```
> library(haven)
                      # load the package that contain functions to read
SAS files into R environment
> library(plm)
                     # package that can process panel data in R
> library(margins) # package that can compute the marginal effect of
mode1
> file<-read_sas("college_admission.sas7bdat") # read SAS file into R</pre>
> file$COLLEGE<-file$PSECHOICE!=1</pre>
> file$COLLEGE<-as.numeric(file$COLLEGE) # set up COLLEGE variable</pre>
> probit<-glm(COLLEGE~GRADES+FAMINC+FAMSIZ+PARCOLL+FEMALE+BLACK, family=b
inomial(link="probit"),data=file)
> summary(probit)
                                    # probit model
Call:
glm(formula = COLLEGE ~ GRADES + FAMINC + FAMSIZ + PARCOLL +
   FEMALE + BLACK, family = binomial(link = "probit"), data = file)
Deviance Residuals:
   Min
           10 Median
                         3Q
                                 Max
-3.6547
         0.0891 0.3615
                         0.6878
                                 1.7145
Coefficients:
          Estimate Std. Error z value Pr(>|z|)
(Intercept) 2.693704  0.285643  9.430 < 2e-16 ***
          -0.294552  0.027452  -10.730 < 2e-16 ***
GRADES
          FAMINC
          FAMSIZ
          0.476540 0.141154 3.376 0.000735 ***
PARCOLL
          0.023788 0.101626 0.234 0.814926
FEMALE
          0.610896  0.215740  2.832  0.004631 **
BLACK
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1058.86 on 999 degrees of freedom
Residual deviance: 832.44 on 993 degrees of freedom
AIC: 846.44
```

Number of Fisher Scoring iterations: 6

```
>
> logit<-glm(COLLEGE~GRADES+FAMINC+FAMSIZ+PARCOLL+FEMALE+BLACK, family=bi
nomial(link="logit"),data=file)
                                     # logit model
> summary(logit)
call:
qlm(formula = COLLEGE ~ GRADES + FAMINC + FAMSIZ + PARCOLL +
   FEMALE + BLACK, family = binomial(link = "logit"), data = file)
Deviance Residuals:
   Min
           10 Median
                           3Q
                                 Max
        0.1252 0.3579
-3.3697
                         0.6656
                                 1.7735
Coefficients:
          Estimate Std. Error z value Pr(>|z|)
(Intercept) 4.541835 0.519120 8.749 < 2e-16 ***
         GRADES
         0.013042  0.003897  3.346  0.000819 ***
FAMINC
FAMSIZ -0.088256 0.064466 -1.369 0.170987
          0.835371 0.262977 3.177 0.001490 **
PARCOLL
FEMALE 0.034279 0.176918 0.194 0.846365
      1.073344 0.372292 2.883 0.003938 **
BLACK
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 1058.86 on 999 degrees of freedom
Residual deviance: 829.95 on 993 degrees of freedom
AIC: 843.95
Number of Fisher Scoring iterations: 6
> sum(file$COLLEGE)/length(file$COLLEGE) # the rate that attend coll
ege
[1] 0.778
> mean(file$FAMINC)
                                    # sample average of family income
variable
[1] 51.3935
```

```
> x<-data.frame("GRADES"=10,"FAMINC"=51.3935,"FAMSIZ"=4,"PARCOLL"=1,"FEM
ALE"=1, "BLACK"=1)
> predict.glm(probit,newdata=x)
                                         # predict the result of a new obs
ervation with probit model
      1
0.924101
> marginal_effects(probit,x)
                                        # see the marginal effect of the
variables in this observation
 dydx_GRADES dydx_FAMINC dydx_FAMSIZ dydx_PARCOLL dydx_FEMALE dydx_BLACK
1 -0.07667187 0.00140358 -0.01382398 0.1240433 0.006192081 0.1590163
>
>
> data<-read_sas("college2.sas7bdat") # read the other SAS file to R</pre>
> pdata<-plm.data(data,index=c("hh","t"))</pre>
                                               # set up panel data
                                                    # ordinary least squar
> ols<-lm(l\sim x, data=pdata)
e model
> pool<-plm(l~x,data=pdata,model="pooling")</pre>
> fixed<-plm(l~x,data=pdata,model="within")</pre>
> summary(fixed)
                                            # fixed effect model
Oneway (individual) effect Within Model
call:
plm(formula = 1 \sim x, data = pdata, model = "within")
Balanced Panel: n=40, T=3, N=120
Residuals:
  Min. 1st Qu. Median 3rd Qu.
                                  Max.
-1.6700 -0.5490 -0.0494 0.5540 2.0500
Coefficients:
 Estimate Std. Error t-value Pr(>|t|)
x 0.020742 0.020908 0.9921 0.3242
Total Sum of Squares: 77.107
Residual Sum of Squares: 76.159
R-Squared
          : 0.012305
     Adj. R-Squared: 0.0081005
F-statistic: 0.984166 on 1 and 79 DF, p-value: 0.3242
```

```
> random<-plm(l~x,data=pdata,model="random")</pre>
> summary(random)
                                          # random effect model
Oneway (individual) effect Random Effect Model
  (Swamy-Arora's transformation)
Call:
plm(formula = 1 \sim x, data = pdata, model = "random")
Balanced Panel: n=40, T=3, N=120
Effects:
              var std.dev share
idiosyncratic 0.9640 0.9819 0.571
individual 0.7251 0.8515 0.429
theta: 0.4459
Residuals:
  Min. 1st Qu. Median 3rd Qu. Max.
-2.2600 -0.6970 0.0787 0.5530 2.2300
Coefficients:
          Estimate Std. Error t-value Pr(>|t|)
(Intercept) 0.9690324 0.5210052 1.8599 0.0653859 .
          0.0265755 0.0070126 3.7897 0.0002391 ***
Χ
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Total Sum of Squares: 126.61
Residual Sum of Squares: 112.88
R-Squared : 0.1085
     Adj. R-Squared : 0.1067
F-statistic: 14.3618 on 1 and 118 DF, p-value: 0.00023906
> pFtest(fixed,pool)
                                       # test the hypothesis that individu
al intercepts all equal
      F test for individual effects
data: 1 \sim x
F = 3.1755, df1 = 39, df2 = 79, p-value = 6.647e-06
alternative hypothesis: significant effects
```