

Introduction to Experiments and Power Analysis (實驗計量與統計檢定力分析)

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Experiments Lecture 1 (實驗計量第一講)

Outline: The Replication Size Trinity

1. **Sample Size n** : # of observations/subjects
2. **Effect Size**: How big is the true result
3. **Power $(1-\beta)$** : How likely will your test show significance if there is truly an effect

Why Do We Care About This?

- ▶ Editor's Preface ([JEEA 2015](#)):
 - ▶ A necessary (but not sufficient) condition for publishing a replication study or null result
 - ▶ will be the presentation of **power calculations**.
- ▶ **Test Resolution**: $\Pr(\text{confirm} \mid \text{infected patient})$
 - ▶ In 2020, Taiwan requires 3 consecutive negatives to discharge for COVID-19, since even PCR has insufficient power (around 70%)...
- ▶ But what about structural estimation?

Key Concepts and Definitions

- ▶ Treatment Test:
 - ▶ Null ($H_0 : \theta = \theta_0$) Hypothesis - No Effect!
 - ▶ Alternative ($H_1 : \theta = \theta_1$) Hypothesis - Effective!
- ▶ **Effect Size** ($\theta_1 - \theta_0$): True size of effect
- ▶ Alternative Hypothesis can be **Directional**:
 - ▶ One-sided Alternative - **One-tailed** test
 - ▶ Usually comes from **prior beliefs** based on theory
 - ▶ Two-sided Alternative - **Two-tailed** test

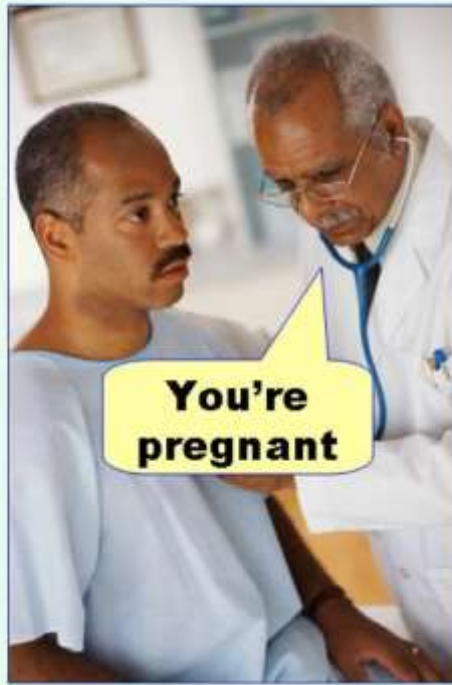
Key Concepts and Definitions

- ▶ Two Stages of the Treatment Test:
 1. Compute Test Statistic of sample size n
 2. Compare Test Statistic with null distribution
- ▶ Rejection Region = Tail of null distribution
 - ▶ of a Size $\alpha = \Pr(\text{reject null} \mid \text{null is true})$
 - ▶ Critical Value: Rejection region starting point
- ▶ p -value = $\Pr(|T| \geq T_{CV} \mid \text{null is true})$
 - ▶ $p < 0.05$ vs. $p < 0.01/0.001$ (strength of evidence)
 - ▶ Evidence vs. Strong/Overwhelming Evidence

Key Concepts and Definitions

- ▶ **Type 1 Error:** $\alpha = \Pr(\text{reject null} \mid \text{null is true})$

Type I error
(false positive)



- ▶ But what is Power?

Type II error
(false negative)



- ▶ **Type 2 Error:** $\beta = \Pr(\text{accept null} \mid \text{null is false})$

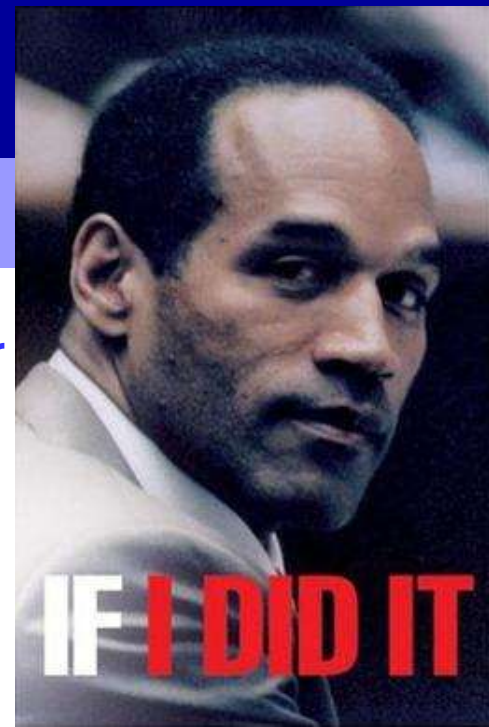
Key Concepts and Definitions

- ▶ **Type 1 Error:** $\alpha = \Pr(\text{reject null} \mid \text{null is true})$
- ▶ **Type 2 Error:** $\beta = \Pr(\text{accept null} \mid \text{null is false})$
- ▶ **Power**(π): $1 - \beta = \Pr(\text{reject null} \mid \text{null is false})$
 1. True effect size $\theta_1 - \theta_0$ (and one/two-tailed)
 2. Sample size n
 3. Size of the test α
- ▶ **Trade-off:** The higher α/n , the higher is π
 1. **Power Analysis:** Compute power $\pi = 1 - \beta$, or
 2. Find n to meet power requirement $\pi(n) \geq \bar{\pi}$

Choosing the Value of α

- ▶ How big can we allow **Type 1 Error**
- ▶ To convict a crime suspect,
 - ▶ Null Hypothesis: Not Guilty
 - ▶ Alternative Hypothesis: Guilty
 - ▶ **Type 1:** $\alpha = \Pr(\text{convict} \mid \text{innocent suspect})$
 - ▶ **Type 2:** $\beta = \Pr(\text{acquit} \mid \text{guilty suspect})$
- ▶ **Type 1 Error** more serious than **Type 2 Error**
 - ▶ Choose very low α at the expense of **power**:

$$1 - \beta = \Pr(\text{convict} \mid \text{guilty suspect})$$



Choosing the Value of α

- ▶ How big can we allow **Type 1 Error**
- ▶ To test for COVID-19,
 - ▶ Null Hypothesis: Healthy
 - ▶ Alternative Hypothesis: Infected by COVID-19
 - ▶ Type 1: $\alpha = \Pr(\text{confirm} \mid \text{healthy patient})$
 - ▶ Type 2: $\beta = \Pr(\text{discharge} \mid \text{infected patient})$
- ▶ **Type 2 Error** more serious than **Type 1 Error**
 - ▶ Choose a higher α so get higher power:
 $1 - \beta = \Pr(\text{confirm} \mid \text{infected patient})$



Choosing the Value of α

- ▶ Type 1 $\alpha = \Pr(\text{confirm} \mid \text{healthy patient})$
- ▶ Type 2 $\beta = \Pr(\text{discharge} \mid \text{infected patient})$
- ▶ Both errors not fatal in

Experimental
Economics,

- ▶ Convention is:

$$\alpha = 0.05$$

$$\pi = 1 - \beta = 0.80$$

$$\beta = 0.20$$

疾病篩檢結果

+

-

	+	-
+	<p>True Positive 真陽性</p> <p>病人真的生病， 檢驗也確實為陽性</p>	<p>False Positive 偽陽性</p> <p>病人沒有生病， 但檢驗結果為陽性</p> 
-	<p>False Negative 偽陰性</p> <p>病人真的生病， 檢驗結果卻為陰性</p> 	<p>True Negative 真陰性</p> <p>病人真的沒生病， 檢驗也確實為陰性</p>

Treatment Testing Toolkit

- ▶ One-sample t-test
 - ▶ Does $WTP = £3$ (= retail value of coffee mug)?
- ▶ Two-sample t-test (with equal variance)
 - ▶ If passes variance ratio test
 - ▶ Can be done using OLS!
- ▶ Two-sample t-test (with unequal variance)
 - ▶ If fails variance ratio test
 - ▶ Skewness-kurtosis test
- ▶ Need CLT: Okay if sufficiently large n ($\geq 30?$)

Treatment Testing Toolkit

- ▶ What if we do not have CLT/large n ?
 - ▶ Use non-parametric tests instead!
- ▶ Mann-Whitney Test (aka ranksum test)
 - ▶ Between-subject non-parametric treatment test
- ▶ Kolmogorov-Smirnov (KS) Test
- ▶ Epps-Singleton Test (discrete KS test)
 - ▶ Tests comparing entire distributions

Treatment Testing: WTP - WTA Gap

- ▶ What if we have within-subject data?
- ▶ Can use within-subject tests!
 - ▶ But, watch out for order effect...
- ▶ Paired t-test (assume CLT)
- ▶ Wilcoxon Signed Rank Test
 - ▶ Within-subject non-parametric treatment test
 - ▶ Assume symmetric distribution around median
 - ▶ (regarding paired difference). Without it, use:
- ▶ Paired-sample sign test

Treatment Testing: WTP - WTA Gap

- ▶ Isoni et al. (AER 2011)
 - ▶ Replicate Plott and Zeiler (AER 2007), which
 - ▶ Replicate Kahneman et al. (JPE 1990) (KKT)
- ▶ Measure WTP and/or WTA
 - ▶ Becker–DeGroot–Marschak (BDM) mechanism
 - ▶ 2nd price auction against (randomizing) computer
- ▶ Treatment Test:
 - ▶ Does WTP or WTA = £3 (= retail value of the coffee mug)?

Power Analysis: Theory

1. **Power Analysis:** Find test power $\pi = 1 - \beta$, or
 2. Find n to meet power requirement $\pi(n) \geq \bar{\pi}$
- ▶ **One-sample t-test**
 - ▶ Rarely used in experimental economics, but...
 - ▶ Isoni et al. (2011) test WTP of coffee mug = £3
 - ▶ Y : Continuous outcome measure with mean μ
 - ▶ Null Hypothesis: $H_0 : \mu = \mu_0$
 - ▶ Alternative Hypothesis: $H_1 : \mu = \mu_1 > \mu_0$
 - ▶ Collect data of sample size n

Power Analysis: Theory

1. What is the power of this test?
2. How big should sample size n be?

▶ Test Size $\alpha = 0.05 = \Pr(\text{reject null} \mid \text{null is true})$

▶ Type 2 $\beta = 0.20 = \Pr(\text{accept null} \mid \text{null is false})$

▶ Power $\pi = 1 - \beta = 0.80$

▶ One-sample t-test

\bar{y} = sample mean
 s^2 = sample variance

▶ Test Statistic: $t = \frac{\bar{y} - \mu_0}{s/\sqrt{n}} \sim t(n - 1)$

▶ Reject if $t > t_{n-1, \alpha}$ ($t > z_\alpha$ for large n)

Power Analysis: Power of the Test

$$\begin{aligned}\pi &= \Pr(t > z_\alpha | \mu = \mu_1) = \Pr\left(\frac{\bar{y} - \mu_0}{s/\sqrt{n}} > z_\alpha \mid \mu = \mu_1\right) \\ &= \Pr\left(\bar{y} > \mu_0 + z_\alpha(s/\sqrt{n}) \mid \mu = \mu_1\right) \quad \begin{array}{l} \mu_0 = 10 \\ \mu_1 = 12 \end{array} \\ &= \Pr\left(\frac{\bar{y} - \mu_1}{s/\sqrt{n}} > \frac{\mu_0 + z_\alpha(s/\sqrt{n}) - \mu_1}{s/\sqrt{n}} \mid \mu = \mu_1\right) \\ &= \Phi\left(\frac{12 - 10 - 1.645(5/\sqrt{30})}{5/\sqrt{30}}\right) \quad \begin{array}{l} z_\alpha = 1.645, \quad s = 5 \\ n = 30, \quad \alpha = 0.05 \end{array} = \underline{0.71}\end{aligned}$$

► What n is required to get $\pi = 0.80$?

Power Analysis: How Big Should n Be?

► Power $\pi = 1 - \beta = \Phi \left(\frac{\mu_1 - \mu_0 - z_\alpha (s/\sqrt{n})}{s/\sqrt{n}} \right)$

$$\Rightarrow z_\beta = \frac{\mu_1 - \mu_0 - z_\alpha (s/\sqrt{n})}{s/\sqrt{n}}$$

$$\alpha = 0.05, \beta = 0.20$$

$$\Rightarrow z_\beta + z_\alpha = \frac{\mu_1 - \mu_0}{s/\sqrt{n}}$$

$$z_\alpha = 1.645, z_\beta = 0.842$$

$$\Rightarrow n = \frac{s^2 (z_\alpha + z_\beta)^2}{(\mu_1 - \mu_0)^2} = \frac{5^2 (1.645 + 0.842)^2}{(12 - 10)^2}$$

$$s = 5$$

$$\mu_0 = 10$$

$$\mu_1 = 12$$

► So we need $n \geq 39$

$$= \underline{38.66}$$

Power Analysis: Power in STATA

- ▶ What is the power for sample size $n = 30$?
 - ▶ STATA command for power calculation

```
power onemean  $\mu_0/\mu_1$  10 12 , sd(5) n(30) oneside
```

sample std; sample size

▶ 1-sample t-test

one-tailed test

Power Analysis: Power Results in STATA

- ▶ What is the power for sample size $n = 30$?

```
power onemean 10 12 , sd(5) n(30) oneside
```

- ▶ STATA Results:

```
Estimated power for a one-sample mean test  
t test
```

```
Ho: m = m0 versus Ha: m > m0
```

```
Study parameters:
```

```
alpha = 0.0500  
N = 30  
delta = 0.4000  
m0 = 10.0000  
ma = 12.0000  
sd = 5.0000
```

```
Estimated power:
```

```
power = 0.6895
```

Slightly different
since STATA did
not use normal
approximation...

Power Analysis: Sample Size in STATA

- ▶ What is the sample size to get power $\pi = 0.80$?

- ▶ STATA command for power calculation

`power onemean μ_0/μ_1 10 12 , sd(5) oneside p(0.8)`

sample std

required power

- ▶ 1-sample t-test

- ▶ one-tailed test

Power Analysis: Sample Size Result/Stata

- ▶ What is the sample size to get power $\pi = 0.80$?

power onemean 10 12 , sd(5) onside p(0.8)

- ▶ STATA
Results:

```
Performing iteration ...  
  
Estimated sample size for a one-sample mean test  
t test  
Ho: m = m0 versus Ha: m > m0  
  
Study parameters:  
  
alpha = 0.0500  
power = 0.8000  
delta = 0.4000  
m0 = 10.0000  
ma = 12.0000  
sd = 5.0000  
  
Estimated sample size:  
  
N = 41
```

Slightly larger n
since STATA did
not use normal
approximation...

Power Analysis: Graph Power in STATA

- ▶ Plot power against sample size with `graph`
- ▶ STATA command for power calculation

$\mu_0/\mu_1 \in [10.5, 12.5]$

```
power onemean 10 (10.5(0.5)12.5), sd(5) n(20(10)200) oneseid graph
```

sample std; $n=20-200$

▶ 1-sample t-test

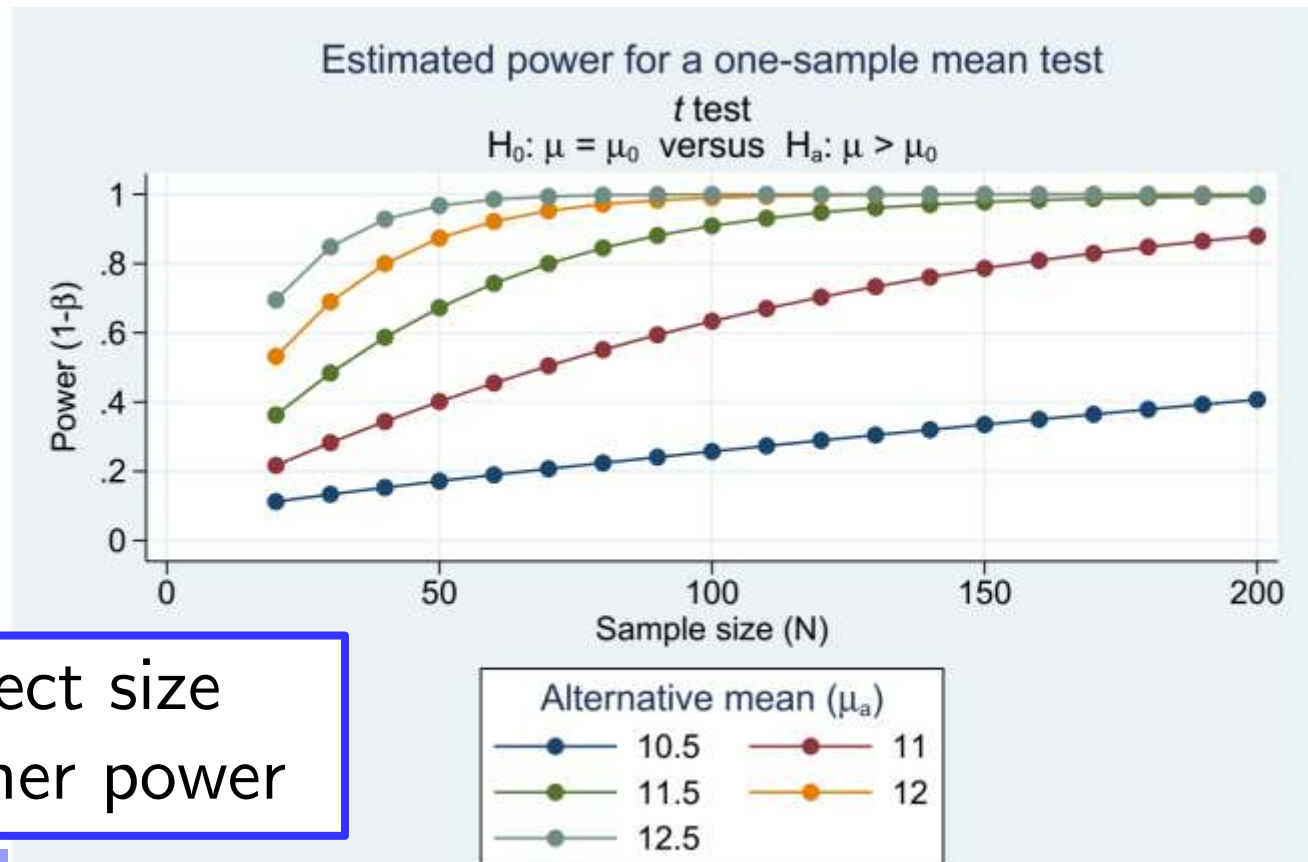
one-tailed test

Power Analysis: Graph Power in STATA

- ▶ Plot power against sample size with `graph`

```
power onemean 10 (10.5(0.5)12.5), sd(5) n(20(10)200) onside graph
```

- ▶ STATA Results:



Larger effect size
yields higher power

Parameters: $\alpha = .05$, $\mu_0 = 10$, $\sigma = 5$

Power Analysis: Graph Sample Size/Stata

- ▶ Plot sample size against effect size
- ▶ STATA command for power calculation

$\mu_0/\mu_1 \in [10.5, 12.5]$
power onemean 10 (10.5(0.25)12.5), sd(5) p(0.6(0.1)0.9) onесide graph

sample std;

power=0.6-0.9

▶ 1-sample t-test

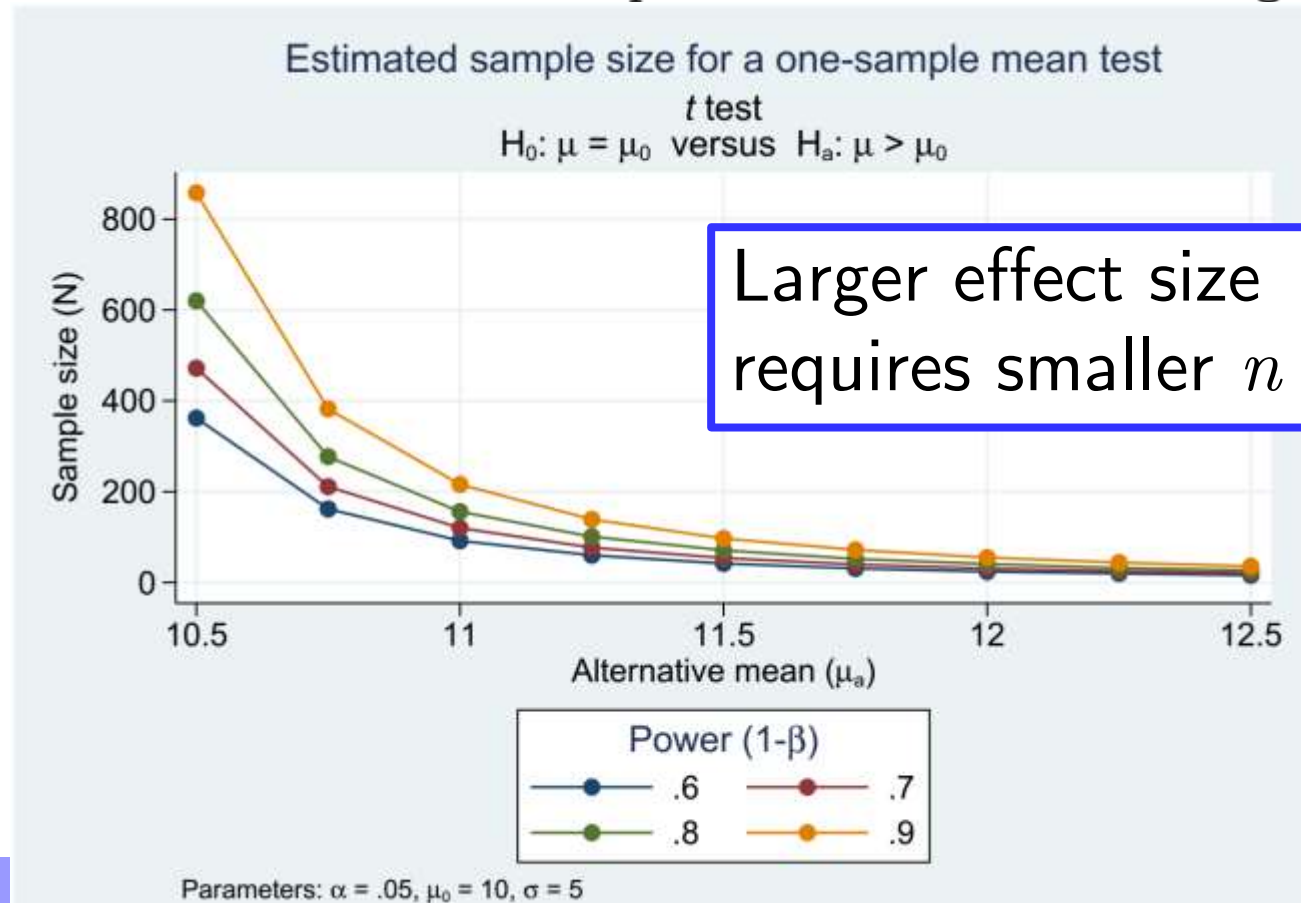
one-tailed test

Power Analysis: Graph Sample Size/Stata

- ▶ Plot sample size against effect size

```
power onemean 10 (10.5(0.25)12.5), sd(5) p(0.6(0.1)0.9) onside graph
```

- ▶ STATA Results:



Power Analysis: Two-sample t-test

1. **Power Analysis:** Find test power $\pi = 1 - \beta$, or
2. Find n to meet power requirement $\pi(n) \geq \bar{\pi}$

▶ Two-sample t-test

- ▶ More common in experimental economics...
- ▶ μ_1 : Population mean of control group
- ▶ μ_2 : Population mean of treatment group
 - ▶ Null Hypothesis: $H_0 : \mu_2 - \mu_1 = 0$
 - ▶ Alternative Hypothesis: $H_1 : \mu_2 - \mu_1 = d$
- ▶ Collect data of sample size n_1 and n_2

Effect Size
from prior

Power Analysis: Two-sample t-test

- ▶ Test Size $\alpha = 0.05$
- ▶ Type 2 $\beta = 0.20$
- ▶ Power $\pi = 1 - \beta = 0.80$

$\bar{y}_1, \bar{y}_2 =$ sample means
 $s_1^2, s_2^2 =$ sample variances

$$s_p = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}}$$

(pooled sample s.d. for equal variance)

▶ Test $t = \frac{\bar{y}_2 - \bar{y}_1}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim t(n_1 + n_2 - 2)$

▶ Statistic: $s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$

- ▶ Reject if $t > t_{n_1+n_2-2, \alpha}$ ($t > z_\alpha$ for large n)

Power Analysis: Two-sample t-test

▶ Equal sample size

$$n_1 = n_2 = n$$

$\bar{y}_1, \bar{y}_2 =$ sample means
 $s_1^2, s_2^2 =$ sample variances

▶ Pooled sample s.d. for equal variance is:

$$s_p = \sqrt{\frac{s_1^2 + s_2^2}{2}}$$

▶ Test $t = \frac{\bar{y}_2 - \bar{y}_1}{s_p \sqrt{\frac{2}{n}}} \sim t(2n - 2)$

▶ Statistic:

$$s_p \sqrt{\frac{2}{n}}$$

▶ Reject if $t > t_{2n-2, \alpha}$ ($t > z_\alpha$ for large n)

Power Analysis: Power of the Test

$$\begin{aligned}\pi &= \Pr(t > z_\alpha | \mu_2 - \mu_1 = d) \\ &= \Pr\left(\frac{\bar{y}_2 - \bar{y}_1}{s_p \sqrt{2/n}} > z_\alpha \mid \mu_2 - \mu_1 = d\right) \\ &= \Pr\left(\bar{y}_2 - \bar{y}_1 > z_\alpha s_p \sqrt{2/n} \mid \mu_2 - \mu_1 = d\right) \\ &= \Pr\left(\frac{\bar{y}_2 - \bar{y}_1 - d}{s_p \sqrt{2/n}} > \frac{z_\alpha s_p \sqrt{2/n} - d}{s_p \sqrt{2/n}} \mid \mu_2 - \mu_1 = d\right) \\ &= \Phi\left(\frac{d - z_\alpha s_p \sqrt{2/n}}{s_p \sqrt{2/n}}\right)\end{aligned}$$

Power Analysis: How Big Should n Be?

► Power $\pi = 1 - \beta = \Phi \left(\frac{d - z_\alpha s_p \sqrt{2/n}}{s_p \sqrt{2/n}} \right)$

$\Rightarrow z_\beta = \frac{d - z_\alpha s_p \sqrt{2/n}}{s_p \sqrt{2/n}}$ $\alpha = 0.05, \beta = 0.20$

$z_\alpha = 1.645, z_\beta = 0.842$

$\Rightarrow z_\beta + z_\alpha = \frac{d}{s_p \sqrt{2/n}}$

$s_1 = 4.0, s_2 = 5.84$

$s_p^2 = \frac{s_1^2 + s_2^2}{2} = 5.0^2$

$\Rightarrow n = \frac{2s_p^2(z_\alpha + z_\beta)^2}{d^2} = \frac{2(5^2)(1.645 + 0.842)^2}{2^2}$

$d = 2$

► So we need $n \geq 78$ $= \underline{77.32}$

Power Analysis: Sample Size in Stata

- ▶ What is the sample size to get power $\pi = 0.80$?
- ▶ STATA command for power calculation

`power twomeans 10 12 , sd1(4.0) sd2(5.84) oneside p(0.8)`

μ_0 / μ_1 2 sample std's

required power

▶ 2-sample t-test

one-tailed test

Power Analysis: Sample Size Result/Stata

- ▶ What is the sample size to get power $\pi = 0.80$?

```
power twomeans 10 12 , sd1(4.0) sd2(5.84) oneside p(0.8)
```

▶ STATA Results:

```
Performing iteration ...
```

```
Estimated sample sizes for a two-sample means test  
Satterthwaite's t test assuming unequal variances  
Ho: m2 = m1 versus Ha: m2 > m1
```

```
Study parameters:
```

```
alpha = 0.0500  
power = 0.8000  
delta = 2.0000  
m1 = 10.0000  
m2 = 12.0000  
sd1 = 4.0000  
sd2 = 5.8400
```

```
Estimated sample sizes:
```

```
N = 158  
N per group = 79
```

Slightly larger n
since STATA did
not use normal
approximation...

Power Analysis: Graph Power in STATA

- ▶ Plot power against sample size with `graph`
- ▶ STATA command for power calculation

```
power twomeans 10 12, sd1(4.0) sd2(5.84) n(20(10)200) oneseide graph
```

μ_0/μ_1

sample std; $n=20-200$

▶ 2-sample t-test

one-tailed test

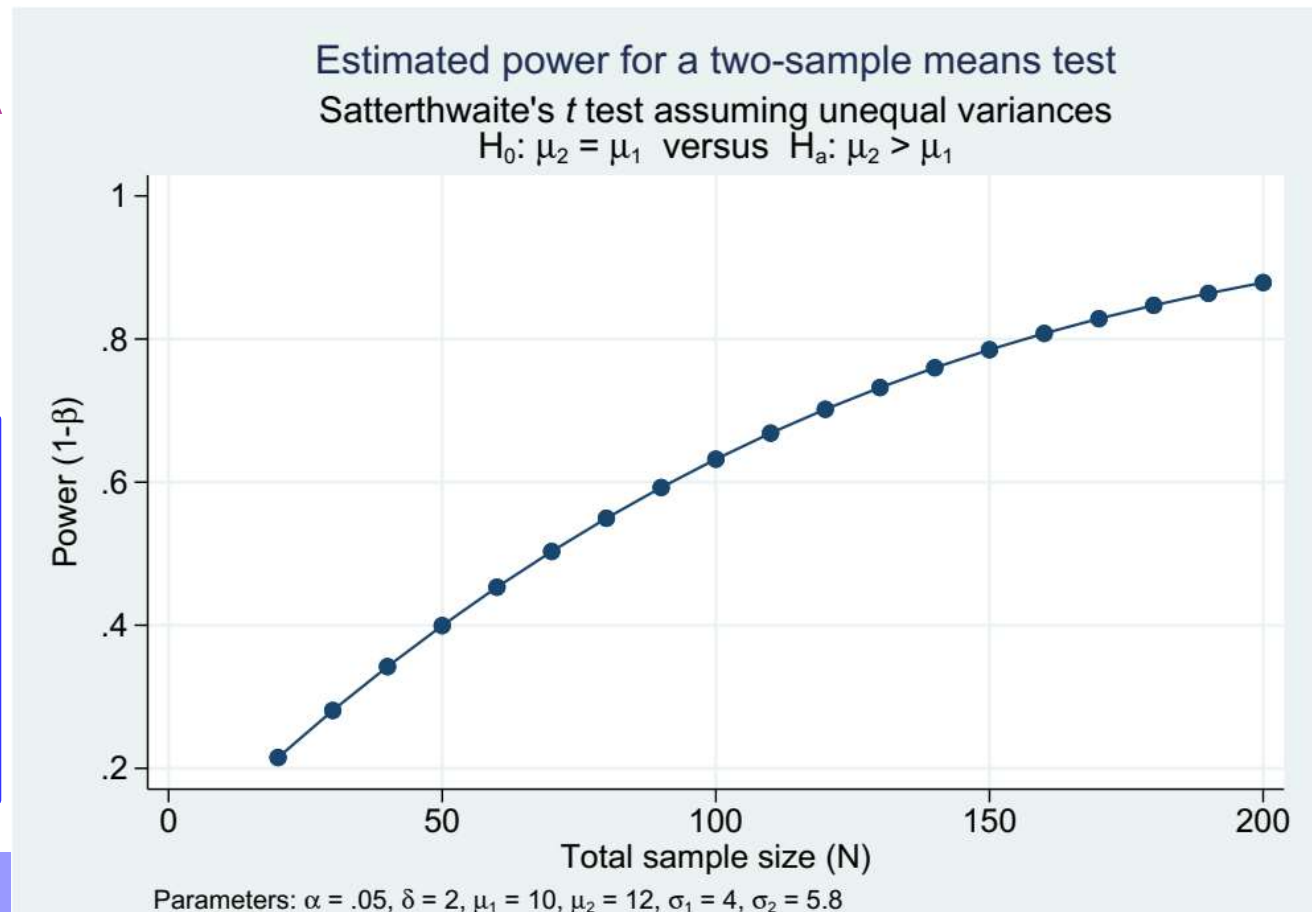
Power Analysis: Graph Power in STATA

- ▶ Plot power against sample size with `graph`

```
power twomeans 10 12, sd1(4.0) sd2(5.84) n(20(10)200) onside graph
```

- ▶ STATA Results:

Larger total sample size yields higher power



Power Analysis: Paired Sample t-test

1. **Power Analysis:** Find test power $\pi = 1 - \beta$, or
2. Find n to meet power requirement $\pi(n) \geq \bar{\pi}$

▶ Paired Sample t-test

- ▶ Observe subjects twice (with vs. w/o treatment)...

▶ μ_1 : Population mean without treatment

▶ μ_2 : Population mean with treatment

Effect Size
from prior

- ▶ Null Hypothesis: $H_0 : \mu_2 - \mu_1 = 0$

- ▶ Alternative Hypothesis: $H_1 : \mu_2 - \mu_1 = d$

▶ Collect data of sample size n (HW: Theory?!)

Power Analysis: Paired Sample t-test

- ▶ What is the sample size to get power $\pi = 0.80$?
- ▶ STATA command for power calculation

μ_0 / μ_1
power pairedmeans 10 12 , sd1(4.0) sd2(5.84) corr(0) onесide

2 sample std's & its correlation

▶ Paired Sample t-test

one-tailed test

Power Analysis: Paired Sample t-test

- ▶ What is the sample size to get power $\pi = 0.80$?

power pairedmeans 10 12 , sd1(4.0) sd2(5.84) corr(0) oneside

▶ STATA Results:

```
Performing iteration ...

Estimated sample size for a two-sample paired-means test
Paired t test
Ho: d = d0 versus Ha: d > d0

Study parameters:

      alpha =    0.0500          ma1 =    10.0000
      power =    0.8000          ma2 =    12.0000
      delta =    0.2825          sd1 =     4.0000
      d0 =     0.0000          sd2 =     5.8400
      da =     2.0000          corr =     0.0000
      sd_d =     7.0785

Estimated sample size:

      N =          79
```

Same n per group as
in two-sample t-test!

Power Analysis: Paired Sample t-test

- ▶ What if subjects' two responses are correlated?
 - ▶ STATA command for power calculation

`power pairedmeans 10 12 , sd1(4.0) sd2(5.84) corr(0.5) onесide`

μ_0 / μ_1 2 sample std's

2 sample's correlation

one-tailed test

- ▶ Paired Sample t-test

Power Analysis: Paired Sample t-test

- ▶ What if subjects' two responses are correlated?

```
power pairedmeans 10 12 , sd1(4.0) sd2(5.84) corr(0.5) oneside
```

▶ STATA
Results:

- ▶ What if
corr. = 1?
- ▶ $n=1$ ok!

```
Performing iteration ...
```

```
Estimated sample size for a two-sample paired-means test
```

```
Paired t test
```

```
Ho: d = d0 versus Ha: d > d0
```

```
Study parameters:
```

alpha =	0.0500	ma1 =	10.0000
power =	0.8000	ma2 =	12.0000
delta =	0.3867		
d0 =	0.0000		
da =	2.0000		
sd_d =	5.1716		

```
Estimated sample size:
```

```
N = 43
```

Paired data positively correlated (from same subjects) require less n

Power Analysis: Paired Sample t-test

- ▶ Paired Tests saves number of subjects!
 - ▶ Since the two responses are highly correlated
- ▶ But can cause **Order Effect**
 - ▶ AB and BA yield different results
- ▶ **Crossover Design**
 - ▶ Half AB vs. Half BA
 - ▶ Measures Order Effects (if any)
 - ▶ Can later control with order dummies, etc.

Real Example: WTP - WTA Gap

- ▶ Isoni et al. (AER 2011)
 - ▶ Replicate Plott and Zeiler (AER 2007), which
 - ▶ Replicate Kahneman et al. (JPE 1990) (KKT)
- ▶ Measure WTP and/or WTA
 - ▶ Becker–DeGroot–Marschak (BDM) mechanism
 - ▶ 2nd price auction against (randomizing) computer
- ▶ Treatment Test:
 - ▶ Does WTP or WTA = £3 (= retail value of the coffee mug)?

WTP - WTA Gap: Summary Statistics

▶ Summary Statistics of N=100

```
summ v_mug
```

▶ STATA Results:

Variable	Obs	Mean	Std. Dev.	Min	Max
v_mug	100	2.0415	1.571287	0	7.5

Is the population mean value of a mug = 3.0?

WTP - WTA Gap: One-sample t-test

▶ Test $H_0 : \mu = 3.0$

```
ttest v_mug=3.0
```

▶ STATA
Results:

```
One-sample t test
-----+-----
Variable |      Obs       Mean   Std. Err.   Std. Dev.   [95% Conf. Interval]
-----+-----
  v_mug |      100    2.0415   .1571287   1.571287   1.729723   2.353277
-----+-----
      mean = mean(v_mug)                                t =  -6.1001
Ho: mean = 3.0                                         degrees of freedom =      99

      Ha: mean < 3.0                                Ha: mean != 3.0                                Ha: mean > 3.0
Pr(T < t) = 0.0000                                Pr(|T| > |t|) = 0.0000                                Pr(T > t) = 1.0000
```

Power Analysis: One-sample t-test

- ▶ What is the **power** of this test?

```
power onemean 3.0 2.04, n(100) sd(1.571)
```

- ▶ STATA
Results:

```
Estimated power for a one-sample mean test
t test
Ho: m = m0 versus Ha: m != m0

Study parameters:

      alpha =    0.0500
        N =     100
      delta =   -0.6111
        m0 =    3.0000
        ma =    2.0400
        sd =    1.5710

Estimated power:

      power =    1.0000
```

Power = 1?
Not surprising
if mean=2.04
& s.d.=1.571

Power Analysis: One-sample t-test

- ▶ What is the **sample size** to get power $\pi = 0.80$?

power onemean 3.0 2.04, **p(0.80)** sd(1.571)

- ▶ STATA
Results:

```
Estimated sample size for a one-sample mean test  
t test
```

```
Ho: m = m0 versus Ha: m != m0
```

```
Study parameters:
```

```
alpha = 0.0500  
power = 0.8000  
delta = -0.6111  
m0 = 3.0000  
ma = 2.0400  
sd = 1.5710
```

```
Estimated sample size:
```

```
N = 24
```

$n=24$ is enough!

Power Analysis: Two-sample t-test

- ▶ Is there sufficient **power** to test WTA vs. WTP?

```
power twomeans 1.86 2.21 , n1(49) n2(51) sd1(1.29) sd2(1.80) oneside
```

▶ STATA
Results:

▶ Power < 0.3!!

```
Estimated power for a two-sample means test  
Satterthwaite's t test assuming unequal variances  
Ho: m2 = m1 versus Ha: m2 > m1
```

```
Study parameters:
```

```
alpha = 0.0500  
N = 100  
N1 = 49  
N2 = 51  
N2/N1 = 1.0408  
delta = 0.3500  
m1 = 1.8600  
m2 = 2.2100  
sd1 = 1.2900  
sd2 = 1.8000
```

```
Estimated power:
```

```
power = 0.2973
```

Power =
Probability of
rejecting null
hypothesis

Power Analysis: Two-sample t-test

- ▶ What is the **sample size** to get power $\pi = 0.80$?

power twomeans 1.86 2.21 , sd1(1.29) sd2(1.80) oneside **p(0.8)**

- ▶ STATA
Results:

```
Performing iteration ...
```

```
Estimated sample sizes for a two-sample means test  
Satterthwaite's t test assuming unequal variances  
Ho: m2 = m1 versus Ha: m2 > m1
```

```
Study parameters:
```

```
alpha = 0.0500  
power = 0.8000  
delta = 0.3500  
m1 = 1.8600  
m2 = 2.2100  
sd1 = 1.2900  
sd2 = 1.8000
```

```
Estimated sample sizes:
```

```
N = 498  
N per group = 249
```

Need really
large n due to
tiny effect size
assumed

Power Analysis: Two-sample t-test

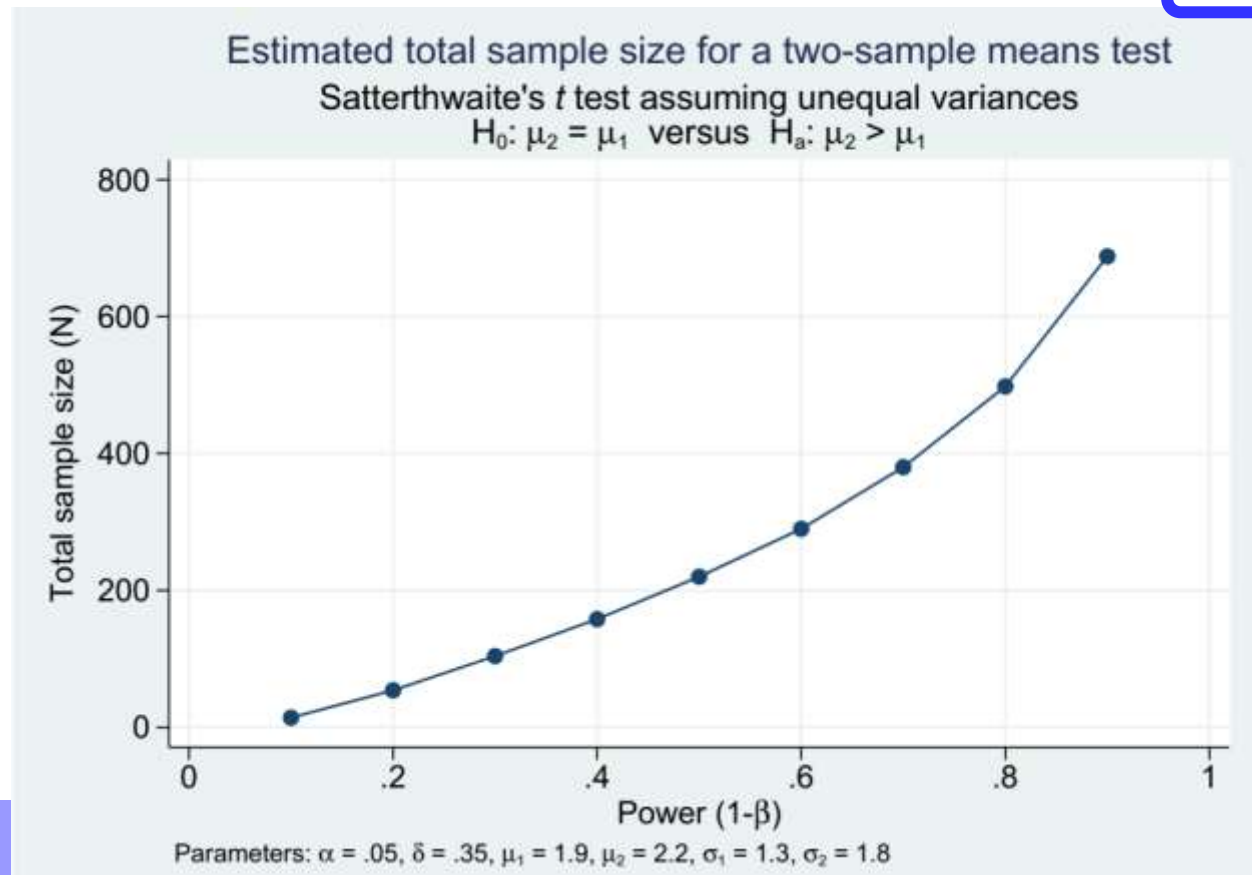
- ▶ Plot power against sample size with `graph`

```
power twomeans 1.86 2.21 , sd1(1.29) sd2(1.80) oneside
```

```
power(0.1(0.1)0.9) graph
```

- ▶ STATA
Results:

Larger total
sample size
yields higher
power



Power Analysis: Equality of Variance Test

▶ Test $H_0 : \sigma_1^2 = \sigma_2^2$

`sdtest v_mug, by(v_type)`

▶ STATA Results:

Significant difference b/w
var(WTP) and var(WTA)!

Variance ratio test

Group	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
WTP	49	1.862245	.184379	1.290653	1.491526	2.232964
WTA	51	2.213725	.2515679	1.796554	1.708436	2.719015
combined	100	2.0415	.1571287	1.571287	1.729723	2.353277

ratio = sd(WTP) / sd(WTA) f = 0.5161
Ho: ratio = 1 degrees of freedom = 48, 50

Ha: ratio < 1
Pr(F < f) = 0.0114

Ha: ratio != 1
2*Pr(F < f) = 0.0229

Ha: ratio > 1
Pr(F > f) = 0.9886

Power Analysis: Equality of Variance Test

- ▶ What is the **power** of this test?

```
power twovariances 1.66 3.24, n1(49) n2(51)
```

- ▶ STATA
Results:

```
Estimated power for a two-sample variances test
F test
Ho: v2 = v1 versus Ha: v2 != v1

Study parameters:

      alpha =      0.0500
       N =      100
      N1 =       49
      N2 =       51
  N2/N1 =     1.0408
   delta =     1.9518
      v1 =     1.6600
      v2 =     3.2400

Estimated power:

      power =     0.6402
```

$$1.29^2 = 1.66,$$
$$1.80^2 = 3.24$$

Larger sample
needed to get
desired power
of 0.8...

Power Analysis: Equality of Variance Test

- ▶ What is the **sample size** to get power $\pi = 0.80$?

power twovariances 1.66 3.24, $p(0.8)$

▶ STATA

Results:

```
Performing iteration ...
```

```
Estimated sample sizes for a two-sample variances test
```

```
F test
```

```
Ho: v2 = v1 versus Ha: v2 != v1
```

```
Study parameters:
```

```
alpha = 0.0500
```

```
power = 0.8000
```

```
delta = 1.9518
```

```
v1 = 1.6600
```

```
v2 = 3.2400
```

```
Estimated sample sizes:
```

```
N = 146
```

```
N per group = 73
```

Need $n = 73!$

Power Analysis: Paired Sample t-test

- ▶ Is there sufficient **power** to test WTA vs. WTP?

power pairedmeans **2.57 2.24**, sd1(0.96) sd2(1.11) corr(0)
n(100) oneside

- ▶ STATA
Results:

```
Estimated power for a two-sample paired-means test  
Paired t test  
Ho: d = d0 versus Ha: d < d0
```

```
Study parameters:
```

```
alpha = 0.0500          ma1 = 2.5700  
N = 100                ma2 = 2.2400  
delta = -0.2249        sd1 = 0.9600  
d0 = 0.0000           sd2 = 1.1100  
da = -0.3300          corr = 0.0000  
sd_d = 1.4675
```

```
Estimated power:
```

```
power = 0.7219
```

Almost 0.8, but not yet!

Power Analysis: Paired Sample t-test

▶ What is the **sample size** to get power $\pi = 0.80$?

power pairedmeans 2.57 2.24 , sd1(0.96) sd2(1.11)

`corr(0)` p(0.80) onside

▶ STATA
Results:

```
Performing iteration ...

Estimated sample size for a two-sample paired-means test
Paired t test
Ho: d = d0 versus Ha: d < d0
Study parameters:

      alpha =    0.0500          ma1 =    2.5700
      power =    0.8000          ma2 =    2.2400
      delta =   -0.2249          sd1 =    0.9600
      d0 =      0.0000          sd2 =    1.1100
      da =     -0.3300          corr =    0.0000
      sd_d =    1.4675

Estimated sample size:

      N =          124
```

$n = 124$ is
slightly
above 100

Power Analysis: Paired Sample t-test

- ▶ Is correlation truly 0?

```
corr WTA WTP
```

- ▶ STATA Results:

```
(obs=100)
```

	WTA	WTP
WTA	1.0000	
WTP	0.0987	1.0000

Should set correlation = +0.1...

Power Analysis: Paired Sample t-test

- ▶ What is the **sample size** to get power $\pi = 0.80$?

power pairedmeans 2.57 2.24 , sd1(0.96) sd2(1.11)

corr(0.10) p(0.80) onside

- ▶ STATA
Results:

```
Performing iteration ...
```

```
Estimated sample size for a two-sample paired-means test
```

```
Paired t test
```

```
Ho: d = d0 versus Ha: d < d0
```

```
Study parameters:
```

alpha =	0.0500	ma1 =	2.5700
power =	0.8000	ma2 =	2.2400
delta =	-0.2369	sd1 =	0.9600
d0 =	0.0000	sd2 =	1.1100
da =	-0.3300	corr =	0.1000
sd_d =	1.3930		

```
Estimated sample size:
```

```
N = 112
```

Only need
 $n = 112!$

Acknowledgment

- ▶ This presentation is based on
 - ▶ Section 1.1-1.4 of the lecture notes on Experimentetrics,
- ▶ prepared for a mini-course taught by Peter G. Moffatt (UEA) at National Taiwan University in Spring 2019