

# Escalamiento

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Veamos cómo se ve la informacion.

```
##          ID          Year_Birth      Education      Marital_Status
##  Min.      :    0      Min.      :1893      Length:2240      Length:2240
##  1st Qu.: 2828      1st Qu.:1959      Class :character      Class :character
##  Median : 5458      Median :1970      Mode  :character      Mode  :character
##  Mean   : 5592      Mean   :1969
##  3rd Qu.: 8428      3rd Qu.:1977
##  Max.   :11191      Max.   :1996
##
##          Income          Kidhome          Teenhome          Dt_Customer
##  Min.      : 1730      Min.      :0.0000      Min.      :0.0000      Length:2240
##  1st Qu.: 35303      1st Qu.:0.0000      1st Qu.:0.0000      Class :character
##  Median : 51382      Median :0.0000      Median :0.0000      Mode  :character
##  Mean   : 52247      Mean   :0.4442      Mean   :0.5062
##  3rd Qu.: 68522      3rd Qu.:1.0000      3rd Qu.:1.0000
##  Max.   :666666      Max.   :2.0000      Max.   :2.0000
##  NA's    :24
##          Recency          MntWines          MntFruits          MntMeatProducts
##  Min.      : 0.00      Min.      : 0.00      Min.      : 0.0      Min.      : 0.0
##  1st Qu.:24.00      1st Qu.: 23.75      1st Qu.: 1.0      1st Qu.: 16.0
##  Median :49.00      Median : 173.50      Median : 8.0      Median : 67.0
##  Mean   :49.11      Mean   : 303.94      Mean   : 26.3      Mean   : 166.9
##  3rd Qu.:74.00      3rd Qu.: 504.25      3rd Qu.: 33.0      3rd Qu.: 232.0
##  Max.   :99.00      Max.   :1493.00      Max.   :199.0      Max.   :1725.0
##
##  MntFishProducts  MntSweetProducts  MntGoldProds  NumDealsPurchases
##  Min.      : 0.00      Min.      : 0.00      Min.      : 0.00      Min.      : 0.000
##  1st Qu.: 3.00      1st Qu.: 1.00      1st Qu.: 9.00      1st Qu.: 1.000
##  Median : 12.00      Median : 8.00      Median : 24.00      Median : 2.000
##  Mean   : 37.53      Mean   : 27.06      Mean   : 44.02      Mean   : 2.325
##  3rd Qu.: 50.00      3rd Qu.: 33.00      3rd Qu.: 56.00      3rd Qu.: 3.000
##  Max.   :259.00      Max.   :263.00      Max.   :362.00      Max.   :15.000
##
##  NumWebPurchases  NumCatalogPurchases  NumStorePurchases  NumWebVisitsMonth
##  Min.      : 0.000      Min.      : 0.000      Min.      : 0.00      Min.      : 0.000
##  1st Qu.: 2.000      1st Qu.: 0.000      1st Qu.: 3.00      1st Qu.: 3.000
##  Median : 4.000      Median : 2.000      Median : 5.00      Median : 6.000
##  Mean   : 4.085      Mean   : 2.662      Mean   : 5.79      Mean   : 5.317
##  3rd Qu.: 6.000      3rd Qu.: 4.000      3rd Qu.: 8.00      3rd Qu.: 7.000
##  Max.   :27.000      Max.   :28.000      Max.   :13.00      Max.   :20.000
##
##  AcceptedCmp3      AcceptedCmp4      AcceptedCmp5      AcceptedCmp1
```

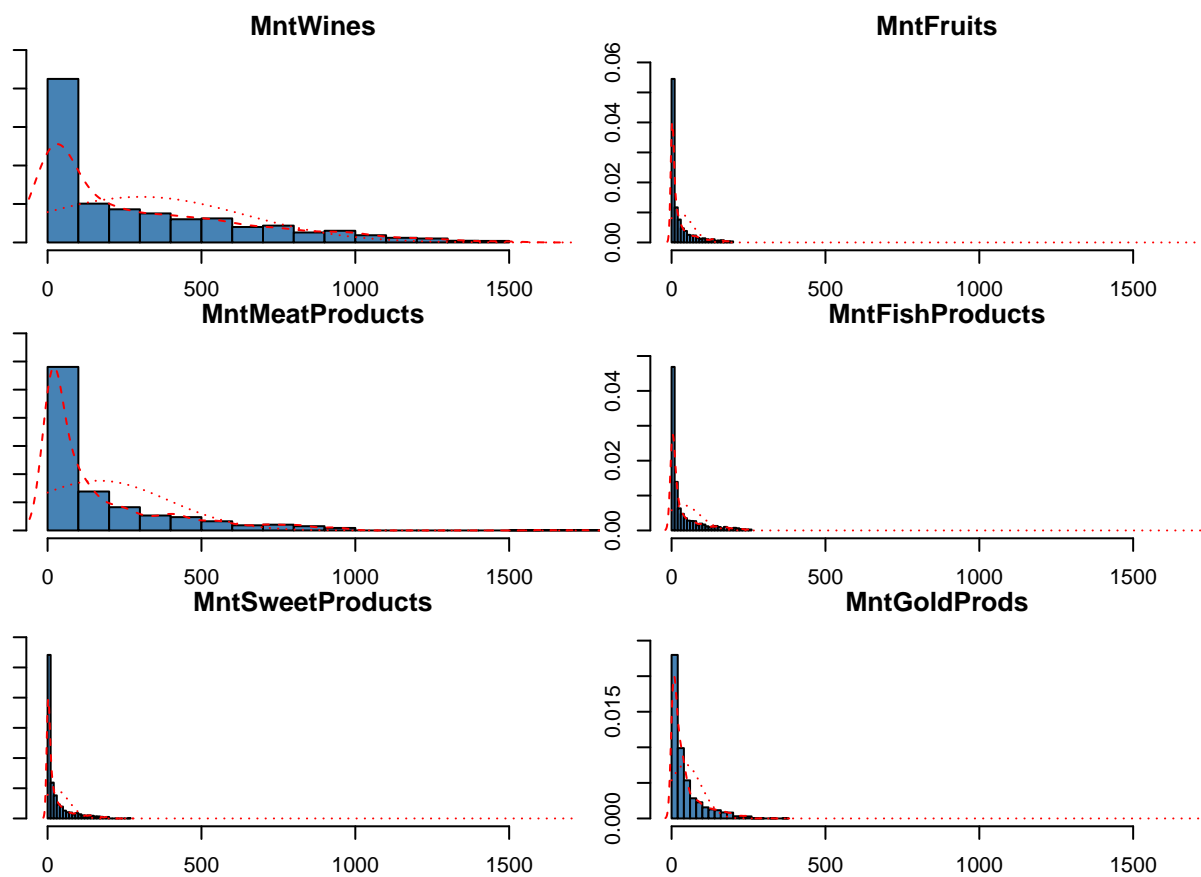
```
## Min. :0.00000 Min. :0.00000 Min. :0.00000 Min. :0.00000
## 1st Qu.:0.00000 1st Qu.:0.00000 1st Qu.:0.00000 1st Qu.:0.00000
## Median :0.00000 Median :0.00000 Median :0.00000 Median :0.00000
## Mean :0.07277 Mean :0.07455 Mean :0.07277 Mean :0.06429
## 3rd Qu.:0.00000 3rd Qu.:0.00000 3rd Qu.:0.00000 3rd Qu.:0.00000
## Max. :1.00000 Max. :1.00000 Max. :1.00000 Max. :1.00000
##
## AcceptedCmp2 Complain Z_CostContact Z_Revenue
## Min. :0.00000 Min. :0.000000 Min. :3 Min. :11
## 1st Qu.:0.00000 1st Qu.:0.000000 1st Qu.:3 1st Qu.:11
## Median :0.00000 Median :0.000000 Median :3 Median :11
## Mean :0.01339 Mean :0.009375 Mean :3 Mean :11
## 3rd Qu.:0.00000 3rd Qu.:0.000000 3rd Qu.:3 3rd Qu.:11
## Max. :1.00000 Max. :1.000000 Max. :3 Max. :11
##
## Response
## Min. :0.0000
## 1st Qu.:0.0000
## Median :0.0000
## Mean :0.1491
## 3rd Qu.:0.0000
## Max. :1.0000
##
```

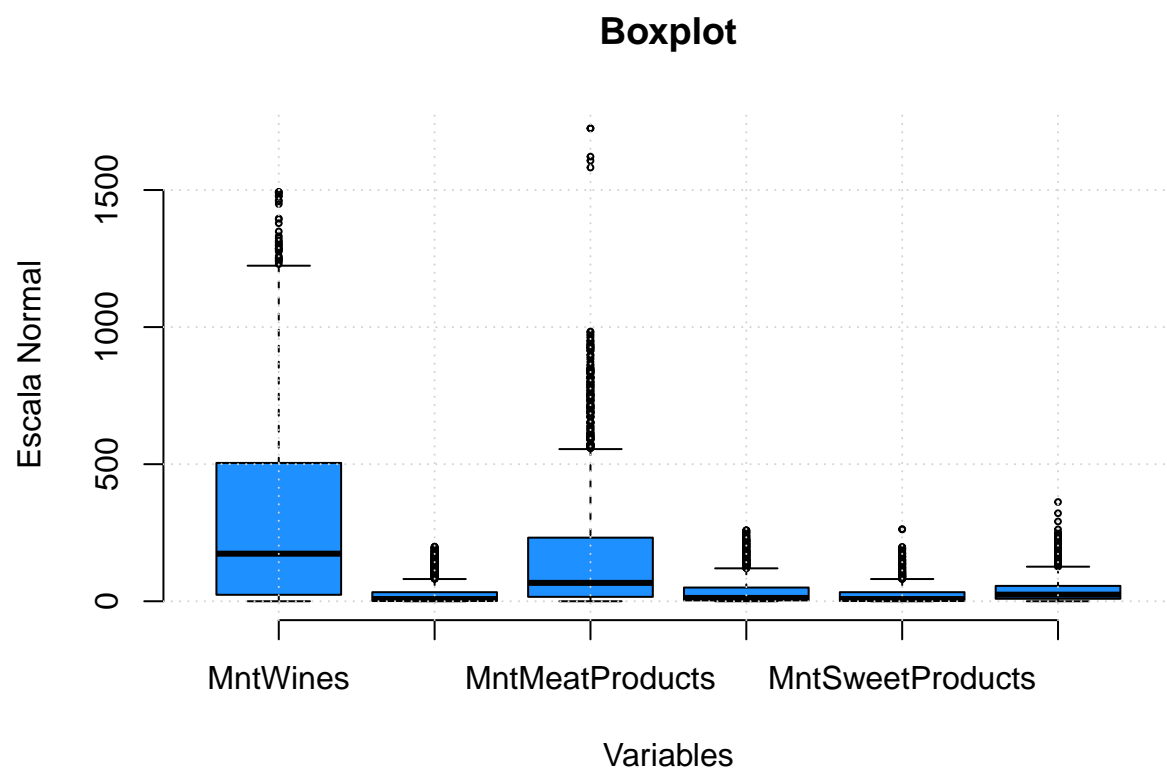
Ya que vimos un resumen rapido de la informacion lo que sigue es tomar las variables que nos sirvan de auxiliares y las que nos sirvan para el analisis.

Extraemos los datos.

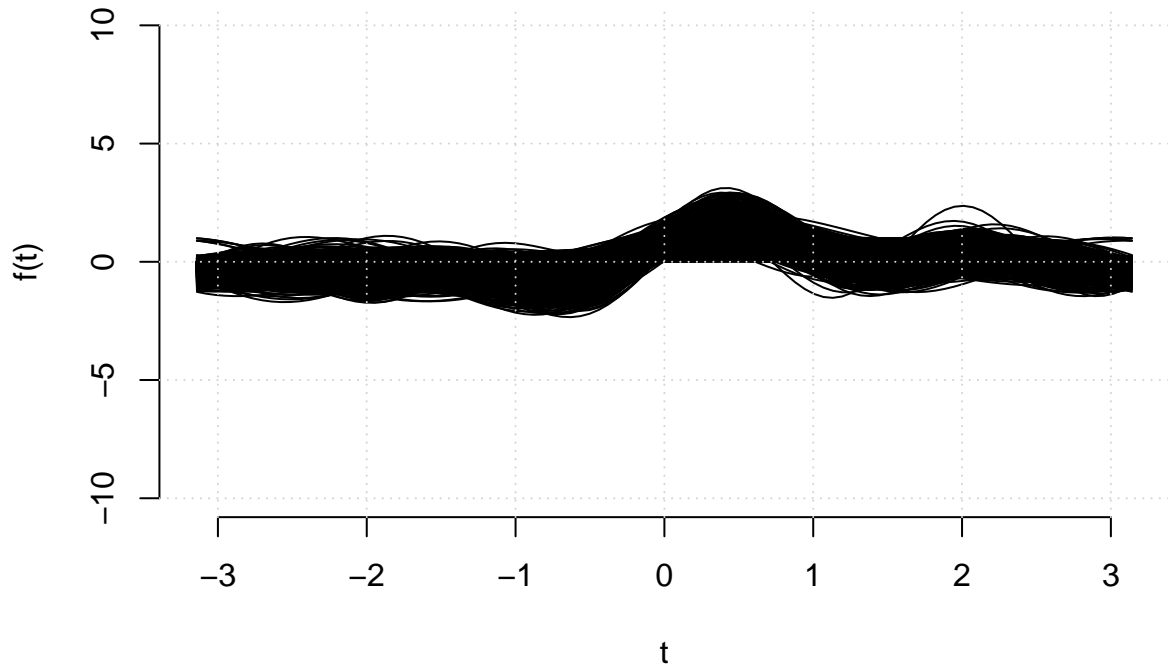
```
## # A tibble: 6 x 7
## ID MntWines MntFruits MntMeatProducts MntFishProducts MntSweetProducts
## <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 5524 635 88 546 172 88
## 2 2174 11 1 6 2 1
## 3 4141 426 49 127 111 21
## 4 6182 11 4 20 10 3
## 5 5324 173 43 118 46 27
## 6 7446 520 42 98 0 42
## # i 1 more variable: MntGoldProds <dbl>
```

Veamos las estadisticas multivariadas con histogramas y box plot.





## Grafico Andrews

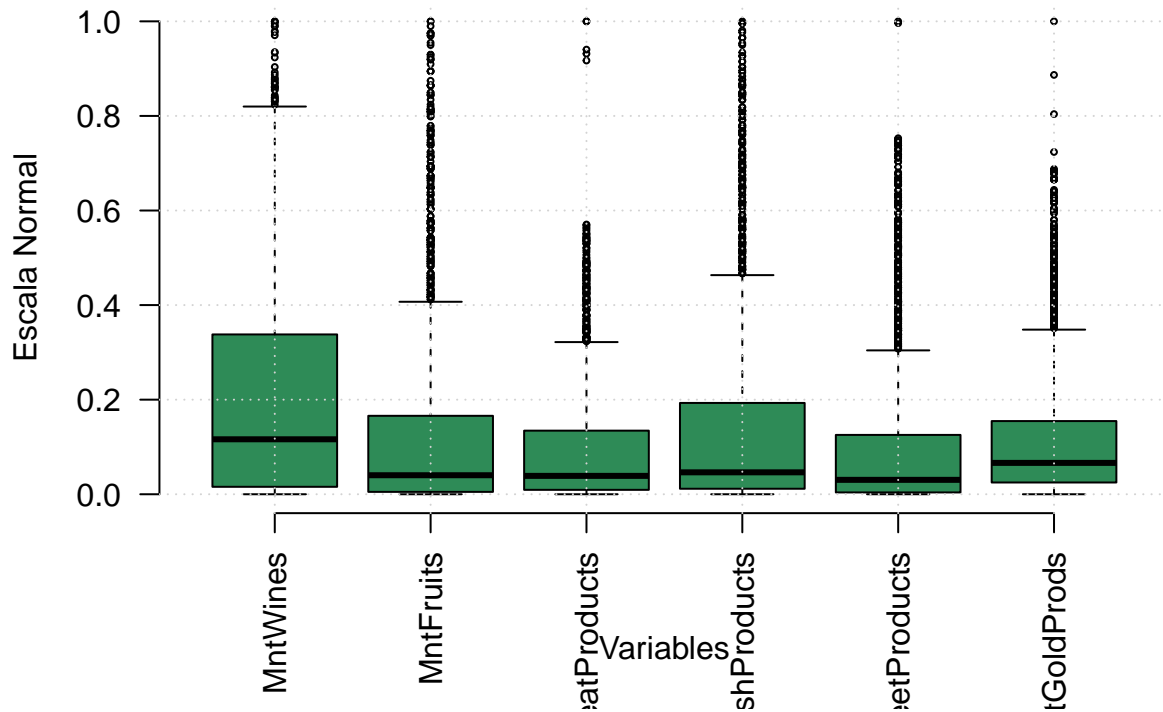


El paso siguiente para nuestro analisis es la normalización de nuestros datos, este paso es muy importante ya que lo obtenido en este punto servirá despues para obtener la escala de lickert que se definirá.

MntWines	MntFruits	MntMeatProducts	MntFishProducts	MntSweetProducts	MntGoldProds
0.4253182	0.4422111	0.3165217	0.6640927	0.3346008	0.2430939
0.0073677	0.0050251	0.0034783	0.0077220	0.0038023	0.0165746
0.2853315	0.2462312	0.0736232	0.4285714	0.0798479	0.1160221
0.0073677	0.0201005	0.0115942	0.0386100	0.0114068	0.0138122
0.1158741	0.2160804	0.0684058	0.1776062	0.1026616	0.0414365
0.3482920	0.2110553	0.0568116	0.0000000	0.1596958	0.0386740

Con los cambios realizados a los datos veremos cómo cambió esto con unos boxplot, en estos vemos que cambió la escala que teníamos antes y tenemos valores fuera del rango del tercer cuantil, lo cual por la forma en que se ven los puntos nos dice que pueden ser valores extremos.

## Caja y Bigotes



## 1 Matriz de datos con escala ordinales

Definicion de rangos para la escala de lickert

0-20 -> 1

21:40 -> 2

40:60 -> 3

61:80 -> 4

81:100 -> 5

Matriz de correlación.

```
R <- cor(datos_new, method = "spearman")
R
```

```
##          MntWines MntFruits MntMeatProducts MntFishProducts
## MntWines      1.000000  0.5169842      0.8190281      0.5266061
## MntFruits      0.5169842  1.0000000      0.7129579      0.7037176
## MntMeatProducts 0.8190281  0.7129579      1.0000000      0.7245835
## MntFishProducts 0.5266061  0.7037176      0.7245835      1.0000000
## MntSweetProducts 0.5222219  0.6940737      0.7058843      0.7016569
## MntGoldProds   0.5733340  0.5660507      0.6346116      0.5631088
##
##          MntSweetProducts MntGoldProds
## MntWines          0.5222219      0.5733340
## MntFruits          0.6940737      0.5660507
```

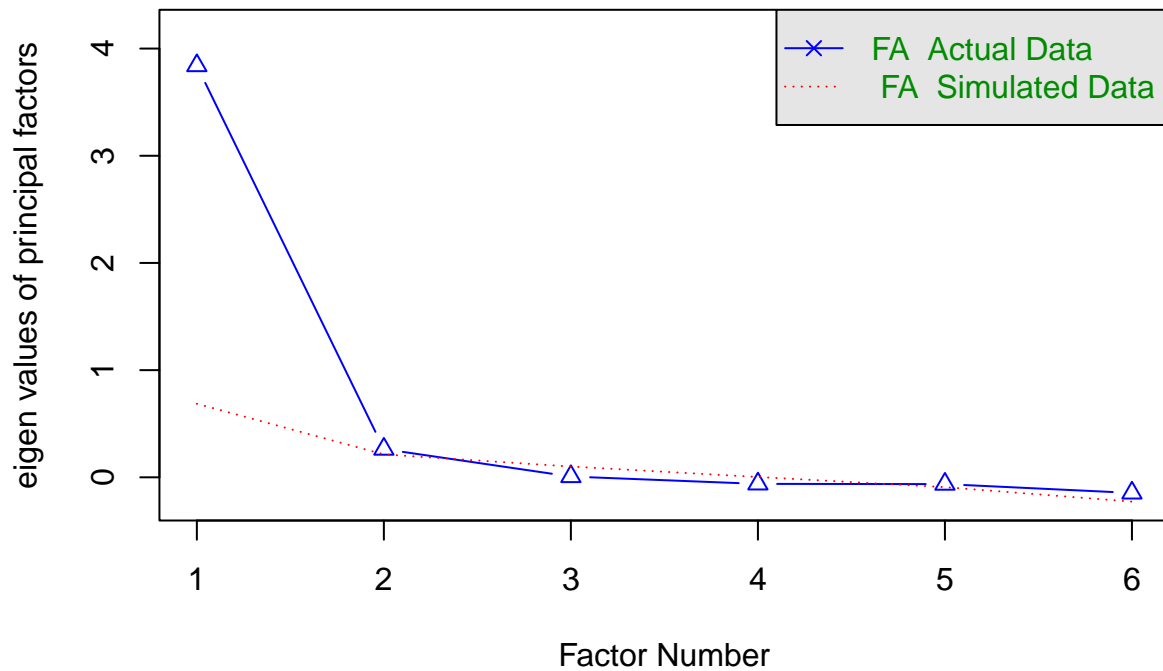
```
## MntMeatProducts      0.7058843      0.6346116
## MntFishProducts      0.7016569      0.5631088
## MntSweetProducts     1.0000000      0.5512815
## MntGoldProds         0.5512815      1.0000000

## [1] TRUE

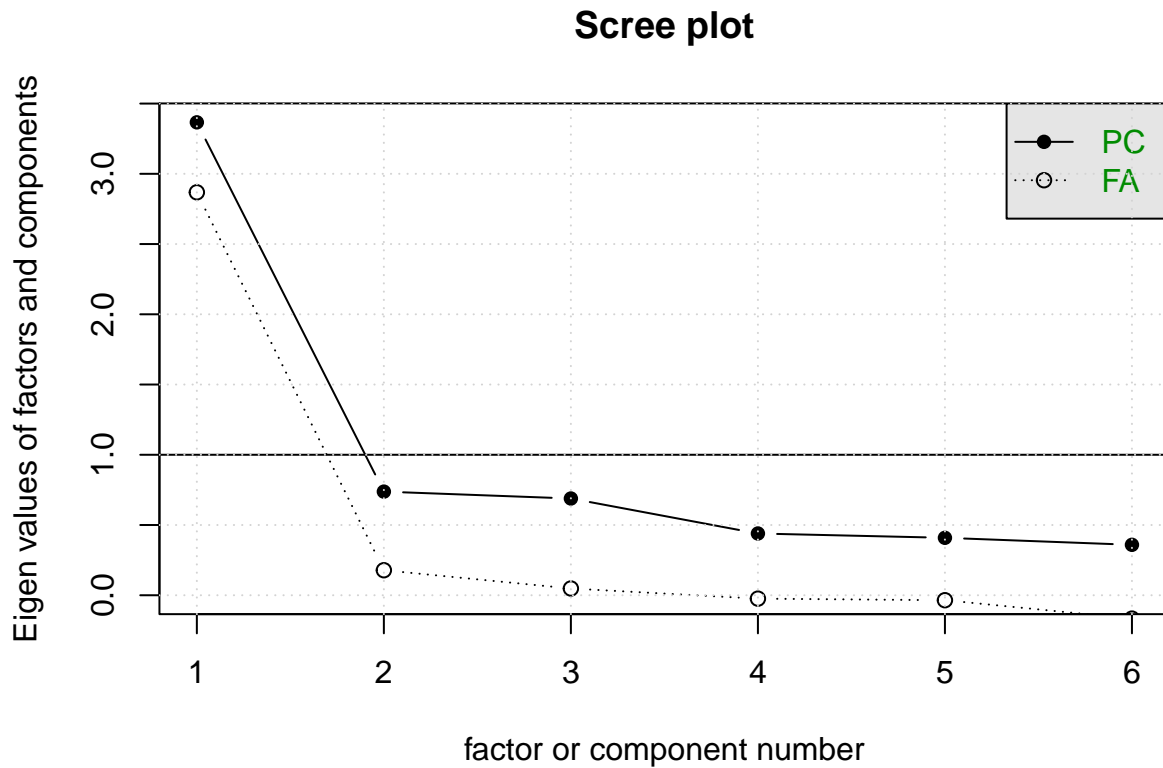
## Kaiser-Meyer-Olkin factor adequacy
## Call: KMO(r = R)
## Overall MSA = 0.86
## MSA for each item =
##      MntWines      MntFruits MntMeatProducts MntFishProducts
##      0.77         0.90         0.79          0.90
## MntSweetProducts MntGoldProds
##      0.91         0.95
```

Del resultados de la prueba de Bartlet tenemos que rechazamos la hipótesis nula, por lo que sí tenemos correlación entre las variables, además con la prueba del KMO obtuvimos un valor de 0.86 lo que significa que tenemos un valor Bueno por lo que sí se puede hacer análisis factorial.

## Parallel Analysis Scree Plots



```
## Parallel analysis suggests that the number of factors = 2 and the number of components = NA
```



Con el screeplot lo que vemos es que solo tenemos un factor significativo por lo que se usará ese valor para el análisis.

```
## Factor Analysis using method = pa
## Call: fa(r = R, nfactors = k, rotate = "none", fm = "pa")
## Standardized loadings (pattern matrix) based upon correlation matrix
##           V PA1  h2  u2 com
## MntMeatProducts 3 0.93 0.86 0.14 1
## MntFishProducts 4 0.81 0.66 0.34 1
## MntFruits        2 0.80 0.64 0.36 1
## MntSweetProducts 5 0.80 0.63 0.37 1
## MntWines         1 0.74 0.55 0.45 1
## MntGoldProds     6 0.71 0.50 0.50 1
##
##           PA1
## SS loadings 3.84
## Proportion Var 0.64
##
## Mean item complexity = 1
## Test of the hypothesis that 1 factor is sufficient.
##
## df null model = 15 with the objective function = 4.32 0.3
## df of the model are 9 and the objective function was 0.46
## 0.3
## The root mean square of the residuals (RMSR) is 0.06
## The df corrected root mean square of the residuals is 0.07
## 0.3
```

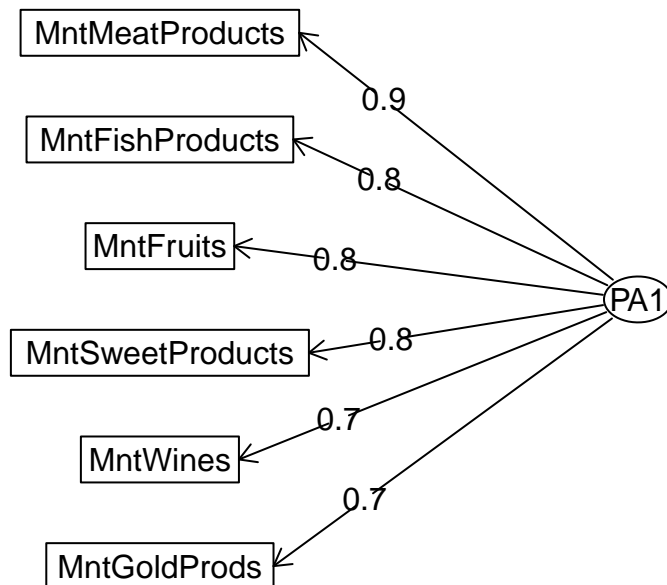


```

## Fit based upon off diagonal values = 0.99
## Measures of factor score adequacy
##
## Correlation of (regression) scores with factors    PA1    0.97
## Multiple R square of scores with factors          0.94
## Minimum correlation of possible factor scores      0.87
##
## Loadings:
##
## MntWines      PA1    0.743
## MntFruits     PA1    0.801
## MntMeatProducts 0.929
## MntFishProducts 0.809
## MntSweetProducts 0.795
## MntGoldProds  0.705
##
##
##          PA1
## SS loadings 3.841
## Proportion Var 0.640

```

## Factor Analysis



```

## Factor Analysis using method = pa
## Call: fa(r = R, nfactors = k, rotate = "none", fm = "pa")
## Standardized loadings (pattern matrix) based upon correlation matrix
##
##          V PA1  h2  u2 com
## MntMeatProducts 3 0.93 0.86 0.14 1
## MntFishProducts 4 0.81 0.66 0.34 1

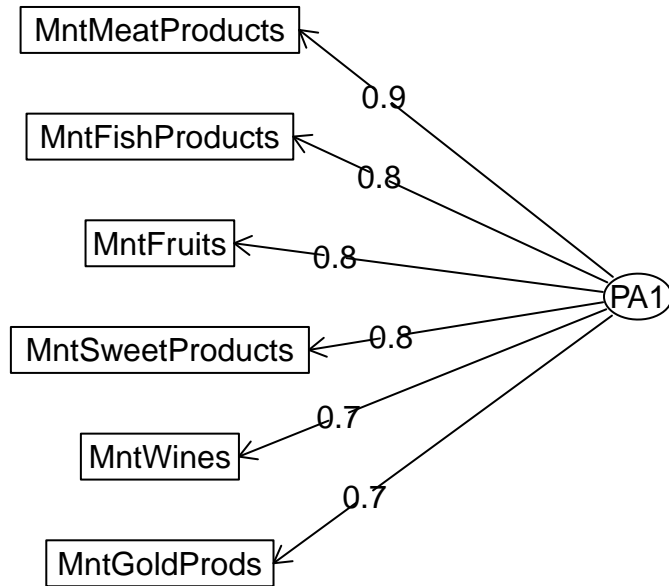
```

```

## MntFruits      2 0.80 0.64 0.36  1
## MntSweetProducts 5 0.80 0.63 0.37  1
## MntWines       1 0.74 0.55 0.45  1
## MntGoldProds   6 0.71 0.50 0.50  1
##
##              PA1
## SS loadings   3.84
## Proportion Var 0.64
##
## Mean item complexity = 1
## Test of the hypothesis that 1 factor is sufficient.
##
## df null model = 15 with the objective function = 4.32 0.3
## df of the model are 9 and the objective function was 0.46
## 0.3
## The root mean square of the residuals (RMSR) is 0.06
## The df corrected root mean square of the residuals is 0.07
## 0.3
## Fit based upon off diagonal values = 0.99
## Measures of factor score adequacy
##
##              PA1
## Correlation of (regression) scores with factors 0.97
## Multiple R square of scores with factors 0.94
## Minimum correlation of possible factor scores 0.87
##
## Loadings:
##              PA1
## MntWines      0.743
## MntFruits     0.801
## MntMeatProducts 0.929
## MntFishProducts 0.809
## MntSweetProducts 0.795
## MntGoldProds  0.705
##
##              PA1
## SS loadings   3.841
## Proportion Var 0.640

```

## Factor Analysis



De los ajustes realizados con el factor que se escogió vemos que si una persona compra Productos como GoldProds, Wines, además de alimentos como FishProducts entre otros es un consumidor que tiene un ingreso mas alto con respecto a los que no compran este tipo de alimentos, por otro lado si tenemos que una persona solo compra cosas como MeatProducts Fruits sin tomar productos como Fish y Wines entonces es un consumidor que solo compra lo necesario o la canasta básica.