# Data Analysis for Mechanical Engineering Type I and II Errors and Power of a Test

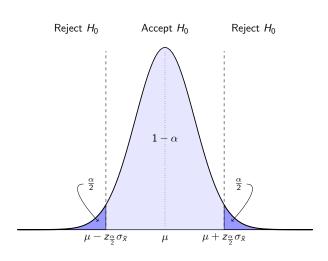
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## Type I Error



$$\alpha = 1 - \Phi(\mu - z_{\frac{\alpha}{2}}\sigma_{\bar{x}} < \bar{x} < \mu + z_{\frac{\alpha}{2}}\sigma_{\bar{x}}; \mu, \sigma_{\bar{x}})$$

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#### Type I Error: False Alarm

Recall, Type I is Rejecting  $H_0$  when the reality is  $H_0$ . The probability of doing this is  $\alpha$ 

$$\alpha = 1 - \Phi(\mu - z_{\frac{\alpha}{2}}\sigma_{\bar{x}} < \bar{X} < \mu + z_{\frac{\alpha}{2}}\sigma_{\bar{x}}; \mu, \sigma_{\bar{x}})$$

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#### Type II Error: Missed Alarm

- Type II is Accepting (or Failing to Reject)  $H_0$  when the reality is  $H_A$ . The probability of doing this is  $\beta$
- One minor issue, you CAN ensure you NEVER make a Type II Error.
- How?
- Not very useful though is it?

#### Type II Error: Example

Let's look at an example to help clarify before formalizing this:

#### Example

- 16 Gears are samples and diameters measured and averaged
- Population variance is known to be 20
- You hypothesize that the population average is 600 at the 0.05 significance level
- What is the chance of making a Type II Error if the real average is 608?

### Type II Error: Example

#### Example

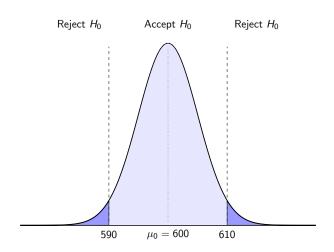
Start with Acceptance Region of  $H_0$ 

$$\mu - z_{0.025} \frac{\sigma}{\sqrt{N}} < \bar{X} < \mu + z_{0.025} \frac{\sigma}{\sqrt{N}}$$

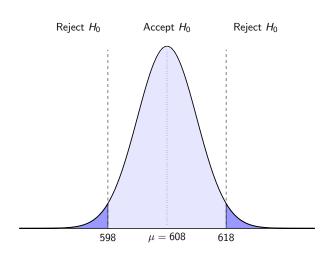
$$0 - (1.96) \frac{20}{\sqrt{N}} < \bar{X} < 600 + (1.96) \frac{\sigma}{\sqrt{N}}$$

$$600 - (1.96)\frac{20}{\sqrt{16}} < \bar{X} < 600 + (1.96)\frac{20}{\sqrt{16}}$$
$$590 < \bar{X} < 610$$

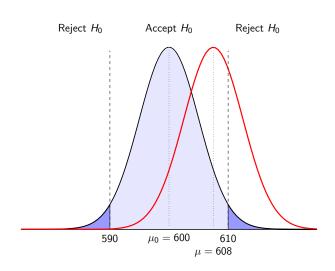
#### So we 'Think' the distribution of sample means should be



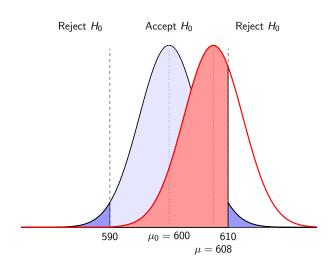
## But the reality is



# Overlaying Them



## Accept $\mu_0$ even though we are wrong!



## Type II Error: Missed Alarm

Type II is Accepting  $H_0$  when the reality is  $H_A$ . The probability of doing this is  $\beta$ 

$$\beta = \Phi(\mu - z_{\frac{\alpha}{2}}\sigma_{\bar{x}} < \bar{X} < \mu + z_{\frac{\alpha}{2}}\sigma_{\bar{x}}; \mu, \sigma_{\bar{x}})$$

$$\beta = \Phi(590 < \bar{X} < 610; 608, 5)$$

$$\beta = \Phi(\frac{590 - 608}{5} < Z < \frac{610 - 608}{5})$$

$$\beta = \Phi(-3.60 < Z < 0.40) = 0.6554$$

65.5% Chance of Type II Error

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## Power of the Test (P)

Another use of  $\beta$  is to determine the power of a test.

#### Power of a Test (P)

The Probability of Detecting a Specified Difference or Effect

$$P = 1 - \beta$$

Power is useful for determining the minimal sample size needed beforehand to have a good shot at detecting a actual significant difference.

# Sample Size

$$N \geqslant \left(\frac{(z_{\beta} + z_{\frac{\alpha}{2}})\sigma}{\delta}\right)^2$$

Note: Assumes you know the population standard deviation, or your sample is above 30.  $\delta$  can either be difference between a known value [one means] or two sample means [two means]

## Sample Size Example

#### Horsepower differences

How many motors are needed to detect a shift from a mean output of 40hp to  $40\pm2hp$  with a probability of 0.90. Significance level needed is 0.05. The population standard deviation is known to be 2.2hp and the distribution of hp is normal.

$$\mu=40$$
 $\delta=2$ 
 $P=1-\beta=0.9$ 
 $\alpha=0.05$ 
 $\sigma=2.2$ 

# Sample Size Example

$$\beta = 0.1 \Rightarrow z_{\beta} = z_{0.1} = 1.28$$

$$\alpha = 0.05 \Rightarrow z_{\frac{\alpha}{2}} = z_{0.025} = 1.96$$

$$N \geqslant \left(\frac{(z_{\beta} + z_{\frac{\alpha}{2}})\sigma}{\delta}\right)^{2}$$

$$N \geqslant \left(\frac{(1.28 + 1.96)(2.2)}{2}\right)^{2}$$

$$N \geqslant (3.24(1.1))^{2} \Rightarrow N \geqslant 12.70 = 13$$

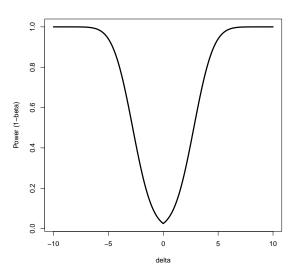
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#### R Code

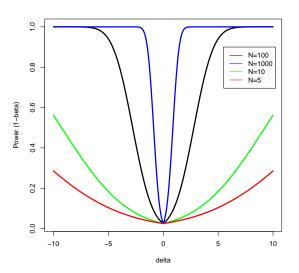
```
power.t.test(100, delta=5, sd=10, sig.level=0.05)
##
##
        Two-sample t test power calculation
##
##
                n = 100
##
             delta = 5
##
                sd = 10
##
         sig.level = 0.05
##
             power = 0.9404
##
       alternative = two.sided
##
## NOTE: n is number in *each* group
```

#### R Code

```
plot (type="1", seq(-10, 10, by=0.1),
  power.t.test(
    100,
    delta=seq(-10, 10, by=0.1),
    sd=10,
    sig.level=0.05) $power,
  xlab="delta",
  ylab="Power (1-beta)",
  lwd="3",
  col="black"
```



# Power Graphs in R - Changing N



# Power Graphs in R - Changing $\sigma$

