

## Power and Sample Size: Issues in Study Design

John McGready

Department of Biostatistics, Bloomberg School

---

---

---

---

---

---

---

### Lecture Topics

- Re-visit concept of statistical power
- Factors influencing power
- Sample size determination when comparing two means
- Sample size determination when comparing two proportions

2

---

---

---

---

---

---

---

### Section A

Power and Its Influences

---

---

---

---

---

---

---

### Example

- Consider the following results from a study done on 29 women, all 35-39 years old

Sample Data			
	n	Mean SBP	SD of SBP
OC users	8	132.8	15.3
Non-OC Users	21	127.4	18.2

4

---

---

---

---

---

---

---

### Example

- Of particular interest is whether OC use is associated with higher blood pressure
- Statistically speaking we are interested in testing :

$$H_0: \mu_{OC} = \mu_{NO\ OC} \quad H_0: \mu_{OC} - \mu_{NO\ OC} = 0$$
$$H_A: \mu_{OC} \neq \mu_{NO\ OC} \quad H_A: \mu_{OC} - \mu_{NO\ OC} \neq 0$$

Here  $\mu_{OC}$  represents (population) mean SBP for OC users,  $\mu_{NO\ OC}$  (population ) mean BP for women not using OC

5

---

---

---

---

---

---

---

### Example

- Let's unveil the results of this test from our data:

Sample Data			
	n	Mean SBP	SD of SBP
OC users	8	132.8	15.3
Non-OC Users	21	127.4	18.2

6

---

---

---

---

---

---

---

### Example

- The sample mean difference in blood pressures is  $132.8 - 127.4 = 5.4$
- This could be considered scientifically significant, however, the result is not statistically significant (or even close to it!) at the  $\alpha = .05$  level
- Suppose, as a researcher, you were concerned about detecting a population difference of this magnitude if it truly existed
- This particular study of 29 women has low power to detect a difference of such magnitude

7

---

---

---

---

---

---

---

---

### Power

- Recall, from lecture four of SR1:

		TRUTH	
		$H_0$	$H_A$
D E C I S I O N	Reject $H_0$	Type I Error alpha-level	Power 1-beta
	Not Reject $H_0$		Type II Error beta

8

---

---

---

---

---

---

---

---

### Power

- Power is a measure of "doing the right thing" when  $H_A$  is true!
- Higher power is better (the closer the power is to 1.0 or 100%)
- We can calculate power for a given study if we specify a specific  $H_A$
- This OC/Blood pressure study has power of .13 to detect a difference in blood pressure of 5.4 or more, if this difference truly exists in the population of women 35-39 years old!
- When power is this low, it is difficult to determine whether there is no statistical difference in population means or we just could not detect it

9

---

---

---

---

---

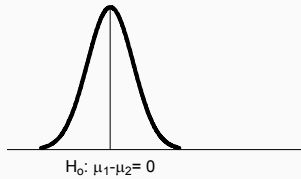
---

---

---

### Power

- Recall, the sampling behavior of  $\bar{x}_1 - \bar{x}_2$  is normally distributed (large samples) with this sampling distributed centered at true mean difference. If  $H_0$  truth, then curve is centered at 0



10

---

---

---

---

---

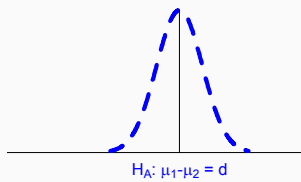
---

---

---

### Power

- Recall, the sampling behavior of  $\bar{x}_1 - \bar{x}_2$  is normally distributed (large samples) with this sampling distributed centered at true mean difference. If  $H_A$  truth, then curve is centered at some value  $d$ ,  $d \neq 0$



11

---

---

---

---

---

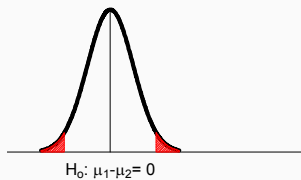
---

---

---

### Power

- $H_0$  will be rejected (for  $\alpha=.05$ ) if the sample result,  $\bar{x}_1 - \bar{x}_2$ , is more than 2 standard errors away from 0, either above or below



12

---

---

---

---

---

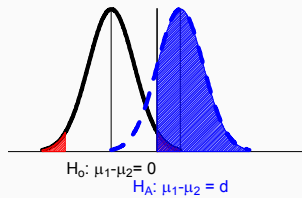
---

---

---

## Power

- $H_0$  will be rejected (for  $\alpha=.05$ ) if the sample result,  $\bar{x}_1 - \bar{x}_2$ , is more than 2 standard errors away from 0, either above or below



13

---

---

---

---

---

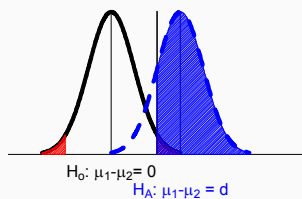
---

---

---

## What Influences Power?

- In order to INCREASE power for a study comparing two group means, we need to do the following:
  1. Change the estimates of  $\mu_1$  and  $\mu_2$  so that the difference between the two ( $\mu_1 - \mu_2$ ) is bigger



14

---

---

---

---

---

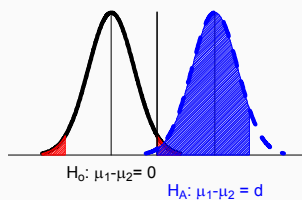
---

---

---

## What Influences Power?

- In order to INCREASE power for a study comparing two group means, we need to do the following:
  1. Change the estimates of  $\mu_1$  and  $\mu_2$  so that the difference between the two ( $\mu_1 - \mu_2$ ) is bigger



15

---

---

---

---

---

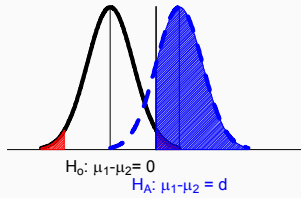
---

---

---

### What Influences Power?

- In order to INCREASE power for a study comparing two group means, we need to do the following:
  2. Increase the sample size in each group



16

---

---

---

---

---

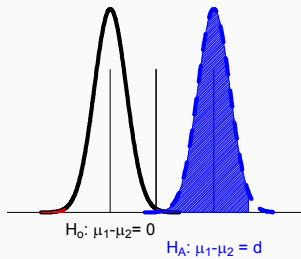
---

---

---

### What Influences Power?

- In order to INCREASE power for a study comparing two group means, we need to do the following:
  2. Increase the sample size in each group



17

---

---

---

---

---

---

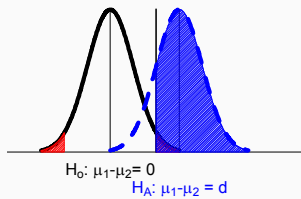
---

---

### What Influences Power?

- In order to INCREASE power for a study comparing two group means, we need to do the following:
  3. Increase the  $\alpha$ -level of the hypothesis test (functionally speaking, make it "easier to reject")

here, with  $\alpha = .05$ :



18

---

---

---

---

---

---

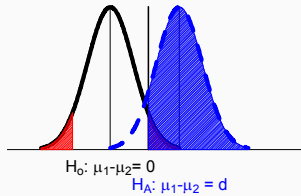
---

---

### What Influences Power?

- In order to INCREASE power for a study comparing two group means, we need to do the following:
  3. Increase the  $\alpha$ -level of the hypothesis test (functionally speaking, make it “easier to reject”)

here, with  $\alpha=.10$ :



19

---

---

---

---

---

---

---

---

### Power and Studies

- Power can be computed after a study is completed
  - Can only be computed for specific  $H_A$ 's: i.e. this study had XX% to detect a difference in population means of YY or greater.
  - Sometimes presented as an “excuse” for non statistically significant finding: “the lack of a statically significant association between A and B could be because of low power (< 15%) to detect a mean difference of YY or greater between..”
  - Can also be presented to corroborate with a non statistically significant result
- Many times, in study design, a required sample size is computed to actually achieve a certain preset power level to find a “Clinically/scientifically” minimal important difference in means
  - In next section we will show how to do this using Stata
  - “Industry standard” for power: 80% (or greater)

20

---

---

---

---

---

---

---

---

### Section B

#### Sample Size Calculations when Comparing Group Means

---

---

---

---

---

---

---

---

### Example

- Blood pressure and oral contraceptives
  - Suppose we used data from the example in Section A to motivate the following question:
  - Is oral contraceptive use associated with higher blood pressure among individuals between the ages of 35-39?

22

---

---

---

---

---

---

---

---

### Pilot Study

- Recall, the data:

Sample Data			
	n	Mean SBP	SD of SBP
OC users	8	132.8	15.3
Non-OC Users	21	127.4	18.2

23

---

---

---

---

---

---

---

---

### Pilot Study

- We think this research has a potentially interesting association
- We want to do a bigger study
  - We want this larger study to have ample power to detect this association, should it really exist in the population
- What we want to do is determine sample sizes needed to detect about a 5mm increase in blood pressure in O.C. users with 80% power at significance level  $\alpha = .05$ 
  - Using pilot data, we estimate that the standard deviations are 15.3 and 18.2 in O.C. and non-O.C. users respectively

24

---

---

---

---

---

---

---

---



### Pilot Study

- Here we have a desired power in mind and want to find the sample sizes necessary to achieve a power of 80% to detect a population difference in blood pressure of five or more mmHg between the two groups

25

---

---

---

---

---

---

---

---

### Pilot Study

- We can find the necessary sample size(s) of this study if we specify .
  - $\alpha$ -level of test (.05)
  - Specific values for  $\mu_1$  and  $\mu_2$  (specific  $H_A$ ) and hence  $d = \mu_1 - \mu_2$ : usually represents the minimum scientific difference of interest)
  - Estimates of  $\sigma_1$  and  $\sigma_2$
  - The desired power(.80)

26

---

---

---

---

---

---

---

---

### Pilot Study

- How can we specify  $d = \mu_1 - \mu_2$  and estimate population SDs?
  - Researcher knowledge—experience makes for good educated guesses
  - Make use of pilot study data!

27

---

---

---

---

---

---

---

---

### Pilot Study

- Fill in blanks from pilot study
  - $\alpha$  -level of test (.05)
  - Specific  $H_A$  ( $\mu_{OC}=132.8$ ,  $\mu_{NO\ OC}=127.4$ ), and hence  $d=\mu_1-\mu_2=5.4$  mmHg
  - Estimates of  $\sigma_{OC}$  (= 15.3) and  $\sigma_{NO\ OC}$  (=18.2)
  - The power we desire (.80)

28

---

---

---

---

---

---

---

---

### Pilot Study

- Given this information, we can use Stata to do the sample size calculation
- “sampsi” command
  - Command syntax (items in *italics* are numbers to be supplied by researcher)

sampsi  $\mu_1$   $\mu_2$ , alpha( $\alpha$ ) power(*power*) sd1( $\sigma_1$ ) sd2( $\sigma_2$ )

29

---

---

---

---

---

---

---

---

### Stata Results

- Blood Pressure/OC example

```
. sampsi 132.8 127.4, alpha(.05) power(.8) sd1(15.3) sd2(18.2)

Estimated sample size for two-sample comparison of means

Test Ho: m1 = m2, where m1 is the mean in population 1
               and m2 is the mean in population 2

Assumptions:

      alpha = 0.0500 (two-sided)
      power = 0.8000
      m1 = 132.8
      m2 = 127.4
      sd1 = 15.3
      sd2 = 18.2
      n2/n1 = 1.00

Estimated required sample sizes:

      n1 = 153
      n2 = 153
```

30

---

---

---

---

---

---

---

---

### Pilot Study/Stata Results

- Our results from Stata suggest that in order to detect a difference in B.P. of 5.4 units (if it really exists in the population) with high (80%) certainty, we would need to enroll 153 O.C. users and 153 non-users
- This assumed that we wanted equal numbers of women in each group!

31

---

---

---

---

---

---

---

---

### Stata Resultss

- Blood Pressure/OC example

```
. sampsi 132.8 127.4, alpha(.05) power(.8) sd1(15.3) sd2(18.2)
Estimated sample size for two-sample comparison of means
Test Ho: m1 = m2, where m1 is the mean in population 1
                    and m2 is the mean in population 2
Assumptions:
      alpha = 0.0500 (two-sided)
      power = 0.8000
      m1 = 132.8
      m2 = 127.4
      sd1 = 15.3
      sd2 = 18.2
      n2/n1 = 1.00
Estimated required sample sizes:
      n1 = 153
      n2 = 153
```

32

---

---

---

---

---

---

---

---

### Pilot Study/Stata Results

- Suppose we estimate that the prevalence of O.C. use amongst women 35-39 years of age is 20%
  - We wanted this reflected in our group sizes
- If O.C. users are 20% of the population, non-OC users are 80%
  - There are four times as many non-users as there are users (4:1 ratio)

33

---

---

---

---

---

---

---

---

### Pilot Study/Stata Results

- We can specify a ratio of group sizes in Stata
  - Again, using “sampsi” command with “ratio” option

sampsi  $\mu_1$   $\mu_2$ , alpha( $\alpha$ ) power( $power$ ) sd1( $\sigma_1$ ) sd2( $\sigma_2$ ) ratio( $n_2/n_1$ )

34

---

---

---

---

---

---

---

---

### Stata Resultss

- Blood Pressure/OC example

```
. sampsi 132.8 127.4, alpha(.05) power(.8) sd1(15.3) sd2(18.2) ratio(4)
```

Estimated sample size for two-sample comparison of means

Test Ho:  $\mu_1 = \mu_2$ , where  $\mu_1$  is the mean in population 1  
and  $\mu_2$  is the mean in population 2

Assumptions:

```
alpha = 0.0500 (two-sided)
power = 0.8000
m1 = 132.8
m2 = 127.4
sd1 = 15.3
```

```
sd2 = 18.2
n2/n1 = 4.000
```

Estimated required sample sizes

```
n1 = 88
n2 = 352
```

35

---

---

---

---

---

---

---

---

## Section C

### Sample Size Determination for Comparing Two Proportions

---

---

---

---

---

---

---

---

### Power for Comparing Two Proportions

- Same ideas as with comparing means

37

---

---

---

---

---

---

---

---

### Pilot Study

- We can find the necessary sample size(s) of this study if we specify .
  - $\alpha$ -level of test
  - Specific values for  $p_1$  and  $p_2$  (specific  $H_A$ ) and hence  $d = p_1 - p_2$ : usually represents the minimum scientific difference of interest)
  - The desired power

38

---

---

---

---

---

---

---

---

### Pilot Study

- Given this information, we can use Stata to do the sample size calculation
  - “samps” command
    - Command syntax (items in italics are numbers to be supplied by researcher)
- samps  $p_1$   $p_2$ , alpha( $\alpha$ ) power(*power*)

39

---

---

---

---

---

---

---

---

### Another Example

- Two drugs for treatment of peptic ulcer compared (Familiari, et al., 1981)
  - The percentage of ulcers healed by pirenzepine ( drug A) and trithiozine ( drug B) was 77% and 58% based on 30 and 31 patients respectively (p-value = .17), 95% CI for difference in proportions healed ( $P_{DRUG A} - P_{DRUG B}$ ) was (-.04, .42)
  - The power to detect a difference as large as the sample results with samples of size 30 and 31 respectively is only 25%

	Healed	Not Healed	Total
Drug A	23	7	30
Drug B	18	13	31

■ ▼ Notes Available

■ Continued 40

---

---

---

---

---

---

---

---

### Example

- As a clinician, you find the sample results intriguing - want to do a larger study to better quantify the difference in proportions healed
- Redesign a new trial, using aforementioned study results to “guestimate” population characteristics
  - Use  $P_{DRUG A} = .77$  and  $P_{DRUG B} = .58$
  - 80% power
  - $\alpha = .05$
- Command in Stata  
samps `.77 .58, alpha (.05) power (.8)`

41

---

---

---

---

---

---

---

---

### Example

- Command in Stata  
samps `.77 .58, alpha (.05) power (.8)`  
  
`. samps .77 .58, alpha (.05) power (.8)`  
  
Estimated sample size for two-sample comparison of proportions  
  
Test Ho:  $p1 = p2$ , where  $p1$  is the proportion in population 1  
and  $p2$  is the proportion in population 2  
  
Assumptions:  
  
alpha = 0.0500 (two-sided)  
power = 0.8000  
p1 = 0.7700  
p2 = 0.5800  
n2/n1 = 1.00  
  
Estimated required sample sizes:  
  

n1 =	105
n2 =	105

42

---

---

---

---

---

---

---

---

### Example

- What would happen if you change power to 90%?  
`sampsi .77 .58, alpha (.05) power (.9)`

```
. sampsi .77 .58, alpha (.05) power (.9)

Estimated sample size for two-sample comparison of proportions
```

```
Test Ho: p1 = p2, where p1 is the proportion in population 1
              and p2 is the proportion in population 2
```

Assumptions:

```
alpha = 0.0500 (two-sided)
power = 0.9000
p1 = 0.7700
p2 = 0.5800
n2/n1 = 1.00
```

Estimated required sample sizes:

n1	=	136
n2	=	136

43

---

---

---

---

---

---

---

---

### Example

- Suppose you wanted two times as many people on trithiozone ("Drug B") as compared to pirenzepine ("Drug A")
  - Here, the ratio of sample size for Group 2 to Group 1 is 2.0
- Can use "ratio" option in "sampsi" command

44

---

---

---

---

---

---

---

---

### Example

- Changing the ratio  
`sampsi .77 .58, alpha (.05) power (.9) ratio(2)`

```
. sampsi .77 .58, alpha (.05) power (.9) ratio(2)

Estimated sample size for two-sample comparison of proportions
```

```
Test Ho: p1 = p2, where p1 is the proportion in population 1
              and p2 is the proportion in population 2
```

Assumptions:

```
alpha = 0.0500 (two-sided)
power = 0.9000
p1 = 0.7700
p2 = 0.5800
n2/n1 = 2.00
```

Estimated required sample sizes:

n1	=	193
n2	=	206

45

---

---

---

---

---

---

---

---

### Sample Size for Comparing Two Proportions

- A randomized trial is being designed to determine if vitamin A supplementation can reduce the risk of breast cancer
  - The study will follow women between the ages of 45-65 for one year
  - Women were randomized between vitamin A and placebo
- What sample sizes are recommended?

46

---

---

---

---

---

---

---

---

### Breast Cancer/Vitamin A Example

- Design a study to have 80% power to detect a 50% *relative reduction* in risk of breast cancer w/vitamin A  
 (i.e.  $RR = \frac{P_{VITA}}{P_{PLACEBO}} = .50$ )
- using a (two-sided) test with significance level  $\alpha$ -level = .05
- To get estimates of proportions of interest:
  - using other studies, the breast cancer rate in the controls can be assumed to be 150/100,000 per year

47

---

---

---

---

---

---

---

---

### Breast Cancer/Vitamin A Example

- A 50% relative reduction: if

$$RR = \frac{P_{VITA}}{P_{PLACEBO}} = .50 \text{ then, } p_{VITA} = .50 \times p_{PLACEBO}$$

So, for this desired difference in the relative scale:

$$p_{VITA} = \frac{150}{100,000} \times 0.5 = \frac{75}{100,000}$$

Notice, that this difference on the absolute scale,  $p_{VITA} - p_{PLACEBO}$ , is much smaller in magnitude:

$$-\frac{75}{100,000} = -0.00075 = -0.075\%$$

48

---

---

---

---

---

---

---

---



### Breast Cancer Sample Size Calculation in Stata

- “sampsi” command

sampsi .00075 .0015, alpha(.05) power(.8)

```
. sampsi .00075 .0015, alpha(.05) power(.8)
```

Estimated sample size for two-sample comparison of proportions

Test Ho: p1 = p2, where p1 is the proportion in population 1  
and p2 is the proportion in population 2

Assumptions:

```
alpha = 0.0500 (two-sided)
power = 0.8000
p1 = 0.0008
p2 = 0.0015
n2/n1 = 1.00
```

Estimated required sample sizes:

```
n1 = 35974
n2 = 35974
```

49

---

---

---

---

---

---

---

---

---

---

### Breast Cancer Sample Size Calculation in Stata

- You would need about 34,000 individuals per group

- Why so many?

- Difference between two hypothesized proportions is very small:  
= .00075

We would expect about 50 cancer cases among the controls and 25 cancer cases among the vitamin A group

$$\text{placebo} : \frac{150}{100,000} \times 34,000 = 51$$

$$\text{vitamin A} : \frac{75}{100,000} \times 34,000 \approx 25$$

50

---

---

---

---

---

---

---

---

---

---

### Breast Cancer/Vitamin A Example

- Suppose you want 80% power to detect only a 20% (relative) reduction in risk associated with vitamin A

A 20% relative reduction: if  $RR = \frac{P_{VITA}}{P_{PLACEBO}} = .80$  then  $P_{VITA} = .80 \times P_{PLACEBO}$

So, for this desired difference in the relative scale:

$$P_{VITA} = \frac{150}{100,000} \times 0.8 = \frac{120}{100,000}$$

Notice, that this difference on the absolute scale,  $P_{VITA} - P_{PLACEBO}$ , is much smaller in magnitude:

$$-\frac{30}{100,000} = -0.0003 = -0.03\%$$

51

---

---

---

---

---

---

---

---

---

---

### Breast Cancer Sample Size Calculation in Stata

```
■ "sampsi" command
  sampsi .0012 .0015, alpha(.05) power(.8)

. sampsi .0012 .0015, alpha(.05) power(.8)

Estimated sample size for two-sample comparison of proportions

Test Ho: p1 = p2, where p1 is the proportion in population 1
              and p2 is the proportion in population 2

Assumptions:

      alpha = 0.0500 (two-sided)
      power = 0.8000
      p1 = 0.0012
      p2 = 0.0015
      n2/n1 = 1.00

Estimated required sample sizes:
```

```
n1 = 241769
n2 = 241769
```

52

---

---

---

---

---

---

---

---

### Breast Cancer—Vitamin A Example Revisited

- You would need about 242,000 per group!
- We would expect 360 cancer cases among the placebo group and 290 among vitamin A group

53

---

---

---

---

---

---

---

---

### An Alternative Approach—Design a Longer Study

- Proposal
  - Five-year follow-up instead of one yearHere:
$$p_{VITA} = 5 \times .0012 = .006$$
$$p_{PLACEBO} = 5 \times .0015 = .0075$$
- Need about 48,000 per group
  - Yields about 290 cases among vitamin A and 360 cases among placebo
- Issue
  - Loss to follow-up

54

---

---

---

---

---

---

---

---

## Section D

### Sample Size and Study Design Principles: A Brief Summary

---

---

---

---

---

---

---

### Designing Your Own Study

- When designing a study, there is a tradeoff between :
  - Power
  - $\alpha$ -level
  - Sample size
  - Minimum detectable difference (specific  $H_a$ )

- Industry standard—80% power,  $\alpha = .05$

56

---

---

---

---

---

---

---

### Designing Your Own Study

- What if sample size calculation yields group sizes that are too big (i.e., can not afford to do study) or are very difficult to recruit subjects for study?
  - Increase minimum difference of interest
  - Increase  $\alpha$ -level
  - Decrease desired power

57

---

---

---

---

---

---

---

### Designing Your Own Study

- Sample size calculations are an important part of study proposal
  - Study funders want to know that the researcher can detect a relationship with a high degree of certainty (should it really exist)
- Even if you anticipate confounding factors, these approaches are the best you can do and are relatively easy
- Accounting for confounders requires more information and sample size has to be done via computer simulation—consult a statistician!

58

---

---

---

---

---

---

---

---

### Designing Your Own Study

- When would you calculate the power of a study?
  - Secondary data analysis
  - Data has already been collected, sample size is fixed
  - *Pilot Study*—to illustrate that low power may be a contributing factor to non-significant results and that a larger study may be appropriate

59

---

---

---

---

---

---

---

---

### Designing Your Own Study

- What is this specific alternative hypothesis?
  - *Power or sample size* can only be calculated for a specific alternative hypothesis
  - When comparing two groups this means estimating the true population means (proportions) for each group

60

---

---

---

---

---

---

---

---

### Designing Your Own Study

- What is this specific alternative hypothesis?
  - Therefore specifying a difference between the two groups
  - This difference is frequently called minimum detectable difference or effect size, referring to the minimum detectable difference with scientific interest

61

---

---

---

---

---

---

---

---

### Designing Your Own Study

- Where does this specific alternative hypothesis come from?
  - Hopefully, not the statistician!
  - As this is generally a quantity of scientific interest, it is best estimated by a knowledgeable researcher or pilot study data
  - This is perhaps the most difficult component of sample size calculations, as there is no magic rule or “industry standard”

62

---

---

---

---

---

---

---

---

### FYI—Using Stata to Compute Power

- I promised you in part A of this lecture that I would eventually show you how to compute the power to detect difference in a study that has already been conducted
- The “sampsi” command is still the command for this—we just need to feed it slightly different information for it to compute power

63

---

---

---

---

---

---

---

---

### Calculating Power

- In order to calculate power for a study comparing two population means, we need the following:
  - Sample size for each group
  - Estimated (population) means,  $\mu_1$  and  $\mu_2$  for each group—these values frame a specific alternative hypothesis (usually minimum difference of scientific interest)
  - Estimated (population) SD's,  $\sigma_1$  and  $\sigma_2$
  - $\alpha$ -level of the hypothesis test

64

---

---

---

---

---

---

---

---

### Calculating Power

- The Blood Pressure/Oral Contraceptive Example

Sample Data			
	n	Mean SBP	SD of SBP
OC users	8	132.8	15.3
Non-OC Users	21	127.4	18.2

65

---

---

---

---

---

---

---

---

### Calculating Power

- Fill in information below with results from this study
  - Sample size for each group (  $n_{OC} = 8$ ,  $n_{NO\ OC} = 21$  )
  - Estimated (population) means,  $\mu_{OC} = 132.8$  and  $\mu_{NO\ OC} = 127.4$
  - Estimated (population) sd's,  $\sigma_{OC} = 15.3$  and  $\sigma_{NO\ OC} = 18.2$  for each group
  - $\alpha$ -level of the hypothesis test (.05)

66

---

---

---

---

---

---

---

---

## Calculating Power

### ■ Using *sampsi* in Stata

```
. sampsi 132.8 127.4, sd1(15.3) sd2(18.2) n1(8) n2(21)

Estimated power for two-sample comparison of means

Test Ho: m1 = m2, where m1 is the mean in population 1
                    and m2 is the mean in population 2

Assumptions:

      alpha = 0.0500 (two-sided)
      m1 = 132.8
      m2 = 127.4
      sd1 = 15.3
      sd2 = 18.2
sample size n1 = 8
              n2 = 21
              n2/n1 = 2.63

Estimated power:

      power = 0.1268
```

67

---

---

---

---

---

---

---

---

## Calculating Power

- In order to calculate power for a study comparing two population proportions, we need :
  - Sample size for each group
  - Estimated (population) proportions,  $p_1$  and  $p_2$  for each group : these values frame a specific alternative hypothesis (it usually is the minimum difference of scientific interest)
  - $\alpha$ -level of the hypothesis test

68

---

---

---

---

---

---

---

---

## Calculating Power

### ■ Ulcer Drug/Healing Example

	Healed	Not Healed	Total
Drug A	23	7	30
Drug B	18	13	31

In this study:

$$\hat{p}_{\text{DRUG A}} = \frac{23}{30} \approx .77$$

$$\hat{p}_{\text{DRUG B}} = \frac{18}{31} \approx .58$$

69

---

---

---

---

---

---

---

---

## Calculating Power

- Fill in information below with results from this study
  - Sample size for each group (  $n_{DRUG\ A} = 30$ ,  $n_{DRUG\ B} = 31$  )
  - Estimated (population) proportions,  $p_{DRUG\ A} = .77$  and  $p_{DRUG\ B} = .58$
  - $\alpha$ -level of the hypothesis test (.05)

70

---

---

---

---

---

---

---

---

## Calculating Power

- Using the sampsi command
  - `sampsi  $p_1$   $p_2$ ,  $n1(n_1)$   $n2(n_2)$   $\alpha(\alpha)$`

```
. sampsi .77 .58, n1(30) n2(31) alpha(.05)
```

Estimated power for two-sample comparison of proportions

Test Ho:  $p_1 = p_2$ , where  $p_1$  is the proportion in population 1  
and  $p_2$  is the proportion in population 2

Assumptions:

```
alpha = 0.0500 (two-sided)
p1 = 0.7700
p2 = 0.5800
sample size n1 = 30
n2 = 31
n2/n1 = 1.03
```

Estimated power:

```
power = 0.2525
```

Note: For the above sample size(s) and proportion(s), the normal approximation to the binomial may not be very accurate. Thus, power calculations are questionable.

71

---

---

---

---

---

---

---

---

## Section E

FYI if Interested

---

---

---

---

---

---

---

---



### Example

- Consider the following results from a study done on 29 women, all 35-39 years old

Sample Data			
	n	Mean SBP	SD of SBP
OC users	8	132.8	15.3
Non-OC Users	21	127.4	18.2

73

---

---

---

---

---

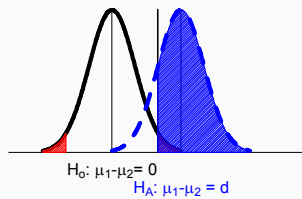
---

---

---

### Example

- Suppose we want to design a study with 80% power to detect a mean difference of at least 5.4 mmHg between the two group
  - will reject at  $\alpha=0.05$  if  $\left| \frac{\bar{x}_{OC} - \bar{x}_{NOOC}}{SE(\bar{x}_{OC} - \bar{x}_{NOOC})} \right| \geq 2$
  - as such, want  $\Pr\left(\left| \frac{\bar{x}_{OC} - \bar{x}_{NOOC}}{SE(\bar{x}_{OC} - \bar{x}_{NOOC})} \right| \geq 2\right) = 0.80$  if  $\mu_{OC} - \mu_{NOOC} \geq 5.4 \text{ mmHg}$



74

---

---

---

---

---

---

---

---

### Example

- Consider  $SE(\bar{x}_{OC} - \bar{x}_{NOOC})$

Using the estimates from the small study for population SDs:

$$SE(\bar{x}_{OC} - \bar{x}_{NOOC}) = \sqrt{\frac{\sigma_{OC}^2}{n_{OC}} + \frac{\sigma_{NOOC}^2}{n_{NOOC}}}$$

with  $n_{OC} = n_{NOOC} = n$  this becomes:

$$SE(\bar{x}_{OC} - \bar{x}_{NOOC}) = \sqrt{\frac{\sigma_{OC}^2}{n} + \frac{\sigma_{NOOC}^2}{n}} = \sqrt{\frac{1}{n}(\sigma_{OC}^2 + \sigma_{NOOC}^2)}$$

75

---

---

---

---

---

---

---

---

### Example

- Suppose we want to design a study with 80% power to detect a mean difference of at least 5.4 mmHg between the two groups

$$\text{i.e. } \Pr\left(\left|\frac{\bar{x}_{OC} - \bar{x}_{NOOC}}{\sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}}\right| \geq 2\right) = 0.80 \text{ if } \mu_{OC} - \mu_{NOOC} \geq 5.4 \text{ mmHg}$$

With some algebra:

$$\Pr\left(|\bar{x}_{OC} - \bar{x}_{NOOC}| \geq 2 \times \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}\right) = 0.80 \text{ if } \mu_{OC} - \mu_{NOOC} \geq 5.4 \text{ mmHg}$$

But if  $\mu_{OC} - \mu_{NOOC} \geq 5.4 \text{ mmHg}$ , then assuming large  $n$ ,  $|\bar{x}_{OC} - \bar{x}_{NOOC}|$  is a normally

distributed process with mean  $\mu_{OC} - \mu_{NOOC}$  and standard error

$$SE(\bar{x}_{OC} - \bar{x}_{NOOC}) = \sqrt{\frac{\sigma^2_{OC}}{n} + \frac{\sigma^2_{NOOC}}{n}} = \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}$$

76

---

---

---

---

---

---

---

---

### Example

- Suppose we want to design a study with 80% power to detect a mean difference of at least 5.4 mmHg between the two groups

$$\text{So } \Pr\left(|\bar{x}_{OC} - \bar{x}_{NOOC}| \geq 2 \times \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}\right) = 0.80 \text{ if } \mu_{OC} - \mu_{NOOC} \geq 5.4 \text{ mmHg}$$

Becomes:

$$\Pr\left(\left|\frac{\bar{x}_{OC} - \bar{x}_{NOOC}}{\sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}}\right| \geq \frac{2 \times \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})} - (\mu_{OC} - \mu_{NOOC})}{\sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}}\right) = 0.8$$

77

---

---

---

---

---

---

---

---

### Example

- Suppose we want to design a study with 80% power to detect a mean difference of at least 5.4 mmHg between the two groups

But on a standard normal curve, the value that cuts off 80% of the area to its right is 0.84.

So we need to solve:

$$\frac{2 \times \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})} - (\mu_{OC} - \mu_{NOOC})}{\sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}} = -0.84$$

Some more beautiful algebra:

$$2 \times \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})} - (\mu_{OC} - \mu_{NOOC}) = -0.84 \times \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOOC})}$$

78

---

---

---

---

---

---

---

---

### Example

- Suppose we want to design a study with 80% power to detect a mean difference of at least 5.4 mmHg between the two groups

Some more beautiful algebra:

$$(2 + 0.84) \times \sqrt{\frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOC})} = (\mu_{OC} - \mu_{NOC})$$

squaring both sides:

$$(2 + 0.84)^2 \times \frac{1}{n}(\sigma^2_{OC} + \sigma^2_{NOC}) = (\mu_{OC} - \mu_{NOC})^2$$

$$\frac{(2 + 0.84)^2 \times (\sigma^2_{OC} + \sigma^2_{NOC})}{(\mu_{OC} - \mu_{NOC})^2} = n$$

79

---

---

---

---

---

---

---

---

### Example

- Suppose we want to design a study with 80% power to detect a mean difference of at least 5.4 mmHg between the two groups

Plugging on our info:

$$\frac{(2 + 0.84)^2 \times (15.3^2 + 18.2^2)}{(5.4)^2} = n$$

$$\frac{8.1 \times 565}{29.2} = n$$

$$157 \approx n$$

80

---

---

---

---

---

---

---

---

### Example

- It is also possible to design a study to estimate a quantity, such as a mean difference or difference in proportions with a desired level of precision
- In other word, the necessary sample sizes can be estimated to try to get a confidence interval of a desired maximum width

81

---

---

---

---

---

---

---

---

### Example

- Suppose we wanted to design a study with equal sample sizes to estimate the mean difference within  $\pm 3$  mmHg
  - i.e. design the study to have a specific precision

Now using the estimates from the small study for population SDs:

$$SE(\bar{x}_{OC} - \bar{x}_{NO OC}) = \sqrt{\frac{\sigma_{OC}^2}{n_{OC}} + \frac{\sigma_{NO OC}^2}{n_{NO OC}}}$$

with  $n_{OC} = n_{NO OC} = n$  this becomes:

$$SE(\bar{x}_{OC} - \bar{x}_{NO OC}) = \sqrt{\frac{\sigma_{OC}^2}{n} + \frac{\sigma_{NO OC}^2}{n}} = \sqrt{\frac{1}{n}(\sigma_{OC}^2 + \sigma_{NO OC}^2)}$$

82

---

---

---

---

---

---

---

---

### Example

- We want

$$2 \times SE(\bar{x}_{OC} - \bar{x}_{NO OC}) = 3$$

$$2 \times \sqrt{\frac{1}{n}(\sigma_{OC}^2 + \sigma_{NO OC}^2)} = 3$$

plugging in pilot data:

$$2 \times \sqrt{\frac{1}{n}(15.3^2 + 18.2^2)} = 3$$

Sample Data			
	n	Mean SBP	SD of SBP
OC users	8	132.8	15.3
Non-OC Users	21	127.4	18.2

83

---

---

---

---

---

---

---

---

### Example

- Solving algebraically

$$2 \times \sqrt{\frac{1}{n}(15.3^2 + 18.2^2)} = 3$$

$$\sqrt{\frac{1}{n}(15.3^2 + 18.2^2)} = \frac{3}{2}$$

$$\frac{1}{n}(15.3^2 + 18.2^2) = \left(\frac{3}{2}\right)^2$$

$$\frac{(15.3^2 + 18.2^2)}{\left(\frac{3}{2}\right)^2} = n$$

$$n \approx 251$$

84

---

---

---

---

---

---

---

---

### Example

- Solving algebraically

$$2 \times \sqrt{\frac{1}{n}(15.3^2 + 18.2^2)} = 3$$

$$\sqrt{\frac{1}{n}(15.3^2 + 18.2^2)} = \frac{3}{2}$$

$$\frac{1}{n}(15.3^2 + 18.2^2) = \left(\frac{3}{2}\right)^2$$

$$\frac{(15.3^2 + 18.2^2)}{\left(\frac{3}{2}\right)^2} = n$$

$$n \approx 251$$

85

---

---

---

---

---

---

---