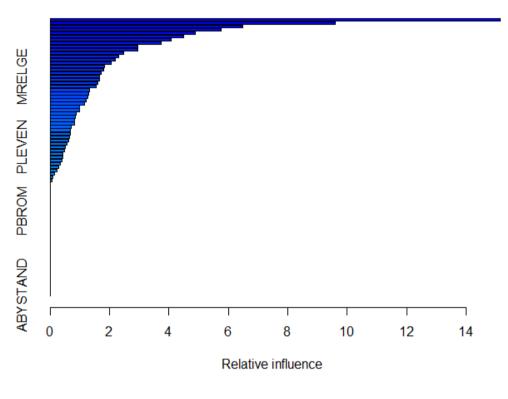
## # question 1

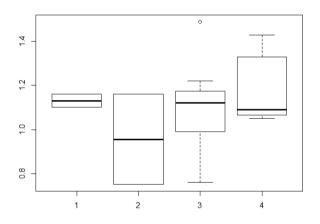
```
# question 1
library(ISLR)
sapply(Caravan,table)
table(Caravan$PVRAAUT)
table(Caravan$AVRAAUT)
Caravan
nrow(Caravan)
colnames(Caravan)
d0 = subset(Caravan, select = c(-PVRAAUT, -AVRAAUT))
ncol(Caravan)
ncol(d0)
#a)
str(d0)
levels(d0$Purchase) = c("0","1")
d0$Purchase=as.numeric(d0$Purchase)-1
d0$Purchase
dtrain = d0[1:1000,]
nrow(d0)
dtest = d0[1001:5822,]
head(d0$Purchase)
d0$Purchase
#b)
library(gbm)
set.seed(1)
boost1 = gbm(Purchase~.,data = dtrain,distribution="bernoulli",n.trees = 1000,shrinkage =
0.01)
summary(boost1)
# PPERSAUT seems to be the most important
```

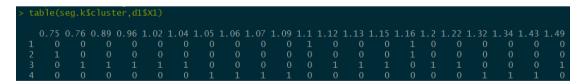


```
#c)
yhat.boost = predict(boost1,newdata = dtest,n.trees=1000,type = "response")
yhat.boost
yhat = ifelse(yhat.boost>0.2,"1","0")
tb = table(yhat,dtest$Purchase)
tb
# yhat
          0
                1
#
     0 4410 256
     1 123
                33
aux = prop.table(tb)
aux
# yhat
                  0
                               1
     0 0.914558275 0.053090004
     1 0.025508088 0.006843633
# 0.025508088 of people predicted to make a purchase do in fact make one
1-sum(diag(aux))
# test error rateis 0.07859809
```

# question 2

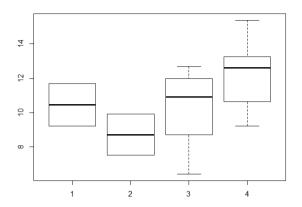
```
#a)
library(cluster)
d0 = utilities
d0$X6 = as.numeric(d0$X6)
d1=d0[,-1]
set.seed(1)
seg.k = kmeans(d1,centers = 4)
seg.k$cluster
summary(seg.k)
table(seg.k$cluster)
#[1]4343343143233432341334
#1 2 3 4
# 2 2 11 7
#b)
table(seg.k$cluster,d1$X1)
boxplot(d1$X1 ~ seg.k$cluster)
```





# group 2 is smaller than others in X1

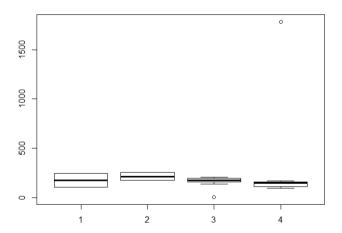
table(seg.k\$cluster,d1\$X2) boxplot(d1\$X2 ~ seg.k\$cluster)



# no big diff

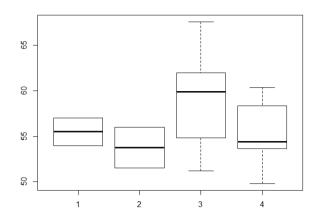
table(seg.k\$cluster,d1\$X3) boxplot(d1\$X3 ~ seg.k\$cluster)

> '	table(seg.k\$cluster,d1\$X3)																					
		1.92	96	104	111	113	136	148	150	151	164	168	173	175	178	197	199	202	204	245	252	1784
:	1	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0
	2	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	1	0
	3	1	0	0	0	0	1	1	0	0	1	1	0	1	1	1	1	1	1	0	0	0
	4	0		0	1	1	0	0	1	1	0	1	0	0	0	0	0	0	0	0	0	1



# no big diff

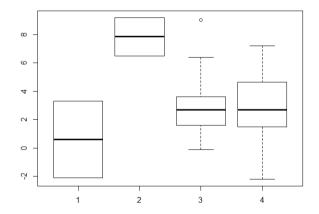
table(seg.k\$cluster,d1\$X4) boxplot(d1\$X4 ~ seg.k\$cluster)



# group 3 is slightly higher than 1,2,4 in X4

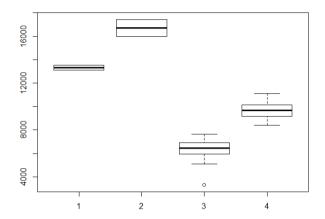
table(seg.k\$cluster,d1\$X5) boxplot(d1\$X5 ~ seg.k\$cluster)

	٠	`		9			,													
> '	table(seg.k\$cluster,d1\$X5)																			
		-2.2	-2.1	-0.1	0.3	1	1.4	1.6	2.2	2.7	3.3	3.4	3.5	3.7	5.9	6.4	6.5	7.2	9	9.2
	1	0	1	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
	2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	1
	3	0	0	1	1	1	0	0	2	1	0	0	2	1	0	1	0	0	1	0
	4	1	0	0	0	0	1	1	0	1	0	1	0	0	1	0	0	1	0	0



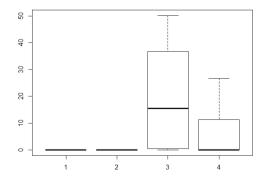
# group 2 is higher than group 1,3,4 in X5

table(seg.k\$cluster,d1\$X6) boxplot(d1\$X6 ~ seg.k\$cluster)



# group 1,2,3 and 4 are all different in X6, group 2 is highest, followed by group 1,4,and 3.

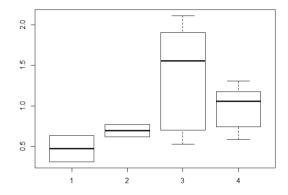
table(seg.k\$cluster,d1\$X7) boxplot(d1\$X7 ~ seg.k\$cluster)



# group 3 is higher than others in X7

## table(seg.k\$cluster,d1\$X8)

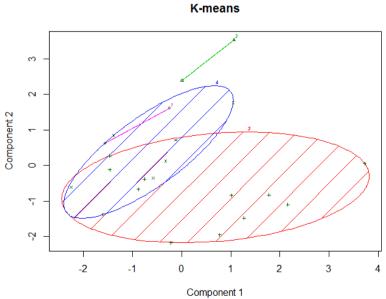
```
boxplot(d1$X8 ~ seg.k$cluster)
```



# group 3 is higher than and group 1 is lower than others in X8

#c)
pic=clusplot(d1,seg.k\$cluster,color=T,shade=T,labels = 1,lines=0,main="K-means",cex=0.5)
# the green ellipse, which is group 2 is different from others, because 1,3,4 are overlapping each other

# and group 1 is different from group 3 and 2. Group 3 and 4 is similar

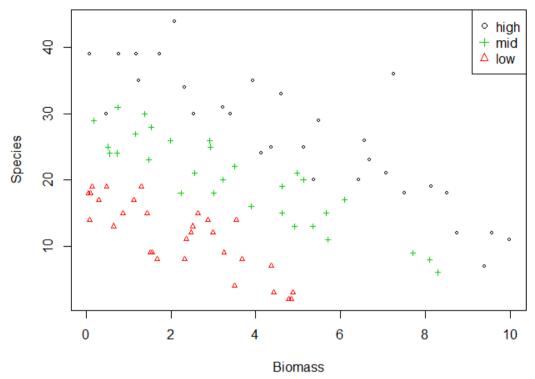


These two components explain 49.6 % of the point variability.

```
#d)
pr = prcomp(d1,scale=T)
rot = pr$rotation
rot
pr$x
as.matrix(scale(d1))%*%rot
PC1.scaled = rot[,1]
PC1.scaled
                  X2
#
        X1
                            X3
                                      X4
                                                X5
                                                          X6
                                                                    X7
X8
0.4557031
# is the PC1 eigonvector for scaled X
# the defining equation of scaled data is:
                       0.4451538*C1+0.5647231*C2-0.1333447*C3-0.3561666*C4-
0.2684790*C5+0.1404794*C6+0.1970019*C7-0.4557031*C8
# and C[i] = (X[i]-meanX[i])/std(X[i])
```

## # question 3

```
#a)
getwd()
d0 = read.table("species.txt")
d1 = d0
colnames(d1) = c("pH","Biomass","Species")
d1$Biomass = as.numeric(d1$Biomass)
d1$Species = as.numeric(d1$Species)
aux = as.numeric(d1$pH)
aux
d1$pH
levels(d1$pH)
plot(Species~Biomass,d1,col = aux,pch = aux,cex = 0.6)
legend("topright",c("high","mid","low"),pch = c(1,3,2),col = c(1,3,2) )
# according to the plot, we can not find evidence saying that the slope for each PH level is different
```



```
#b)
m1 = glm(Species~Biomass*pH,d1,family = poisson)
#c)
m2 = glm(Species~.,d1,family = poisson)
anova(m2,m1, test = "Chisq")
# Analysis of Deviance Table
# Model 1: Species ~ pH + Biomass
# Model 2: Species ~ Biomass * pH
# Resid. Df Resid. Dev Df Deviance Pr(>Chi)
# 1
                    99.242
            86
# 2
            84
                    83.201 2
                                  16.04 0.0003288 ***
   Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
# P-value is small, we can not reject m1(Model 2: Species ~ Biomass * pH) is better than
```

m2(Model 1: Species ~ pH + Biomass)

```
#d)

llow = rep("low",101)

lmid = rep("mid",101)

lhigh = rep("high",101)

lbio = seq(0,10,0.1)

dlow = data.frame(pH = llow,Biomass = lbio)

dmid = data.frame(pH = lmid,Biomass = lbio)

dhigh = data.frame(pH = lhigh,Biomass = lbio)

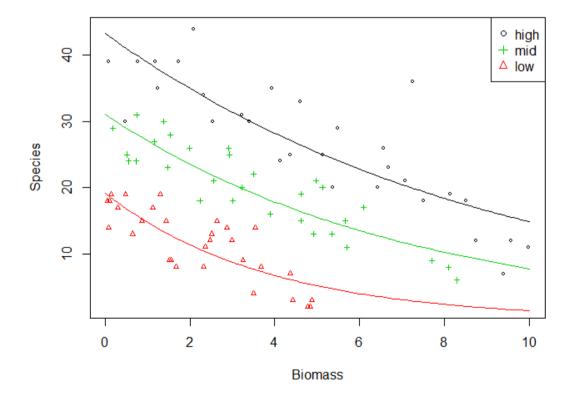
ylow = predict(m1,dlow,type = "response")

ymid = predict(m1,dmid,type = "response")

yhigh = predict(m1,dhigh,type = "response")

lines(ylow~lbio,col= 2,pch=".",cex = 0.6,xlim = c(0,10),ylim = c(0,40))

lines(yhigh~lbio,col= 1,pch=".",cex = 0.6,xlim = c(0,10),ylim = c(0,40))
```



```
#e)
newval = data.frame(pH = "low",Biomass = 2.0)
alpha = 0.05
yhat = predict(m1, newval, se.fit=T, type="link")
lower95 = exp(yhat$fit - qnorm(1-alpha/2)*yhat$se.fit)
lower95 = round(lower95,3)
```

```
upper95 = exp(yhat$fit + qnorm(1-alpha/2)*yhat$se.fit)
upper95 = round(upper95,3)
cat("(",lower95,",",upper95,")\n")
# ( 10.173 , 12.639 )

#f)
yhat = predict(m1,d1,type = "response")
res = yhat-d1$Species
which.max(abs(res))
d1[18,]
# pH Biomass Species
# 18 high 7.242062 36
# the number of species is 36, which associated with largest residual
```