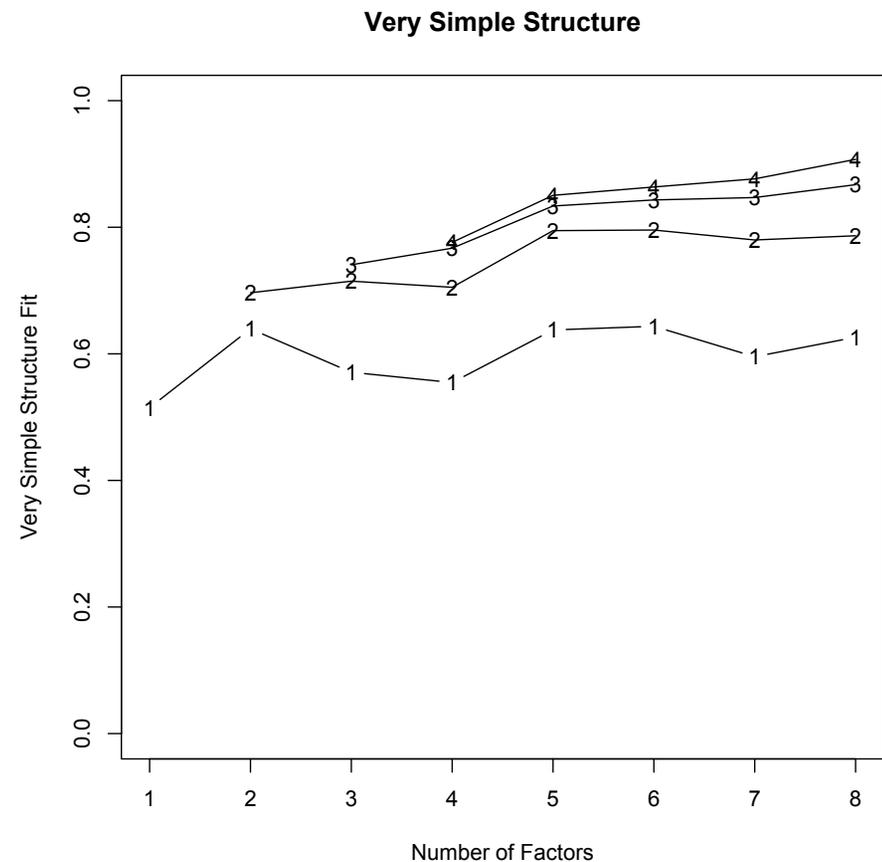
	raw_alpha	std.alpha	G6(smc)	average_r		n	r	r.cor	mean	sd
A1	0.72	0.72	0.74	0.23	A1	1000	0.41	0.29	2.3	1.3
A2	0.70	0.70	0.72	0.20	A2	994	0.55	0.49	4.8	1.1
A3	0.70	0.70	0.72	0.20	A3	989	0.55	0.49	4.6	1.2
A4	0.71	0.71	0.74	0.22	A4	993	0.46	0.35	4.8	1.4
A5	0.70	0.70	0.72	0.21	A5	988	0.53	0.46	4.6	1.2
C1	0.70	0.70	0.72	0.21	C1	997	0.52	0.46	4.4	1.2
C2	0.70	0.70	0.72	0.21	C2	997	0.54	0.47	4.2	1.3
C3	0.69	0.69	0.71	0.20	C3	995	0.59	0.54	4.3	1.3
C4	0.67	0.68	0.70	0.19	C4	986	0.64	0.61	2.6	1.4
C5	0.70	0.70	0.72	0.20	C5	997	0.55	0.49	3.5	1.5

Examine the items

VSS suggests 2 factors!

The items

- A1
Am indifferent to the feelings of others.
- A2
Inquire about others' well-being.
- A3
Know how to comfort others.
- A4
Love children.
- A5
Make people feel at ease.
- C1
Am exacting in my work.
- C2
Continue until everything is perfect.
- C3
Do things according to a plan.
- C4
Do things in a half-way manner.
- C5
Waste my time.



Omega reliability

```
> om2 <- omega(bfi[1:10],2)
```

```
Warning messages:
```

```
1: In schmid(m, nfactors, pc, digits, rotate = rotate, n.obs =  
n.obs,  :
```

```
   Three factors are required for identification -- general factor  
loadings set to be equal. Proceed with caution.
```

```
2: In schmid(m, nfactors, pc, digits = digits, n.obs = n.obs, ...) :
```

```
   Three factors are required for identification -- general factor  
loadings set to be equal. Proceed with caution.
```

```
> om2
Omega
Call: omega(m = bfi[1:10], nfactors = 2)
Alpha: 0.72
G.6: 0.75
Omega Hierarchical: 0.36
Omega Total 0.77
```

Schmid Leiman Factor loadings greater than 0.2

	g	F1*	F2*	h2	u2
A1-	0.21		0.30		0.87
A2	0.36		0.53	0.41	0.59
A3	0.36		0.55	0.43	0.57
A4	0.24		0.28		0.86
A5	0.36		0.54	0.42	0.58
C1	0.30	0.51		0.36	0.64
C2	0.30	0.48		0.32	0.68
C3	0.37	0.48		0.37	0.63
C4-	0.40	0.58		0.50	0.50
C5-	0.33	0.48		0.33	0.67

With eigenvalues of:

	g	F1*	F2*
	1.1	1.3	1.1

Omega h is
low

```
> keys <- make.keys(10,list(all=c(-1,2:8,-9,-10),agree=c(-1,2:5),con=c(6:8,-9,-10)))
> score.items(keys,bfi[1:10])
Call: score.items(keys = keys, items = bfi[1:10])
```

(Unstandardized) Alpha:

	all	agree	con
alpha	0.72	0.65	0.74

Average item correlation:

	all	agree	con
average.r	0.2	0.27	0.36

Guttman 6* reliability:

	all	agree	con
Lambda.6	0.74	0.62	0.72

Scale intercorrelations corrected for attenuation

raw correlations below the diagonal, alpha on the diagonal
corrected correlations above the diagonal:

	all	agree	con
all	0.72	1.10	1.13
agree	0.75	0.65	0.36
con	0.83	0.25	0.74

Score 3 scales

Item scale correlations

Item by scale correlations:

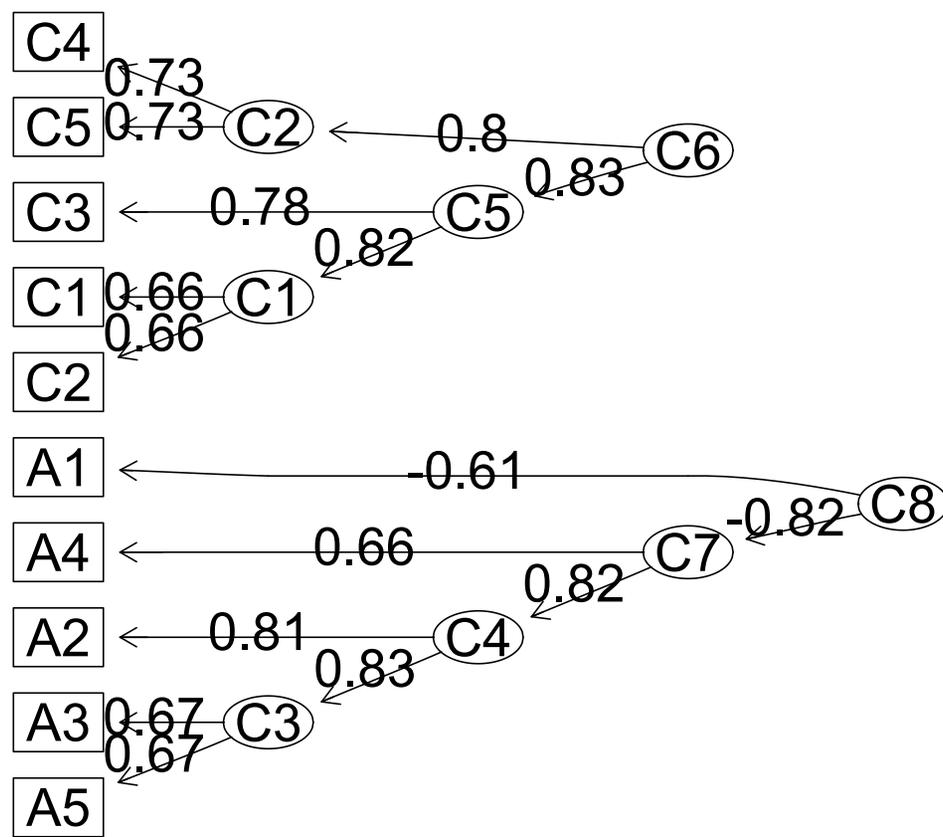
corrected for item overlap and scale reliability

	all	agree	con
A1	-0.28	-0.37	-0.12
A2	0.47	0.63	0.21
A3	0.47	0.63	0.21
A4	0.34	0.37	0.22
A5	0.44	0.58	0.20
C1	0.46	0.13	0.59
C2	0.47	0.19	0.55
C3	0.54	0.25	0.60
C4	-0.62	-0.30	-0.69
C5	-0.50	-0.23	-0.56

>

ICLUST shows 2 scales

ICLUST



Structural Equation modeling in R

I. sem by John Fox

II. does not do multiple group analyses

III. Mx in R is a coming attraction

IV. Using psych as a front end to sem to
generate the model commands

Using psych as front end to sem

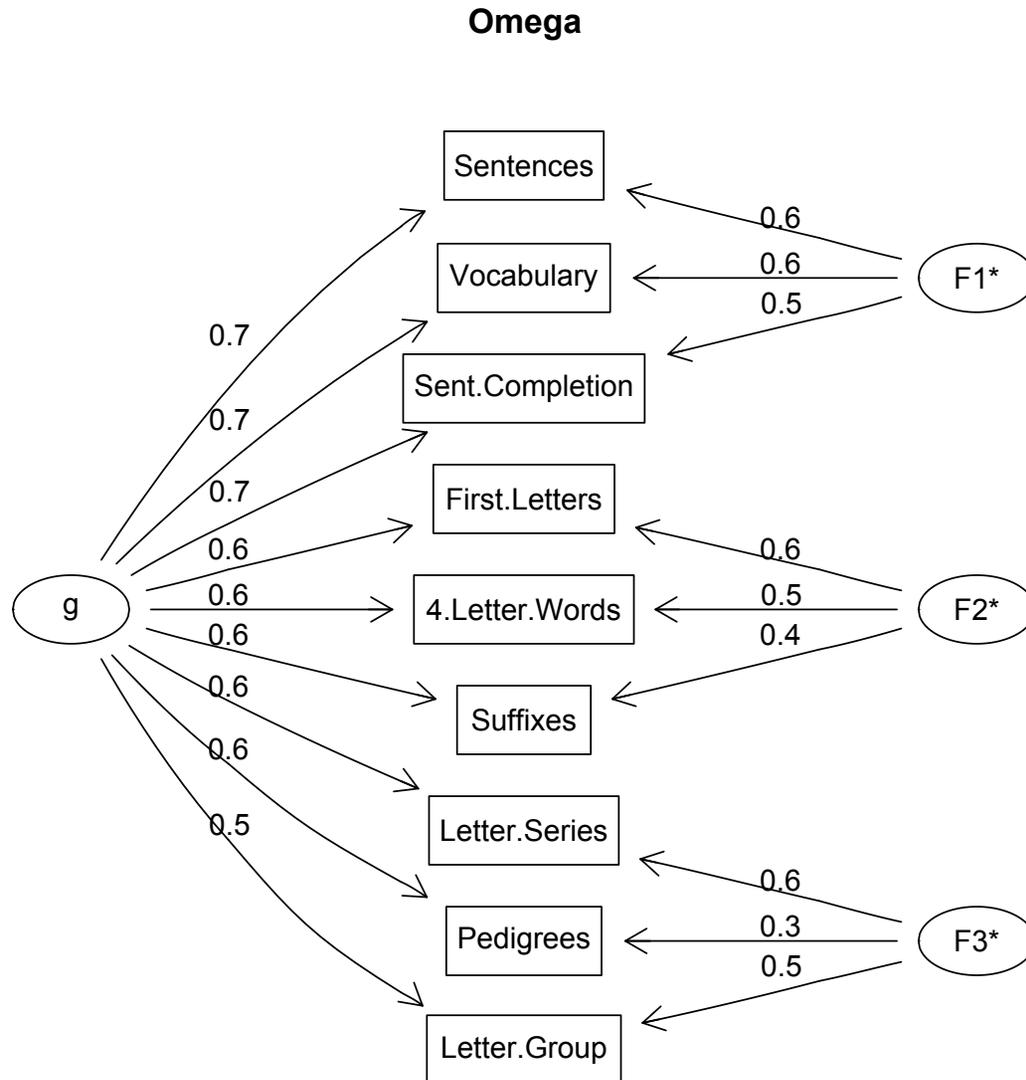
- I. Do the exploratory analysis (fa or omega) in psych
- II. output includes the sem model instructions
- III. run sem
- IV. (see the vignette on using psych for sem)

```
> om <- omega(Thurstone) #creates the path model and the model commands
> om$model
```

Value	Path	Parameter	Initial
[1,]	"g->Sentences"	"Sentences"	NA
[2,]	"g->Vocabulary"	"Vocabulary"	NA
[3,]	"g->Sent.Completion"	"Sent.Completion"	NA
[4,]	"g->First.Letters"	"First.Letters"	NA
[5,]	"g->4.Letter.Words"	"4.Letter.Words"	NA
[6,]	"g->Suffixes"	"Suffixes"	NA
[7,]	"g->Letter.Series"	"Letter.Series"	NA
[8,]	"g->Pedigrees"	"Pedigrees"	NA
[9,]	"g->Letter.Group"	"Letter.Group"	NA
[10,]	"F1*->Sentences"	"F1*Sentences"	NA
[11,]	"F1*->Vocabulary"	"F1*Vocabulary"	NA
[12,]	"F1*->Sent.Completion"	"F1*Sent.Completion"	NA
[13,]	"F2*->First.Letters"	"F2*First.Letters"	NA
[14,]	"F2*->4.Letter.Words"	"F2*4.Letter.Words"	NA
[15,]	"F2*->Suffixes"	"F2*Suffixes"	NA
[16,]	"F3*->Letter.Series"	"F3*Letter.Series"	NA
[17,]	"F3*->Pedigrees"	"F3*Pedigrees"	NA
[18,]	"F3*->Letter.Group"	"F3*Letter.Group"	NA
[19,]	"Sentences<->Sentences"	"e1"	NA
[20,]	"Vocabulary<->Vocabulary"	"e2"	NA
[21,]	"Sent.Completion<->Sent.Completion"	"e3"	NA
[22,]	"First.Letters<->First.Letters"	"e4"	NA
[23,]	"4.Letter.Words<->4.Letter.Words"	"e5"	NA
[24,]	"Suffixes<->Suffixes"	"e6"	NA
[25,]	"Letter.Series<->Letter.Series"	"e7"	NA
[26,]	"Pedigrees<->Pedigrees"	"e8"	NA
[27,]	"Letter.Group<->Letter.Group"	"e9"	NA

The model

The model



Do the sem

```
> library(sem)
> sem.bf <- sem(om$model,Thurstone,213)
> summary(sem.bf,digits=2)
```

```
Model Chisquare = 24    Df = 18 Pr(>Chisq) = 0.15
Chisquare (null model) = 1102    Df = 36
Goodness-of-fit index = 0.98
Adjusted goodness-of-fit index = 0.94
RMSEA index = 0.04    90% CI: (NA, 0.078)
Bentler-Bonnett NFI = 0.98
Tucker-Lewis NNFI = 0.99
Bentler CFI = 1
SRMR = 0.035
BIC = -72
```

Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-8.2e-01	-3.3e-01	-8.9e-07	2.8e-02	1.6e-01	1.8e+00

Parameter values

Parameter Estimates

	Estimate	Std Error	z value	Pr(> z)	
Sentences	0.77	0.073	10.57	0.0e+00	Sentences <--- g
Vocabulary	0.79	0.072	10.92	0.0e+00	Vocabulary <--- g
Sent.Completion	0.75	0.073	10.27	0.0e+00	Sent.Completion <--- g
First.Letters	0.61	0.072	8.43	0.0e+00	First.Letters <--- g
4.Letter.Words	0.60	0.074	8.09	6.7e-16	4.Letter.Words <--- g
Suffixes	0.57	0.071	8.00	1.3e-15	Suffixes <--- g
Letter.Series	0.57	0.074	7.63	2.3e-14	Letter.Series <--- g
Pedigrees	0.66	0.069	9.55	0.0e+00	Pedigrees <--- g
Letter.Group	0.53	0.079	6.71	2.0e-11	Letter.Group <--- g
F1*Sentences	0.49	0.085	5.71	1.1e-08	Sentences <--- F1*
F1*Vocabulary	0.45	0.090	5.00	5.7e-07	Vocabulary <--- F1*
F1*Sent.Completion	0.40	0.093	4.33	1.5e-05	Sent.Completion <--- F1*
F2*First.Letters	0.61	0.086	7.16	8.2e-13	First.Letters <--- F2*
F2*4.Letter.Words	0.51	0.085	5.96	2.5e-09	4.Letter.Words <--- F2*
F2*Suffixes	0.39	0.078	5.04	4.7e-07	Suffixes <--- F2*
F3*Letter.Series	0.73	0.159	4.56	5.1e-06	Letter.Series <--- F3*
F3*Pedigrees	0.25	0.089	2.77	5.6e-03	Pedigrees <--- F3*
F3*Letter.Group	0.41	0.122	3.35	8.1e-04	Letter.Group <--- F3*
e1	0.17	0.034	5.05	4.4e-07	Sentences <--> Sentences
e2	0.17	0.030	5.65	1.6e-08	Vocabulary <--> Vocabulary
e3	0.27	0.033	8.09	6.7e-16	Sent.Completion <--> Sent.Comp
e4	0.25	0.079	3.18	1.5e-03	First.Letters <-First.Letters
e5	0.39	0.063	6.13	8.8e-10	4.Letter.Words <--> 4.Letter.Wo
e6	0.52	0.060	8.68	0.0e+00	Suffixes <--> Suffixes
e7	0.15	0.223	0.67	5.0e-01	Letter.Series <--> Letter.Ser
e8	0.50	0.060	8.39	0.0e+00	Pedigrees <--> Pedigrees

Programming in R

I. Very high level language

A. interpreted at run time

B. can integrate Fortran or C++ code

II. 3 ways of developing code

A. cut and paste from an editor

B. Modify prior code by adding or changing a function

C. source from a file

D. build a package for local or global distribution

Programming in R

I. functions and data structures

A. The output of all functions is an object that will have a certain structure

B. Part of the structure might be invisible but can be shown with the `str()` command.

1. `f3 <- fa(Thurstone,3,rotate=oblmin)`

2. `str(f3)` #will show more than just asking for `f3`

selected output

```
> f3 <- fa(Thurstone,3,rotate="oblimin")
> f3
Factor Analysis using method = minres
Call: fa(r = Thurstone, nfactors = 3, rotate = "oblimin")

      item  MR1  MR2  MR3  h2  u2
Sentences      1  0.90          0.82 0.18
Vocabulary     2  0.89          0.84 0.16
Sent.Completion 3  0.84          0.74 0.26
First.Letters  4          0.85  0.73 0.27
4.Letter.Words 5          0.75  0.63 0.37
Suffixes       6          0.63  0.50 0.50
Letter.Series  7          0.84  0.72 0.28
Pedigrees     8  0.38          0.47 0.50 0.50
Letter.Group   9          0.63  0.52 0.48

      MR1  MR2  MR3
SS loadings  2.64 1.87 1.49
Proportion Var 0.29 0.21 0.17
Cumulative Var 0.29 0.50 0.67
  With factor correlations of
      MR1  MR2  MR3
MR1 1.00 0.59 0.53
MR2 0.59 1.00 0.52
MR3 0.53 0.52 1.00
Test of the hypothesis that 3 factors are sufficient.
The degrees of freedom for the model is 12 and the objective function was 0.01
Fit based upon off diagonal values = 1
Measures of factor score adequacy

      [,1] [,2] [,3]
Correlation of scores with factors 0.96 0.92 0.90
Multiple R square of scores with factors 0.93 0.85 0.82
Minimum correlation of factor score estimates 0.86 0.71 0.63
```

All the output (too much)

```
> print(f3,all=TRUE)
Factor Analysis using method = minres
Call: fa(r = Thurstone, nfactors = 3, rotate = "oblimin")

      item  MR1  MR2  MR3  h2  u2
Sentences      1  0.90      0.82 0.18
Vocabulary      2  0.89      0.84 0.16
Sent.Completion  3  0.84      0.74 0.26
First.Letters   4      0.85      0.73 0.27
4.Letter.Words  5      0.75      0.63 0.37
Suffixes        6      0.63      0.50 0.50
Letter.Series   7      0.84 0.72 0.28
Pedigrees      8  0.38      0.47 0.50 0.50
Letter.Group    9      0.63 0.52 0.48

      MR1  MR2  MR3
SS loadings  2.64 1.87 1.49
Proportion Var 0.29 0.21 0.17
Cumulative Var 0.29 0.50 0.67

With factor correlations of
  MR1  MR2  MR3
MR1 1.00 0.59 0.53
MR2 0.59 1.00 0.52
MR3 0.53 0.52 1.00

Test of the hypothesis that 3 factors are sufficient.

The degrees of freedom for the model is 12 and the objective function was 0.01

Fit based upon off diagonal values = 1
Measures of factor score adequacy

Correlation of scores with factors      [,1] [,2] [,3]
Multiple R square of scores with factors 0.96 0.92 0.90
Minimum correlation of factor score estimates 0.93 0.85 0.82
$residual

      Sentences  Vocabulary Sent.Completion First.Letters 4.Letter.Words
Sentences      0.1797970835  0.003569352  0.001056412 -0.0053521359  0.0054931042
Vocabulary      0.0035693518  0.164376546  -0.004726931  0.0015287573  -0.0013163850
Sent.Completion  0.0010564117 -0.004726931  0.264475270  0.0064257822  -0.0061643080
First.Letters   -0.0053521359  0.001528757  0.006425782  0.2708794227  -0.0007462237
4.Letter.Words  0.0054931042  -0.001316385  -0.006164308  -0.0007462237  0.3691635597
Suffixes        -0.0005413284  0.003307528  -0.006202623  0.0009046706  -0.0007702702
Letter.Series   -0.0001193089  0.005601586  -0.009592364  0.0036158583  -0.0048190635
Pedigrees       -0.0103409056 -0.003359723  0.020557454  -0.0045780810  0.0019664267
Letter.Group    0.0074095333 -0.010637633  0.007277838  -0.0037248900  0.0083593427

      Suffixes Letter.Series  Pedigrees Letter.Group
Sentences -0.0005413284 -0.0001193089 -0.010340906  0.0074095333

$fit
[1] 0.9569096

$fit.off
[1] 0.9998501

$nfactors
[1] 3

$n.obs
[1] NA

$PVAL
[1] NA

$dof
[1] 12

$objective
[1] 0.01390055

$criteria
objective
0.01390055      NA      NA

$Call
fa(r = Thurstone, nfactors = 3, rotate = "oblimin")

$r.scores
      [,1]      [,2]      [,3]
[1,] 1.0000000 0.6615062 0.6122765
[2,] 0.6615062 1.0000000 0.6134362
[3,] 0.6122765 0.6134362 1.0000000

$R2
[1] 0.9285797 0.8525374 0.8161127

$valid
[1] 0.9598034 0.9078568 0.8823116
```

```

> f3 <- fa(Thurstone,3,rotate="oblimin")
> str(f3)
List of 22
 $ residual      : num [1:9, 1:9] 0.1798 0.00357 0.00106 -0.00535 0.00549 ...
  ..- attr(*, "dimnames")=List of 2
  .. ..$ : chr [1:9] "Sentences" "Vocabulary" "Sent.Completion" "First.Letters" ...
  .. ..$ : chr [1:9] "Sentences" "Vocabulary" "Sent.Completion" "First.Letters" ...
 $ fit           : num 0.957
 $ fit.off       : num 1
 $ factors       : num 3
 $ n.obs         : logi NA
 $ PVAL          : logi NA
 $ dof          : num 12
 $ objective     : num 0.0139
 $ criteria      : Named num [1:3] 0.0139 NA NA
  ..- attr(*, "names")= chr [1:3] "objective" "" ""
 $ Call          : language fa(r = Thurstone, nfactors = 3, rotate = "oblimin")
 $ r.scores      : num [1:3, 1:3] 1 0.662 0.612 0.662 1 ...
 $ R2            : num [1:3] 0.929 0.853 0.816
 $ valid         : num [1:3] 0.96 0.908 0.882
 $ score.cor     : num [1:3, 1:3] 1 0.571 0.574 0.571 1 ...
 $ weights       : num [1:9, 1:3] 0.35332 0.38596 0.2284 0.01086 0.00503 ...
  ..- attr(*, "dimnames")=List of 2
  .. ..$ : chr [1:9] "Sentences" "Vocabulary" "Sent.Completion" "First.Letters" ...
  .. ..$ : NULL
 $ communality   : Named num [1:9] 0.82 0.84 0.74 0.73 0.63 0.5 0.72 0.5 0.52
  ..- attr(*, "names")= chr [1:9] "Sentences" "Vocabulary" "Sent.Completion" "First.Letters" ...
 $ uniquenesses : Named num [1:9] 0.18 0.16 0.26 0.27 0.37 0.5 0.28 0.5 0.48
  ..- attr(*, "names")= chr [1:9] "Sentences" "Vocabulary" "Sent.Completion" "First.Letters" ...
 $ values        : num [1:9] 4.85 1.09 1.04 0.48 0.45 0.37 0.32 0.23 0.17
 $ loadings      : loadings [1:9, 1:3] 0.90356 0.889 0.83522 -0.00297 -0.01535 ...
  ..- attr(*, "dimnames")=List of 2
  .. ..$ : chr [1:9] "Sentences" "Vocabulary" "Sent.Completion" "First.Letters" ...
  .. ..$ : chr [1:3] "MR1" "MR2" "MR3"
 $ fm            : chr "minres"
 $ Phi           : num [1:3, 1:3] 1 0.592 0.535 0.592 1 ...
 $ fn            : chr "fa"
 - attr(*, "class")= chr [1:2] "psych" "fa"

```

Using
str to
show
what is
there

Programming in R

I. Data types

II. operators

III. simple functions

IV. Writing functions

Data structures

I. elements: logical, integer, real, character, factor

II. vectors: ordered sets of elements of one type

III. matrices: ordered sets of vectors (all of one type)

IV. data.frames: ordered sets of vectors, may be
different types

V. lists: ordered set of anything

Operators

I. arithmetical

1. $+$, $-$, $*$, $/$, $^$, $\% \%$

2. $a + b$, $a - b$, $a * b$, a / b , $a ^ b$, $a \% \% b$

II. Logical

A. $a == b$, $!a$, $a != b$, $a > b$, $a < b$, $a >= b$, $a <= b$

III. Matrix

A. $\% * \%$ is matrix multiplication

B. $\% o \%$ is outer product

```
> a <- 2      example operations
> b <- 3
> v <- 5:10
> w <- 6:7
> v
[1] 5 6 7 8 9 10
> w
[1] 6 7
> v ^ a
[1] 25 36 49 64 81 100
> w * b
[1] 18 21
> w * v
[1] 30 42 42 56 54 70
```

Matrix operations

```
> v
[1] 5 6 7 8 9 10
> t(v)
      [,1] [,2] [,3] [,4] [,5] [,6]
[1,]    5    6    7    8    9   10
> t(v)%*% v
      [,1]
[1,]  355
> v %*% t(v)
      [,1] [,2] [,3] [,4] [,5] [,6]
[1,]   25   30   35   40   45   50
[2,]   30   36   42   48   54   60
[3,]   35   42   49   56   63   70
[4,]   40   48   56   64   72   80
[5,]   45   54   63   72   81   90
[6,]   50   60   70   80   90  100
```

Additional matrix operators

outer product

```
> v
[1] 5 6 7 8 9 10
> w
[1] 6 7
> v %O% w
      [,1] [,2]
[1,] 30 35
[2,] 36 42
[3,] 42 49
[4,] 48 56
[5,] 54 63
[6,] 60 70
```

kroncker

matrix “addition” (psych)

```
> x <- seq(4,8,2)
> x
[1] 4 6 8
> x %+% t(x)
      [,1] [,2] [,3]
[1,]  8 10 12
[2,] 10 12 14
[3,] 12 14 16
```

Functions

I. Operate on an object and provide a new object

II. e.g., `f <- function(x) {x * 2}`

```
> f <- function(x) {x * 2}
```

```
> f(43)
```

```
[1] 86
```

```
> x
```

```
[1] 4 6 8
```

```
> f(x)
```

```
[1] 8 12 16
```

```
> f( v %O% w)
```

```
      [,1] [,2]
```

```
[1,]    60    70
```

```
[2,]    72    84
```

```
[3,]    84    98
```

```
[4,]    96   112
```

```
[5,]   108   126
```

```
[6,]   120   140
```

Simple functions

a subset of useful functions

I. `is.na()`, `is.null()`, `is.vector()`, `is.matrix()`, `is.list()`

II. `sum()`, `rowSums()`, `colSums()`, `mean(x)`,
`rowMeans()`, `colMeans()`, `max`, `min`, `median`
(these work on the entire matrix)

III. `var`, `cov`, `cor`, `sd` (these work on the columns
of the matrix/data.frame)

IV. `help.start()` brings up a web page of manuals

More useful functions

I. `rep(x,n)` (repeats the value x n times)

II. `c(x,y)` (combines x with y)

III. `cbind(x,y)` combines column wise

IV. `rbind(x,y)` combines rowwise

V. `seq(a,b,c)` sequence from a to b stepping by c

sums on matrices and data.frames

```
> z <- f( v %O% w)
```

```
> z
```

```
      [,1] [,2]
[1,]   60   70
[2,]   72   84
[3,]   84   98
[4,]   96  112
[5,]  108  126
[6,]  120  140
```

```
> sum(z)
```

```
[1] 1170
```

```
> min(z)
```

```
[1] 60
```

```
> max(z)
```

```
[1] 140
```

```
> median(z)
```

```
[1] 97
```

```
> rowSums(z)
```

```
[1] 130 156 182 208 234 260
```

```
> colSums(z)
```

```
[1] 540 630
```

```
> mean(z)
```

```
[1] 97.5
```

```
> rowMeans(z)
```

```
[1] 65 78 91 104 117 130
```

Basic stats functions, part 2

```
> var(z)
      [,1] [,2]
[1,]  504  588
[2,]  588  686
> cov(z)
      [,1] [,2]
[1,]  504  588
[2,]  588  686
> cor(z)
      [,1] [,2]
[1,]    1    1
[2,]    1    1
> sd(z)
[1] 22.44994 26.19160
> z
      [,1] [,2]
[1,]   60   70
[2,]   72   84
[3,]   84   98
[4,]   96  112
[5,]  108  126
[6,]  120  140
```

?cor

```
var(x, y = NULL, na.rm = FALSE, use)
```

```
cov(x, y = NULL, use = "everything",  
     method = c("pearson", "kendall",  
                "spearman"))
```

```
cor(x, y = NULL, use = "everything",  
     method = c("pearson", "kendall",  
                "spearman"))
```

```
cov2cor(V)
```

More on cor

x

a numeric vector, matrix or data frame.

y

NULL (default) or a vector, matrix or data frame with compatible dimensions to x. The default is equivalent to $y = x$ (but more efficient).

na.rm

logical. Should missing values be removed?

use

an optional character string giving a method for computing covariances in the presence of missing values. This must be (an abbreviation of) one of the strings "everything", "all.obs", "complete.obs", "na.or.complete", or "pairwise.complete.obs".

method

a character string indicating which correlation coefficient (or covariance) is to be computed. One of "pearson" (default), "kendall", or "spearman", can be abbreviated.

V

symmetric numeric matrix, usually positive definite such as a covariance matrix.

row and col as

functions

```
> r <- .8  
> R <- diag(1,8)  
> R
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]
[1,]	1	0	0	0	0	0	0	0
[2,]	0	1	0	0	0	0	0	0
[3,]	0	0	1	0	0	0	0	0
[4,]	0	0	0	1	0	0	0	0
[5,]	0	0	0	0	1	0	0	0
[6,]	0	0	0	0	0	1	0	0
[7,]	0	0	0	0	0	0	1	0
[8,]	0	0	0	0	0	0	0	1

```
> R <- r^(abs(row(R))-col(R))
```

```
> round(R,2)
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]
[1,]	1.00	0.80	0.64	0.51	0.41	0.33	0.26	0.21
[2,]	0.80	1.00	0.80	0.64	0.51	0.41	0.33	0.26
[3,]	0.64	0.80	1.00	0.80	0.64	0.51	0.41	0.33
[4,]	0.51	0.64	0.80	1.00	0.80	0.64	0.51	0.41
[5,]	0.41	0.51	0.64	0.80	1.00	0.80	0.64	0.51
[6,]	0.33	0.41	0.51	0.64	0.80	1.00	0.80	0.64
[7,]	0.26	0.33	0.41	0.51	0.64	0.80	1.00	0.80
[8,]	0.21	0.26	0.33	0.41	0.51	0.64	0.80	1.00

Yet more stats functions

I. `sample(n, N, replace=TRUE)`

II. `eigen(X)` (eigen value decomposition of X)

III. `solve(X)` (inverse of X)

IV. `solve(X, Y)` Regression of Y on X

```
> x <- matrix(sample(10,50,replace=TRUE),ncol=5)
```

```
> x
```

```
      [,1] [,2] [,3] [,4] [,5]
[1,]   10    3    4    4    6
[2,]    3   10    8    8    9
[3,]    1    6    5    8    5
[4,]    9    1    3    5    3
[5,]    6    8    3    5    1
[6,]    8    6   10    1   10
[7,]   10    5   10    2    1
[8,]    9    3    2    2    9
[9,]    6   10    2    9    4
[10,]   1    8    8    2    6
```

Creating a matrix

```
> z <- scale(x)
```

```
> z
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]
[1,]  1.04838349 -0.9819805 -0.4678877 -0.2059329  0.1852621
[2,] -0.93504473  1.3093073  0.7798129  1.1669533  1.1115724
[3,] -1.50173851  0.0000000 -0.1559626  1.1669533 -0.1235080
...
[9,] -0.08500407  1.3093073 -1.0917380  1.5101749 -0.4322782
[10,] -1.50173851  0.6546537  0.7798129 -0.8923761  0.1852621
attr(,"scaled:center")
[1] 6.3 6.0 5.5 4.6 5.4
attr(,"scaled:scale")
[1] 3.529243 3.055050 3.205897 2.913570 3.238655
```

standardize it

Just center it

```
> c <- scale(x,scale=FALSE)
```

```
> c
```

```
      [,1] [,2] [,3] [,4] [,5]
[1,]  3.7  -3  -1.5 -0.6  0.6
[2,] -3.3   4   2.5  3.4  3.6
[3,] -5.3   0  -0.5  3.4 -0.4
[4,]  2.7  -5  -2.5  0.4 -2.4
[5,] -0.3   2  -2.5  0.4 -4.4
[6,]  1.7   0   4.5 -3.6  4.6
[7,]  3.7  -1   4.5 -2.6 -4.4
[8,]  2.7  -3  -3.5 -2.6  3.6
[9,] -0.3   4  -3.5  4.4 -1.4
[10,] -5.3   2   2.5 -2.6  0.6
attr(,"scaled:center")
[1] 6.3 6.0 5.5 4.6 5.4
```

Find the covariance and inverse

```
> c <- cov(x)
> round(c,2)
      [,1] [,2] [,3] [,4] [,5]
[1,] 12.46 -6.89 -1.61 -4.53 -1.58
[2,] -6.89  9.33  2.11  4.11  0.56
[3,] -1.61  2.11 10.28 -3.89  2.22
[4,] -4.53  4.11 -3.89  8.49 -1.60
[5,] -1.58  0.56  2.22 -1.60 10.49
```

```
> round(solve(c),2)
      [,1] [,2] [,3] [,4] [,5]
[1,] 0.15  0.08  0.02  0.06  0.02
[2,] 0.08  0.23 -0.07 -0.10  0.00
[3,] 0.02 -0.07  0.16  0.12 -0.01
[4,] 0.06 -0.10  0.12  0.26  0.03
[5,] 0.02  0.00 -0.01  0.03  0.10
```

Flow control

I. if(condition) {then do this} else {do this}

II. for (condition) do {expression}

A. for (i in 1:n} do {x <- x + 1}

III.while (condition) {expression}

conditionals

I. (a & b) vs. (a && b)

II. (a | b) vs. (a || b)

```
a <- 1
> if (a & b) {print ("hello")} else {print("goodby")}
Error: object 'b' not found
> if (a && b ) {print ("hello")} else {print("goodby")}
[1] "goodby"
> if (a | b) {print ("hello")} else {print("goodby")}
Error: object 'b' not found

> if (a || b) {print ("hello")} else {print("goodby")}
Error: object 'b' not found
> a <- 1
> if (a || b) {print ("hello")} else {print("goodby")}
[1] "hello"
>
```

simple control

```
> a <- 1
> b <- 2
> c <- 3
> k <- 10
> x <-1
> if(x == a) {print("x is the same as a and has
value",x)} else {print ("x is not equal to a")}

> x <- 3
> if(x == a) {print("x is the same as a and has
value",x)} else {print ("x is not equal to a")}
[1] "x is not equal to a"
>
```

Make that a function

```
> f1 <- function(x,y) {if(x == y) {print  
("x is the same as y and has value",x)}  
else {print ("x is not equal to y")}}  
> f1(3,4)  
[1] "x is not equal to y"  
> f1(5,5)  
[1] "x is the same as y and has value"
```

Simple functions:part 2

- I. Find the squared multiple correlation of a variable with all the other variables in a matrix.
- II. The R^2 is 1- the residual variance

The essence of the function

```
SMC <- function(R) {  
  R.inv <- solve(R)  
  SMC <- 1 - 1/diag(R.inv)}
```

```
> S <- cor(attitude)  
> SMC(S)      #does not show anything
```

```
> round(SMC(S),2) #but this does
```

	rating	complaints	privileges	learning	raises
critical	0.73	0.77	0.38	0.62	0.68
advance	0.19	0.52			

Add a return

```
SMC <- function(R) {  
  R.inv <- solve(R)  
  SMC <- 1 - 1/diag(R.inv)  
  return(SMC)}  

```

```
> SMC(S)
```

```
      rating complaints privileges   learning   raises  
critical   advance  
  0.7326020  0.7700868  0.3831176  0.6194561  0.6770498  
0.1881465  0.5186447
```

Allow it to find R

```
SMC <- function(R) {  
  if(dim(R)[1] !=dim(R)[2]) {R <-cor(R)}  
  R.inv <- solve(R)  
  SMC <- 1 - 1/diag(R.inv)  
  return(SMC)}  

```

```
> SMC(attitude)  
      rating complaints privileges    learning    raises  
critical    advance  
  0.7326020  0.7700868  0.3831176  0.6194561  0.6770498  
0.1881465  0.5186447
```

Clean up the output

```
SMC <- function(R,digits=2) {  
  if(dim(R)[1] !=dim(R)[2]) {R <-cor(R)}  
  R.inv <- solve(R)  
  SMC <- 1 - 1/diag(R.inv)  
  return(round(SMC,digits))}
```

```
> SMC(attitude)  
      rating complaints privileges   learning   raises   critical  
advance  
      0.73      0.77      0.38      0.62      0.68      0.19  
0.52  
>
```

Check for poor input

```
> att <- data.frame(attitude[1:3],attitude[1:3])
> SMC(att)
Error in solve.default(R) :
  Lapack routine dgesv: system is exactly singular
```

Add checks for weird matrices

```
SMC <- function(R,digits=2) {  
  p <- dim(R)[2]  
  if (dim(R)[1] != p) {R <-cor(R)}  
  R.inv <- try(solve(R),TRUE)  
  if(class(R.inv)== as.character("try-error")) {SMC <- rep(1,p)  
    warning("Correlation matrix not invertible, smc's returned as 1s")}  
  else {smc <- 1 -1/diag(R.inv)}  
  SMC <- 1 - 1/diag(R.inv)}  
  return(round(SMC,digits))}
```

```
> SMC(att)
```

```
[1] 1 1 1 1 1 1
```

```
Warning message:
```

```
In SMC(att) : Correlation matrix not invertible, smc's returned as  
1s
```

```
> SMC(attitude)
```

rating	complaints	privileges	learning	raises	critical
0.73	0.77	0.38	0.62	0.68	0.19

Further checks

Input is a covariance matrix

```
> SMC(cov(attitude))
  rating complaints privileges learning raises critical
advance
  -38.62    -39.76    -91.35    -51.42    -33.91    -78.49
-49.96
```

Input is raw data or correlations

```
> SMC(cor(attitude))
  rating complaints privileges learning raises critical
advance
  0.73    0.77    0.38    0.62    0.68    0.19
0.52
```

```
> SMC(attitude)
  rating complaints privileges learning raises critical
advance
  0.73    0.77    0.38    0.62    0.68    0.19
0.52
```

Final version

```
#modified Dec 10, 2008 to return 1 on diagonal if non-invertible
#modified March 20, 2009 to return smcs * variance if covariance matrix
is desired
#modified April 8, 2009 to remove bug introduced March 10 when using
covar from data
"smc" <-
function(R,covar =FALSE) {
failed=FALSE
  p <- dim(R)[2]
  if (dim(R)[1] != p) {if(covar) {C <- cov(R, use="pairwise")
                                vari <- diag(C)
                                R <- cov2cor(C)
                                } else {R <- cor(R,use="pairwise")}}
else {vari <- diag(R)
      R <- cov2cor(R)
      if (!is.matrix(R)) R <- as.matrix(R)}
R.inv <- try(solve(R),TRUE)
if(class(R.inv)== as.character("try-error")) {smc <- rep(1,p)
warning("Correlation matrix not invertible, smc's returned as 1s")}
else {smc <- 1 -1/diag(R.inv)
      if(covar) {smc <- smc * vari}}
return(smc)
}
```

Creating a new function

- I. Is there a base function to modify?
- II. Consider the case of modifying Promax rotation to allow for any target matrix
- III. original promax (inside the factanal package) had been modified to report the factor correlation.
- IV. This version was created with the assistance of Pat Shrout and Steve Miller

promax

```
> promax
function (x, m = 4)
{
  if (ncol(x) < 2)
    return(x)
  dn <- dimnames(x)
  xx <- varimax(x)
  x <- xx$loadings
  Q <- x * abs(x)^(m - 1)
  U <- lm.fit(x, Q)$coefficients
  d <- diag(solve(t(U) %*% U))
  U <- U %*% diag(sqrt(d))
  dimnames(U) <- NULL
  z <- x %*% U
  U <- xx$rotmat %*% U
  dimnames(z) <- dn
  class(z) <- "loadings"
  list(loadings = z, rotmat = U)
}
<environment: namespace:stats>
```

Promax

```
> Promax
function (x, m = 4)
{
  if (!is.matrix(x) & !is.data.frame(x)) {
    if (!is.null(x$loadings))
      x <- as.matrix(x$loadings)
  }
  else {
    x <- x
  }
  if (ncol(x) < 2)
    return(x)
  dn <- dimnames(x)
  xx <- varimax(x)
  x <- xx$loadings
  Q <- x * abs(x)^(m - 1)
  U <- lm.fit(x, Q)$coefficients
  d <- diag(solve(t(U) %*% U))
  U <- U %*% diag(sqrt(d))
  dimnames(U) <- NULL
  z <- x %*% U
  U <- xx$rotmat %*% U
  ui <- solve(U)
  Phi <- ui %*% t(ui)
  dimnames(z) <- dn
  class(z) <- "loadings"
  result <- list(loadings = z, rotmat = U, Phi = Phi)
  class(result) <- c("psych", "fa")
  return(result)}
}
```

```

"target.rot" <-
function (x, keys=NULL,m = 4)
{
  if(!is.matrix(x) & !is.data.frame(x) ) {
    if(!is.null(x$loadings)) x <- as.matrix(x$loadings)
  } else {x <- x}
  if (ncol(x) < 2)
    return(x)
  dn <- dimnames(x)
  if(is.null(keys)) {xx <- varimax(x)
  x <- xx$loadings
  Q <- x * abs(x)^(m - 1)} else {Q <- keys}
  U <- lm.fit(x, Q)$coefficients
  d <- diag(solve(t(U) %*% U))
  U <- U %*% diag(sqrt(d))
  dimnames(U) <- NULL
  z <- x %*% U
  if (is.null(keys)) {U <- xx$rotmat %*% U } else {U <- U}
  ui <- solve(U)
  Phi <- ui %*% t(ui)
  dimnames(z) <- dn
  class(z) <- "loadings"
  result <- list(loadings = z, rotmat = U,Phi = Phi)
  class(result) <- c("psych","fa")
  return(result)
}

```

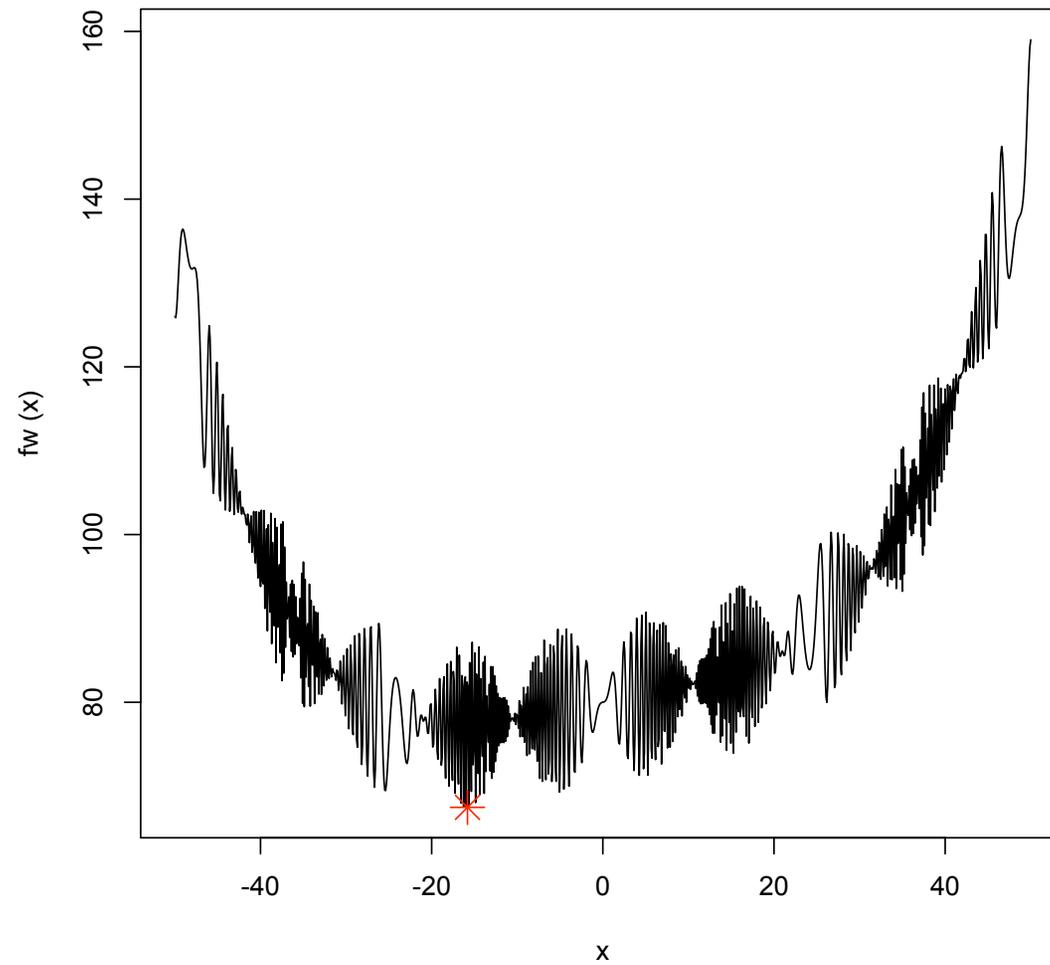
target.rot

optim as “solver” for R

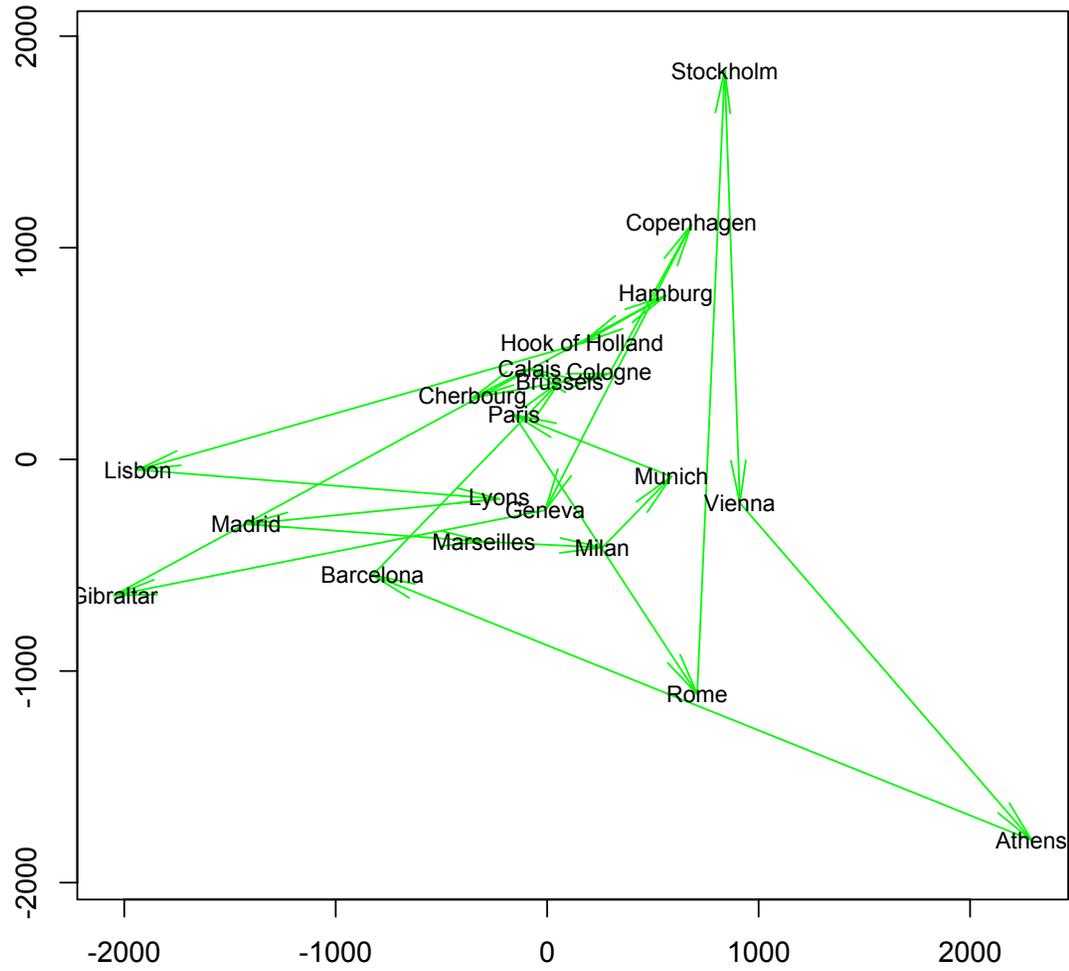
- I. Many statistical functions are not closed form but rather are solved iteratively.
- II. We start with a good guess and then minimize the function
- III. optim will do this for functions where you manipulate one vector (which can of course actually be a matrix)

optim

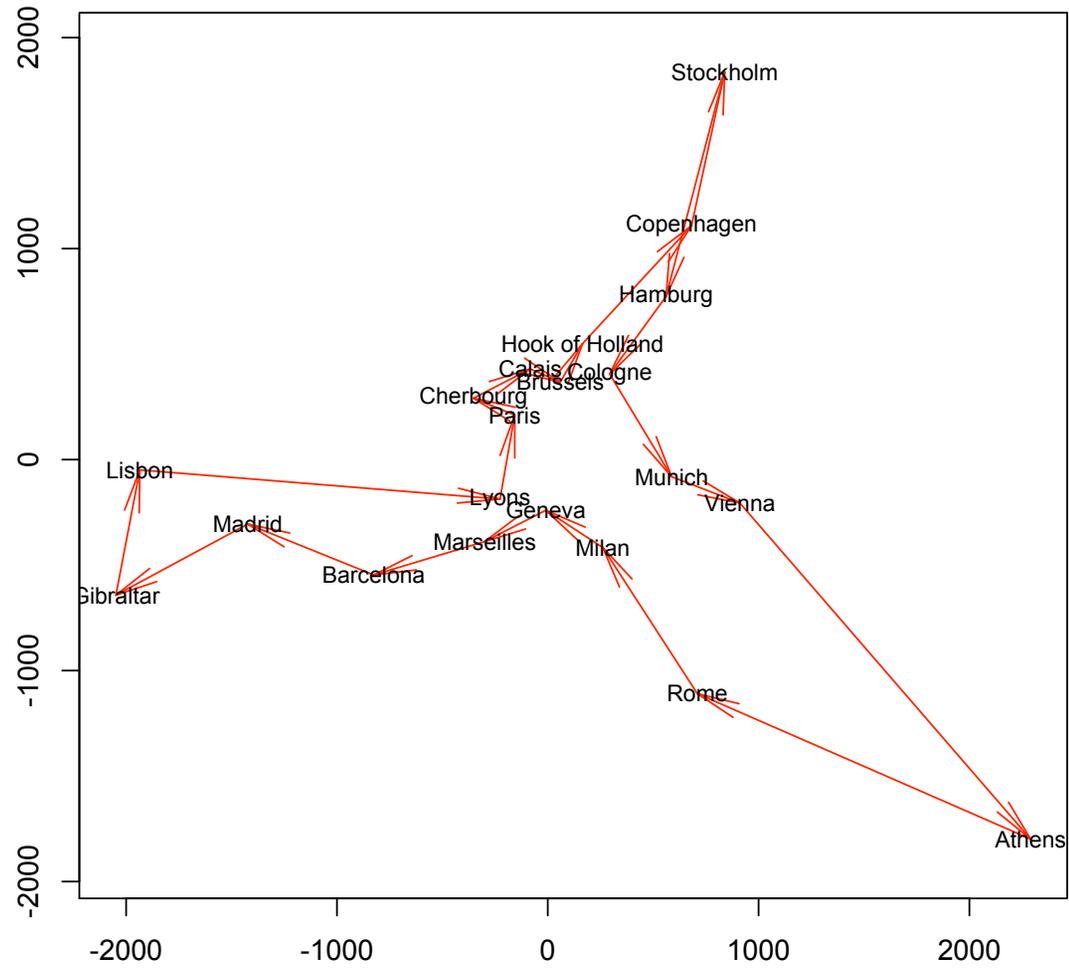
optim() minimising 'wild function'



initial solution of traveling salesman problem



optim() 'solving' traveling salesman problem



Trying to make a new function to do OLS FA

I. First, look at current ML FA function

A. `factanal`

B. It turns out that the critical optimization is done in `factanal.fit.mle`, but where is that?

II. `getAnywhere(factanal.fit.mle)`

A. then look at the code

B. scratch your head and try running it

Sharing your code

- I. Post the source code on your web site
- II. develop a package which you keep on a “repository” on your web site
- III. develop a package and upload it to CRAN

Package development

- I. a somewhat dated tutorial is found at <http://personality-project.org/r/makingpackages.html>

Package development

I. `package.skeleton(yourpackage)` #creates directories and subdirectories for a package

A.this includes a number of different subdirectories and files.

II.`prompt(yourfunction)` #creates a draft help file for your function

III.Document as you go

in X11

I. R CMD check mypackage #makes sure all the code is correct, checks for matches with documentation, runs all the example,

II.R CMD make mypackage #builds the package, but does not check except for working R

A.this will convert the helpfiles from Rd files to html, LaTeX, and pdf

III.R CMD install mypackage will add any changes to your current package

Using X11 to check packages

```
bash-3.2$ cd /Volumes/Test
bash-3.2$ R CMD check psych
* checking for working pdflatex ... OK
* using log directory '/Volumes/Test/psych.Rcheck'
* using R version 2.9.2 Patched (2009-09-24 r49893)
* using session charset: ASCII
* checking for file 'psych/DESCRIPTION' ... OK
* this is package 'psych' version '1.0-82'
* checking package name space information ... OK
* checking package dependencies ... OK
* checking if this is a source package ... OK
* checking for executable files ... OK
* checking whether package 'psych' can be installed ... OK
* checking package directory ... OK
* checking for portable file names ... OK
* checking for sufficient/correct file permissions ... OK
* checking DESCRIPTION meta-information ... OK
* checking top-level files ... OK
* checking index information ... OK
* checking package subdirectories ... OK
* checking R files for non-ASCII characters ... OK
* checking R files for syntax errors ... OK
* checking whether the package can be loaded ... OK
* checking whether the package can be loaded with stated dependencies ... OK
* checking whether the name space can be loaded with stated dependencies ... OK
* checking for unstated dependencies in R code ... OK
* checking S3 generic/method consistency ... OK
* checking replacement functions ... OK
* checking foreign function calls ... OK
* checking R code for possible problems ... OK
* checking Rd files ... OK
* checking Rd files against version 2 parser ... OK
* checking Rd cross-references ... OK
* checking for missing documentation entries ... OK
* checking for code/documentation mismatches ... OK
* checking Rd \usage sections ... OK
* checking data for non-ASCII characters ... OK
* checking examples ... OK
* checking package vignettes in 'inst/doc' ... OK
* checking PDF version of manual ... OK

bash-3.2$ █
```

The structure of a package

▼ psych.Rcheck	Today, 3:16 PM	--	Folder
psych-manual.pdf	Today, 3:15 PM	1 MB	Adobe PDF document
psych-manual.log	Today, 3:15 PM	70 KB	BBEdit text document
00check.log	Today, 3:15 PM	4 KB	BBEdit text document
psych-Ex.Rout	Today, 3:14 PM	180 KB	Document
psych-Ex.ps	Today, 3:14 PM	6.5 MB	PostScript
▼ inst	Today, 3:14 PM	--	Folder
▶ doc	Today, 3:14 PM	--	Folder
test.ICLUST.graph.dot	Today, 3:13 PM	4 KB	Microsoft Word document
psych-Ex.R	Today, 3:12 PM	70 KB	BBEdit text document
R.css	Today, 3:12 PM	4 KB	CSS style sheet
00install.out	Today, 3:12 PM	12 KB	gmon.out
▼ psych	Today, 3:12 PM	--	Folder
▶ Meta	Today, 3:12 PM	--	Folder
INDEX	Today, 3:12 PM	16 KB	Document
▶ doc	Today, 3:12 PM	--	Folder
▶ R-ex	Today, 3:12 PM	--	Folder
▶ latex	Today, 3:12 PM	--	Folder
▶ html	Today, 3:12 PM	--	Folder
▶ help	Today, 3:12 PM	--	Folder
CONTENTS	Today, 3:12 PM	25 KB	Document
▶ R	Today, 3:11 PM	--	Folder
▶ man	Today, 3:11 PM	--	Folder
NEWS	Today, 3:11 PM	12 KB	SimpleText Format
DESCRIPTION	Today, 3:11 PM	4 KB	Document
▶ data	Today, 3:11 PM	--	Folder
NAMESPACE	Today, 3:11 PM	4 KB	Document

The package

▼	psych	Aug 28, 2009 11:02 AM	--	Folder
	CHANGES	Oct 5, 2009 9:48 AM	12 KB	SimpleText Format
	DESCRIPTION	Oct 5, 2009 9:48 AM	8 KB	BEdit text document
▶	man	Oct 4, 2009 7:39 PM	--	Folder
	NAMESPACE	Sep 29, 2009 10:04 PM	4 KB	SimpleText Format
▶	R	Sep 27, 2009 6:34 PM	--	Folder
	sort.rowcol.R	Aug 28, 2009 11:02 AM	4 KB	BEdit text document
▶	data	Aug 23, 2009 12:31 PM	--	Folder
▼	inst	Jun 30, 2009 7:39 PM	--	Folder
▼	doc	Yesterday, 2:16 PM	--	Folder
	psych_for_sem.pdf	Yesterday, 2:16 PM	414 KB	Adobe PDF document
	overview.pdf	Yesterday, 2:15 PM	2.3 MB	Adobe PDF document
	psych_for_sem	Sep 30, 2009 9:25 AM	37 KB	Rnw File
	overview	Sep 30, 2009 9:00 AM	66 KB	Rnw File
	psych_manual.pdf	Jul 26, 2009 12:23 PM	1 MB	Adobe PDF document
	all	Jun 30, 2009 7:36 PM	1.5 MB	BibTeX
	apa.bst	Mar 21, 2007 4:33 PM	25 KB	bst File
	NEWS	Sep 30, 2009 7:45 PM	12 KB	SimpleText Format

```

mat.sort          text  html  latex  example
matrix.addition  text  html  latex  example
msq              text  html  latex  example
multi.hist       text  html  latex  example
neo             text  html  latex  example
omega           text  html  latex  example
omega.graph     text  html  latex  example
p.rep          text  html  latex  example
paired,r        text  html  latex  example
pairs.panels    text  html  latex  example
partial.r       text  html  latex  example
peas           text  html  latex  example
phi            text  html  latex  example
phi,demo       text  html  latex  example
phi2poly       text  html  latex  example
plot.psych     text  html  latex  example
polar          text  html  latex  example
poly.mat       text  html  latex  example
polychor.matrix text  html  latex  example
principal      text  html  latex  example
print.psych    text  html  latex  example
r.test         text  html  latex  example
read.clipboard text  html  latex  example
rescale        text  html  latex  example
sat.act        text  html  latex  example
scaling.fits   text  html  latex  example
schmid         text  html  latex  example
score.alpha    text  html  latex  example
score.items    text  html  latex  example
score,multiple,choice text  html  latex  example
sim            text  html  latex  example
sim,VSS        text  html  latex  example
sim.anova      text  html  latex  example
sim,congeneric text  html  latex  example
sim,hierarchical text  html  latex  example
sim,item       text  html  latex  example
sim,structural text  html  latex  example
simulation.circ text  html  latex  example
skew           text  html  latex  example
smc            text  html  latex  example
structure.graph text  html  latex  example
structure.list text  html  latex  example
super.matrix   text  html  latex  example
table2matrix   text  html  latex  example
test.psych     text  html  latex  example
thurstone     text  html  latex  example
tr            text  html  latex  example
vegetables     text  html  latex  example
winsor        text  html  latex  example
* building package indices ...
^ DONE (psych)
^ creating vignettes ... OK
^ removing junk files
^ excluding invalid files from 'psych'
subdirectory 'R' contains invalid file names:
  correlations neo pearson power reliability.data.sets
^ checking for LF line-endings in source and make files
^ checking for empty or unneeded directories
^ building 'psych_1.0-82.tar.gz'

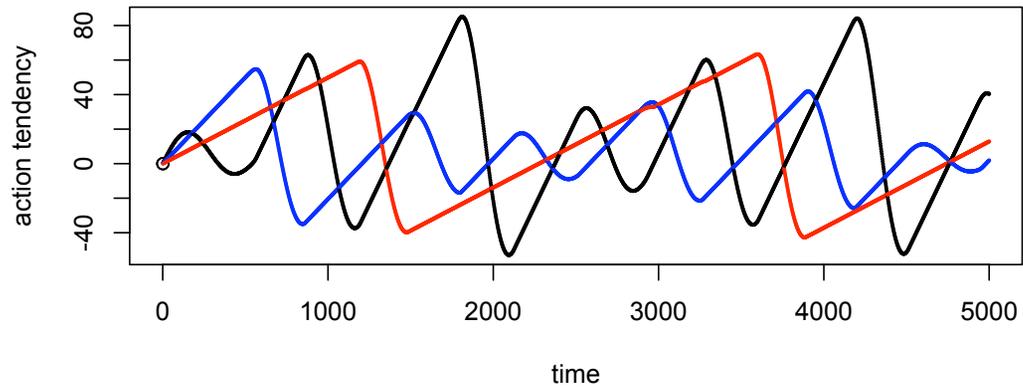
```

R CMD build psych

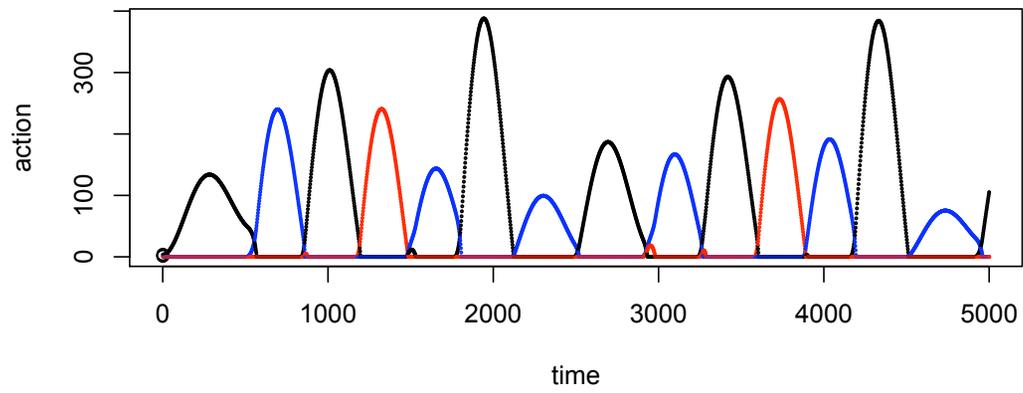
R in the classroom

- I. Undergraduate statistics and research methods
 - A. describe, pairs.panels, anova, lm
 - B. plot, curve, etc.
 - C. see tutorials for 205 and 371
 - D. simulations of data for simulated studies
 - E. Examples of complex models

Action Tendencies over time



Actions over time



R in the classroom

I. Graduate

A. data simulations

B. data analysis

C. longer tutorial