

# Power Analysis with Monte Carlo 使用蒙地卡羅法 進行統計檢定力分析

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Experimentics Lecture 2 (實驗計量第二講)

# Power Analysis for Another Test?

- ▶ STATA has the `power` command for pre-set tests, but what if I want to run another test?
  - ▶ Use **Monte Carlo** to perform power calculation!
- ▶ Can do **Treatment vs. Control** by comparing:
  1. 2 means: Two-sample t-test
  2. 2 medians: Mann-Whitney Test
  3. 2 distributions: Kolmogorov-Smirnov Test
- ▶ Which to use?
  - ▶ The one with **desired size** and **highest power**!

# DGP with Normally Distributed Errors

$$x_i = 10 + \underbrace{\delta \cdot d_i}_{\uparrow} + \epsilon_i, \quad i = 1, \dots, n = 100$$

▶ [Treatment Effect] × [Treatment dummy]

- ▶ Control:  $d_i = 0$  if  $i \leq \frac{n}{2} = 50$
- ▶ Treatment:  $d_i = 1$  if  $i > \frac{n}{2} = 50$
- ▶ Error:  $\epsilon_i \sim N(0, 1)$ ,  $E(\epsilon_i) = 0$ ,  $V(\epsilon_i) = 1$
- ▶ What is the size of each test?
  - ▶ % of resamples that “reject null | null is true”

# DGP with Normally Distributed Errors

- ▶ What is the **size** and **power** (at  $\delta = 0.5$ )?
  - ▶ `do-file_2.do`: Monte Carlo procedure
- ▶ Results of 1,000 replications are:

All three unbiased (properly sized)

	Size	Power
t-Test	0.052 <sup>u</sup>	0.702
M-W Test	0.053 <sup>u</sup>	0.683
K-S Test	0.040 <sup>u</sup>	0.513

High to Low...

u: Not significantly different from 0.05

# DGP with Normally Distributed Errors

- ▶ Same as power analysis of t-Test via STATA?
  - ▶ STATA command for power calculation

`power twomeans  $\mu_0/\mu_1$  10 10.5 , sd(1) n(100)`

sample std; sample size

- ▶ 2-sample t-test

# DGP with Normally Distributed Errors

- ▶ Same as power analysis of t-Test via STATA?

```
power twomeans 10 10.5 , sd(1) n(100)
```

- ▶ STATA Results:

```
Estimated power for a two-sample means test
t test assuming sd1 = sd2 = sd
Ho: m2 = m1 versus Ha: m2 != m1

Study parameters:

      alpha =      0.0500
        N =       100
N per group =       50
      delta =      0.5000
        m1 =     10.0000
        m2 =     10.5000
        sd =      1.0000

Estimated power:

      power =      0.6969
```

Very close to our Monte Carlo results of 0.702...

t-Test best due to Normality?

# DGP w/ Non-Normally Distributed Errors

$$x_i = 10 + \underbrace{\delta \cdot d_i}_{\uparrow} + \epsilon_i, \quad i = 1, \dots, n = 100$$

▶ [Treatment Effect] × [Treatment dummy]

- ▶ Control:  $d_i = 0$  if  $i \leq \frac{n}{2} = 50$
- ▶ Treatment:  $d_i = 1$  if  $i > \frac{n}{2} = 50$
- ▶ Error 1:  $\epsilon_i \sim \text{Uniform}[-2, 2]$ ,  $E(\epsilon_i) = 0$
- ▶ Error 2:  $\epsilon_i \sim \text{std } \chi^2(3)$  w/  $E(\epsilon_i) = 0$ ,  $V(\epsilon_i) = 1$
- ▶ What is the size and power (at  $\delta = 0.5$ )?

# DGP w/ Non-Normally Distributed Errors

- ▶ What is the **size** and **power** (at  $\delta = 0.5$ )?
- ▶ **Error 1:**  $\epsilon_i \sim \text{Uniform}[-2, 2]$ ,  $E(\epsilon_i) = 0$ 
  - ▶ Symmetric errors: Not skewed

	Size	Power
t-Test	0.056 <sup>u</sup>	0.566
M-W Test	0.056 <sup>u</sup>	0.526
K-S Test	0.039 <sup>u</sup>	0.306

All three unbiased (properly sized)

High to Low...

u: Not significantly different from 0.05



# DGP w/ Non-Normally Distributed Errors

- ▶ What is the **size** and **power** (at  $\delta = 0.5$ )?
- ▶ **Error 2:**  $\epsilon_i \sim \text{std } \chi^2(3)$  w/  $E(\epsilon_i) = 0, V(\epsilon_i) = 1$ 
  - ▶ Skewed error

	Size	Power
t-Test	0.061 <sup>u</sup>	0.705
M-W Test	0.067	0.867
K-S Test	0.052 <sup>u</sup>	0.862

M-W Test biased!

K-S test the best!

u: Not significantly different from 0.05

# Homework for Section 2.1

$$x_i = 10 + \underbrace{\delta \cdot d_i}_{\uparrow} + \epsilon_i, \quad i = 1, \dots, n = 100$$

▶ [Treatment Effect] × [Treatment dummy]

▶ What if skewed opposite like Error 3:

$$-\epsilon_i \sim \text{std } \chi^2(3) \text{ w/ } E(\epsilon_i) = 0, V(\epsilon_i) = 1$$

▶ Hint: Is M-W test better than K-S test here?

▶ Can we try the Epps-Singleton test?

▶ Hint: See `do-file_2a.do`

# Treatment Testing w/ Multi-Level Data

- ▶ Experimental data dependent at multi-levels:
  - ▶ Same Subject (with repeated observations)
  - ▶ Same Group (in interactive experiments)
  - ▶ Same Session (with re-matching of groups)
- ▶ How serious is ignoring these clustering?
  - ▶ `do-file_2b.do`: Use Monte Carlo to tell!
- ▶ Evaluate **Treatment Effect** with t-test for:
  - ▶ Between-Subject (Treat Half of the Subjects)
  - ▶ Within-Subject (Treat Half of the Tasks)

# Evaluate Treatment Effect with t-test in:

1. OLS (no clustering)
  2. OLS clustering at subject level
  3. OLS clustering at group level
  4. RE (no clustering)
  5. RE clustering at subject level
  6. RE clustering at group level
  7. Multi-Level Model (subject RE and group RE)
- ▶ Which are **correctly sized**?
- ▶ Among these, which has **highest power**?

# Treatment Testing with Multi-Level Data

Example:

40 Subjects of  
50 Rounds each  
(10 Groups of 4)

- ▶ Levels: Skrondal and Rabe-Hesketh (2004)
  - ▶ One-Level:  $T$  observations of a single subject
  - ▶ Two-Level:  $T$  observations for each of  $N$  subjects
  - ▶ Three-Level:  $T$  observations for each of  $N$  subjects in each of  $J$  groups:

$$y_{ij t} = \alpha + \delta d_i + \beta x_{ij t} + u_i + v_j + \epsilon_{ij t}$$

$$V(u_i) = \sigma_u, \quad V(v_j) = \sigma_v, \quad V(\epsilon_{ij t}) = \sigma_\epsilon$$
$$i = 1, \dots, n, \quad j = 1, \dots, J, \quad t = 1, \dots, T$$

- ▶ xtmixed for Subject RE + Group RE in STATA

# Example: Experimental Auction Data

$y_{ijt}$  : Bid of Subject  $i$  of Group  $j$  in Round  $t$

$x_{ijt}$  : Private Signal of Subject  $i$  of Group  $j$  in Round  $t$

$d_i$  : Treatment Dummy (like Auction Format)

$$y_{ijt} = \alpha + \delta d_i + \beta x_{ijt} + u_i + v_j + \epsilon_{ijt}$$

Example:

40 Subjects of  
50 Rounds each  
(10 Groups of 4)

▶ Three-Level Model:

▶  $u_i$  : Subject-specific RE

▶  $v_j$  : Group-specific RE

▶  $\epsilon_{ijt}$  : Observation-specific error

# RE: Special Case of Multi-Level Model

$y_{ijt}$  : Bid of Subject  $i$  ~~of Group  $j$~~  in Round  $t$

$x_{ijt}$  : Private Signal of Subject  $i$  ~~of Group  $j$~~  in Round  $t$

$d_i$  : Treatment Dummy (like Auction Format)

$$y_{ijt} = \alpha + \delta d_i + \beta x_{ijt} + u_i + \cancel{u_{ij}} + \epsilon_{ijt}$$

▶ Random Effect (RE) Model:

▶  $u_i$  : Subject-specific RE

▶  $e_{ijt}$  : Observation-specific error

# OLS: Special Case of RE Model

$y_{ijt}$  : Bid ~~of Subject  $i$  of Group  $j$~~  in Round  $t$

$x_{ijt}$  : Private Signal ~~of Subject  $i$  of Group  $j$~~  in Round  $t$

$d_i$  : Treatment Dummy (like Auction Format)

$$y_{ijt} = \alpha + \delta d_i + \beta x_{ijt} + \text{ ~~} \text{ ~~} + \epsilon_{ijt}~~~~$$

► Linear Regression (OLS) Model:

►  $e_{ijt}$  : Observation-specific error



# Between-Subject vs. Within-Subject Treatment Effects

$d_i$  : (Between-Subject) Treatment Dummy

$d_i = 0$  for Subject  $i = 1-20$

$d_i = 1$  for Subject  $i = 21-40$

▶  $d_{it}$  : Within-Subject Treatment Dummy

$$y_{ijt} = \alpha + \delta d_{it} + \beta x_{ijt} + u_i + v_j + \epsilon_{ijt}$$

▶ Three-Level Model:

▶  $u_i$  : Subject-specific RE

▶  $v_j$  : Group-specific RE

▶  $e_{ijt}$  : Observation-specific error

$d_{it} = 0$  for Round  $t = 1-25$

$d_{it} = 1$  for Round  $t = 26-50$

Example:

40 Subjects of  
50 Rounds each  
(10 Groups of 4)

# Multi-Level Models in STATA (Cluster at 1/2 Levels)

- ▶ **40 Subjects** of 50 Rounds each (10 Groups of 4)
- ▶ `egen i=seq(), f(1) b(50)` (or `egen i=seq(), from(1) by(50)`)
  - ▶ “from 1 by 50” means (1,...,1, 2,...,2, 3,...,3, 4,...,4, ...)
- ▶ `egen i=seq(), f(1) t(50)` (or `egen i=seq(), from(1) to(50)`)
  - ▶ “from(1) to(50)” means (1,2,3,4,...,50, 1,2,...,50, 1,2,...,50, ...)
- ▶ STATA Command:
  - ▶ OLS: Omitted (Review your Econometrics Class Notes!)
  - ▶ 1-Level: `xtmixed y d x || i:` **Cluster at Subject  $i$**
  - ▶ 2-Level: `xtmixed y d x || j: || i:` **Cluster at Group  $j$  and Subject  $i$**

# Three-Level Model Using STATA (Clustered at 2 Levels)

► STATA

```
xtmixed y d x || j: || i:
```

Results:

Performing EM optimization:

Cluster at Group  $j$  and Subject  $i$

Performing gradient-based optimization:

Iteration 0: log likelihood = -2959.3982

Iteration 1: log likelihood = -2959.3978

Iteration 2: log likelihood = -2959.3978

40 Subjects of  
50 Rounds each  
(10 Groups of 4)

Computing standard errors:

Mixed-effects ML regression

Number of obs = 2,000

Group Variable		No. of Groups	Observations per Group		
			Minimum	Average	Maximum
j		10	200	200.0	200
i		40	50	50.0	50

# Three-Level

## STATA Results:

40 Subjects of  
50 Rounds each  
(10 Groups of 4)

Error STD for  
Group  $j$  and  
Subject  $i$  and  
Residual  $e$

		Wald chi2(2) = 155.37		Prob > chi2 = 0.0000		
Log likelihood = -2959.3978						
<b><math>d</math> : Treatment increases bid by 0.148</b>						
y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
<b>d</b>	.1482739	.0454989	3.26	0.001	.0590978	.23745
<b>x</b>	.0955655	.0079035	12.09	0.000	.0800749	.111056
_cons	-.1241784	.247917	-0.50	0.616	-.6100867	.3617299
<b><math>x</math> : How values affect bids</b>						
Random-effects Parameters		Estimate	Std. Err.	[95% Conf. Interval]		
<b>j: Identity</b>	$\hat{\sigma}_v$ sd(_cons)	.4820359	.292011	.1470391	1.580251	
<b>i: Identity</b>	$\hat{\sigma}_u$ sd(_cons)	1.193918	.156372	.9236118	1.543333	
	$\hat{\sigma}_\epsilon$ sd(Residual)	1.017198	.0162466	.9858481	1.049544	
LR test vs. linear model: chi2(2) = 1737.24				Prob > chi2 = 0.0000		
Note: LR test is conservative and provided only for reference.						

# Between-Subject 100 Monte Carlo Results ( $\delta = 0.5$ )

Unbiased if cluster at group (not subject) level	Size: $d = 0$	Power: $\delta = 0.5$
OLS	0.46 <del>XXX</del>	<del>0.68</del>
OLS clustering at subject level	0.15 <del>X</del>	<del>0.41</del>
OLS clustering at <b>group level</b>	0.07 <sup>u</sup>	0.25
RE (no clustering)	0.13 <del>X</del>	<del>0.41</del>
RE clustering at subject level	0.15 <del>X</del>	<del>0.41</del>
RE clustering at <b>group level</b>	0.07 <sup>u</sup>	0.25
<b>Multi-Level</b> (subject and <b>group level</b> )	0.08 <sup>u</sup>	<b>0.27</b>

u: Not significantly different from 0.05

**Multi-Level highest (but low)**

# Within-Subject 100 Monte Carlo Results ( $\delta = 0.05$ )

	Size: $d = 0$	Power: $\delta = 0.05$
All 7 unbiased (with 100 replications)		
OLS	0.02 <sup>u</sup>	<del>0.07</del>
OLS clustering at subject level	0.09 <sup>u</sup>	0.31
OLS clustering at group level	0.09 <sup>u</sup>	0.33
RE (no clustering)	0.05 <sup>u</sup>	0.31
RE clustering at subject level	0.09 <sup>u</sup>	0.31
RE clustering at group level	0.08 <sup>u</sup>	0.33
Multi-Level (subject and group level)	0.05 <sup>u</sup>	0.31

u: Not significantly different from 0.05

No Cluster = Low Power

# Conclusion

- ▶ Between-Subject:
  - ▶ Size: Cluster at Highest Level possible
  - ▶ Power: Multi-Level model is best
- ▶ Between-Subject:
  - ▶ Size: All models able to detect small treatment
  - ▶ Power: All but OLS is good
- ▶ HW: What if we make group effect = 0.1 instead of 1?
  - ▶ Is size good now? `gen y=0.5+delta*d+0.1*x+u+0.1v+e`
  - ▶ What about power?

## Increase $n$ and $T$ of Between-Subject Multi-Level Model

- ▶ Multi-Level best with  $n=40$  Subjects of  $T=50$  Rounds each
- ▶ How to increase **power** of Multi-Level with  $n$  and  $T$  ?
  - ▶ `do-file_2c.do`: Monte Carlo procedure
  - ▶ Typo: “`1`” in wrong place for STATA command `gen d=i/2`
- ▶ Double or Triple  $n$  and/or  $T$  for:
  - ▶ Between-Subject at  $\delta = 0.5$
  - ▶ Within-Subject at  $\delta = 0.05$



# Increase $n$ and $T$ of Between-Subject Multi-Level Model

- ▶ Double or Triple  $n$  and/or  $T$  for:
  - ▶ Between-Subject at  $\delta = 0.5$

Multi-Level	$T = 50$	$T = 100$	$T = 150$
$n = 40$	0.24	0.26	0.28
$n = 80$	0.25	0.36	0.35
$n = 120$	0.39	0.38	0.35

Modest  
Gains  
( $n > T$ )

Power  
Ceiling  
at 0.40

# Increase $n$ and $T$ of Between-Subject Multi-Level Model

- ▶ Double or Triple  $n$  and/or  $T$  for:
  - ▶ Within-Subject at  $\delta = 0.05$

Steep  
Gains!!  
( $T > n$ )

Multi-Level	$T = 50$	$T = 100$	$T = 150$
$n = 40$	0.20	0.47	0.75
$n = 80$	0.44	0.71	0.91
$n = 120$	0.67	0.81	0.97

Power  
close to 1  
if increase  
both  $n, T$

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