

This is an R HTML document. When you click the **Knit HTML** button a web page will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
# Factor Analysis
#Exploratory Factor Analysis (EFA) is a statistical technique that is used to identify
#the Latent relational structure among a set of variables and narrow down to smaller number
#of variables. This essentially means that the variance of Large number of variables can be
#described by few summary variables, i.e., factors.
# Importing the dataset

#####reading the dataset#####

who <- read.csv("C:/Users/prera/Downloads/Life_Expectancy_Data.csv")
head(who)

##      Country Year      Status Life.expectancy Adult.Mortality infant.deaths
## 1 Afghanistan 2015 Developing           65.0           263           62
## 2 Afghanistan 2014 Developing           59.9           271           64
## 3 Afghanistan 2013 Developing           59.9           268           66
## 4 Afghanistan 2012 Developing           59.5           272           69
## 5 Afghanistan 2011 Developing           59.2           275           71
## 6 Afghanistan 2010 Developing           58.8           279           74
##      Alcohol percentage.expenditure Hepatitis.B Measles BMI under.five.deaths
## 1      0.01           71.279624           65      1154 19.1           83
## 2      0.01           73.523582           62      492 18.6           86
## 3      0.01           73.219243           64      430 18.1           89
## 4      0.01           78.184215           67      2787 17.6           93
## 5      0.01           7.097109           68      3013 17.2           97
## 6      0.01           79.679367           66      1989 16.7          102
##      Polio Total.expenditure Diphtheria HIV.AIDS GDP Population
## 1      6           8.16           65      0.1 584.25921 33736494
## 2     58           8.18           62      0.1 612.69651 327582
## 3     62           8.13           64      0.1 631.74498 31731688
## 4     67           8.52           67      0.1 669.95900 3696958
## 5     68           7.87           68      0.1 63.53723 2978599
## 6     66           9.20           66      0.1 553.32894 2883167
##      thinness..1.19.years thinness.5.9.years Income.composition.of.resources
## 1           17.2           17.3           0.479
## 2           17.5           17.5           0.476
## 3           17.7           17.7           0.470
## 4           17.9           18.0           0.463
## 5           18.2           18.2           0.454
## 6           18.4           18.4           0.448
##      Schooling
## 1      10.1
## 2      10.0
## 3       9.9
## 4       9.8
## 5       9.5
## 6       9.2

#####Dimesion of the dataset#####

dim(who)

## [1] 2938 22

##### TOP 10 DEVELOPED & DEVELOPING Countires #####

status.of.countries <- who[(who$Status %in% c("Developing") & who$Life.expectancy<55) | (who$Status %in% c("Developed") & who$Life.expectancy>80) ,]
dim(status.of.countries)

## [1] 509 22

#View(status.of.countries)

class(status.of.countries)

## [1] "data.frame"

head(status.of.countries)

##      Country Year      Status Life.expectancy Adult.Mortality infant.deaths
## 16 Afghanistan 2000 Developing           54.8           321           88
## 49      Angola 2015 Developing           52.4           335           66
## 50      Angola 2014 Developing           51.7           348           67
## 51      Angola 2013 Developing           51.1           355           69
## 53      Angola 2011 Developing           51.0           361           75
## 54      Angola 2010 Developing           49.6           365           78
##      Alcohol percentage.expenditure Hepatitis.B Measles BMI under.five.deaths
## 16      0.01           10.42496           62      6532 12.2          122
## 49      NA           0.00000           64      118 23.3           98
## 50      8.33           23.96561           64     11699 22.7          101
## 51      8.10           35.95857           77      8523 22.1          105
## 53      8.06           239.89139           72     1449 21.0          115
## 54      7.80           191.65374           77     1190 2.4           121
##      Polio Total.expenditure Diphtheria HIV.AIDS GDP Population
## 16     24           8.20           24      0.1 114.5600 293756
```

```
## 49      7      NA      64      1.9 3695.7937      2785935
## 50     68     3.31     64     2.0 479.3122      2692466
## 51     67     4.26     77     2.3 484.6169      2599834
## 53     73     3.38     71     2.5 4299.1289      24218565
## 54     81     3.39     77     2.5 3529.5348      23369131
##      thinness..1.19.years thinness.5.9.years Income.composition.of.resources
## 16      2.3      2.5      0.338
## 49      8.3      8.2      0.531
## 50      8.5      8.3      0.527
## 51      8.6      8.5      0.523
## 53      8.9      8.8      0.495
## 54      9.1      9.0      0.488
##      Schooling
## 16      5.5
## 49     11.4
## 50     11.4
## 51     11.4
## 53      9.4
## 54      9.0
```

```
#View(status.of.countries)
WHONew<-status.of.countries
#resting the index values
row.names(WHONew) <- NULL
#View(WHONew)
dim(WHONew)
```

```
## [1] 509  22
```

```
##### CLEANING THE DATA #####
# For 347 rows running the for loop for chechking any NA values and replacing it with the mean of the
# particular country.
for(i in 1:347)
{
  if(is.na(WHONew$Alcohol[i]))
  {
    WHONew$Alcohol[i] <- with(WHONew, mean(WHONew$Alcohol[Country == WHONew$Country[i]], na.rm = TRUE))
  }
}
for(i in 1:347)
{
  if(is.na(WHONew$Hepatitis.B[i]))
  {
    WHONew$Hepatitis.B[i] <- with(WHONew, mean(WHONew$Hepatitis.B[Country == WHONew$Country[i]], na.rm = TRUE))
  }
}
for(i in 1:347)
{
  if(is.na(WHONew$Total.expenditure[i]))
  {
    WHONew$Total.expenditure[i] <- with(WHONew, mean(WHONew$Total.expenditure[Country == WHONew$Country[i]], na.rm = TRUE))
  }
}
dim(WHONew)
```

```
## [1] 509  22
```

```
#View(WHONew)

# Deleting the Empty rows where there is no data present.
new.life<- na.omit(WHONew)
dim(new.life)
```

```
## [1] 285  22
```

```
View(new.life)

fac.eff <- new.life[,c(4,5,7,11,16,17,19,20,21,22)]

View(fac.eff)
attach(fac.eff)
head(fac.eff[1])
```

```
##      Life.expectancy
## 1      54.8
## 2      52.4
## 3      51.7
## 4      51.1
## 5      51.0
## 6      49.6
```

```
# Computing Correlation Matrix
corrm.fac.eff <- cor(fac.eff[-1])
corrm.fac.eff
```

```
##      Adult.Mortality      Alcohol      BMI
## Adult.Mortality      1.0000000 -0.4530275 -0.4153265
## Alcohol      -0.4530275      1.0000000      0.5762918
```

```
## BMI -0.4153265 0.5762918 1.0000000
## HIV.AIDS 0.5676648 -0.3364188 -0.2739551
## GDP -0.4023978 0.5137808 0.4557127
## thinness..1.19.years 0.5345618 -0.3924699 -0.5408240
## thinness.5.9.years 0.5538395 -0.3747142 -0.5176442
## Income.composition.of.resources -0.4775719 0.6980880 0.7668355
## Schooling -0.4277133 0.7156089 0.7487213
## HIV.AIDS GDP thinness..1.19.years
## Adult.Mortality 0.5676648 -0.4023978 0.5345618
## Alcohol -0.3364188 0.5137808 -0.3924699
## BMI -0.2739551 0.4557127 -0.5408240
## HIV.AIDS 1.0000000 -0.3260591 0.4341843
## GDP -0.3260591 1.0000000 -0.4571052
## thinness..1.19.years 0.4341843 -0.4571052 1.0000000
## thinness.5.9.years 0.4631060 -0.4526723 0.9188703
## Income.composition.of.resources -0.3450488 0.6115709 -0.6638266
## Schooling -0.2729460 0.5859284 -0.5865751
## thinness.5.9.years
## Adult.Mortality 0.5538395
## Alcohol -0.3747142
## BMI -0.5176442
## HIV.AIDS 0.4631060
## GDP -0.4526723
## thinness..1.19.years 0.9188703
## thinness.5.9.years 1.0000000
## Income.composition.of.resources -0.6422292
## Schooling -0.5678306
## Income.composition.of.resources Schooling
## Adult.Mortality -0.4775719 -0.4277133
## Alcohol 0.6980880 0.7156089
## BMI 0.7668355 0.7487213
## HIV.AIDS -0.3450488 -0.2729460
## GDP 0.6115709 0.5859284
## thinness..1.19.years -0.6638266 -0.5865751
## thinness.5.9.years -0.6422292 -0.5678306
## Income.composition.of.resources 1.0000000 0.9347573
## Schooling 0.9347573 1.0000000
```

You can also embed plots, for example:

```
install.packages("corrplot")

## Installing package into 'C:/Users/prera/OneDrive/Documents/R/win-library/3.6'
## (as 'Lib' is unspecified)

## Error in contrib.url(repos, "source"): trying to use CRAN without setting a mirror

library(corrplot)

## Warning: package 'corrplot' was built under R version 3.6.3

## corrplot 0.84 Loaded

library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##   filter, lag

## The following objects are masked from 'package:base':
##   intersect, setdiff, setequal, union

#For plotting the scatter density plots
library(GGally)

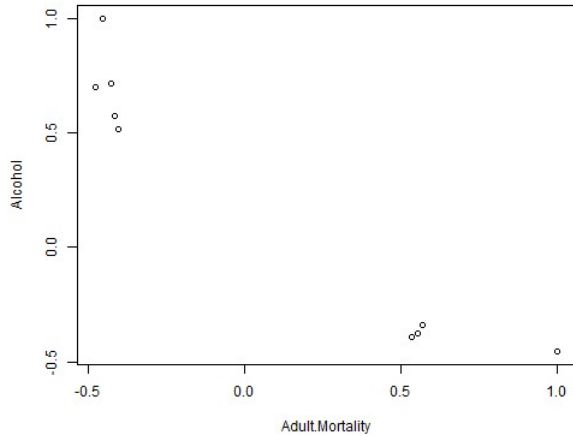
## Loading required package: ggplot2

## Registered S3 method overwritten by 'GGally':
##   method from
##   +.gg ggplot2

##
## Attaching package: 'GGally'

## The following object is masked from 'package:dplyr':
##   nasa
```

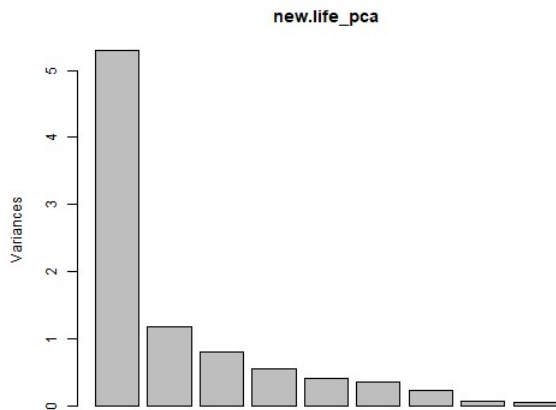
```
plot(corr.fac.eff)
```



```
new.life_pca <- prcomp(fac.eff[-1], scale=TRUE)
summary(new.life_pca)
```

```
## Importance of components:
##          PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation  2.3031  1.0868  0.89854  0.75009  0.64178  0.59893  0.48578
## Proportion of Variance 0.5894  0.1312  0.08971  0.06252  0.04576  0.03986  0.02622
## Cumulative Proportion 0.5894  0.7206  0.81032  0.87283  0.91860  0.95845  0.98467
##          PC8      PC9
## Standard deviation  0.28324  0.24023
## Proportion of Variance 0.00891  0.00641
## Cumulative Proportion 0.99359  1.00000
```

```
plot(new.life_pca)
```



```
# A table containing eigenvalues and %'s accounted, follows. Eigenvalues are the sdev^2
(eigen.fac.eff <- round(new.life_pca$sdev^2,2))
```

```
## [1] 5.30 1.18 0.81 0.56 0.41 0.36 0.24 0.08 0.06
```

```
names(eigen.fac.eff) <- paste("PC",1:9,sep="")
eigen.fac.eff
```

```
## PC1 PC2 PC3 PC4 PC5 PC6 PC7 PC8 PC9
## 5.30 1.18 0.81 0.56 0.41 0.36 0.24 0.08 0.06
```

```
sumlambdas <- sum(eigen.fac.eff)
sumlambdas
```

```
## [1] 9
```

```
cumvar.fac.eff <- cumsum(propvar)
```

```
## Error in eval(expr, envir, enclos): object 'propvar' not found
```

```
propvar <- round(eigen.fac.eff/sumlambdas,2)
propvar
```

```
## PC1 PC2 PC3 PC4 PC5 PC6 PC7 PC8 PC9
## 0.59 0.13 0.09 0.06 0.05 0.04 0.03 0.01 0.01
```

```
cumvar.fac.eff <- cumsum(propvar)
cumvar.fac.eff
```

```
## PC1 PC2 PC3 PC4 PC5 PC6 PC7 PC8 PC9
## 0.59 0.72 0.81 0.87 0.92 0.96 0.99 1.00 1.01
```

```
matlambdas <- rbind(eigen.fac.eff,propvar,cumvar.fac.eff)
matlambdas
```

```
##          PC1 PC2 PC3 PC4 PC5 PC6 PC7 PC8 PC9
## eigen.fac.eff 5.30 1.18 0.81 0.56 0.41 0.36 0.24 0.08 0.06
## propvar      0.59 0.13 0.09 0.06 0.05 0.04 0.03 0.01 0.01
## cumvar.fac.eff 0.59 0.72 0.81 0.87 0.92 0.96 0.99 1.00 1.01
```

```
rownames(matlambdas) <- c("Eigenvalues","Prop. variance","Cum. prop. variance")
rownames(matlambdas)
```

```
## [1] "Eigenvalues"          "Prop. variance"          "Cum. prop. variance"
```

```
eigvec.emp <- new.life_pca$rotation
print(new.life_pca)
```

```
## Standard deviations (1, ..., p=9):
## [1] 2.3031386 1.0867502 0.8985405 0.7500902 0.6417774 0.5989313 0.4857819
## [8] 0.2832368 0.2402331
```

```
## Rotation (n x k) = (9 x 9):
```

```
##          PC1      PC2      PC3      PC4
## Adult.Mortality -0.2963146 0.3933735 0.32256464 -0.18714120
## Alcohol         0.3209942 0.2975146 -0.41755945 0.16611060
## BMI             0.3411230 0.2733771 0.09577648 0.36221151
## HIV.AIDS        -0.2374740 0.5433865 0.46615149 -0.08782235
## GDP             0.3019017 0.1277741 -0.21359316 -0.88233421
## thinness..1.19.years -0.3519585 0.2828611 -0.48138414 0.05968413
## thinness.5.9.years -0.3485108 0.3260066 -0.45961828 0.06859044
## Income.composition.of.resources 0.3960508 0.2543053 0.07153997 0.06224593
## Schooling       0.3782407 0.3456159 0.03749890 0.08843326
##          PC5      PC6      PC7
## Adult.Mortality 0.739488143 -0.21630783 0.15597097
## Alcohol         -0.084880468 -0.62808722 0.45128571
## BMI             0.133005364 0.65476669 0.46983943
## HIV.AIDS        -0.642072242 -0.04957061 0.08119827
## GDP             -0.039747971 0.22195785 0.13095165
## thinness..1.19.years 0.005153132 0.19628897 -0.16025755
## thinness.5.9.years 0.006619857 0.18872320 -0.12723461
## Income.composition.of.resources 0.099567043 -0.01561493 -0.45582342
## Schooling       0.066644790 -0.06233604 -0.52714093
##          PC8      PC9
## Adult.Mortality -0.0204520520 -0.002848616
## Alcohol         -0.0141814025 -0.030675532
## BMI             -0.0246304217 0.027643013
## HIV.AIDS        -0.0186365347 -0.040898855
## GDP             -0.0003984935 0.019994326
## thinness..1.19.years -0.6916551644 -0.134895083
## thinness.5.9.years 0.7075942789 0.061381753
## Income.composition.of.resources 0.1034032951 -0.735428718
## Schooling       -0.0930467252 0.658327024
```

```
#Considering the observation of the standard deviation, we initially tried considering four
#components as the SD after four components are comparatively higher also considering the
#elbow diagram done in the previous section justifies for taking into account four components.
# Taking the first four PCs to generate linear combinations for all the variables with four factors
pcafactors.emp <- eigvec.emp[,1:4]
pcafactors.emp
```

```
##          PC1      PC2      PC3      PC4
## Adult.Mortality -0.2963146 0.3933735 0.32256464 -0.18714120
## Alcohol         0.3209942 0.2975146 -0.41755945 0.16611060
## BMI             0.3411230 0.2733771 0.09577648 0.36221151
## HIV.AIDS        -0.2374740 0.5433865 0.46615149 -0.08782235
## GDP             0.3019017 0.1277741 -0.21359316 -0.88233421
## thinness..1.19.years -0.3519585 0.2828611 -0.48138414 0.05968413
## thinness.5.9.years -0.3485108 0.3260066 -0.45961828 0.06859044
## Income.composition.of.resources 0.3960508 0.2543053 0.07153997 0.06224593
## Schooling       0.3782407 0.3456159 0.03749890 0.08843326
```

```
# Multiplying each column of the eigenvector's matrix by the square-root of the corresponding eigenvalue in order to get the factor loadings
unrot.fact.emp <- sweep(pcafactors.emp,MARGIN=2,new.life_pca$sdev[1:4],`*`)
unrot.fact.emp
```

```
##          PC1      PC2      PC3      PC4
## Adult.Mortality -0.6824537 0.4274988 0.28983739 -0.14037279
## Alcohol         0.7392942 0.3233240 -0.37519408 0.12459794
## BMI             0.7856536 0.2970926 0.08605905 0.27169132
## HIV.AIDS        -0.5469356 0.5905254 0.41885599 -0.06587469
## GDP             0.6953215 0.1388585 -0.19192210 -0.66183028
## thinness..1.19.years -0.8106092 0.3073993 -0.43254314 0.04476849
## thinness.5.9.years -0.8026688 0.3542877 -0.41298564 0.05144902
## Income.composition.of.resources 0.9121599 0.2763663 0.06428156 0.04669007
## Schooling       0.8711407 0.3755981 0.03369428 0.06633293
```

```
# Computing communalities
communalities.emp <- rowSums(unrot.fact.emp^2)
communalities.emp
```

```
##          Adult.Mortality          Alcohol
##          0.7522084          0.8073897
##          BMI          HIV.AIDS
##          0.7867380          0.8276387
##          GDP          thinness..1.19.years
##          0.9776071          0.9406794
##          thinness.5.9.years Income.composition.of.resources
##          0.9430011          0.9147262
##          Schooling
##          0.9054955
```

```
# Performing the varimax rotation. The default in the varimax function is norm=TRUE thus, Kaiser normalization is carried out
rot.fact.emp <- varimax(unrot.fact.emp)
View(unrot.fact.emp)
rot.fact.emp
```

```
## $loadings
##
## Loadings:
##          PC1      PC2      PC3      PC4
## Adult.Mortality -0.288 0.759 -0.293
## Alcohol         0.795 -0.330 -0.255
## BMI             0.812 -0.116 0.337
## HIV.AIDS        0.877 -0.194 0.123
## GDP             0.340 -0.179 0.213 -0.886
## thinness..1.19.years -0.299 0.261 -0.872 0.150
## thinness.5.9.years -0.265 0.300 -0.872 0.151
## Income.composition.of.resources 0.805 -0.152 0.411 -0.271
## Schooling       0.849 0.327 -0.262
##
##          PC1      PC2      PC3      PC4
## SS loadings 3.026 1.688 2.080 1.062
## Proportion Var 0.336 0.188 0.231 0.118
## Cumulative Var 0.336 0.524 0.755 0.873
##
## $rotmat
##          [,1]      [,2]      [,3]      [,4]
## [1,] 0.6848287 -0.4010275 0.52019938 -0.3155618
## [2,] 0.6184353 0.6653274 -0.39108202 -0.1480948
## [3,] -0.1140636 0.6018484 0.75535942 0.2328092
## [4,] 0.3681535 -0.1851886 -0.07667864 0.9079034
```

```
# The print method of varimax omits Loadings less than abs(0.1). In order to display all the Loadings, it is necessary to ask explicitly the contents
fact.load.emp <- rot.fact.emp$loadings[1:9,1:4]
fact.load.emp
```

```
##          PC1      PC2      PC3      PC4
## Adult.Mortality -0.28772215 0.75854298 -0.29250406 0.09207786
## Alcohol         0.79491204 -0.33024507 -0.03482619 -0.25540135
## BMI             0.81197857 -0.11592453 0.33668157 -0.02521530
## HIV.AIDS        -0.08138367 0.87651605 -0.19402144 0.12284395
## GDP             0.34028725 -0.17940131 0.21317882 -0.88554030
## thinness..1.19.years -0.29920268 0.26098181 -0.87205507 0.15021846
## thinness.5.9.years -0.26453881 0.29952710 -0.87200099 0.15138731
## Income.composition.of.resources 0.80544487 -0.15188587 0.41139868 -0.27141580
## Schooling       0.84944267 -0.09146092 0.32664215 -0.26245460
```

```
scale.emp <- scale(fac.eff[-1])
head(scale.emp)
```

```
## Adult.Mortality Alcohol BMI HIV.AIDS GDP
## 1 0.3025231 -1.46317163 -0.9386706 -0.7355316 -0.4957402
## 2 0.3661937 -0.04793103 -0.3991077 -0.5888599 -0.3173613
## 3 0.4253164 0.65759545 -0.4282733 -0.5807114 -0.4775722
## 4 0.4571517 0.59896848 -0.4574388 -0.5562661 -0.4773079
## 5 0.4844391 0.58877248 -0.5109090 -0.5399693 -0.2873095
## 6 0.5026307 0.52249851 -1.4150414 -0.5399693 -0.3256425
## thinness..1.19.years thinness.5.9.years Income.composition.of.resources
## 1 -0.6650011 -0.6259672 -0.8934474
```

```
## 2      0.8032389      0.7360523      -0.1276533
## 3      0.8521802      0.7599473      -0.1435246
## 4      0.8766509      0.8077375      -0.1593960
## 5      0.9500629      0.8794227      -0.2704957
## 6      0.9990042      0.9272129      -0.2982706
##      Schooling
## 1 -1.30128198
## 2  0.03496574
## 3  0.03496574
## 4  0.03496574
## 5 -0.41799959
## 6 -0.50859265
```

```
#as.matrix(scale.emp)%fact.Load.emp%%solve(t(fact.Load.emp)%*fact.Load.emp)
```

```
library(psych)
```

```
## Warning: package 'psych' was built under R version 3.6.3
```

```
##
## Attaching package: 'psych'
```

```
## The following objects are masked from 'package:ggplot2':
##
##      %%, alpha
```

```
#install.packages("psych", Lib="/Library/Frameworks/R.framework/Versions/3.5/Resources/Library")
library(psych)
# On 4 factors analysis
fit.pc <- principal(fac.eff[-1], nfactors=4, rotate="varimax")
# On 3 factors analysis
fit.pc1 <- principal(fac.eff[-1], nfactors=3, rotate="varimax")

fit.pc
```

```
## Principal Components Analysis
## Call: principal(r = fac.eff[-1], nfactors = 4, rotate = "varimax")
## Standardized loadings (pattern matrix) based upon correlation matrix
##
##      RC1  RC3  RC2  RC4  h2  u2 com
## Adult.Mortality      -0.29  0.29  0.76 -0.09 0.75 0.248 1.6
## Alcohol              0.79  0.03 -0.33  0.26 0.81 0.193 1.6
## BMI                  0.81 -0.34 -0.12  0.03 0.79 0.213 1.4
## HIV.AIDS             -0.08  0.19  0.88 -0.12 0.83 0.172 1.2
## GDP                  0.34 -0.21 -0.18  0.89 0.98 0.022 1.5
## thinness..1.19.years -0.30  0.87  0.26 -0.15 0.94 0.059 1.5
## thinness.5.9.years   -0.26  0.87  0.30 -0.15 0.94 0.057 1.5
## Income.composition.of.resources 0.81 -0.41 -0.15  0.27 0.91 0.085 1.8
## Schooling            0.85 -0.33 -0.09  0.26 0.91 0.095 1.5
##
##      RC1  RC3  RC2  RC4
## SS loadings      3.03 2.08 1.69 1.06
## Proportion Var    0.34 0.23 0.19 0.12
## Cumulative Var    0.34 0.57 0.75 0.87
## Proportion Explained 0.39 0.26 0.21 0.14
## Cumulative Proportion 0.39 0.65 0.86 1.00
##
## Mean item complexity = 1.5
## Test of the hypothesis that 4 components are sufficient.
##
## The root mean square of the residuals (RMSR) is 0.05
## with the empirical chi square 47.49 with prob < 1.5e-08
##
## Fit based upon off diagonal values = 0.99
```

```
fit.pc1
```

```
## Principal Components Analysis
## Call: principal(r = fac.eff[-1], nfactors = 3, rotate = "varimax")
## Standardized loadings (pattern matrix) based upon correlation matrix
##
##      RC1  RC3  RC2  h2  u2 com
## Adult.Mortality      -0.29  0.30  0.75 0.73 0.267 1.7
## Alcohol              0.83  0.01 -0.31 0.79 0.208 1.3
## BMI                  0.75 -0.38 -0.07 0.71 0.287 1.5
## HIV.AIDS             -0.12  0.19  0.88 0.82 0.177 1.1
## GDP                  0.64 -0.18 -0.31 0.54 0.460 1.6
## thinness..1.19.years -0.31  0.88  0.27 0.94 0.061 1.5
## thinness.5.9.years   -0.28  0.87  0.31 0.94 0.060 1.5
## Income.composition.of.resources 0.84 -0.44 -0.15 0.91 0.087 1.6
## Schooling            0.88 -0.35 -0.08 0.90 0.099 1.3
##
##      RC1  RC3  RC2
## SS loadings      3.42 2.15 1.73
## Proportion Var    0.38 0.24 0.19
## Cumulative Var    0.38 0.62 0.81
## Proportion Explained 0.47 0.29 0.24
## Cumulative Proportion 0.47 0.76 1.00
##
## Mean item complexity = 1.4
## 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 61.3 with prob < 1.3e-08
##
## Fit based upon off diagonal values = 0.99
```

```
round(fit.pc$values, 3)
```

```
## [1] 5.304 1.181 0.807 0.563 0.412 0.359 0.236 0.080 0.058
```

```
round(fit.pc1$values, 3)
```

```
## [1] 5.304 1.181 0.807 0.563 0.412 0.359 0.236 0.080 0.058
```

```
fit.pc$loadings
```

```
##
## Loadings:
##              RC1   RC3   RC2   RC4
## Adult.Mortality -0.288 0.293 0.759
## Alcohol          0.795        -0.330 0.255
## BMI              0.812 -0.337 -0.116
## HIV.AIDS         0.194 0.877 -0.123
## GDP              0.340 -0.213 -0.179 0.886
## thinness..1.19.years -0.299 0.872 0.261 -0.150
## thinness.5.9.years -0.265 0.872 0.300 -0.151
## Income.composition.of.resources 0.805 -0.411 -0.152 0.271
## Schooling        0.849 -0.327        0.262
##
##              RC1   RC3   RC2   RC4
## SS loadings  3.026 2.080 1.688 1.062
## Proportion Var 0.336 0.231 0.188 0.118
## Cumulative Var 0.336 0.567 0.755 0.873
```

```
fit.pc1$loadings
```

```
##
## Loadings:
##              RC1   RC3   RC2
## Adult.Mortality -0.293 0.303 0.745
## Alcohol          0.833        -0.314
## BMI              0.753 -0.377
## HIV.AIDS         -0.115 0.192 0.879
## GDP              0.643 -0.176 -0.308
## thinness..1.19.years -0.311 0.876 0.272
## thinness.5.9.years -0.280 0.874 0.312
## Income.composition.of.resources 0.838 -0.436 -0.146
## Schooling        0.877 -0.354
##
##              RC1   RC3   RC2
## SS loadings  3.419 2.149 1.725
## Proportion Var 0.380 0.239 0.192
## Cumulative Var 0.380 0.619 0.810
```

```
# Loadings with more digits
for (i in c(1,3,2,4)) { print(fit.pc$loadings[[1,i]])} # 4 components
```

```
## [1] -0.2877222
## [1] 0.758543
## [1] 0.2925041
## [1] -0.09207786
```

```
for (i in c(1,3,2,4)) { print(fit.pc1$loadings[[1,i]])} # 3 components
```

```
## [1] -0.2925753
## [1] 0.745044
## [1] 0.3030066
```

```
## Error in fit.pc1$loadings[[1, i]]: subscript out of bounds
```

```
# Communalities
fit.pc$communality
```

```
##              Adult.Mortality              Alcohol
##              0.7522084              0.8073897
##              BMI              HIV.AIDS
##              0.7867380              0.8276387
##              GDP              thinness..1.19.years
##              0.9776071              0.9406794
##              thinness.5.9.years Income.composition.of.resources
##              0.9430011              0.9147262
##              Schooling
##              0.9054955
```



```
fit.pc1$communality
```

```
##          Adult.Mortality          Alcohol
##          0.7325039          0.7918650
##          BMI          HIV.AIDS
##          0.7129218          0.8232992
##          GDP          thinness..1.19.years
##          0.5395878          0.9386752
##          thinness.5.9.years Income.composition.of.resources
##          0.9403541          0.9125462
##          Schooling
##          0.9010954
```

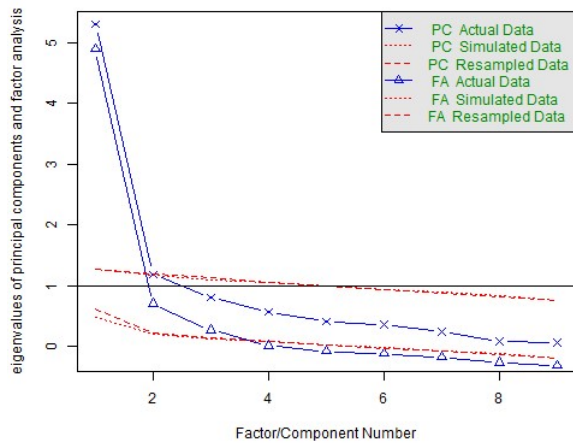
```
# Rotated factor scores, Notice the columns ordering: RC1, RC3, RC2 and RC4
head(fit.pc$scores)
```

```
##          RC1          RC3          RC2          RC4
## 1 -1.87231179 -1.2526059 -0.3875991 -0.21184580
## 2  0.03201845  0.9441829 -0.4093176 -0.12371144
## 3  0.34852090  1.2559007 -0.5349032 -0.29552668
## 4  0.32354251  1.2701292 -0.4984147 -0.27534799
## 5  0.08707079  1.3592718 -0.5480157 -0.01157608
## 6 -0.33601387  1.4056528 -0.6285787  0.29479789
```

```
# Play with FA utilities
```

```
fa.parallel(fac.eff[-1]) # See factor recommendation
```

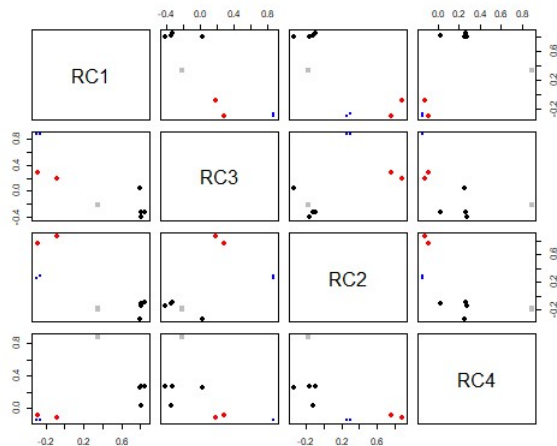
### Parallel Analysis Scree Plots



```
## Parallel analysis suggests that the number of factors = 3 and the number of components = 1
```

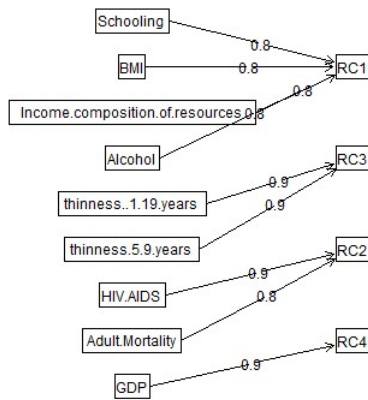
```
fa.plot(fit.pc) # See Correlations within Factors
```

### Principal Component Analysis



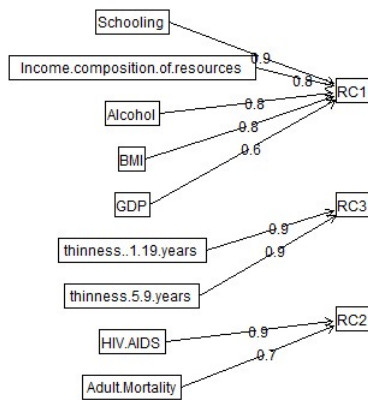
```
fa.diagram(fit.pc)
```

### Components Analysis



```
fa.diagram(fit.pcl)# Visualize the relationship for 4 components analysis
```

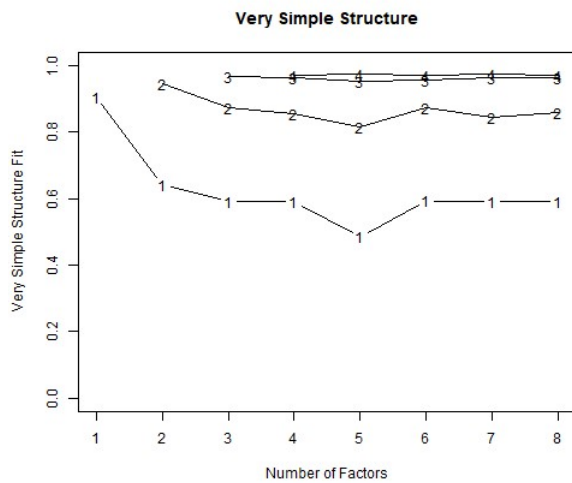
### Components Analysis



#We have observed 9 variables schooling, BMI, income composition of resources, Alcohol, #thinness 5.9 years, thinness 1.91 years, HIV/AIDS, Adult Mortality, GDP. We hypothesize #that there are four unobserved latent factors (RC1, RC2, RC3, RC4) that underly the observed #variables as described in this diagram. Schooling, BMI, income composition of resources, #Alcohol Loads on RC1 with Loadings 0.8, 0.8, 0.8 and 0.9 respectively. Thinness 5.9 years, #thinness 1.91 years Loads on RC2 with Loadings 0.9 and 0.9. HIV/AIDS, Adult Mortality on #RC3 with Loadings 0.9 and 0.8. GDP Loads on RC4 with Loadings of 0.9.

#In a three-component analysis, we see that the common variance for the GDP and adult mortality #is comparatively Less considering the four component analysis for the same. As the unique #variance of 0.4 GDP and unique variance of 0.3 Adult mortality is given away in 3 component #analysis. Where as, in 4 component analysis we observe the unique variance of GDP, Adult Mortality #is only 0.1 and 0.2 respectively is given away. Therefore, we take Component as 4 for analysis.

```
vss(fac.eff[-1]) # See Factor recommendations for a simple structure
```



```
##
## Very Simple Structure
## Call: vss(x = fac.eff[-1])
## VSS complexity 1 achieves a maximum of 0.9 with 1 factors
## VSS complexity 2 achieves a maximum of 0.94 with 2 factors
##
## The Velicer MAP achieves a minimum of 0.08 with 3 factors
## BIC achieves a minimum of NA with 3 factors
## Sample Size adjusted BIC achieves a minimum of NA with 3 factors
##
## Statistics by number of factors
## vss1 vss2 map dof chisq prob sqresid fit RMSEA BIC SABIC complex
## 1 0.90 0.00 0.110 27 6.7e+02 1.4e-123 2.95 0.90 0.289 516.0 601.6 1.0
## 2 0.64 0.94 0.091 19 1.6e+02 1.6e-24 1.73 0.94 0.162 53.7 113.9 1.4
## 3 0.59 0.87 0.084 12 1.0e+01 6.1e-01 1.02 0.97 0.000 -57.8 -19.7 1.5
## 4 0.59 0.86 0.140 6 7.3e+00 2.9e-01 0.89 0.97 0.027 -26.6 -7.6 1.6
## 5 0.49 0.82 0.257 1 8.6e-01 3.5e-01 0.74 0.98 0.000 -4.8 -1.6 1.7
## 6 0.59 0.87 0.508 -3 1.1e-05 NA 0.86 0.97 NA NA NA 1.6
## 7 0.59 0.84 0.448 -6 1.5e-07 NA 0.83 0.97 NA NA NA 1.7
## 8 0.59 0.86 1.000 -8 0.0e+00 NA 0.83 0.97 NA NA NA 1.7
## eChisq SRMR eCRMS eBIC
## 1 2.3e+02 1.1e-01 0.1215 74.5
## 2 5.7e+01 5.3e-02 0.0724 -50.6
## 3 1.5e+00 8.5e-03 0.0148 -66.3
## 4 3.1e-01 3.9e-03 0.0095 -33.6
## 5 9.4e-03 6.8e-04 0.0041 -5.6
## 6 8.1e-07 6.3e-06 NA NA
## 7 1.6e-08 8.9e-07 NA NA
## 8 2.5e-17 3.5e-11 NA NA
```