Practical Rubrics for Informal Science Education Studies: (1) a STEM Research Design Rubric for Assessing Study Design and a (2) STEM Impact Rubric for Measuring Evidence of Impact

Bobby Habig

Informal learning institutions, such as museums, science centers, and community-based organizations, play a critical role in providing opportunities for students to engage in science, technology, engineering, and mathematics (STEM) activities during out-of-school time hours. In recent years, thousands of studies, evaluations, and conference proceedings have been published measuring the impact that these programs have had on their participants. However, because studies of informal science education (ISE) programs vary considerably in how they are designed and in the quality of their designs, it is often quite difficult to assess their impact on participants. Knowing whether the outcomes reported by these studies are supported with sufficient evidence is important not only for maximizing participant impact, but also because there are considerable economic and human resources invested to support informal learning initiatives. To address this problem, I used the theories of impact analysis and triangulation as a framework for developing user-friendly rubrics for assessing quality of research designs and evidence of impact. I used two main sources, research-based recommendations from STEM governing bodies and feedback from a focus group, to identify criteria indicative of high-quality STEM research and study design. Accordingly, I developed three STEM Research Design Rubrics, one for quantitative studies, one for qualitative studies, and another for mixed methods studies, that can be used by ISE researchers, practitioners, and evaluators to assess research design quality. Likewise, I developed three STEM Impact Rubrics, one for quantitative studies, one for qualitative studies, and another for mixed methods studies, that can be used by ISE researchers, practitioners, and evaluators to assess evidence of outcomes. The rubrics developed in this study are practical tools that can be used by ISE researchers, practitioners, and evaluators to improve the field of informal science learning by increasing the quality of study design and for discerning whether studies or program evaluations are providing sufficient evidence of impact.

Keywords: informal science education, museum education, research design, STEM, rubric design, evidence-based outcomes, out-of-school time
INTRODUCTION

Informal science education (ISE) programs can be important vehicles for facilitating interest in science, technology, engineering, and mathematics (STEM) (National Research Council, 2009; Young et al., 2017; Habig et al., 2018). Indeed, in the last few decades, multiple studies and evaluations have reported evidence that involvement in informal, out-of-school time (OST) STEM programs is linked to participants’ awareness, interest, and engagement in STEM majors and careers (e.g., Fadigán and Hammrich, 2004; Schumacher et al., 2009; Winkleby et al., 2009; McCreedy and Dierking, 2013). Because many of these studies and evaluations vary considerably in how they are designed and in the quality of their designs, it is often difficult to gauge whether the outcomes reported are supported with sufficient evidence (Institute for Learning Innovation, 2007). This is particularly a dilemma for studies of ISE programs because participation is voluntary and thus more fluid than formal programs, and there is also considerable variation in the number of contact hours between programs and among participants (Institute for Learning Innovation, 2007; National Research Council, 2009). Therefore, a continuous challenge faced by the ISE community is how to gauge whether and to what extent the outcomes reported by studies and evaluations are supported by evidence. To address this problem, the goal of this study was to develop user-friendly rubrics that can be used to assess research designs and STEM outcomes of ISE studies. These rubrics, in turn, can be used to discern whether individual research studies or program evaluations are providing sufficient evidence supporting claims such as increased awareness, interest, and engagement in STEM majors and careers.

Over the past decade, there have been multiple initiatives carried out by several national agencies, including the National Research Council, the United States Department of Education, and the National Science Foundation, with the goal of identifying characteristics of high-impact STEM programs (e.g., U.S. Department of Education, 2007; What Works Clearinghouse, 2008; National Research Council, 2011, 2013). Many of these agencies have established criteria for assessing a range of STEM outcomes including the mastery of twenty-first century skills (National Research Council, 2011), the implementation of Next Generation Science Standards (National Research Council, 2014), and the impact of teacher professional development programs on student achievement (Yoon et al., 2007). Additionally, the Committee on Highly Successful Schools or Programs for K-12 STEM Education established criteria for identifying the effectiveness of STEM-focused schools and associated student outcomes (National Research Council, 2011). In recent years, various agencies have turned their attention to ISE programs. For example, the U.S. Department of Education (2007), in a report by the Academic Competitive Council, advanced as a national goal to improve awareness, interest, and engagement of STEM careers in the context of informal education. Additionally, the Institute for Learning Innovation (2007) was charged with assessing the quality and strength of evidence of STEM outcomes of ISE programs and of its participants. Further work has also been completed by several national agencies outlining criteria that can be used to identify effective OST STEM projects (Friedman, 2008; National Research Council, 2010, 2014, 2015; Krishnamurthi et al., 2014). Overall, there have been many strides made by science education stakeholders in terms of identifying characteristics of high-quality studies and for recommending criteria for assessing program outcomes. However, for studies of informal science projects, there remains a need for developing accessible methodologies for assessing the effectiveness of STEM interventions and the strength of the evidence of ISE research outcomes.

Due to the unique characteristics of informal learning environments, it is quite challenging to develop evidence-based criteria to assess whether a research study or evaluation has achieved specific outcomes (National Research Council, 2010). Informal science education, defined as voluntary participation in science during out-of-school time hours, typically occurs after school, on weekends, and during the summer in a variety of settings including but not limited to museums, zoos, universities, and, non-profit organizations (Blanchard et al., 2020). By design, ISE programs are voluntary, inquiry-based, and emphasize choice learning (National Research Council, 2009). Thus, for many programs, the random assignment of participants to treatment and control groups, often considered the gold standard in research design (What Works Clearinghouse, 2008), is often logistically infeasible, potentially upsetting to learners, and may jeopardize the validity of certain studies (National Research Council, 2009, 2015). An additional challenge is that the durations of many OST experiences are short-term making it difficult to measure program impact especially if the evidence of effects occur downstream of the experience (National Research Council, 2015). Furthermore, because many ISE OST programs are designed with the specific intent of differentiating from formal school programs, program leaders often avoid the administration of written assessments (National Research Council, 2015). Lastly, the Academic Competitive Council (U.S. Department of Education, 2007) recognizes that informal learning experiences are highly individualized, complex, and multifaceted, and suggest that due to the modest scale of many of these programs, they may not warrant a costly assessment approach. Therefore, the application of criteria typically used to assess the effectiveness of outcomes of formal science programs and participants (e.g., What Works Clearinghouse, 2008) might not be feasible for assessing the effectiveness of research studies or program evaluations designed to measure outcomes of participants of ISE programs. Promisingly, in recent years, many ISE stakeholders have developed rigorous research designs that are alternatives to random control trials including the employment of mixed methods and triangulation designs (Institute for Learning Innovation, 2007; Flick, 2018a,b). Nonetheless, the unique nature of ISE programs must be accounted for when developing methodologies for assessing the effectiveness of STEM interventions in an informal setting.

Despite the inherent challenges of assessing the effectiveness of research studies and program evaluations of ISE participants, knowing whether the outcomes reported by these studies are supported with sufficient evidence is important for several
reasons. First, there is an economic justification for gauging what works and what doesn’t work because many ISE institutions, non-profit organizations, and governmental agencies invest considerable monetary and human resources to support informal STEM education initiatives (Wilkerson and Haden, 2014). Therefore, to convince funding agencies, policy makers, and the public at-large that investing in ISE OST programs is important, the ISE community needs to show that these programs are helping young people to persist in STEM (U.S. Department of Education, 2007; Wilkerson and Haden, 2014). Second, the use of user-friendly tools can serve as a guide to inspire researchers to design rigorous studies that maximize evidence-based outcomes (Institute for Learning Innovation, 2007; Panadero and Jonsson, 2013). Consequently, program leaders can more confidently use the information from these research studies to revise, redesign, and continuously improve their programs. Lastly, as more programs provide evidence of high impact, the ISE community can extract program design principles from highly effective programs and where appropriate, ISE OST program leaders can adopt and adapt these principles across institutions (Klein et al., 2017).

One area of interest by STEM stakeholders related to assessment is how and to what extent ISE programs augment participants’ STEM major and STEM career outcomes (e.g., Cuddeback et al., 2019; Chan et al., 2020). Indeed, over the past decade, a major goal set forth by the National Research Council and the United States Department of Education is to inspire and motivate students to consider a STEM pathway (U.S. Department of Education, 2007; National Research Council, 2011, 2013, 2015). In the context of informal education and outreach, the Academic Competitiveness Council (U.S. Department of Education, 2007) identified increased awareness, interest, and engagement in STEM majors and careers as priorities; each outcome is defined below:

- **STEM major awareness**: increased knowledge and awareness of the various STEM disciplines available as fields of study at institutions of higher education
- **STEM major interest**: increased curiosity, motivation, and attention toward a STEM discipline as a focus of study at an institution of higher education
- **STEM major engagement**: a formal commitment to a STEM discipline as a focus of study in an institution of higher education
- **STEM career awareness**: increased knowledge and understanding of various STEM professions
- **STEM career interest**: increased curiosity, motivation, and attention toward STEM professions
- **STEM career engagement**: employment in a STEM profession

In response to this challenge, many ISE programs have provided outreach and programming specifically designed to augment students’ awareness, interest, and engagement in a STEM pathway. Moreover, hundreds of studies and evaluations have been carried out to assess participants’ outcomes. Unfortunately, a tool to assess research design and to test whether these studies are supported with sufficient evidence is lacking, hence the focus of this study.

Our understanding of ISE OST programs is derived from two forms of published knowledge—studies that have been published in peer-reviewed journals and studies that are the result of internal and external program evaluation (National Research Council, 2015). Peer review is considered the cornerstone of academic research because research methods and findings are subject to critical examination by experts within a discipline. Peer-reviewed studies are essential for answering scholarly questions and for communicating meaningful research (Gannon, 2001). As an alternative to peer review, internal and external program evaluations are also valuable for documenting program outcomes and for informing stakeholders on how to improve program design. Of these two forms of evaluation, external evaluations are often preferred by stakeholders because during the evaluation process, a program of interest is subject to independent analysis by an objective third party; however, external evaluations are typically more expensive than internal evaluations (U.S. Department of Education, 2007; National Research Council, 2015). Of the many types of program evaluations, the two most common conducted in ISE research are formative and summative. Formative evaluations typically occur during the developmental stage of a program and are particularly informative for improving program design. Summative evaluations are conducted after the completion of a program and are useful for assessing whether the program outcomes align with the project goals and objectives (Institute for Learning Innovation, 2007). The choice to conduct peer-reviewed research or an internal or external evaluation depends on many factors including available financial resources, the nature and duration of the program, and the goals of the stakeholders. Regardless of which form of published knowledge is selected, it is critical that research studies and program evaluations are rigorously designed to ensure the validity of research outcomes.

The aim of this study was to develop user-friendly rubrics to assess research design and to gauge whether a research study or program evaluation provided sufficient evidence to support specific claims (e.g., increased awareness, interest, and engagement in STEM majors and careers). To accomplish this goal, first I reviewed what experts consider to be evidence of high-quality research design and evidence of impact. Based on these research-based recommendations, I created a STEM Research Design Rubric and a STEM Impact Rubric tailored specifically for quantitative, qualitative, and mixed methods studies and evaluations of ISE OST programs. Second, I tested these rubrics for user-friendliness, reliability, and validity. Based on feedback from STEM researchers and practitioners, I made revisions to the rubrics when appropriate. Lastly, I assessed specific ISE OST studies and evaluations using these rubrics and provided case studies that illustrate how these tools can be used to evaluate research design and evidence of impact. Through this process, I developed practical tools that can be used by ISE researchers, STEM practitioners, and other stakeholders to assess the effectiveness of STEM interventions and evidence of research outcomes for both research studies and evaluations.
THEORETICAL FRAMEWORK

The theory of impact analysis is described as a “rigorous and parsimonious” framework for mapping the assessment of impacts (Mohr, 1995 p. 55). The theory, which stems from the seminal work of Campbell and Stanley (1963) and later refined by Mohr (1995), considers three basic experimental designs: (1) experimental; (2) quasi-experimental; and (3) retrospective. In a true experimental design, the researcher sets up one or more subjects to receive the treatment (participation in the program) and another group in which one or more subjects do not receive the treatment (the control). As a benchmark, the control and treatment groups are assigned randomly, and an adequate number of subjects are assigned to participate in the study. As a best practice, the theory of impact analysis recommends when possible, the use of larger treatment and control groups in order to help increase the sensitivity of a study. For example, the Institute for Learning Innovation (2007), while assessing the impact of different program evaluations of various ISE programs, defined an adequate sample size as a minimum of 50 subjects. Two examples of a true experimental design are the pre-test, post-test design with random assignment and the post-test only design with random assignment (Table 1). The second design considered in the theory of impact analysis is the quasi-experimental design. In the quasi-experimental design, researchers set up intervention and comparison groups, but the groups are not assigned randomly. If the study is designed so that comparison groups are closely matched in key characteristics, evidence suggests that a quasi-experimental study can yield strong evidence of the intervention’s impact. Examples of quasi-experimental studies include post-test only designs, comparative change, and comparative time series (Table 1). Lastly, in a retrospective design, also known as ex post facto (“after the fact”), the subjects that received treatment were not assigned by the experimenter. For example, if some youth register for an ISE program and others do not, the selection of participants was not assigned by the researcher; rather, the participants underwent self-selection.

In alignment with the theory of impact analysis, the U.S. Department of Education (2007) proposed a “Hierarchy of Study Designs for Evaluating the Effectiveness of a STEM Educational Intervention” consisting of three hierarchical levels—(1) experimental designs (randomized controlled trials), (2) quasi-experimental designs (well-matched comparison groups), and (3) other designs (e.g., pre/post studies, comparison groups without careful matching). In this hierarchy, a well-designed randomized controlled trial is the preferred method; quasi-experimental designs are preferred when experimental designs are not feasible, and other designs are considered when the first two designs are not feasible. Thus, according to the theory of impact analysis, the best designed studies are those that exhibit internal validity (i.e., studies that can make a causal link between treatment and outcomes) and those that exhibit external validity (studies that are generalizable to other programs and populations). Threats to internal validity include non-equivalent comparison groups, non-random assignment of subjects, and confounded experimental treatments (Fuchs and Fuchs, 1986). Threats to external validity include the timing of the study, the setting in which it occurred, the study subjects, and treatment conditions (Mohr, 1995). The theory of impact analysis thus describes a framework for reducing threats to internal and external validity and for developing experimental designs that measure the extent to which a study provides evidence of impact.

One limitation of the theory of impact analysis is that it might underestimate the power of qualitative studies. Qualitative studies are important because they emphasize depth of understanding and provide rich information on how participants have interpreted their STEM experience (Diamond et al., 2016). Because qualitative research relies on open-ended questions and in-depth responses, researchers often use this approach to identify patterns and to develop emerging themes (Jackson et al., 2007). According to Creswell and Poth (2018), high quality qualitative studies share common characteristics including the employment of rigorous methodological, data collection, and data analysis protocols, and the incorporation of one of the five following approaches to qualitative inquiry: (1) a narrative study, (2) a phenomenological study, (3) a grounded theory study, (4) an ethnographic study, and (5) a case study (Table 1). When qualitative studies are rigorously designed, they can be effective means for probing how ISE OST experiences impact participants’ STEM pathways (U.S. Department of Education, 2007; National Research Council, 2015). One way to include qualitative research in an impact analysis is to allow for the combination of quantitative and qualitative data, which is the basis of the theory of triangulation (Greene and McClintock, 1985; Creswell and Poth, 2018).

The theory of triangulation is based on the idea that the collection of multiple sources of quantitative and qualitative data helps to increase study validity allowing researchers to gain a more complete picture of participants’ outcomes (Denzin, 1970; Ammenwerth et al., 2003; Flick, 2018a,b). The concept of triangulation can be traced to Campbell and Fiske (1959), who developed the idea of “multiple operationalism,” which argues that the use of multiple methods ensures that the variance reflects the study’s outcomes and not its methodology. Triangulation has also been described in the literature as convergent methodology, convergent validation, and mixed methods (Hussein, 2009). The triangulation metaphor stems from military and navigational strategy where multiple reference points are used to identify an object's exact position (Smith, 1975). By applying basic principles of geometry, multiple perspectives allow for greater accuracy. Similarly, researchers can improve the accuracy of their interpretations by collecting multiple sources of data assessing the same phenomenon (Jick, 1979). Together, the theory of impact analysis and the theory of triangulation are helpful lenses for informing the development of a STEM Research Design Rubric and a STEM Impact Rubric.

METHODS

The primary objectives of this study were to develop a user-friendly STEM Research Design Rubric and a STEM Impact Rubric for ISE researchers, evaluators, and practitioners. To do
### TABLE 1 | Study designs for assessing the effectiveness of a STEM intervention.

| Quantitative study design            | Examples                                      | Diagram          | Advantages/Disadvantages                                                                 |
|-------------------------------------|-----------------------------------------------|------------------|-----------------------------------------------------------------------------------------|
|                                     |                                               |                  | X = STEM intervention                                                                  |
|                                     |                                               |                  | C = comparison group                                                                   |
|                                     |                                               |                  | T = treatment group                                                                    |
|                                     |                                               |                  | R = random assignment                                                                  |
|                                     |                                               |                  | O = outcome measure/evidence                                                            |
|                                     |                                               |                  | NA = not applicable                                                                     |
|                                     |                                               |                  |                                                                                            |
| **Experimental design**             |                                               |                  |                                                                                            |
| **(randomized controlled trials)**  | 1. Pre-test, post-test design                  | T(R): OXO        | 1. Reduces threats to internal validity/doesn’t control for effect of pre-test          |
|                                     | with random assignment                        | C(R): OO         | 2. Controls for pre-test effects/doesn’t measure change over time                      |
|                                     | 2. Post-test only with random assignment      | T(R): XO         | 3. Strongest quantitative design for reducing threats to validity                        |
|                                     |                                               | C(R): O          |                                                                                            |
|                                     | 3. Solomon four group design                  | T(R): OXO        |                                                                                            |
|                                     | *(Solomon, 1949)*                            | C(R): OO         |                                                                                            |
|                                     |                                               | T(R): XO         |                                                                                            |
|                                     |                                               | C(R): O          |                                                                                            |
| **Quasi-experimental designs**      |                                               |                  |                                                                                            |
| **(well-matched comparison groups)**| 1. Pre-test, post-test design                  | T: OXO           | 1. Reduces threats to internal validity/doesn’t control for effect of pre-test          |
|                                     | with comparison group                         | C: OO            | 2. Assesses change over time/non-random assignment increases threats to validity         |
|                                     | 2. Post-test only with comparison group       | T: XO            | 3. Assesses longer term change/non-random assignment increases threats to validity       |
|                                     |                                               | C: O             |                                                                                            |
|                                     | 3. Time series with                          | T: OXXXO         |                                                                                            |
|                                     | comparison group                             | C: OOO           |                                                                                            |
| **Other quantitative designs**      |                                               |                  |                                                                                            |
| 1. Pre-test, post-test design       | 1. T: OXO                                    |                  | 1. Assesses change over time/lack of control increases threats to validity               |
| without comparison group            |                                               |                  | 2. Provides a snapshot/lack of control increases threats to validity                     |
| 2. Post-test only without comparison group | 2. T: XO                                    |                  | 3. Assesses longer term change/lack of control increases threats to validity             |
| 3. Time series without comparison group | 3. Example: T: OXXXO                          |                  |                                                                                            |
|                                     |                                               | C: OOO           |                                                                                            |
| **Qualitative study design**        |                                               |                  |                                                                                            |
| **Qualitative design**              | 1. Narrative study                            | 1. Researcher extracts themes from narratives of one or more individuals                  |
|                                     | 2. Phenomenological study                     | 2. Researchers study several individuals with shared experiences to analyze a phenomenon of interest |
|                                     | 3. Grounded theory study                      | 3. Researcher extracts data from interviews of ~20–60 individuals and uses systematic coding to develop a unified theoretical explanation |
|                                     | 4. Ethnographic study                         | 4. Researcher extracts themes by describing and interpreting patterns of a shared culture of group |
|                                     | 5. Case study                                 | 5. Researchers conduct an in-depth analysis of one or multiple cases                       |
| **Mixed methods design**            | Uses one or more quantitative and qualitative design | Researchers collect, analyze, and integrate quantitative and qualitative data             |
|                                     |                                               |                  | Allows for triangulation of data, which counteracts disadvantages of individual designs/may be difficult to interpret if there are conflicting outcomes |

so, first I identified national agencies and STEM governing bodies comprised of experts specifically charged with identifying criteria indicative of high-quality STEM research and study design. This was accomplished by searching documents in the Center for the Advancement of Informal Science Education repository (informalscience.org) and by contacting three distinguished ISE researchers (two university-based ISE researchers and one research director at a large ISE institution) for recommended sources of information. Through this process, seven agencies were identified as potential sources of information: (1) National
Research Council; (2) United States Department of Education; (3) National Science Foundation; (4) Institute of Learning Innovation; (5) Afterschool Alliance; (6) The PEAR Institute: Partnerships in Education and Resilience; and (7) What Works Clearinghouse. Second, I used Google Scholar and informalscience.org to identify and extract publications from these agencies with a specific focus on documents that provide research-based recommendations on how to assess research design and evidence of impact from studies and evaluations of STEM participants. Accordingly, the criteria that I identified to develop the STEM Research Design Rubric and STEM Impact Rubric were based on a synthesis of recommendations from published reports from multiple governing bodies. Importantly, these criteria stemmed from the counsel of leading ISE researchers, statisticians, and policymakers based on our current knowledge of best practices for research design and assessment. Thus, to ensure content validity, I adopted research-based recommendations from ISE stakeholders because these experts specifically considered the unique characteristics of ISE programs when making recommendations (e.g., Institute for Learning Innovation, 2007; U.S. Department of Education, 2007; Friedman, 2008; National Research Council, 2009, 2015), and I used the theories of impact analysis and triangulation to inform this process. Lastly, based on these recommendations, I developed rubrics that researchers, STEM practitioners, and other stakeholders can use to assess research design and whether a research study or program evaluation has provided sufficient evidence to support a claim made about a particular STEM outcome.

During the design of these rubrics, I constructed a rating scale for the STEM Research Design Rubric and another for the STEM Impact Rubric. For the STEM Research Design Rubric, the rating scale was divided into four different levels: (1) a rating of one was indicative of a study or evaluation in which there was a weak research design; (2) a rating of two was indicative of a study or evaluation in which there was an adequate research design; (3) a rating of three was indicative of a study or evaluation in which there was a strong research design; and (4) a rating of four was indicative of a study or evaluation in which there was an exemplary research design. For the STEM Impact Rubric, the rating scale was also divided into four levels: (1) a rating of one was indicative of a study or evaluation in which there was little or no evidence of impact; (2) a rating of two was indicative of a study or evaluation in which there was moderate evidence of impact; (3) a rating of three was indicative of a study or evaluation in which there was strong evidence of impact; and (4) a rating of four was indicative of a study or evaluation in which there was exemplary evidence of impact. A four-point rubric was applied because this rating scale is considered the gold standard in rubric design (Phillip, 2002) and is commonly applied by STEM stakeholders in a myriad of contexts (e.g., What Works Clearinghouse, 2008; Singer et al., 2012; The PEAR Institute: Partnerships in Education and Resilience, 2017).

After designing prototypes of the STEM Impact Rubric, I recruited eight ISE stakeholders to review the STEM Impact Rubric and to participate in a focus group. All eight participants are ISE educators; five are active STEM researchers with a PhD; three hold a Masters in a STEM-related field. Of the eight focus group participants, one is a director of research at one of the largest informal science education institutions in the world; two are postdoctoral fellows both with a background in informal science education and mixed methods research. The remaining five participants are program managers at various ISE institutions; three hold Master’s in museum science education, one of these three has a background in statistics and two in anthropology. The other two participants are directors of a high school science research mentoring program; both have PhDs in Biology. All eight focus group members have published ISE research, and in addition to their status as STEM stakeholders, are also intended users of the rubrics developed in this study.

The focus group members were provided with rubrics and asked to conduct an initial review and to test out the rubrics on a study or evaluation on the STEM outcomes of ISE participants. Each volunteer was given 1 week to review the document, and then invited to a focus group meeting to provide their feedback. During the focus group, I conducted a semi-structured group interview that included the following questions for discussion: (1) What are your impressions of the rubrics? (2) Do you think that this tool is user-friendly for ISE stakeholders (researchers, evaluators, practitioners)? Why or why not? (3) What changes or tweaks would you make so that these rubrics are more user-friendly? The focus group discussion was also guided by the theories of impact analysis and triangulation as the focus group participants used the hierarchy of study designs and mixed methods triangulation as a lens for assessing and practicing the rubrics (Denzin, 1970; Mohr, 1995; Ammenwerth et al., 2003; U.S. Department of Education, 2007; Flick, 2018a,b). While the focus group members did not have overlap with the experts who helped inform the initial rubric design, their feedback was important for improving the categories and metrics used for each rubric. Based on the recommendations of the focus group, I revised the rubrics and next tested the tools for reliability and validity.

To test for reliability, I worked with a graduate research assistant and together we used the STEM Research Design Rubric to rate 25 ISE studies (i.e., a combination of peer-reviewed studies, program evaluations, and conference proceedings) independently. Additionally, we used the STEM Impact Rubric to rate 47 outcomes independently. To assess consistency across raters (reliability), I calculated inter-rater agreement using both percent agreement (Lombard et al., 2002) and Cohen's kappa (Cohen, 1968). Percent agreement for the STEM Research Design Rubric was 92.0% and for the STEM Impact Rubric 80.9%. Cohen's kappa (k) for the STEM Research Design Rubric was 0.89, and for the STEM Impact Rubric 0.74. These measures were indicative of substantial agreement (Cohen, 1968). Lastly, to increase content validity of our rubric, we incorporated triangulation methods (Creswell and Miller, 2000; Creswell and Poth, 2018). Because the most valid assessments stem from "the collective judgment of recognized experts in that field" (Baer and McKool, 2014, p. 82), during triangulation, I extracted and synthesized data from multiple sources including recommendations from national agencies and STEM governing bodies comprised of experts specifically charged with identifying
Study Title:  
Authors:  

Use this worksheet to help you complete the STEM Research Design Rubric for quantitative studies.

1A. Does the study design include a random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group (e.g.: pre-test/post-test design with comparison; post-test only design with comparison)? If yes, go to 1B. If no, go to 2A.

1B. Is the study design grounded in a theoretical framework? If yes, go to 1C. If no, go to 2A.

1C. Does the sample size consist of a minimum of 50 subjects per group? If yes, this is an exemplary research design (rubric score: 4). If no, go to 2A.

2A. Does the study design include one of the following: random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group (e.g. pre-test/post-test, post-test only, and/or time series design(s) without comparison)? If yes, go to 2B. If no, go to 3A.

2B. Is the study design grounded in a theoretical framework? If yes, go to 2C. If no, go to 3A.

2C. Does the sample size consist of a minimum of 40 subjects? If yes, this is a strong research design (rubric score: 3). If no, go to 3A.

3A. Does the study design include one of the following: random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group (e.g. pre-test/post-test, post-test only, and/or time series design(s) without comparison)? If yes, go to 3B. If no, go to 4.

3B. Is the study design grounded in a theoretical framework? If yes, go to 3C. If no, go to 3C.

3C. Does the sample size consist of a minimum of 25 subjects? If yes, this is an adequate research design (rubric score: 2). If no, go to 4.

4. This study is a poor research design (rubric score: 1).

FIGURE 1 | STEM Research Design Rubric worksheet for quantitative studies.

criteria indicative of high-quality STEM research and from focus group participants consisting of experienced ISE STEM researchers and practitioners.

RESULTS

Focus Group Results

Four key recommendations were made by the focus group. First, the focus group recommended that I design six distinct rubrics separated into two categories: (1) three research design rubrics, one for quantitative studies, one for qualitative studies, and one for mixed methods studies, and (2) three evidence of impact rubrics, one for quantitative studies, one for qualitative studies, and one for mixed methods studies. Initially, I provided the focus group with three rubrics that combined research design and evidence of impact together. However, the focus group unanimously agreed that research design and evidence of impact are distinct criteria warranting separate analysis. Thus, based on this recommendation, in the final iteration, I developed six distinct rubrics to assess quantitative, qualitative, and mixed methods studies: three STEM Research Design Rubrics and three STEM Impact Rubrics. Second, to make the rubrics more user friendly, the focus group recommended that instead of presenting the four levels of evidence vertically, I should present these criteria horizontally. This recommendation was adopted in the final draft. Third, because some researchers might have difficulties using each rubric as a stand-alone tool, the focus group suggested that I develop a worksheet with a key to help guide researchers, practitioners, and evaluators through the assessment process when using each rubric. This recommendation was also adopted (see Figures 1–6). Lastly, there was some debate between focus group members on whether a research design not grounded in theory should automatically be rated as an adequate research design (a score of 2 out of 4). The focus group did not reach consensus with respect to this question and instead recommended that if I retain this rating in the final draft, then I should also emphasize in the Discussion that
Use this worksheet to help you complete the STEM Impact Rubric for quantitative studies. Before completing the rubric, you need to identify the STEM outcome that you are assessing in the study (e.g. STEM career interest). Because some studies use multiple criteria to assess an outcome, you may need to repeat this process for multiple instruments and then average the results at the end.

List all criteria used to assess the outcome of interest (e.g. survey question 1, survey question 2, open-ended question 1, open-ended question 2, etc.)

1. Is there a statistically significant difference between comparison (control) and treatment (program participants)? If yes, this is an example exemplary evidence of impact (rubric score: 4). If no (or if the study did not include a treatment and comparison group), then go to 2.

2. Is there a statistically significant difference between pre- and post-survey OR did a minimum of 75% of participants indicate higher than median (e.g. very likely; likely) outcome on post-program survey. If yes, this is an example of strong evidence of impact (rubric score: 3). If no, then go to 3.

3. Did 40-75% of participants indicate a higher than median (e.g. very likely; likely) outcome on post-program surveys OR was a STEM outcome maintained in comparisons of pre- and post-surveys? If yes, then this is an example of moderate evidence of impact (rubric score: 2). If no, then go to 4.

4. This is an example of little or no evidence of impact (rubric score: 1).

Repeat this process for each criterion and then average the results to attain your STEM Impact Rubric score.
As a key to guide researchers, practitioners, or evaluators through the process of assessing research design quality (Figure 1).

**A STEM Impact Rubric for Quantitative Studies**

A STEM Impact Rubric for quantitative studies (Table 3) was designed in accordance with research-based recommendations from ISE experts (Institute for Learning Innovation, 2007; U.S. Department of Education, 2007; Friedman, 2008; National Research Council, 2009, 2015). Based on these recommendations, a study outcome indicative of exemplary evidence of impact needs to report a statistically significant difference between comparison and treatment groups (U.S. Department of Education, 2007; National Research Council, 2013, 2015). Therefore, a study that measured STEM career interest (or some other outcome of interest) and found that program participants exhibited significantly higher interest than a well-matched comparison group would be an example of an outcome indicative of exemplary evidence of impact. Based on additional recommendations from ISE experts, I also developed criteria indicative of strong, moderate, and little or no evidence of impact for quantitative studies and evaluations (Table 3). Lastly, as recommended by the focus group, I developed a STEM Impact Worksheet for quantitative studies as a key to guide researchers, practitioners, or evaluators through the process of assessing the impact of specific outcomes (Figure 2).

**A STEM Research Design Rubric for Qualitative Studies**

A STEM Research Design Rubric for qualitative studies (Table 4) was designed in accordance with research-based recommendations from ISE experts (Institute for Learning Innovation, 2007; U.S. Department of Education, 2007; Friedman, 2008; National Research Council, 2013, 2015). In alignment with the theory of impact analysis (Mohr, 1995), I found that a qualitative study or evaluation (e.g., narrative study; phenomenological study; grounded theory study; ethnographic study) that provides evidence of exemplary research design is one that includes a control (well-matched comparison) and treatment group (program participants) selected randomly (experimental) or non-randomly (quasi-experimental) (U.S. Department of Education, 2007; National Research Council, 2013, 2015). ISE experts also recommend that a qualitative study or evaluation with evidence of exemplary research design should
be grounded in a theoretical framework (Institute for Learning Innovation, 2007) and consist of a sample size of twenty or more subjects per group (Creswell and Poth, 2018). Based on additional recommendations from ISE experts, I also developed criteria indicative of strong, adequate, and weak research design for qualitative studies and evaluations (Table 4). Lastly, as recommended by the focus group, I developed a STEM Research Design Worksheet for qualitative studies as a key to guide researchers, practitioners, or evaluators through the process of assessing research design quality (Figure 3).

**A STEM Impact Rubric for Qualitative Studies**

A STEM Impact Rubric for qualitative studies (Table 5) was designed in accordance with research-based recommendations from ISE experts (Institute for Learning Innovation, 2007; U.S. Department of Education, 2007; Friedman, 2008; National Research Council, 2009, 2015). Based on these recommendations, a qualitative study indicative of exemplary evidence of impact may demonstrate this in one of two ways: (1) a study or evaluation in which researchers identify the outcome of interest in the treatment but not the control as an emerging theme or (2) a study or evaluation in which a minimum of 75% of data (e.g., interview data, response to open-ended question) is indicative of the outcome of interest in the treatment but not the control; this difference needs to be statistically significant. Based on additional recommendations from ISE experts, I also developed criteria indicative of strong, moderate, and little or no evidence of impact for qualitative studies and evaluations (Table 5). Lastly, as recommended by the focus group, I developed a STEM Impact Worksheet for qualitative studies as a key to guide researchers, practitioners, or evaluators through the process of assessing the impact of specific outcomes (Figure 4).

**A STEM Research Design Rubric for Mixed Methods Studies**

A STEM Research Design Rubric for mixed methods studies (Table 6) was designed in accordance with research-based recommendations from ISE experts (Institute for Learning Innovation, 2007; U.S. Department of Education, 2007; Friedman, 2008; National Research Council, 2009, 2015). In alignment with the theories of impact analysis (Mohr, 1995) and triangulation (Campbell and Fiske, 1959), I found that a mixed methods study (i.e., a study or evaluation that incorporates both quantitative and qualitative analyses) provides exemplary evidence of quality research design if the study or evaluation meets the benchmarks for exemplary evidence described for quantitative (Table 2) and qualitative (Table 4) analyses. In terms of evidence of exemplary research design, the study or evaluation must also be grounded in a theoretical framework (Institute for Learning Innovation, 2007) and the sample size needs to be comprised of fifty or more subjects per group for the quantitative analysis and twenty or more subjects per group for the qualitative analysis (Institute for Learning Innovation, 2007; Diamond et al., 2016; Creswell and Poth, 2018). Based on additional recommendations from ISE experts, I also developed criteria indicative of strong, adequate, and weak research design for mixed methods studies and evaluations (Table 6). Lastly, as

![Figure 4](https://example.com/figure4.png)
recommended by the focus group, I developed a STEM Research Design Worksheet for mixed methods studies as a key to guide researchers, practitioners, or evaluators through the process of assessing research design quality (Figure 5).

**A STEM Impact Rubric for Mixed Methods Studies**

Lastly, a STEM Impact Rubric for mixed methods studies (Table 7) was designed in accordance with research-based recommendations from ISE experts (Institute for Learning Innovation, 2007; U.S. Department of Education, 2007; Friedman, 2008; National Research Council, 2009, 2015). In terms of exemplary evidence of impact, the mixed methods study or evaluation must demonstrate the following: (1) a quantitative analysis in which there is a significant difference between the comparison (well-matched control) and treatment (program participants) groups and (2) a qualitative analysis in which researchers identify the outcome of interest as an emerging theme in the treatment but not in the comparison (well-matched control) group or a qualitative analysis in which a minimum of 75% of data (e.g., interview data, response to opened question) is indicative of the outcome of interest in the treatment but not the comparison (well-matched control) group; this difference needs to be statistically significant. Based on additional recommendations from ISE experts, I also developed criteria indicative of strong, moderate, and little or no evidence of impact for mixed methods studies and evaluations (Table 7). Lastly, as recommended by the focus group, I developed a STEM Impact Worksheet for mixed methods studies as a key to guide researchers, practitioners, or evaluators through the process of assessing the impact of specific outcomes (Figure 6).

**Case Studies**

In the next section, I present case studies demonstrating how the STEM Research Design and STEM Impact Rubrics can be used to assess research design quality and measure evidence of impact for specific outcomes. In accordance with recommendations from the Academic Competitiveness Council (U.S. Department of Education, 2007), I identified three select studies that measure one of the following: STEM major or STEM career awareness, interest, or engagement. The first case study provides an example.
of how to use the rubrics to assess a quantitative study; the second case study provides an example of how to assess a qualitative study; the third case study provides an example of how to assess a mixed methods study. While these case studies focus specifically on awareness, interest, and engagement, a STEM practitioner may use these rubrics to assess any outcome of interest and to compare studies that assess comparable outcomes.

Case Study 1: Stanford Medical Youth Science Program

The Stanford Medical Youth Science Program is a biomedical pipeline program for high school students. The goal of this program is to diversify participation in the health professions (Winkleby, 2007). This 5 week residential summer program includes classroom-based workshops, anatomy and pathology practicums, hospital field placements, research projects, and college readiness advisement. In 2009, Winkleby et al. (2009) published a quantitative study of the STEM outcomes of program participants. Two specific outcomes measured in this study were whether alumni of this program (1) majored in a STEM discipline or (2) engaged in a STEM career. I used the STEM Research Design Rubric (Table 2) to assess quality of research design and the STEM Impact Rubric (Table 3) to measure evidence of outcome (engagement in a STEM major; engagement in a STEM career). This process is also depicted graphically in Figure 7A.

In terms of research design, first I examined evidence of exemplary research design (Table 2, column 1). Since the study did not include either a random or non-random comparison group, I moved on to the second column: strong research design. First, I checked the first bullet point in column two: Since the study was a quantitative design without a comparison group (post-test only), it met the first criterion for strong research design. Next, I checked the second bullet point in column two. Since the study was grounded in two theoretical frameworks: (1) Cognitive Apprenticeship (Collins et al., 1991) and (2) Situated Learning (Lave and Wenger, 1991), it met the second criterion for strong research design. Lastly, I checked the third bullet point in column two. Since the sample size was well-exceeded the minimum of 40, it met the third criterion for strong research design. Thus, based on the STEM Research Design Rubric, this study was classified as a strong research design.

In terms of evidence of outcome (engagement in a STEM major or STEM career), first I assessed criteria for exemplary evidence (Table 3, column 1). Since there was no statistical comparison between program participants and a comparison group, I moved on to the second column (strong evidence). The criteria for strong evidence of impact included two possible
TABLE 2 | STEM research design rubric for quantitative studies.

| (4) Exemplary research design | (3) Strong research design | (2) Adequate research design | (1) Weak research design |
|-------------------------------|---------------------------|-----------------------------|-------------------------|
| • random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group (e.g., pre-test/post-test design with comparison; post-test only design with comparison) | • random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group (e.g., pre-test/post-test, post-test only, and/or time series design(s) without comparison) | • random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group (e.g., pre-test/post-test, post-test only, and/or time series design(s) without comparison) | • random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group (e.g., pre-test/post-test, post-test only, and/or time series design(s) without comparison) |
| • study grounded in a theoretical framework | • study grounded in a theoretical framework | • study may or may not be grounded in a theoretical framework | • study may or may not be grounded in a theoretical framework |
| • minimum of 50 subjects per group | • minimum of 40 subjects | • minimum of 25 subjects | • <25 subjects or not reported |

Helpful shortcuts:
✓ If there is no comparison group, the starting value is 3
✓ If there is no theoretical framework, the starting value is 2
✓ The sample size represents the number of subjects analyzed in the study, not the number of program participants

Rubric Score: _________

Note: In order to receive a rubric score (4, 3, 2, or 1), a study must meet the criteria of all three items in a given column.

outcomes: (1) statistically significant difference between pre- and post-survey or (2) minimum of 75% of participants indicate higher than median (e.g., very likely; likely) outcome on post-program survey. Neither of these outcomes was found for STEM major engagement or STEM career engagement. Next, I moved on to the third column (moderate evidence of impact). A survey of alumni of the Stanford Medical Youth Science Program indicated that 57.1% were engaged in a STEM major, which met the criteria for moderate evidence (a favorable response by 40–75% of participants). However, only 33.1% of alumni were engaged in a STEM career; this met the criteria for little or no evidence (Table 3, column 4). Overall, this study provided evidence of a strong research design and moderate evidence of impact, in terms of engagement in a STEM major, and little or no evidence of impact in terms of STEM career engagement.

Case Study 2: The Source (Game Changer Chicago Design Lab, University of Chicago)
The Source is a 5 week summer program that uses alternative reality games for teaching engineering concepts to high school students (Gilliam et al., 2017). The weekly program includes 1 day of online activities off-campus and 3 to 4 days of on-campus activities focused on workshops in different STEM subject areas. Gilliam et al. (2017) published a qualitative study describing the STEM outcomes of high school participants of this program. To assess the quality of the research design of this study and to measure evidence of impact (in this case STEM career awareness), I used both the STEM Research Design Rubric for qualitative studies (Table 4) and the STEM Impact Rubric for qualitative studies (Table 5) as described below. This process is also depicted graphically in Figure 7B.

In terms of research design, first I examined evidence of exemplary research design (Table 4, column 1). Since the study did not include either a random or nonrandom comparison group, I moved on to the second column: strong research design. First, I checked the first bullet point in column two. Since the study was a qualitative design without comparison group, it met the first criterion for strong research design. Next, I checked the second bullet point in column two. Since the study was grounded in multiple theoretical frameworks, including Situated Learning Theory (Lave and Wenger, 1991), it met the second criterion for strong research design. Lastly, I checked the third bullet point in column two. Since 43 students were interviewed, the sample size well-exceeded the minimum of 15 subjects; thus, it met the third criterion for strong research design. Therefore, based on the STEM Research Design Rubric for qualitative studies, this study was classified as a strong research design.

In terms of evidence of outcome (in this case, STEM career awareness), first I used the STEM Impact Rubric for qualitative studies to assess criteria for exemplary evidence (Table 5, column 1). Since there was no comparison group in this analysis, I moved on to the second column and tested for strong evidence of impact. One criterion indicative of strong evidence of impact is if the researchers identify outcome of interest as an emerging theme during qualitative analyses. A theme that emerged from this study was “Mentoring and Exposure to STEM Professionals,” which provided evidence that participants became more aware of STEM career opportunities as a result of their experiences in the program. Thus, this study provides strong evidence of impact. Overall, this study provided evidence of a strong research design and strong evidence of outcome, in this case, STEM career awareness.

Case Study 3: The Lang Science Program (American Museum of Natural History)
The Lang Science Program is a comprehensive 7 year program for middle school and high school students facilitated by the American Museum of Natural History. The program takes place...
on alternating Saturdays during the academic year and for 3 weeks during the summer for a minimum of 165 contact hours per year. The program centers on the research disciplines of the Museum—the biological sciences, Earth and planetary sciences, and anthropological sciences. As students transition from middle school to high school, they increasingly engage in authentic research projects alongside scientists and educators as well as career and college readiness workshops (Habig et al., 2018). A mixed methods study of the Lang Program centered on STEM major engagement and STEM career awareness (Habig et al., 2018). To assess the quality of the research design of this study and to measure evidence of impact, I used both the STEM Research Design Rubric for mixed methods studies (Table 6) and the STEM Impact Rubric for mixed methods studies (Table 7) as described below. This process is depicted graphically in Figure 7C.
Table 5: STEM impact rubric for qualitative studies.

| Title of study or evaluation: STEM outcome under review |
|--------------------------------------------------------|
| List all method(s) used to assess the STEM outcome under review (e.g., survey question 1, survey question 2, etc.): |
| 1. 2. 3. 4. 5. 6. 7. 8. 9. 10. |

Directions:
1. Identify the STEM outcome under review (e.g., increased STEM career interest)
2. List all criteria used to assess the STEM outcome under review. For example, if there were four survey questions that evaluated STEM career interest, list all four questions (e.g., open-ended question 1, open-ended question 2, etc.)
3. Use the rubric to evaluate each criterion used to assess the STEM outcome under review starting from left (exemplary evidence) to right (little or no evidence)
4. Calculate the average rubric score by dividing the sum of all rubric scores by the number of criteria used to assess STEM outcomes

| (4) Exemplary evidence | (3) Strong evidence | (2) Moderate evidence | (1) Little or No evidence |
|------------------------|---------------------|-----------------------|--------------------------|
| • researchers identify outcome of interest as an emerging theme during qualitative analyses for treatment but not comparison OR minimum of 75% of data (e.g., interview data; responses to open-ended questions) indicate that the program impacted an outcome of interest in the treatment but not control, and there was a statistically significant difference between these two groups |
| • If there is no comparison group, starting value is 3 |
| • 40–75% of data (e.g., interview data; responses to open-ended questions) indicate that the program impacted STEM outcome of interest |
| • less than 40% of data (e.g., interview data; responses to open-ended questions) indicate that the program impacted STEM outcome of interest OR anecdotal evidence of STEM outcomes (e.g., handful of participants quotes, but no systematic analysis) OR no difference in emerging themes between treatment and comparison groups |

In terms of research design, first I examined evidence of exemplary research design by assessing the first bullet point in column 1 (Table 6). Since the study design met the criteria of the first bullet point (a quantitative analysis that included a comparison group), I moved on to the second bullet point in column 1. Since there was no comparison group for the qualitative analysis, this study did not meet the criteria for an exemplary research design. Therefore, I moved on to the second column: strong research design. Since the study met the criteria for the first two bullet points in column 2 (a quantitative analysis with non-random assignment (quasi-experimental) of treatment and comparison group; a qualitative design without comparison group), I next checked the third bullet point in column 2. Since the study was grounded in two theoretical frameworks: (1) communities of practice (Lave and Wenger, 1991) and (2) possible selves (Markus and Nurius, 1986); it met the third criterion for strong research design. Next, I checked bullet points five and six (sample sizes). Since the sample sizes exceeded the minimum of 40 subjects for quantitative analyses and 15 subjects for qualitative analyses, this study met the fifth and sixth criteria for strong research design. Therefore, based on the STEM Research Design Rubric for mixed methods studies, this study was classified as a strong research design.

In terms of evidence of outcome, first I assessed the first outcome of interest: STEM major engagement. Four quantitative criteria were used to measure STEM major engagement: (1) percentage of STEM majors; (2) percentage of STEM majors compared to college students nationally; (3) percentage of STEM majors compared to college students in New York City; (4) percentage of STEM majors compared to students who attended specialized science, mathematics, and technology high schools. Since 80.3% of alumni engaged in a STEM major, this outcome was indicative of strong evidence of impact (Table 7, column 2). Since the percentage of STEM majors of the Lang program was significantly higher than college students nationally and locally, these two outcomes were indicative of exemplary evidence of impact (Table 7, column 1). However, since there was no significant difference in the percentage of STEM majors when comparing Lang alumni to alumni of specialized science, mathematics, and technology high schools, this last outcome was indicative of little or no evidence of impact. After averaging the four outcomes, the mean STEM rubric score was 3 suggestive of strong evidence of impact overall. For the second outcome, STEM career awareness, a qualitative analysis was used to measure evidence of impact. Based on the STEM Impact Rubric (Table 7, column 2), one criterion indicative of strong evidence of impact is if the researchers identify outcome of interest as an emerging theme during qualitative analyses. A theme that emerged from this study was “Discovering Possible Selves,” which provided evidence that participants were exposed to and became more aware of STEM career opportunities as a result of their experiences in the program. Thus, this outcome was indicative of strong evidence of impact. Overall, this study provided evidence of a strong research design and strong evidence of outcomes in terms of STEM
major engagement and STEM career awareness, although the former outcome varied considerably based on the analysis used to measure impact.

### DISCUSSION

The purpose of this study was to develop user-friendly rubrics that can be used by ISE STEM researchers, practitioners, and evaluators to improve quality of research designs and to measure evidence of outcomes. By consulting with informal learning experts and by leveraging several sources of data from STEM governing bodies, non-profit organizations, and ISE institutions, I identified research-based recommendations on how to assess research designs and on how to measure evidence of impact. With the feedback of STEM practitioners and recommendations from STEM stakeholders, the end products were two types of rubrics for quantitative, qualitative, and mixed methods ISE studies: (1) a STEM Research Design Rubric and (2) a STEM Impact Rubric. These tools were found to be especially applicable for assessing informal learning studies designed to inspire and motivate students to consider a STEM pathway. Here, I discuss what was learned from this process, the general applicability of these rubrics, and their respective limitations.

When designing the STEM Research Design and STEM Impact Rubrics, I used the theories of impact analysis (Mohr, 1995) and triangulation (Denzin, 1970; Ammenwerth et al., 2003; Flick, 2018a,b) to inform my research. By espousing the principles of impact analysis in alignment with triangulation theory, this study allowed for ISE educators to carefully consider different hierarchical levels for designing effective studies, ranging from higher-level designs (i.e., experimental and quasi-experimental designs) to lower-level designs (pre/post studies; comparison groups without careful matching, etc.), while simultaneously considering the unique characteristics of ISE programs (National Research Council, 2009, 2010). Moreover, because many ISE studies use qualitative designs to assess participants’ outcomes, the application of impact analysis theory, coupled with research-based recommendations, were particularly instructional during the process of developing rubrics for qualitative study design and impact.

Based on these results, ISE practitioners might consider the inclusion of comparison groups in qualitative study designs (Mohr, 1995). For example, the use of open-ended survey questions or semi-structured interviews that qualitatively compare program participants to well-matched comparisons are practices that will help to increase the internal validity of ISE studies (Mohr, 1995). Lastly, the theories of impact analysis and triangulation were particularly informative when designing rubrics for mixed methods studies as these studies provide multiple sources of data, both quantitative and qualitative, and help to increase validity and paint a more complete picture of participants’ STEM outcomes (Denzin, 1970; Ammenwerth et al., 2003).

The STEM Research Design and STEM Impact Rubrics developed in this study are practical tools that can be used by ISE researchers, practitioners, and evaluators to improve

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### Table 6: STEM Research Design Rubric (Mixed methods studies)

| (4) Exemplary research design | (3) Strong research design | (2) Adequate research design | (1) Weak research design |
|-------------------------------|---------------------------|-----------------------------|-------------------------|
| • quantitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group (e.g., pre-test/post-test design with comparison; post-test only design with comparison) | • quantitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group (e.g., pre-test/post-test, post-test only, and/or time series design(s) without comparison) | • quantitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group | • quantitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) of treatment and comparison group OR quantitative design without comparison group (e.g., pre-test/post-test, post-test only, and/or time series design(s) without comparison) |
| • qualitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) using one of the following designs: narrative study, phenomenological study, grounded theory study, ethnographic study | • qualitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) using one of the following designs: narrative study, phenomenological study, grounded theory study, ethnographic study OR qualitative design without comparison group | • qualitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) using one of the following designs: narrative study, phenomenological study, grounded theory study, ethnographic study OR qualitative design without comparison group | • qualitative analysis: random assignment (experimental) OR non-random assignment (quasi-experimental) using one of the following designs: narrative study, phenomenological study, grounded theory study, ethnographic study OR qualitative design without comparison group |
| • study grounded in a theoretical framework | • study grounded in a theoretical framework | • study grounded in a theoretical framework | • study grounded in a theoretical framework |
| • quantitative analyses: minimum of 50 subjects per group | • quantitative analyses: minimum of 40 subjects per group | • quantitative analyses: minimum of 25 subjects per group | • quantitative analyses: < 10 subjects per group |
| • qualitative analyses: minimum of 20 subjects per group | • qualitative analyses: minimum of 15 subjects per group | | |
| Note: If there is no comparison group, starting value is 3 | | | |

Note: In order to receive a rubric score (4, 3, 2, or 1), a study must meet the criteria of all items in a given column.
the field of informal science learning. First, depending on the goals of a study, the STEM Research Design Rubric is a useful tool for designing a study and for deciding which of the three hierarchical levels of study design is most appropriate for a given study (U.S. Department of Education, 2007). In some cases, especially studies receiving external funding and that focus on long-term outcomes, it might be most appropriate to design an experimental or quasi-experimental study. For example, in the third case study presented in this paper (Habig et al., 2018), a quasi-experimental design was applied to compare STEM major engagement outcomes between museum program participants and comparison groups. If the research team that conducted this study had access to the STEM Research Design Rubric prior to conducting this study, the authors might have opted to design their study differently. Specifically, the researchers might have matched program participants to a comparison group with shared demographic characteristics and used propensity score analysis to compare STEM major outcomes between groups (Rosenbaum and Rubin, 1983; Hahs-Vaughn and Onwuegbuzie, 2006). Alternatively, prior to the study, the research team might opt to randomly select students to participate in the program via lottery and then compare the outcomes between the treatment and comparison groups (e.g., Hubelbank et al., 2007). In other cases, especially when funding is limited and the study is focusing on a short-term outcome, it might be appropriate to design a study using one of the lower hierarchical levels. For example, a study of a 1 day engineering outreach event for 4th–7th grade Girl Scouts used a pre/post study design to assess whether participation in this program increased participants’ awareness of engineering careers (Christman et al., 2008). Thus, for programs with short-term outcomes and/or little or no funding, it might be most appropriate to design quasi-experimental studies or more simple designs such as a pre/post study design. Practically speaking, studies that assess short-term outcomes, such as STEM major and STEM career awareness, might operate at lower hierarchical levels (e.g., pre/post study design); studies that assess intermediate STEM outcomes, such as STEM major and STEM career interest, might operate at middle hierarchical levels (e.g., quasi-experimental), and programs that assess long-term STEM outcomes, such as STEM major and STEM career engagement, might operate at higher hierarchical levels (e.g., experimental design) (Cooper et al., 2000; Wilkerson and Haden, 2014).

Second, while the STEM Research Design Rubric is useful for informing study design, the STEM Impact Rubric is an important

### TABLE 7 | STEM impact rubric for mixed methods studies.

| Directions: |
|---|
| 1. Identify the STEM outcome under review (e.g., increased STEM career interest). |
| 2. List all criteria used to assess the STEM outcome under review. For example, if there were four survey questions that evaluated STEM career interest, list all four questions (e.g., survey question 1, survey question 2, open-ended question 1, etc.). |
| 3. Use the rubric to evaluate each criterion used to assess the STEM outcome under review starting from left (exemplary evidence) to right (little or no evidence) |
| 4. Calculate the average rubric score by dividing the sum of all rubric scores by the number of criteria used to assess STEM outcomes |

| STEM Impact Rubric (Mixed methods studies) |
|---|
| **Title of study or evaluation:** |
| **STEM outcome under review:** |
| List all method(s) used to assess the STEM outcome under review (e.g., survey question 1, survey question 2, etc.): |
| 1. ____________________ 2. ____________________ 3. ____________________ 4. ____________________ 5. ____________________ |
| 6. ____________________ 7. ____________________ 8. ____________________ 9. ____________________ 10. ____________________ |

**Directions:**

1. Identify the STEM outcome under review (e.g., increased STEM career interest).
2. List all criteria used to assess the STEM outcome under review. For example, if there were four survey questions that evaluated STEM career interest, list all four questions (e.g., survey question 1, survey question 2, open-ended question 1, etc.).
3. Use the rubric to evaluate each criterion used to assess the STEM outcome under review starting from left (exemplary evidence) to right (little or no evidence).
4. Calculate the average rubric score by dividing the sum of all rubric scores by the number of criteria used to assess STEM outcomes.

**Average Rubric Score (sum of all rubric scores ÷ number of criteria used to assess STEM outcomes):**
A Graphical Depiction of Case Study 1

Winkleby, M. A., Ned, J., Ahn, D., Koehler, A., & Kennedy, J. D. (2009). Increasing diversity in science and health professions: A 21-year longitudinal study documenting college and career success. Journal of Science Education and Technology, 18(6), 535-545.

![Graphical depiction of case study 1.](A)

![Graphical depiction of case study 2.](B)

![Graphical depiction of case study 3.](C)

B Graphical Depiction of Case Study 2

Gilliam, M., Jagoda, P., Fabiyi, C., Lyman, P., Wilson, C., Hilt, B., & Bouris, A. (2017). Alternate reality games as an informal learning tool for generating STEM engagement among underrepresented youth: A qualitative evaluation of the source. Journal of Science Education and Technology, 26(3), 295-308.

![Graphical depiction of case study 2.](B)

C Graphical Depiction of Case Study 3

Habig, B., Gupta, P., Levine, B., & Adams, J. (2018). An informal science education program’s impact on STEM major and STEM career outcomes. Research in Science Education, 1-24.

FIGURE 7 | (A) Graphical depiction of case study 1. (B) Graphical depiction of case study 2. (C) Graphical depiction of case study 3.
companion tool for helping ISE researchers, practitioners, and evaluators assess evidence of impact. Critically, the more rigorous the research design (exemplary research design), the more likely you can trust the study's validity (Mohr, 1995). If a study provides evidence of exemplary research design, then the results of the STEM Impact Rubric can be used to more confidently claim evidence of outcome. Thus, the STEM Research Design Rubric is a practical tool for developing rigorous study designs and in turn, the STEM Impact Rubric is a practical tool for providing evidence of impact. Importantly, the use of these rubrics across different ISE studies will ensure consistency in research design and measurement of impact, and when applicable, the results of these rubrics can be used to refine programs especially when there is little or no evidence of impact. Notably, when a STEM Impact Rubric is informed by a study with evidence of strong or exemplary research design, these results can be used for reports to government officials and funding agencies and can be potential sources for increased funding.

Finally, while the development of STEM Research Design and STEM Impact Rubrics are critical for improving the field of informal science learning and for promoting consistency between studies, it should be noted that there are several limitations of these tools. First, if a study is not designed rigorously, then we cannot reliably infer evidence of outcome. In data science, the term “garbage-in, garbage-out” is used to describe a situation in which the quality of the output is linked to the quality of the input (Rose and Fischer, 2011). Analogously, evidence of STEM outcomes (based on the STEM Impact Rubric) is linked to the quality of study design (based on the STEM Research Design Rubric). Thus, without a well-designed study, it is virtually impossible to confidently infer evidence of impact. In support, the National Research Council (2013) recommends that results from non-rigorous study designs (e.g., pre/post study design; comparison group without careful comparison) are appropriate for refining hypotheses that can be used to inform more rigorous study designs conducted in the future; however, these results should not be interpreted as conclusive evidence of impact. A second limitation is that STEM stakeholders might not always agree on the criteria used to assess research design or evidence of impact. For example, in the present study, there were some disagreements among the focus group members about whether or not a research design needs to be grounded in a theoretical framework as recommended by the Institute for Learning Innovation (2007). Thus, it is important to emphasize that these rubrics should be viewed as heuristics—tools for aiding ISE stakeholders in evaluating research design and evidence of impact—and that in some cases, depending on the goals of a particular study, it might be appropriate to alter certain criteria. Lastly, a third limitation is the challenges of applying these rubrics universally. Some ISE studies are very complex consisting of varied analyses where it might be quite difficult to assess study design and impact. For example, the third case study (Habig et al., 2018) used four different methods to assess evidence of impact with respect to STEM major engagement. In two cases, the rubric indicated that there was exemplary evidence of impact; in one case, the rubric indicated that there was strong evidence of impact, and in the final case; the rubric indicated that there was little or no evidence of impact. While averaging the rubric score provided a rough estimate of overall impact, the choice of an inappropriate analysis might skew the results. One of the comparison groups in this quasi-experimental study was students who attended specialized science, mathematics, and technology high schools. While there were no significant differences in STEM major engagement between participants of the informal museum program and these high school students, this analysis might not accurately measure the outcome of interest—whether participation in the ISE museum program had an impact on participants’ decisions to major in STEM. Thus, it is critical that a study design is aligned to the outcome of interest and that the appropriate comparison group is considered carefully.

CONCLUSIONS

The STEM Research Design and a STEM Impact Rubrics developed in this study are potentially useful tools for ISE researchers, practitioners, and evaluators for improving study design and for assessing the effectiveness of STEM interventions. There are several possible applications for these rubrics especially in terms of areas of future research. First, in future studies, ISE researchers can measure whether large scale use of the STEM Research Design Rubric results in overall improvements in research design across the informal learning community. Second, researchers can measure whether the STEM Impact Rubric helps to inform program design and in turn, helps ISE institutions to improve specific outcomes such as persistence in STEM. More specifically, STEM researchers can assess studies yielding exemplary evidence of impact, extract program design principles from these studies, and share best practices across institutions. Critically, and as recommended by the National Research Council (2013), researchers can also track outcomes of interest by race, ethnicity, language status, and socioeconomic status to ensure that programs are effective across different populations. Lastly, these rubrics can be used in meta-analyses to quantitatively compare research design quality and evidence of impact across studies and outcomes of interest. In summary, the large-scale application of the STEM Research Design Rubric and the STEM Impact Rubric has the potential to transform research design quality and more confidently measure evidence of outcomes.

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

ETHICS STATEMENT

Ethical review and approval was not required for the study on human participants in accordance with the local legislation and institutional requirements. Written informed consent was not required to participate in this study in accordance with the national legislation and the institutional requirements.
AUTHOR CONTRIBUTIONS
The author conceived, designed, and wrote the manuscript for this study.

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