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First Class

Part I -- β

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Intro to class - walk through syllabus

How to think about regression coefficients

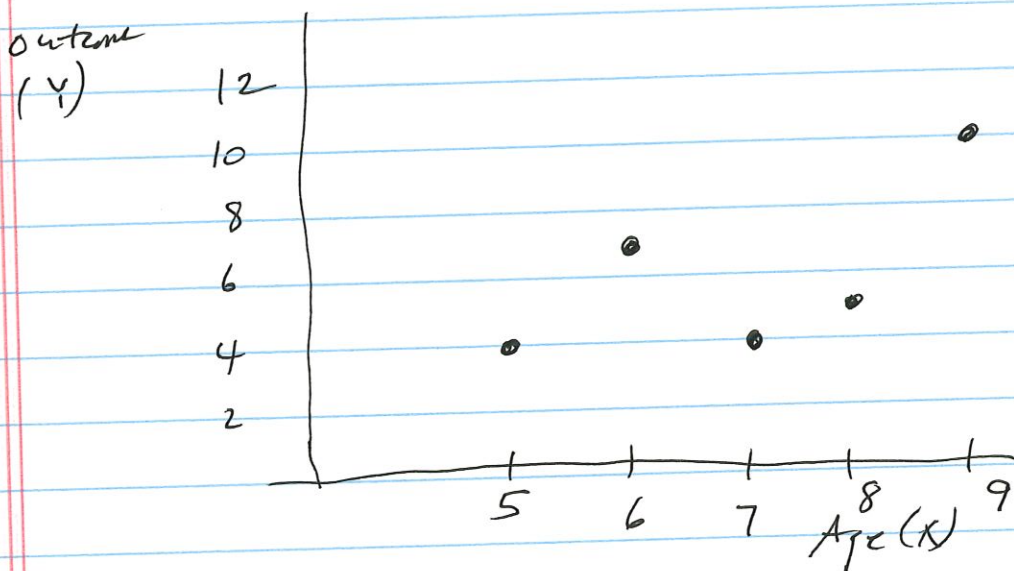
Data

| Age (x) | School outcome (Y) |
|---------|--------------------|
| 5 | 4 |
| 6 | 7 |
| 7 | 4 |
| 8 | 5 |
| 9 | 10 |

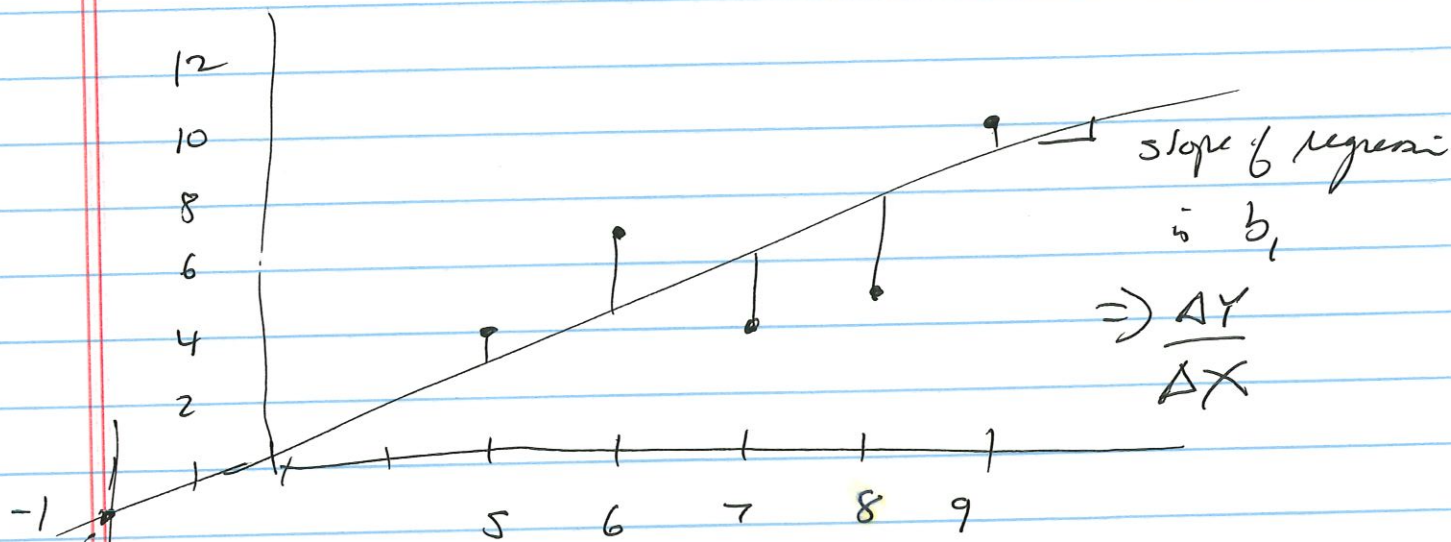
$$\bar{x} = 7$$
$$\bar{y} = 6 \left(= \frac{30}{5} \right)$$

A ~~simple~~ ^{bivariate} regression tells you the "best fitting" linear relationship between $x : Y$

Plot it:



if I hand you a thin rod, how are you going to place it to best fit it to these data?



If you run a linear regression, you get a & b , that best fits the data

$$Y = a + b_1 X$$

"Best fit" is based on a "least squares" criterion -

take all of the vertical distances, square them and make that as small as possible

It turns out that "best fitting" equation is $a = -1$
 $b = +1$

$$Y = -1 + 1 \cdot \text{Age}$$

how to interpret 1.0 slope?

Each additional unit of X is associated with a 1.0 unit increase in Y

How to interpret the ~~coef~~? -1

$$Y = a + b_1 X$$

$$\Rightarrow a = Y \text{ when } X = 0$$

\Rightarrow predict test score when age = 0!

The constant term (a) rarely tells you anything useful

All of our attention is focused on the slope coefficients

Slope coefficient

5 observations

Formula?

$$b = \frac{\sum_{i=1}^5 (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^5 (x_i - \bar{x})^2}$$

That doesn't help very much!

Apply it to our data. First off, "center" the data by subtracting \bar{x} from x_i 's

| i | new X | Age (x) | Y |
|---|-------|---------|----|
| 1 | -2 | 5 | 4 |
| 2 | -1 | 6 | 7 |
| 3 | 0 | 7 | 4 |
| 4 | 1 | 8 | 5 |
| 5 | 2 | 9 | 10 |

Let's first do the denominator:

$$\sum_{i=-2}^2 (x_i - \bar{x})^2 = (-2)^2 + (-1)^2 + 0^2 + 1^2 + 2^2$$

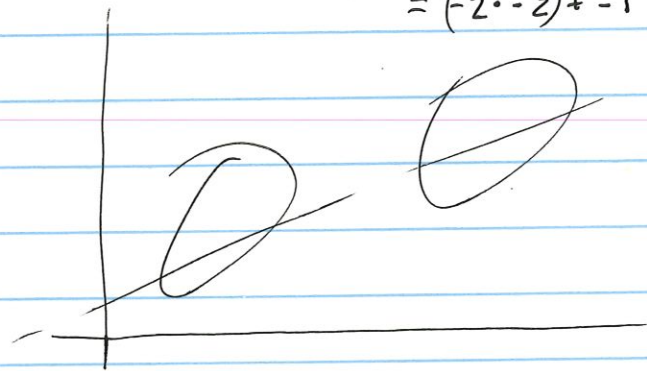
$$= 4 + 1 + 0 + 1 + 4$$

$$= 10$$

⇒ a number that is used to scale the numerator

So $b = \frac{\text{bunch of stuff}}{10}$

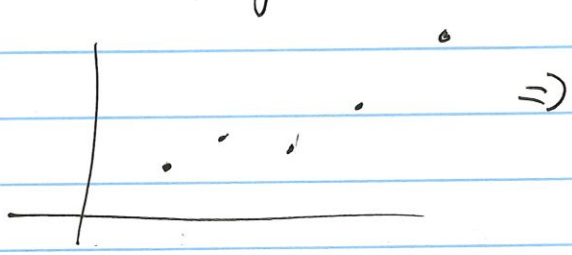
Numerator: $-2 \cdot (4-6) + (-1)(7-6) + 0(4-6)$
 $(+1)(5-6) + 2(10-6)$
 $= (-2 \cdot -2) + -1 \cdot -1 + 8 = 10$



Really just a bunch
of stuff on the right
half means a
bunch of stuff on
the left

\Rightarrow a weighted average of high

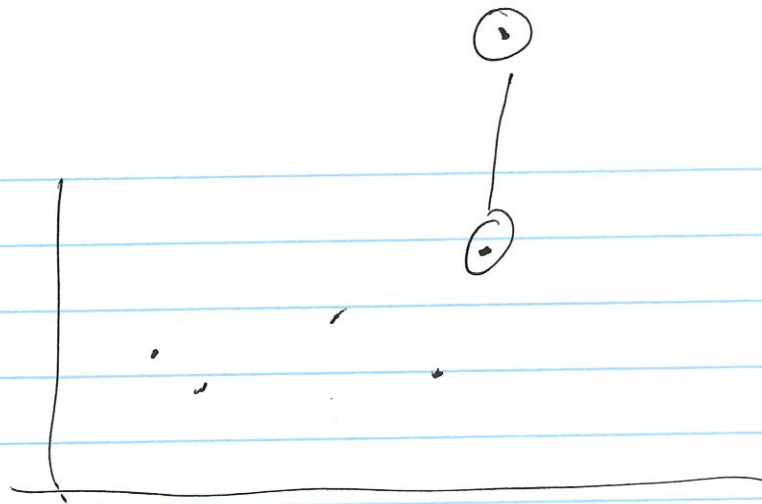
Note: more extreme Xs get more weight in determining
the slope



\Rightarrow why not just sample
very high and very
low X's

\Rightarrow measure Y at ages 5
and 9 but not
in between

\Rightarrow why not outliers associated with very large or
very small X values.



week 1
-7-

huge influence

What about the influence of outliers at \bar{X} ?

BREAKOUT ROOM

\Rightarrow outliers at \bar{X} have no effect on the slope

why gonna anything at age 7??

So what is b_1 and a

$$\text{Numerator} = \begin{matrix} 4 & -1 & -1 & +8 \\ (-2 \cdot -2) & -1(7-6) & +1(-1) & +2 \cdot 4 = 10 \end{matrix}$$

Denominator = 10

Slope = 1

$$\Rightarrow y = a + X$$

intercept

First class Part II - Tables

Now let's take a look at Table 3

Regression:

$$Ed = a + b_1 Inc + b_2 Controls$$

in years

Harcove's:

1. Theory 101
2. Table 3
3. Syllabus

Controls = child race and gender, race, maternal schooling, age of mother at birth, region, single-parent family structure, maternal employment.

b_1 is .14 \Rightarrow a one unit change in income (.02) is associated with a .14 unit change in education ($\frac{1}{7}$ th of a year)

What is a one-unit change in income?

\$10,000!

So, all else constant, an additional \$10,000/year is associated with a $\frac{1}{7}$ th of a year increase in schooling

for 15 years

\Rightarrow small!!

Is it causal?

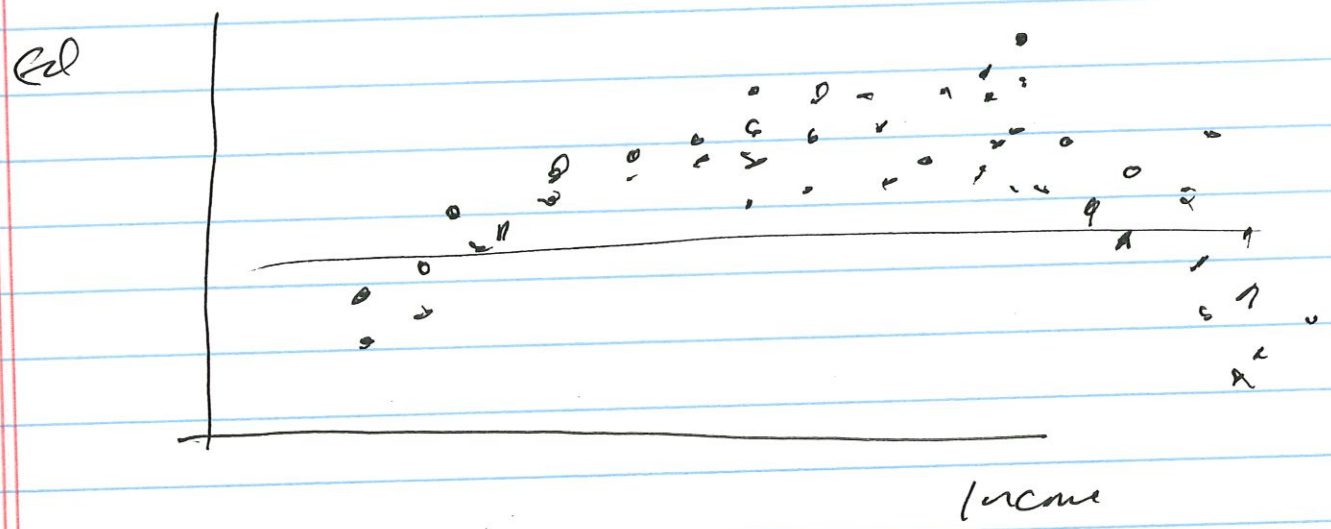
GREAT QUESTION!

(1) identification; (2) functional form

First class Part II Table 3

Don't stop with just $Y = a + bX$
and $b = .14$

What if non-linear relationship?



In this case, best fitting line is flat
 \Rightarrow look like no relationship

What fits better? A parabola

$$Y = a + b_1 \text{inc} + b_2 \text{inc}^2 + \text{Constant}$$

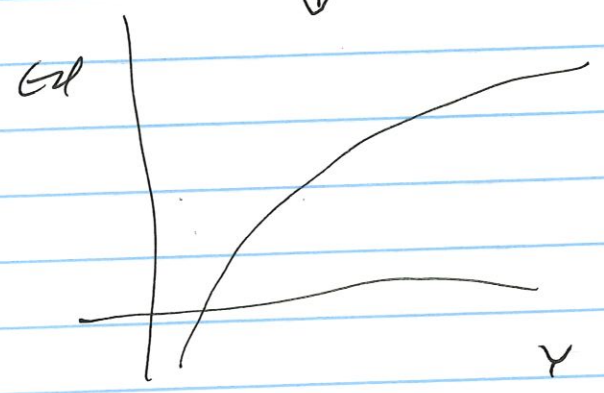
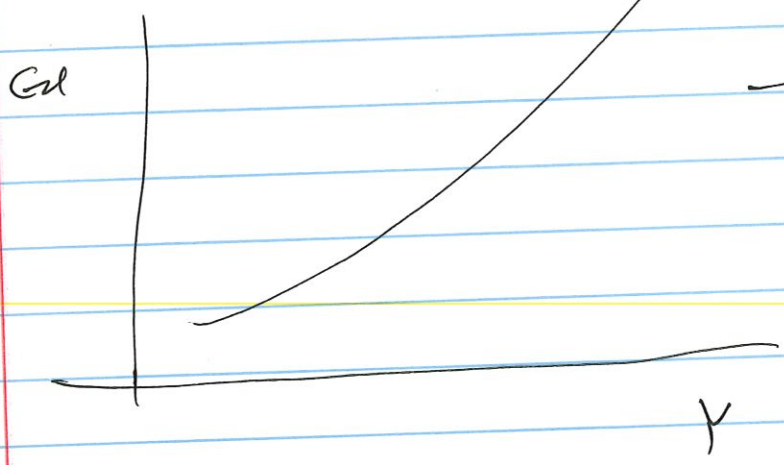
157 p. II Table 3

In the article we do 3 things

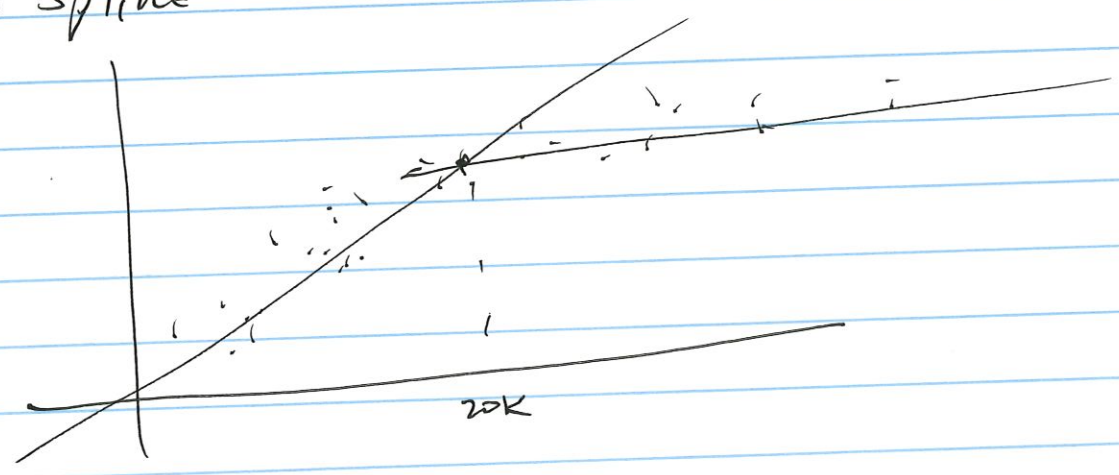
① Fit a logit curve function

$$Y = a + b_1 \log \text{income} + \text{constants} \quad \downarrow$$

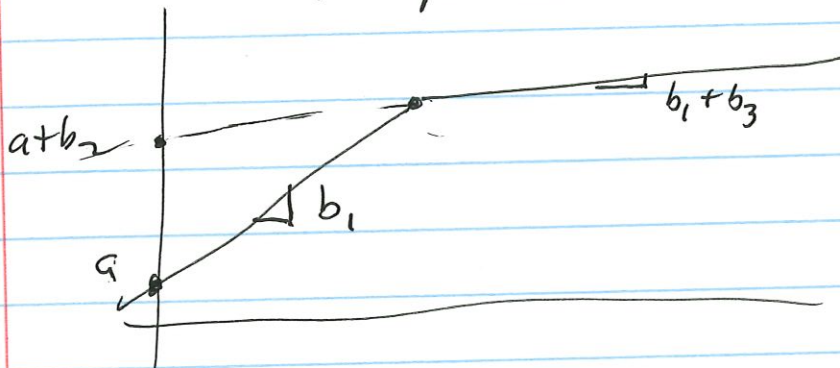
If $\ln Y = a + b_1 \text{income} + \text{constants}$



② Fit a spline



1st part II Table 3



how to do this?

$$(1) \quad Y = a + b_1 \text{ income} + b_2 (1,0) \text{ when income} > 20K \\ + b_3 (1,0) \text{ when} > 20K \cdot \text{income}$$

how to make sense out of this?

Suppose income is $< 20K$, the (1) gives you

$$Y = a + b_1 \text{ income} + 0 + 0$$

if income $> 20K$ then

$$Y = a + b_1 \text{ income} + b_2 \cdot 1 + b_3 \cdot \text{income}$$

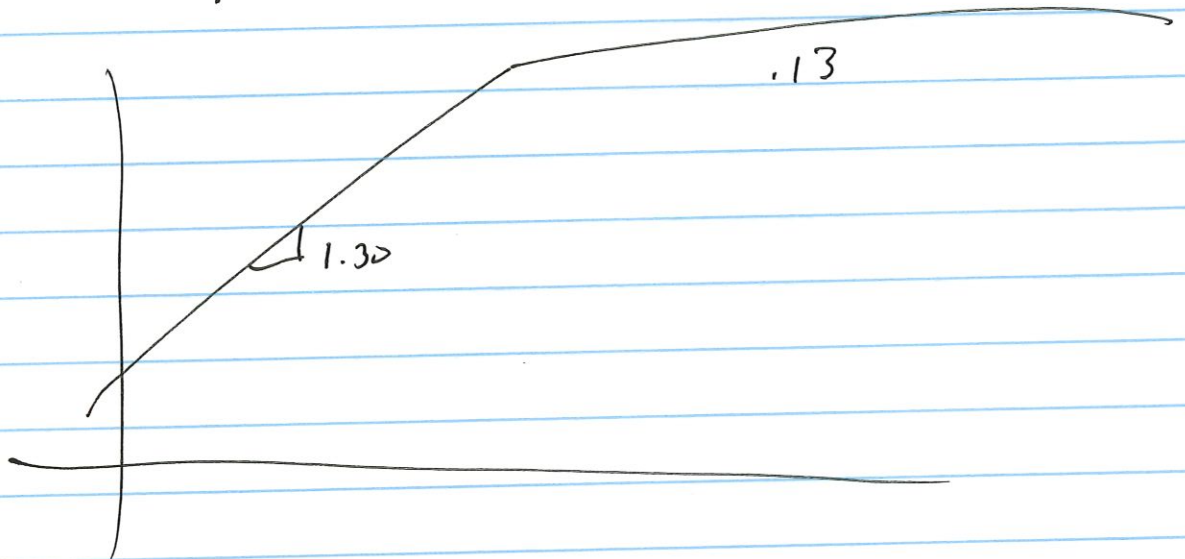
$$\Rightarrow Y = (a + b_2) + (b_1 + b_3) \text{ income}$$

1st part II Table 3

-5-

look at Table $b_1 = 1.30$! much bigger
than .14

$$b_1 + b_0 = 1.30 - 1.17 = .13$$



What about ~~dummy~~ column 4?

$$Y = a + b_1 (1,0) \text{ Inc}(15-25) + b_2 (1,0) \text{ Inc}(25-35) \\ + b_3 (1,0) \text{ Inc}(35-50) + b_4 (1,0) \text{ Inc } 50+ + \text{controls}$$

Hmm...

For someone with a 20K income:

$$Y = a + b_1 \cdot 1 + 0 + 0 + 0 + \text{controls} \\ .82$$

... 30K

$$Y = a + 0 + b_2 + 0 + 0 + \text{controls}$$

$$a + 1.41$$

$$a + 1.69$$

$$a + 2.35$$

Someone with a 10K income?

$$Y = a + 0 + 0 + 0 + 0 + \text{controls}$$

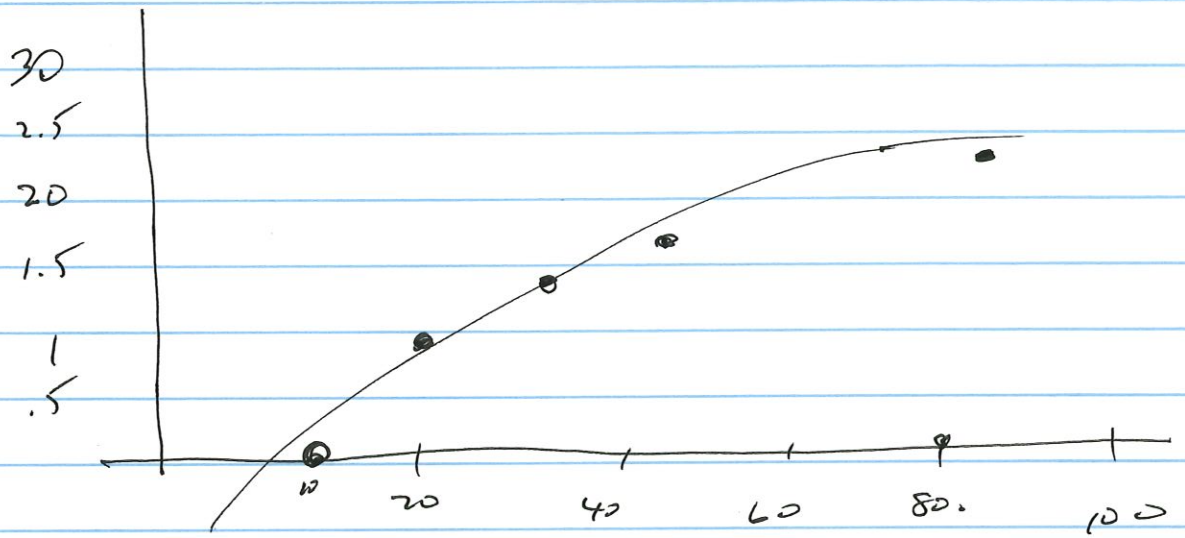
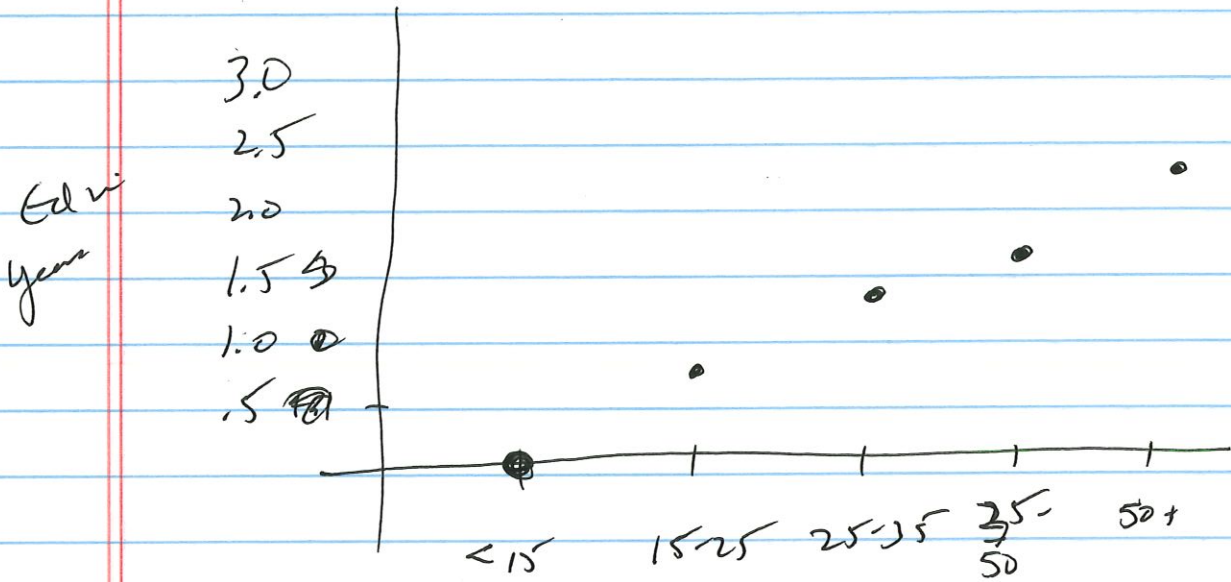
a is the ed level associated with very low income

b_1 shows difference between 20K and 10K
.82 years

b_2 shows diff between 30K and low income

$$1.41$$

1st Part II Table 3



most flexible

Other columns : (1,0) dependent variable \rightarrow logit regression
 1,0 dependent variable with censoring
 event history