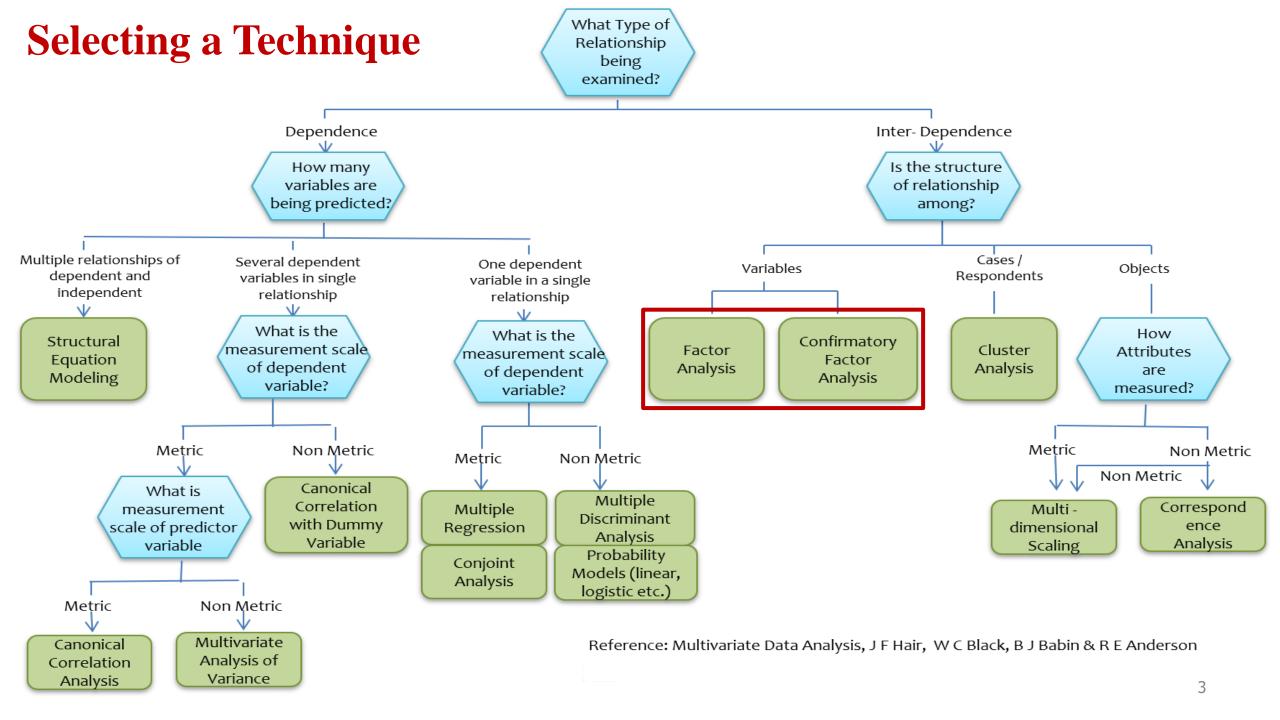
Multivariate Analysis

Exploratory Factor Analysis Using R

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Outline

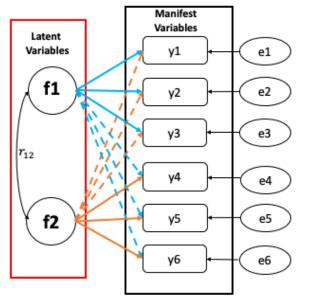
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 - Initial preparation and analysis
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 - Orthogonal rotation
 - Oblique rotation
 - Factor Score
 - o How to report factor analysis?



Factor Analysis & It's Types

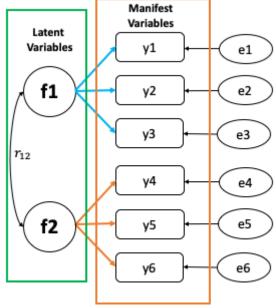
Factor Analysis A method for *modeling observed variables*, and their covariance structure, in terms of a smaller number of underlying unobservable (latent) "factors."

Exploratory



- EFA is used to discover the factor structure of a construct and examine its reliability.
- It is data driven.

Confirmatory



- CFA is used to confirm the fit of the hypothesized factor structure to observed (sample) data.
- It is theory driven.

Exploratory Factor Analysis (EFA): Definitions

- Factor: A factor is an underlying dimension that account for several observed variables.
- Factor-loadings: Values that explain how closely the variables are related to each one of the factors discovered. They are also known as factor-variable correlations.
 - o It is the absolute size (rather than the signs, plus or minus) of the loadings.
- Variance Types
 - o *Common variance* = overlapping variance between items (systematic variance)
 - *Unique variance* = variance only related to that item (error variance)
 - \circ *Communality* (h^2) the common variance for the item (how much of each variable is accounted for by the underlying factor)
 - We can think it as R² for that item (The estimate of variance in each variable that is explained by the factors)
 - e.g. For variable A, communality of 0.65 indicates that 65% of the variance in variable A is explained in the terms of factors

Variance of Variable 3

Variable Variable 3

Variable 4

 h^2 of the ith Variable = $(i^{th}$ factor loading of factor $A)^2 + (i^{th}$ factor loading of factor $B)^2 + ...$

- *Eigen value (or latent root):* The sum of squared values of factor loadings relating to a factor (referred to as Eigen Value)
 - o Eigen value indicates the relative importance of each factor in accounting for the particular set of variables being analyzed.

EFA Using R: The R anxiety questionnaire (RAQ)

SD	= Strongly Disagree, D = Disagree, N = Neither, A = Agree, SA =	Strongl	y Agree			
		SD	D	N	Α	SA
1	Statistics make me cry	0	0	0	0	0
2	My friends will think I'm stupid for not being able to cope with R	0	0	0	0	0
3	Standard deviations excite me	0	0	0	0	0
4	I dream that Pearson is attacking me with correlation coefficients	0	0	0	0	0
5	I don't understand statistics	0	0	0	0	0
6	I have little experience of computers	0	0	0	0	0
7	All computers hate me	0	0	0	0	0
8	I have never been good at mathematics	0	0	0	0	0
9	My friends are better at statistics than me	0	0	0	0	0
10	Computers are useful only for playing games	0	0	0	0	0
11	I did badly at mathematics at school	0	0	0	0	0
12	People try to tell you that R makes statistics easier to understand but it doesn't	0	0	0	0	0
13	I worry that I will cause irreparable damage because of my incompetence with computers	0	0	0	0	0
14	Computers have minds of their own and deliberately go wrong whenever I use them	0	0	0	0	0
15	Computers are out to get me	0	0	0	0	0
16	I weep openly at the mention of central tendency	0	0	0	0	0
17	I slip into a coma whenever I see an equation	0	0	0	0	0
18	R always crashes when I try to use it	0	0	0	0	0
19	Everybody looks at me when I use R	0	0	0	0	0
20	I can't sleep for thoughts of eigenvectors	0	0	0	0	0
21	I wake up under my duvet thinking that I am trapped under a normal distribution	0	0	0	0	0
22	My friends are better at R than I am	0	0	0	0	0
23	If I am good at statistics people will think I am a nerd	0	0	0	0	0

Data file → raq.dat

EFA Using R: Initial preparation and analysis

1. Correlation matrix:

- o Check the correlations between variables
 - There are essentially two potential problems:
 - 1. Correlations that are not high enough
 - 2. Correlations that are too high.

- o Compares an observed correlation matrix to the identity matrix
 - \circ H_0 : R Matrix = Identity Matrix (i.e., There is No Correlation Between Variables)
 - \circ H_1 : R Matrix \neq Identity Matrix (i.e., There is a Correlation Between Variables)

```
$chisq
[1] 19334.49
$p.value
[1] 0
$df
[1] 253
```

■ P-Value < 0.01; so, Bartlett's test is highly significant (i.e., *R*-matrix is not an identity matrix); $\gamma_2(253) = 19334.49$, p < .01, and therefore factor analysis is appropriate.

The Determinant of the *R*-matrix should be greater than 0.00001

> # Determinant of the R-matrix
> det(raqMatrix)
[1] 0.0005271037

EFA Using R: Initial preparation and analysis...

3. Sample Size:

KMO Test (Kaiser-Meyer-Olkin (KMO) Measure of Sampling Adequacy):

KMO \rightarrow The ratio of the squared correlation between variables to the squared partial correlation between variables.

- The KMO statistic varies between 0 and 1.
- KMO = 0 indicates diffusion in the pattern of correlations \rightarrow factor analysis is likely to be inappropriate
- KMO close to 1 indicates that patterns of correlations are relatively compact → factor analysis should yield distinct and reliable factors.

Kaiser-Meyer-Olkin factor adequacy
Call: KMO(r = raqData)
Overall MSA = 0.93
MSA for each item =
Q01 Q02 Q03 Q04 Q05 Q06 Q07 Q08 Q09 Q10 Q11 Q12 Q13 Q14 Q15 Q16 Q17 Q18
0.93 0.87 0.95 0.96 0.96 0.89 0.94 0.87 0.83 0.95 0.91 0.95 0.95 0.97 0.94 0.93 0.95
Q19 Q20 Q21 Q22 Q23
0.94 0.89 0.93 0.88 0.77

KMO Value	Level of Acceptance
Above 0.90	Superb
0.80 to 0.90	Great
0.70 to 0.80	Good
0.50 to 0.70	Mediocre
Below 0.50	Unacceptable

Source: Field [56].

- Because Overall KMO = 0.93 & values of KMO for all the variables greater than 0.5 so, sample size and the data are adequate for factor analysis
- If any variables with KMO values below .5 then you should consider excluding them from the analysis.

EFA Using R: Factor extraction

- Methods of Factor Extraction
 - PCA Method
 - Maximum Likelihood
 - Generalized Least Square
 - Alpha Factoring

EFA Using R: Factor extraction

Factor Loading Matrix

```
Principal Components Analysis
Call: principal(r = raqData, nfactors = 23, rotate = "none")
Standardized loadings (pattern matrix) based upon correlation matrix
                       PC5 PC6 PC7 PC8 PC9 PC10 PC11
                                                0.00 -0.13  0.20  0.24 -0.03  0.08  0.20 -0.14
                                                0.37 -0.22 -0.11 -0.21 -0.17 -0.15 -0.07
                                               -0.14 0.00 0.03 0.02 0.03 -0.02 0.07 -0.05
                                                0.41 0.15 0.09 -0.09 0.16 0.16 0.06 -0.12 -0.10 -0.04
                                                0.15  0.16 -0.19  0.12 -0.08  0.06 -0.22 -0.03
                                      0.03 0.33 0.02 0.21 0.04 0.17 0.07 0.05
                                      0.10 0.12 0.08 -0.02 0.04 0.03 -0.15 -0.04 -0.27 0.20 -0.03 -0.11
                            0.12 0.31 0.12 -0.41 -0.39 -0.19 -0.10 0.08 0.15 0.09 0.01
                                                          0.13 -0.01 -0.07 -0.12 -0.06 -0.03 0.05 -0.03
                  PC1 PC2 PC3 PC4 PC5 PC6 PC7 PC8 PC9 PC10 PC11 PC12 PC13 PC14 PC15 PC16 PC17 PC18 PC19 PC20 PC21 PC22 PC23
                  7.29 1.74 1.32 1.23 0.99 0.90 0.81 0.78 0.75 0.72 0.68 0.67 0.61 0.58 0.55 0.52 0.51 0.46 0.42 0.41 0.38 0.36 0.33
SS loadings
Proportion Var
Cumulative Var
                  0.32 0.39 0.45 0.50 0.55 0.59 0.62 0.65 0.69 0.72 0.75 0.78 0.80 0.83 0.85 0.88 0.90 0.92 0.94 0.95 0.97 0.99
Cumulative Proportion 0.32 0.39 0.45 0.50 0.55 0.59 0.62 0.65 0.69 0.72 0.75 0.78 0.80 0.83 0.85 0.88 0.90 0.92 0.94 0.95 0.97 0.99 1.00
Mean item complexity = 5
Test of the hypothesis that 23 components are sufficient.
The root mean square of the residuals (RMSR) is 0
with the empirical chi square 0 with prob < NA
Fit based upon off diagonal values = 1
```

PCA Method:

- By extracting the factors, inspect two columns, labelled
 - h2 (communalities) → All equal to 1 (Explained all of the variance in every variable)
 - Because Factor extracted 23 = number of variables
 - When we extract fewer factors (or components) we'll have lower communalities.
 - u2 (amount of unique variance for each variable) \rightarrow (1 communalities) \rightarrow all of the uniqueness's are 0

EFA Using R: Factor Extraction

Eigen Values:

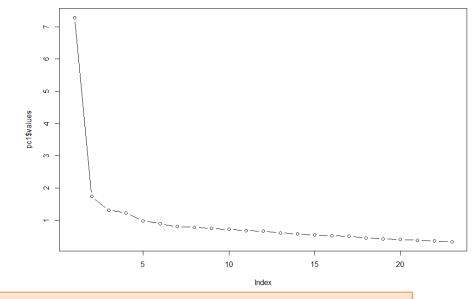
- o The eigenvalues associated with each factor represent the *variance explained by that particular linear component*.
- o **R** calls these SS loadings (sums of squared loadings)

• Factor 1 \rightarrow explains 7.29 units of variance out of a possible 23 (the number of factors) so as a proportion this is 7.29/23 = 0.22. The factor 1 applicant 220% of the total region as

0.32; so, factor 1 explains 32% of the total variance.

Factor Extraction Criteria:

- According to Kaiser's criterion (eigenvalues > 1) → We can pick four components (or factors)
- o By Jolliffe's criterion (retain factors with eigenvalues > 0.7) \rightarrow we can pick 10 factors,
- We should also consider the scree plot.



EFA Using R: Factor Extraction Tips

- To decide how many factors to extract:
 - o look at the eigenvalues and the scree plot.
- If number of variables < 30 then using eigenvalues greater than 1 is OK (Kaiser's criterion) as long as your communalities are all over .7
- Likewise, if your sample size exceeds 250 and the average of the communalities is .6 or greater then this is also fine.
- Alternatively, with 200 or more participants the scree plot can be used.

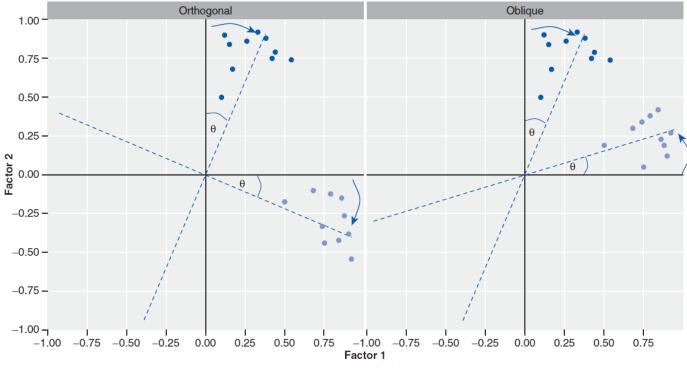
EFA Using R: Factor Rotation

After the factors extraction:

• Factor Loading: Calculate to what degree variables load on these factors (i.e., calculate the loading of the variable on each factor).

o Generally, most variables have high loadings on the most important factor and small loadings on all other

factors.



- Orthogonal rotation ensures that the factors remain independent or uncorrelated (perpendicular).
- Oblique rotation allow the factors to correlate (hence, do not perpendicular).

Rotation maximizes the loading of each variable on one of the extracted factors while minimizing the loading on all other factors.

EFA Using R: Factor Rotation → Orthogonal rotation (varimax)

- Having theoretical grounds to think that the factors are independent (unrelated) then
 - Choose one of the orthogonal rotations
 - Varimax (Recommended), quartimax, BentlerT & geominT
- pc3 <- principal(raqData, nfactors = 4, rotate = "varimax")</pre>
- print.psych(pc3, cut = 0.3, sort = TRUE)
 - i. Remove loadings that are below the cut point
 - ii. Reorders the items to try to put them into their factors, using the sort option

SD:	= Strongly Disagree, D = Disagree, N = Neither, A = Agree, SA =	Strongl	y Agree			
		SD	D	N	Α	SA
1	Statistics make me cry	0	0	0	0	0
2	My friends will think I'm stupid for not being able to cope with R	0	0	0	0	0
3	Standard deviations excite me	0	0	0	0	0
4	I dream that Pearson is attacking me with correlation coefficients	0	0	0	0	0
5	I don't understand statistics	0	0	0	0	0
6	I have little experience of computers	0	0	0	0	0
7	All computers hate me	0	0	0	0	0
8	I have never been good at mathematics	0	0	0	0	0
9	My friends are better at statistics than me	0	0	0	0	0
10	Computers are useful only for playing games	0	0	0	0	0
11	I did badly at mathematics at school	0	0	0	0	0
12	People try to tell you that R makes statistics easier to understand but it doesn't	0	0	0	0	0
13	I worry that I will cause irreparable damage because of my incompetence with computers	0	0	0	0	0
14	Computers have minds of their own and deliberately go wrong whenever I use them	0	0	0	0	0
15	Computers are out to get me	0	0	0	0	0
16	I weep openly at the mention of central tendency	0	0	0	0	0
17	I slip into a coma whenever I see an equation	0	0	0	0	0
18	R always crashes when I try to use it	0	0	0	0	0
19	Everybody looks at me when I use R	0	0	0	0	0
20	I can't sleep for thoughts of eigenvectors	0	0	0	0	0
21	I wake up under my duvet thinking that I am trapped under a normal distribution	0	0	0	0	0
22	My friends are better at R than I am	0	0	0	0	0
23	If I am good at statistics people will think I am a nerd	0	0	0	0	0

- All these questions related to using computers or R.
 - o we can label this factor fear of computers.

```
Principal Components Analysis
Call: principal(r = raqData, nfactors = 4, rotate = "varimax")
Standardized loadings (pattern matrix) based upon correlation matrix
                 RC1 RC4 RC2 h2 u2 com
           RC3
       6
         0.80
                                 0.65 0.35 1.0
     18
         0.68
               0.33
                                 0.60 0.40 1.5
      13 0.65
                                 0.54 0.46 1.6
          0.64
               0.33
                                 0.55 0.45 1.7
      14
         0.58
                                 0.49 0.51 1.8
     10 0.55
                                 0.33 0.67 1.2
     15 0.46
                                 0.38 0.62 2.6
                0.68
      ZU
                                 0.48 0.52 1.1
Q21
      21
                0.66
                                0.55 0.45 1.5
               -0.57
                            0.37 0.53 0.47 2.3
     12 0.47 0.52
                                0.51 0.49 2.1
         0.32 0.52 0.31
                                 0.47 0.53 2.4
      16 0.33 0.51 0.31
                                 0.49 0.51 2.6
Q01
                0.50 0.36
                                 0.43 0.57 2.4
         0.32 0.43
                                 0.34 0.66 2.5
                     0.83
                                 0.74 0.26 1.1
Q17
      17
                      0.75
                                0.68 0.32 1.5
     11
Q11
                      0.75
                                0.69 0.31 1.5
                            0.65 0.48 0.52 1.3
     22
Q22
                            0.65 0.46 0.54 1.2
      23
Q23
                            0.59 0.41 0.59 1.4
              -0.34
                            0.54 0.41 0.59 1.7
019
     19
               -0.37
                            0.43 0.34 0.66 2.2
                       RC3 RC1 RC4 RC2
                      3.73 3.34 2.55 1.95
SS loadings
Proportion Var
                      0.16 0.15 0.11 0.08
Cumulative Var
                      0.16 0.31 0.42 0.50
Proportion Explained 0.32 0.29 0.22 0.17
Cumulative Proportion 0.32 0.61 0.83 1.00
Mean item complexity = 1.8
Test of the hypothesis that 4 components are sufficient.
The root mean square of the residuals (RMSR) is 0.06
 with the empirical chi square 4006.15 with prob < 0
Fit based upon off diagonal values = 0.96
```

EFA Using R: Factor Rotation \rightarrow **Orthogonal rotation (varimax)**

\mathbf{c}						
SD = Strongly Disagree, D = Disagree, N = Neither, A = Agree, SA = Strongly Agree						
		SD	D	N	Α	SA
1	Statistics make me cry	0	0	0	0	0
2	My friends will think I'm stupid for not being able to cope with R	0	0	0	0	0
3	Standard deviations excite me	0	0	0	0	0
4	I dream that Pearson is attacking me with correlation coefficients	0	0	0	0	0
5	I don't understand statistics	0	0	0	0	0
6	I have little experience of computers	0	0	0	0	0
7	All computers hate me	0	0	0	0	0
8	I have never been good at mathematics	0	0	0	0	0
9	My friends are better at statistics than me	0	0	0	0	0
10	Computers are useful only for playing games	0	0	0	0	0
11	I did badly at mathematics at school	0	0	0	0	0
12	People try to tell you that R makes statistics easier to understand but it doesn't	0	0	0	0	0
13	I worry that I will cause irreparable damage because of my incompetence with computers	0	0	0	0	0
14	Computers have minds of their own and deliberately go wrong whenever I use them	0	0	0	0	0
15	Computers are out to get me	0	0	0	0	0
16	I weep openly at the mention of central tendency	0	0	0	0	0
17	I slip into a coma whenever I see an equation	0	0	0	0	0
18	R always crashes when I try to use it	0	0	0	0	0
19	Everybody looks at me when I use R	0	0	0	0	0
20	I can't sleep for thoughts of eigenvectors	0	0	0	0	0
21	I wake up under my duvet thinking that I am trapped under a normal distribution	0	0	0	0	0
22	My friends are better at R than I am	0	0	0	0	0
23	If I am good at statistics people will think I am a nerd	0	0	0	0	0

```
Principal Components Analysis
Call: principal(r = ragData, nfactors = 4, rotate = "varimax")
Standardized loadings (pattern matrix) based upon correlation matrix
    item RC3 RC1 RC4 RC2 h2 u2 com
      6 0.80
                               0.65 0.35 1.0
     18 0.68 0.33
                               0.60 0.40 1.5
     13 0.65
                               0.54 0.46 1.6
         0.64 0.33
                               0.55 0.45 1.7
     14 0.58 0.36
                               0.49 0.51 1.8
     10 0.55
                               0.33 0.67 1.2
015 15 0 46
                               0.38 0.62 2.6
               0.68
                               0.48 0.52 1.1
     21
               0.66
                               0.55 0.45 1.5
              -0.57
                          0.37 0.53 0.47 2.3
     12 0.47 0.52
                               0.51 0.49 2.1
      4 0.32 0.52 0.31
                               0.47 0.53 2.4
     16 0.33 0.51 0.31
                               0.49 0.51 2.6
               0.50 0.36
                               0.43 0.57 2.4
      5 0.32 0.43
                               0.34 0.66 2.5
                     0.83
                               0.74 0.26 1.1
                     0.75
                               0.68 0.32 1.5
011 11
                     0.75
                               0.69 0.31 1.5
                           0.65 0.48 0.52 1.3
                          0.65 0.46 0.54 1.2
                          0.59 0.41 0.59 1.4
```

- All these questions related to different aspects of statistics.
 - o we can label this factor *fear of Statistics*.

```
RC3 RC1 RC4 RC2

SS loadings 3.73 3.34 2.55 1.95

Proportion Var 0.16 0.15 0.11 0.08

Cumulative Var 0.16 0.31 0.42 0.50

Proportion Explained 0.32 0.29 0.22 0.17

Cumulative Proportion 0.32 0.61 0.83 1.00

Mean item complexity = 1.8

Test of the hypothesis that 4 components are sufficient.

The root mean square of the residuals (RMSR) is 0.06

with the empirical chi square 4006.15 with prob < 0

Fit based upon off diagonal values = 0.96
```

0.54 0.41 0.59 1.7

0.43 0.34 0.66 2.2

EFA Using R: Factor Rotation \rightarrow Orthogonal rotation (varimax)

SD = Strongly Disagree, D = Disagree, N = Neither, A = Agree, SA = Strongly Agree						
		SD	D	N	Α	SA
1	Statistics make me cry	0	0	0	0	0
2	My friends will think I'm stupid for not being able to cope with R	0	0	0	0	0
3	Standard deviations excite me	0	0	0	0	0
4	I dream that Pearson is attacking me with correlation coefficients	0	0	0	0	0
5	I don't understand statistics	0	0	0	0	0
6	I have little experience of computers	0	0	0	0	0
7	All computers hate me	0	0	0	0	0
8	I have never been good at mathematics	0	0	0	0	0
9	My friends are better at statistics than me	0	0	0	0	0
10	Computers are useful only for playing games	0	0	0	0	0
11	I did badly at mathematics at school	0	0	0	0	0
12	People try to tell you that R makes statistics easier to understand but it doesn't	0	0	0	0	0
13	I worry that I will cause irreparable damage because of my incompetence with computers	0	0	0	0	0
14	Computers have minds of their own and deliberately go wrong whenever I use them	0	0	0	0	0
15	Computers are out to get me	0	0	0	0	0
16	I weep openly at the mention of central tendency	0	0	0	0	0
17	I slip into a coma whenever I see an equation	0	0	0	0	0
18	R always crashes when I try to use it	0	0	0	0	0
19	Everybody looks at me when I use R	0	0	0	0	0
20	I can't sleep for thoughts of eigenvectors	0	0	0	0	0
21	I wake up under my duvet thinking that I am trapped under a normal distribution	0	0	0	0	0
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```
Principal Components Analysis
                               Call: principal(r = ragData, nfactors = 4, rotate = "varimax")
                               Standardized loadings (pattern matrix) based upon correlation matrix
                                   item RC3 RC1 RC4 RC2 h2 u2 com
                                     6 0.80
                                                               0.65 0.35 1.0
                                    18 0.68 0.33
                                                               0.60 0.40 1.5
                                    13 0.65
                                                               0.54 0.46 1.6
                                        0.64 0.33
                                                               0.55 0.45 1.7
                                     14 0.58 0.36
                                                               0.49 0.51 1.8
                                    10 0.55
                                                               0.33 0.67 1.2
                                    15 0.46
                                                              0.38 0.62 2.6
                                    20
                                              0.68
                                                               0.48 0.52 1.1
                                     21
                                              0.66
                                                               0.55 0.45 1.5
                                             -0.57
                                                          0.37 0.53 0.47 2.3
                                    12 0.47 0.52
                                                               0.51 0.49 2.1
                                        0.32 0.52 0.31
                                                               0.47 0.53 2.4
                                    16 0.33 0.51 0.31
                                                               0.49 0.51 2.6
                                              0.50 0.36
                                                               0.43 0.57 2.4
                                     5 0.32 0.43
                                                               0.34 0.66 2.5
                                                    0.83
                                                               0.74 0.26 1.1
                                    17
                                                    0.75
                                                               0.68 0.32 1.5
                               011 11
                                                    0.75
                                                               0.69 0.31 1.5
                                                          v.65 0.48 0.52 1.3
                               Q22
                                    22
                                                          0.65 0.46 0.54 1.2
                                    23
                               Q23
                                                          0.59 0.41 0.59 1.4
                                             -0.34
                                                          0.54 0.41 0.59 1.7
                                                         0.43 0.34 0.66 2.2

    All these questions related to mathematics.

                                                          RC1 RC4 RC2
    o Label this factor fear of Mathematics.
                                                          .34 2.55 1.95
                                                    0.10 0.15 0.11 0.08
                               Cumulative Var
                                                    0.16 0.31 0.42 0.50
                              Proportion Explained 0.32 0.29 0.22 0.17
                               Cumulative Proportion 0.32 0.61 0.83 1.00
                               Mean item complexity = 1.8
                               Test of the hypothesis that 4 components are sufficient.
                               The root mean square of the residuals (RMSR) is 0.06
                                with the empirical chi square 4006.15 with prob < 0
                              Fit based upon off diagonal values = 0.96
```

EFA Using R: Factor Rotation \rightarrow **Orthogonal rotation (varimax)**

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3	Standard deviations excite me	0	0	0	0	0
4	I dream that Pearson is attacking me with correlation coefficients	0	0	0	0	0
5	I don't understand statistics	0	0	0	0	0
6	I have little experience of computers	0	0	0	0	0
7	All computers hate me	0	0	0	0	0
8	I have never been good at mathematics	0	0	0	0	0
9	My friends are better at statistics than me	0	0	0	0	0
10	Computers are useful only for playing games	0	0	0	0	0
11	I did badly at mathematics at school	0	0	0	0	0
12	People try to tell you that R makes statistics easier to understand but it doesn't	0	0	0	0	0
13	I worry that I will cause irreparable damage because of my incompetence with computers	0	0	0	0	0
14	Computers have minds of their own and deliberately go wrong whenever I use them	0	0	0	0	0
15	Computers are out to get me	0	0	0	0	0
16	I weep openly at the mention of central tendency	0	0	0	0	0
17	I slip into a coma whenever I see an equation	0	0	0	0	0
18	R always crashes when I try to use it	0	0	0	0	0
19	Everybody looks at me when I use R	0	0	0	0	0
20	I can't sleep for thoughts of eigenvectors	0	0	0	0	0
21	I wake up under my duvet thinking that I am trapped under a normal distribution	0	0	0	0	0
22	My friends are better at R than I am	0	0	0	0	0
23	If I am good at statistics people will think I am a nerd	0	0	0	0	0

```
Principal Components Analysis
                           Call: principal(r = ragData, nfactors = 4, rotate = "varimax")
                           Standardized loadings (pattern matrix) based upon correlation matrix
                               item RC3 RC1 RC4 RC2 h2 u2 com
                                 6 0.80
                                                          0.65 0.35 1.0
                                18 0.68 0.33
                                                          0.60 0.40 1.5
                                13 0.65
                                                          0.54 0.46 1.6
                                    0.64 0.33
                                                          0.55 0.45 1.7
                                14 0.58 0.36
                                                          0.49 0.51 1.8
                                10 0.55
                                                          0.33 0.67 1.2
                                                          0.38 0.62 2.6
                                15 0.46
                                20
                                          0.68
                                                          0.48 0.52 1.1
                                                          0.55 0.45 1.5
                                         -0.57
                                                      0.37 0.53 0.47 2.3
                                                           0.51 0.49 2.1
                                 4 0.32 0.52 0.31
                                                          0.47 0.53 2.4
                                 16 0.33 0.51 0.31
                                                          0.49 0.51 2.6
                                          0.50 0.36
                                                          0.43 0.57 2.4
                                  5 0.32 0.43
                                                           0.34 0.66 2.5
                                                          0.74 0.26 1.1
All these items contain some component of
                                                          0.68 0.32 1.5
social evaluation from friends.
                                                          0.69 0.31 1.5
                                                      0.65 0.48 0.52 1.3
• Label this factor peer evaluation.
                                                      0.65 0.46 0.54 1.2
                                                      0.59 0.41 0.59 1.4
                           Q02
                                         -0.34
                                                      0.54 0.41 0.59 1.7
                           019
                                19
                                         -0.37
                                                      0.43 0.34 0.66 2.2
                                                 RC3 RC1 RC4 RC2
                           SS loadings
                                                3.73 3.34 2.55 1.95
                                                0.16 0.15 0.11 0.08
                           Proportion Var
                           Cumulative Var
                                                0.16 0.31 0.42 0.50
                          Proportion Explained 0.32 0.29 0.22 0.17
                           Cumulative Proportion 0.32 0.61 0.83 1.00
                           Mean item complexity = 1.8
                           Test of the hypothesis that 4 components are sufficient.
                           The root mean square of the residuals (RMSR) is 0.06
                            with the empirical chi square 4006.15 with prob < 0
                          Fit based upon off diagonal values = 0.96
```

EFA Using R: Factor Rotation \rightarrow oblique rotation (oblimin)

- Having theoretical grounds to think that the factors are related then
 - Choose one of the orthogonal rotations
 - **oblimin**, promax, simplimax, BentlerQ and geominQ

```
Principal Components Analysis
Call: principal(r = raqData, nfactors = 4, rotate = "oblimin")
Standardized loadings (pattern matrix) based upon correlation matrix
                TC4 TC3
                            TC2
          TC1
                                  h2
                                 0.65 0.35 1.1
                                 0.60 0.40 1.1
         0.64
                                 0.55 0.45 1.2
     13 0.64
     10 0.57
                                 0.33 0.67 1.2
     14 0.57
                                 0.49 0.51 1.3
     12 0.45
                      0.43
                                 0.51 0.49 2.0
     15 0.40
                                 0.38 0.62 1.9
                0.90
                                 0.74 0.26 1.0
Q11
     11
                0.78
                                 0.69 0.31 1.0
     17
Q17
                0.78
Q20
                      0.71
                                 0.48 0.52 1.1
     21
Q21
                      0.60
                     -0.51
                                 0.53 0.47 1.8
Q04
                      0.41
                                 0.47 0.53 2.6
Q16
     16
                      0.41
                                 0.49 0.51 2.4
Q01
                0.33 0.40
                                 0.43 0.57 2.4
                      0.34
                                 0.34 0.66 2.7
Q22
     22
                            0.65 0.46 0.54 1.2
                            0.63 0.48 0.52 1.4
Q23
     23
                            0.61 0.41 0.59 1.6
Q02
                     -0.36 0.51 0.41 0.59 1.9
019
     19
                     -0.35 0.38 0.34 0.66 2.1
                       TC1 TC4 TC3 TC2
SS loadings
Proportion Var
                      0.17 0.13 0.13 0.08
Cumulative Var
                      0.17 0.29 0.42 0.50
Proportion Explained 0.34 0.25 0.25 0.16
Cumulative Proportion 0.34 0.59 0.84 1.00
```

In our Study:

All of our factors related to fear; It's likely that these will be correlated: people with fear of one thing might have fear of other things.

The **same four factors** seem to have emerged although they are in a different order.

- Factor 1 seems to represent fear of computers,
- factor 2 represents fear of peer evaluation,
- factor 3 represents fear of statistics and
- factor 4 represents fear of mathematics.

```
With component correlations of

TC1 TC4 TC3 TC2

TC1 1.00 0.44 0.36 -0.18

TC4 0.44 1.00 0.31 -0.10

TC3 0.36 0.31 1.00 -0.17

TC2 -0.18 -0.10 -0.17 1.00
```

 These correlations exist tell us that the constructs measured can be interrelated

EFA Using R: Factor Score

```
# Factor Score
'-----'
pc5 <- principal(raqData, nfactors = 4, rotate = "oblimin", scores = TRUE)
pc5$scores
raqData <- cbind(raqData, pc5$scores)
raqData
```

- Factor scores can be used in this way to assess the relative fear of one person compared to another.
- After saving Factor score they can be further used as
 - Independent variable
 - Dependent variable
- Used in ANOVA / MANOVA / Regression Analysis
- Used in Cluster Analysis

How to report factor analysis?

Summary of exploratory factor analysis results for the **R** anxiety questionnaire (N = 2571)

	Varimax rotated factor loadings				
	Fear of	Fear of	Peer	Fear of	
Item	computers	statistics	evaluation	maths	
I have little experience of computers	.80	01	07	.10	
R always crashes when I try to use it	.68	.33	08	.13	
I worry that I will cause irreparable damage because of my incompetence with computers	.65	.23	10	.23	
All computers hate me	.64	.33	08	.16	
Computers have minds of their own and deliberately go wrong whenever I use them	.58	.36	07	.14	
Computers are useful only for playing games	.55	.00	12	.13	
Computers are out to get me	.46	.22	19	.29	
I can't sleep for thoughts of eigen vectors	04	.68	14	.08	
I wake up under my duvet thinking that I am trapped under a normal distribution	.29	.66	07	.16	
Standard deviations excite me	20	57	.37	18	
People try to tell you that R makes statistics easier to understand but it doesn't	.47	.52	08	.10	
I dream that Pearson is attacking me with correlation coefficients	.32	.52	.04	.31	
I weep openly at the mention of central tendency	.33	.51	12	.31	
Statistics makes me cry	.24	.50	.06	.36	
I don't understand statistics	.32	.43	.02	.24	
I have never been good at mathematics	.13	.17	.01	.83	
I slip into a coma whenever I see an equation	.27	.22	04	.75	
I did badly at mathematics at school	.26	.21	14	.75	
My friends are better at statistics than me	09	20	.65	.12	
My friends are better at R than I am	19	.03	.65	10	
If I'm good at statistics my friends will think I'm a nerd	02	.17	.59	20	
My friends will think I'm stupid for not being able to cope with R	01	34	.54	.07	
Everybody looks at me when I use R	15	37	.43	03	
Eigenvalues	3.73	3.34	1.95	2.55	
% of variance	16.22	14.52	8.48	11.10	
α	.82	.82	.57	.82	

Note: Factor loadings over .40 appear in bold.

QUESTIONS

