

## problem4-4

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## Chemical Features of Wine.

Table 4.13 shows the PCA output on data non normalized in which the variables represent chemical characteristics of wine, and each case is a different wine.

**TABLE 4.13** PRINCIPAL COMPONENTS OF NON-NORMALIZED WINE DATA



code for running PCA on the wine data

```
wine.df <- read.csv("Wine.csv")
pcs.cor <- prcomp(wine.df[, -1])
summary(pcs.cor)
pcs.cor$rot[, 1:4]
```

Output

```
> summary(pcs.cor)
```

importance of components:

	PC1	PC2	PC3	PC4	PC5
Standard deviation	314.9632	13.13527	3.07215	2.23409	1.10853
Proportion of Variance	0.9981	0.00174	0.00009	0.00005	0.00001
Cumulative Proportion	0.9981	0.99983	0.99992	0.99997	0.99998
	PC6	PC7	PC8	PC9	PC10
Standard deviation	0.91710	0.5282	0.3891	0.3348	0.2678
Proportion of Variance	0.00001	0.0000	0.0000	0.0000	0.0000
Cumulative Proportion	0.99999	1.0000	1.0000	1.0000	1.0000
	PC11	PC12	PC13		
Standard deviation	0.1938	0.1452	0.09057		
Proportion of Variance	0.0000	0.0000	0.00000		
Cumulative Proportion	1.0000	1.0000	1.00000		

```
> pcs.cor$rot[, 1:4]
```

	PC1	PC2	PC3	PC4
Alcohol	-0.0016592647	-1.203406e-03	-0.016873809	0.141446778
Malic_Acid	0.0006810156	-2.154982e-03	-0.122003373	0.160389543
Ash	-0.0001949057	-4.593693e-03	-0.051987430	-0.009772810
Ash_Alcalinity	0.0046713006	-2.645039e-02	-0.938593003	-0.330965260
Magnesium	-0.0178680075	-9.993442e-01	0.029780248	-0.005393756
Total_Phenols	-0.0009898297	-8.779622e-04	0.040484644	-0.074584656
Flavanoids	-0.0015672883	5.185073e-05	0.085443339	-0.169086724
Nonflavanoid_Phenols	0.0001230867	1.354479e-03	-0.013510780	0.010805561
Proanthocyanins	-0.0006006078	-5.004400e-03	0.024659382	-0.050120952
Color_Intensity	-0.0023271432	-1.510035e-02	-0.291398464	0.878893693
Hue	-0.0001713800	7.626731e-04	0.025977662	-0.060034945
OD280_OD315	-0.0007049316	3.495364e-03	0.070323969	-0.178200254
Proline	-0.9998229365	1.777381e-02	-0.004528682	-0.003112916

- a. The data are in the file Wine.csv. Consider the rows labeled Proportion of Variance. Explain why the value for PC1 is so much greater than that of any other column.
- b. Comment on the use of normalization (standardization) in part (a).

a)

According to output ( above table ) we can see summary and the rotation outputs.

the summary proportion of variance for PC1 is the biggest and is equal to 0.9981, it means the first components coverage about 99.81% of all variables variance and we can reduce our dimension to one dimension with PC1 rotation and new values.

the coefficient of each variable for PC1 is equal to first column of second output(`pcs$cor$rot[,1:4]`)

b)

i think that we should use principal components with normalized data cause we have a lot of variables and they are with different measure and scale and when we normalizing them we are using scale and center on them and it's better for doing principal component method. for doing these code we just need to :

```
#wine.df <- read.csv("Wine.csv")  
#pcs.cor.norm <- prcomp(wine.df[, -1] ,  
#center = TRUE, scale. = FALSE )  
#summary(pcs.cor.norm)  
#pcs.cor.norm$rot[, 1:4]
```