

Module 3: Concepts and Procedures in G Theory

1

Main topics

- Describing a one-facet model
- Running the programs
 - Using one facet design as the example
 - Spss program
 - Genova program

2

Review of concepts

- Universe score and multifaceted measurement error
 - Universe is defined by facets which can be fixed or random
- Relative and absolute decisions
 - Purpose/use of scores
 - Statistical methods involved
- G coefficients
 - $V(p) / (V(p) + \text{variance of measurement error})$
- G and D studies

3

Describing the model

- Only one facet is involved in the design. For example, 5 tasks were administered to 50 students ($p \times t$), or 2 raters reviewed 80 students' research papers ($p \times r$).
- The facet is random.
- The universe to be generalized is defined as all the conditions of this single facet.

Q: What about measurement errors?

4

Describing the model: Equations

Take the $p \times i$ as an example:

- Observed score

$X_{pi} =$	μ	[grand mean]
	$+ X_p - \mu$	[person effect]
	$+ X_i - \mu$	[item effect]
	$+ X_{pi} - X_p - X_i + \mu$	[residual effect]

- Variance

$V(X_{pi}) = V(\mu_p) + V(\mu_i) + V(\mu_{pi,e})$ or

$$\sigma^2(X_{pi}) = \sigma_p^2 + \sigma_i^2 + \sigma_{pi,e}^2$$

5

Calculating G coefficient(s)

- Take the $p \times i$ design (random effect) as an example

Effect	SS	MS	Expected MS	Est. VC using 1 item	Est. G/D Study V Components
Person(p)	SS_p	$MS_p = SS_p/df_p$	$E(MS_p) = \sigma_{pi,e}^2 + \sigma_p^2$	$\sigma_p^2 = [MS_p - MS_{pi,e}]/n_i$	σ_p^2
Item(i)	SS_i	$MS_i = SS_i/df_i$	$E(MS_i) = \sigma_{pi,e}^2 + n_p \sigma_i^2$	$\sigma_i^2 = [MS_i - MS_{pi,e}]/n_p$	σ_i^2/n_i
pi, e	$SS_{pi,e}$	$MS_{pi,e} = SS_{pi,e}/df_{pi}$	$E(MS_{pi,e}) = \sigma_{pi,e}^2$	$\sigma_{pi,e}^2 = MS_{pi,e}$	$\sigma_{pi,e}^2/n_i$

6

Notations again

- Error variances
 - $\sigma^2(\delta)$ refers to relative error
 - $\sigma^2(\Delta)$ refers to absolute error
- G coefficients
 - ρ^2 refers to relative G coefficient
 - ϕ refers to absolute G coefficient

7

Running the Programs to Conduct G and D Studies

8

Three main ways

- Obtain the ANOVA results and then calculate the variance components – an indirect way

Two direct ways

- Run the SPSS - Variance Components
- Run the Genova

9

Obtain the EMS from the anova output

- You can run the anova or scale in spss

10

Comparison of SPSS and Genova

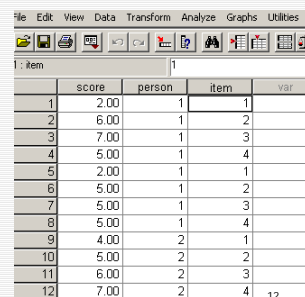
- When SPSS
 - Data structure has all the scores of all test takers just in a column
 - Includes missing value
 - Only handles the crossed design
- When Genova
 - Data structure has all the scores of one test taker in a row
 - Does not include any missing values
 - Involves the nested design (or crossed design)

11

Running SPSS: data preparation

The data should be organized using the structure for running anova, e.g.,

- A column for item scores
- A column for stu id
- A column for item id

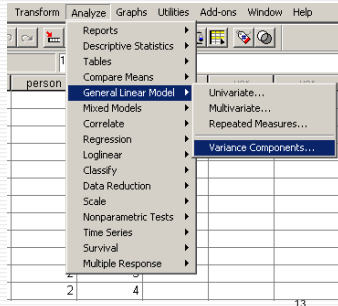


	score	person	item	var
1	2.00	1	1	
2	6.00	1	2	
3	7.00	1	3	
4	5.00	1	4	
5	2.00	1	1	
6	5.00	1	2	
7	5.00	1	3	
8	5.00	1	4	
9	4.00	2	1	
10	5.00	2	2	
11	6.00	2	3	
12	7.00	2	4	12

Running SPSS: analyzing

Use the menu to run the analysis

- (1) Analyze -> GLM -> Variance Components...

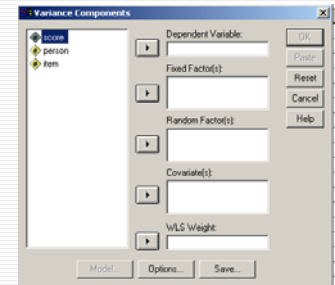


Running SPSS: analyzing

Then, you will see a small window

You need to:

- (1) Select "score" as Dependent Variable;
- (2) Select person and item as "Random Factor(s)"



The result!

- Identify the three variance components from your SPSS output
- Use Excel to calculate
 - the relative and absolute measurement errors
 - the relative and absolute coefficients

Component	Estimate
Var(item)	.968
Var(person)	.824
Var(item * person)	1.540
Var(Error)	.000 ^a

Dependent Variable: score
Method: Minimum Norm Quadratic Unbiased Estimation (Weight = 1 for Random Effects and Residual)
^a. This estimate is set to zero because it is redundant.

Try with mini-project 1 data

- Now use the spss VC to generate the G study results when using only one judge with your homework data

Mini-project 1 again: go through the genova program with the data

- Now use the genova to generate the G study results when using only one judge with project 1 data

One Facet Model: Cross and Nested Design

Main topics

- ☐ Review of the crossed design
 - Sharing your output
 - Interpreting the results from Genova
- ☐ Basic descriptions of the nested design
 - Describing the model
 - Estimating the G coefficient(s)
- ☐ Comparison of the crossed and nested designs
 - Ways to identify the variance components
 - Homework problems

19

Interpret the Genova output (1)

Pages for you to double check your program

- ☐ P1 - show a proportion of your card file
- ☐ P2 - indicate the design and df
- ☐ P3 - list the scores
- ☐ P4 - print the means

20

Interpret the Genova output (2)

Pages about the process of estimating the VCs

- ☐ P5 – report the anova results
- ☐ P6 - two procedures for estimating the VCs and SEs
- ☐ P7 - equations that link estimated MS with variances
- ☐ P8 - variance-covariance matrix

21

Interpret the Genova output (3)

Pages that report the G/D study results

- ☐ P9 - another proportion of your card file
- ☐ P10 - the output of your G study
 - It includes all the information that you need.
- ☐ Other pages related to D studies and summaries

22

Intro of the Nested Design

23

Describe the model

- ☐ Only one facet is involved in the design. For example, each of the 50 students took a different set of 10 items (i:p), 20 research papers submitted to a journal and each was reviewed by 2 different reviewers (r:p).
- ☐ The universe to be generalized is defined as conditions from one single facet.

24

Describe the model: examples

Determine what G design can be used to estimate the reliability.

- A set of four tasks were given to 30 students.
- In a study to evaluate teachers' effectiveness, each of 50 teachers randomly selected from WA state was observed by a different researcher.
- 30 students were administered a different set of two essay items.
- In a wine competition, each type of wine was evaluated by four judges.

25

Describe the model: equations

Take the i:p as an example:

- Observed score

$$X_{i:p} = \mu \quad [\text{grand mean}]$$

$$+ \mu_p - \mu \quad [\text{person effect}]$$

$$+ X_{i:p} - \mu_p \quad [\text{residual effect}]$$

- Variance

$$V(X_{i:p}) = V(\mu_p) + V(\mu_{i:p, e}) \text{ or}$$

$$\sigma^2(X_{i:p}) = \sigma_p^2 + \sigma_{i:p, e}^2$$

26

Estimate G coefficient(s)

- Take the i:p design (random effect) as an example

Effect	SS	MS	Expected MS	Est. VC Using 1 item	Est. G/D Study VC
Person(p)	SS _p	MS _p = SS _p /df _p	E(MS _p) = $\sigma_p^2 + \eta_i \sigma_p^2$	σ_p^2 = [MS _p - MS _{i:p}]/ η_i	σ_p^2
res (i:p,e)	SS _{i:p,e}	MS _{i:p} = SS _{i:p} /df _{i:p}	E(MS _{i:p}) = $\sigma_{i:p,e}^2$	$\sigma_{i:p,e}^2$ = MS _{i:p}	$\sigma_{i:p,e}^2 / \eta_i$
G coefficient = $\sigma_p^2 / (\sigma_p^2 + \sigma_{i:p,e}^2)$					

27

Final note on the nested design

The nested design is

- not possible to obtain a separate estimate for all the variance components, such as V(i)
- less flexible in the D studies

Then, why nested design?

- Logistic concerns
- Some designs by nature are nested

28

Ways to identify variance components: notation

Rules

- When a facet (B) is nested with another facet (A), B:A, this means that main effect of B cannot be a stand alone. It is always subsumed in B:A.

29

Ways to identify variance components: notation

p x (i:t)

- Start with the nested part
- Decompose the nested part as two
 - t is a main effect and i is nested with t, so we should expect t main effect plus i:t nested effect.
- Continue with cross design
 - Since p is cross with (i:t), we should expect: p main effect, p cross with t, and p cross with i:t.
 - Regarding notations, they are p, pt, and i:pt, e.

30

Now your turn

i:t:p

p x (r:i)

31

Ways to identify variance components: diagram

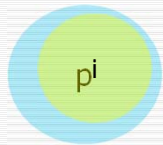
Rules:

- Each circle represents a facet.
- When two facets are crossed, the two circles are overlapped.
- When one facet (B) is nested within another facet (A), then Circle B is completely covered (or included) by Circle A.

32

Ways to identify variance components: diagram

Persons (p) were given a set of items (i).

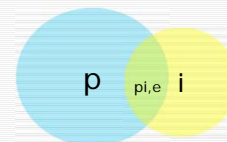


33

Ways to identify variance components: diagram

p x i -

persons were given the same set of items.



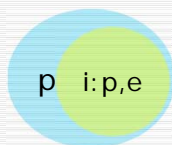
How many colors do we have now?

34

Ways to identify variance components: diagram

i:p -

persons were given different sets of items.



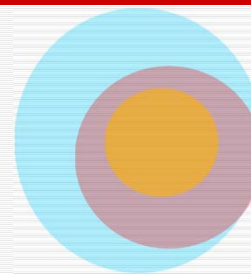
How many colors do we have now?

35

Let's figure out p x (i:t)

p

t



i

36

Now your turn

i:t:p

p x (r:i)

37

Compare Crossed and Nested Designs

38

If time allows (see the example that I pass out), let's compare the crossed and nested models ...

- ☐ Run two D studies from a p x j design using the data in mini-project 1.
 - First D study includes 3 judges, using p x j design.
 - Second D study includes 3 judges too, using j:p design.
- Compare the two sets of D study outputs.
 - ☐ What is the relationship between p variance?
 - ☐ What is the relationship between the abs. G coefficient?
 - ☐ Any other relationship?

39

Planning: what data do you have?

- ☐ Go and check the data file that you have
- ☐ Start to describe the measurement procedure involved
 - what facets are you interested in investigating?
 - are they crossed or nested?
 - what design will you consider in your G study?

40