

What do we get from these assumptions? We get relations between our estimates and our population parameters. That is, we get INTERNAL VALIDITY.

→ Unbiasedness:

$$E[\hat{\beta}_n] = \beta_n$$

→ Consistency:

$$\text{As } n \rightarrow \infty \quad \hat{\beta}_n \rightarrow \beta_n$$

More formally, for all $\epsilon > 0$,

$$P[|\hat{\beta}_n - \beta_n| < \epsilon] \rightarrow 1 \quad \text{as } n \rightarrow \infty$$

Additionally, these assumptions guarantee that

$$\frac{\hat{\beta}_n - \beta_n}{SE(\hat{\beta}_n)} \xrightarrow{d} N(0, 1)$$

Population Param.

- 2 more
- Asymptotic Dist. of $\hat{\beta}$
 - Testing
 - Controls.

So, we can test null hypotheses of " $\beta_n = a$ " by asking, "given what we see, how likely is it that $\frac{\hat{\beta}_n - a}{SE(\hat{\beta}_n)}$ truly is $\rightarrow N(0, 1)$ "

SAVE UNTIL LATER *