



**Georgia Institute  
of Technology**

# Statistics for Economics

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# Today

- Review T-test
- Review confidence intervals
- Introduce binomial t-test

What are the three things that economists are looking for in applied analysis?

desired alpha level	two-sided test	one-sided test
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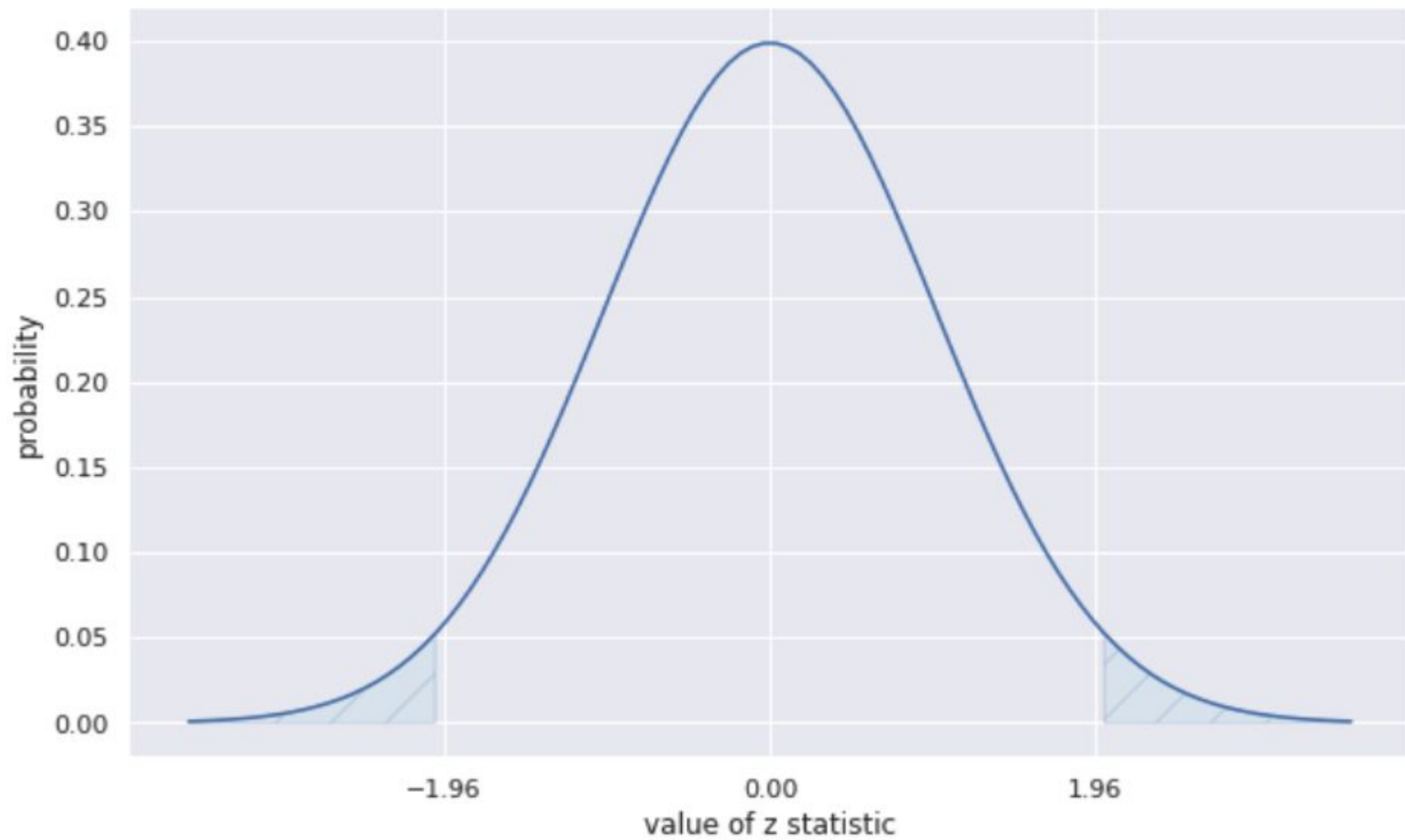
0.100	1.644854	1.281552
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0.050	1.959964	1.644854
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0.010	2.575829	2.326348
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0.001	3.290527	3.090232
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Two Sided Test



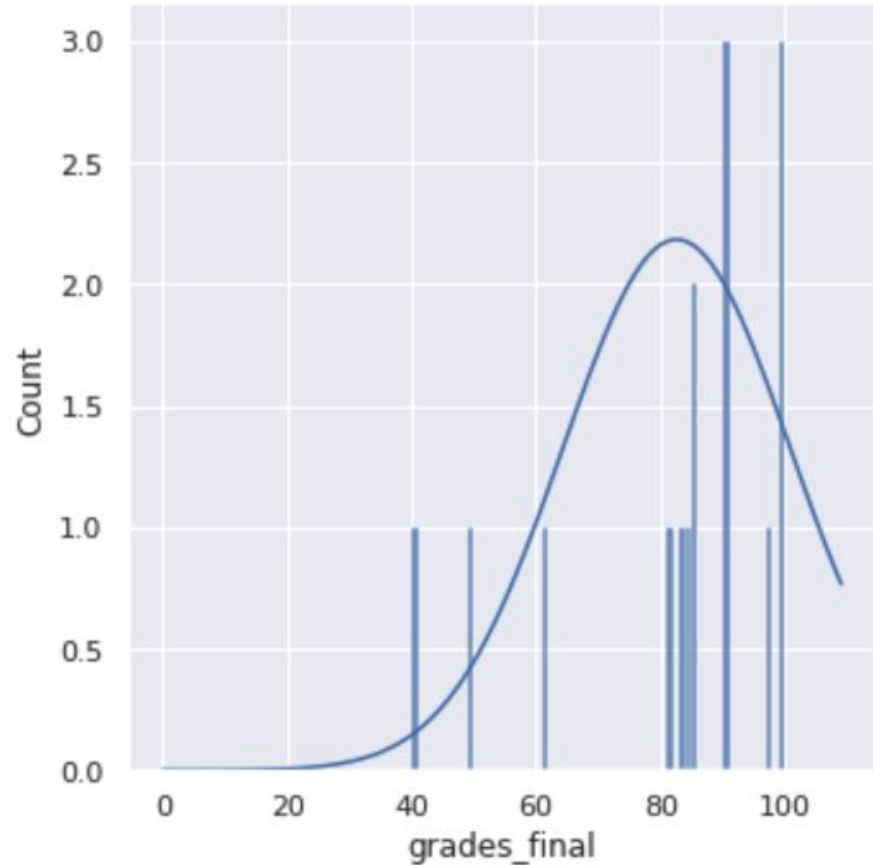
(simple) T Tests = z-test

$$Z_{\bar{X}} = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{N}}$$

#1. We can make a normal with the mean and the standard deviation

$$\bar{x} = \frac{\sum x}{15} = \frac{1238}{15} = 82$$

$$\sigma = \sqrt{\frac{(\sum x - \bar{x})^2}{N - 1}} = \sqrt{\frac{4672}{14}} = 18.2$$

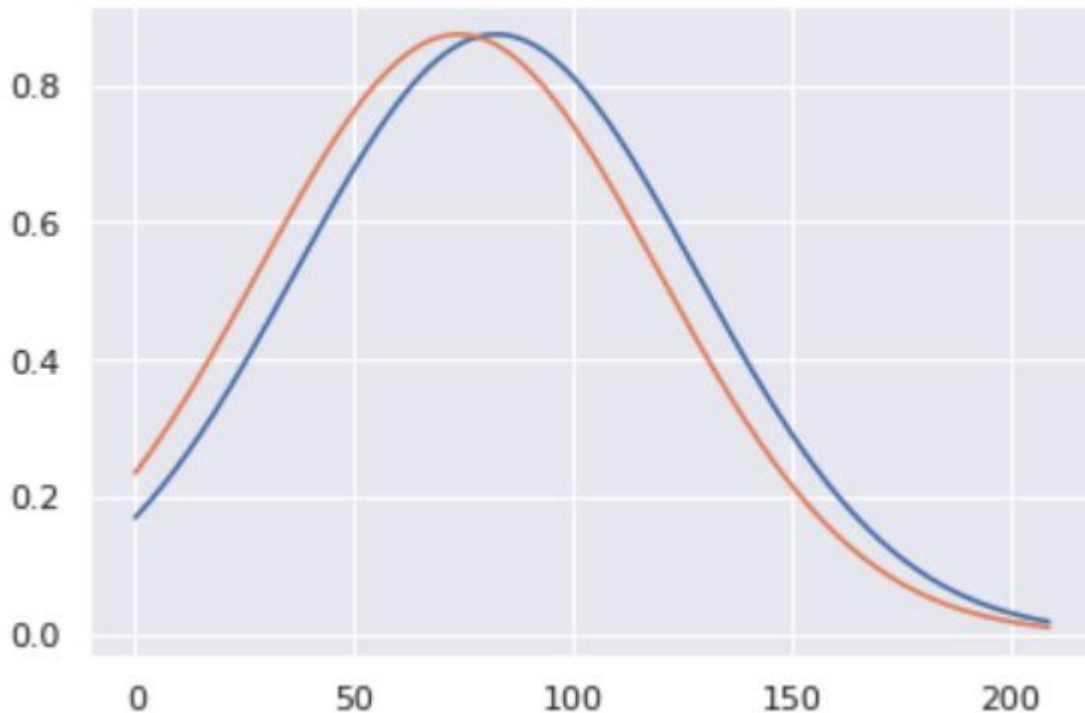


# Can use the abstraction to compare to other abstractions:

Compare 2 classes:

Class1: 82 (18)

Class2: 72 (18)





## T-test

$$t = \frac{\bar{X}_1 - \bar{X}_2}{SE}$$

where

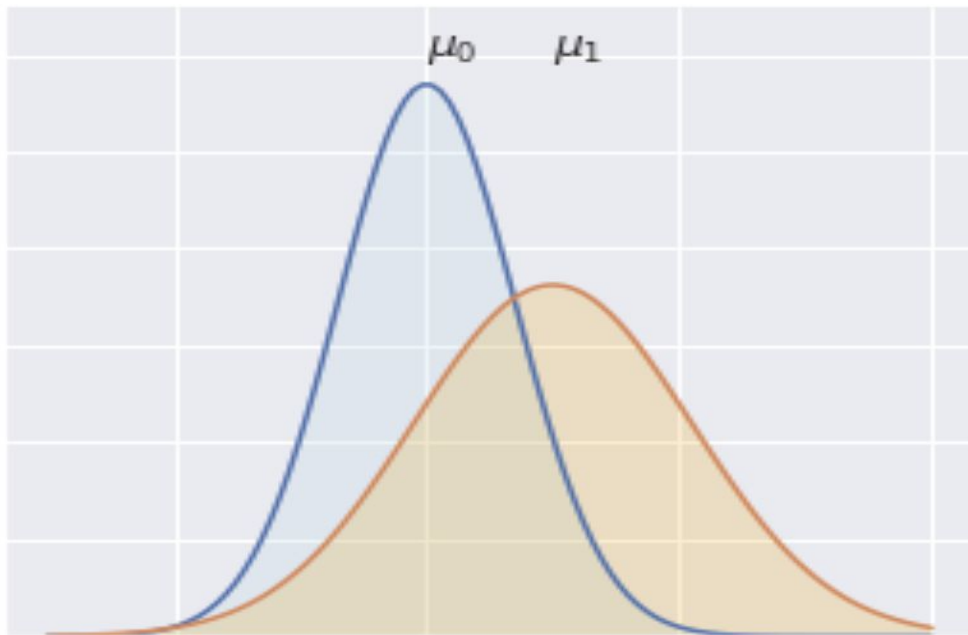
$$SE = \sqrt{\frac{\sigma_1^2}{N_1} + \frac{\sigma_2^2}{N_2}}$$

# Difference between two groups with different sample sizes and variance

$$H_0 : \mu_1 = \mu_2$$

$$H_1 : \mu_1 \neq \mu_2$$

- Classroom 1
  - Avg: 82
  - Stdev: 18
  - Size: 15
- Classroom 2
  - Avg: 72
  - Stdev: 22
  - Size: 30



- Classroom 1
  - Avg: 82
  - Stdev: 18
  - Size: 15
- Classroom 2
  - Avg: 72
  - Stdev: 22
  - Size: 24

$$ttest = \frac{\mu_1 - \mu_2}{\sqrt{\frac{stdev_1^2}{n_1} + \frac{stdev_2^2}{n_2}}}$$

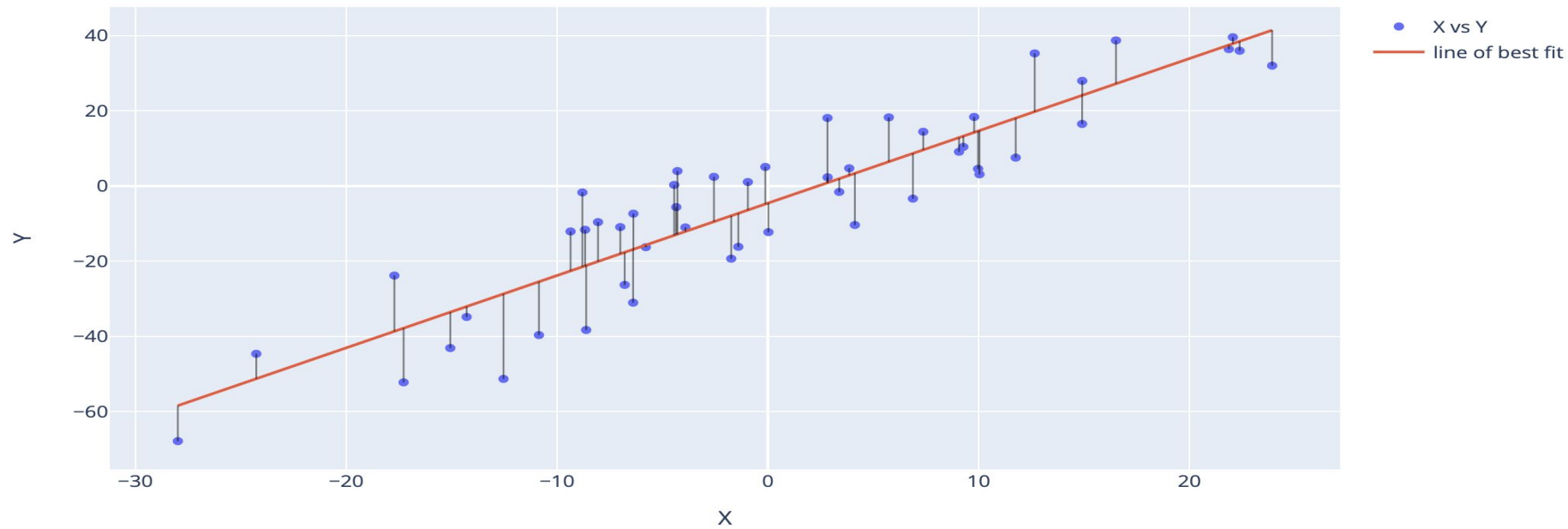
$$t - stat = \frac{82 - 72}{\sqrt{\frac{18^2}{15} + \frac{22^2}{24}}} = \frac{10}{6.46} = 1.54$$

## Linear Regression

$$y_i = a + b * x_i + u_i$$

$$\hat{y}_i = \hat{a} + \hat{b} * x_i$$

$$\hat{u}_i = y_i - \hat{y}_i$$



$$\hat{y}_i = \hat{a} + \hat{b} * x_i$$

$$\hat{b} = \frac{\text{cov}(x, y)}{\text{var}(x)} = \frac{\sum (x_i - \bar{x}) \sum (y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$

$$\hat{a} = \bar{y} - \hat{b} * \bar{x}$$

# Walk through example

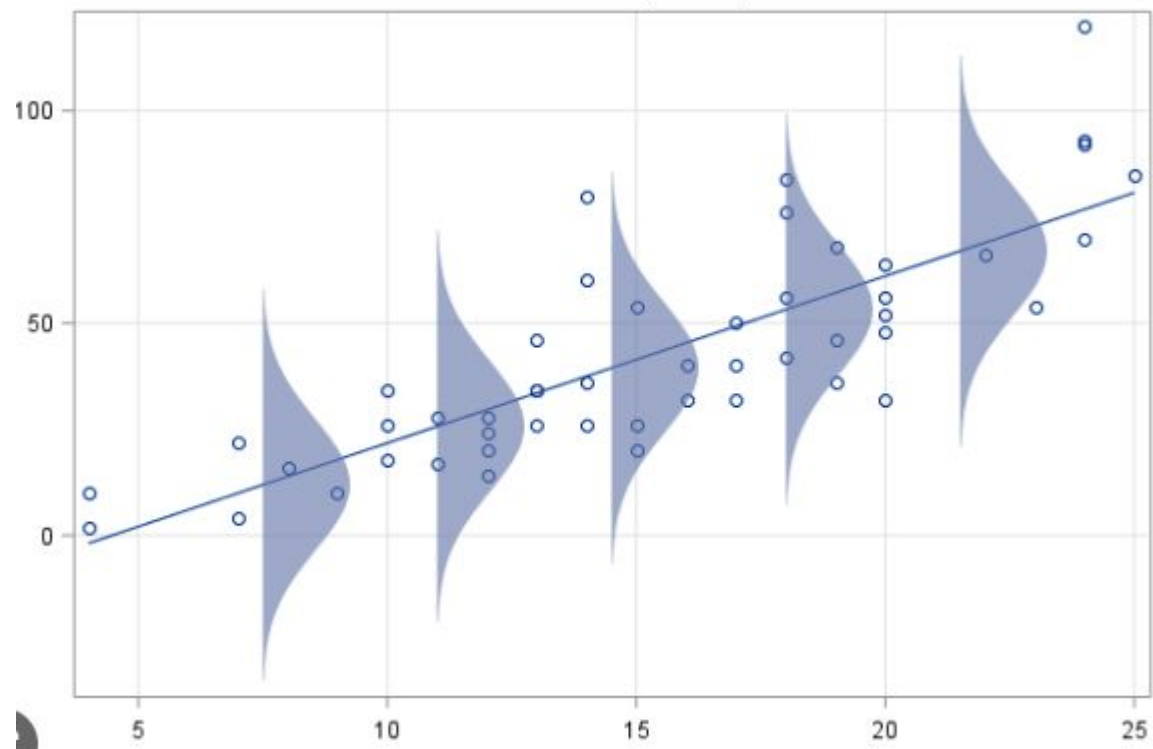
x	y	demean_x	demean_x_sq	demean_y	demean_y_sq	demean_x*demean_y
3	6	-2.4	5.76	1.6	2.56	-3.84
5	9	-0.4	0.16	4.6	21.16	-1.84
2	2	-3.4	11.56	-2.4	5.76	8.16
8	1	2.6	6.76	-3.4	11.56	-8.84
9	4	3.6	12.96	-0.4	0.16	-1.44
			<b>37.2</b>		<b>41.2</b>	-7.8
						<b>-1.95</b>
						sum
						sum/(n-1)

mean_y	4.4
mean_x	5.4

yhat	y-yhat
b0 + b1*x	error
4.90	1.10
4.48	4.52
5.11	-3.11
3.85	-2.85
3.65	0.35

numerator	<b>-1.95</b>	-1.95	<b>-0.199</b>	<b>correlation</b>
denom	sqrt( <b>37.2/4 * 41.2/4</b> )		9.79	

slope	-0.210
intercept	5.53





# Confidence Interval

$H_0: b_1 = 0$  (we don't believe there is a slope (relationship))

$H_A: b_1 \text{ not equal } 0$  (we believe there is a slope (relationship))

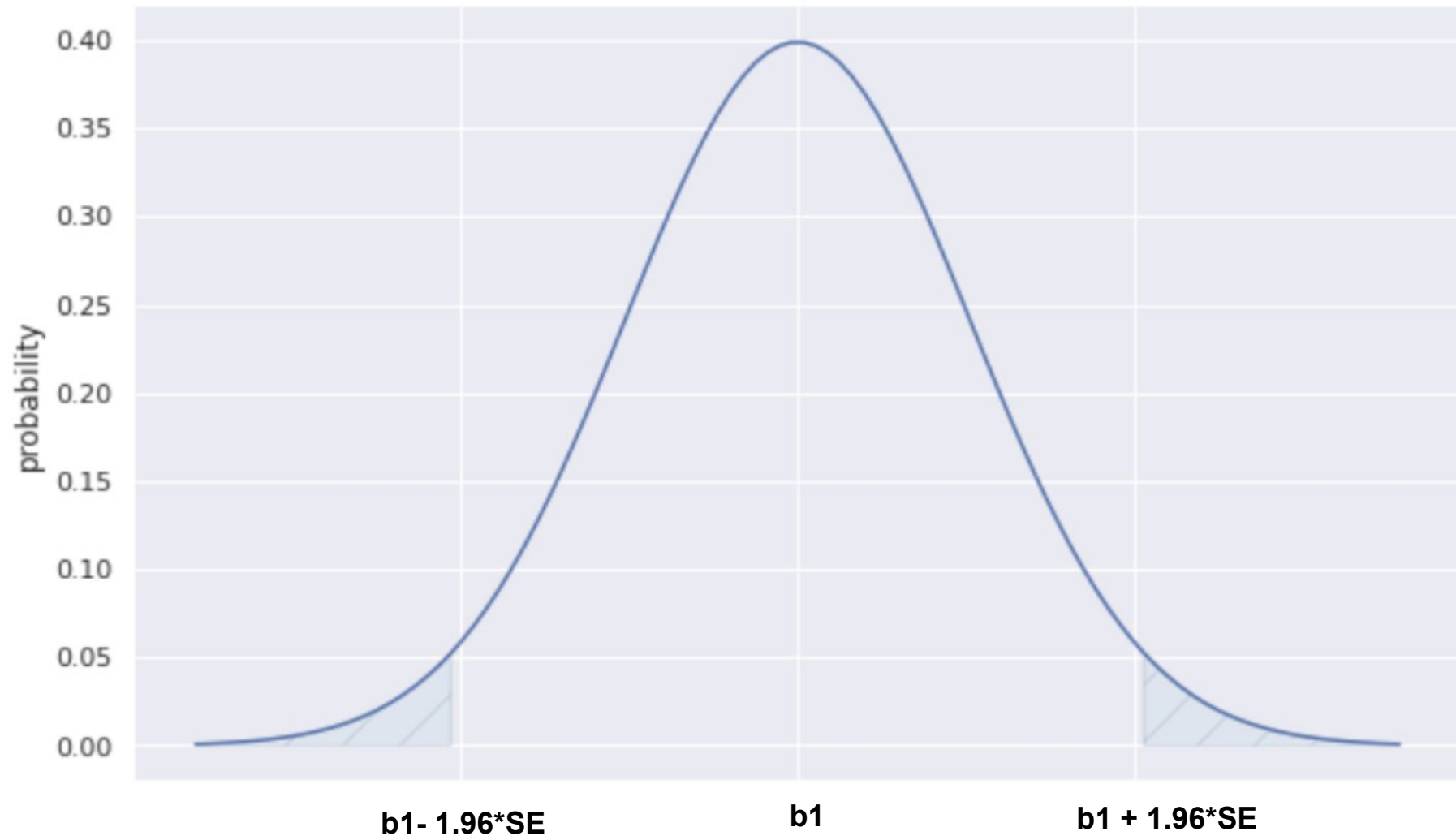
$$t = \frac{b_1 - 0}{SE_{b_1}} = \frac{b_1}{SE_{b_1}}$$

Confidence interval is  $b_1 \pm CV^* SE$ , where CV is critical value. This is easier to show than anything else (see notebook).

$$t = \frac{\bar{x}_1 - \bar{x}_2}{SE}$$

$$(\bar{x}_1 - \bar{x}_2) = SE * t$$

$$(\bar{x}_1 - \bar{x}_2) \pm SE * t = 0$$



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=====
Dep. Variable:          y      R-squared:          0.269
Model:                  OLS    Adj. R-squared:       0.254
Method:                 Least Squares    F-statistic:       17.68
Date:                   Tue, 04 Apr 2023    Prob (F-statistic): 0.000113
Time:                   12:59:08    Log-Likelihood:    -294.80
No. Observations:      50    AIC:              593.6
Df Residuals:          48    BIC:              597.4
Df Model:               1
Covariance Type:       nonrobust
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              coef      std err          t      P>|t|      [0.025      0.975]
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Intercept    15.4229      12.700        1.214      0.231     -10.113      40.959
x            -5.7164       1.360       -4.204      0.000      -8.450      -2.983
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Omnibus:          1.825    Durbin-Watson:          1.558
Prob(Omnibus):    0.401    Jarque-Bera (JB):        1.428
Skew:             -0.414    Prob(JB):                0.490
Kurtosis:         2.975    Cond. No.                9.34
=====

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	coef	std err	t	P> t	[0.025	0.975]
Intercept	15.4229	12.700	1.214	0.231	-10.113	40.959
x	$\frac{Cov(x, y)}{Var(x)}$	1.360	$\frac{b1}{SE}$	0.000	$b1-1.96*b1$	$b1+1.96*b1$

## Binary Standard Error

$$SE = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

When an estimate is of a binomial (1,0) there are some simplifications in the variance.

See here if you are interested in the derivation

[https://en.wikipedia.org/wiki/Binomial\\_proportion\\_confidence\\_interval](https://en.wikipedia.org/wiki/Binomial_proportion_confidence_interval)

$$SE = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

$$\hat{p} \pm z \sqrt{\frac{\hat{p} (1 - \hat{p})}{n}}$$

A newspaper says a randomized national survey found that 40% of the population has covid.

What are the confident intervals with different  $n$ ?

$$n = 10, \quad \sqrt{\frac{0.4(1-0.4)}{10}} = 0.15, CI = [10\%, 70\%]$$

$$n = 100, \quad \sqrt{\frac{0.4(1-0.4)}{100}} = 0.05, CI = [30\%, 50\%]$$

$$n = 1000, \quad \sqrt{\frac{0.4(1-0.4)}{1,000}} = 0.015, CI = [37\%, 43\%]$$

$$n = 10,000, \quad \sqrt{\frac{0.4(1-0.4)}{10,000}} = 0.015, CI = [39\%, 41\%]$$