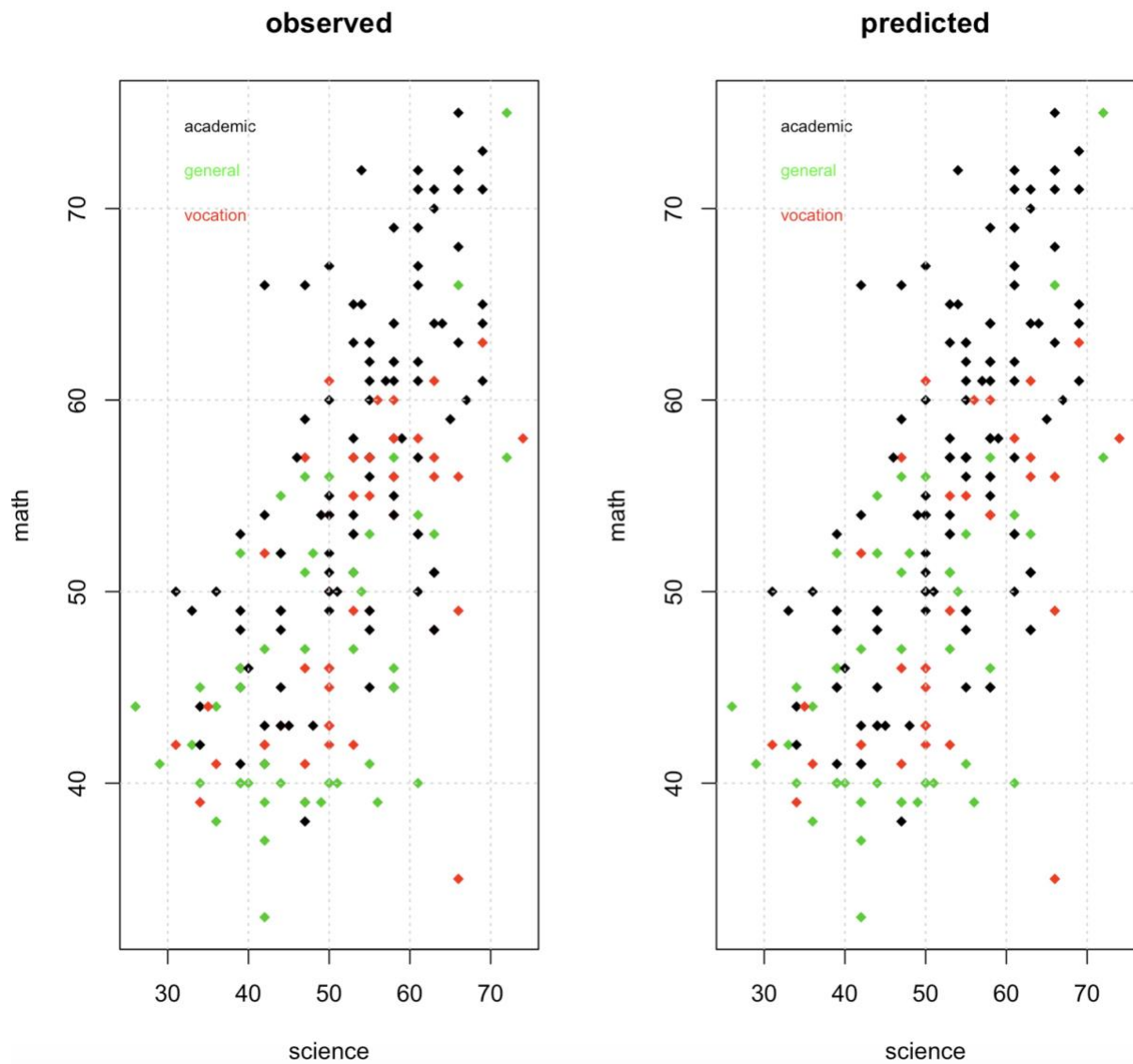


Shuting Chen

HW6

```
> library(faraway)
> library(carData)
> library(car)
> #1
> d0=data.frame(hsb)
> dim(d0)
[1] 200 11
> head(d0)
  id gender race  ses schtyp  prog read write math science socst
1  70  male white  low public general  57  52  41   47  57
2 121 female white middle public vocation  68  59  53   63  61
3  86  male white  high public general  44  33  54   58  31
4 141  male white  high public vocation  63  44  47   53  56
5 172  male white middle public academic  47  52  57   53  61
6 113  male white middle public academic  44  52  51   63  61
> set.seed(1)
> bag1=randomForest(prog~math+science,d0,mtry=2,importance=T)
> par(mfrow=c(1,2))
> ypred=predict(bag1,newdata = d0)
> labels=c("academic","general","vocation")
> colors = c("black","green","red")
> plot(math~science,d0,col=d0$prog,pch=18,main="observed")
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)
> grid()
> plot(math~science,d0,col=ypred,pch=18,main="predicted")
> legend("topleft",legend=labels,bty="n",text.col = colors,cex = 0.7)
> grid()
> par(mfrow=c(1,1))
> 1-sum(diag(prop.table(table(d0$prog,ypred))))
[1] 0.09
> #error rate is 0.09
```



> #2

> #a)

> library(ISLR)

> library(randomForest)

> d0=data.frame(Caravan)

> sapply(Caravan,table)

\$MOSTYPE

```

1 2 3 4 5 6 7 8 9 10 11 12 13 15 16 17 18 19 20
124 82 249 52 45 119 44 339 278 165 153 111 179 5 16 9 19 3 25
21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39

```

15 98 251 180 82 48 50 25 86 118 205 141 810 182 214 225 132 339 328
40 41
71 205

\$MAANTHUI

1 2 3 4 5 6 7 8 10
5267 505 39 2 1 1 5 1 1

\$MGEMOMV

1 2 3 4 5
284 2131 2646 693 68

\$MGEMLEEF

1 2 3 4 5 6
74 1452 3000 1073 193 30

\$MOSHOOFD

1 2 3 4 5 6 7 8 9 10
552 502 886 52 569 205 550 1563 667 276

.....

Does not show all the result

.....

```
> dim(d0)
[1] 5822 86
> d0=d0[,-which(names(d0)%in% c("PVRAAUT","AVRAAUT"))]
> train=1:1000
> d0train=d0[train,]
> d0test=d0[-train]
> set.seed(1)
> forest1=randomForest(Purchase~.,d0train,mtry=83,importance=T)
> importance(forest1)
```

	No	Yes	MeanDecreaseAccuracy	MeanDecreaseGini
MOSTYPE	9.36748604	-3.36760792	9.25053009	4.517191004
MAANTHUI	1.95373123	-0.06535802	1.91469335	1.151085123
MGEMOMV	7.83167926	-2.72448820	7.33494542	1.140477450
MGEMLEEF	4.10784489	-0.12529425	4.14172311	0.775692035
MOSHOOFD	3.77920421	0.51099817	3.97966332	1.451814974
MGODRK	4.08285089	-2.46532927	3.59848175	1.461014416

MGODPR	7.95704145	2.42704345	8.21955518	3.546813354
MGODOV	2.12706237	1.56932854	2.46568801	1.310703255

.....

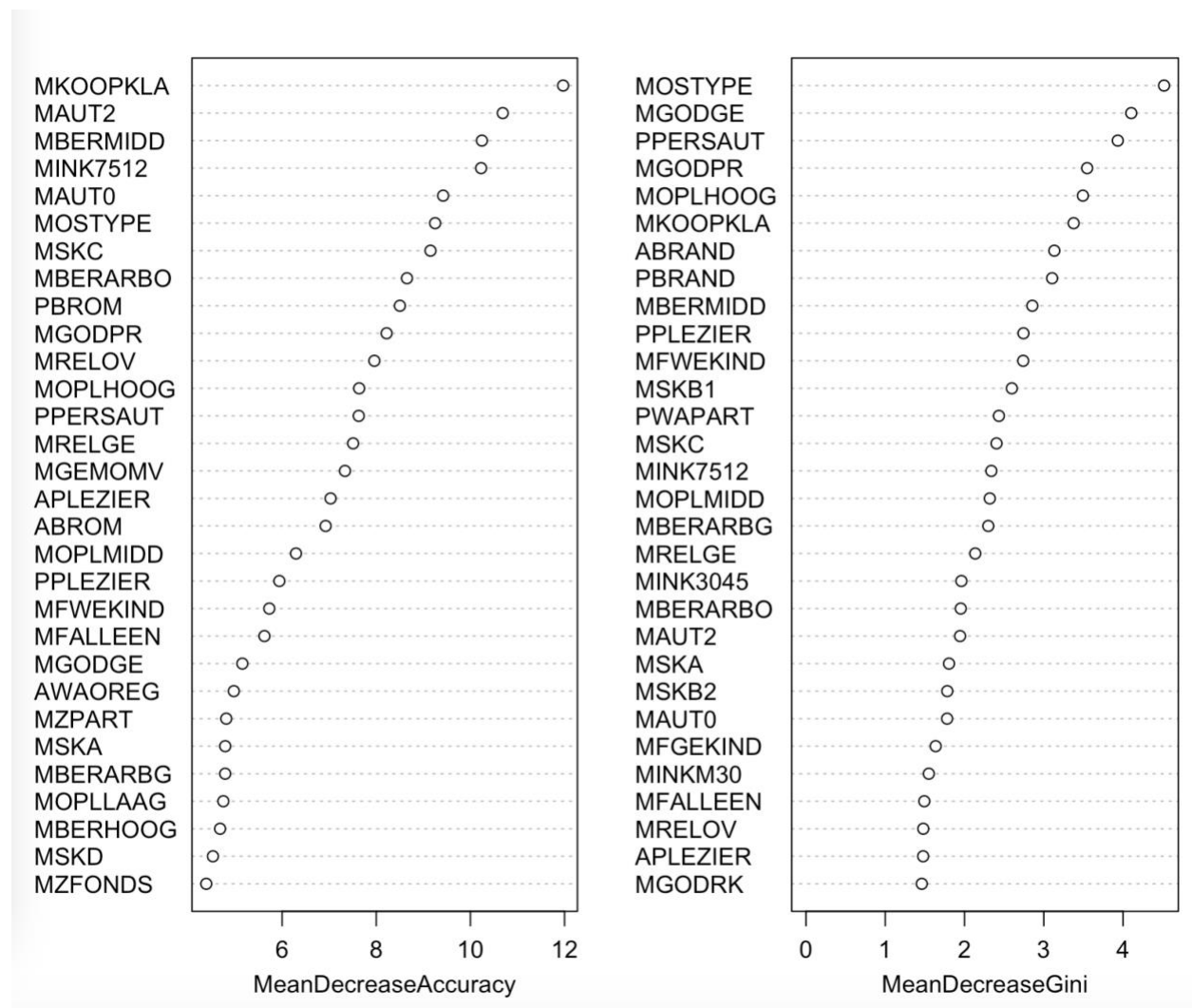
Does not show all the result

.....

AFIETS	-0.56967023	-2.62350658	-1.20496853	0.793364260
AINBOED	0.05537678	0.00000000	0.05606031	0.031434525
ABYSTAND	1.09725797	-1.09627420	0.86464848	0.248707510

> varImpPlot(forest1,main="")

> #the best predictor is MOSTYPE



> #b)

```
> library(gbm)
> set.seed(1)
> n=nrow(d0)
> y=rep(0,n)
> y[d0$Purchase=="Yes"]=1
> d0$Purchase=y
> train=1:1000
> d0train=d0[train,]
> d0test=d0[-train,]
> boost1 = gbm(Purchase~.,data=d0[train,],distribution="bernoulli",
n.trees=1000,shrinkage=0.01)
> summary(boost1)
      var  rel.inf
PPERSAUT PPERSAUT 14.63504779
MKOOPKLA MKOOPKLA  9.47091649
MOPLHOOG MOPLHOOG  7.31457416
MBERMIDD MBERMIDD  6.08651965
PBRAND   PBRAND   4.66766122
MGODGE   MGODGE   4.49463264
ABRAND   ABRAND   4.32427755
MINK3045 MINK3045  4.17590619
MOSTYPE  MOSTYPE  2.86402583
PWAPART  PWAPART  2.78191075
MAUT1    MAUT1    2.61929152
MBERARBG MBERARBG  2.10480508
MSKA     MSKA     2.10185152
MAUT2    MAUT2    2.02172510
.....
```

Does not show all the result

```
.....
AGEZONG  AGEZONG  0.00000000
AWAOREG  AWAOREG  0.00000000
AZEILPL  AZEILPL  0.00000000
APLEZIER APLEZIER  0.00000000
AFIETS   AFIETS   0.00000000
AINBOED  AINBOED  0.00000000
ABYSTAND ABYSTAND  0.00000000
> grid()
>
> #the most important predictor is PPERSAUT
```

