

**Making Sense of SCIENCE: The Challenges of Impacts and Assessments**

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## **Making Sense of SCIENCE: The Challenges of Impacts and Assessments**

### **Background.**

With the introduction of the Next Generation Science Standards there is an urgency to evaluate impacts of programs aligned to the new Standards and to ensure that assessments validly reflect the core concepts and skills.

We investigate the impact of an NGSS-aligned program, Making Sense of SCIENCE (MSS) funded through the Investing in Innovation (i3) program. We report the main impact findings; however, our primary focus is on the challenge of assessment. That is, without a valid assessment, determination of “practical significance and meaningful effects”, both central to the theme of SREE 2020, cannot be achieved.

We faced the challenge that no assessments were available by the end of the study that adequately tapped the Depth Of Knowledge (DOK) and higher-order thinking skills that NGSS and MSS promote. Therefore, science content experts on the evaluation team, independently developed an assessment written to general specifications that drew on established previously operational items from multiple sources (e.g., NAEP and MOSART).

The test, rapidly developed, satisfied content needs, but presented challenges by introducing uncertainty in reliability, difficulty, and discriminability. At the same time, availability of item specific outcomes created a valuable opportunity to investigate the robustness of impact results to alternative approaches to scaling item responses. We focus on these aspects.

### **Research Questions**

**Confirmatory questions:** What is the impact of MSS on student science achievement in 4<sup>th</sup> and 5<sup>th</sup> grade after two years of implementation compared to business as usual? What is the impact on students below the first tertile of incoming achievement?

**Exploratory questions:** Did impacts vary depending on state (Wisconsin versus California) and grade (4<sup>th</sup> vs 5<sup>th</sup>)?

**Questions about the assessment:** What were the test’s properties, including its difficulty, discriminability and reliability? How robust are impact findings to alternative methods to score scaling (e.g., use of Item Response Theory versus percent correct, and alternative approaches to handling missing responses).

### **Setting**

The study was conducted in five districts in California and two in Wisconsin. The districts were a mixture of rural and urban.

### **Participants**

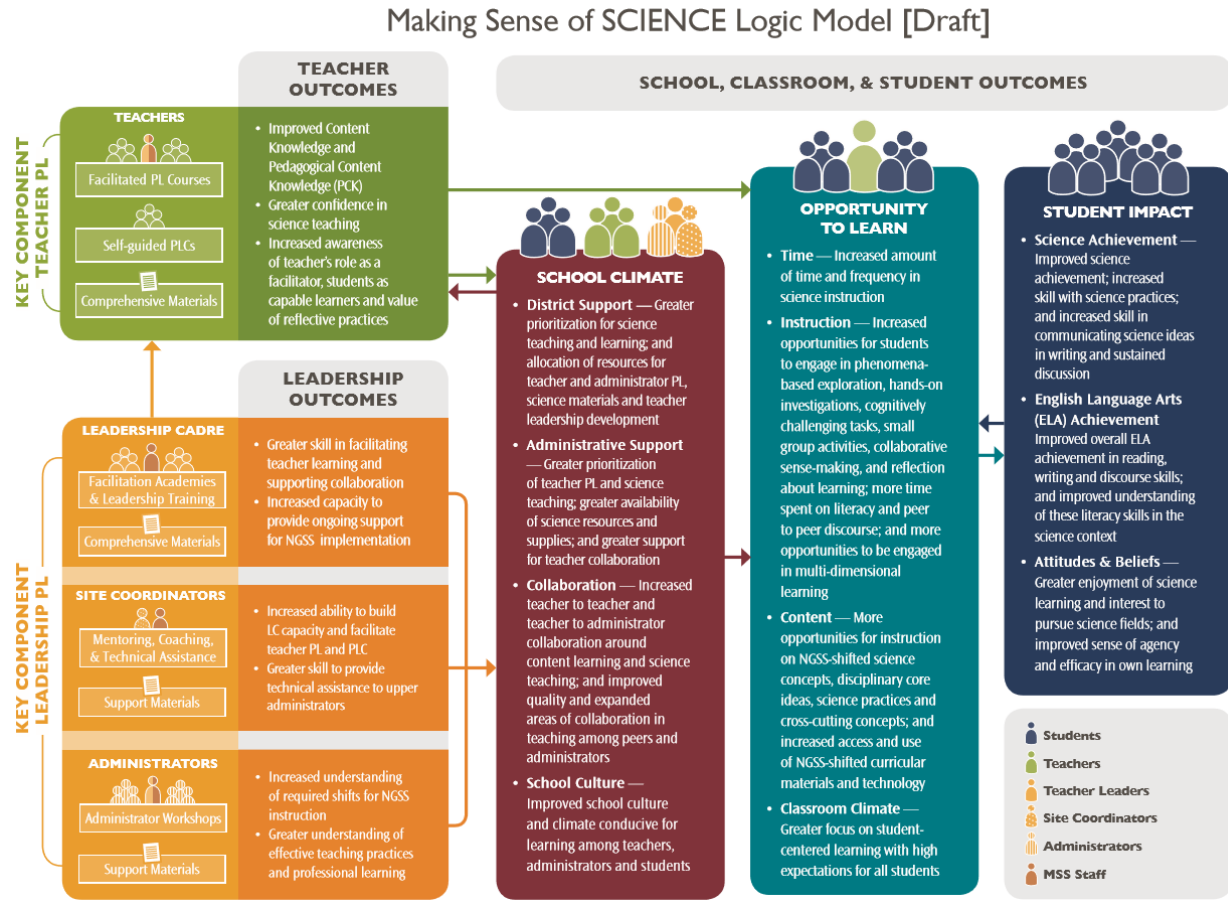
Impacts were assessed on 2,140 4<sup>th</sup> and 5<sup>th</sup> grade students in 55 schools. Student ethnicity was as follows: Asian (N=240, 11.2%), Black (N=318, 14.86%), Hispanic (N=881, 41.17%), White (N=448, 20.93%), Other and unknown (253, 11.80%).

### **Program**

MSS is a PD program aimed at raising students’ science achievement through improving science instruction. It focuses on practices supporting the implementation of Next Generation Science Standards (NGSS). Professional learning activities for teachers include 30 hours of

professional learning each summer for two summers and 12 hours of Professional Learning Communities (PLCs) each year. (See logic model in Figure 1, with key components at left).

Figure 1. Logic Model Making for Sense of SCIENCE



### Research Design

The study randomized 60 schools to MSS or a two-year wait-list. Impacts on students on the main science assessment were assessed after two years. (Randomization was in winter 2016, with outcomes assessed in spring 2018.) Attrition of schools was low, with impact assessed on 2,140 students in classes of 174 teachers and 55 schools. Student and teacher joiners were allowed to enter the study to keep up the sample sizes, making this an RCT that will be reclassified as a QED for review under WWC 4.0.

### Data Collection and Analysis

The assessments was administered online to students in Spring 2018. We estimated impact using 3-level Hierarchical Linear Models (HLMs) (student, cluster, randomized blocks), with individual student scores regressed against baseline covariates, a dummy variable indicating treatment assignment at the cluster level (MSS=1, control=0), and random effects at student, cluster, and block levels.

**Results**

***Confirmatory Impact Findings***

Impact of MSS on students is displayed in Table 1. We show results of progressively adding covariates to the model. The benchmark result (Model 9) shows no impact of MSS on student achievement after two years of implementation with standardized effects size .076 ( $p=.25$ ). Baseline equivalence was achieved on both pretests:  $-.13$  SD ( $p=.18$ ) on the 3<sup>rd</sup> grade state ELA test, and  $-.07$  SD ( $p=.48$ ) on the math pretest. We also did not observe a significant effect on students below the first tertile, with effect size .085 ( $p=.158$ ).

*Table 1. Impact on Student Content Knowledge (n= 2,140, J= 55 )*

Effect	BENCHMARK								
	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6	Model 7	Model 8	Model 9
Intercept	-0.059	0.028	-0.075	0.038	0.066	0.984**	-0.037	0.702**	<b>0.998**</b>
In Treatment group	-0.010	0.053	-0.010	0.054	0.062	0.062	0.064	0.051	<b>0.074</b>
Effect size	-0.010	0.054	-0.010	0.055	0.064	0.064	0.066	0.052	<b>0.076</b>
Pretests		X		X	X	X	X	X	X
Grade 4					X				X
State CA			X	X	X	X	X	X	X
Other student covariates					X				X
Set 1 (teacher covariates)						X	X		X
Set 2 (teacher covariates)						X		X	X

\* $p < .10$  ; \*\* $p < .05$  ; \*\*\* $p < .01$  ; \*\*\*\* $p < .001$ ; We do not show variance component estimates in this table.

***Exploratory Questions.*** Fidelity Of Implementation (attendance at PD, PLC) was high, but just below required threshold. There was no difference in impact between WI and CA. We observed positive impact of the program in 5<sup>th</sup> grade (Effect size=.144,  $p=.045$ ), but not in 4<sup>th</sup> grade (Effect size=-.032,  $p=.735$ ). (Baseline equivalence on the pretest was achieved in each case.) We will report at SREE impacts depending on exposure, dosage and fidelity of implementation.

***Questions about the Assessment.***

We used a new assessment, well aligned to NGSS standards, and drawing on established items used in previous operational assessments. It was designed to allow students to demonstrate Depth Of Knowledge targeted by NGSS and MSS. The assessment was difficult (with averages of 42% and 38% in percent correct in 4<sup>th</sup> and 5<sup>th</sup> grades respectively) and with moderate reliability (Cronbach Alpha=.63). A test written to more challenging content standards was hard for lower

achieving students, as supported by the test score standard error calculated across ability levels as displayed in Figure 2.

Figure 2. Standard error of the student science assessment

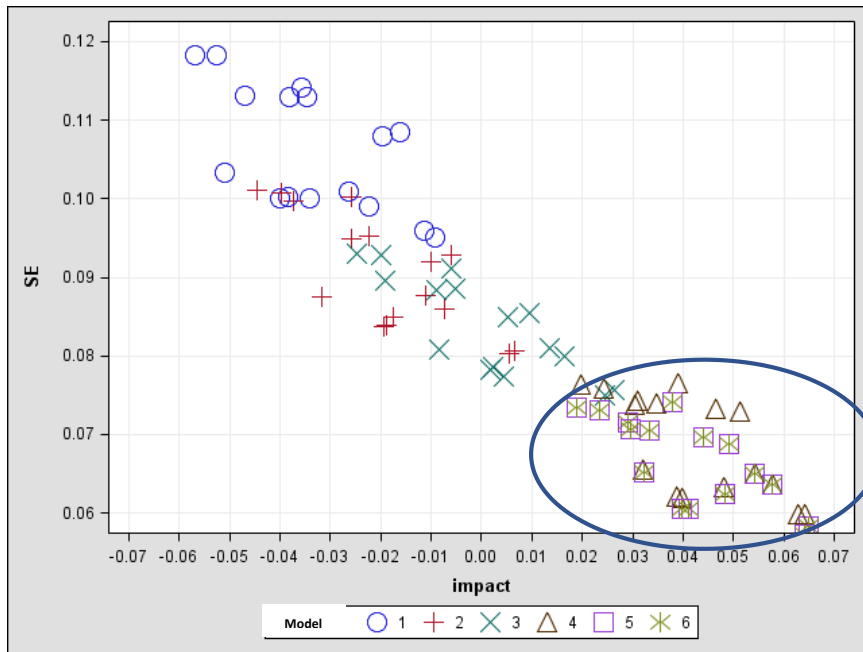


We addressed the general technical question of whether impact results depend on different approaches to scaling posttest scores. (This question is not normally asked, as testing companies typically provide ready scale scores.) Brevity of this abstract precludes full discussion, but for each of six families of impact models (Table 2) we examined impacts across 8 combinations of scaling factors (2 strategies for missing item responses  $\times$  2 ways to address items showing moderate Differential Item Functioning  $\times$  2 ways of scaling responses (3PL IRT calibration versus percent correct)), and 2 samples (including students in classes of teacher joiners or not). We focus on the oval in Figure 3, which shows that for each of Models 4, 5 and 6, **impacts ranged up to .05 standardized effect size units depending on how item responses were scaled**. This is an important finding given that impacts as small as .05 SD can be educationally meaningful (Newman et al., 2012).

Table 2. Characteristics of Six Basic Impact Models

Dimension	Description	Model					
		1	2	3	4	5	6
	Base: indicator of random assignment status, random cluster and student effects	X	X	X	X	X	X
Compositional	Including a pretest in analysis				X	X	X
	Including a block-level random effect		X	X		X	X
Structural	Using block random effects instead of cluster level random effects			X			X

Figure 3. Impacts and their standard errors with changes of structural, compositional and scaling factors of impact models



### Conclusions

This work reminds us that impacts of programs cannot be separated from their contexts (Cronbach, 1975, 1982) including the assessment used. The attempt to evaluate deeper knowledge in this experiment led to a difficult test that may have restricted the range over which valid responses could be measured (Shadish, Cook and Campbell, 2002). Yet developing a NGSS-responsive assessment was critical. It is hard to establish practical significance of effects in a shifting educational landscape where science content, assessment of its learning, and educational policy driving it, are evolving.

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