

## HW-07

**Task 1.**

$$y_i = x_i' \beta + \nu_i - u_i,$$

where

$$\nu_i \sim N(0, \sigma_\nu^2),$$

$$u_i \sim N^+(0, \sigma^2(z)),$$

$$\sigma_i^2 = \sigma^2(z_i) = \exp(\delta + \gamma z_i)$$

Analytical expression for inefficiency measures:

(i)

$$\text{Let } b_1 = \delta, b_2 = \gamma$$

$$\text{Then } \ln \sigma^2 = \delta + \gamma z_i = b'z \text{ and } \sigma = \exp(1/2 * b'z)$$

Therefore

$$E(u|z) = \int_0^\infty u \frac{2}{\sqrt{2\pi}\sigma} \exp\left(-\frac{u^2}{2\sigma^2}\right) du = \sqrt{\frac{2}{\pi}} * \sigma = \sqrt{\frac{2}{\pi}} * \exp(1/2 * b'z)$$

(ii)

$$E(\exp(-u)|z) = \int_0^\infty \exp(-u) \frac{2}{\sqrt{2\pi}\sigma} \exp\left(-\frac{u^2}{2\sigma^2}\right) du = 2 \exp\left(\frac{\sigma^2(b'z)}{2}\right) (1 - \Phi(\sigma(b'z)))$$

(iii)

Marginal effects:

(a)

$$\begin{aligned} \frac{\partial}{\partial z_i} E(u_i|z_i) &= \frac{\partial}{\partial z_i} \sqrt{\frac{2}{\pi}} * \sigma(z'b) = \sqrt{\frac{2}{\pi}} \frac{\partial}{\partial z_i} \exp(1/2 * b'z) = \sqrt{\frac{2}{\pi}} \exp(1/2 * b'z) * 1/2 b_2 = \\ &= \frac{b_2}{\sqrt{2\pi}} \exp(1/2 * b'z) = \frac{\gamma \sigma(b'z)}{\sqrt{2\pi}} \end{aligned}$$

(b)

$$\begin{aligned} \frac{\partial}{\partial z_i} \exp(-E(u_i|z_i)) &= \frac{\partial}{\partial z_i} \exp\left(-\sqrt{\frac{2}{\pi}} * \sigma(z'b)\right) = \\ &= \exp\left(-\sqrt{\frac{2}{\pi}} * \sigma(z'b)\right) * \left(-\sqrt{\frac{2}{\pi}} \exp(1/2 * b'z) * 1/2 * b_2\right) = \end{aligned}$$

$$= -\frac{\gamma\sigma(b'z)}{\sqrt{2\pi}} \exp\left(-\sqrt{\frac{2}{\pi}} * \sigma(z'b)\right)$$

$\frac{\partial}{\partial z_i} E(u_i|z_i)$  and  $\frac{\partial}{\partial z_i} \exp(-E(u_i|z_i))$  has opposite signs.

(c)

$$\frac{\partial}{\partial z_i} (E(\exp(-u_i)|z_i)) = \frac{\partial}{\partial z_i} 2 \exp\left(\frac{\sigma^2(b'z)}{2}\right) (1 - \Phi(\sigma(b'z))) = 2 * \exp\left(\frac{\sigma^2(b'z)}{2}\right) * \sigma(b'z) *$$

$$\exp(1/2 * b'z) * 1/2 * b_2 * (1 - \Phi(\sigma(b'z))) + 2 * \exp\left(\frac{\sigma^2(b'z)}{2}\right) * (-\phi(\sigma(b'z))) * \exp(1/2 * b'z) *$$

$$1/2 * b_2) = \gamma\sigma(b'z) \exp\left(\frac{\sigma^2(b'z)}{2}\right) (\sigma(b'z) * (1 - \Phi(\sigma(b'z))) - \phi(\sigma(b'z)))$$

$\frac{\partial}{\partial z_i} (E(\exp(-u_i)|z_i))$  and  $\frac{\partial}{\partial z_i} \exp(-E(u_i|z_i))$  has same signs,

$$\text{because } \frac{\phi(\sigma(x))}{1 - \Phi(\sigma(x))} > \sigma(x) \Rightarrow \frac{\phi(\sigma(b'z))}{1 - \Phi(\sigma(b'z))} > \sigma(b'z) \Rightarrow$$

$$(\sigma(b'z) * (1 - \Phi(\sigma(b'z))) - \phi(\sigma(b'z))) < 0.$$

## Task 2.

First, we estimate quadratic regression:

$$lwage_i = const + exp_i + exp_i^2 + e_i$$

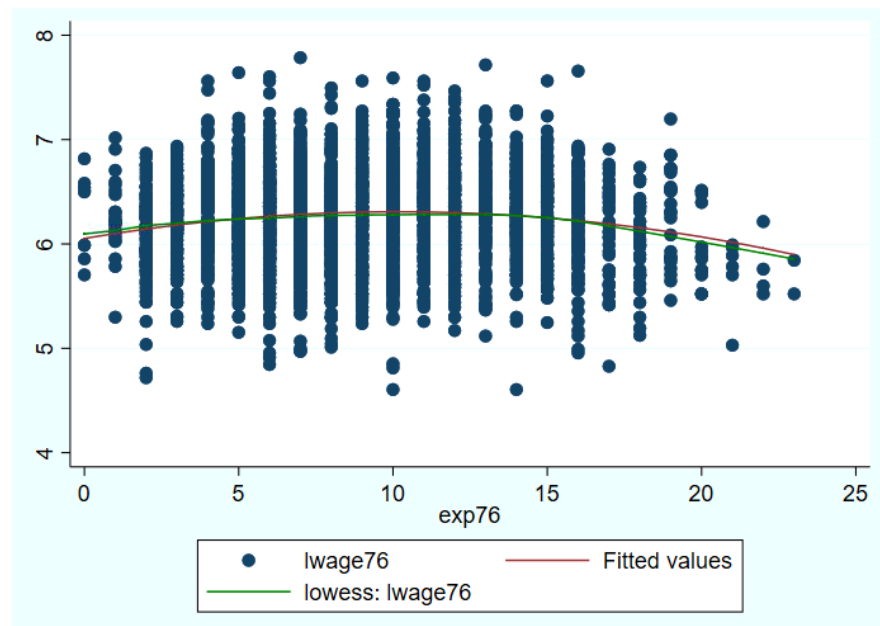
Source	SS	df	MS	Number of obs	=	3,010
Model	8.67672483	2	4.33836241	F(2, 3007)	=	22.34
Residual	583.964921	3,007	.194201836	Prob > F	=	0.0000
				R-squared	=	0.0146
				Adj R-squared	=	0.0140
Total	592.641646	3,009	.196956346	Root MSE	=	.44068

lwage76	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
exp76	.0504443	.0076365	6.61	0.000	.035471	.0654176
exp762	-.002485	.0003738	-6.65	0.000	-.0032179	-.0017522
_cons	6.052608	.0352271	171.82	0.000	5.983536	6.121679

Second, we run locally weighted scatterplot smoothing (LOWESS) with bandwidth 0.8.

Plotting both predicted values on a same plot, where red line represents polynomial fit estimated with OLS and a green line represents the smoothed curve by non-parametric method, we can see, that the predictions differ insignificantly. That means that second-order polynomial is a reasonable model, as the shape of a curve is exactly the same as a non-parametric model would suggest. However, LOWESS fit shows, that a linear fit would not be a suitable model here.



STATA code:

```
reg lwage76 exp76 exp762
predict preg
lowess lwage76 exp76, generate (pnonp)
twoway (scatter lwage76 exp76) (line preg exp76, sort)
(line pnonp exp76, sort lcolor(green))
```