

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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EE-BGT, Lecture 1c (Experimetrics Module 2)

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 ^u	0.702
M-W Test	0.053 ^u	0.683
K-S Test	0.040 ^u	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 10 10.5 , sd(1) n(100)`

- ▶ μ_0 / μ_1
sample std; sample size

- ▶ 2-sample t-test

DGP with Normally Distributed Errors

▶ Same as power analysis

▶ STATA Results:

```
power twomeans 1
```

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

```
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
```

t-Test best due to Normality?

DGP w/ Non-Normally Distributed Errors

$$x_i = 10 + \underbrace{\delta \cdot d_i}_{\substack{\uparrow \\ \text{[Treatment Effect]} \times \text{[Treatment dummy]}}} + \epsilon_i, \quad i = 1, \dots, n = 100$$

- ▶ [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

All three
unbiased
(properly sized)

	Size	Power
t-Test	0.056 ^u	0.566
M-W Test	0.056 ^u	0.526
K-S Test	0.039 ^u	0.306

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 ^u	0.705
M-W Test	0.067	0.867
K-S Test	0.052 ^u	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

- ▶ 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:
40 Subjects
of 50 Rounds
each (10
Groups of 4)

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)

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

▶ Three-Level Model:

▶ u_i : **Subject-specific RE**

▶ v_j : **Group-specific RE**

▶ ϵ_{ijt} : **Observation-specific error**

Example:
40 Subjects of
50 Rounds
each (**10**
Groups of 4)

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)

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

► 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)

$$y_{ijt} = \alpha + \delta d_i + \beta x_{ijt} + \cancel{\alpha_i} + \cancel{\alpha_j} + \epsilon_{ijt}$$

► Linear Regression (OLS) Model:

► e_{ijt} : Observation-specific error

Between-Subject vs. Within-Subject Treatment Effects

$d_i = 0$ for Subject $i = 1-20$; $d_i = 1$ for Subject $i = 21-40$

d_i : (Between-Subject) Treatment Dummy

▶ d_{it} : Within-Subject Treatment Dummy

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

Example:
40 Subjects

▶ Three-Level Model:

▶ u_i : Subject-specific RE

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

$d_{it} = 1$ for Round $t = 26-50$ 10

▶ v_j : Group-specific RE

Groups of 4)

▶ e_{ijkt} : Observation-specific error

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
Results:

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

Cluster at Group j
and Subject i

```
Performing EM optimization:
```

```
Performing gradient-based optimization:
```

```
Iteration 0: log likelihood = -2959.3982  
Iteration 1: log likelihood = -2959.3978  
Iteration 2: log likelihood = -2959.3978
```

```
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

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

Three-Level

► STATA
Results:

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

Error STD
for Group j
and Subject i
& Residual e

```
Log likelihood = -2959.3978      Wald chi2(2)      =      155.37
                                Prob > chi2           =      0.0000
```

y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
d	.1482739	.0454989	3.26	0.001	.0590978	.23745
x	.0955655	.0079035	12.09	0.000	.0800749	.111056
_cons	-.1241784	.247917	-0.50	0.616	-.6100867	.3617299

```
Random-effects Parameters      |      Estimate      Std. Err.      [95% Conf. Interval]
```

j: Identity	$\hat{\sigma}_v$				
	sd(_cons)	.4820359	.292011	.1470391	1.580251
i: Identity	$\hat{\sigma}_u$				
	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.

d : Treatment increases bid by 0.148

x : How values affect bids

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

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

u: Not significantly different from 0.05

Multi-Level highest (but low)

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

All 7 unbiased (with 100 replications)

Size: $d = 0$

Power: $\delta = 0.05$

OLS

0.02^u

~~0.07~~

OLS clustering at subject level

0.09^u

0.31

OLS clustering at group level

0.09^u

0.33

RE (no clustering)

0.05^u

0.31

RE clustering at subject level

0.09^u

0.31

RE clustering at group level

0.08^u

0.33

Multi-Level (subject and group level)

0.05^u

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
- ▶ Within-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?
gen $y=0.5+\text{delta}*d+0.1*x+u+0.1v+e$
 - ▶ Is size good now?
 - ▶ 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 $\delta = 0.5$

Modest
Gains
($n > T$)

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

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 $\delta = 0.05$

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

Steep
Gains!!
($T > n$)

Power
close to 1
if increase
both n, T

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