

# Principal Components and Factor Analysis

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## Initial Principal Components and Factor Analysis of GCA data.

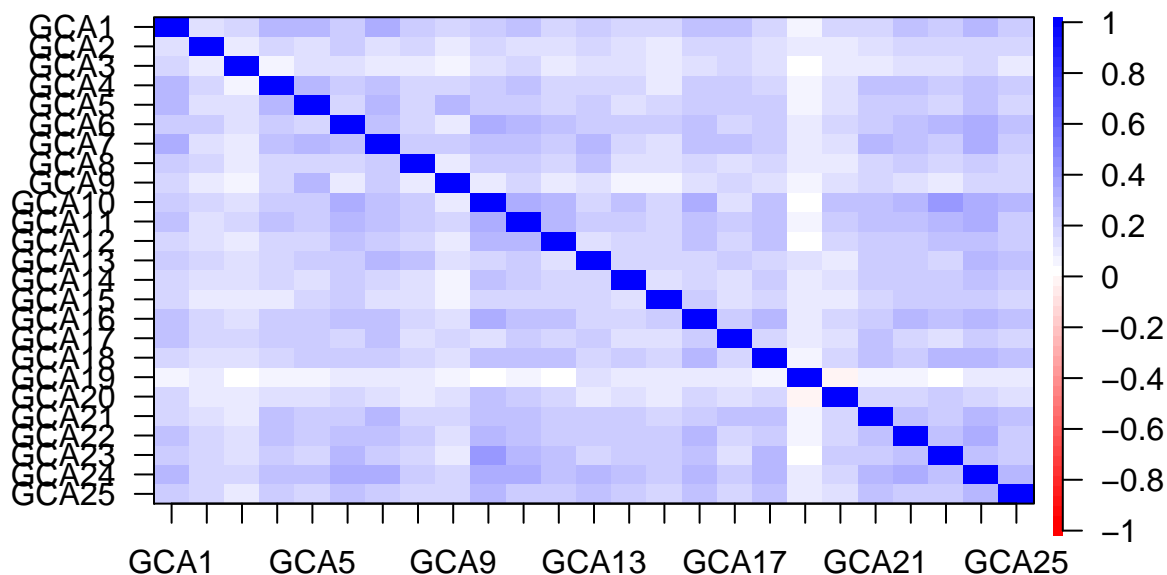
The X1516GCA\_FA data set is all Fall 15 and Spring 16 pre and post GCA scores (2346 case).

```
X1516GCA_FA <- read.csv(file = "1516GCA.csv", header = TRUE)
```

First look at the correlaton matrix for GCA items:

```
library(psych)
library(GPArotation)
corPlot(X1516GCA_FA)
```

## Correlation plot

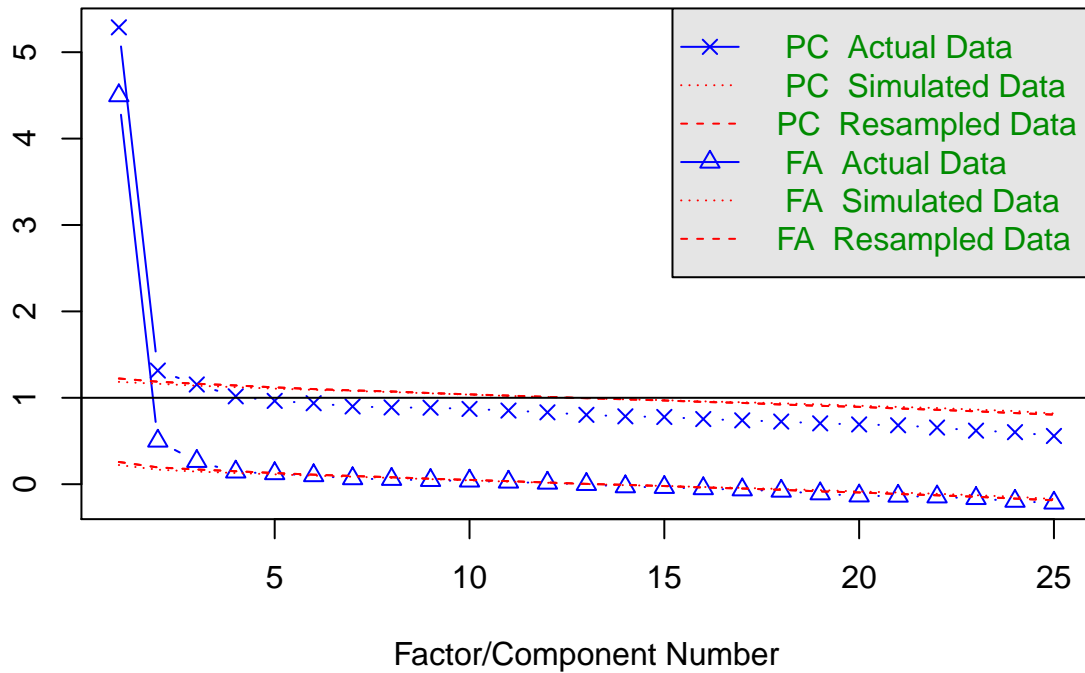


One way to determine the number of factors is to compare the solution to a set of simulated random data with properties similar to the GCA data set (a parallel analysis). Running this parallel analysis also produces the scree plot:

```
fa.parallel((X1516GCA_FA))
```

eigenvalues of principal components and factor analysis

## Parallel Analysis Scree Plots



## Parallel analysis suggests that the number of factors = 6 and the number of components = 3

The parallel analysis suggest 3 components and 5 factors and the scree plot shows 3-4 components with eigenvalue > 1, so run PCA (descriptive model) with 3 factors, varimax rotation

```
principal(X1516GCA_FA, nfactors=3, rotate = "varimax")
```

## Principal Components Analysis

## Call: principal(r = X1516GCA\_FA, nfactors = 3, rotate = "varimax")

## Standardized loadings (pattern matrix) based upon correlation matrix

	RC1	RC2	RC3	h2	u2	com
## GCA1	0.21	0.52	0.13	0.332	0.67	1.5
## GCA2	0.21	0.12	0.34	0.175	0.83	1.9
## GCA3	0.18	0.22	0.03	0.081	0.92	1.9
## GCA4	0.19	0.51	0.08	0.301	0.70	1.3
## GCA5	0.13	0.63	0.02	0.410	0.59	1.1
## GCA6	0.49	0.14	0.27	0.334	0.67	1.7
## GCA7	0.19	0.53	0.23	0.370	0.63	1.6
## GCA8	0.24	0.17	0.37	0.225	0.78	2.2
## GCA9	-0.05	0.61	-0.05	0.378	0.62	1.0
## GCA10	0.68	0.12	0.00	0.477	0.52	1.1
## GCA11	0.48	0.32	0.01	0.335	0.66	1.7
## GCA12	0.51	0.13	0.04	0.278	0.72	1.1
## GCA13	0.16	0.33	0.43	0.316	0.68	2.2
## GCA14	0.42	0.06	0.22	0.225	0.77	1.5
## GCA15	0.38	0.03	0.24	0.199	0.80	1.7
## GCA16	0.46	0.22	0.20	0.296	0.70	1.9

```

## GCA17  0.09  0.51  0.17  0.292  0.71  1.3
## GCA18  0.47  0.17  0.16  0.272  0.73  1.5
## GCA19 -0.12 -0.04  0.73  0.552  0.45  1.1
## GCA20  0.43  0.20 -0.26  0.299  0.70  2.1
## GCA21  0.29  0.41  0.10  0.262  0.74  1.9
## GCA22  0.42  0.25  0.20  0.280  0.72  2.1
## GCA23  0.64  0.08 -0.01  0.422  0.58  1.0
## GCA24  0.43  0.34  0.27  0.372  0.63  2.6
## GCA25  0.34  0.21  0.33  0.274  0.73  2.7
##
##
##          RC1  RC2  RC3
## SS loadings      3.43 2.70 1.63
## Proportion Var    0.14 0.11 0.07
## Cumulative Var    0.14 0.25 0.31
## Proportion Explained 0.44 0.35 0.21
## Cumulative Proportion 0.44 0.79 1.00
##
## Mean item complexity = 1.7
## Test of the hypothesis that 3 components are sufficient.
##
## The root mean square of the residuals (RMSR) is 0.05
## with the empirical chi square 4018.23 with prob < 0
##
## Fit based upon off diagonal values = 0.91

```

Now compare to PCA with 4 factors:

```
principal(X1516GCA_FA, nfactors=4, rotate = "varimax")
```

```

## Principal Components Analysis
## Call: principal(r = X1516GCA_FA, nfactors = 4, rotate = "varimax")
## Standardized loadings (pattern matrix) based upon correlation matrix
##          RC1  RC2  RC3  RC4  h2  u2 com
## GCA1  0.20  0.46  0.11  0.29  0.35  0.65 2.3
## GCA2  0.23  0.11  0.33  0.05  0.18  0.82 2.1
## GCA3  0.10 -0.01 -0.04  0.86  0.75  0.25 1.0
## GCA4  0.22  0.54  0.08 -0.05  0.35  0.65 1.4
## GCA5  0.14  0.61  0.02  0.13  0.41  0.59 1.2
## GCA6  0.51  0.12  0.25  0.06  0.34  0.66 1.6
## GCA7  0.20  0.52  0.22  0.12  0.37  0.63 1.8
## GCA8  0.25  0.13  0.35  0.17  0.23  0.77 2.6
## GCA9 -0.02  0.66 -0.03 -0.11  0.45  0.55 1.1
## GCA10 0.68  0.10 -0.03  0.06  0.48  0.52 1.1
## GCA11 0.47  0.27 -0.02  0.22  0.34  0.66 2.0
## GCA12 0.52  0.13  0.02  0.00  0.29  0.71 1.1
## GCA13 0.16  0.25  0.40  0.32  0.35  0.65 3.0
## GCA14 0.41  0.01  0.19  0.18  0.24  0.76 1.8
## GCA15 0.39  0.01  0.22  0.04  0.20  0.80 1.6
## GCA16 0.47  0.21  0.18  0.07  0.30  0.70 1.8
## GCA17 0.07  0.42  0.14  0.39  0.35  0.65 2.3
## GCA18 0.48  0.16  0.13  0.04  0.28  0.72 1.4
## GCA19 -0.08 -0.01  0.75 -0.10  0.57  0.43 1.1
## GCA20 0.43  0.20 -0.28  0.02  0.30  0.70 2.2
## GCA21 0.30  0.42  0.09  0.04  0.27  0.73 2.0
## GCA22 0.43  0.23  0.18  0.11  0.28  0.72 2.0

```

```
## GCA23  0.65  0.07 -0.04  0.02  0.43  0.57  1.0
## GCA24  0.44  0.30  0.24  0.16  0.37  0.63  2.7
## GCA25  0.37  0.23  0.32 -0.04  0.30  0.70  2.7
##
##              RC1  RC2  RC3  RC4
## SS loadings      3.51 2.46 1.52 1.29
## Proportion Var    0.14 0.10 0.06 0.05
## Cumulative Var    0.14 0.24 0.30 0.35
## Proportion Explained 0.40 0.28 0.17 0.15
## Cumulative Proportion 0.40 0.68 0.85 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 4496.83 with prob < 0
##
## Fit based upon off diagonal values = 0.9
```

4 factors only accounts for 4% more variance than the 3 factor PCA. But look at communalities for GCA3 in each solution (0.081 in 3 factor, 0.75 in 4 factor)?

Now compare the PCAs to a factor analysis (structural model) specifying 3 factors, varimax rotation, do not impute values for missing, use minimum residual factoring method (default) and view loading matrix

```
fa(X1516GCA_FA, nfactors = 3, rotate = "varimax")
```

```
## Factor Analysis using method = minres
## Call: fa(r = X1516GCA_FA, nfactors = 3, rotate = "varimax")
## Standardized loadings (pattern matrix) based upon correlation matrix
##      MR1 MR2 MR3  h2  u2 com
## GCA1  0.23 0.43 0.14 0.257 0.74 1.8
## GCA2  0.21 0.15 0.19 0.102 0.90 2.8
## GCA3  0.17 0.16 0.06 0.059 0.94 2.3
## GCA4  0.23 0.40 0.08 0.221 0.78 1.7
## GCA5  0.18 0.51 0.04 0.292 0.71 1.2
## GCA6  0.44 0.18 0.21 0.266 0.73 1.8
## GCA7  0.23 0.44 0.21 0.291 0.71 2.0
## GCA8  0.24 0.19 0.23 0.148 0.85 2.9
## GCA9  0.06 0.40 0.02 0.165 0.83 1.0
## GCA10 0.62 0.14 0.00 0.405 0.60 1.1
## GCA11 0.43 0.29 0.05 0.272 0.73 1.8
## GCA12 0.41 0.17 0.06 0.200 0.80 1.4
## GCA13 0.21 0.28 0.31 0.222 0.78 2.7
## GCA14 0.34 0.12 0.16 0.156 0.84 1.7
## GCA15 0.30 0.11 0.15 0.124 0.88 1.8
## GCA16 0.41 0.23 0.16 0.247 0.75 1.9
## GCA17 0.15 0.38 0.16 0.191 0.81 1.7
## GCA18 0.40 0.19 0.14 0.216 0.78 1.7
## GCA19 -0.02 0.03 0.37 0.142 0.86 1.0
## GCA20 0.32 0.18 -0.10 0.145 0.86 1.7
## GCA21 0.29 0.34 0.12 0.209 0.79 2.2
## GCA22 0.38 0.25 0.17 0.235 0.76 2.2
## GCA23 0.55 0.12 0.00 0.322 0.68 1.1
## GCA24 0.42 0.31 0.24 0.333 0.67 2.5
## GCA25 0.33 0.22 0.25 0.219 0.78 2.7
```

```
##
##              MR1  MR2  MR3
## SS loadings      2.77 1.92 0.74
## Proportion Var    0.11 0.08 0.03
## Cumulative Var    0.11 0.19 0.22
## Proportion Explained 0.51 0.35 0.14
## Cumulative Proportion 0.51 0.86 1.00
##
## Mean item complexity = 1.9
## Test of the hypothesis that 3 factors are sufficient.
##
## The degrees of freedom for the null model are 300 and the objective function was 3.54 with Chi Sq
## The degrees of freedom for the model are 228 and the objective function was 0.17
##
## The root mean square of the residuals (RMSR) is 0.02
## The df corrected root mean square of the residuals is 0.02
##
## The harmonic number of observations is 2056 with the empirical chi square 422.44 with prob < 8.8
## The total number of observations was 2346 with Likelihood Chi Square = 401.54 with prob < 1e-11
##
## Tucker Lewis Index of factoring reliability = 0.971
## RMSEA index = 0.018 and the 90 % confidence intervals are 0.015 0.021
## BIC = -1367.85
## Fit based upon off diagonal values = 0.99
## Measures of factor score adequacy
##
##              MR1  MR2  MR3
## Correlation of scores with factors      0.82 0.75 0.60
## Multiple R square of scores with factors      0.68 0.56 0.36
## Minimum correlation of possible factor scores 0.36 0.12 -0.29
```

And finally run the FA with 4 factors:

```
fa(X1516GCA_FA, nfactors = 4, rotate = "varimax")
```

```
## Factor Analysis using method = minres
## Call: fa(r = X1516GCA_FA, nfactors = 4, rotate = "varimax")
## Standardized loadings (pattern matrix) based upon correlation matrix
##      MR1  MR2  MR3  MR4  h2  u2 com
## GCA1  0.24 0.43 0.17 -0.07 0.278 0.72 2.0
## GCA2  0.21 0.14 0.19 0.00 0.103 0.90 2.8
## GCA3  0.17 0.17 0.08 -0.05 0.066 0.93 2.6
## GCA4  0.23 0.40 0.10 0.03 0.220 0.78 1.7
## GCA5  0.18 0.50 0.04 0.06 0.291 0.71 1.3
## GCA6  0.44 0.17 0.23 0.01 0.269 0.73 1.8
## GCA7  0.23 0.44 0.22 0.03 0.292 0.71 2.0
## GCA8  0.25 0.19 0.25 -0.05 0.161 0.84 2.9
## GCA9  0.04 0.41 -0.01 0.18 0.202 0.80 1.4
## GCA10 0.61 0.13 0.01 0.11 0.407 0.59 1.2
## GCA11 0.43 0.29 0.07 0.03 0.273 0.73 1.8
## GCA12 0.41 0.16 0.06 0.09 0.202 0.80 1.5
## GCA13 0.20 0.28 0.32 0.03 0.224 0.78 2.7
## GCA14 0.33 0.11 0.16 0.10 0.157 0.84 1.9
## GCA15 0.30 0.10 0.16 0.01 0.125 0.87 1.8
## GCA16 0.40 0.22 0.16 0.11 0.249 0.75 2.1
## GCA17 0.16 0.38 0.17 0.00 0.194 0.81 1.8
```

```

## GCA18  0.39 0.18  0.13  0.14 0.223 0.78 2.0
## GCA19 -0.03 0.03  0.36  0.05 0.131 0.87 1.1
## GCA20  0.33 0.18 -0.08 -0.01 0.146 0.85 1.7
## GCA21  0.27 0.33  0.11  0.19 0.228 0.77 2.8
## GCA22  0.38 0.25  0.19 -0.03 0.245 0.75 2.2
## GCA23  0.56 0.12  0.01 -0.02 0.332 0.67 1.1
## GCA24  0.41 0.30  0.25  0.12 0.333 0.67 2.8
## GCA25  0.30 0.18  0.26  0.35 0.313 0.69 3.4
##
##
##          MR1  MR2  MR3  MR4
## SS loadings      2.72 1.86 0.79 0.29
## Proportion Var    0.11 0.07 0.03 0.01
## Cumulative Var    0.11 0.18 0.21 0.23
## Proportion Explained 0.48 0.33 0.14 0.05
## Cumulative Proportion 0.48 0.81 0.95 1.00
##
## Mean item complexity = 2
## Test of the hypothesis that 4 factors are sufficient.
##
## The degrees of freedom for the null model are 300 and the objective function was 3.54 with Chi Sq
## The degrees of freedom for the model are 206 and the objective function was 0.15
##
## The root mean square of the residuals (RMSR) is 0.02
## The df corrected root mean square of the residuals is 0.02
##
## The harmonic number of observations is 2056 with the empirical chi square 355.87 with prob < 4.3
## The total number of observations was 2346 with Likelihood Chi Square = 345.33 with prob < 4e-09
##
## Tucker Lewis Index of factoring reliability = 0.975
## RMSEA index = 0.017 and the 90 % confidence intervals are 0.014 0.02
## BIC = -1253.32
## Fit based upon off diagonal values = 0.99
## Measures of factor score adequacy
##
##          MR1  MR2  MR3  MR4
## Correlation of scores with factors    0.82 0.75 0.60 0.48
## Multiple R square of scores with factors    0.67 0.56 0.36 0.23
## Minimum correlation of possible factor scores 0.35 0.11 -0.27 -0.53

```

The 4 factor FA only accounts for 1% more variance than the 3 factor FA, and is likely harder to interpret. Also the communalities (variance accounted for in each item by all factors in the solution) does not increase much for 4 factors versus 3.