Enhancing student learning in a Large-Enrollment, Scientific Computing Workshop in Seismology





Seismology Skill Building Workshop (SSBW)



cal thinking skills while answering questions and receiving immediate feedba

Study Motivation

Contextual Focus:

Centers on computer programming in the context of seismology

Module and Assignment Structure:

- 6 modules with approximately 6 tutorial assignments per module
- Comprises a total of 1084 questions
- Immediate Feedback:
- Provides immediate feedback to students
- Allows for reattempting questions based on feedback

Learning Management System (LMS):

- Utilizes the Moodle LMS for interactivity and data collection
- Increasingly digital nature of assignments provides an opportunity for in-depth assessment of student learning and course design.
- The SSBW was chosen for analysis due to its high student enrollment, facilitating trend identification within the data.
- There is a need for a shift to hands-on, computational coursework in geoscience and the SSBW addresses this need on a large scale.

Who The SSBW Reaches

- Since 2020, international participation in the workshop has surged from 46% to 83% by 2024.
- US participants have included ~60% women and ~30% marginalized racial and ethnic groups, surpassing diversity levels in undergraduate geoscience degrees awarded to women in 2019 46%) and doubling those awarded to marginalized groups in 2016 (15%).
- Learn more Tuesday 8:30—12:20 at Poster Board Number: ED21F-2372

Skill and Taxonomy Characterization



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Distribution of Question Type:

- Most common type of question was multiple choice (86%), focused on remembering (64%) factual knowledge (81%)
- The high proportion of remembering questions was not due to a reliance on the multiple choice format as understanding and analyzing used this format more often

Graphical Literacy

- Computer programming was the most common skill (37%), but computational thinking was the least common (4%)
- Seismology skills were half as common as computing (17%), Although basic instructions are needed in tutorial assignments, no skill questions (20%) were a missed opportunity
- No skill questions had higher facility than our target (91%), while questions requiring 2+ skills were more challenging (82%) than 1

Skill Required to Answer a Question:

Definition	Example
Requiring students to perform multiple steps to arrive at the right solution	"What is the correct order of com- mands below to create, check, and then enter these sac and act1 directories?"
Requiring students' under- standing of programing lan- guage and applications to an- swer the question	"Which command would print just the earthquake depths stored in this file?"
Requiring students to partici- pate in an element of explora- tion or navigation within an outside source or reference in finding an answer	"Take a minute to use one of the many tools we have learned in this course to find out the rest of the information for this station, assuming we want data from a broadband, vertical channel, re- cording at 40 samples per second."
Requiring students to learn or rely on previous seismology knowledge to arrive at the cor- rect solution	"Why should we not use a latitude range that goes all the way from -90 to 90?"
Requiring students' under- standing of math within a con- text	"If there are 200 points with an interval of 0.01 seconds, what is the total length of the signal?"
Requiring students' under- tanding of a graph or visual de- piction	"What is the general trend in the ampli- tude vs. frequency plot?"

Conclusions

- <u>Clear Improvement</u>: Mean scores improved from pre-test to post-test across both 2023 and 2024, showing effective student learning gains.
- Survey feedback aligns with these findings, with 69% of students reporting that they learned "A lot" or "A great deal" in seismology, and 76% expressing similar growth in scientific computing skills.
- **Reliable Findings**: Robust sample sizes in both years indicate results are representative of the student population. Meaningful Growth: High normalized gains and moderate to large effect sizes indicate that the instructional approach significantly boosted understanding, especially in areas with lower initial scores.
- This is supported by survey responses, with 88% of students agreeing that the workshop enabled clearer understanding of seismology concepts, and 96% noting that it improved their scientific computing abilities. · Aligns with surveys showing high normalized gains for scientific computing skills, reflecting growth in self-
- perception of computing capabilities. Minimal gains for our control validates the survey as a measure of impact. Analyzing vs. Remembering: "Analyzing" questions generally saw higher gains and effect sizes than "Remembering" questions, suggesting students improved more in higher-order thinking skills. • Consistent Impact: Similar gains in both years suggest a consistently effective instructional approach across cohorts.

Resulting Changes from Previous Analysis

- Investigating Bloom's taxonomy and skill usage in our assignments identified that our active e-learning design had less higher order learning and balanced skill development than we sought.
- · Changes enacted to implement more higher order thinking and multi-skill requirements. Learning analytics presents an excellent opportunity as courses become more digital to achieve evaluation of instructional design to ensure alignment with student learning outcomes and author designed goals.



completed for each tutorial. The y axis is the averaged taxonomy level across each tutorial in module. 1 is remembering, 2 is understanding, 3 is applying and 4 is analyzing. The thin line shows the averaged taxonomy for each percent of tutorial and the thicker straight line shows the trend of that data. Light green is the taxonomy for 2022 and blue is the taxonomy for 2023. We worked to reduce prompting and increase taxonomy as students progress within modules and from module to module.

2022-2024 Surveys

Which of the following best describes how much scientific computing you learned from this workshop?



The workshop design developed my seismology abilities and skills



Which of the following best describes how much seismology content you learned from this workshop?



Figure 5. Skill requirement changes (2022 to 2023) for (a) number of skills and (b) auestions addressing each skill. We removed excessive no skill questions when possible



and added skills to questions that were previously lacking

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ormalized Gains (%) based on Post compared to Registration for the same user: Туре Normalized Gain 28.4% Linux GMT 26.7% 5.1 4.9 33.8% 30.2% Python 6.1 28.0% Jupyter Matlab (Control) 4.6 13.6%

ore Scale	1	2	3 4		5
escription	Not at all interested/ prepared	Not so interested/ prepared	Somewhat interested/ Very interested/ prepared prepared		Extremely interested/ prepared
	Category	Be	efore Survey	After Survey	Normalized Gain
Stud	dent Interest Av	eraged	3.76	4.29	42.4%
nterest	in Seismology/G	Seophysics	3.97	4.4	41.8%
Interest in Scientific Computing			3.65 4.29		47.1%
Interest in Grad School			3.67	37.5%	
Interest in Employment			3.76 4.3		43.3%
Studen	It Preparedness	Averaged	2.85	3.65	37.3%
Preparedness for Grad School			2.85	3.69	39.1%
Preparedness for Employment			2.84	3.61	35.5%

2023-2024 Evaluation of Un-Paired Tests

post-tests indicates improvement in student performance across the assessments, d pite the unpaired nature of the data 2023-2024 Unpaired Pre and Post



Statistical Analysis: Independent t-test (t = -9.68, p = 2.04 x 10⁻¹⁸) and Mann-Whitney U test (U = 6106.0, p = 2.16×10^{-17}) both show a significant difference from pre- and posttest averages.

2023-2024 Evaluation of Paired Skill Gains



st-tests indicates improvement in student performance across the assessmer



Figure 8. Combined skill gains for 2023 and 2024 cohorts (total of 194 students) during the SSBW valuated through matched pre- and post-test questions. List blue is the pre-test and dark blueshows the post-test. Students that scored below average on either the pre-test or the post-test are to the right. Students who scored above average or average on both the pre-test and the post-test are in the middle. All students are shown to the left.

Skill Domain	Taxonomy	Pre-Mean	Post-Mean	Cohen's D	Pre-Test Std.	Post-Test Std.	Normalized Gain	n (Students)	# of Questions
Quantitative Literacy	Remembering	96.9	96.9	0	17.4	17.4	0.0%	194	11
Quantitative Literacy	Analyzing	66.0	79.4	0.30	47.5	40.6	39.4%	194	15
Graphical Literacy	Remembering	96.4	92.8	-0.16	18.7	25.9	-100.0%	194	103
Graphical Literacy	Analyzing	64.9	75.8	0.24	47.8	43.0	30.9%	194	45
Computational Thinking	Remembering	80.4	75.8	-0.11	39.8	43.0	-23.7%	194	26
Computational Thinking	Analyzing	42.3	64.4	0.46	49.5	48.0	38.4%	194	9
Navigation and Exploration	Remembering	90.7	92.8	0.08	29.1	25.9	22.2%	194	63
Navigation and Exploration	Analyzing	35.1	68.6	0.71	47.8	46.6	51.6%	194	6
Computer programming	Remembering	71.6	82.5	0.26	45.2	38.1	38.2%	194	232
Computer programming	Analyzing	55.2	80.4	0.56	49.9	39.8	56.3%	194	13
Seismology Content	Remembering	54.1	72.2	0.38	50.0	44.9	39.3%	194	53
Seismology Content	Analyzing	68.0	79.4	0.26	46.8	40.6	35.5%	194	62
Python	Remembering	63.4	83	0.45	48.3	37.7	53.5%	194	72

show post-test averages, with error bars indicating standard error. Dark blue circles indicate Cohen's D effect sizes, and pink circles represent normalized gain values.







Most questions show increased mean scores by approximately 10-20% from pre-test to post-test, indicating better student performance after instruction.

• Moderate to large Cohen's D values on many questions highlight significant improvements, with some areas like analyzing navigation and analyzing computer programming showing particularly strong gains.

• Post-test scores often have lower standard deviations, suggesting more consistent understanding among students after instruction. Only three questions have higher standard deviation with the two greatest being within the remembering taxonomy.

Normalized Gain: Many questions display moderate to high gains, showing effective learning, though some areas may need targeted reinforcement.

Test	Mean	Median	Standard Deviation	n	Cohen's D	Normalized Gain
Pre-Test	65.1	69.2	17.5	769	0 87	44.9%
Post-Test	80.8	84.6	18.7	241		

Statistical Analysis: Paired t-test (t = -5.05, p = 2.06 x 10^{-6}) and Wilcoxon signed-rank test $(W = 705.0, p = 8.51 \times 10^{-7})$ both show a significant improvement from pre- to posttest averages.

Test	Mean	Median	Standard Deviation	n	Cohen's D	Normalized Gain
Pre-Test	68.1	69.2	17.3	194	0.00	38.3%
Post-Test	80.3	84.6	18.2	194	0.09	

• Assessed skill gain in 82 students during the 2023 SSBW and 112 students during the 2024 SSBW using matched pre-/ post-test questions.

• Analyzing questions generally exhibit higher normalized gains compared to remembering questions.

• Strong gains observed in programming and seismology skills.

• Despite fewer higher-order thinking questions, students show significant improvement in analyzing skills, suggesting that the active e-learning pedagogy effectively supports the development of higher-order skills.



Pre-mean Post-mean Ochen's D Normalized Gain

Figure 9. Combined skill gains for 2023 and 2024 cohorts (total of 194 students) during the SSBW, evaluated through matched pre- and post-test questions. Average question scores are displayed for each skill and cognitive level, divided into "Remembering" and "Analyzing" skills across various domains (Quantitative, Graphical, Computational Thinking, Navigation, Programming, Seismology, and Python). Light blue bars represent pre-test averages, while dark blue bars