

HW5_Prob1

Haoyu_Zhao

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Problem 1

a All the R code will shown in the report.

b

```
#Load the data
#Skip the step that convert the
#xlsx file into the csv file.
flavor <- read.csv("flavor.csv", sep=";")
#Delete the column that contains the variety
flavor <- flavor[,2:9]
#Compute the Covariance
Cov <- cov(flavor)
#do the PC analysis
flavor.pc <- princomp(flavor)
summary(flavor.pc)

## Importance of components:
##               Comp.1   Comp.2   Comp.3   Comp.4   Comp.5
## Standard deviation    8.3357813 4.6912363 3.1474171 1.55063523 1.29621790
## Proportion of Variance 0.6454225 0.2044213 0.0920153 0.02233422 0.01560657
## Cumulative Proportion 0.6454225 0.8498438 0.9418591 0.96419328 0.97979985
##               Comp.6   Comp.7   Comp.8
## Standard deviation    1.1164306 0.717358144 0.643194062
## Proportion of Variance 0.0115775 0.004779952 0.003842691
## Cumulative Proportion 0.9913774 0.996157309 1.000000000

lds <- flavor.pc$loadings
sdevs <- flavor.pc$sdev
#the loading for the factor analysis
l1 <- lds[,1] * sdevs[1]
l2 <- lds[,2] * sdevs[2]
l3 <- lds[,3] * sdevs[3]

#For m=1, the loadings are
ld <- matrix(l1, 8, 1)
print(ld)

##               [,1]
## [1,] -5.04730017
## [2,] -4.32490311
## [3,] -3.38622039
## [4,] -0.66521459
## [5,] -0.90355038
## [6,] -0.04860147
## [7,] -1.13673758
## [8,] -3.35935823
```

```
#The communality
print(ld * ld)
```

```
##           [,1]
## [1,] 25.475239039
## [2,] 18.704786937
## [3,] 11.466488504
## [4,]  0.442510448
## [5,]  0.816403290
## [6,]  0.002362103
## [7,]  1.292172315
## [8,] 11.285287703
```

```
#The matrix Cov - LL'
Cov <- matrix(Cov, 8, 8)
temp <- Cov - ld %*% t(ld)
#The specific variance
sv <- temp * diag(8)
sv
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]
## [1,] 3.644457 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
## [2,] 0.000000 8.441239 0.000000 0.000000 0.000000 0.000000 0.000000
## [3,] 0.000000 0.000000 5.300912 0.000000 0.000000 0.000000 0.000000
## [4,] 0.000000 0.000000 0.000000 8.550805 0.000000 0.000000 0.000000
## [5,] 0.000000 0.000000 0.000000 0.000000 2.130377 0.000000 0.000000
## [6,] 0.000000 0.000000 0.000000 0.000000 0.000000 2.865403 0.000000
## [7,] 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 1.791081
## [8,] 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
##           [,8]
## [1,] 0.000000
## [2,] 0.000000
## [3,] 0.000000
## [4,] 0.000000
## [5,] 0.000000
## [6,] 0.000000
## [7,] 0.000000
## [8,] 6.67242
```

```
#the portion of variance
sdevs[1] / sum(sdevs)
```

```
##      Comp.1
## 0.3877419
```

```
#For m=2, the loadings are
ld <- cbind(l1,l2)
print(ld)
```

```
##           11      12
## OVERALL.LIKING    -5.04730017  1.3074429
## TEXTURE          -4.32490311  2.0579311
## Sweetness        -3.38622039 -1.1337442
## Sourness         -0.66521459 -2.2846747
## Salty            -0.90355038 -1.2070271
## Bitter           -0.04860147 -1.2959698
```

```
## Umami -1.13673758 -0.8280021
## Overall.Flavor.Intensity -3.35935823 -2.3950067
```

```
#The communality
```

```
print(rowSums(ld *ld))
```

```
##          OVERALL.LIKING          TEXTURE          Sweetness
##          27.184646          22.939867          12.751864
##          Sourness          Salty          Bitter
##          5.662249          2.273318          1.681900
##          Umami Overall.Flavor.Intensity
##          1.977760          17.021345
```

```
#The matrix Cov - LL'
```

```
Cov <- matrix(Cov, 8, 8)
```

```
temp <- Cov - ld %*% t(ld)
```

```
#The specific variance
```

```
sv <- temp * diag(8)
```

```
sv
```

```
##          OVERALL.LIKING  TEXTURE Sweetness Sourness
## OVERALL.LIKING          1.935051 0.000000 0.000000 0.000000
## TEXTURE                  0.000000 4.206158 0.000000 0.000000
## Sweetness                0.000000 0.000000 4.015537 0.000000
## Sourness                 0.000000 0.000000 0.000000 3.331067
## Salty                    0.000000 0.000000 0.000000 0.000000
## Bitter                   0.000000 0.000000 0.000000 0.000000
## Umami                    0.000000 0.000000 0.000000 0.000000
## Overall.Flavor.Intensity 0.000000 0.000000 0.000000 0.000000
##          Salty  Bitter  Umami
## OVERALL.LIKING 0.0000000 0.0000000 0.0000000
## TEXTURE        0.0000000 0.0000000 0.0000000
## Sweetness      0.0000000 0.0000000 0.0000000
## Sourness       0.0000000 0.0000000 0.0000000
## Salty          0.6734624 0.0000000 0.0000000
## Bitter         0.0000000 1.185866 0.0000000
## Umami          0.0000000 0.0000000 1.105493
## Overall.Flavor.Intensity 0.0000000 0.0000000 0.0000000
##          Overall.Flavor.Intensity
## OVERALL.LIKING          0.0000000
## TEXTURE                 0.0000000
## Sweetness               0.0000000
## Sourness                0.0000000
## Salty                   0.0000000
## Bitter                  0.0000000
## Umami                   0.0000000
## Overall.Flavor.Intensity 0.9363624
```

```
#the portion of variance
```

```
sdevs[1] / sum(sdevs)
```

```
##      Comp.1
```

```
## 0.3877419
```

```
sdevs[2] / sum(sdevs)
```

```
##      Comp.2
```

```
## 0.2182146
```

```
#For m=3
```

```
ld <- cbind(l1,l2,l3)
```

```
t(ld) %*% ld
```

```
##           l1           l2           l3
## 11 6.948525e+01 0.000000e+00 3.747003e-16
## 12 0.000000e+00 2.200770e+01 -1.050549e-14
## 13 3.747003e-16 -1.050549e-14 9.906234e+00
```

```
print(ld)
```

```
##           l1           l2           l3
## OVERALL.LIKING      -5.04730017  1.3074429  0.73606614
## TEXTURE             -4.32490311  2.0579311 -1.86881816
## Sweetness           -3.38622039 -1.1337442  1.69603162
## Sourness            -0.66521459 -2.2846747 -1.57587743
## Salty               -0.90355038 -1.2070271 -0.39811211
## Bitter              -0.04860147 -1.2959698 -0.59291472
## Umami               -1.13673758 -0.8280021 -0.03365409
## Overall.Flavor.Intensity -3.35935823 -2.3950067  0.02954650
```

```
#The communality
```

```
print(rowSums(ld *ld))
```

```
##           OVERALL.LIKING           TEXTURE           Sweetness
##           27.726439           26.432349           15.628388
##           Sourness           Salty           Bitter
##           8.145639           2.431811           2.033448
##           Umami Overall.Flavor.Intensity
##           1.978892           17.022218
```

```
#The matrix Cov - LL'
```

```
Cov <- matrix(Cov, 8, 8)
```

```
temp <- Cov - ld %*% t(ld)
```

```
#The specific variance
```

```
sv <- temp * diag(8)
```

```
sv
```

```
##           OVERALL.LIKING  TEXTURE Sweetness  Sourness
## OVERALL.LIKING           1.393257 0.0000000 0.000000 0.0000000
## TEXTURE                  0.000000 0.7136771 0.000000 0.0000000
## Sweetness                0.000000 0.0000000 1.139013 0.0000000
## Sourness                 0.000000 0.0000000 0.000000 0.8476773
## Salty                    0.000000 0.0000000 0.000000 0.0000000
## Bitter                   0.000000 0.0000000 0.000000 0.0000000
## Umami                    0.000000 0.0000000 0.000000 0.0000000
## Overall.Flavor.Intensity 0.000000 0.0000000 0.000000 0.0000000
##           Salty  Bitter  Umami
## OVERALL.LIKING 0.000000 0.000000 0.000000
## TEXTURE        0.000000 0.000000 0.000000
## Sweetness      0.000000 0.000000 0.000000
## Sourness       0.000000 0.000000 0.000000
## Salty          0.5149692 0.000000 0.000000
## Bitter         0.000000 0.8343177 0.000000
## Umami          0.000000 0.000000 1.10436
```

```
## Overall.Flavor.Intensity 0.0000000 0.0000000 0.00000
## Overall.Flavor.Intensity
## OVERALL.LIKING 0.0000000
## TEXTURE 0.0000000
## Sweetness 0.0000000
## Sourness 0.0000000
## Salty 0.0000000
## Bitter 0.0000000
## Umami 0.0000000
## Overall.Flavor.Intensity 0.9354894
```

```
#the portion of variance
sdevs[1] / sum(sdevs)
```

```
## Comp.1
## 0.3877419
```

```
sdevs[2] / sum(sdevs)
```

```
## Comp.2
## 0.2182146
```

```
sdevs[3] / sum(sdevs)
```

```
## Comp.3
## 0.1464033
```

```
#Then we do the factor analysis using the MLE.
#We provide the result of m = 1,2,3.
```

```
flavor.fa1 <- factanal(flavor, factors = 1)
flavor.fa2 <- factanal(flavor, factors = 2)
flavor.fa3 <- factanal(flavor, factors = 3)
#print the result
flavor.fa1
```

```
##
## Call:
## factanal(x = flavor, factors = 1)
##
## Uniquenesses:
## OVERALL.LIKING TEXTURE Sweetness
## 0.635 0.804 0.352
## Sourness Salty Bitter
## 0.668 0.352 0.814
## Umami Overall.Flavor.Intensity
## 0.397 0.005
##
## Loadings:
## Factor1
## OVERALL.LIKING 0.604
## TEXTURE 0.443
## Sweetness 0.805
## Sourness 0.576
## Salty 0.805
## Bitter 0.431
## Umami 0.776
## Overall.Flavor.Intensity 0.997
##
```

```
##          Factor1
## SS loadings      3.972
## Proportion Var   0.497
##
## Test of the hypothesis that 1 factor is sufficient.
## The chi square statistic is 317.85 on 20 degrees of freedom.
## The p-value is 1.8e-55
```

```
flavor.fa2
```

```
##
## Call:
## factanal(x = flavor, factors = 2)
##
## Uniquenesses:
##          OVERALL.LIKING          TEXTURE          Sweetness
##              0.005              0.346              0.331
##          Sourness              Salty              Bitter
##              0.286              0.129              0.273
##          Umami Overall.Flavor.Intensity
##              0.316              0.141
##
## Loadings:
##              Factor1 Factor2
## OVERALL.LIKING              0.997
## TEXTURE              0.807
## Sweetness              0.323  0.752
## Sourness              0.845
## Salty              0.867  0.346
## Bitter              0.835 -0.170
## Umami              0.624  0.543
## Overall.Flavor.Intensity 0.681  0.629
##
##              Factor1 Factor2
## SS loadings      3.125  3.048
## Proportion Var   0.391  0.381
## Cumulative Var   0.391  0.772
##
## Test of the hypothesis that 2 factors are sufficient.
## The chi square statistic is 86.93 on 13 degrees of freedom.
## The p-value is 5.38e-13
```

```
flavor.fa3
```

```
##
## Call:
## factanal(x = flavor, factors = 3)
##
## Uniquenesses:
##          OVERALL.LIKING          TEXTURE          Sweetness
##              0.005              0.281              0.022
##          Sourness              Salty              Bitter
##              0.232              0.131              0.265
##          Umami Overall.Flavor.Intensity
##              0.327              0.093
##
```

```
## Loadings:
##
##          Factor1 Factor2 Factor3
## OVERALL.LIKING          0.880  0.468
## TEXTURE          0.831  0.167
## Sweetness        0.154  0.364  0.906
## Sourness         0.873
## Salty            0.839  0.219  0.341
## Bitter           0.834 -0.198
## Umami            0.583  0.397  0.418
## Overall.Flavor.Intensity 0.602  0.379  0.633
##
##          Factor1 Factor2 Factor3
## SS loadings      2.891  1.987  1.766
## Proportion Var   0.361  0.248  0.221
## Cumulative Var   0.361  0.610  0.830
##
## Test of the hypothesis that 3 factors are sufficient.
## The chi square statistic is 34.24 on 7 degrees of freedom.
## The p-value is 1.56e-05
```

c The factorization of the PCA and MLE approach are much different. First, the PCA uses the covariance matrix, but in the 'factanal' in R language, it uses the correlation matrix to approximate the model.

When we look at the portion of variance explained by each factor, we can see that the PCA approach is much better than the MLE approach. There are 2 reasons, the first is that PCA aims to find the direction or loadings that can best explain the variance so it is absolutely the best if we just look at the portion of variance. The second reason is that in the 'factanal' function in R language, it uses the correlation matrix to compute, but in the PCA, we use the covariance matrix to compute. This will cause some bias on the certain factor and thus increase the portion of variance that the first few factors can analyze.

d In the PCA approach, the loadings of the first factor is a combination of all the features, and 1 of the features takes small weights. The loadings of the second factor contrast the first 2 features with the other features, the savor. The third factor contrast the first and the third features with other features.