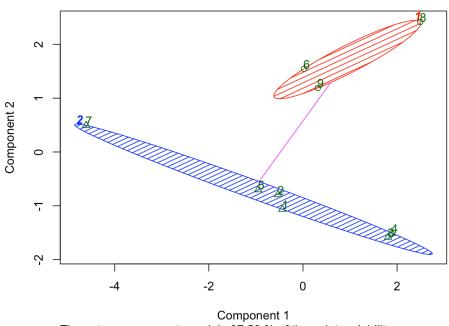
Problem 1

(a) As the professor mentioned in the KLMS, I assume that the number of clusters is 2.

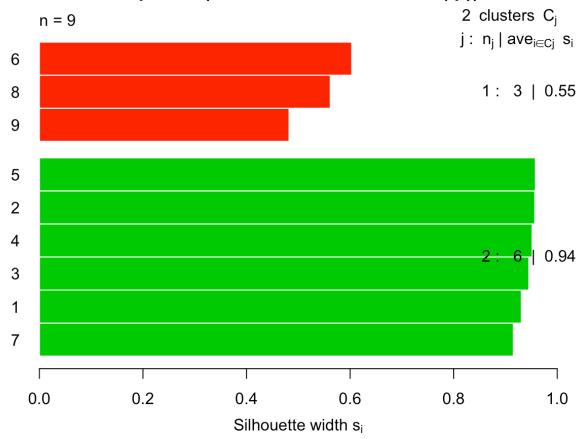
```
> (kc = kmeans(tp, centers=2, nstart=3))
K-means clustering with 2 clusters of sizes 3, 6
Cluster means:
        nΑ
              meanA
                         sdA
                                   nВ
                                         meanB
1 861.3333 3.806667 4.410000 843.6667 4.046667 4.660000
2 168.8333 5.271667 4.458333 160.1667 5.368333 4.803333
Clustering vector:
1 2 3 4 5 6 7 8 9
2 2 2 2 2 1 2 1 1
Within cluster sum of squares by cluster:
[1] 246870.421 9793.245
(between_SS / total_SS = 88.1 %)
Available components:
[1] "cluster"
                   "centers"
                                  "totss"
                                                 "withinss"
                                                                 "tot.withinss" "betwe
enss"
         "size"
[8] "iter"
                   "ifault"
```

CLUSPLOT(tp)



These two components explain 97.53 % of the point variability.

Silhouette plot of (x = kc\$cluster, dist = dist(tp))



Average silhouette width: 0.81

Problem 2

(a)

> (kc = kmeans(pec, centers=2, nstart=3))

K-means clustering with 2 clusters of sizes 13, 12

Cluster means:

RedMeat WhiteMeat Eggs Milk Fish Cereals Starch Nuts 1 11.807692 9.607692 3.707692 22.08462 5.023077 24.06923 4.761538 1.692308 2 7.683333 6.041667 2.100000 11.72500 3.483333 41.10833 3.750000 4.566667 FruitVeg

1 3.500

2 4.825

Clustering vector:

Within cluster sum of squares by cluster:

[1] 808.6554 1668.0933

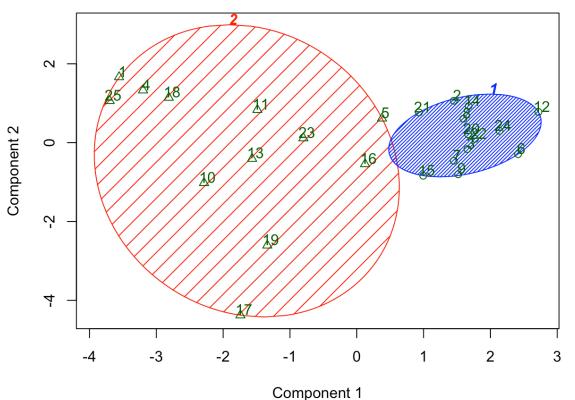
(between_SS / total_SS = 52.8 %)

Available components:

[1] "cluster" "centers" "totss" "withinss" "tot.withinss"

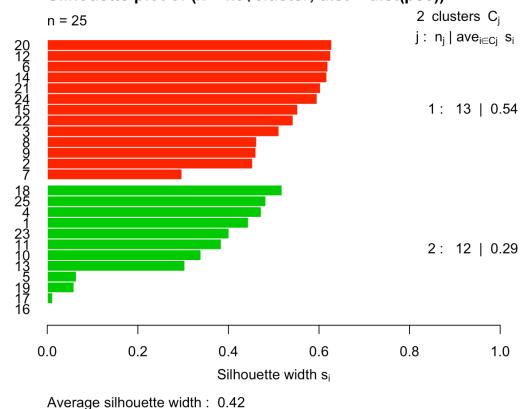
[6] "betweenss" "size" "iter" "ifault"

CLUSPLOT(pec)



These two components explain 62.68 % of the point variability.

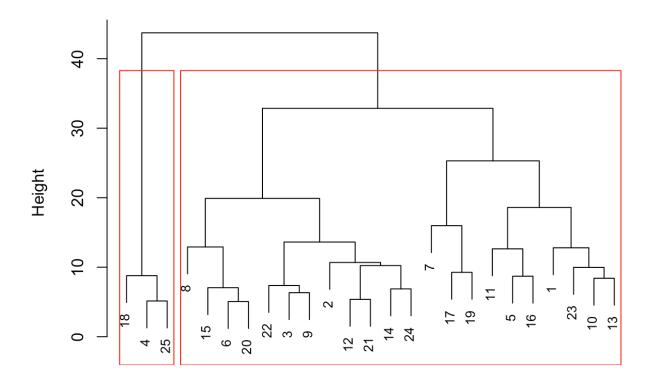
(b) Silhouette plot of (x = kc\$cluster, dist = dist(pec))



```
(C)
> cl1 = subset(pec, kc$cluster=="1")
> cl2 = subset(pec, kc$cluster=="2")
> t.test(cl1[,1],cl2[,1])$p.value #
[1] 0.0007830653
> t.test(cl1[,2],cl2[,2])$p.value #
[1] 0.01310718
> t.test(cl1[,3],cl2[,3])$p.value #
[1] 5.03499e-05
> t.test(cl1[,4],cl2[,4])$p.value #
[1] 1.930472e-05
> t.test(cl1[,5],cl2[,5])$p.value
[1] 0.2768672
> t.test(cl1[,6],cl2[,6])$p.value #
[1] 4.407262e-05
> t.test(cl1[,7],cl2[,7])$p.value
[1] 0.1388205
> t.test(cl1[,8],cl2[,8])$p.value #
[1] 0.0001519818
> t.test(cl1[,9],cl2[,9])$p.value
[1] 0.07255526
```

Hence, except for "Fish", "Starch", and "FruitVeg", all categories between two clusters show significant difference.

Cluster Dendrogram



ds hclust (*, "complete")