

# Lecture notes to Stock and Watson chapter 5

## Inference in linear regression with a single regressor

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## 4.3

## 4.4 d (an extra question)

Show that  $\beta = 1$  for all assets in a CAPM model with equal weights,  $R_m = \frac{1}{N} \sum_{i=1}^N R_i$ , and equally variable and symmetrically correlated excess earnings  $X_i = R_i - R_f$  :  $var(X_i) = \sigma^2$ ,  $corr(X_i, X_j) = \rho$   $i \neq j$

# Inference in regression

Model:  $E[Y|X = x] = \beta_0 + \beta_1 x$

Typical questions referring to the model, and empirical answers:

- What is the marginal expected effect on  $Y$  of a change in  $X$ ?  
Answer:  $\beta_1$ 
  - Estimate or confidence interval for  $\beta_1$
- What is the expected effect of increasing  $X$  from  $x_1$  to  $x_1 + \Delta x$ ? Answer:  $\beta_1 \cdot \Delta x$ 
  - Estimate or confidence interval for  $\theta = \beta_1 \cdot \Delta x$
- Is  $Y$  affected by  $X$  in the mean (is there correlation)?
  - Test  $H_0 : \beta_1 = 0$  vs.  $H_1 : \beta_1 \neq 0$
- What is the mean response of  $Y$  when  $X = x$ ?
  - Estimate or confidence interval for  $\theta = \beta_0 + \beta_1 x$

- Linear parameters  $\theta = a\beta_0 + b\beta_1$  are estimated by  $\hat{\theta} = a\hat{\beta}_0 + b\hat{\beta}_1$ ,  $\hat{\beta}_0$  and  $\hat{\beta}_1$  are the OLS estimators,  $a$  and  $b$  are numbers.
- $var\hat{\theta} = a^2 var(\hat{\beta}_0) + b^2 var(\hat{\beta}_1) + 2ab \cdot cov(\hat{\beta}_0, \hat{\beta}_1)$
- Estimates of  $\sqrt{var(\hat{\beta}_0)}$  and  $\sqrt{var(\hat{\beta}_1)}$ ,  $SE(\hat{\beta}_0)$  and  $SE(\hat{\beta}_1)$  are given in computer output, along with an estimate of  $\rho = cor(\hat{\beta}_0, \hat{\beta}_1)$ 
  - compute  $SE(\hat{\theta})$  from these numbers.
  - STATA provides inference on linear parameters

- For large samples, and when the basic assumptions are satisfied (Key Concepts 4.3 in SW: Linear regression, random sample, large outliers are unlikely):
  - Two-sided confidence interval of degree  $1 - 2\alpha$  for  $\theta$  is

$$\hat{\theta} \pm z_{1-\alpha} SE(\hat{\theta}) = \left( \hat{\theta} - z_{1-\alpha} SE(\hat{\theta}), \hat{\theta} + z_{1-\alpha} SE(\hat{\theta}) \right)$$

where  $z_{1-\alpha} = \Phi^{-1}(1 - \alpha)$  is the  $1 - \alpha$  quantile of the standard normal distribution  $z_{.95} = 1.64$

- Reject  $H_0 : \theta = \theta_0$  versus the two-sided alternative, at level  $\alpha$ , if

$$\left| \frac{\hat{\theta} - \theta_0}{SE(\hat{\theta})} \right| > z_{1-\alpha} \Leftrightarrow \theta_0 \text{ is not in } \hat{\theta} \pm z_{1-\alpha} SE(\hat{\theta})$$

- the p-value is  $p = 2 \cdot \min \left\{ \Phi \left( \frac{\hat{\theta} - \theta_0}{SE(\hat{\theta})} \right), 1 - \Phi \left( \frac{\hat{\theta} - \theta_0}{SE(\hat{\theta})} \right) \right\}$

One-sided confidence interval of degree  $1 - \alpha$  for  $\theta$

is  $(\hat{\theta} - z_{1-\alpha} SE(\hat{\theta}), \infty)$

- the p-value for the one-sided test problem

$H_0 : \theta \leq \theta_0$  versus  $H_1 : \theta > \theta_0$  is

$$p = 1 - \Phi \left( \frac{\hat{\theta} - \theta_0}{SE(\hat{\theta})} \right).$$

- Reject  $H_0$  at level  $\alpha$  if

$$p < \alpha$$

$$\Leftrightarrow \frac{\hat{\theta} - \theta_0}{SE(\hat{\theta})} > z_{1-\alpha}$$

$$\Leftrightarrow \theta_0 \text{ is not in } (\hat{\theta} - z_{1-\alpha} SE(\hat{\theta}), \infty)$$

# Methods for estimating standard errors for OLS regression parameters

Calculate empirical residuals

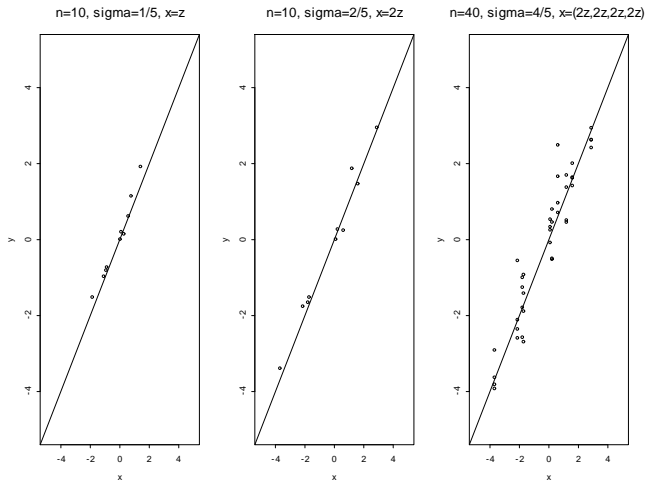
$$\hat{u}_i = Y_i - \hat{Y}_i = Y_i - \left( \hat{\beta}_0 + \hat{\beta}_1 X_i \right) \quad i = 1, \dots, n$$

- Heteroskedasticity-robust standard error:

$$SE \left( \hat{\beta}_1 \right) = \frac{\frac{1}{n-2} \sum (X_i - \bar{X})^2 \hat{u}_i^2}{n \left[ \frac{1}{n} \sum (X_i - \bar{X})^2 \right]^2}$$

- The robust standard errors are obtained by STATA command `regress . . . , robust`
- $SE \left( \hat{\beta}_1 \right)$  is small when the empirical residuals  $\hat{u}_i$  are small and/or when  $\sum (X_i - \bar{X})^2$  is large!

$\text{var}(\hat{\beta}_1)$  is identical in the three graphs





# Properties of OLS

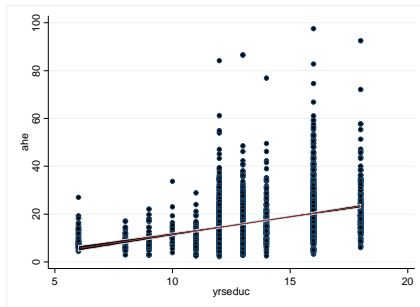
- Linearity,  $E[Y_i|X_i] = \hat{\beta}_0 + \hat{\beta}_1 X_i \Rightarrow$  linear parameters  $\hat{\theta} = a\hat{\beta}_0 + b\hat{\beta}_1$  are unbiased, i.e.  $E\hat{\theta} = \theta$
- Linearity + homoskedasticity + independent normally distributed residuals,  $u_1, \dots, u_n$  are iid  $N(0, \sigma^2) \Rightarrow \frac{\hat{\theta} - \theta}{SE(\hat{\theta})}$  is t-distributed with  $n - 2$  degrees of freedom
- Linearity + random sample + large outliers unlikely  $\Rightarrow \frac{\hat{\theta} - \theta}{SE(\hat{\theta})}$  is approximately standard normally distributed
- approximation better the larger  $n$  is and/or the closer to normality the residuals are distributed.

```
use D:\Schweder\undervisning\Master\ECON4135_2009\ch5_cps_box.dta
reg ahe yrseduc , robust
```

```
Linear regression      Number of obs =   2950
                      F( 1, 2948) = 415.76
                      Prob > F   = 0.0000
                      R-squared   = 0.1304
                      Root MSE  = 8.7694
```

ahe	Robust Coef.	Std. Err.	t	P>t	[95% Conf. Interval]
yrseduc	1.466925	.0719423	20.39	0.000	1.325863 1.607988
_cons	-3.134371	.9258849	-3.39	0.001	-4.949818 -1.318925

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twoway (lfitci ahe yrseduc) (scatter ahe yrseduc)
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# Problems to be done in class

- SW: 5.4
- SW: 5.8
- SW: 5.15