## Assignment 1

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## Question 1: The sample selection model.

A researcher aims to gain insight in the potential earnings of the non-employed. (In the data, the nonemployed can be identified by a missing value for the earnings variable). She realizes that the sample of observed wages may be subject to sample selection.

(a) Run an OLS regression for log-earnings on schooling, age, and age squared. Present the results and comment on the estimates.

```
model <- lm(data = data, formula = logWage ~ schooling + age + age2)
stargazer(model, style = "AER",
          font.size = "small",
         header = F)
```

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	$\log$ Wage
schooling	0.216***
	(0.032)
n mo	-0.342
age	(0.521)
	(0.021)
age2	-0.011
	(0.008)
Constant	26.409***
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	(8.057)
Observations	416
R <sup>2</sup>	0.815
Adjusted R <sup>2</sup>	0.813
Residual Std. Error	1.499 (df = 412)
F Statistic	$604.261^{***}$ (df = 3; 412)
Notes:	***Significant at the 1 percent level.

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The results show that 1 additional year of schooling has an effect of 0.216 on log(Wage), which means that it has an effect of 1.2411024 on schooling. Hence, each year of schooling supposedly increases earnings with 1.2411024.

<sup>\*\*</sup>Significant at the 5 percent level.

<sup>\*</sup>Significant at the 10 percent level.

(b) Briefly discuss the sample selection problem that may arise in using these OLS estimates for the purpose of predicting the potential earnings of the non-employed. Formulate the sample selection model. In your answer, include an explanation why OLS may fail in this context.

An individual is only in this dataset if they earn wages, i.e. if they are employed. Being employed itself is not randomly allocated, but rather, a function of e.g. age, age^2, and schooling. Hence, the estimates of schooling on earnings are conditional on having earnings to begin with, whereas unbiased estimates must also include those individuals

Formally,  $\mathbb{E}[\text{Earnings}|\text{Having a job}] \cdot \mathbb{P}[\text{Having a job}] + \mathbb{E}[\text{Earnings}|\text{Not having a job}] \cdot (1 - \mathbb{P}[\text{Having a job}]).$ 

## Question 2: Earnings and Schooling

The same researcher is interested in estimating the causal effect of schooling on earnings for employed individuals only. As a consequence, she performs the subsequent analysis on the (sub)sample of employed individuals.

(a) Discuss the estimation of the causal effect of schooling on earnings by OLS. In particular, address whether or not it is plausible that regularity conditions for applying OLS are satisfied.

It is not plausible that the regularity conditions are satisfied. In particular, an observable such as an individual's **ability** might be correlated with the wage, but also with the decision to live close to a school. Hence, the estimates suffer from endogeneity.

- (b) The researcher has collected data on two potential instrumental variables subsidy and distance for years of schooling.
  - distance measures the distance between the school location and the residence of the individual while at school-going age.
  - subsidy is an indicator depending on regional subsidies of families for covering school expenses.

The researcher has the option to use only distance as an instrumental variable, or to use only the instrumental variable subsidy, or to use both distance and subsidy as instrumental variables. Perform instrumental variables estimation for these three options. Which option do you prefer? Include in your answer the necessary analyses and numbers on which you base your choice.

We consider that the second option, to include only subsidy as an instrument, is the best option. The reason is that distance is unlikely to satisfy the exclusion restriction: distance is (to a certain extent) an

Table 2:

	$\log$ Wage		
	(1)	(2)	(3)
age	-0.192	-0.233	-0.229
	(0.587)	(0.546)	(0.547)
age2	-0.014	-0.013	-0.013
	(0.010)	(0.009)	(0.009)
schooling	0.470	0.401***	0.408***
	(0.299)	(0.106)	(0.102)
Constant	22.681**	23.694***	23.589***
	(9.704)	(8.517)	(8.530)
Observations	416	416	416
$\mathbb{R}^2$	0.786	0.799	0.798
Adjusted $R^2$	0.784	0.798	0.797
Residual Std. Error ( $df = 412$ )	1.613	1.560	1.565
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Notes:

endogenous variable: wealthier (or more able) parents may choose to live closer to school, and invest more in the education of their children (or genetically transmit ability). Since a potentially endogenous instrument must not be used as such, we prefer the estimates in equation 2. However, we see that the results show that distance has no predictive power in schooling, thus showing that the endogeneity is very small. Conditional on subsidy being a good instrument, then, the potential endogeneity does not substantially changes the estimates of schooling on earnings.

 $<sup>\</sup>ensuremath{^{***}}\mathbf{Significant}$  at the 1 percent level.

<sup>\*\*</sup>Significant at the 5 percent level.

<sup>\*</sup>Significant at the 10 percent level.