

# POWER AND SAMPLE SIZE

## BIOS 662

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2008-10-31 14:06

# Outline

- Introduction
- One sample:
  - continuous outcome, Z test
  - continuous outcome, t test
  - binary outcome, Z test
  - binary outcome, exact test

## Introduction

- Read Chs 5.8 and 6.3.3. of text
- In a test of a hypothesis, we are testing whether some population parameter has a particular value

$$H_0 : \theta = \theta_0,$$

where  $\theta_0$  is a known constant

- Generally,

$$H_A : \theta \neq \theta_0$$

## Introduction

- Once the data are collected, we will compute a statistic related to  $\theta$ , say  $S(\hat{\theta})$
- $S(\hat{\theta})$  is a random variable, since it is computed from a sample (and hence it has a probability distribution)

## Power

$$\Pr[\text{Type I error}] = \alpha = \Pr[S(\hat{\theta}) \in C_\alpha | H_0]$$

$$\Pr[\text{Type II error}] = \beta = \Pr[S(\hat{\theta}) \notin C_\alpha | H_A]$$

$$\text{Power} = 1 - \beta = \Pr[S(\hat{\theta}) \in C_\alpha | H_A]$$

## One sample Z test

- Example: One sample test
- Study: collect data on continuous outcome ( $Y$ ) on  $N$  individuals
- $E(Y) = \mu, V(Y) = \sigma^2$

$$H_0 : \mu = \mu_0 \text{ vs } H_A : \mu > \mu_0$$

$$S(\hat{\theta}) = Z = \frac{\bar{Y} - \mu_0}{\sigma / \sqrt{N}}$$

## One sample Z test

$$\Pr[S(\hat{\theta}) \in C_\alpha | H_0] = \Pr[Z > z_{1-\alpha} | H_0] = \alpha$$

$$\Pr[S(\hat{\theta}) \in C_\alpha | H_A] = \Pr[Z > z_{1-\alpha} | H_A] = 1 - \beta$$

- Choose a value  $\mu_A \in H_A$
- Q: what sample size do we need to detect this alternative with  $1 - \beta$  power?

## One sample Z test

- Under  $H_A : \mu = \mu_A$

$$Z' = \frac{\bar{Y} - \mu_A}{\sigma/\sqrt{N}} \sim N(0, 1)$$

- Note

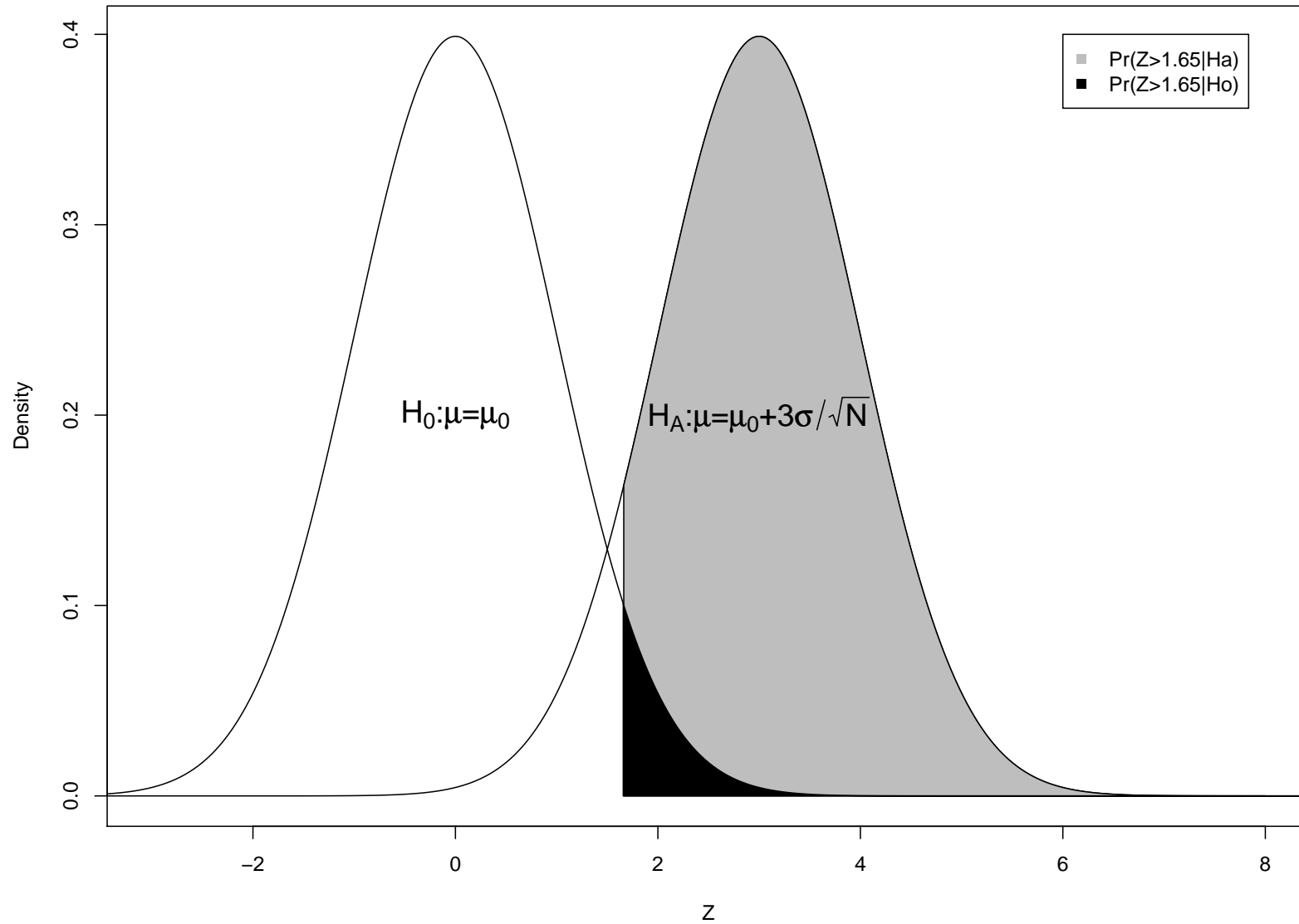
$$Z' + \frac{\mu_A - \mu_0}{\sigma/\sqrt{N}} = \frac{\bar{Y} - \mu_0}{\sigma/\sqrt{N}} = Z$$

- Thus

$$Z \sim N\left(\frac{\mu_A - \mu_0}{\sigma/\sqrt{N}}, 1\right)$$



# Distribution of $Z$ -statistic under $H_0$ and $H_A$



## One sample Z test

- Power is given by

$$\begin{aligned}1 - \beta &= \Pr[Z > z_{1-\alpha} | \mu = \mu_A] \\ &= \Pr[Z' > z_{1-\alpha} + \frac{\sqrt{N}(\mu_0 - \mu_A)}{\sigma} | \mu = \mu_A]\end{aligned}$$

- Therefore

$$z_{1-\alpha} + \frac{\sqrt{N}(\mu_0 - \mu_A)}{\sigma} = z_{\beta} = -z_{1-\beta}$$

## One sample Z test

- Equivalently

$$N = \frac{\sigma^2(z_{1-\alpha} + z_{1-\beta})^2}{(\mu_0 - \mu_A)^2}$$

- For a two sided test

$$N = \frac{\sigma^2(z_{1-\alpha/2} + z_{1-\beta})^2}{(\mu_0 - \mu_A)^2}$$

## One sample Z test

- Equivalent form

$$N = \left( \frac{z_{1-\alpha/2} + z_{1-\beta}}{\Delta} \right)^2$$

where

$$\Delta = \frac{|\mu_0 - \mu_A|}{\sigma}$$

- $\Delta$  is called the *standardized distance* or *difference*

## One sample Z test

- To apply this formula, we need to know  $\alpha$ ,  $\beta$ ,  $|\mu_0 - \mu_A|$ , and  $\sigma^2$
- Example: a study to determine the effect of a drug on blood pressure.
- BP measured, drug administered, and BP measured 2 hours later
- $Y = (BP_{after} - BP_{before})$

$$H_0 : \mu = 0 \text{ vs } H_A : \mu \neq 0$$

## One sample Z test

- Choose  $\alpha = 0.05$ ,  $1 - \beta = 0.9$
- Estimate of  $\sigma^2$ : need data from literature or pilot study; note we need variance of difference after-before
- The choice of  $\mu_A$  is a subject matter decision. In example: a drug that changes BP only 1 mmHg would not be of practical importance, but a drug that changes BP 5 mmHg might be of interest

## One sample Z test

- Suppose  $\alpha = 0.05$ ,  $1 - \beta = .9$ ,  $\sigma^2 = 225$ , and  $\mu_A = 5$

$$N = \frac{225(1.96 + 1.28)^2}{5^2} = 94.5 \approx 95$$

- Often compute  $N$  for various different values of  $\alpha$ ,  $\beta$ ,  $\sigma^2$ , and  $\mu_A$
- Note  $(1.96 + 1.28)^2 \approx 10.5$ , so that

$$N \approx \frac{10.5}{\Delta^2}$$

## One sample Z test

$\alpha$	$1 - \beta$	$\sigma^2$	$\mu_A$	$N$	$\alpha$	$1 - \beta$	$\sigma^2$	$\mu_A$	$N$
0.05	0.90	225	5	95	0.01	0.90	225	5	133
		225	6	66			225	6	93
		256	5	108			256	5	151
		256	6	75			256	6	105
0.05	0.80	225	5	71	0.01	0.80	225	5	105
		225	6	49			225	6	73
		256	5	81			256	5	119
		256	6	56			256	6	83



## One sample Z test

$$\alpha \downarrow \Rightarrow N \uparrow$$

$$1 - \beta \uparrow \Rightarrow N \uparrow$$

$$\sigma^2 \uparrow \Rightarrow N \uparrow$$

$$|\mu_0 - \mu_A| \downarrow \Rightarrow N \uparrow$$

## One sample Z test

- Sometimes  $N$  is fixed and we estimate the power of the test

$$N = \frac{\sigma^2(z_{1-\alpha/2} + z_{1-\beta})^2}{(\mu_0 - \mu_A)^2}$$

$$z_{1-\beta} = \frac{|\mu_0 - \mu_A|\sqrt{N}}{\sigma} - z_{1-\alpha/2}$$

$$1 - \beta = \Phi \left\{ \frac{|\mu_0 - \mu_A|\sqrt{N}}{\sigma} - z_{1-\alpha/2} \right\}$$

## One sample Z test

- Investigator says there are only 50 patients,  $\alpha = .05$ ,  
 $\sigma^2 = 225$ ,  $|\mu_0 - \mu_A| = 5$

$$z_{1-\beta} = \frac{5\sqrt{50}}{15} - 1.96 = 0.40$$

$$1 - \beta = 0.6554$$

## One sample t test

- In practice,  $\sigma$  is not know, so we use a  $t$  test instead of a  $Z$  test
- Results above should be viewed as approximations in this case
- Power of  $t$  test is as follows

## One sample $t$ test

- Need the following result: if  $U \sim N(\lambda, 1)$  and  $V \sim \chi_\nu^2$  where  $U \perp V$ , then

$$\frac{U}{\sqrt{V/\nu}} \sim t_{\nu, \lambda}$$

i.e., a non-central  $t$  distribution with df  $\nu$  and non-centrality parameter  $\lambda$

## One sample $t$ test

- Consider  $H_0 : \mu = 0$  versus  $H_A : \mu = \mu_A$  for  $\mu_A \neq 0$
- Under  $H_A$ ,

$$\bar{Y} \sim N(\mu_A, \sigma^2/N)$$

- Thus

$$\frac{\bar{Y}\sqrt{N}}{\sigma} \sim N\left(\mu_A\frac{\sqrt{N}}{\sigma}, 1\right)$$

- Recall

$$\frac{(N-1)s^2}{\sigma^2} \sim \chi_{N-1}^2$$

## One sample $t$ test

- Since  $\bar{Y} \perp s^2$ ,

$$T = \frac{\bar{Y}}{s/\sqrt{N}} \sim t_{N-1,\lambda}$$

where  $\lambda = \mu_A \sqrt{N}/\sigma$

- So power of two-sided  $t$ -test for  $\mu_A > 0$

$$\Pr[T \geq t_{N-1,0;1-\alpha/2}]$$

where  $T \sim t_{N-1,\mu_A \sqrt{N}/\sigma}$

## One sample $t$ test: R

```
# by hand
> 1-pt(qt(.975,49), 49, 5/15*sqrt(50))
[1] 0.6370846

> power.t.test(n=50, sd=15, delta=5, type="one.sample")
```

One-sample t test power calculation

```
      n = 50
  delta = 5
     sd = 15
sig.level = 0.05
  power = 0.6370846
alternative = two.sided
```



# One sample $t$ test: SAS

```
proc power;  
  onesamplemeans  
    mean = 5  
    ntotal = 50  
    stddev = 15  
    power = .; run;
```

The POWER Procedure  
One-sample t Test for Mean

## Fixed Scenario Elements

Distribution	Normal
Method	Exact
Mean	5
Standard Deviation	15
Total Sample Size	50
Number of Sides	2
Null Mean	0
Alpha	0.05

Computed Power

Power

0.637

## One sample Z test: Binary Outcome

- Null hypothesis

$$H_0 : \pi = \pi_0$$

- Test statistic

$$Z = \frac{\hat{p} - \pi_0}{\sqrt{\pi_0(1 - \pi_0)/N}}$$

- Sample size formula

$$N = \frac{\left( z_{1-\alpha/2} \sqrt{\pi_0(1 - \pi_0)} + z_{1-\beta} \sqrt{\pi_A(1 - \pi_A)} \right)^2}{(\pi_A - \pi_0)^2}$$

## One sample Z test: Binary Outcome

- Example: Study of risk of breast cancer in sibs.
- Prevalence in general population of women 50-54 years old: 2%
- Sample sisters of women with breast cancer

$$H_0 : \pi = 0.02 \text{ vs } H_A : \pi \neq .02$$

## One sample Z test: Binary Outcome

- Suppose  $\alpha = .05$ ,  $1 - \beta = .9$ ,  $\pi_A = .05$

- Then

$$N = \frac{\left(1.96\sqrt{.02(.98)} + 1.28\sqrt{.05(.95)}\right)^2}{(.05 - .02)^2} = 340.65 \approx 341$$

# One sample Z test with binary outcome: SAS

```
proc power;  
  onesamplefreq test = z  
  method = normal  
  nullp = .02  
  p = .05  
  power = .9  
  ntotal = .;  
run;
```

The POWER Procedure

Z Test for Binomial Proportion

## Fixed Scenario Elements

Method	Normal approximation
Null Proportion	0.02
Binomial Proportion	0.05
Nominal Power	0.9
Number of Sides	2
Alpha	0.05

Actual	N
Power	Total
0.900	341

## One sample exact test: binary outcome

- What is power of exact test?
- Let  $Y$ , the number of successes, be test statistic.
- Under  $H_0$ ,  $Y \sim \text{Binomial}(N, \pi_0)$   
Under  $H_A$ ,  $Y \sim \text{Binomial}(N, \pi_A)$
- Power

$$\Pr[Y \geq y_{1-\alpha/2} | \pi = \pi_A] + \Pr[Y \leq y_{\alpha/2} | \pi = \pi_A]$$

where  $y_{\alpha/2}$  and  $y_{1-\alpha/2}$  determined as in Counting Data slides

## One sample exact test: binary outcome

- Exact example. Suppose  $N = 20$ ,  $\pi_0 = .2$ ,  $\pi_A = .5$ ,  $\alpha = 0.05$ . What is exact power of two-sided test?
- Based on CDF of *Binomial*(20, .2), choose  $y_{\alpha/2} = 0$  and  $y_{1-\alpha/2} = 9$  such that power equals

$$\Pr[Y \geq 9 | \pi = .5] + \Pr[Y \leq 0 | \pi = .5] = 0.748$$

# One sample exact test, binary outcome: SAS

```
proc power;  
  onesamplefreq test = exact  
  method = exact  
  nullp = .2  
  p = .5  
  power = .  
  ntotal = 20;  
run;
```

Computed Power			
Lower	Upper		
Crit	Crit	Actual	
Val	Val	Alpha	Power
0	9	0.0215	0.748