

# POLS 309: Hypothesis Testing

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# Crafting Hypotheses

- When we craft our hypotheses in political science, we often have some criteria we consider desirable for a good hypothesis:
  - Directionality ( $\uparrow X \rightarrow \uparrow Y$  or  $\uparrow X \rightarrow \downarrow Y$ )
  - Causality (our theory normally is a logical, plausible, falsifiable explanation of how x should cause y)

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- We define our hypotheses this way so that they are exclusive and exhaustive (i.e. there is no third option, only one of these can be true)

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- If we think economic perceptions **do not** influence vote choice, that is our null hypothesis
  - ▶ This is equivalent to setting  $\beta_1 = 0$ , we are saying deterministically economic perception's effect on vote choice is non-existent

# Restricted vs Unrestricted Hypothesis Test

- Whenever we test our null hypotheses, we are effectively testing if imposing this restriction of  $\beta = 0$  is an appropriate restriction OR should we let  $\beta$  be free and allowed to vary
- When our  $p$ -value is lower than some threshold,  $\alpha$ , we reject that this restriction is appropriate and we say: let this parameter  $\beta$  vary.
- When we test our theories, what this means for us is that we are testing: "is making the assumption that our regressor  $\beta$  has no effect, a safe bet?"
  - ▶ That is why we **do not** want to prove our null, we *want* to find evidence against it because a causal and directional theory about  $\beta$  means we want  $\beta$  to influence how  $Y$  changes

# *t*-Tests

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- Definition: A *t*-test is a hypothesis test for hypotheses about a normal RV with an *estimated* standard error
- The *t*-test is considered a generalization of the *z*-test. ***z-tests*** are used when we know our **population** (then we know the true variance and we can estimate  $\frac{\beta}{SE(\beta)}$ ), when we deal with **samples**, we have to rely on the ***t-test***.

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  - N.B. As  $N \rightarrow \infty$ ,  $t \rightarrow z$ .

## t-tests

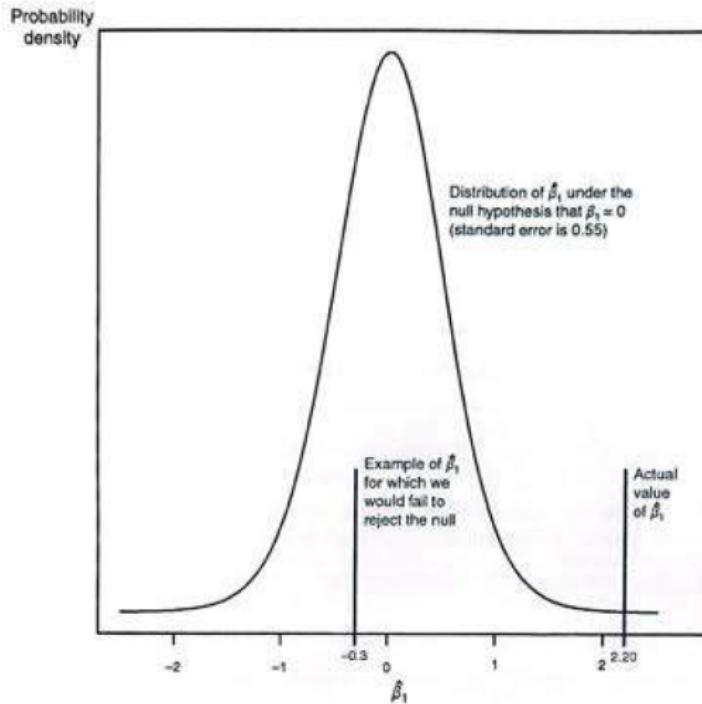


Figure: Distribution of  $\hat{\beta}$  under a null hypothesis (p.99 Bailey 2021)

## *p*-values

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- Our value of  $\alpha$  determines the cut-off at which point we say, observing this value in our distribution is probabilistically, pretty unlikely
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- Why?
- By definition, our *t*-statistic's distribution is normal-ish (with higher kurtosis) and our estimated  $\beta$  is in the tails, that means our estimated  $\beta$  is really unlikely to have occurred.
- So  $\beta$  is unlikely to have really come from this *t*-distribution, meaning it is really unlikely  $\beta$  is 0.

# Quick Review

## General Steps for Hypothesis Testing

- Choose a null hypothesis  $H_0$  and alternative  $H_A$
- Choose a test statistic and its corresponding distribution (e.g.  $t$ -test)
- Choose your significance level  $\alpha$
- Determine your rejection region (one-tailed or two-tailed test)
- Calculate the test statistic ( $t = \frac{\beta}{SE(\beta)}$ )
- Reject  $H_0$  if test statistic falls in rejection region