

Factor Analysis

Yuchen Hu

11/26/2018

```
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
```

```
library(ggplot2)  
library(data.table)
```

```
##  
## Attaching package: 'data.table'  
## The following objects are masked from 'package:dplyr':  
##  
##   between, first, last
```

```
library(psych)
```

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

```
library(nFactors)
```

```
## Loading required package: MASS  
##  
## Attaching package: 'MASS'  
## The following object is masked from 'package:dplyr':  
##  
##   select  
## Loading required package: boot  
##  
## Attaching package: 'boot'  
## The following object is masked from 'package:psych':  
##  
##   logit  
## Loading required package: lattice
```

```
##
## Attaching package: 'lattice'

## The following object is masked from 'package:boot':
##
##      melanoma

##
## Attaching package: 'nFactors'

## The following object is masked from 'package:lattice':
##
##      parallel
library(reshape2)

##
## Attaching package: 'reshape2'

## The following objects are masked from 'package:data.table':
##
##      dcast, melt
library(tidyr)

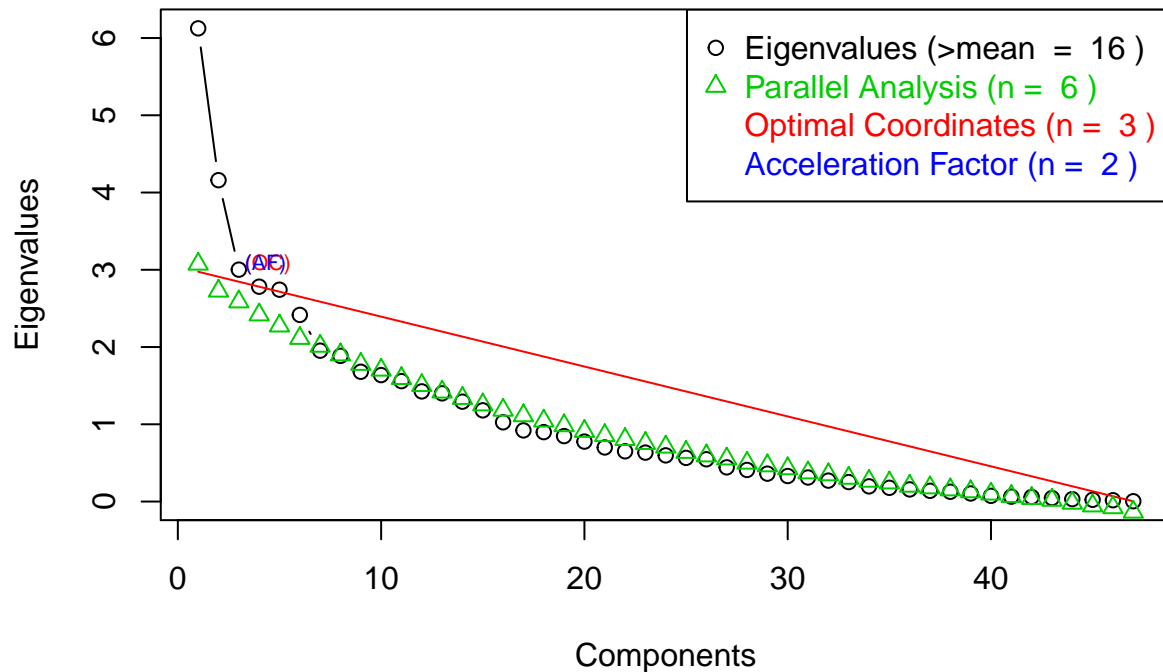
##
## Attaching package: 'tidyr'

## The following object is masked from 'package:reshape2':
##
##      smiths
food <- read.csv("food_coded.csv",stringsAsFactors = FALSE)
#summary(food)
```

basic FA

```
food_numeric <- food %>% select_if(is.numeric)
food_numeric <- food_numeric[complete.cases(food_numeric), ]
ev <- eigen(cor(food_numeric))
ap <- parallel(subject=nrow(food_numeric),var=ncol(food_numeric),
               rep=100,cent=.05)
nS <- nScree(x=ev$values, aparallel=ap$eigen$gevpea)
plotnScree(nS)
```

Non Graphical Solutions to Scree Test



```
fa_numeric <- factanal(food_numeric, 5, rotation="varimax", scores="regression")
print(fa_numeric, digits=2, cutoff=.3, sort=TRUE)
```

```
##
## Call:
## factanal(x = food_numeric, factors = 5, scores = "regression",      rotation = "varimax")
##
## Uniquenesses:
##           Gender          breakfast
##           0.86           0.62
##   calories_chicken  calories_day
##           0.50           0.51
##   calories_scone    coffee
##           0.70           0.87
##   comfort_food_reasons_coded  cook
##           0.03           0.78
##   comfort_food_reasons_coded.1  cuisine
##           0.00           0.88
##   diet_current_coded  drink
##           0.89           0.83
##   eating_changes_coded  eating_changes_coded1
##           0.85           0.97
##   eating_out  employment
##           0.72           0.84
##   ethnic_food  exercise
##           0.43           0.80
##   father_education  fav_cuisine_coded
##           0.79           0.81
##   fav_food  fries
##           0.81           0.91
```

##	fruit_day	grade_level
##	0.34	0.38
##	greek_food	healthy_feeling
##	0.32	0.94
##	ideal_diet_coded	income
##	0.93	0.59
##	indian_food	italian_food
##	0.11	0.80
##	life_rewarding	marital_status
##	0.88	0.81
##	mother_education	nutritional_check
##	0.93	0.76
##	on_off_campus	parents_cook
##	0.76	0.90
##	pay_meal_out	persian_food
##	0.89	0.18
##	self_perception_weight	soup
##	0.80	0.85
##	sports	thai_food
##	0.84	0.20
##	tortilla_calories	turkey_calories
##	0.57	0.62
##	veggies_day	vitamins
##	0.24	0.69
##	waffle_calories	
##	0.52	

Loadings:

##	Factor1	Factor2	Factor3	Factor4	Factor5
## ethnic_food	0.74				
## greek_food	0.80				
## indian_food	0.93				
## persian_food	0.86				
## thai_food	0.87				
## calories_chicken		0.56	0.35		
## tortilla_calories		0.64			
## turkey_calories		0.58			
## waffle_calories		0.69			
## fruit_day	0.35		0.68		
## veggies_day	0.30		0.79		
## comfort_food_reasons_coded				0.97	
## comfort_food_reasons_coded.1				0.98	
## grade_level					0.69
## income					-0.56
## Gender					
## breakfast			-0.50		
## calories_day	0.42	0.39	0.39		
## calories_scone		0.35		0.37	
## coffee					
## cook					-0.41
## cuisine					
## diet_current_coded					
## drink					
## eating_changes_coded			0.33		

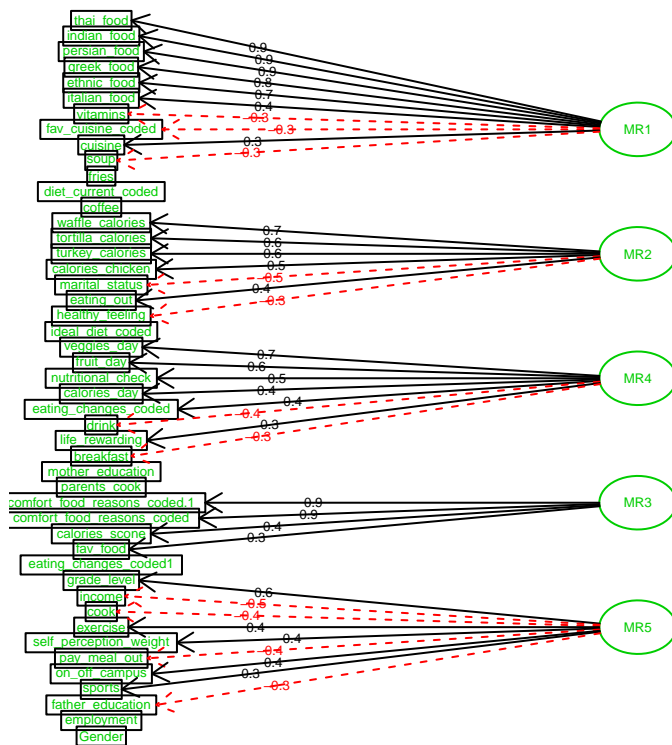
```

## eating_changes_coded1
## eating_out                0.46
## employment
## exercise                  -0.36
## father_education          -0.31
## fav_cuisine_coded
## fav_food
## fries
## healthy_feeling
## ideal_diet_coded
## italian_food              0.38
## life_rewarding
## marital_status            -0.38
## mother_education
## nutritional_check          0.39
## on_off_campus              0.44
## parents_cook
## pay_meal_out
## self_perception_weight    -0.30
## soup
## sports
## vitamins                  -0.43
##
##               Factor1 Factor2 Factor3 Factor4 Factor5
## SS loadings    5.12    3.01    2.86    2.69    2.06
## Proportion Var  0.11    0.06    0.06    0.06    0.04
## Cumulative Var  0.11    0.17    0.23    0.29    0.33
##
## Test of the hypothesis that 5 factors are sufficient.
## The chi square statistic is 946.87 on 856 degrees of freedom.
## The p-value is 0.0162
# factor1: food choices; 2:calories guess; 3:life style ; 4:comfort food; 5:education
fa_numeric <- fa(food_numeric, 5)

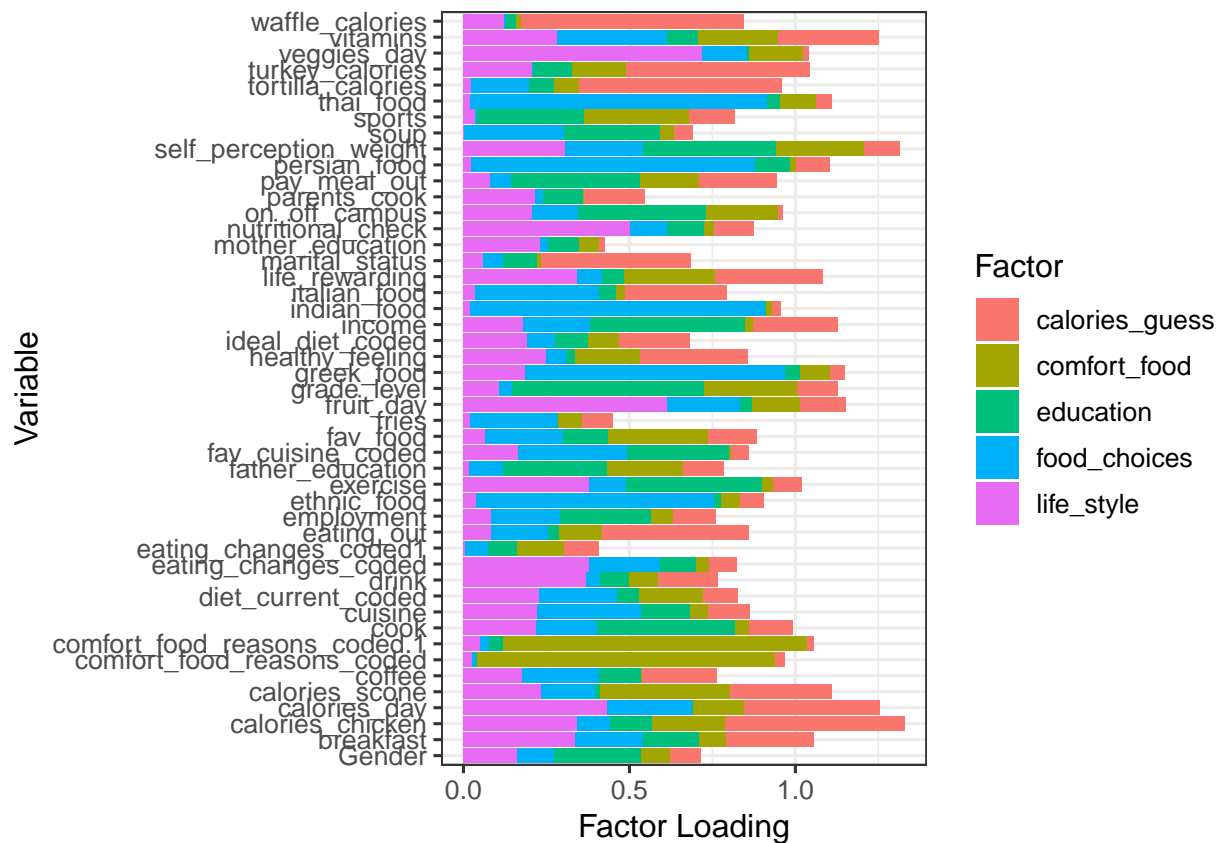
## Loading required namespace: GPArotation
fa.diagram(fa_numeric,side=1)

```

Factor Analysis



```
# plot the result
gathering <- as.data.frame(matrix(nrow = dim(food_numeric)[2], ncol = 5))
gathering$Variable <- colnames(food_numeric)
for (i in 1:5) {
  for (j in 1:dim(food_numeric)[2]) {
    gathering[j, i] <- fa_numeric$loadings[j, i]
  }
}
colnames(gathering) <- c("food_choices", "calories_guess",
                        "life_style", "comfort_food", "education", "Variable")
gathering <- gathering %>% gather("Factor", "Value", 1:5)
ggplot(gathering, aes(Variable, abs(Value), fill=Factor)) +
  geom_bar(stat="identity") + coord_flip() +
  ylab("Factor Loading") +
  theme_bw(base_size=12)
```



run regression on score

```
factor_gpa <- merge(as.numeric(food$GPA),fa_numeric$scores,
                    by="row.names",all.x=FALSE)[,-1]

## Warning in merge(as.numeric(food$GPA), fa_numeric$scores, by =
## "row.names", : NAs introduced by coercion

names(factor_gpa) <- c("GPA","food_choices","calories_guess","life_style",
                      "comfort_food","education")
# regression based on numerical GPA
fit.factor <- lm(GPA~food_choices+calories_guess+life_style+comfort_food+education,
                 data=factor_gpa)
summary(fit.factor)

##
## Call:
## lm(formula = GPA ~ food_choices + calories_guess + life_style +
##     comfort_food + education, data = factor_gpa)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.64709 -0.23254  0.06347  0.17809  0.67057
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)

```

```
## (Intercept)      3.45408      0.04089  84.464 < 2e-16 ***
## food_choices     0.02948      0.04334   0.680  0.49933
## calories_guess   0.15661      0.04671   3.353  0.00148 **
## life_style       -0.06132      0.04638  -1.322  0.19179
## comfort_food     -0.11280      0.04160  -2.712  0.00900 **
## education         0.02096      0.04648   0.451  0.65389
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3139 on 53 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.2728, Adjusted R-squared:  0.2042
## F-statistic: 3.977 on 5 and 53 DF,  p-value: 0.003891

# regression based on categorical GPA
factor_gpa$GPA <- cut(factor_gpa$GPA, c(0,2.5,3,3.25,3.5,3.75,4.0))
factor_gpa$GPA <- as.numeric(factor_gpa$GPA)
fit.factor <- lm(GPA~food_choices+calories_guess+life_style+comfort_food+education,
                 data=factor_gpa)
summary(fit.factor)

##
## Call:
## lm(formula = GPA ~ food_choices + calories_guess + life_style +
##     comfort_food + education, data = factor_gpa)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2939 -0.8507  0.1867  0.9839  2.3880
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    4.22067    0.16666  25.325 < 2e-16 ***
## food_choices    0.12323    0.17664   0.698  0.48846
## calories_guess  0.54759    0.19035   2.877  0.00578 **
## life_style     -0.25663    0.18902  -1.358  0.18030
## comfort_food   -0.36102    0.16952  -2.130  0.03786 *
## education       0.04968    0.18943   0.262  0.79412
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.279 on 53 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.2085, Adjusted R-squared:  0.1338
## F-statistic: 2.792 on 5 and 53 DF,  p-value: 0.02605
```