O10.18:

- (a)12.15% of mothers have some college education
 - 11.68% of fathers have some college education.

```
> mean(mroz_sub$MOTHERCOLL, na.rm = TRUE)
[1] 0.1214953
> mean(mroz_sub$FATHERCOLL, na.rm = TRUE)
```

(B)The correlation between EDUC and MOTHERCOLL is 0.3595. The correlation between EDUC and FATHERCOLL is 0.3985. The education level from parents show positive correlation with education level for married women.

(c) The 95% confidence interval for one IV model

```
> confint(iv_model, level = 0.95)["educ", ]
2.5 % 97.5 %
-0.001219763 0.153255678
```

(d) the first-stage equation for one IV

```
\hat{EDUC} = 12.0791 + 0.0562 \cdot EXPER - 0.00196 \cdot EXPER^2 + 2.5171 \cdot MOTHER COLL + \nu
```

```
> summary(first_stage)
lm(formula = educ ~ exper + I(exper^2) + MOTHERCOLL, data = mroz_sub)
Residuals:
           10 Median
                         30
                               Max
-7.4267 -0.4826 -0.3731 1.0000 4.9353
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
MOTHERCOLL 2.517068 0.315713 7.973 1.46e-14 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 2.133 on 424 degrees of freedom
Multiple R-squared: 0.1347,
                           Adjusted R-squared: 0.1285
F-statistic: 21.99 on 3 and 424 DF, p-value: 2.965e-13
```

null hypothesis H_0 : $\beta_{MOTHERCOLL} = 0$

Alternative hypothesis $H_1: \beta_{MOTHERCOLL} \neq 0$

The F value for MOTHERCOLL is 63.5631 larger than 10.We fail to reject H_0 , this indicates that MOTHERCOLL has a significant effect on EDUC in the regression model. MOTHERCOLL is a strong instrumental variable.

(e) The 95% confidence interval for two IV model

The 95% confidence interval of model use two instrumental variable is narrower than the model use only one instrumental variable

	2.5%	97.5%
one IV model	-0.00122	0.15326
two IV model	0.027512	0.14818

(f) the first-stage equation for one IV $E\hat{DUC} = 11.89026 + 0.04915 \cdot EXPER - 0.00145 \cdot EXPER^2 + 1.74995 \cdot MOTHERCOLL$

```
Call:
lm(formula = educ ~ exper + exper2 + MOTHERCOLL + FATHERCOLL,
   data = mroz_sub)
Residuals:
   Min
           1Q Median
                         3Q
                               Max
-7.2152 -0.3056 -0.2152 0.7627 5.0620
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
0.049149 0.040133
exper
                             1.225
                                     0.221
          -0.001449 0.001199 -1.209
exper2
                                      0.227
MOTHERCOLL 1.749947 0.322347 5.429 9.58e-08 ***
FATHERCOLL 2.186612 0.329917 6.628 1.04e-10 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 2.033 on 423 degrees of freedom
Multiple R-squared: 0.2161,
                           Adjusted R-squared: 0.2086
F-statistic: 29.15 on 4 and 423 DF, p-value: < 2.2e-16
```

 $+2.18661 \cdot FATHERCOLL + \nu$

The F-test statistic of the joint significance of MOTHERCOLL and FATHERCOLL is 56.96 > 10, so we reject the null hypothesis that the instruments are weak.

(g) The p-value is 0.6281333, indicating that your instrument variables do not have any issues in the regression model and there are no overidentifying restrictions.

O10.20:

(a) Based on this output, Microsoft stock appears to be risky relative to the market, as indicated by the positive and significant beta of 1.201840. The market return is a significant factor in explaining the excess return on Microsoft stock.

```
Call:
lm(formula = msft_excess ~ mkt_excess, data = capm5)
Residuals:
    Min
              1Q Median
                               3Q
-0.27424 -0.04744 -0.00820 0.03869 0.35801
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.003250 0.006036 0.538
                                       0.591
mkt_excess 1.201840 0.122152 9.839
                                       <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.08083 on 178 degrees of freedom
Multiple R-squared: 0.3523,
                              Adjusted R-squared: 0.3486
F-statistic: 96.8 on 1 and 178 DF, p-value: < 2.2e-16
```

(b) The F-statistic is 1857.6, which is far greater than the common rule-of-thumb threshold of 10. This indicates that RANKis a strong instrument for market excess return.

Thus, based on the first-stage regression output, RANK can be regarded as a strong IV for the market return variable in the CAPM model.

(c) The hypothesis that the market return is exogenous is not strongly supported at the 99% confidence level. However, there is some evidence of endogeneity at the 95% confidence level. Therefore, it's safe to conclude that the market return could be endogenous, but the evidence isn't overwhelmingly strong at the 1% significance level.

```
lm(formula = msft_excess ~ mkt_excess + v_hat, data = capm5)
Residuals:
                               3Q
              1Q Median
    Min
                                      Max
-0.27140 -0.04213 -0.00911 0.03423 0.34887
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.003018 0.005984 0.504 0.6146
mkt_excess 1.278318 0.126749 10.085
                                        <2e-16 ***
v_hat
          -0.874599 0.428626 -2.040 0.0428 *
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.08012 on 177 degrees of freedom
Multiple R-squared: 0.3672, Adjusted R-squared: 0.36
F-statistic: 51.34 on 2 and 177 DF, p-value: < 2.2e-16
```

(d) The IV estimation gives a slightly higher coefficient for mkt_excess compared to the OLS model, which suggests that the IV model might be correcting for endogeneity, leading to a slightly stronger estimated relationship between the market excess return and Microsoft excess return.

```
Call:
ivreg(formula = msft_excess ~ mkt_excess | RANK, data = capm5)
Residuals:
     Min
              1Q
                    Median
                                  3Q
-0.271625 -0.049675 -0.009693 0.037683 0.355579
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.003018 0.006044 0.499
mkt_excess 1.278318 0.128011 9.986 <2e-16 ***
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' '1
Residual standard error: 0.08092 on 178 degrees of freedom
Multiple R-Squared: 0.3508,
                             Adjusted R-squared: 0.3472
Wald test: 99.72 on 1 and 178 DF, p-value: < 2.2e-16
```

(e)The F-test statistic for the joint significance of these variables is 951.26>10, indicating that RANK and pos are jointly significantly different from zero. Therefore, the null hypothesis that RANK and pos are both equal to zero can be rejected, suggesting that these variables are meaningful in the first stage of your regression.

(f)At the 1% level we fail to reject the null hypothesis that the market return is exogenous.

(g)The IV/2SLS estimate gives a slightly higher coefficient than the OLS estimate (1.2831 vs. 1.2018). This suggests that the instrument (v_hat2) might be capturing some part of the variation that OLS does not, leading to a higher estimate.

```
Call:
ivreg(formula = msft_excess ~ mkt_excess | rank + pos, data = capm5)
Residuals:
    Min
              1Q Median
                                3Q
                                        Max
-0.27168 -0.04960 -0.00983 0.03762 0.35543
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.003004 0.006044
                                0.497
                                           0.62
mkt_excess 1.283118 0.127866 10.035
                                         <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 0.08093 on 178 degrees of freedom
Multiple R-Squared: 0.3507,
                               Adjusted R-squared: 0.347
Wald test: 100.7 on 1 and 178 DF, p-value: < 2.2e-16
```

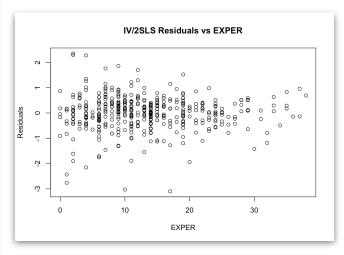
(h)At the 5% significance level, we fail to reject the null hypothesis, which means that both RANK and POS are valid instrumental variables and can be considered exogenous variables.

```
R squared= 0.003102574
nRsquared= 0.5584634
p-value= 0.45488
```

Q10.24:

(a) The points appear to be randomly scattered without any visible trend or systematic structure, which suggests that the assumption of homoskedasticity holds.

```
Call:
ivreg(formula = lwage ~ exper + exper2 + educ | exper + exper2 +
   mothereduc + fathereduc, data = mroz_sub)
Residuals:
   Min
            10 Median
                            30
                                   Max
-3.0986 -0.3196 0.0551 0.3689 2.3493
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept) 0.0481003 0.4003281 0.120 0.90442
exper
            0.0441704 0.0134325
                                 3.288 0.00109 **
exper2
           -0.0008990 0.0004017 -2.238 0.02574 *
            0.0613966 0.0314367 1.953 0.05147 .
educ
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.6747 on 424 degrees of freedom
Multiple R-Squared: 0.1357, Adjusted R-squared: 0.1296
Wald test: 8.141 on 3 and 424 DF, p-value: 2.787e-05
```



(b) We reject the null hypothesis of homoskedasticity.

```
> nR2
[1] 7.438552
> p_value
[1] 0.006384122
```

(c)The 95% Confidence Interval (using Robust SE)=[-0.00393,0.12673]

	Estimate	Baseline_SE	${\tt Robust_SE}$	Increased_SE
(Intercept)	0.04810	0.40033	0.42980	Yes
exper	0.04417	0.01343	0.01555	Yes
exper2	-0.00090	0.00040	0.00043	Yes
educ	0.06140	0.03144	0.03334	Yes

(d)Bootstrap SE is larger than the baseline SE but smaller than the robust SE. The 95% Confidence Interval (using Bootstrap SE)=[-0.002,0.1248]

```
> cat(" 95% CI for EDUC (bootstrap) = [",
+    round(ci_boot[1],4), ", ", round(ci_boot[2],4), "]\n")
95% CI for EDUC (bootstrap) = [ -0.002 , 0.1248 ]
```