

```

> library(faraway)
> d0=data.frame(hsb)
> d0 = hsb[,-1]
> str(d0)
'data.frame': 200 obs. of 10 variables:
 $ gender : Factor w/ 2 levels "female","male": 2 1 2 2 2 2 2 2 2 ...
 $ race   : Factor w/ 4 levels "african-amer",...: 4 4 4 4 4 4 1 3 4 1 ...
 $ ses    : Factor w/ 3 levels "high","low","middle": 2 3 1 1 3 3 3 3 3 ...
 $ schtyp : Factor w/ 2 levels "private","public": 2 2 2 2 2 2 2 2 2 ...
 $ prog   : Factor w/ 3 levels "academic","general",...: 2 3 2 3 1 1 2 1 2 1 ...
 $ read   : int 57 68 44 63 47 44 50 34 63 57 ...
 $ write  : int 52 59 33 44 52 52 59 46 57 55 ...
 $ math   : int 41 53 54 47 57 51 42 45 54 52 ...
 $ science: int 47 63 58 53 53 63 53 39 58 50 ...
 $ socst  : int 57 61 31 56 61 61 61 36 51 51 ...

```

**> #a)**

```
> table(d0$gender,d0$prog)
```

	academic	general	vocation
female	58	24	27
male	47	21	23

```
> round(prop.table(table(d0$gender,d0$prog),2)*100,2)
```

	academic	general	vocation
female	55.24	53.33	54.00
male	44.76	46.67	46.00

```
> # The proportion of females choosing the three different programs is almost similar,
55.24%, 53.33%, 54.00%
```

```
> # Likewise,the proportion of males choosing the three different programs is almost similar,
44.76%, 46.67%, 46.00%
```

```
> # Therefore, for different levels of program, proportions of female and male is close to
proportion of female and male in total population
```

```
> # Therefore, gender is not a good predictor.
```

```
>
```

```
> table(d0$ses,d0$prog)
```

	academic	general	vocation
high	42	9	7
low	19	16	12
middle	44	20	31

```
> round(prop.table(table(d0$ses,d0$prog),2)*100,2)
```

```
      academic general vocation
high   40.00  20.00  14.00
low    18.10  35.56  24.00
middle 41.90  44.44  62.00
> # The proportion of high SES choosing the three different program is different,40.00%,
20.00%, 14.00%
> #Likewise, the proportion of low and middle SES choosing the three different program are
different
> # Therefore, SES is a good predictor to predict the program they choose.
>
```

**> #b)**

```
> library(nnet)
> m1=multinom(prog~.,data=d0)
# weights: 42 (26 variable)
initial value 219.722458
iter 10 value 171.814970
iter 20 value 153.793692
iter 30 value 152.935260
final value 152.935256
converged
> summary(m1)
Call:
multinom(formula = prog ~ ., data = d0)
```

Coefficients:

```
(Intercept) gendermale raceasian racehispanic racewhite  seslow
general    3.631901 -0.09264717 1.352739 -0.6322019 0.2965156 1.09864111
vocation   7.481381 -0.32104341 -0.700070 -0.1993556 0.3358881 0.04747323
  sesmiddle schtyppublic  read  write  math  science
general  0.7029621  0.5845405 -0.04418353 -0.03627381 -0.1092888 0.10193746
vocation 1.1815808  2.0553336 -0.03481202 -0.03166001 -0.1139877 0.05229938
  socst
general -0.01976995
vocation -0.08040129
```

Std. Errors:

```
(Intercept) gendermale raceasian racehispanic racewhite  seslow
general    1.823452 0.4548778 1.058754  0.8935504 0.7354829 0.6066763
vocation   2.104698 0.5021132 1.470176  0.8393676 0.7480573 0.7045772
  sesmiddle schtyppublic  read  write  math  science
```

```
general 0.5045938 0.5642925 0.03103707 0.03381324 0.03522441 0.03274038
vocation 0.5700833 0.8348229 0.03422409 0.03585729 0.03885131 0.03424763
      socst
general 0.02712589
vocation 0.02938212
```

```
Residual Deviance: 305.8705
AIC: 357.8705
```

```
#Coefficients read  write  math  science  socst
#general -0.05445264 -0.03716360 -0.1037470 0.1065258 -0.01786542
#vocation -0.04078359 -0.03220268 -0.1099712 0.0537472 -0.07959798
#the unexpected coefficients is the what science subject have
#which is the only one subject have positive influence
#to choose general or vocation program rather than academic program
#(Since the base level of program is academic, so when other predictors hold
#the score of subject science is high, the probability of choosing general and vocation program
will increase
#And because the sum of probability of three program is 1
#the probability of choosing academic program will decreases.)
```

**> #c)**

```
> n1=nrow(d0)
> step1=stepAIC(m1,k=log(n1))
Start: AIC=443.63
prog ~ gender + race + ses + schtyp + read + write + math + science +
      socst
```

```
# weights: 39 (24 variable)
initial value 219.722458
iter 10 value 171.468391
iter 20 value 153.592758
final value 153.142827
converged
# weights: 33 (20 variable)
initial value 219.722458
iter 10 value 172.925326
iter 20 value 156.065379
final value 155.776076
```

```
#didn't show all
```

```
Call:
```

```
multinom(formula = prog ~ ., data = d0)
```

```
Step: AIC=374.52  
prog ~ math + socst
```

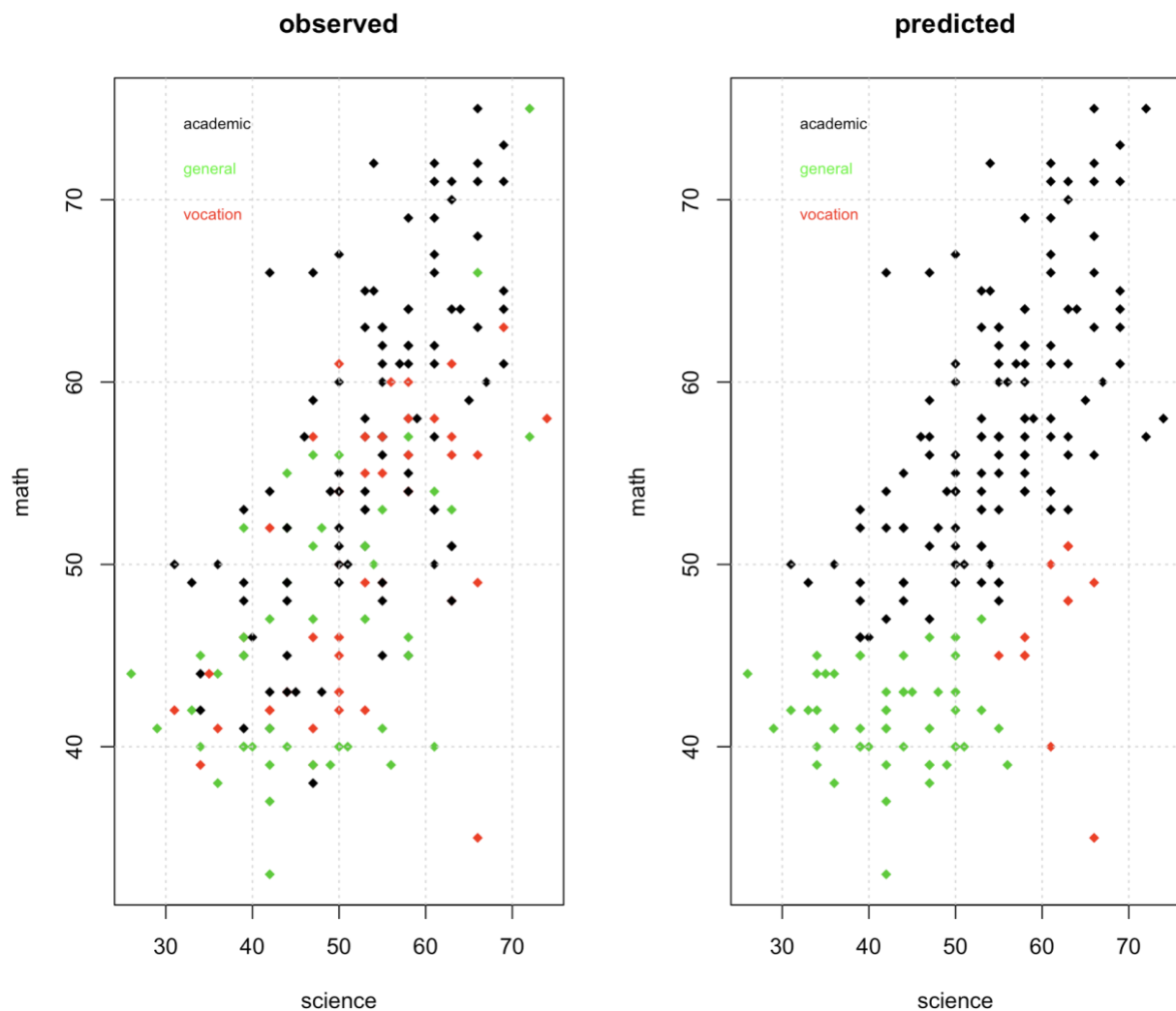
```
# weights: 9 (4 variable)  
initial value 219.722458  
final value 181.550192  
converged  
# weights: 9 (4 variable)  
initial value 219.722458  
final value 178.114227  
converged
```

```
      Df  AIC  
<none> 374.52  
- socst 2 377.42  
- math  2 384.29  
>
```

#iteration stop when BIC=374.52, the variables I choose is math and socst

**> #d)**

```
> m3=multinom(prog~math+science,data=d0)  
# weights: 12 (6 variable)  
initial value 219.722458  
iter 10 value 175.074512  
final value 175.074335  
converged  
> par(mfrow=c(1,2))  
> #predict probabilities  
> pi.hat=predict(m3,newdata = d0,type="probs")  
> ypred = apply(pi.hat,1,which.max)  
> labels=c("academic","general","vocation")  
> colors = c("black","green","red")  
> plot(math~science,d0,col=d0$prog,pch=18,main="observed")  
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)  
> grid()  
> plot(math~science,d0,col=ypred,pch=18,main="predicted")  
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)  
> grid()  
>
```



```
> #error rate
> 1-sum(diag(prop.table(table(d0$prog,ypred))))
[1] 0.425
> #so the error rate is 42.5%
>
```

**> #e)**

```
> library(class)
> y = d0$prog
> str(d0)
'data.frame': 200 obs. of 10 variables:
 $ gender : Factor w/ 2 levels "female","male": 2 1 2 2 2 2 2 2 2 ...
 $ race   : Factor w/ 4 levels "african-amer",...: 4 4 4 4 4 4 1 3 4 1 ...
 $ ses    : Factor w/ 3 levels "high","low","middle": 2 3 1 1 3 3 3 3 3 ...
 $ schtyp : Factor w/ 2 levels "private","public": 2 2 2 2 2 2 2 2 2 ...
 $ prog   : Factor w/ 3 levels "academic","general",...: 2 3 2 3 1 1 2 1 2 1 ...
```

```

$ read : int 57 68 44 63 47 44 50 34 63 57 ...
$ write : int 52 59 33 44 52 52 59 46 57 55 ...
$ math : int 41 53 54 47 57 51 42 45 54 52 ...
$ science: int 47 63 58 53 53 63 53 39 58 50 ...
$ socst : int 57 61 31 56 61 61 61 36 51 51 ...
> x = d0[,c(8,9)] #math and science
> str(x)
'data.frame': 200 obs. of 2 variables:
 $ math : int 41 53 54 47 57 51 42 45 54 52 ...
 $ science: int 47 63 58 53 53 63 53 39 58 50 ...
> x=scale(x)
> head(x)
      math science
1 -1.24300207 -0.4898549
2  0.03789315  1.1261613
3  0.14463442  0.6211562
4 -0.60255446  0.1161512
5  0.46485822  0.1161512
6 -0.17558939  1.1261613
> #k=3
> set.seed(1)
> ypred3=knn(x,x,y,3)
> par(mfrow=c(1,2))
> plot(math~science,d0,col=d0$prog,pch=18,main="observed")
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)
> grid()
> plot(math~science,d0,col=ypred3,pch=18,main="predicted")
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)
> grid()
> #error rate
> 1-sum(diag(prop.table(table(y,ypred3))))
[1] 0.31
> #so the error rate is 31%

> #k=5
> set.seed(1)
> ypred5=knn(x,x,y,5)
> plot(math~science,d0,col=d0$prog,pch=18,main="observed")
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)
> grid()
> plot(math~science,d0,col=ypred5,pch=18,main="predicted")
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)
> grid()
> #error rate

```

```
> 1-sum(diag(prop.table(table(y,ypred5))))  
[1] 0.325  
> #so the error rate is 32.5%
```

