### Aggregating quantitative data to draw meaningful conclusions

EGAD Workshop June 2019 Peter Ostafichuk, Margaret Gwyn, and John Donald with Brian Frank, Nerissa Mulligan and Jake Kaupp



### **Session focus**

Comparing approaches to aggregating data, going

From: Task-level scores
 assigned to individual students ---- (which most of us are doing)
 To: Demonstrating that a student cohort possesses a graduate attribute

#### Canadian Engineering Accreditation Board

Bureau canadien d'agrément des programmes de génie

2018 Accreditation Criteria and Procedures • Normes et procédures d'agrément 2018 Revised November 2018 / Révisé en novembre 2018

**3.1.5 Assessment results**: At least one set of assessment results must be obtained for all twelve attributes over a cycle of six years or less. The **results should provide clear evidence that graduates of a program possess the above list of attributes**. **3.2.1 Improvement process:** There must be processes in place that demonstrate that program outcomes are being assessed in the context of the graduate attributes, and that the **results are validated, analyzed** and applied to the further development of the program.

# When you **aggregate data**, you replace **groups of observations** with **summary statistics** based on those observations.



#### Amount of data conveyed



"It is not possible to carry out meaningful statistical analysis of data that is fundamentally inaccurate."

https://totalinternalreflectionblog.com/2018/09/11/garbage-in-garbage-out/

# We need *reliable data* if we are to draw *valid conclusions*.

**Task 1:** In your groups, discuss what does it mean to say data is "reliable."

# Task 1 discussion: To draw valid conclusions we need reliable data.

- Reliability of data relies on consistency, which can be measured as:
- Consistency over time
  - i.e. test-retest reliability
- Consistency between graders
  - i.e. inter-rater reliability
- Internal consistency
  - i.e. inter-item reliability

Validity of conclusions depends on:

Measuring the right things (e.g. indicators) THER
 Using appropriate approaches to MACORKSHOPS
 Agreement with conclusions drawn from other approaches (students, employers, alumni,...)

### An analogy to test understanding...

Inter-rater reliability:



Inter-item reliability:



Validity:



Test-retest reliability:



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Inter-item reliability:



### Characterizing reliability could involve comparing over time (e.g. multiple tests):

Comparison over time



### Characterizing reliability could involve comparing grader agreement:



**Task 2:** In your group, identify what you are doing, or could do, in your GA process to make your data consistent, considering three possible risks:

- 1. Consistency over time (test-retest reliability)
- 2. Consistency between graders (inter-rater reliability)
- 3. Internal consistency (inter-item reliability)

Task 2 discussion: what you are doing, or could do, in your GA process to make your data consistent, considering three possible risks:

- 1. Consistency over time (test-retest reliability)
- 2. Consistency between graders (inter-rater reliability)
- 3. Internal consistency (inter-item reliability)

Student ID	Progra m	Year of Study	Course	Attribute	Indicator	Assessment	Score
A	ENGR	1	ENGR 101	КВ	ENGR-KB-1	Midterm	3
A	ENGR	1	ENGR 101	КВ	ENGR-KB-1	Final	5
В	ENGR	1	ENGR 101	КВ	ENGR-KB-1	Midterm	4
В	ENGR	1	ENGR 101	КВ	ENGR-KB-1	Final	4
С	ENGR	1	ENGR 101	КВ	ENGR-KB-1	Midterm	4
С	ENGR	1	ENGR 101	КВ	ENGR-KB-1	Final	1

#### group\_by(Student ID, Course, Attribute, Indicator)

#### summarize(Score = mean(Score)

Student ID	Course	Attribute	Indicator	Score
А	ENGR 101	KB	ENGR-KB-1	4
В	ENGR 101	KB	ENGR-KB-1	4
С	ENGR 101	КВ	ENGR-KB-1	2.5

#### group\_by(Course, Attribute, Indicator)

Student ID	Course	Attribute	Indicator	Score		
A	ENGR 101	KB	ENGR-KB-1	4		
В	ENGR 101	КВ	ENGR-KB-1	4		
С	ENGR 101	KB	ENGR-KB-1	2.5		

#### summarize(Score = mean(Score)

Course	Attribute	Indicator	Score		
ENGR 101	КВ	ENGR-KB-1	3.5		



Raw Data

Distribution

### Framework for comparing aggregation approaches

Factor	Possible options
Aggregation target	<ul> <li>single value (e.g. Design = 3.6/5)</li> <li>distribution of performance, (e.g. histogram of student performance)</li> <li>qualitative description (textual based analysis of results)</li> </ul>
Aggregation level	<ul> <li>up to attribute (e.g. "Design")</li> <li>up to indicator within each attribute (e.g. "Problem definition")</li> <li>up to course-level learning outcome / sub-indicator (e.g. aspect of "problem definition")</li> </ul>
Differentiation factors	<ul> <li>Year of Program (Year 1 to 4)</li> <li>IDA level (Introduce, Developed, Applied)</li> <li>Program option (e.g. biomechanics vs. materials)</li> <li>Summative vs. Formative</li> <li>Assessment type (e.g. final report, exam, lab simulation, portfolio)</li> <li>Student groups (first in family, racialized, Indigenous)</li> </ul>
Reliability measure	<ul> <li>Correlation between tasks (e.g. correlation between three measures of "problem definition")</li> <li>Correlation between years (e.g. correlation between scores in 2016, 2017, and 2018)</li> <li>Correlation between multiple ways of measuring an indicator</li> </ul>

### Examples



Self-Assess Survey(4th yr)



	KB		PA		1	N	D	E	E	Т	Т	W	C	S	P	R	I.	S	E	E	E	С	ш	
٠	Val	Targ	Val	Targ	Val	Targ	Val	Targ	Val	Targ	Val	Targ	Val	Torg	Val	Torg	Val	Targ	Val	Targ	Val	Targ	Val	Tar
1	11.1	80	82.6	80	86.7	80	100.0	80	- <b>m</b> ax	80	97.2		91.6	80	-95.2	80	- 899	80	45.0	80	12.8	80	73.7	
2	75.9		67.0		90.3	80	87.2	80	- 16.4	80	-954-	80	-98.5	80	- 46.2	80	91.8	80	84.7		\$7.5		8.1	80
3	. 93.9	80	-98.5	80	83.5	80	- 92.4	80	-	80	70.8		-82.8	80	- 95.5	80	- 812	80	13.9	80	76.5	80	70.0	-
4	10.5	80	46.8	80	93.5	80	- 17.4	80	10.4	80	2.1	80	46.7	80	74.2		108.0	80	49.7	80	94.4	80	ma	80
5	1015	80	52.6		-	80	300.0	80	10.1	80	18871	80	17.4	80	-00.0	80	8.5	80		80	10000	80	86.6	80
6	73.7		-45.1	80	62.5		96.9	80	54.3	(#)	993	80	100.0	80			85.5		19.4	80	66.8		43.3	-
7	107.5	80	-	80	867	80	100.0	80	- 100	80	1963	80	66.8	. *			904	80	83.1	1.00	10.5	80	15.4	80
	40.7	80	60.2				-84	80	1		994.	80	864	80					81.2	80			- 19.0	80
9	73.9		- 89.6	80			-100.0	80			12.0	80											- 100	80
10	100.4	80	79.9	BO			47.6				1983.	80											73.9	
11	48.0		54.1				16.6	80			11.6	80											-	80
12	16.7	80	87.0	80							93.5													
13	36.4	80									95.8	80							Tar	get m	et or e	xceed	led	
14	88.5	80									75.5								Per	forma	ance be	elow t	arget	E.
15	78.6	80									142.5	80												5
16	-	80																	Missing data					
17	80.7	80																	No target					
18	18.1	80																L				_		

Factor	Approach						
Aggregation target	Single values: Mean and % meeting target						
Aggregation level	Attribute, but only for CEAB reporting, not internal use						
Differentiation factors	Direct/peer/self assessment						
Reliability measure	Factor analysis						

### GA4 (QR4) by year



			Introd	uctory Developing	Adva	anced				-			
100% 90%				*********	4: Exceeds expectations						Factor	Approach	
<b>9</b> 80%							Ħ.	2: 140	otc own	octations			
ad 60%	-		•••••				<b>!!</b> '	<b>5.</b> IVIE	ersexp		Aggregation	Frequency	
te 50%							tt ·	2: Ma	rginally	meets	target	distribution	
30%	10					H	H.	expect	tations is to me	et expectations	laryet		
20%					HH	HH	HĽ.	2. 101	0 00 111		1	of performance	
0%						Ш	Ц,	0: No	demon	strated	A 1*		
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	BWB	BWB	BWI BWI BWI BWI BWI BWI BWI		BWB	BWI	BWI			-	level	(learning	
				Assessment Tool						-		outcome within	
										-		indicator)	
Tool	Ind	Level	Assessor	Question or course learning outcome	# of s	students at LOM % of students				% of students		mulcatory	
				•	0	1	2	3	4	over threshold	Differentiation		
BWE336	а	I	1. Instructor	CLO #5 (Awesome assessment method #5)	7	47	88	93	25	45%	factore		
<b>BWE477</b>	а	1	1. Instructor	CLO #4 (Awesome assessment method #5)	18	54	93	44	50	36%	Tactors		
BWE106	а	1	1. Instructor	CLO #11 (Awesome assessment method #7)	14	35	45	27	71	51%			
BWE181	а	1	1. Instructor	CLO #11 (Awesome assessment method #6)	27	75	8	6	60	38%			
BWE237	а	I	1. Instructor	CLO #11 (Awesome assessment method #1)	25	46	26	28	83	53%			
BWE426	а		1. Instructor	CLO #7 (Awesome assessment method #8)	43	91	49	15	82	35%			
BWE442	a		1. Instructor	CLO #6 (Awesome assessment method #1)	37	48	44	27	57	39%			
BWE310	a		1. Instructor	CLO #3 (Awesome assessment method #2)	61	8/	81	20	// 61	38%	Deliability		
BWE380	a		1. Instructor	CLO #8 (Awesome assessment method #1)	37	00	35	29	10	40%	Reliability		
BWEZZ4	a		1 Instructor	CLO #2 (Awesome assessment method #6)	35	41	54	8	44	35%	measure		
BWF241	a	1	1. Instructor	CLO #6 (Awesome assessment method #3)	35	93	72	1	75	28%			
BWE182	a	1	1. Instructor	CLO #10 (Awesome assessment method #3)	77	89	6	47	14	26%			
	1999	100 Sa 6						96.5.0			i i i i i i i i i i i i i i i i i i i		



#### **Design Overview** Approach Factor Aggregation Distribution Mastery of performance target High Quality ¢ Meets Expectations Aggregation Indicator Marginal level Not Demonstrated Differentiation Semesters factors 8 Semester Samples of Design Indicators Design Conceptual, Convergent & Performance Process creativity Divergent evaluation Reliability Mastery measure High Quality Meets Expectations

12345678

12345678

Semester

12345678

Marginal

Not Demonstrated

2345678



**Reliability measure** 

Median performance change from year 1 to 4

% agreement (Inter-rater reliability)



### GA4.3 (QA4.3), same group, civil engineering



# Task 3: Connect the factors to current institutional approaches

- Identify how well the four factors describe the institutions' approaches represented within the group. Is there a key factor that is not captured by that list of four? If so add it to your list of factors to consider.
- 2. Briefly describe each institution's approach using the factors.

# Task 3 discussion: Connect the factors to current institutional approaches

**Report out:** 

- 1. Were there any key factors emerge that were not captured by the original list of four?
- 2. How well are institutional approaches captured by the factors? Are there two extremely different approaches between institutions represented at your table?

### **Task 4:** Consider what aggregation means to key stakeholders

As a table group, identify what key stakeholders are looking for from aggregation. Consider

- Course instructors
- Department administration
- Faculty administration
- CEAB visiting team

# Task 4 discussion: Consider what aggregation means to key stakeholders

Report out: what are key stakeholders are looking for from aggregation?

- Course instructors
- Department committees, staff, and administration (which may also consult with broader stakeholders)
- Faculty committees, staff, and administration (which may also consult with broader stakeholders)
- CEAB visiting team

### **Task 5:** What mix of aggregation approaches would meet the collective needs of stakeholders?

#### **Consider needs of key stakeholders:**

- Course instructors
- Department committees, staff, and administration (which may also consult with broader stakeholders)
- Faculty committees, staff, and administration (which may also consult with broader stakeholders)
- CEAB visiting team

#### Consider factors in aggregation

- Aggregation target: single value, distribution of performance, or qualitative description
- 2. Aggregated level: up to attribute, up to indicator within each attribute, up to task within indicator within attribute
- Differentiation factor: differentiate by year, IDA level, program, student sub-group, student
- 4. Reliability measure: qualitative or quantitative instructor rating, correlation between tasks or years

# **Task 5 discussion:** What mix of aggregation approaches would meet the collective needs of stakeholders?

#### **Consider needs of key stakeholders:**

- Course instructors
- Department committees, staff, and administration (which may also consult with broader stakeholders)
- Faculty committees, staff, and administration (which may also consult with broader stakeholders)
- CEAB visiting team

#### **Consider factors in aggregation**

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- 2. Aggregated level: up to attribute, up to indicator within each attribute, up to task within indicator within attribute
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Slides available

bit.ly/EGAD-2019

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