

# **Advanced Econometrics: Exercise #8**

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*Professor Paulo Parente*

**Amy Qian**

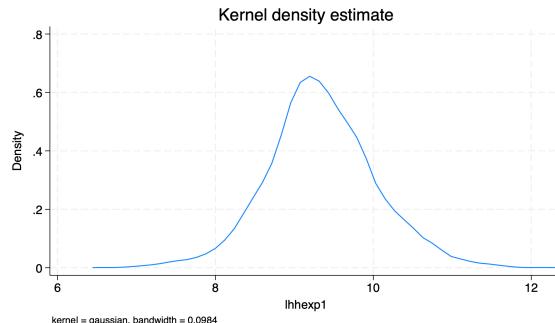
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## Problem 1

Obtain a non-parametric estimate of the density of lhhepx1. Compare the bandwidth used by STATA with the one suggested by Silverman's rule of thumb.

### STATA

```
kdensity lhhepx1, kernel(gaussian)
```



Bandwidth used by STATA: 0.0984

### Silverman's Rule of Thumb

```
sum lhhepx1, d
```

This gives us

- Standard deviation: 0.6877458
- n: 5,999
- IQR:  $9.759566 - 8.919547 = 0.840019$

To calculate Silverman's bandwidth, we use the formula:  $h = \min(\sigma, \frac{IQR}{1.34}) \frac{0.9}{n^{-1/5}}$

Plugging in our values, we get:

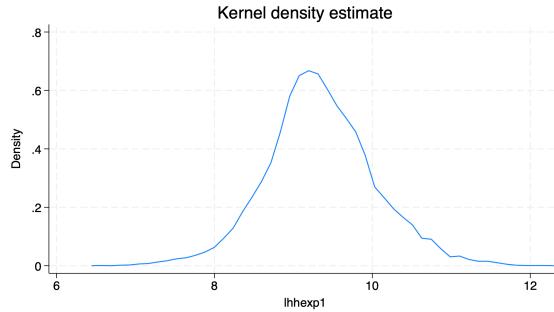
- $\text{display } ((.6877458*0.9)/5999^{(0.2)}) = .10865624$
- $\text{display } (((9.759566 - 8.919547)/1.34*0.9)/5999^{(0.2)}) = .09904009$

So the bandwidth suggested by Silverman's rule of thumb is 0.09904009, which is very close to the bandwidth used by STATA.

## Problem 2

Repeat the previous exercise using different kernels.

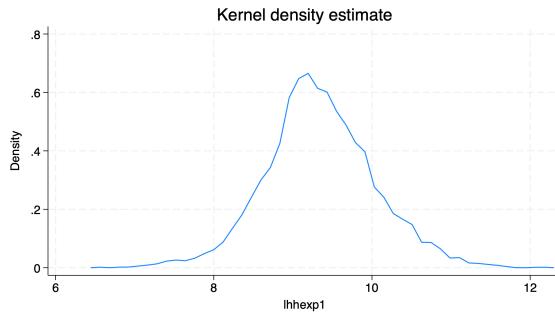
### Rectangular Kernel



`kdensity lhexp1, kernel(rec)`

This is less smooth than the Gaussian kernel.

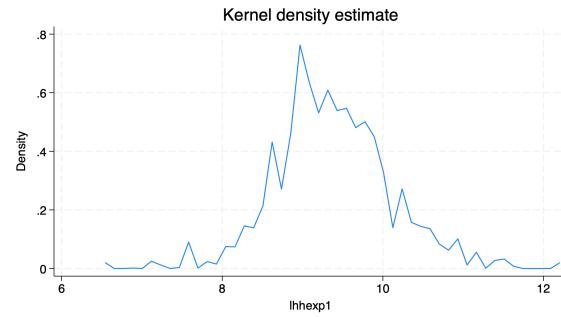
### Parzen Kernel



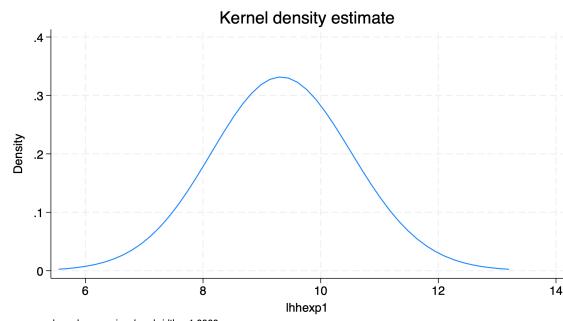
`kdensity lhexp1, kernel(parz)`

This is also less smooth than the Gaussian kernel.

### Gaussian with different bandwidth



`kdensity lhexp1, kernel(gaussian) bwidth(0.002)`



`kdensity lhexp1, kernel(gaussian) bwidth(1)`

In general, the bandwidth is more important than the kernel. We want it to be smooth enough but not too smooth that we miss important features of the data.

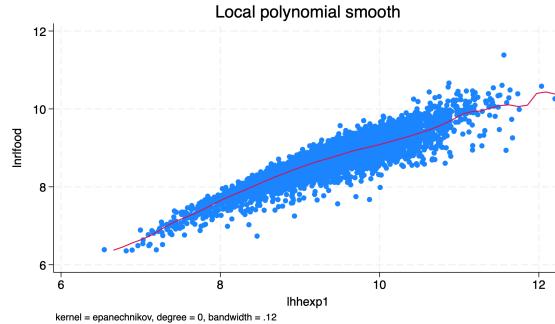
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## Problem 3

Obtain the Nadaraya-Watson estimator of the conditional expectation of lnrlfood, the log household expenditure on food, given lhhexp1. Does the result suggest that this conditional expectation is linear?

### Solution

We want to estimate  $E[y_i|x_i] = m(x_i)$  where  $y_i$  is lnrlfood.



`lpoly lnrlfood lhhexp1`

Blue dots are the data points, the red line is the estimated conditional expectation (the function). Could be linear looking at the graph, but not exactly linear.

## Problem 4

Using OLS, estimate a regression of lnrlfood on lhhexp1, age and its square. Is the effect of age linear?

### Solution

First generate the new variable age squared.

`gen age2 = age^2`

Then run the OLS regression using robust standard errors to control for heteroskedasticity.

`reg lnrlfood lhhexp1 age age2, r`

We want to test the null hypothesis that the effect of age is linear, which is equivalent to testing if the coefficient of age2 is zero.

$$H_0 : \beta_{age2} = 0 \text{ vs. } H_a : \beta_{age2} \neq 0$$

From the regression output, we see that the p-value is 0.000, which is less than 0.05, so we reject  $H_0$  in favor of  $H_1$  at 5% confidence level.

## Problem 5

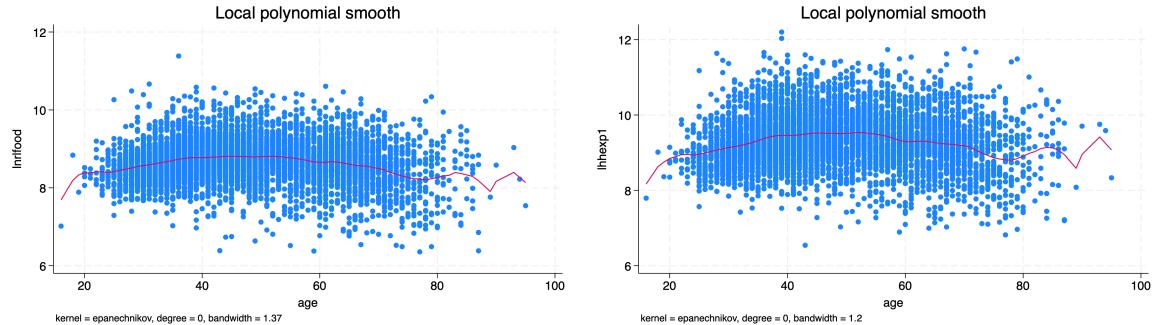
Use the partially linear estimator to estimate in the following model:  $\lnrlfood_i = \beta lhhexp1_i + g(\text{age}_i) + u_i$

### Solution

$G(\cdot)$  is now an unknown function.

Get the estimates

- $E[y_i|\text{age}_i] \Rightarrow \text{foodage}$ . `lpoly lnrlfood age, generate(foodage) at (age)`
- $E[x_i|\text{age}_i] \Rightarrow \text{expage}$ . `lpoly lhhexp1 age, generate(expage) at (age)`



Now we compute the differences.

- gen y= lnrlfood- foodage
- gen x= lhhepl- expage

Finally, we run the regression of y on x, not assuming anything on the distribution of the data.

```

Source |      SS          df        MS   Number of obs = 
> 5,999 |               F(1, 5998) =    235
> 64.09 |               Prob > F =     0
> .0000 |      Model | 1274.81681      1 1274.81681 R-squared =    0
> .0000 |      Residual | 324.491644  5,998  .054099974 Adj R-squared =    0
> .7971 |               Root MSE =   .
> .7971 |               Total | 1599.30845  5,999  .266595841
> 23259 |               .
> _____ |               y | Coefficient  Std. err.      t    P>|t| [95% conf. inte
> rval]
> _____ |               x |  .695743   .0045324  153.51  0.000    .686858   .70
> 46281 |               .
reg y x, noconstant

```

## Problem 6

Compare the results.

Linear regression		Number of obs	=	5,999
		F(3, 5995)	=	5484.44
		Prob > F	=	0.0000
		R-squared	=	0.8112
		Root MSE	=	.2333
<hr/>				
lnrlfood				
Robust				
lnrlfood	Coefficient	std. err.	t	P> t  [95% conf. interval]
lhhepl	.6942858	.005711	121.57	0.000 .6830901 .7054815
age	.005095	.0016213	3.14	0.002 .0019168 .0082733
age2	-.0000716	.0000161	-4.45	0.000 -.0001031 -.00004
_cons	2.1277	.0561216	37.91	0.000 2.017681 2.237718

Actually quite close.