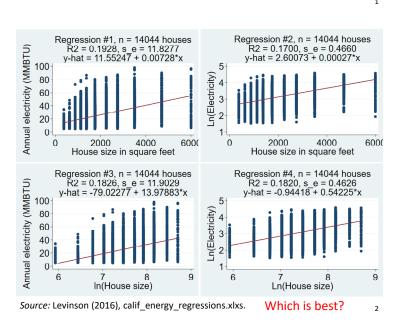
# Statistical Inference with Regression

Lecture 19

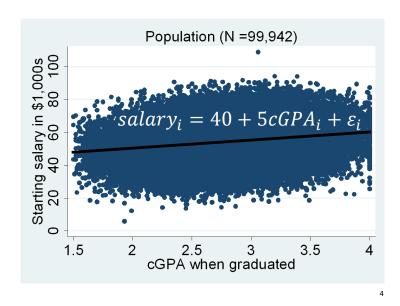
Reading: Sections 18.3 – 18.6, 19.2



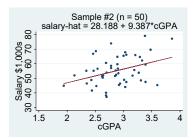
## Inference with Regression

- Linear regression model:  $y_i = \alpha + \beta x_i + \varepsilon_i$ 
  - Which parameters are we interested in?
    - E.g. How much does earning good marks in university affect your starting salary?
- OLS gives a and b:  $\hat{y}_i = a + bx_i$ 
  - Recall:  $b=\frac{s_{\chi y}}{s_{\chi}^2}=r\frac{s_y}{s_{\chi}}$  and  $a=\bar{y}-b\bar{x}$ 
    - Are these sample statistics or parameters?
    - What do we need to know about a and b so that we may use them for inference?

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Sample #1 (n = 50) salary-hat = 40.475 + 4.335\*cGPA



Sample #3 (n = 50) salary-hat = 44.148 + 3.723\*cGPA

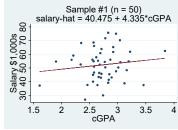
How many samples will you actually have?

What important concept do these graphs illustrate?

# Standard Error of OLS Slope

- SE(b<sub>1</sub>) reflects size of sampling error and depends on:
  - 1) Sample size (n)
  - 2) Amount of scattering about line  $(s_e)$
  - 3) How much x-variable varies in the data  $(s_x)$

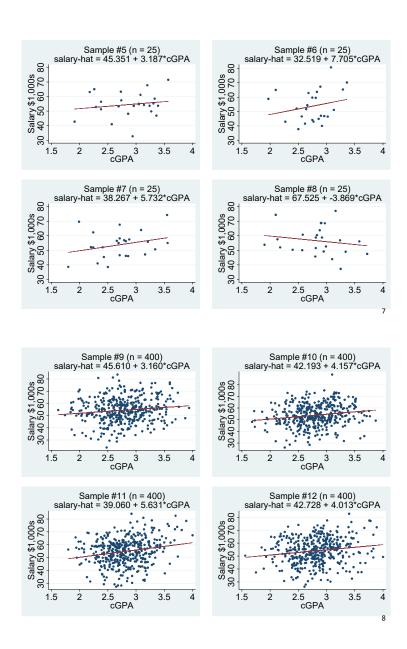
• 
$$SE(b_1)=s_{b_1}=rac{s_e}{s_\chi\sqrt{n-1}}$$
 Recall:  $s_e=\sqrt{rac{\sum_{i=1}^ne_i^2}{n-2}}$ 



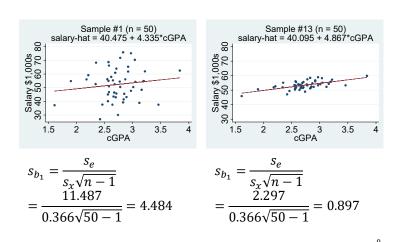
Often s.e. in parentheses below the point estimate:

Salary-hat = 40.475 + 4.335\*cGPA (4.484)

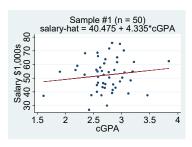
6

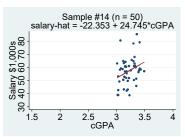


# Main Difference? Effect on s.e.?



### Main Difference? Effect on s.e.?





$$s_{b_1} = \frac{s_e}{s_x \sqrt{n-1}}$$

$$= \frac{11.487}{0.366\sqrt{50-1}} = 4.484$$

$$s_{b_1} = \frac{s_e}{s_x \sqrt{n-1}}$$

$$= \frac{9.842}{0.127\sqrt{50-1}} = 11.028$$

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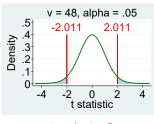
## Inference about $\beta$ : $y_i = \alpha + \beta x_i + \varepsilon_i$

- $H_0$ :  $\beta = \beta_0$ ;  $H_1$ :  $\beta \neq \beta_0$ - Test statistic:  $t = \frac{b - \beta_0}{s_b}$ - Student t, v = n - 2
- E.g. Is slope statistically significant in Reg. #1?
- "Statistical significance"
- Salary-hat = 40.475 + 4.335 \* cGPA (4.484)
- If  $H_0$ :  $\beta = 0$ ;  $H_1$ :  $\beta \neq 0$ then  $t = \frac{b}{s_b}$
- $-H_0: \beta = 0$ -  $H_1: \beta \neq 0$
- Roughly, need  $t \ge 2$  or  $t \le -2$  at  $\alpha = 0.05$
- $-t = \frac{b-\beta_0}{s_b} = \frac{4.335-0}{4.484} =$
- Conclusion?

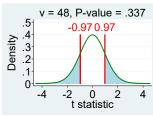
Note: Section 18.5 "A Hypothesis Test for Correlation" is redundant: same result as a test of the slope coefficient. However, we'll see the t test statistic alternate formula (on p. 617) later when we talk about the F test.

### t = 0.97, Not Statistically Significant

Rejection Region Approach



P-value Approach



Conclusion?

Conclusion?

#### Recall Frozen Pizza in Denver

Statistically significant?

$$-H_0: \beta_1 = 0$$
  
$$-H_1: \beta_1 \neq 0$$

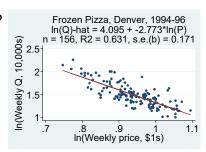
• 
$$t = \frac{-2.773 - 0}{0.171} = -16.2$$

- Is "slope" coefficient less than -1?

$$- H_0: \beta_1 = -1$$

$$- H_1: \beta_1 < -1$$

• 
$$t = \frac{-2.773 - -1}{0.171} = -10.$$



Does this analysis imply that the demand for frozen pizza in Denver in the mid-1990s was •  $t = \frac{-2.773 - -1}{0.171} = -10.4$  elastic? Why or why not?

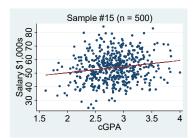
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## Confidence Interval (CI) Estimate

• CI estimator of  $\beta$ :

$$b \pm t_{\alpha/2} s_b$$

- Confidence level:  $1-\alpha$
- Degrees of freedom:  $\nu = n - 2$
- Interpretation?



Sal-hat = 41.376 + 4.479\*cGPA (1.169)

95% CI of slope: LCL = 2.182 and UCL = 6.776

## Reading STATA Output

#### . regress salary cGPA;

Source		df	MS	Number of obs = $F(1, 498) =$	500
Model   Residual	1523.07551 51656.349	1 <b>498</b>	1523.07551 103.727608	Prob > F = R-squared =	0.0001 0.0286
Total	53179.4245			Adj R-squared = Root MSE =	

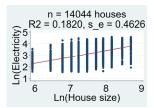
salary					[95% Conf.	_
cGPA	4.479379 41.37579	1.168972	3.83	0.000	2.182653 34.84368	6.776105

Recall: 
$$s_y^2 = \frac{\sum_{i=1}^n (y_i - \bar{y})^2}{n-1} = \frac{SST}{n-1}$$

Recall: 
$$s_e = \sqrt{\frac{\sum_{l=1}^{n}(e_l-0)^2}{n-2}} = \sqrt{\frac{SSE}{n-2}}$$
 What does 10.185 mean? Units?

#### Recap w/ CA Elec. OLS results

ln(elec)-hat = -0.9442 + 0.5423\*ln(size) (0.0097)



. regress ln\_elec\_mmbtu ln\_sq\_feet;

Source	ss	df	MS	Number of obs	= =	,
Model   Residual	668.541968 3005.32471	1 14,042	668.541968 .214023979	Prob > F R-squared	=	0.0000 0.1820
Total	3673.86668		.261615515	)	=	
ln_elec_mm~u	Coef.	Std. Err.		P> t  [95% C	onf.	Interval]
ln_sq_feet   _cons	.5422451 944182	.009702 .0724338	55.89	0.000 .52322 0.000 -1.0861		.5612624 802202

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#### World Happiness Reports (2012, 2013)

mean\_happy\_10\_12: Mean reply in a country to "Please imagine a ladder with steps numbered from zero at the bottom to 10 at the top. The top of the ladder represents the best possible life for you and the bottom of the ladder represents the worst possible life for you. On which step of the ladder would you say you personally feel you stand at this time?" asked in 2010, 2011, and 2012

Note: "We average the three most recent years (2010-12)." p. 9 (2013)

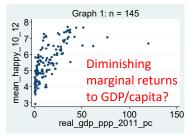
real\_gdp\_ppp\_2011\_pc: Real GDP per capita at current PPP (purchasing power parity) in 2011 \$1,000s US

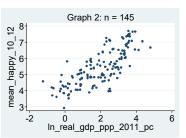
In\_real\_gdp\_ppp\_2011\_pc, Natural logarithm of real\_gdp\_ppp\_2011\_pc

Variable		mean	median	s.d.	min.	max.	
mean_happy_10_12	145	5.440	5.345	1.119	2.936	7.693	
real_gdp_ppp_2011_pc	145	14.169	8.360	16.642	0.288	120.172	
In_real_gdp_ppp_2011_pc	145	1.953	2.123	1.312	-1.245	4.789	

http://worldhappiness.report/download/; "One Question" Test #1, Nov. 2013 17

## Natural Log: Straighten Scatter Plot





Is there an issue with outliers?

Does Assumption 1 (linearity) hold for Graph 2?

Does Assumption 2 (homoscedasticity) hold for Graph 2?

Does Assumption 6 (exogeneity) ( $COV(x_i, \varepsilon_i) = 0$ ) hold?

Remember we studied this case with more recent data in Week 6.

#### What do the coefficients mean?

. regress mean\_happy\_10\_12 ln\_real\_gdp\_ppp\_2011\_pc;

	ss +				Number of obs		
Model Residual	107.855278	1 143	107.855278 .505602162		Prob > F  R-squared  Adj R-squared	= =	0.0000 <b>0.5987</b>
Total	180.156387	144	1.25108602		Root MSE	=	.71106
mean_hap_~12		Std. I	Err. t	P> t	[95% Conf.		
ln_real_g~pc _cons	. 6597638	.0451	723 14.61	0.000	.5704721 3.942095		7490556 . 361773

In 2010/12, countries with real GDP per capita that is 10% higher have mean happiness (on a 0-10 scale) that is approximately units higher on average.

In 2010/12, countries with real GDP per capita of \$1,000 have mean happiness (on a 0-10 scale) that is on average \_\_\_\_

#### **Point Prediction**

- Point prediction: Use estimated model to predict y (y-hat) for a given value of x
  - Ex: Salary-hat = 41.376 + 4.479\*cGPA (3.325) (1.169)
  - If cGPA is 3.2 then salary-hat is 55.709
  - How to interpret 55.709?
- Even with huge n, we cannot precisely predict y given x because  $y_i = \alpha + \beta x_i + \varepsilon_i$

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#### Prediction Interval vs. Confidence Interval

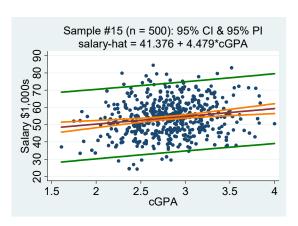
• Prediction Interval: [individual] contains y for

a given 
$$x_g$$
 with confidence  $1-\alpha$  
$$\hat{y}_{x_g} \pm t_{\alpha/2} s_e \sqrt{1 + \frac{1}{n} + \frac{\left(x_g - \bar{X}\right)^2}{(n-1)s_x^2}} \quad \text{with } \nu = n-2$$

• Confidence Interval: [mean] contains E[y] for a given  $x_g$  with confidence 1-lpha

$$\hat{y}_{x_g} \pm t_{\alpha/2} s_e \sqrt{\frac{1}{n} + \frac{\left(x_g - \bar{X}\right)^2}{(n-1)s_x^2}} \qquad \text{with } \nu = n-2$$

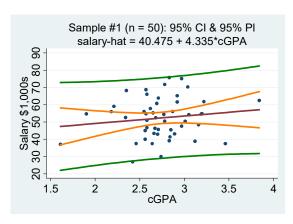
For alternate (mathematically identical) versions, see textbook 21



For a cGPA of 3.2 ( $x_g = 3.2$ ) the 95% PI is (36.0, 76.2)

For a cGPA of 3.2 ( $x_g = 3.2$ ) the 95% CI is (54.7, 57.5)

Which should contain  $\sim$ 95% of dots in scatter diagram above?



For a cGPA of 3.2 ( $x_g = 3.2$ ) the 95% PI is (31.6, 81.4)

For a cGPA of 3.2 ( $x_g = 3.2$ ) the 95% CI is (47.2, 65.8)

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## Recap: Three Kinds of Intervals

• 2010/12 happiness OLS results:

Happiness-hat = 4.151934 + 0.6597638\*In(GDP per capita)(0.0451723)

• Three different intervals:

$$b \pm t_{\alpha/2} s_b$$

$$\hat{y}_{x_g} \pm t_{\alpha/2} s_e \sqrt{1 + \frac{1}{n} + \frac{(x_g - \bar{X})^2}{(n-1)s_x^2}}$$

$$\hat{y}_{x_g} \pm t_{\alpha/2} s_e \sqrt{\frac{1}{n} + \frac{\left(x_g - \bar{X}\right)^2}{(n-1)s_x^2}}$$