

Power

Statistical Inference

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Power

- Power is the probability of rejecting the null hypothesis when it is false
- Ergo, power (as its name would suggest) is a good thing; you want more power
- A type II error (a bad thing, as its name would suggest) is failing to reject the null hypothesis when it's false; the probability of a type II error is usually called β
- Note Power $=1-\beta$

Notes

- Consider our previous example involving RDI
- + $H_0: \mu=30$ versus $H_a: \mu>30$
- Then power is

$$Pigg(rac{ar{X}-30}{s/\sqrt{n}}>t_{1-lpha,n-1}\ ;\ \mu=\mu_aigg)$$

- Note that this is a function that depends on the specific value of $\mu_a!$
- Notice as μ_a approaches 30 the power approaches α

Calculating power for Gaussian data

- We reject if $rac{ar{X}-30}{\sigma/\sqrt{n}}>z_{1-lpha}$
 - Equivalently if $ar{X} > 30 + Z_{1-lpha}\,rac{\sigma}{\sqrt{n}}$
- Under $H_0: ar{X} \sim N(\mu_0, \sigma^2/n)$
- Under $H_a:ar{X}\sim N(\mu_a,\sigma^2/n)$
- So we want

Example continued

```
\cdot \ \mu_a = 32, \, \mu_0 = 30, \, n = 16, \, \sigma = 4
```

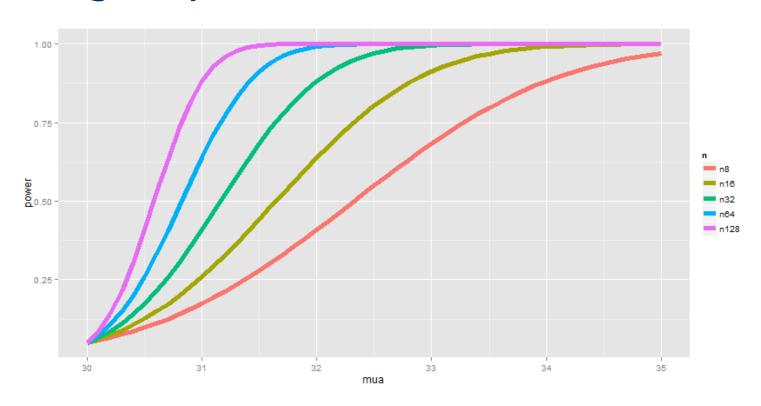
```
mu0 = 30
mua = 32
sigma = 4
n = 16
z = qnorm(1 - alpha)
pnorm(mu0 + z * sigma/sqrt(n), mean = mu0, sd = sigma/sqrt(n), lower.tail = FALSE)
```

```
## [1] 0.05
```

```
pnorm(mu0 + z * sigma/sqrt(n), mean = mua, sd = sigma/sqrt(n), lower.tail = FALSE)
```

```
## [1] 0.6388
```

Plotting the power curve



Graphical Depiction of Power

```
library(manipulate)
mu0 = 30
myplot <- function(sigma, mua, n, alpha) {
    g = ggplot(data.frame(mu = c(27, 36)), aes(x = mu))
    g = g + stat_function(fun = dnorm, geom = "line", args = list(mean = mu0,
        sd = sigma/sqrt(n)), size = 2, col = "red")
    g = g + stat_function(fun = dnorm, geom = "line", args = list(mean = mua,
        sd = sigma/sqrt(n)), size = 2, col = "blue")
    xitc = mu0 + qnorm(1 - alpha) * sigma/sqrt(n)
    q = q + qeom vline(xintercept = xitc, size = 3)
manipulate(myplot(sigma, mua, n, alpha), sigma = slider(1, 10, step = 1, initial = 4),
    mua = slider(30, 35, step = 1, initial = 32), n = slider(1, 50, step = 1,
        initial = 16), alpha = slider(0.01, 0.1, step = 0.01, initial = 0.05))
```

Question

- When testing $H_a: \mu > \mu_0$, notice if power is 1-eta, then

$$1-eta=Pigg(ar{X}>\mu_0+z_{1-lpha}\,rac{\sigma}{\sqrt{n}}\,;\mu=\mu_aigg)$$

- where $ar{X} \sim N(\mu_a, \sigma^2/n)$
- Unknowns: μ_a , σ , n, β
- Knowns: μ_0 , α
- Specify any 3 of the unknowns and you can solve for the remainder

Notes

- The calculation for $H_a: \mu < \mu_0$ is similar
- For $H_a: \mu \neq \mu_0$ calculate the one sided power using $\alpha/2$ (this is only approximately right, it excludes the probability of getting a large TS in the opposite direction of the truth)
- Power goes up as α gets larger
- Power of a one sided test is greater than the power of the associated two sided test
- Power goes up as μ_1 gets further away from μ_0
- Power goes up as n goes up
- Power doesn't need μ_a , σ and n, instead only $\frac{\sqrt{n}(\mu_a-\mu_0)}{\sigma}$
 - The quantity $\frac{\mu_a-\mu_0}{\sigma}$ is called the effect size, the difference in the means in standard deviation units.
 - Being unit free, it has some hope of interpretability across settings

T-test power

- Consider calculating power for a Gossett's T test for our example
- The power is

$$Pigg(rac{ar{X}-\mu_0}{S/\sqrt{n}}>t_{1-lpha,n-1}\ ;\ \mu=\mu_aigg)$$

- Calcuting this requires the non-central t distribution.
- power.t.test does this very well
 - Omit one of the arguments and it solves for it

Example

```
power.t.test(n = 16, delta = 2/4, sd = 1, type = "one.sample", alt = "one.sided")power
```

```
## [1] 0.604
```

```
power.t.test(n = 16, delta = 2, sd = 4, type = "one.sample", alt = "one.sided")$power
```

```
## [1] 0.604
```

```
power.t.test(n = 16, delta = 100, sd = 200, type = "one.sample", alt = "one.sided")$power
```

```
## [1] 0.604
```

Example

```
power.t.test(power = 0.8, delta = 2/4, sd = 1, type = "one.sample", alt = "one.sided")$n
```

```
## [1] 26.14
```

```
power.t.test(power = 0.8, delta = 2, sd = 4, type = "one.sample", alt = "one.sided")$n
```

```
## [1] 26.14
```

```
power.t.test(power = 0.8, delta = 100, sd = 200, type = "one.sample", alt = "one.sided")$n
```

```
## [1] 26.14
```