

cereal factor analysis

September 17, 2021

1 Cereal Data Factor Analysis

Objective: Comprehending the behaviour of customers on taking a particular cereal brands into consideration, based on their attributes. Based on the analysis, trying to understand the underlying similarities between the variables and converting them into factors for further analysis using statistical techniques.

```
[47]: cereal <- read.csv("C:/Users/amite/Downloads/cereal.csv", header = TRUE)
      dim(cereal)
```

1. 235 2. 26

2 Exploratory Data Analysis

2.0.1 Summary of the data

Summarize the data, to see if the data is structured as per the expectations.

```
[3]: summary(cereal)
```

Cereals	Filling	Natural	Fibre
CornFlakes :27	Min. :1.000	Min. :1.000	Min. :1.000
Weetabix :27	1st Qu.:3.000	1st Qu.:3.000	1st Qu.:3.000
Vitabrit :25	Median :4.000	Median :4.000	Median :4.000
NutriGrain :24	Mean :3.881	Mean :3.783	Mean :3.528
SpecialK :23	3rd Qu.:4.500	3rd Qu.:4.000	3rd Qu.:4.000
RiceBubbles:21	Max. :5.000	Max. :5.000	Max. :5.000
(Other) :88			
Sweet	Easy	Salt	Satisfying
Min. :1.000	Min. :1.000	Min. :1.000	Min. :2.000
1st Qu.:2.000	1st Qu.:4.000	1st Qu.:1.000	1st Qu.:3.000
Median :2.000	Median :5.000	Median :2.000	Median :4.000
Mean :2.506	Mean :4.532	Mean :1.991	Mean :4.004
3rd Qu.:3.000	3rd Qu.:5.000	3rd Qu.:3.000	3rd Qu.:5.000
Max. :5.000	Max. :6.000	Max. :4.000	Max. :6.000
Energy	Fun	Kids	Soggy
Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000
1st Qu.:3.000	1st Qu.:2.000	1st Qu.:3.000	1st Qu.:1.000

Median :4.000	Median :2.000	Median :4.000	Median :2.000
Mean :3.643	Mean :2.617	Mean :3.843	Mean :2.255
3rd Qu.:4.000	3rd Qu.:3.000	3rd Qu.:5.000	3rd Qu.:3.000
Max. :5.000	Max. :5.000	Max. :6.000	Max. :5.000

Economical	Health	Family	Calories
Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000
1st Qu.:3.000	1st Qu.:3.000	1st Qu.:3.000	1st Qu.:2.000
Median :3.000	Median :4.000	Median :4.000	Median :3.000
Mean :3.217	Mean :3.809	Mean :3.877	Mean :2.702
3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:5.000	3rd Qu.:3.000
Max. :5.000	Max. :5.000	Max. :6.000	Max. :5.000

Plain	Crisp	Regular	Sugar
Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000
1st Qu.:1.000	1st Qu.:2.000	1st Qu.:2.000	1st Qu.:1.000
Median :2.000	Median :3.000	Median :3.000	Median :2.000
Mean :2.268	Mean :3.204	Mean :3.072	Mean :2.145
3rd Qu.:3.000	3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:3.000
Max. :5.000	Max. :6.000	Max. :5.000	Max. :5.000

Fruit	Process	Quality	Treat	Boring
Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.00	Min. :1.00
1st Qu.:1.000	1st Qu.:2.000	1st Qu.:3.000	1st Qu.:2.00	1st Qu.:1.00
Median :1.000	Median :3.000	Median :4.000	Median :3.00	Median :2.00
Mean :1.694	Mean :2.936	Mean :3.694	Mean :2.63	Mean :1.83
3rd Qu.:3.000	3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:3.00	3rd Qu.:2.00
Max. :5.000	Max. :6.000	Max. :5.000	Max. :6.00	Max. :5.00

Nutritious
Min. :1.000
1st Qu.:3.000
Median :4.000
Mean :3.664
3rd Qu.:4.000
Max. :5.000

It has been observed that there are 235 observations and 26 variables of the 12 different brands of cereals. All the attributes are 'ordinal' and has been marked on the scale between 1-5, but few variables mistakenly entered as 6. Hence, need to be corrected.

```
[41]: cereal[cereal==6] <- 5
```

Structure of data

```
[42]: str(cereal)
```

```
'data.frame': 235 obs. of 26 variables:
```

```

$ Cereals      : Factor w/ 12 levels "AllBran","CMuesli",...: 12 9 9 2 3 8 9 9 8 3
...
$ Filling      : int   5 1 5 5 4 4 4 4 4 4 ...
$ Natural      : int   5 2 4 5 5 4 4 3 3 3 ...
$ Fibre        : int   5 2 5 5 3 4 3 3 3 3 ...
$ Sweet        : int   1 1 5 3 2 2 2 2 2 2 ...
$ Easy         : num   2 5 5 5 5 5 5 5 5 5 ...
$ Salt         : int   1 2 3 2 2 2 1 1 1 1 ...
$ Satisfying   : num   5 5 5 5 5 5 5 5 5 5 ...
$ Energy       : int   4 1 5 5 4 4 5 4 4 4 ...
$ Fun          : int   1 1 5 5 5 5 5 4 4 4 ...
$ Kids         : num   4 5 5 5 5 5 5 5 5 5 ...
$ Soggy        : int   5 3 3 3 1 1 1 1 1 1 ...
$ Economical   : int   5 5 3 3 5 5 5 3 3 3 ...
$ Health       : int   5 2 5 5 5 4 5 4 4 4 ...
$ Family       : num   5 5 5 5 3 5 5 5 5 5 ...
$ Calories     : int   1 1 1 1 3 3 3 2 2 2 ...
$ Plain        : int   3 5 1 1 1 1 1 3 3 3 ...
$ Crisp        : num   1 5 5 1 5 5 5 4 4 4 ...
$ Regular      : int   4 1 4 4 3 3 3 4 4 4 ...
$ Sugar        : int   1 2 3 2 1 2 2 1 1 1 ...
$ Fruit        : int   1 1 1 5 1 1 1 1 1 1 ...
$ Process      : num   3 5 2 2 3 3 3 2 2 2 ...
$ Quality      : int   5 2 5 5 5 5 5 4 4 4 ...
$ Treat        : num   1 1 4 5 5 5 5 2 2 2 ...
$ Boring       : int   1 1 1 1 1 1 1 1 1 1 ...
$ Nutritious   : int   5 3 5 5 4 4 4 3 3 3 ...

```

2.1 Creating a correlation matrix and correlation plot

```

[6]: cereal1 <- cereal[,-1]
cor.cereal <- cor(cereal1)
round(cor.cereal,2)

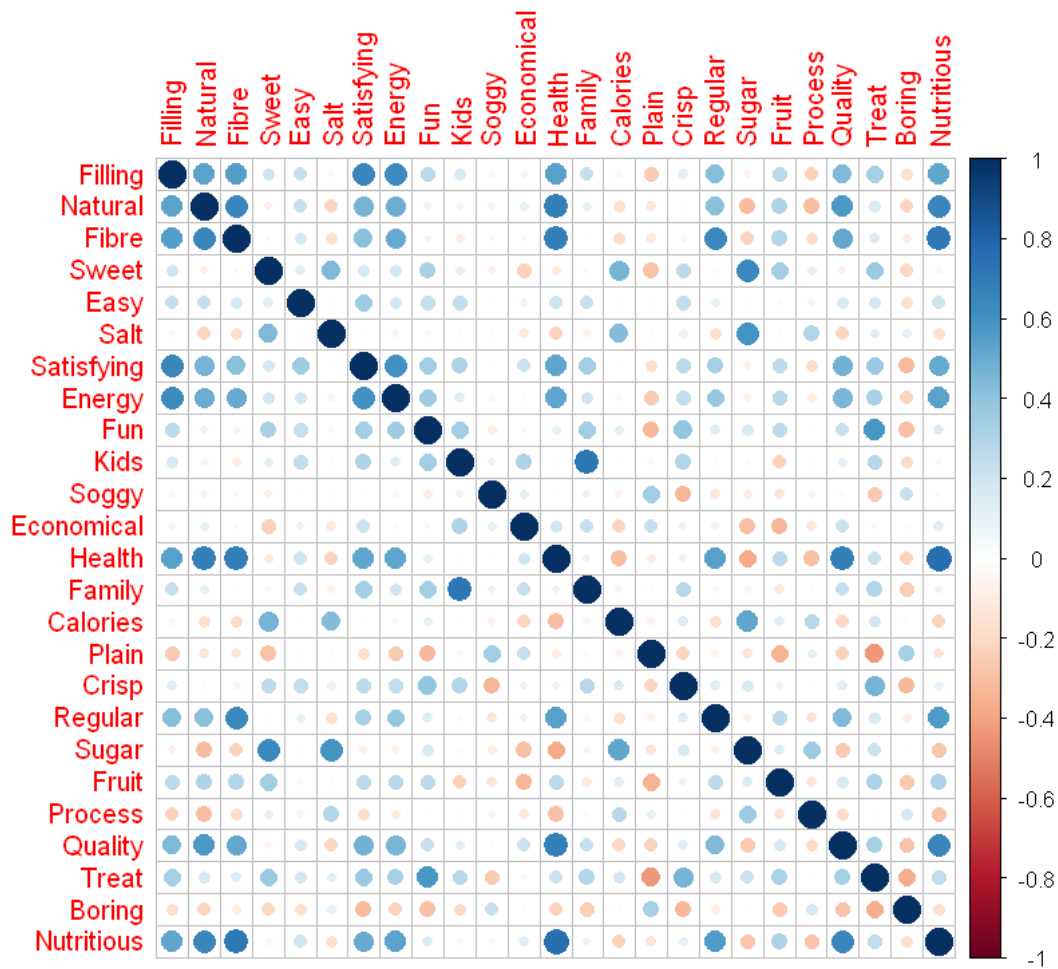
```

	Filling	Natural	Fibre	Sweet	Easy	Salt	Satisfying	Energy	Fun	Kids	...	Plain
Filling	1.00	0.54	0.55	0.19	0.24	-0.04	0.65	0.64	0.27	0.16	...	-0.25
Natural	0.54	1.00	0.65	-0.09	0.23	-0.22	0.47	0.49	0.08	0.06	...	-0.14
Fibre	0.55	0.65	1.00	-0.04	0.18	-0.17	0.41	0.50	0.06	-0.09	...	-0.12
Sweet	0.19	-0.09	-0.04	1.00	0.12	0.44	0.18	0.18	0.33	0.12	...	-0.29
Easy	0.24	0.23	0.18	0.12	1.00	0.01	0.35	0.18	0.24	0.24	...	0.02
Salt	-0.04	-0.22	-0.17	0.44	0.01	1.00	-0.01	-0.07	0.03	0.02	...	0.02
Satisfying	0.65	0.47	0.41	0.18	0.35	-0.01	1.00	0.60	0.35	0.30	...	-0.18
Energy	0.64	0.49	0.50	0.18	0.18	-0.07	0.60	1.00	0.35	0.13	...	-0.26
Fun	0.27	0.08	0.06	0.33	0.24	0.03	0.35	0.35	1.00	0.34	...	-0.32
Kids	0.16	0.06	-0.09	0.12	0.24	0.02	0.30	0.13	0.34	1.00	...	0.03
Soggy	-0.06	0.07	-0.04	-0.08	-0.01	0.02	-0.01	-0.05	-0.10	0.09	...	0.35
Economical	0.05	0.10	-0.03	-0.24	0.09	-0.13	0.21	0.03	0.04	0.30	...	0.23
Health	0.55	0.69	0.68	-0.12	0.20	-0.23	0.52	0.52	0.10	-0.01	...	-0.10
Family	0.23	0.11	-0.01	0.04	0.23	-0.09	0.35	0.19	0.35	0.72	...	-0.03
Calories	0.05	-0.16	-0.19	0.47	-0.02	0.44	0.01	0.03	0.11	0.01	...	-0.08
Plain	-0.25	-0.14	-0.12	-0.29	0.02	0.02	-0.18	-0.26	-0.32	0.03	...	1.00
Crisp	0.13	0.02	0.05	0.26	0.24	0.10	0.26	0.25	0.40	0.29	...	-0.21
Regular	0.42	0.42	0.65	-0.03	0.11	-0.16	0.33	0.39	0.14	-0.03	...	-0.08
Sugar	-0.08	-0.32	-0.23	0.65	-0.02	0.59	-0.09	-0.09	0.17	-0.02	...	-0.15
Fruit	0.26	0.30	0.29	0.35	0.04	0.03	0.25	0.27	0.25	-0.23	...	-0.34
Process	-0.23	-0.31	-0.19	0.11	-0.07	0.30	-0.19	-0.10	-0.01	0.01	...	0.12
Quality	0.44	0.58	0.51	-0.08	0.16	-0.22	0.47	0.46	0.22	0.11	...	-0.23
Treat	0.34	0.17	0.14	0.37	0.18	0.12	0.37	0.32	0.58	0.28	...	-0.43
Boring	-0.18	-0.22	-0.10	-0.20	-0.17	0.11	-0.32	-0.22	-0.30	-0.20	...	0.33
Nutritious	0.53	0.65	0.71	-0.05	0.20	-0.16	0.50	0.54	0.16	0.03	...	-0.14

If the relationship is very strong, the value of the output to be closer to a positive one. If the relationship is inverse, then the value corresponds to a negative value, basically implying that the two quantities are inversely proportional. And, anything closer to zero, implies that the quantities are not related at all.

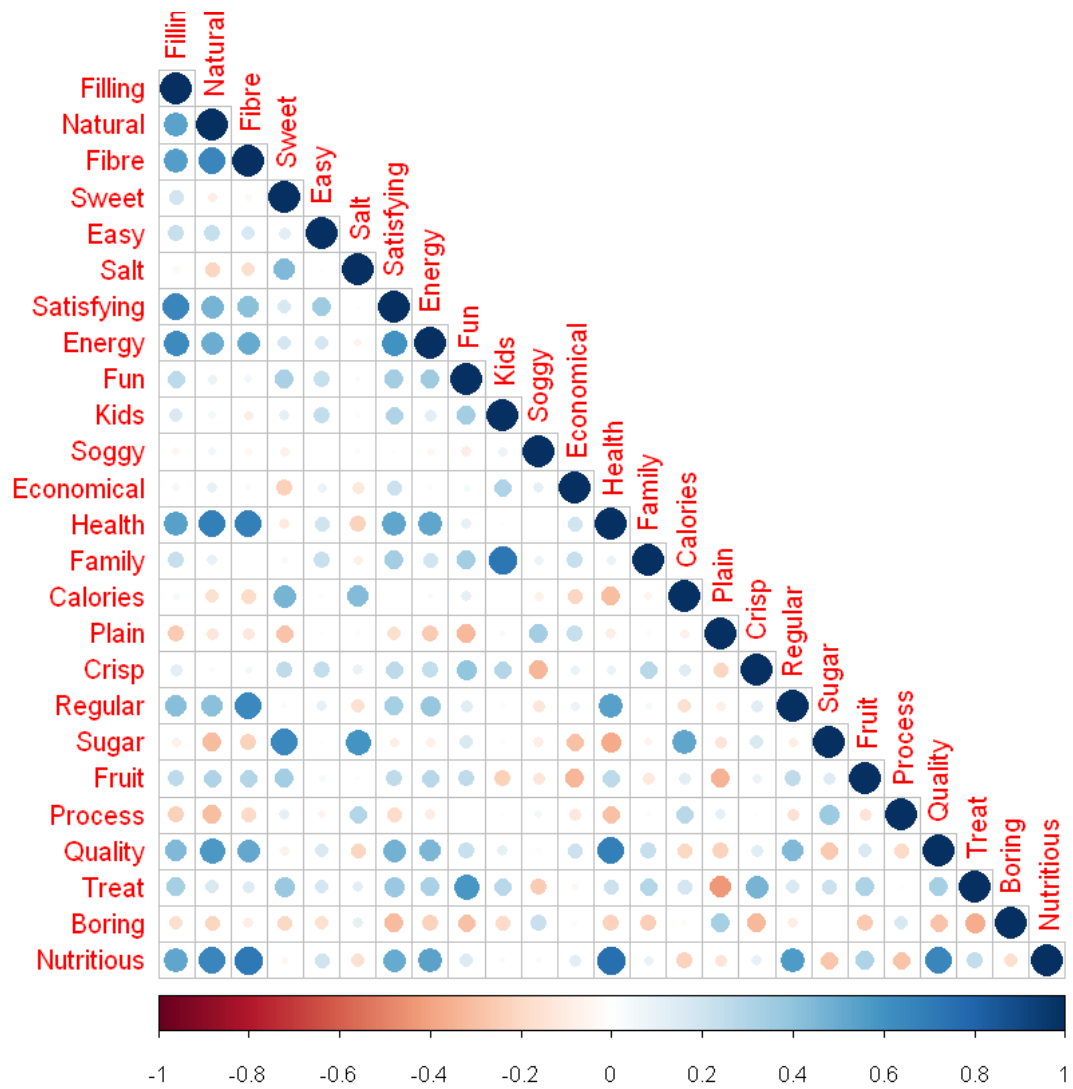
```
[40]: library(corrplot)
```

```
[39]: corrplot(cor.cereal, method = "circle",bg = "white")
```

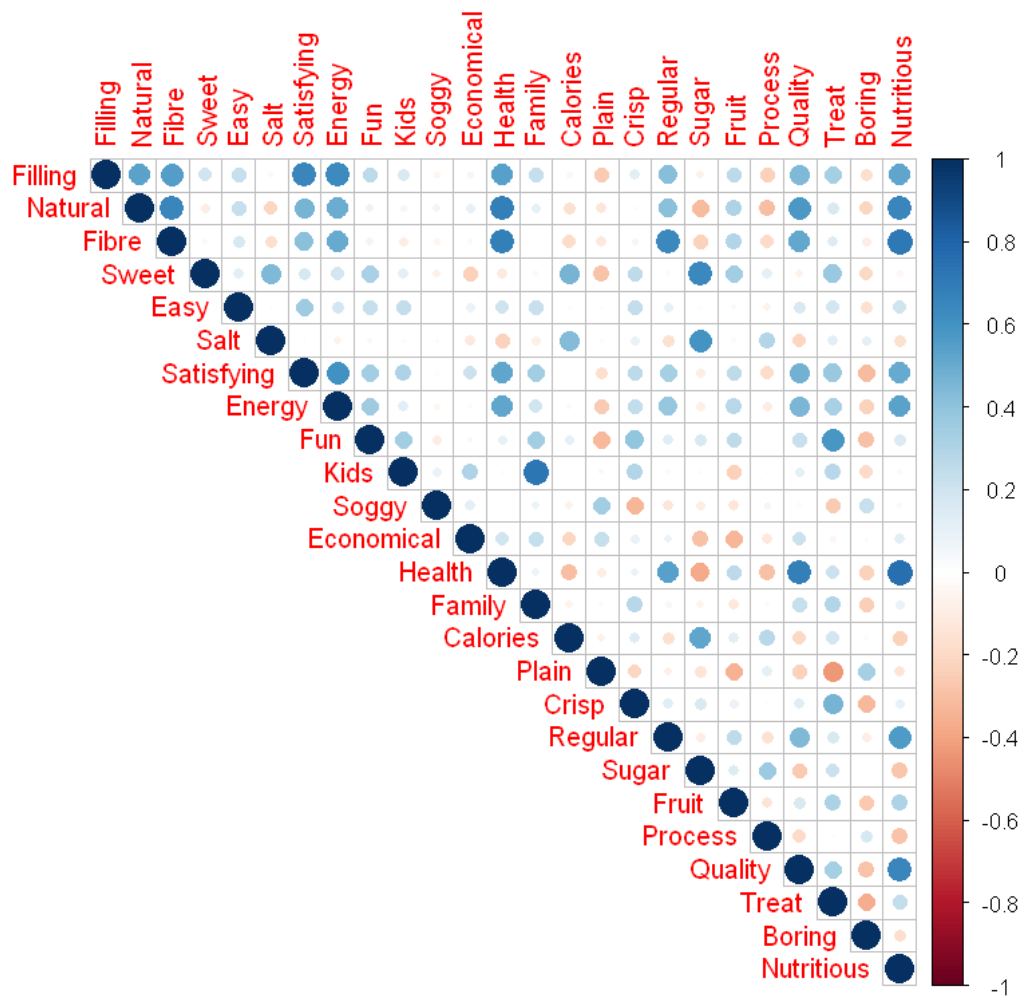


Anything that is in blue is highly correlated, anything that's in red or closer to red, is inversely correlated and if it's white it basically means it's closer to zero(no strong relationship).

```
[36]: corrplot(cor.cereal, type = "lower",bg = "white")
```



```
[38]: corrplot(cor.cereal, type = "upper",bg = "white")
```



2.2 Testing the assumptions to do factor analysis

KMO test for Measuring Sampling Accuracy

KMO test for measuring sampling accuracy(MSA): Kaiser-Meyer-Olkin test gives us the measures to check whether the data is suitable for data analysis or not. Observing each variable, check the model that is present, these are the two adequacy checkpoints which are present. While observing the variable we are trying to estimate the proportion of the variance that is present between all the observed variables. Proportion of variance is basically confirming whether the variance is very high or the variance is very low, high isn't preferable.

```
[15]: library(psych)
      cereal.KMO <- KMO(cor.cereal)
```

```
cereal.KMO
```

Kaiser-Meyer-Olkin factor adequacy

Call: KMO(r = cor.cereal)

Overall MSA = 0.85

MSA for each item =

Filling	Natural	Fibre	Sweet	Easy	Salt	Satisfying
0.89	0.90	0.88	0.78	0.83	0.82	0.91
Energy	Fun	Kids	Soggy	Economical	Health	Family
0.91	0.85	0.67	0.63	0.73	0.92	0.73
Calories	Plain	Crisp	Regular	Sugar	Fruit	Process
0.86	0.82	0.83	0.87	0.78	0.77	0.80
Quality	Treat	Boring	Nutritious			
0.91	0.88	0.87	0.92			

According to Kaiser himself a KMO in the .80's is meritorious!. Thus, our sample is large enough for factor analysis or principal component analysis.

```
[16]: cereal.barlett <- cortest.bartlett(cor.cereal, nrow(cereal))
cereal.barlett
```

\$chisq 2877.73856777721

\$p.value 0

\$df 300

Bartlett Test of Spehericity - The p-value is <.001, thus the null hypothesis is rejected (The null hypothesis is that the corelation matrix is an identity matrix i.e. there is no scope for dimentionality reduction.). Thus, the dimensionality reduction is a possibility using Factor analysis or principal component analysis.

2.3 Running PCA to identify the number of factors

```
[17]: cereal.pca <- princomp(cereal1,scores = TRUE, cor = TRUE)
summary(cereal.pca)
```

Importance of components:

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5
Standard deviation	2.5515645	1.9473508	1.57931249	1.29699429	1.04196617
Proportion of Variance	0.2604193	0.1516870	0.09976912	0.06728777	0.04342774
Cumulative Proportion	0.2604193	0.4121063	0.51187538	0.57916315	0.62259089
	Comp.6	Comp.7	Comp.8	Comp.9	Comp.10
Standard deviation	0.97215571	0.92371682	0.88941254	0.85594266	0.83528812
Proportion of Variance	0.03780347	0.03413011	0.03164219	0.02930551	0.02790825
Cumulative Proportion	0.66039436	0.69452447	0.72616666	0.75547217	0.78338042
	Comp.11	Comp.12	Comp.13	Comp.14	Comp.15
Standard deviation	0.80508009	0.74210795	0.72900837	0.69819277	0.6456121
Proportion of Variance	0.02592616	0.02202897	0.02125813	0.01949893	0.0166726
Cumulative Proportion	0.80930658	0.83133555	0.85259368	0.87209260	0.8887652
	Comp.16	Comp.17	Comp.18	Comp.19	Comp.20

Standard deviation	0.62203549	0.60340597	0.60072703	0.55329582	0.52496345
Proportion of Variance	0.01547713	0.01456395	0.01443492	0.01224545	0.01102346
Cumulative Proportion	0.90424233	0.91880628	0.93324119	0.94548664	0.95651011
	Comp.21	Comp.22	Comp.23	Comp.24	
Standard deviation	0.51267066	0.492791255	0.467311607	0.445682156	
Proportion of Variance	0.01051325	0.009713729	0.008735206	0.007945303	
Cumulative Proportion	0.96702336	0.976737086	0.985472292	0.993417595	
	Comp.25				
Standard deviation	0.405660102				
Proportion of Variance	0.006582405				
Cumulative Proportion	1.000000000				

Interpretations from the above values,

According to KMO statistics concept, all principal component values which has an Eigen value of greater than 1, can be taken into consideration for the dimension reduction technique.

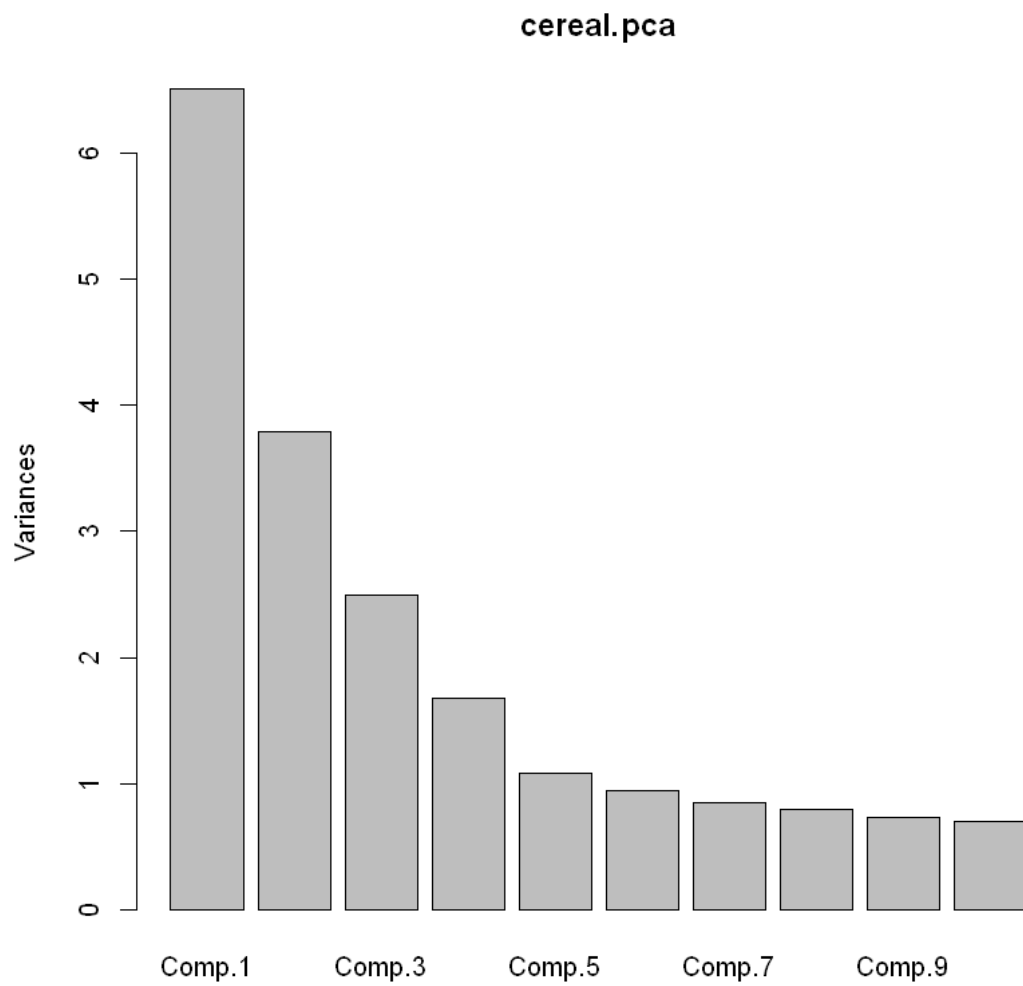
Based on that criteria, in our case we will take first 5 principal components into consideration.

From the output values, Comp 1 explains 26% of variation of the whole data set, likewise Comp 2 explains 15%, Comp 3 explains 10%, Comp 4 explains 7% and Comp 5 explains 4% of variation of the entire data set.

Therefore, it implies PCA1 to PCA5 itself are able to explain 62% of the data.

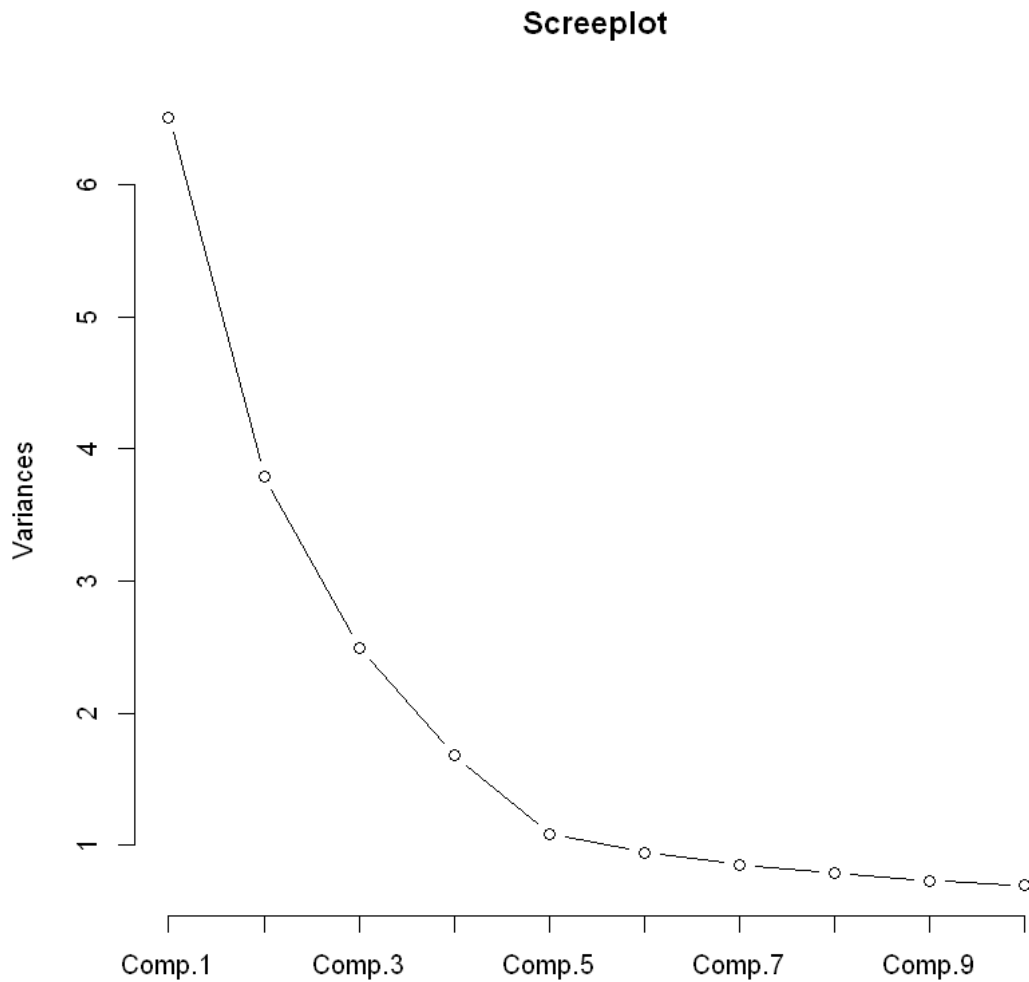
Plot of principal components with variance A PCA plot converts the correlation among all of the cells into 2D graph. Values that are highly correlated cluster together.

```
[19]: plot(cereal.pca,col = "grey")
```



After PC5 the bar flattens and there is no further steep.

```
[20]: screeplot(cereal.pca,type = "lines",main = "Screeplot")
```



In the above Screeplot , after the PC5 which is above Eigen value(Variances) 1,the “Elbow” shape becomes a line and flattens. The analysis suggests that the number of factors = 5.

```
[21]: loadings(cereal.pca)
```

Loadings:

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8	Comp.9
Filling	0.293			0.176	0.107		0.232	0.406	
Natural	0.294	-0.132		0.101	0.139				0.210
Fibre	0.287	-0.123	0.210	0.138	-0.155		-0.148	0.119	
Sweet		0.398	0.117	0.143	0.155	0.114			-0.178
Easy	0.136		-0.171	0.121		0.667	-0.478		0.163
Salt		0.280		0.373	-0.127	0.109	0.210	-0.220	

Satisfying	0.292		-0.108	0.152	0.101	0.124	0.182	0.110	
Energy	0.285			0.131		-0.128	0.112	0.297	0.104
Fun	0.161	0.270	-0.162	-0.113		-0.175	-0.230		-0.353
Kids		0.129	-0.498			-0.103		0.139	
Soggy		-0.142	-0.113	0.446	0.479	-0.173	-0.173	-0.311	-0.103
Economical		-0.147	-0.366		-0.237		0.458	-0.349	-0.162
Health	0.318	-0.161						-0.193	
Family	0.124		-0.460		0.137	-0.199	-0.120	0.172	
Calories		0.324	0.110	0.216			0.280	0.101	0.259
Plain	-0.129	-0.208	-0.158	0.374	-0.143	0.232	-0.120		
Crisp	0.121	0.252	-0.171	-0.185	-0.402	0.186			
Regular	0.243		0.142		-0.381		-0.142		-0.359
Sugar		0.384	0.142	0.201				-0.114	-0.113
Fruit	0.154	0.148	0.342	-0.111	0.282		-0.185	-0.189	-0.178
Process	-0.134	0.155		0.263	-0.339	-0.409	-0.364		0.480
Quality	0.295					-0.225		-0.307	0.206
Treat	0.190	0.302		-0.150		-0.190		-0.155	
Boring	-0.162	-0.152		0.334	-0.158	-0.153		0.361	-0.418
Nutritious	0.316	-0.116	0.102	0.114				-0.167	
	Comp.10	Comp.11	Comp.12	Comp.13	Comp.14	Comp.15	Comp.16	Comp.17	
Filling			0.109		0.108	0.111		0.278	
Natural			-0.281	0.110		-0.272	-0.379	0.167	
Fibre	-0.177				-0.139		-0.213	0.197	
Sweet	-0.130		0.227		-0.214	-0.132	0.294	0.306	
Easy	0.248	0.230	0.114	0.188					
Salt		0.352	-0.158	-0.282	0.306	0.206	-0.379	-0.218	
Satisfying	0.127	-0.124	0.232	-0.108	0.348		0.170	-0.137	
Energy	0.231	-0.283	0.117	-0.347	-0.160	0.107		-0.157	
Fun	0.390	-0.167		0.163		0.312	-0.344	-0.172	
Kids	-0.298	0.116	-0.108			-0.132	-0.189		
Soggy		-0.121		-0.171	-0.348				
Economical	0.223		0.261	0.205		-0.356	-0.132	0.102	
Health							0.124	0.109	
Family	-0.369				0.141		0.144	-0.243	
Calories		-0.310	-0.343	0.520	-0.233			-0.201	
Plain	-0.173	-0.458	-0.179		0.261	0.378	0.205	0.143	
Crisp		-0.261	-0.358	-0.465	-0.247	-0.227	0.106		
Regular	-0.345	-0.103	0.262	0.243	-0.104			-0.117	
Sugar	-0.190	0.148	0.218		-0.141		0.118		
Fruit		-0.302			0.462	-0.427		-0.166	
Process	0.131	-0.132	0.238		0.173	-0.227		0.137	
Quality		0.225		0.159	-0.109	0.102	0.425	-0.362	
Treat	0.198	0.124	-0.300	0.179	0.191	0.181	0.163	0.512	
Boring	0.359	0.227	-0.299			-0.330	0.218		
Nutritious		0.105	-0.156					-0.125	
	Comp.18	Comp.19	Comp.20	Comp.21	Comp.22	Comp.23	Comp.24	Comp.25	
Filling	0.188	0.268		0.416	0.124	0.413	0.230		
Natural	-0.223	0.437		-0.335				-0.315	

Fibre			0.339	0.183	-0.277	-0.503	0.189	0.344
Sweet	-0.381	-0.169	0.120	0.101	0.227	-0.213		-0.331
Easy			-0.191					
Salt			-0.121	0.120	0.159	-0.131		
Satisfying	0.313		0.412	-0.465		-0.150		
Energy	-0.314		-0.519	-0.141	-0.112	-0.115		0.116
Fun	-0.182	0.135	0.324	0.105			-0.102	
Kids	-0.203	-0.356		-0.221	0.245	0.248		0.394
Soggy	0.419							
Economical			-0.117	0.160	-0.178			
Health				0.192	0.226		-0.767	0.270
Family		0.219	-0.143	0.309	-0.267	-0.272	-0.186	-0.248
Calories	0.121	-0.203					-0.160	
Plain	-0.283	0.135					0.125	
Crisp	0.210	0.174	0.126		0.108			
Regular	0.283		-0.270	-0.178	0.305			-0.188
Sugar		0.395		-0.252	-0.421	0.268	-0.136	0.338
Fruit			-0.127	0.144			0.160	0.217
Process				0.102				-0.118
Quality	-0.160	0.160			0.261		0.380	0.136
Treat	0.204		-0.316	-0.211	-0.151	-0.149		
Boring	-0.101							
Nutritious		-0.461			-0.461	0.453		-0.332

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8	Comp.9
SS loadings	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
Proportion Var	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04
Cumulative Var	0.04	0.08	0.12	0.16	0.20	0.24	0.28	0.32	0.36

	Comp.10	Comp.11	Comp.12	Comp.13	Comp.14	Comp.15	Comp.16	Comp.17
SS loadings	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
Proportion Var	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04
Cumulative Var	0.40	0.44	0.48	0.52	0.56	0.60	0.64	0.68

	Comp.18	Comp.19	Comp.20	Comp.21	Comp.22	Comp.23	Comp.24	Comp.25
SS loadings	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
Proportion Var	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04
Cumulative Var	0.72	0.76	0.80	0.84	0.88	0.92	0.96	1.00

```
[22]: print(cereal.pca$loadings, cutoff = 0.2)
```

Loadings:

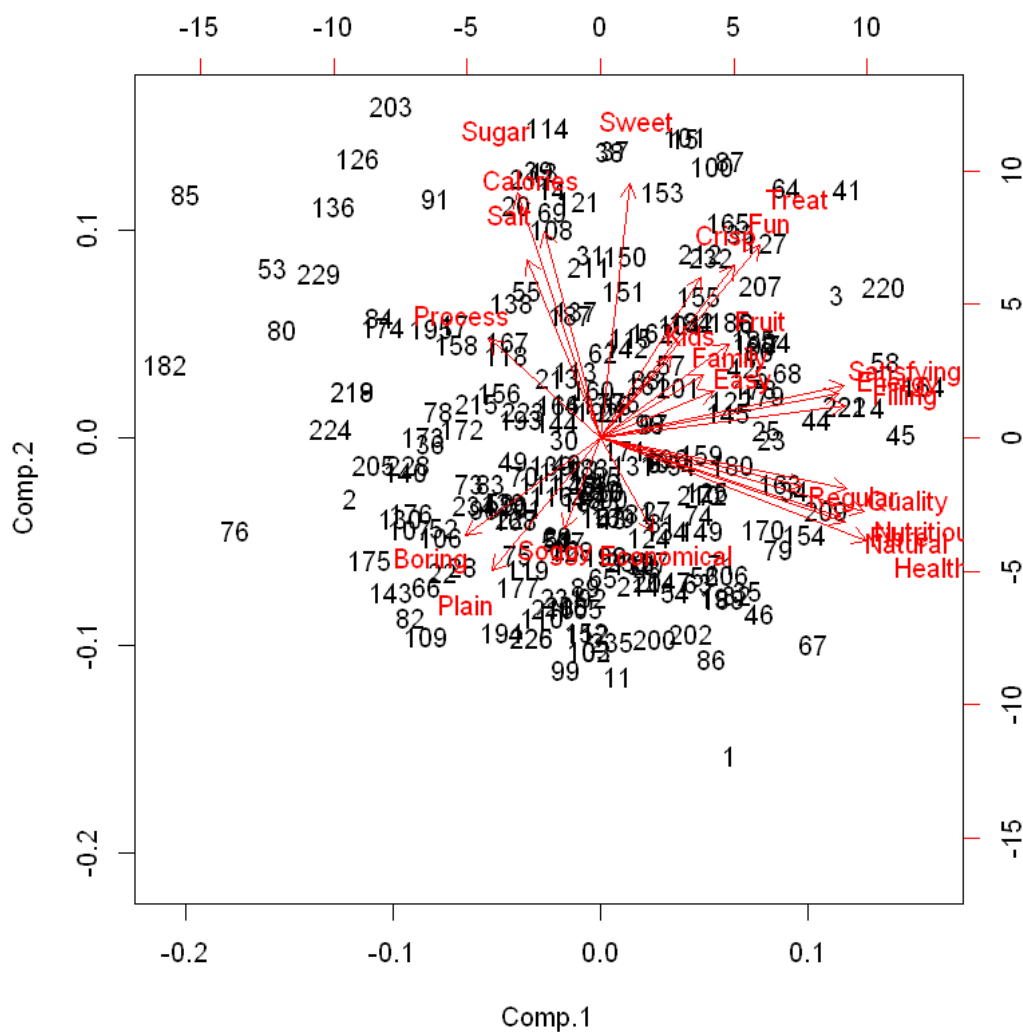
	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8	Comp.9
Filling	0.293						0.232	0.406	
Natural	0.294								0.210
Fibre	0.287		0.210						
Sweet		0.398							
Easy						0.667	-0.478		
Salt		0.280		0.373			0.210	-0.220	

Satisfying	0.292								
Energy	0.285						0.297		
Fun		0.270					-0.230		-0.353
Kids			-0.498						
Soggy				0.446	0.479			-0.311	
Economical			-0.366		-0.237		0.458	-0.349	
Health	0.318								
Family			-0.460						
Calories		0.324		0.216			0.280		0.259
Plain		-0.208		0.374		0.232			
Crisp		0.252			-0.402				
Regular	0.243				-0.381				-0.359
Sugar		0.384		0.201					
Fruit			0.342		0.282				
Process				0.263	-0.339	-0.409	-0.364		0.480
Quality	0.295					-0.225		-0.307	0.206
Treat		0.302							
Boring				0.334				0.361	-0.418
Nutritious	0.316								
	Comp.10	Comp.11	Comp.12	Comp.13	Comp.14	Comp.15	Comp.16	Comp.17	
Filling									0.278
Natural			-0.281			-0.272	-0.379		
Fibre							-0.213		
Sweet			0.227		-0.214		0.294	0.306	
Easy	0.248	0.230							
Salt		0.352		-0.282	0.306	0.206	-0.379	-0.218	
Satisfying			0.232		0.348				
Energy	0.231	-0.283		-0.347					
Fun	0.390					0.312	-0.344		
Kids	-0.298								
Soggy					-0.348				
Economical	0.223		0.261	0.205		-0.356			
Health									
Family	-0.369								-0.243
Calories		-0.310	-0.343	0.520	-0.233				-0.201
Plain		-0.458			0.261	0.378	0.205		
Crisp		-0.261	-0.358	-0.465	-0.247	-0.227			
Regular	-0.345		0.262	0.243					
Sugar			0.218						
Fruit		-0.302			0.462	-0.427			
Process			0.238			-0.227			
Quality		0.225					0.425	-0.362	
Treat			-0.300					0.512	
Boring	0.359	0.227	-0.299			-0.330	0.218		
Nutritious									
	Comp.18	Comp.19	Comp.20	Comp.21	Comp.22	Comp.23	Comp.24	Comp.25	
Filling		0.268		0.416		0.413	0.230		
Natural	-0.223	0.437		-0.335					-0.315

Fibre			0.339		-0.277	-0.503		0.344
Sweet	-0.381				0.227	-0.213		-0.331
Easy								
Salt								
Satisfying	0.313		0.412	-0.465				
Energy	-0.314		-0.519					
Fun			0.324					
Kids	-0.203	-0.356		-0.221	0.245	0.248		0.394
Soggy	0.419							
Economical								
Health					0.226		-0.767	0.270
Family		0.219		0.309	-0.267	-0.272		-0.248
Calories		-0.203						
Plain	-0.283							
Crisp	0.210							
Regular	0.283		-0.270		0.305			
Sugar		0.395		-0.252	-0.421	0.268		0.338
Fruit								0.217
Process								
Quality					0.261		0.380	
Treat	0.204		-0.316	-0.211				
Boring								
Nutritious		-0.461			-0.461	0.453		-0.332

	Comp.1	Comp.2	Comp.3	Comp.4	Comp.5	Comp.6	Comp.7	Comp.8	Comp.9
SS loadings	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
Proportion Var	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04
Cumulative Var	0.04	0.08	0.12	0.16	0.20	0.24	0.28	0.32	0.36
	Comp.10	Comp.11	Comp.12	Comp.13	Comp.14	Comp.15	Comp.16	Comp.17	
SS loadings	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	
Proportion Var	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04	
Cumulative Var	0.40	0.44	0.48	0.52	0.56	0.60	0.64	0.68	
	Comp.18	Comp.19	Comp.20	Comp.21	Comp.22	Comp.23	Comp.24	Comp.25	
SS loadings	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	
Proportion Var	0.04	0.04	0.04	0.04	0.04	0.04	0.04	0.04	
Cumulative Var	0.72	0.76	0.80	0.84	0.88	0.92	0.96	1.00	

```
[23]: biplot(cereal.pca, choices = c(1,2))
```



From the loadings, we are able to interpret that Filling, Natural, Fibre, Satisfying, Energy, Health, Regular, Quality, Nutritious are the variables that are added and gives the weightage to Comp 1. Similarly, we can identify the variables that adds weightage in the Comps 2,3,4,5. As a result, we have 4 components determining the variability of our data.

2.3.1 Factor Analysis

Factor analysis without rotation

```
[24]: factanal(cereal1,factors = 5,rotation = "none")
```

Call:

```
factanal(x = cereal1, factors = 5, rotation = "none")
```


Uniquenesses:

Filling	Natural	Fibre	Sweet	Easy	Salt	Satisfying
0.286	0.389	0.312	0.360	0.847	0.517	0.369
Energy	Fun	Kids	Soggy	Economical	Health	Family
0.431	0.525	0.239	0.775	0.709	0.214	0.352
Calories	Plain	Crisp	Regular	Sugar	Fruit	Process
0.578	0.549	0.648	0.550	0.199	0.562	0.764
Quality	Treat	Boring	Nutritious			
0.392	0.391	0.671	0.241			

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5
Filling	0.719	0.243		0.143	-0.338
Natural	0.763	-0.116			
Fibre	0.757	-0.113	-0.283	0.137	
Sweet		0.757	-0.244		
Easy	0.289	0.188	0.168		
Salt	-0.258	0.495	-0.215	0.341	
Satisfying	0.690	0.295	0.132		-0.205
Energy	0.682	0.237			-0.204
Fun	0.287	0.533	0.191	-0.266	
Kids	0.164	0.365	0.762	0.135	
Soggy		-0.177	0.143	0.413	
Economical	0.183	-0.172	0.440	0.143	0.117
Health	0.854	-0.177	-0.112		0.110
Family	0.260	0.310	0.694		
Calories	-0.237	0.536	-0.206	0.131	-0.134
Plain	-0.232	-0.352	0.187	0.480	
Crisp	0.197	0.450	0.172	-0.233	0.164
Regular	0.605		-0.195		0.196
Sugar	-0.356	0.716	-0.325	0.185	0.147
Fruit	0.324	0.257	-0.447	-0.247	
Process	-0.336	0.227		0.198	0.176
Quality	0.746				0.213
Treat	0.368	0.589		-0.328	0.124
Boring	-0.314	-0.283	-0.114	0.370	
Nutritious	0.834		-0.142		0.170

	Factor1	Factor2	Factor3	Factor4	Factor5
SS loadings	6.049	3.368	2.062	1.165	0.486
Proportion Var	0.242	0.135	0.082	0.047	0.019
Cumulative Var	0.242	0.377	0.459	0.506	0.525

Test of the hypothesis that 5 factors are sufficient.

The chi square statistic is 318.9 on 185 degrees of freedom.

The p-value is 3.48e-09

From the above, we observe that the variables “Easy”, “Soggy”, “Economical”, “Crisp”, “Process” and “Boring” have high 'Uniquenesses' - means it does not correlate with other variables and can be used as it is. We observe that the loadings are not very clear, hence it has to be rotated to increase the very clear interpretability. It is done to maximize the high loadings and minimize low loadings.

2.3.2 Factor loadings with Varimax rotation

```
[25]: factanal(cereal1,factors = 5,rotation = "varimax")
```

Call:

```
factanal(x = cereal1, factors = 5, rotation = "varimax")
```

Uniquenesses:

Filling	Natural	Fibre	Sweet	Easy	Salt	Satisfying
0.286	0.389	0.312	0.360	0.847	0.517	0.369
Energy	Fun	Kids	Soggy	Economical	Health	Family
0.431	0.525	0.239	0.775	0.709	0.214	0.352
Calories	Plain	Crisp	Regular	Sugar	Fruit	Process
0.578	0.549	0.648	0.550	0.199	0.562	0.764
Quality	Treat	Boring	Nutritious			
0.392	0.391	0.671	0.241			

Loadings:

	Factor1	Factor2	Factor3	Factor4	Factor5
Filling	0.646		0.147	0.186	0.486
Natural	0.731	-0.214			0.155
Fibre	0.816				
Sweet		0.696	0.354		0.164
Easy	0.228			0.292	
Salt		0.686			
Satisfying	0.571		0.202	0.369	0.355
Energy	0.610		0.228	0.161	0.343
Fun	0.126	0.155	0.545	0.364	
Kids				0.868	
Soggy			-0.452	0.138	
Economical		-0.257	-0.190	0.410	-0.104
Health	0.839	-0.271			
Family			0.133	0.789	
Calories	-0.157	0.591	0.122		0.179
Plain	-0.115		-0.635	0.100	-0.147
Crisp		0.148	0.465	0.318	
Regular	0.658				
Sugar	-0.177	0.854	0.171		
Fruit	0.339	0.159	0.433	-0.294	0.155
Process	-0.218	0.380	-0.105		-0.181
Quality	0.682	-0.222	0.221	0.190	

Treat	0.234	0.208	0.659	0.278
Boring	-0.150		-0.513	-0.193
Nutritious	0.849	-0.154		

	Factor1	Factor2	Factor3	Factor4	Factor5
SS loadings	5.044	2.583	2.440	2.353	0.711
Proportion Var	0.202	0.103	0.098	0.094	0.028
Cumulative Var	0.202	0.305	0.403	0.497	0.525

Test of the hypothesis that 5 factors are sufficient.
The chi square statistic is 318.9 on 185 degrees of freedom.
The p-value is 3.48e-09

After the rotation we observe that the 5 factors are able to explain about 52% of the data. We infer that based on the latent relation between the variables they are grouped into as “Factors”. The factor loadings explains the weightage of each variable has in the Factors. We can also see that the some of the variables are present in more than one factor, which is known as the ‘Communality’ effect. As a result, the factor scores for all the 235 observations are known,from which prediction or classification can be done.

The variable that depend highly on factor -

Factor 1 -‘State of well being’, Filling,Natural,Fibre,Satisfying,Energy,Health,Regular,Quality,Nutritious

Factor 2 - ‘Nutritional Information’, Sweet,Salt, Calories, Sugar, Process,

Factor 3 - Fun, Soggy, Crisp, Fruit, Treat, Boring,Plain - ‘Experience’

Factor 4 - Kids, Economical,Family - ‘Family’

Factor 5 - Filling, Satisfying, Energy - ‘Supporting variables’