

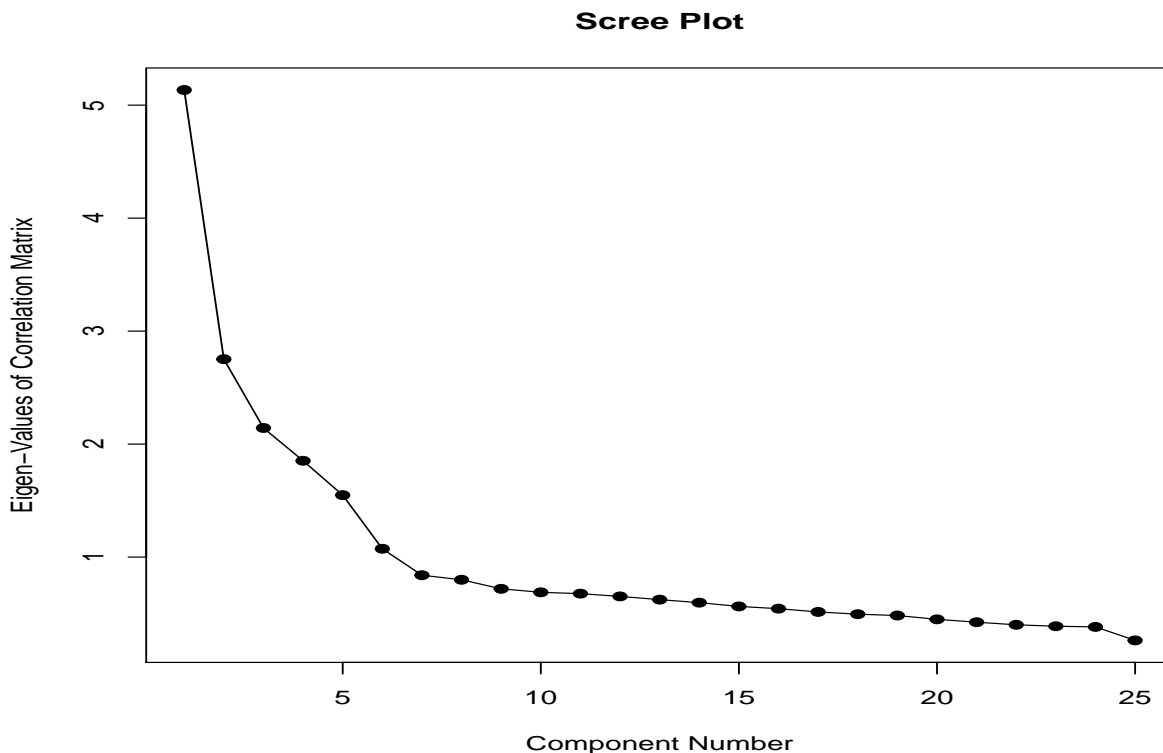
Illustration of Factor Analysis

Example 1: The first example involves the `bfi` data, pertaining to responses to 25 Personality Test Items taken by 2436 (probably neurotic) individuals, which we have already studied for PCA. Recall that these 25 items are supposed to belong to 5 broad dimensions measuring **Agreeability** (items -A1, A2, A3, A4 & A5), **Conscientious** (items C1, C2, C3, -C4 & -C5), **Extraversion** (items -E1, -E2, E3, E4 & E5), **Neuroticism** (items N1, N2, N3, N4 & N5) and **Openness** (items O1, -O2, O3, O4 & -O5). We begin where we left off in the PCA handout with this data set's PCA.

```
> pca.bfi<-principal(X,nfactors=25) #library(psych)

> pca.bfi$values
 [1] 5.1343112 2.7518867 2.1427020 1.8523276 1.5481628 1.0735825 0.8395389
 [8] 0.7992062 0.7189892 0.6880888 0.6763734 0.6517998 0.6232530 0.5965628
[15] 0.5630908 0.5433053 0.5145175 0.4945031 0.4826395 0.4489210 0.4233661
[22] 0.4006715 0.3878045 0.3818568 0.2625390

> plot(1:25,pca.bfi$values,pch=19,xlab="Component Number",ylab="Eigen-Values
+ of Correlation Matrix",main="Scree Plot")
> lines(c(1:25),pca2.cor$sd^2)
```



```
> pca.bfi$loadings
```

```
Loadings:
```

	RC2	RC6	RC8	RC7	RC5	RC9	RC10	RC14	RC4	RC3	RC12
A1					0.973						
A2				0.148	-0.177						
A3				0.165	-0.128						
A4				0.948							
A5				0.125							
C1						0.131				0.943	
C2						0.157				0.197	
C3						0.954				0.127	
C4						-0.155				-0.154	
C5				-0.101		-0.154					
E1		0.941									
E2		0.219	0.111								
E3		-0.126						0.141			
E4		-0.187		0.128							
E5		-0.123						0.120		0.105	
N1	0.414		0.163								
N2	0.899		0.137								
N3	0.247		0.189								
N4	0.148	0.101	0.175				0.102				
N5	0.132		0.941								
O1								0.951			-0.101
O2									0.965		0.146
O3								0.170	-0.133		-0.139
O4							0.974				
O5									0.148		0.962
	RC23	RC13	RC17	RC16	RC11	RC19	RC18	RC20	RC15	RC21	RC22
A1						-0.145				-0.100	
A2				0.111	0.140	0.913				0.200	
A3					0.209	0.214	0.142			0.894	
A4					0.107	0.133				0.139	
A5					0.897	0.147	0.154			0.207	
C1			0.188								-0.135
C2		-0.110	0.929								-0.160
C3		-0.136	0.145								-0.132
C4		0.209	-0.173								0.905
C5		0.916	-0.115						0.128		0.202
E1				-0.112			-0.111				
E2				-0.147	-0.103		-0.138		0.116		
E3				0.145	0.152		0.903	0.167		0.138	
E4				0.105	0.192		0.154		-0.108	0.133	
E5				0.915		0.111	0.141	0.109			
N1	0.249									0.150	
N2	0.224									0.132	

N3	0.892									0.218
N4	0.221	0.139								0.891
N5	0.163									0.148
O1			0.106			0.121	0.154			
O2							-0.116			
O3			0.106			0.157	0.922			
O4										
O5							-0.123			

RC24 RC1 RC25

A1		
A2		
A3	0.125	
A4	0.102	
A5	0.178	
C1		
C2		
C3		
C4		
C5		
E1	-0.150	0.176
E2	-0.205	0.879
E3	0.141	-0.128
E4	0.876	-0.208
E5		-0.131

N1		0.825	
N2		0.263	
N3		0.191	
N4	-0.104	0.112	0.120
N5		0.111	

O1
O2
O3
O4
O5

	RC2	RC6	RC8	RC7	RC5	RC9	RC10	RC14	RC4	RC3
SS loadings	1.116	1.043	1.039	1.030	1.026	1.022	1.019	1.015	1.014	1.013
Proportion Var	0.045	0.042	0.042	0.041	0.041	0.041	0.041	0.041	0.041	0.041
Cumulative Var	0.045	0.086	0.128	0.169	0.210	0.251	0.292	0.332	0.373	0.413
	RC12	RC23	RC13	RC17	RC16	RC11	RC19	RC18	RC20	RC15
SS loadings	1.012	1.004	0.999	0.999	0.998	0.994	0.993	0.992	0.991	0.982
Proportion Var	0.040	0.040	0.040	0.040	0.040	0.040	0.040	0.040	0.040	0.039
Cumulative Var	0.454	0.494	0.534	0.574	0.614	0.654	0.693	0.733	0.773	0.812
	RC21	RC22	RC24	RC1	RC25					
SS loadings	0.977	0.975	0.957	0.957	0.834					
Proportion Var	0.039	0.039	0.038	0.038	0.033					
Cumulative Var	0.851	0.890	0.928	0.967	1.000					

```
> fa5.bfi<-factanal(X,5)
> fa5.bfi
```

Uniquenesses:

	A1	A2	A3	A4	A5	C1	C2	C3	C4	C5	E1	E2	E3
	0.830	0.576	0.466	0.691	0.512	0.660	0.569	0.677	0.510	0.557	0.634	0.454	0.558
	E4	E5	N1	N2	N3	N4	N5	O1	O2	O3	O4	O5	
	0.468	0.592	0.271	0.337	0.478	0.507	0.664	0.675	0.744	0.518	0.752	0.726	

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5
A1	0.104			-0.393	
A2		0.191	0.144	0.601	
A3		0.280	0.110	0.662	
A4		0.181	0.234	0.454	-0.109
A5	-0.124	0.351		0.580	
C1			0.533		0.221
C2			0.624	0.127	0.140
C3			0.554	0.122	
C4	0.218		-0.653		
C5	0.272	-0.190	-0.573		
E1		-0.587		-0.120	
E2	0.233	-0.674	-0.106	-0.151	
E3		0.490		0.315	0.313
E4	-0.121	0.613		0.363	
E5		0.491	0.310	0.120	0.234
N1	0.816			-0.214	
N2	0.787			-0.202	
N3	0.714				
N4	0.562	-0.367	-0.192		
N5	0.518	-0.187		0.106	-0.137
O1		0.182	0.103		0.524
O2	0.163		-0.113	0.102	-0.454
O3		0.276		0.153	0.614
O4	0.207	-0.220		0.144	0.368
O5					-0.512

	Factor1	Factor2	Factor3	Factor4	Factor5
SS loadings	2.687	2.320	2.034	1.978	1.557
Proportion Var	0.107	0.093	0.081	0.079	0.062
Cumulative Var	0.107	0.200	0.282	0.361	0.423

Test of the hypothesis that 5 factors are sufficient.
 The chi square statistic is 1490.59 on 185 degrees of freedom.
 The p-value is 1.22e-202

```

> fa6.bfi<-factanal(X,6)
> fa6.bfi
Call:
factanal(x = X, factors = 6)

```

Uniquenesses:

A1	A2	A3	A4	A5	C1	C2	C3	C4	C5	E1	E2	E3
0.675	0.482	0.473	0.696	0.516	0.637	0.498	0.685	0.424	0.566	0.613	0.453	0.522
E4	E5	N1	N2	N3	N4	N5	O1	O2	O3	O4	O5	
0.432	0.599	0.273	0.302	0.480	0.504	0.652	0.663	0.703	0.510	0.756	0.640	

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6
A1	0.106			-0.530	-0.105	0.135
A2		0.239	0.128	0.663		
A3		0.359	0.122	0.601		0.143
A4		0.231	0.243	0.401	-0.129	
A5	-0.142	0.435	0.107	0.461		0.224
C1			0.557		0.189	
C2			0.677			0.160
C3			0.548	0.107		
C4	0.219		-0.638	-0.105	-0.133	0.300
C5	0.278	-0.182	-0.546			0.143
E1		-0.583		-0.120		0.156
E2	0.239	-0.674		-0.110		0.114
E3		0.566	0.102	0.170	0.251	0.237
E4	-0.136	0.674	0.115	0.222	-0.107	0.148
E5		0.506	0.303		0.207	
N1	0.819			-0.167		-0.125
N2	0.805			-0.128		-0.176
N3	0.710					
N4	0.561	-0.344	-0.162			0.181
N5	0.512	-0.163			-0.158	0.165
O1		0.234	0.127		0.483	0.181
O2	0.158				-0.494	0.137
O3		0.337			0.572	0.193
O4	0.205	-0.169		0.133	0.352	0.177
O5					-0.572	0.146

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6
SS loadings	2.722	2.652	2.075	1.672	1.512	0.614
Proportion Var	0.109	0.106	0.083	0.067	0.060	0.025
Cumulative Var	0.109	0.215	0.298	0.365	0.425	0.450

Test of the hypothesis that 6 factors are sufficient.
The chi square statistic is 896.7 on 165 degrees of freedom.
The p-value is 5.73e-101

```

> factanal(X,12)$PVAL
0.008732393
> factanal(X,13)$PVAL
0.07265011
> factanal(X,14)$PVAL
0.2868025

```

```

> factanal(X,13)

```

Call:

```

factanal(x = X, factors = 14)

```

Uniquenesses:

A1	A2	A3	A4	A5	C1	C2	C3	C4	C5	E1	E2	E3
0.694	0.357	0.373	0.005	0.489	0.590	0.447	0.670	0.382	0.265	0.494	0.388	0.425
E4	E5	N1	N2	N3	N4	N5	O1	O2	O3	O4	O5	
0.307	0.503	0.255	0.257	0.397	0.409	0.465	0.638	0.666	0.457	0.720	0.626	

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6	Factor7	Factor8	Factor9	Factor10	Factor11	Factor12	Factor13
A1	0.110					-0.519							
A2		0.186	0.150		0.203	0.692	0.124						
A3		0.156	0.111		0.560	0.481	0.153						
A4		0.125	0.151		0.176	0.220	0.928						
A5	-0.156	0.269	0.110		0.539	0.300	0.112						
C1			0.600	-0.166									
C2			0.707		0.102		0.123						
C3			0.526			0.104		-0.159					
C4	0.166	-0.127	-0.600	0.171		-0.130		0.237	0.157				
C5	0.224	-0.122	-0.431					0.682					
E1		-0.670		0.100		-0.131							
E2	0.178	-0.682			-0.181			0.118	0.160				
E3		0.361		-0.225	0.558								
E4	-0.120	0.621	0.129		0.412		0.128						
E5		0.398	0.320	-0.160	0.190							-0.147	
N1	0.833					-0.125							
N2	0.830												
N3	0.711											0.139	
N4	0.526	-0.319	-0.108					0.177	0.186				
N5	0.463	-0.110		0.181					0.510				
O1			0.134	-0.445	0.238								
O2	0.113			0.538									
O3		0.200	0.114	-0.579	0.299						0.113		
O4	0.165	-0.215		-0.302		0.112		0.130	0.100				
O5				0.582		-0.108							

```

A1
A2 0.199
A3
A4
A5
C1
C2
C3 0.100
C4 0.147          0.225
C5
E1
E2
E3 0.232          -0.105
E4          0.198    0.172
E5 0.370
N1
N2      -0.169
N3      0.241
N4      0.354
N5
O1 0.259
O2
O3 0.155          0.150
O4          0.251
O5          0.101

```

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6	Factor7	Factor8
SS loadings	2.594	1.982	1.944	1.473	1.432	1.235	0.973	0.641
Proportion Var	0.104	0.079	0.078	0.059	0.057	0.049	0.039	0.026
Cumulative Var	0.104	0.183	0.261	0.320	0.377	0.426	0.465	0.491

	Factor9	Factor10	Factor11	Factor12	Factor13
SS loadings	0.448	0.397	0.257	0.189	0.155
Proportion Var	0.018	0.016	0.010	0.008	0.006
Cumulative Var	0.509	0.525	0.535	0.543	0.549

Test of the hypothesis that 13 factors are sufficient.
The chi square statistic is 68.66 on 53 degrees of freedom.
The p-value is 0.0727

```
> factanal(X,14)
```

```
Call:
factanal(x = X, factors = 14)
```

```
Uniquenesses:
  A1  A2  A3  A4  A5  C1  C2  C3  C4  C5  E1  E2  E3
```

0.719	0.251	0.290	0.005	0.519	0.595	0.465	0.670	0.380	0.316	0.450	0.432	0.396
E4	E5	N1	N2	N3	N4	N5	01	02	03	04	05	
0.163	0.516	0.278	0.217	0.396	0.417	0.452	0.648	0.645	0.478	0.005	0.634	

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6	Factor7	Factor8	Factor9
A1	0.116			-0.478		0.116			
A2		0.152	-0.144	0.784	0.180			0.109	
A3		0.109	-0.116	0.578	0.347			0.159	0.135
A4		0.154		0.250				0.931	
A5	-0.169	0.104	-0.200	0.382	0.379			0.126	0.246
C1		0.599				-0.136			
C2		0.698			0.133			0.123	
C3		0.532		0.108					
C4	0.163	-0.605	0.141	-0.105	0.121	0.231			
C5	0.226	-0.445	0.103					-0.103	
E1			0.703	-0.135	-0.132				
E2	0.182		0.604	-0.117	-0.234		0.104		-0.156
E3			-0.294	0.114	0.677				
E4	-0.125	0.117	-0.493	0.149	0.280	0.123		0.127	0.654
E5		0.329	-0.367	0.109	0.380				
N1	0.821			-0.130					
N2	0.847								
N3	0.707								
N4	0.526	-0.114	0.305						
N5	0.453					0.192			
01		0.138			0.424	-0.344			
02	0.103					0.573			
03		0.117	-0.159		0.462	-0.468			
04	0.131		0.127			-0.161	0.956		
05						0.571			

Factor10 Factor11 Factor12 Factor13 Factor14

A1					
A2			-0.175		0.105
A3			0.422		
A4					
A5			0.168		
C1					
C2					
C3	-0.156				
C4	0.227	0.124			0.235
C5	0.629				
E1					
E2	0.125	0.171			
E3					-0.112
E4					
E5		-0.207			0.122


```

N1
N2          -0.204
N3      0.154      0.249
N4  0.181      0.200      0.329
N5          0.532
O1                      0.114
O2
O3                      0.147
O4
O5

```

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6	Factor7	Factor8
SS loadings	2.584	1.961	1.594	1.532	1.525	1.208	0.988	0.986
Proportion Var	0.103	0.078	0.064	0.061	0.061	0.048	0.040	0.039
Cumulative Var	0.103	0.182	0.246	0.307	0.368	0.416	0.456	0.495

	Factor9	Factor10	Factor11	Factor12	Factor13	Factor14
SS loadings	0.574	0.559	0.478	0.267	0.247	0.161
Proportion Var	0.023	0.022	0.019	0.011	0.010	0.006
Cumulative Var	0.518	0.540	0.560	0.570	0.580	0.586

Test of the hypothesis that 14 factors are sufficient.
The chi square statistic is 45.6 on 41 degrees of freedom.
The p-value is 0.287

```

> fa(X,6) # library(psych)
Loading required namespace: GPArotation

```

```

Factor Analysis using method = minres
Call: fa(r = X, nfactors = 6)

```

Standardized loadings (pattern matrix) based upon correlation matrix

	MR2	MR3	MR1	MR5	MR4	MR6	h2	u2	com
A1	0.09	0.07	-0.11	-0.56	0.03	0.29	0.33	0.67	1.7
A2	0.05	0.07	-0.04	0.70	0.00	-0.07	0.52	0.48	1.1
A3	-0.03	0.03	-0.10	0.63	0.06	0.12	0.53	0.47	1.2
A4	-0.07	0.19	-0.08	0.41	-0.13	0.13	0.30	0.70	2.0
A5	-0.17	0.01	-0.19	0.46	0.11	0.22	0.48	0.52	2.3
C1	0.02	0.55	0.07	-0.04	0.19	0.09	0.36	0.64	1.3
C2	0.07	0.67	0.14	0.02	0.10	0.18	0.50	0.50	1.3
C3	0.01	0.56	0.06	0.07	-0.05	0.06	0.32	0.68	1.1
C4	0.05	-0.67	0.08	-0.05	0.04	0.29	0.59	0.41	1.4
C5	0.15	-0.55	0.18	0.00	0.10	0.04	0.43	0.57	1.5
E1	-0.14	0.10	0.60	-0.13	-0.08	0.09	0.40	0.60	1.4
E2	0.05	-0.03	0.68	-0.08	-0.07	0.02	0.56	0.44	1.1
E3	0.01	0.00	-0.33	0.15	0.39	0.24	0.48	0.52	3.0
E4	-0.05	0.03	-0.53	0.22	0.01	0.28	0.56	0.44	1.9
E5	0.15	0.26	-0.39	0.05	0.24	0.05	0.40	0.60	2.9

N1	0.85	0.00	-0.10	-0.07	-0.05	0.00	0.70	0.30	1.0
N2	0.83	0.02	-0.05	-0.04	0.00	-0.08	0.67	0.33	1.0
N3	0.67	-0.03	0.14	0.08	0.03	0.08	0.54	0.46	1.1
N4	0.43	-0.14	0.43	0.08	0.09	0.06	0.51	0.49	2.4
N5	0.44	0.00	0.24	0.17	-0.13	0.15	0.35	0.65	2.4
O1	-0.05	0.07	-0.03	-0.04	0.57	0.07	0.35	0.65	1.1
O2	0.12	-0.08	-0.02	0.09	-0.42	0.31	0.30	0.70	2.2
O3	-0.01	0.01	-0.10	0.04	0.66	0.04	0.49	0.51	1.1
O4	0.08	-0.03	0.35	0.15	0.36	-0.01	0.25	0.75	2.5
O5	0.04	-0.03	-0.05	-0.03	-0.49	0.35	0.36	0.64	1.9

		MR2	MR3	MR1	MR5	MR4	MR6
SS loadings		2.50	2.11	2.18	1.96	1.74	0.76
Proportion Var		0.10	0.08	0.09	0.08	0.07	0.03
Cumulative Var		0.10	0.18	0.27	0.35	0.42	0.45
Proportion Explained		0.22	0.19	0.19	0.17	0.15	0.07
Cumulative Proportion		0.22	0.41	0.60	0.78	0.93	1.00

With factor correlations of

	MR2	MR3	MR1	MR5	MR4	MR6
MR2	1.00	-0.19	0.25	-0.10	0.01	0.14
MR3	-0.19	1.00	-0.22	0.18	0.20	0.00
MR1	0.25	-0.22	1.00	-0.30	-0.19	-0.11
MR5	-0.10	0.18	-0.30	1.00	0.23	0.18
MR4	0.01	0.20	-0.19	0.23	1.00	0.04
MR6	0.14	0.00	-0.11	0.18	0.04	1.00

Mean item complexity = 1.7

Test of the hypothesis that 6 factors are sufficient.

The degrees of freedom for the null model are 300 and the objective function was 7.48 with Chi Square of 18146.07

The degrees of freedom for the model are 165 and the objective function was 0.38

The root mean square of the residuals (RMSR) is 0.02

The df corrected root mean square of the residuals is 0.03

The harmonic number of observations is 2436 with the empirical chi square 552.63 with prob < 2.5e-43

The total number of observations was 2436 with Likelihood Chi Square = 916.62 with prob < 1.6e-104

Tucker Lewis Index of factoring reliability = 0.923

RMSEA index = 0.043 and the 90 % confidence intervals are 0.041 0.046

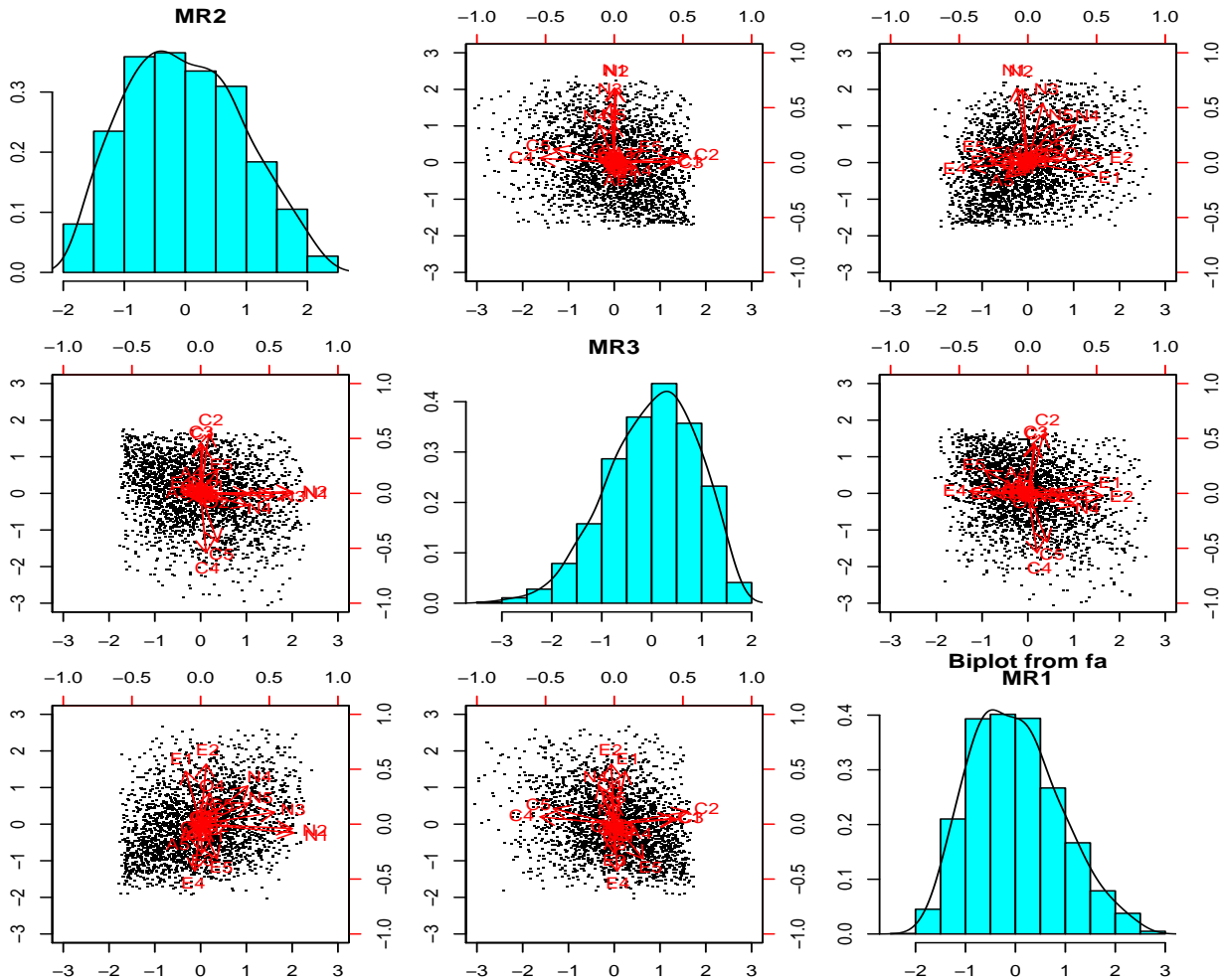
BIC = -370.06

Fit based upon off diagonal values = 0.99

Measures of factor score adequacy

	MR2	MR3	MR1	MR5	MR4	MR6
Correlation of (regression) scores with factors	0.93	0.89	0.89	0.88	0.86	0.76
Multiple R square of scores with factors	0.87	0.79	0.79	0.78	0.74	0.57
Minimum correlation of possible factor scores	0.73	0.59	0.59	0.55	0.48	0.15

```
> biplot.psych(fa(X,6),pch=".",choose=c(1,2,3)) # library(psych)
```



Example 2: The second data set is downloaded from the UC Irvine’s Machine Learning data repository. The data set is comprised of observations on 13 quantitative variables resulting from chemical analysis of wines grown in the same region in Italy but derived from three different cultivars (cultivar labels are also given *i.e.* it’s a labeled data set). Very brief descriptions (essentially just the names) of these 13 variables may be found in <https://archive.ics.uci.edu/ml/datasets/Wine> (click/tap it!). While the actual task is one of supervised learning of being able to classify the three cultivars from these 13 variables, here we first undertake the (unsupervised learning task of) feasibility of reducing these (13) dimensions for basic interpretative purpose (which may or may not be useful for the subsequent classification task).

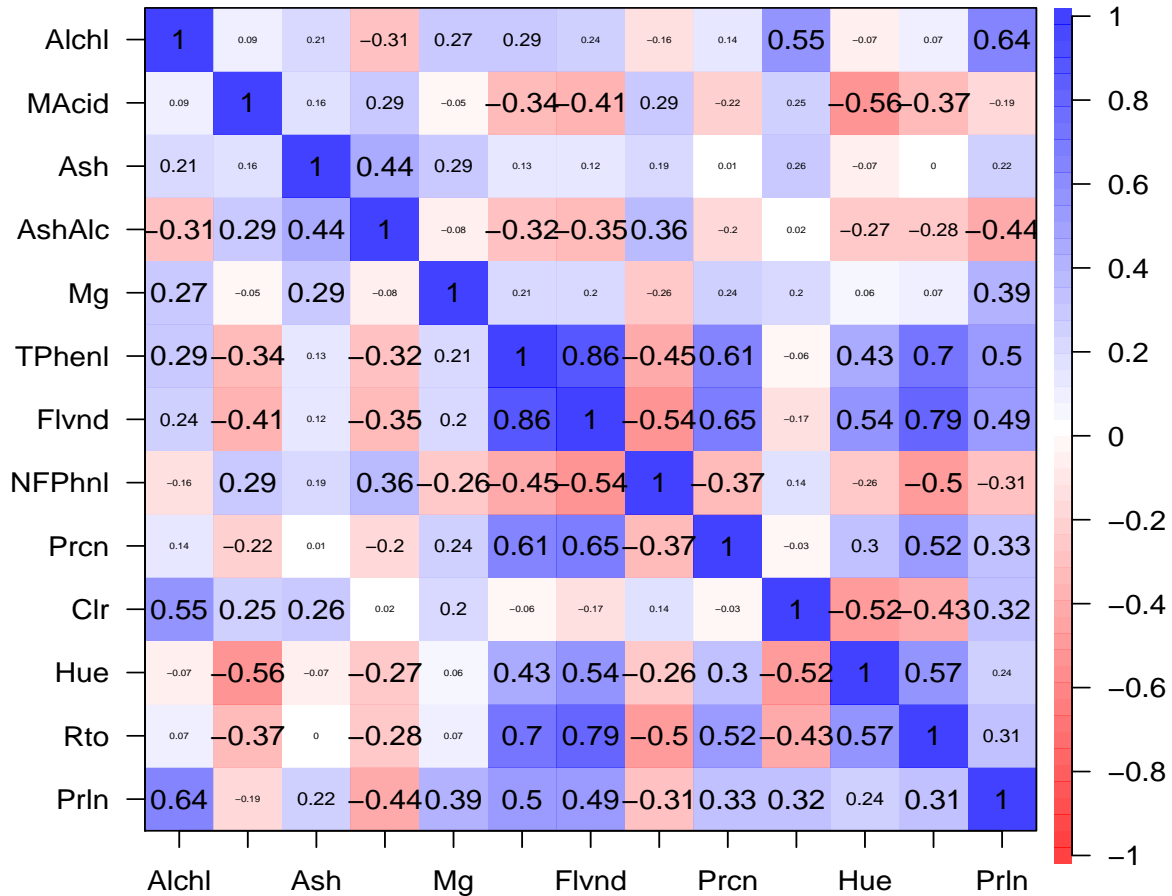
```

> d<-read.table("wine.data")
> X<-d[,2:14]
> names(X)<-c("Alchl", "MAcid", "Ash", "AshAlc", "Mg", "TPhenl", "Flvnd", "NFPhnl",
+ "Prcn", "Clr", "Hue", "Rto", "Prln")

> cor.plot(X)      # library(psych)

```

Correlation plot



```

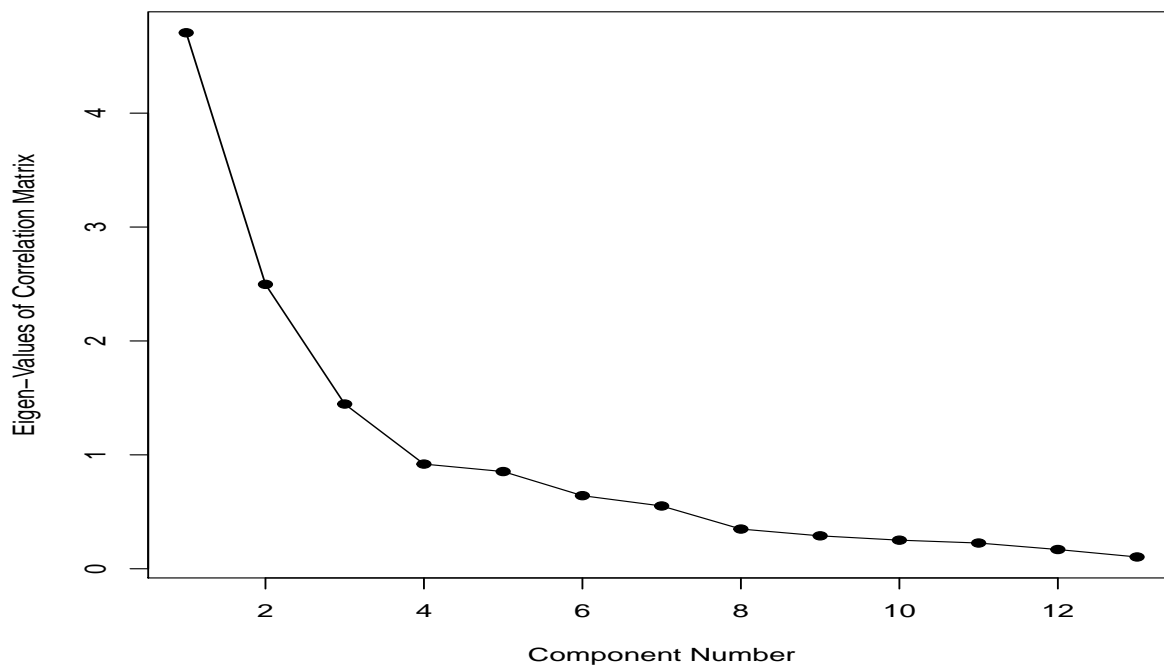
> pc.c<-princomp(X,cor=T)

> pc.c$sd^2
  Comp.1   Comp.2   Comp.3   Comp.4   Comp.5   Comp.6   Comp.7   Comp.8
4.7058503 2.4969737 1.4460720 0.9189739 0.8532282 0.6416570 0.5510283 0.3484974
  Comp.9   Comp.10  Comp.11  Comp.12  Comp.13
0.2888799 0.2509025 0.2257886 0.1687702 0.1033779

> plot(1:13,pc.c$sd^2,pch=19,xlab="Component Number",ylab="Eigen-Values of Correlation
Matrix",main="Scree Plot")
> lines(c(1:13),pc.c$sd^2)

```

Scree Plot



```
> summary(pc.c)
```

Importance of components:

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5
Standard deviation	2.1692972	1.5801816	1.2025273	0.9586313	0.92370351
Proportion of Variance	0.3619885	0.1920749	0.1112363	0.0706903	0.06563294
Cumulative Proportion	0.3619885	0.5540634	0.6652997	0.7359900	0.80162293

	Comp.6	Comp.7	Comp.8	Comp.9	Comp.10
Standard deviation	0.80103498	0.74231281	0.59033665	0.53747553	0.50090167
Proportion of Variance	0.04935823	0.04238679	0.02680749	0.02222153	0.01930019
Cumulative Proportion	0.85098116	0.89336795	0.92017544	0.94239698	0.96169717

	Comp.11	Comp.12	Comp.13
Standard deviation	0.47517222	0.41081655	0.321524394
Proportion of Variance	0.01736836	0.01298233	0.007952149
Cumulative Proportion	0.97906553	0.99204785	1.000000000

```
> pc.c$loadings
```

Loadings:

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8	Comp.9	Comp.10
Alchl	0.144	0.484	0.207		0.266	0.214		0.396	0.509	0.212
MAcid	-0.245	0.225		-0.537		0.537	-0.421			-0.309
Ash		0.316	-0.626	0.214	0.143	0.154	0.149	-0.170	-0.308	
AshAlc	-0.239		-0.612			-0.101	0.287	0.428	0.200	

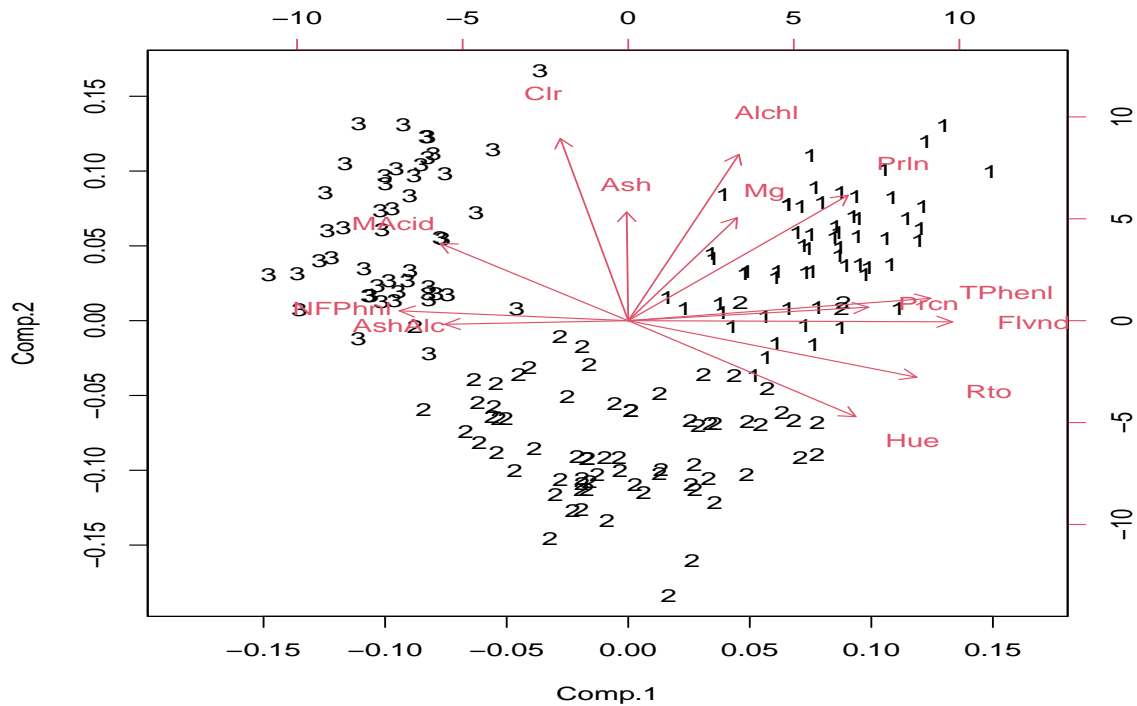
Mg	0.142	0.300	-0.131	0.352	-0.727		-0.323	-0.156	0.271	
TPhenl	0.395		-0.146	-0.198	0.149			-0.406	0.286	-0.320
Flvnd	0.423		-0.151	-0.152	0.109			-0.187		-0.163
NFPhnl	-0.299		-0.170	0.203	0.501	-0.259	-0.595	-0.233	0.196	0.216
Prcn	0.313		-0.149	-0.399	-0.137	-0.534	-0.372	0.368	-0.209	0.134
Clr		0.530	0.137			-0.419	0.228			-0.291
Hue	0.297	-0.279		0.428	0.174	0.106	-0.232	0.437		-0.522
Rto	0.376	-0.164	-0.166	-0.184	0.101	0.266			0.137	0.524
Prln	0.287	0.365	0.127	0.232	0.158	0.120		0.120	-0.576	0.162

	Comp.11	Comp.12	Comp.13
Alchl	0.226	0.266	
MAcid		-0.122	
Ash	0.499		-0.141
AshAlc	-0.479		
Mg			
TPhenl	-0.304	0.304	-0.464
Flvnd			0.832
NFPhnl	-0.117		0.114
Prcn	0.237		-0.117
Clr		-0.604	
Hue		-0.259	
Rto		-0.601	-0.157
Prln	-0.539		

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8	Comp.9
SS loadings	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
Proportion Var	0.077	0.077	0.077	0.077	0.077	0.077	0.077	0.077	0.077
Cumulative Var	0.077	0.154	0.231	0.308	0.385	0.462	0.538	0.615	0.692

	Comp.10	Comp.11	Comp.12	Comp.13
SS loadings	1.000	1.000	1.000	1.000
Proportion Var	0.077	0.077	0.077	0.077
Cumulative Var	0.769	0.846	0.923	1.000

```
> cultivar<-c(rep("1",59),rep("2",71),rep("3",48))
> biplot(pc.c,xlabs=cultivar)
```



```

> factanal(X,2)$PVAL
1.485595e-32
> factanal(X,3)$PVAL
1.959095e-15
> factanal(X,4)$PVAL
1.444642e-05
> factanal(X,5)$PVAL
0.02056416
> factanal(X,6)$PVAL
0.3093393

```

```

> factanal(X,6)

```

Call:

```

factanal(x = X, factors = 6)

```

Uniquenesses:

Alchl	MAcid	Ash	AshAlc	Mg	TPhenl	Flvnd	NFPhnl	Prcn	Clr	Hue
0.308	0.285	0.005	0.393	0.005	0.198	0.055	0.583	0.482	0.005	0.349
Rto	Prln									
0.247	0.239									

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6
Alchl	0.142	0.763		0.254	0.124	
MAcid	-0.245		0.129		0.793	

Ash		0.177	0.967				0.138
AshAlc	-0.233	-0.463	0.541		0.193		
Mg	0.148	0.196	0.114				0.958
TPhenl	0.841	0.252			-0.165		
Flvnd	0.907	0.221		-0.131	-0.233		
NFPhnl	-0.516	-0.143	0.261	0.108	0.114	-0.194	
Prcn	0.697						0.129
Clr	-0.133	0.410	0.126	0.875	0.149		
Hue	0.379			-0.475	-0.524		
Rto	0.750			-0.385	-0.177		
Prln	0.322	0.769			-0.155	0.191	

	Factor1	Factor2	Factor3	Factor4	Factor5	Factor6
SS loadings	3.268	1.775	1.355	1.265	1.138	1.046
Proportion Var	0.251	0.137	0.104	0.097	0.088	0.080
Cumulative Var	0.251	0.388	0.492	0.589	0.677	0.757

Test of the hypothesis that 6 factors are sufficient.
The chi square statistic is 17.16 on 15 degrees of freedom.

> factanal(X,5)

Call:

factanal(x = X, factors = 5)

Uniquenesses:

Alchl	MAcid	Ash	AshAlc	Mg	TPhenl	Flvnd	NFPhnl	Prcn	Clr	Hue
0.348	0.107	0.005	0.399	0.792	0.198	0.055	0.623	0.512	0.005	0.384
Rto	Prln									
0.252	0.204									

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5
Alchl	0.138	0.737		0.264	0.125
MAcid	-0.247		0.131	0.132	0.893
Ash		0.352	0.931		
AshAlc	-0.246	-0.367	0.613		0.146
Mg	0.112	0.412	0.150		
TPhenl	0.842	0.270			-0.133
Flvnd	0.912	0.238		-0.146	-0.186
NFPhnl	-0.505	-0.153	0.266	0.113	0.122
Prcn	0.685	0.102			
Clr	-0.147	0.448		0.869	0.103
Hue	0.405			-0.508	-0.428
Rto	0.756			-0.388	-0.137
Prln	0.310	0.826			-0.109

	Factor1	Factor2	Factor3	Factor4	Factor5
SS loadings	3.274	2.030	1.381	1.298	1.135
Proportion Var	0.252	0.156	0.106	0.100	0.087
Cumulative Var	0.252	0.408	0.514	0.614	0.701

Test of the hypothesis that 5 factors are sufficient.
The chi square statistic is 38.86 on 23 degrees of freedom.
The p-value is 0.0206

> factanal(X,4)

Call:

factanal(x = X, factors = 4)

Uniquenesses:

Alchl	MAcid	Ash	AshAlc	Mg	TPhenl	Flvnd	NFPhnl	Prcn	Clr	Hue
0.372	0.703	0.005	0.402	0.790	0.195	0.056	0.625	0.511	0.163	0.378
Rto	Prln									
0.255	0.194									

Loadings:

	Factor1	Factor2	Factor3	Factor4
Alchl	0.143	0.743	-0.224	
MAcid	-0.302		-0.409	0.189
Ash		0.355		0.931
AshAlc	-0.227	-0.359	-0.208	0.612
Mg	0.113	0.420		0.144
TPhenl	0.833	0.281	0.179	
Flvnd	0.891	0.242	0.304	
NFPhnl	-0.494	-0.146	-0.199	0.264
Prcn	0.685	0.122		
Clr		0.529	-0.740	
Hue	0.355		0.698	
Rto	0.708		0.492	
Prln	0.288	0.836	0.128	

	Factor1	Factor2	Factor3	Factor4
SS loadings	3.095	2.143	1.722	1.391
Proportion Var	0.238	0.165	0.132	0.107
Cumulative Var	0.238	0.403	0.535	0.642

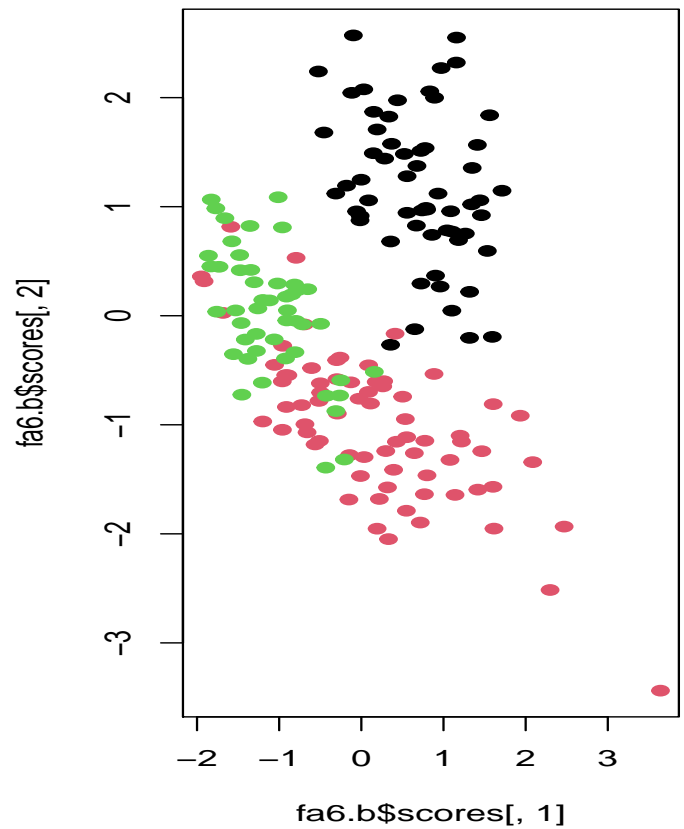
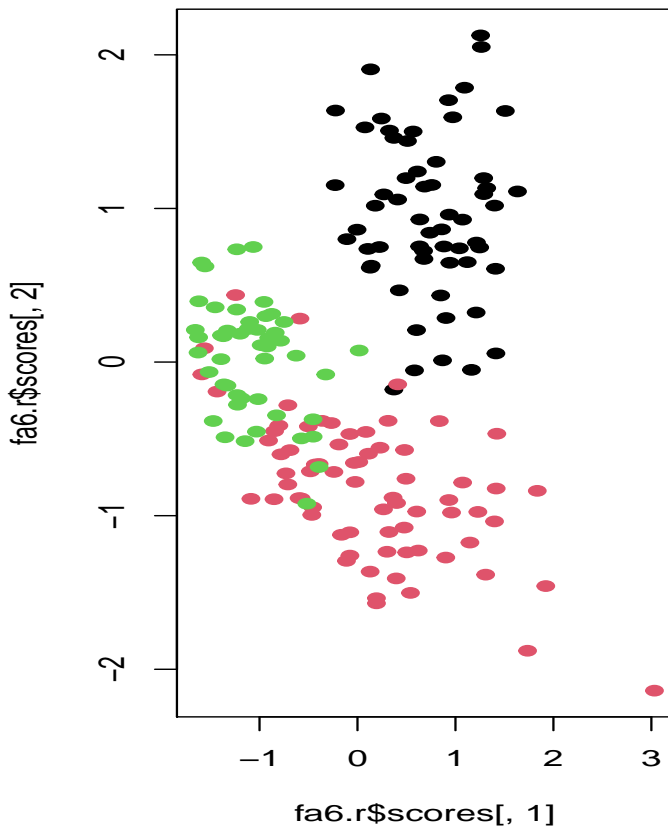
Test of the hypothesis that 4 factors are sufficient.
The chi square statistic is 76.92 on 32 degrees of freedom.
The p-value is 1.44e-05

```

> fa6.r<-factanal(X,6,scores="regression")
> fa6.b<-factanal(X,6,scores="Bartlett")

> plot(fa6.r$scores[,1],fa6.r$scores[,2],col=cultivar,pch=19)
> plot(fa6.b$scores[,1],fa6.b$scores[,2],col=cultivar,pch=19)

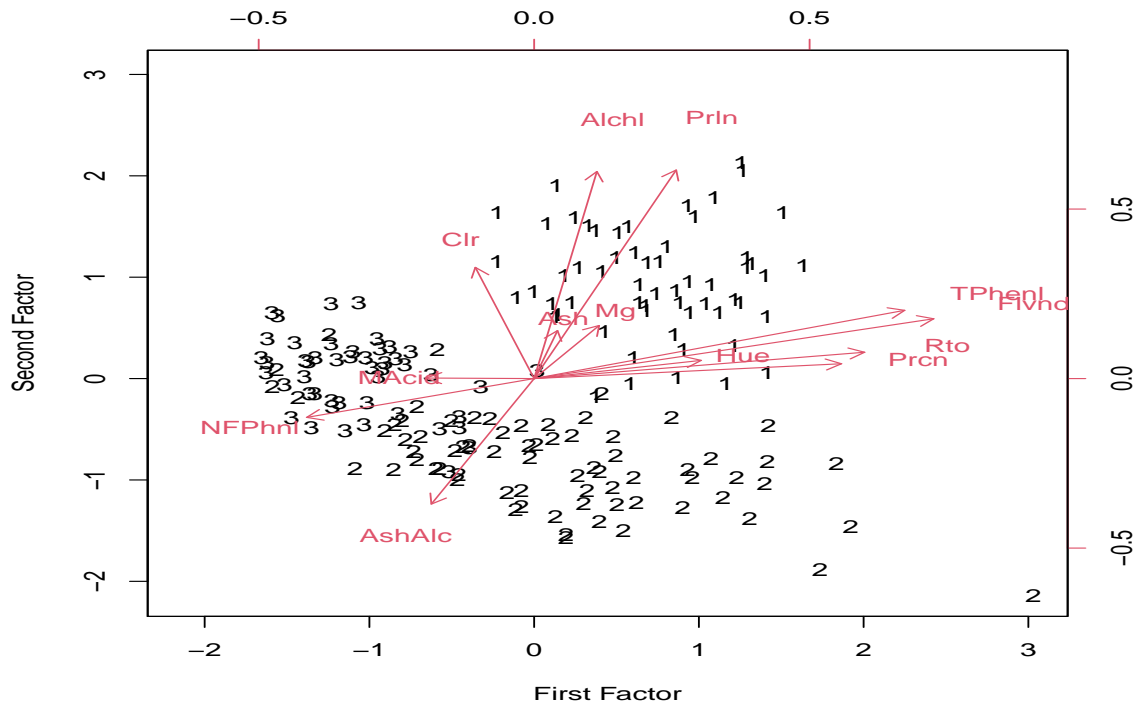
```



```

> biplot(matrix(c(fa6.r$scores[,1],fa6.r$scores[,2]),ncol=2),
+ matrix(c(fa6.r$loadings[,1],fa6.r$loadings[,2]),ncol=2),
+ xlabs=cultivar,ylabs=names(X),xlab="First Factor",ylab="Second Factor")

```



```
> biplot(matrix(c(fa6.r$scores[,3],fa6.r$scores[,4]),ncol=2),
+ matrix(c(fa6.r$loadings[,3],fa6.r$loadings[,4]),ncol=2),
+ xlab=cultivar,ylab=names(X),xlab="Third Factor",ylab="Fourth Factor")
```

