

Ch. 8 Heteroskedasticity

Consequences:

$$\text{var}(u | x_1, \dots, x_k) \neq \sigma^2$$

but depends on x_j

$$\text{price} = \beta_0 + \beta_1 \text{lotsize} + \beta_2 \text{sq.ft.} + \beta_3 \text{bdrms.} + u$$

(e.g. quality)

$$\text{var}(u | \text{lotsize}, \text{sq.ft.}, \text{bdrms})$$

dep. on sq.ft.

under assⁿs for unbiasedness, $\hat{\beta}_j$ unbiased for β_j .

usual std. errors & test statistics $\rightarrow$ no longer valid

Heteroskedasticity - robust Inference

Heteroskedasticity of unknown form.

" robust std. errors available
in case of large samples.

Simple regression:

$$se(\hat{\beta}_1) = \frac{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \hat{u}_i^2}}{SST_x}$$

↓

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

Multiple regression

$$\text{score} = \beta_0 + \beta_1 \text{class-size} \\ + \beta_2 \text{PC stud. exp.} \\ + \beta_3 \text{\% minority} \\ + u$$

$$se(\hat{\beta}_j) = \sqrt{\frac{\sum_{i=1}^n \hat{r}_{ij}^2}{SST_j (1 - R_j^2)}}$$

e.g. $\hat{\beta}_1$

$$\sum_{i=1}^n (x_{ij} - \bar{x}_j)^2$$

R_j^2 from reg. of
class-size
 x_j on all other
 x 's
PC stud. exp. &
\% minority

$\hat{r}_{ij}$: residual from reg. of
class-size
 x_j on all other x 's
for obs. i
PC stud. exp. &
\% minority

Testing

Model (with assⁿs up to those reqd. for unbiasedness of $\hat{\beta}$)

$$y = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k + u$$

Null hyp. of homosk.

$$H_0 : \text{Var}(u | x_1, \dots, x_k) = \sigma^2$$

$$\text{or, } H_0 : E(u^2 | x_1, \dots, x_k) = \sigma^2$$

Suppose

$$u^2 = \delta_0 + \delta_1 x_1 + \dots + \delta_k x_k + v$$

$$H_0 : \delta_1 = 0$$

$$\delta_k = 0$$

$$\text{can estimate: } \hat{u}^2 = \delta_0 + \delta_1 x_1 + \dots + \delta_k x_k + v$$

Joint test : H_0

Can also include quadratics, ^{x_1^2, x_2^2} interactions, & ^{$x_1 \cdot x_2$} fitted values.
Works only in case of ^{$\hat{y}$} large samples.

```
. reg price lotsize sqrft bdrms
```

Source	SS	df	MS	Number of obs	=	88
				F(3, 84)	=	57.46
Model	617130.701	3	205710.234	Prob > F	=	0.0000
Residual	300723.805	84	3580.0453	R-squared	=	0.6724
				Adj R-squared	=	0.6607
Total	917854.506	87	10550.0518	Root MSE	=	59.833

price	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
lotsize	.0020677	.0006421	3.22	0.002	.0007908	.0033446
sqrft	.1227782	.0132374	9.28	0.000	.0964541	.1491022
bdrms	13.85252	9.010145	1.54	0.128	-4.065141	31.77018
_cons	-21.77031	29.47504	-0.74	0.462	-80.38466	36.84405

```
. predict uhat, res
```

```
. g uhatsq = uhat^2
```

```
. reg uhatsq lotsize sqrft bdrms
```

Source	SS	df	MS	Number of obs	=	88
				F(3, 84)	=	5.34
Model	701213780	3	233737927	Prob > F	=	0.0020
Residual	3.6775e+09	84	43780003.5	R-squared	=	0.1601
				Adj R-squared	=	0.1301
Total	4.3787e+09	87	50330276.7	Root MSE	=	6616.6

uhatsq	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
lotsize	.2015209	.0710091	2.84	0.006	.0603116	.3427302
sqrft	1.691037	1.46385	1.16	0.251	-1.219989	4.602063
bdrms	1041.76	996.381	1.05	0.299	-939.6526	3023.173
_cons	-5522.795	3259.478	-1.69	0.094	-12004.62	959.0348

```
. test lotsize sqrft bdrms
```

```
( 1) lotsize = 0
( 2) sqrft = 0
( 3) bdrms = 0

F( 3, 84) = 5.34
Prob > F = 0.0020
```

Weighted Least Squares

$$\text{Model : } y = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k + u$$

$$\text{say } \text{var}(u | x_1, \dots, x_k) = \sigma^2 h$$

Transformed model:

$$\frac{y}{\sqrt{h}} = \beta_0 \cdot \frac{1}{\sqrt{h}} + \beta_1 \frac{x_1}{\sqrt{h}} + \dots + \beta_k \frac{x_k}{\sqrt{h}} + \frac{u}{\sqrt{h}}$$

$$\text{var}\left(\frac{u}{\sqrt{h}} \mid x_1, \dots, x_k\right) = \frac{\sigma^2 h}{h} = \sigma^2$$

Generalized / Weighted Least Squares (GLS/WLS).

Feasible / Estimated GLS $\because$ h is estimated