

# Examining Equity in an Elementary School Computer Science Intervention Using Component-Based Research

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DOI: 10.21585/ijcses.v5i1.121

## Abstract

This article reports on implementation of a problem-based learning intervention developed with the intention of finding time for computer science (CS) in the elementary school day. This study investigated differences in effects on students in particular socio-demographic groups using a quasi-experimental design. We first provide an overview of the perennial problem of group differences or “gaps” in student outcomes. Then we illustrate how, using component-based research (CBR), we moved beyond the question of whether the intervention worked, to focus on which *parts* of the intervention worked, for *whom*, and under what *conditions*. Using hierarchical linear modeling, this study draws from a sample of 16 elementary schools with 321 teachers and 5791 students in Broward County, Florida, the sixth largest school system in the United States. This study complements a previous paper (Century, Ferris, Zuo, 2020), which examined associations between intervention components and student outcomes by investigating how outcomes differ for students in different socio-demographic groups and whether the presence of particular intervention components amplify or reduce differences. Through CBR, our work illustrates that CS interventions which may appear to benefit students overall, may be less beneficial or even detrimental to particular groups.

**Keywords:** computer science, elementary, component-based research, equity, implementation research

## 1. Introduction

### 1.1 *Differences in Academic Outcomes by Group Exist and Persist*

Differences in academic outcomes between groups of learners are well-documented across racial, ethnic, and gender identities, and socioeconomic, disability, English proficiency, and migrant status (Celeste et al., 2019; Chmielewski, 2019; Gilmour, et al., 2019; Hung et al., 2020; Owens, 2018). Although limited publications report on group differences in computer science (CS) outcomes at the primary grade levels, these differences are clear in other subjects. In the US, the 2019 National Assessment of Educational Progress (NAEP) showed that White students scored 26 and 28 points higher than their Black peers in 4<sup>th</sup> grade and 8<sup>th</sup> grade reading assessments, respectively, and 25 and 32 points higher in 4<sup>th</sup> and 8<sup>th</sup> grade math assessments, respectively (NAEP, 2019). Group differences in achievement outcomes also exist between Hispanic and White students with 21 and 20 point differences for reading in 4<sup>th</sup> and 8<sup>th</sup> grades, respectively, and 18 and 24 point differences for 4<sup>th</sup> and 8<sup>th</sup> grade mathematics, respectively (NAEP, 2019). These racial and ethnic group differences can be explained, in part, by differential access to high quality education and inequitable learning experiences. In the US, students who are Black or Hispanic are more likely to attend high-

poverty schools (Orfield et al., 2016; Orfield & Lee, 2005) and be denied access to STEM and CS courses and inclusive pedagogy, which often results in segregated CS education and career pathways (Margolis et al., 2017). School poverty rates are also associated with the size of the difference in achievement by racial group and the rate at which these differences increase over time (Reardon et al., 2019).

Given the strong associations between achievement and school-level poverty, it is also important to consider achievement differences across income groups. One study found that for every \$1,000 increase in family income, there was an associated 2.1% increase in students' math scores and a 3.6% increase in students' reading scores (Dahl & Lochner, 2012). Additionally, academic outcome differences between students of high- and low-SES groups tend to increase as students get older (Caro, 2009). Unsurprisingly, the 2018 Program for International Student Assessment indicates that students from low SES backgrounds lag far behind more advantaged students in their reading, mathematics, and science performance (OECD, 2020).

Disparities in academic achievement between gender groups also persist. For instance, only 19% of female students scored at a proficient level on the science NAEP test compared to 25% of male students (NAEP, 2019). Similar patterns were observed for scores on the NAEP mathematics test, with only 38% of female students scoring at or above proficiency, compared to 44% of males (NAEP, 2019). Differences also emerged based on English-language proficiency levels. Regardless of grade level, English Language Learners (ELLs) performed lower on reading and mathematics assessments compared to native English speakers, and such differences were particularly pronounced in domains that require more use of the English language, such as reading and writing (Abedi, 2002; Fry, 2007).

### *1.2 Group Differences in Attitudes*

In STEM subjects, research on group differences in attitudes have had mixed results, and some group comparisons have yet to be investigated (e.g., ELLs and native English speakers). For instance, some studies indicate that Black students report lower science ability and mathematics beliefs compared to White students (Britner & Pajares, 2001; Pajares & Kranzler, 1995), whereas others have found that, compared with White students, Black students demonstrate higher mathematics attitudes (e.g., mathematics values), and Hispanic students exhibit higher mathematics interest and intrinsic motivation (Else-Quest et al., 2013; Stevens, Olivárez Jr, & Hamman, 2006). Another study found that Black and Hispanic boys specifically, reported lower levels of educational aspirations, persistence, and self-efficacy related to science compared to boys from more economically advantaged backgrounds, White boys, and girls from all backgrounds (Perry et al., 2012). Still other research indicates both Black and Hispanic students have a similar level of interest in STEM majors as their White counterparts (Anderson & Kim, 2006; Hurtado, et al., 2010). Moreover, female students generally report less motivation and confidence in their CS and mathematics-related skills and abilities to learn these subjects compared to male students (Beyer, 2014; Cunningham et al., 2015; Doubé & Lang, 2012; Hur et al., 2017; Maste et al., 2016). Girls and women are also more likely to show math and spatial anxiety (Ferguson et al., 2015; Maloney et al., 2012; Sokolowski et al., 2019). Together, this collection of results underscores the importance of pursuing more research on discrepancies in STEM-related attitudes between different groups of students.

### *1.3 The Importance of Addressing Group Differences in Computer Science*

The need to increase the number of learners interested in CS has never been greater (Scott et al., 2017; Vakil, 2018). Computing knowledge is essential to address the growing demand for CS skills in STEM jobs, yet we must also acknowledge that women and underrepresented minorities earn far few degrees in computing fields at this writing (Sax et al., 2018). Moreover, discrepancies in access to technology in the home and CS educational experiences persist, and are particularly pronounced for Black and Hispanic students and youth from lower SES families (Wang et al., 2016). Together, these findings suggests that research attention must be paid to diverse learners' experiences with STEM and CS, teachers' and administrators' roles in broadening participation in CS, instructional practices facilitating inclusive practice and CS learning, and educational structures and policies contributing to equitable CS educational opportunities (Ryoo et al., 2019; Santo et al., 2019).

It is also essential to begin CS education in the early grades. Introducing young learners to even basic CS concepts (i.e., plugged and offline, unplugged lessons) is crucial during the elementary school years for several reasons. First, doing so helps form the foundation for future learning in the subject, and second, it allows students to experiment with a complex and challenging subject before they form opinions of their abilities (Prottzman, 2014). There is little research on CS education in the early grades and even less that examines which specific learning experiences contribute to diverse students' pursuit of CS. The current study addresses these gaps in the literature by exploring the CS experiences of diverse learners in the early grades. This work will further increase our understanding of how to promote educational equity in CS education as measured by academic and attitudinal outcomes.

#### 1.4 Approaches to Examining Group Differences in Outcomes

Prior research considering group differences in academic achievement and attitudes has employed a variety of research approaches. However, many studies simply *describe* differences in outcomes and studies that focus on comparing outcomes, often privilege certain socio-demographic groups, namely White, male students, over others (Gutiérrez & Dixon-Román, 2011). Such 'gap gazing' illuminates the presence of differences, but does not provide any insights as to the nature of disparities that can inform efforts to change them (Lubienski, 2008).

Research that goes beyond 'gap gazing' operates on the *assumption* that instructional practices, programs, and educational interventions will *not impact all students in the same way* because members of diverse populations have unique needs (Gutiérrez, & Jaramillo, 2006; Rodríguez et al., 2016). Research methods need to account for the nuances that happen when programs are implemented (Vossoughi et al., 2016). To do so, researchers must utilize diverse methodological approaches that move beyond a sole focus on "what works" (Bullock, 2012) to also consider what works *for whom and under what conditions*. Approaching the study of CS achievement and attitudinal outcomes in this manner will elevate the importance of achieving equitable outcomes across diverse groups of learners.

## 2. Material Studied

This research was carried out through a research-practice partnership (RPP) focused on finding time for CS in the elementary school day, which is often dominated by literacy and mathematics instruction. The strategy was to create interdisciplinary problem-based learning modules that would be taught during the literacy block. The modules were organized around a science or social studies theme and had science, English/Language Arts, and associated CS lessons. The CS lessons were part of the Code.org courses available worldwide.

We referred to the modules as "Time4CS" Modules (see Century, Ferris, & Zuo, 2020 for a more detailed discussion of the modules). During the study, third-fifth grade teachers used two Time4CS Modules during the 2016-2017 school year. Each module was taught over a 5-8 week period and was composed of five lesson "collections." Each collection had four 60-minute lessons written in the 5E Model (i.e., Engage, Explore, Explain, Extend, Evaluate; Bybee, 2009) and had at least one CS lesson. Each module ended with a culminating project that entailed using CS to create a final product highlighting what students learned.

Following a quasi-experimental design, the project's primary research questions explored outcome differences for students in classrooms where teachers implemented and did not implement the Time4CS Modules. These analyses found that teachers who used the Time4CS Modules taught more CS than teachers in the comparison group. However, no differences in attitudinal or academic achievement outcomes were observed between students in treatment and comparison groups (Century, Ferris, & Zuo, 2020). Because this was the case, we utilized *component-based research* (CBR) to conduct further, in-depth analysis to explore associations between *specific components* of the Time4CS Modules and student outcomes. Using CBR revealed positive associations between particular Time4CS Module components and students' attitudinal and achievement outcomes. For example, we found positive associations between teachers' use of interdisciplinary teaching practices and students' achievement outcomes (Century, Ferris, & Zuo, 2020). In the findings reported here, we used CBR to conduct follow-up analyses that move beyond the aforementioned findings to deeply examine differences in student outcomes based on socio-demographic characteristics *and* the effect of teachers' instructional practices on outcomes for students in different groups. To do so, we investigated two additional research questions:

1. Are there differences in students' attitudinal and academic achievement outcomes based on socio-demographic characteristics?
2. Does the implementation of particular Time4CS Module components amplify or reduce group differences in students' attitudinal and academic achievement outcomes?

### 3. Area and Sample Description

The RPP team included researchers from the University of Chicago and practitioners from Broward County Public Schools (BCPS) in Broward County, Florida. BCPS is the sixth-largest school district in the United States consisting of 234 schools (i.e., 136 elementary, 37 middle, and 33 high schools) with a diverse student population (i.e., 51.2% White, 40.4% Black, 34.7% Hispanic, and 3.8% Asian; 11.9% English Language learners and 13.4% Exceptional students [students with Individualized Education Plans (IEPs)]). Sixteen BCPS elementary schools with 321 teachers and 5791 students from diverse racial, ethnic, and socioeconomic backgrounds participated in the study (Table 1; Table 2). A randomized block design (e.g., Dunlap, Cortina, Vaslow, & Burke, 1996; Kirk, 2007) was used to assign schools to treatment (i.e., eight schools; use of the Time4CS module) or comparison (i.e., 8 schools; regular classroom instructional practice) conditions.

Table 1. Student demographic characteristics for the treatment and comparison conditions

Demographics		Treatment	Control
Gender	Male	1035	932
	Female	962	878
Race	White	958	854
	African American	418	345
	Multi-ethnic	331	321
	Asian	111	74
	Other	220	186
Ethnicity	Hispanic	1081	1012
	Non-Hispanic	1859	1835
Grade Level	3rd	775	549
	4th	672	711
	5th	726	687
SES	Students not receiving free and reduced price lunch	1064	1091
	Students receiving free and reduced price lunch	917	704
English language proficiency	Instructed on acquiring English as a second language	290	269
	Still being monitored/exiting the program to learn English	171	171
	Native English speaker	1393	1245
Experience with Code.org	Previous experience completing Code.org activities prior to completing the questionnaires	1784	1408
	No previous experience completing Code.org activities	93	198

Table 2. Teacher demographic characteristics for the treatment and comparison conditions

Demographics		Treatment	Comparison
Gender	Male	3	20
	Female	67	134
Race/Ethnicity	White	38	80
	Black or African American	8	24
	Hispanic or Latino/Latina	18	30
	Other	NA	4
	Prefer not to answer	3	10
	Multiethnic	3	9
Grade Level	3rd	31	62
	4th	17	49
	5th	22	51
Highest Degree Earned	Bachelor's Degree	43	81
	Master's degree	25	72
	Doctoral or Professional Degree (Ph.D., Ed.D., MD, JD)	2	6
Degree in CS	Yes	1	2
	No	69	156

*Note.* The average years of teaching experience for treatment group teachers was 16.86 years, and for comparison group teachers was 14.38 years. The average years of teaching CS at any grade level for treatment group teachers was 1.41 years, and for comparison group teachers was 1.91 years. The average age for treatment group teachers was 42.66 years, and for comparison group teachers was 41.19 years.

## 4. Methods

### 4.1 Theoretical Approach – Component-Based Research

This study used a component-based research (CBR) approach. CBR deconstructs an innovation into its components, organizes those components into categories, and uses the categories as a foundation for looking beyond *whether* the innovation “worked,” to examine *what parts* of an innovation work, *for whom*, and *under what conditions*. In this study, CBR was guided by an innovation implementation framework (Century, Cassata, Rudnick, & Freeman, 2012) that organized the components of the Time4CS Modules into two main categories —structural (e.g., number of CS lessons taught, lesson omission) and interactional (e.g., teacher facilitation of cognitively-demanding work, teacher facilitation of student intellectual risk taking). See details in Table 3. We also measured contextual factors hypothesized to influence teachers’ Time4CS Module implementation (e.g., teachers’ years of teaching CS, perception of their own innovativeness or resourcefulness) utilizing a framework that outlines factors that influence implementation (Century & Cassata, 2014). By using CBR, we were able to identify the impact of specific innovation components on students’ attitudinal and achievement outcomes *and* disentangle associations between specific components and outcomes for particular student groups. Employing CBR enabled us to move beyond “gap gazing” to generate greater understanding about potential ways to reduce group differences in attitudinal and achievement outcomes across diverse groups of students.

Table 3. Component-based research: Structural and interactional components

<b>Structural Components</b>	<b>Measure</b>
Time4CS Module Completion	Yes/No
Grade-level assigned Code.org CS lessons	Percentage of lessons taught from the Code.org Fundamentals course assigned to the grade
Non-grade-level Code.org CS lessons	Percentage of lessons taught from the other Code.org courses available (Above or below grade level)
Additional CS (Any other non-Code.org CS lessons)	Yes/No
<b>Interactional Components</b>	<b>Sample Items (5-point scale: never-always)</b>
Teacher facilitation of group work	How often did you encourage group members to solve problems together?
Teacher facilitation of cognitively demanding work	How often did you ask students to explain how they solved a problem?
Teacher facilitation of students taking intellectual risks	How often did you ask students to answer a question even if they were unsure?
Teacher use of interdisciplinary teaching practices	How often did you explicitly cover standards from multiple subject areas in the same lessons?
Teacher facilitation of group work	How often did you encourage group members to solve problems together?

#### 4.2 Measures and Data Sources

**Teacher Implementation Questionnaire:** This questionnaire, administered in spring 2017 using Qualtrics ©, assessed Time4CS Module structural and interactional component implementation and factors influencing implementation, such as teacher attitudes and demographic characteristics. All items and response scales included in the Time4CS Teacher Questionnaire are listed in Appendix A.

**Student Attitude Questionnaire:** Students completed questionnaires at the beginning and end of the 2016-2017 academic year using Qualtrics ©. These questionnaires assessed students' attitudes about school (i.e., school affinity, school self-efficacy) and CS (i.e., CS affinity, CS self-efficacy, CS identity, and CS utility) using items scored on a five-point Likert scale. These questionnaires also asked about previous experience with Code.org CS activities (i.e. yes/no) prior to the study. All items and response scales included in the Time4CS Student Questionnaire are listed in Appendix B.

**Florida Standards Assessment (FSA) and Achieve3000 scores:** Measures of student academic achievement included the Florida Standards Assessment (FSA) and a diagnostic tool BCPS used called Achieve3000. The state administers FSA assessments in ELA and mathematics in 3<sup>rd</sup>, 4<sup>th</sup> and 5<sup>th</sup> grades. The FSA science test is administered in 5<sup>th</sup> grade, only. The FSA assesses student knowledge and critical thinking skills. The Achieve3000 literacy test is administered three times a year to assess students' reading level as a means to support differentiated instruction. In both cases, BCPS de-identified and shared these data for the 2016-2017 academic year with the research team.

**Administrative Data:** BCPS also shared students' de-identified socio-demographic data, including gender, ethnicity (i.e., non-Hispanic and Hispanic status), race, ELL proficiency, and socio-economic status (i.e., free and reduced-price lunch [FRL]). Students also self-reported their gender, ethnicity, and race, and we cross-referenced these data with the district data to account for missing values.

#### 4.3 Using CBR to Measure Time4CS Module Implementation

Studies of CS and other school interventions do not typically measure intervention implementation and when they do, implementation measurement is only conducted with treatment groups. However, CBR recognizes that some intervention components may in fact be accessible to teachers *in the comparison group*, and therefore, must be measured as well. This was the case in this study. Specifically, the Code.org CS lessons associated with the Time4CS Modules were publicly available to *all* teachers in the district. Similarly, the Time4CS Modules' interactional components included instructional practices (e.g., facilitating group work, using interdisciplinary teaching practices) that any teacher can use, with or without the Time4CS Modules. Therefore, while teachers in the treatment condition were asked to report on implementation of *all* module components, teachers in the comparison condition were asked to report on the components (and related factors) they could have enacted even though they did not implement the modules.

**Structural implementation components:** Teachers in the treatment group reported whether they completed (i.e., yes/no) the Time4CS Module with their students. They also reported on how many module lessons they completed and the sections of those lessons they used or omitted. Moreover, because we used a CBR approach, teachers in *both* treatment and comparison groups reported the number of grade-level assigned *and* non-grade-level assigned Code.org CS lessons they completed as well as whether they completed (i.e., yes/no) any additional, non-Code.org CS activities. The number of grade-level assigned, and non-grade-level assigned Code.org CS lessons were converted into completion percentage scores by dividing the number of lessons teachers reported completing by the total number of lessons available to complete. Higher completion percentage scores represented greater CS exposure for students.

**Interactional implementation components:** All teachers reported on their use of interactional Time4CS Module components. These components entailed the instructional strategies embedded in the module, but as explained above, could also be carried out in “business-as-usual” (comparison) classroom instruction. The questionnaires asked teachers to report on use of interdisciplinary teaching practices (e.g., How often did you explicitly cover standards from multiple subject areas in the same lesson?) as well as their facilitation of group work (e.g., How often do you encourage group members to work to solve problems together?), cognitively-demanding work (How often did you ask students to explain how they solved a problem?), and intellectual risk-taking (e.g., How often did you ask students to answer a question even if they were unsure?). Each scale used a five-point Likert scale (1 = *Never*, 5 = *Always*) with higher scores representing greater use of each component.

**Influential Factors:** The factor framework mentioned in section 4.1 organizes influential factors (i.e., contexts and conditions) into several levels (i.e., individual, organizational, and environmental) (Century & Cassata, 2014). The questionnaires asked teachers to report on three *individual* (teacher) factors, including innovativeness (e.g., I experiment with new practices all the time), resourcefulness and coping (e.g., I am able to manage the pressure and stress at my school well), and years of teaching CS. Each scale used a six-point Likert scale (1 = *Strongly disagree*, 6 = *Strongly agree*) with higher scores representing greater use of these practices.

#### 4.6 Analytic Strategy

To conduct our analysis, we used Hierarchical Linear Modeling (HLM) with HLM7 software (Raudenbush, Bryk, Cheong, Congdon, & Toit, 2011). First, two-level, intercept-as-outcome models were conducted, which identified significant group differences in achievement and attitudinal outcomes based on students' socio-demographic characteristics (i.e., Level-1 variables)<sup>1</sup>. Next, two-level, random-intercept and coefficient-as-outcome models were conducted to examine the interactive effects between teacher-level implementation of particular interactional Time4CS Module components (i.e., specific instructional practice variables) and students' socio-demographic characteristics (Raudenbush & Bryk, 2002). In this second set of analyses, the Level-1 intercepts were no longer modeled as a function of structural or interactional component implementation *or* factors at Level-2 as in the first set of analyses; instead, they were included as random effects. More specifically, significant Level-1 coefficients for

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<sup>1</sup> In Century, Ferris, & Zuo (2020), we reported on associations between Level-2 variables and students' attitudinal and academic achievement outcomes.

students' socio-demographic characteristics identified in the first set of analyses were modeled as a function of structural or interactional component or factor variables at Level-2 in the second set of analyses.

In all analytic models, students' socio-demographic characteristic variables were entered at Level-1. These variables were dummy coded as follows: gender (reference group: male); race (reference group: White students); ethnicity (reference group: non-Hispanic students); SES (reference group: students who *do not* receive free and reduced-price lunch); and ELL proficiency (reference group: native English speakers). Students' previous experience completing Code.org activities was included at Level-1 as a control variable as were their socio-demographic characteristics. Students' attitudes about school and CS collected during pre-questionnaire administration (i.e., Fall 2016) were also included at Level 1 in models where attitudinal outcomes were examined. Level-2 teacher variables included structural and interactional implementation components and factors. All categorical variables were uncentered, whereas continuous variables were centered around the group mean for Level-1 variables and the grand mean for Level-2 variables.

## 5. Results

Significant findings are reported in text. Standardize beta coefficients and standard error terms are reported for all associations between students' socio-demographic characteristics (i.e., Level-1 variables) and their academic achievement and attitudinal outcomes are included in Table 4 and 5, respectively. In Table 6 and 7, we report all standardized beta coefficients and standard error terms for the interactive effects between students' socio-demographic characteristics (i.e., Level-1 variables) and structural and interactional components of Time4CS module implementation and factors (i.e., Level-2 variables)<sup>2</sup>. The significant group differences reported in-text and in Tables 4-7 emerged when controlling for all other student socio-demographic characteristic variables in the analytic model.

### 5.1 Academic Achievement Outcomes

Gender: On the science FSA, female students scored significantly lower than males ( $\beta = -5.52, p < 0.01$ ).

Race: On the mathematics FSA, Asian students ( $\beta = 5.92, p < 0.01$ ) scored significantly higher, and Black students ( $\beta = -8.42, p < 0.01$ ) scored significantly lower than White students. On the ELA FSA, Black students scored significantly lower than White students. This main effect was qualified by a significant interaction, which indicated that this score differential (i.e., gap) on the ELA FSA between Black and White students was **larger** in classrooms where teachers reported completing a **higher percentage of grade-level assigned CS lessons** (a structural component) ( $\beta = -34.69, p < 0.01$ ).

Table 4. Results from the two-level, intercept-as-outcome models investigating associations between level-1 variables and students' academic achievement outcomes

	Mathematics FSA	ELA FSA	Science FSA	Achieve 3000 scores
Asian students	5.92 (1.69)*	0.85 (1.79)	8.23 (3.60)*	9.71 (10.33)
African American students	-8.42 (1.48)*	-7.54 (1.59)*	-8.46 (2.52)*	-8.96 (7.99)
Hispanic students	-0.06 (1.16)	0.70 (1.07)	2.55 (1.81)	1.07 (5.29)
Female students	-0.77 (0.94)	0.78 (0.89)	-5.52 (1.55)*	-8.79 (4.47)*
Students receiving FRL	-5.70 (1.12)*	-5.80 (1.15)*	-6.58 (1.93)*	-31.03 (5.72)*

<sup>2</sup> Table 6 and 7 highlight interactive effects on Science FSA and CS Affinity. We have only included tables for these two outcomes due to space limitations. Tables illustrating interactive effects for additional academic achievement (i.e., Mathematics FSA, ELA FSA, and Achieve3000) and attitudinal (i.e., School Affinity, CS Utility, and CS Identity) outcomes are available from the authors upon request.



English language learners	-8.08 (1.60)*	-13.05 (1.79)*	-8.78 (3.03)*	-8.71 (8.48)
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Note. \*  $p < .05$ . Standardized beta coefficients and standard error (SE) terms are presented in the table with SE reported in parentheses.

Table 5. Results from the two-level, intercept-as-outcome models investigating associations between level-1 variables and students' attitudinal outcomes

	School Affinity	CS Affinity	CS Identity	CS Utility
Female students	0.01 (0.07)	-0.13 (0.08)	-0.10 (0.08)	-0.03 (0.07)
Asian students	0.22 (0.13)	0.01 (0.18)	0.28 (0.14)*	0.01 (0.14)
African American students	-0.09 (0.12)	-0.10 (0.14)	-0.08 (0.13)	-0.10 (0.11)
Hispanic students	-0.03 (0.07)	-0.09 (0.07)	0.03 (0.07)	0.07 (0.08)
Students receiving FRL	0.10 (0.08)	0.16 (0.08)*	0.16 (0.08)*	0.11 (0.07)
English-language learners	0.03 (0.11)	0.06 (0.11)	-0.08 (0.14)	-0.04 (0.1)

Note. \*  $p < .05$ . Standardized beta coefficients and standard error (SE) terms are presented in the table with SE reported in parentheses.

On the science FSA, Asian students scored significantly higher compared to White students. This main effect was qualified by a significant interaction indicating that this score differential on the science FSA between Asian and White students was **larger** in classrooms where teachers reported greater use of **interdisciplinary teaching practices** (interactional component) ( $\beta = 55.60, p < 0.05$ ). On the other hand, Black students scored significantly lower on the science FSA compared to White students. This main effect was qualified by a significant interaction, which indicated that this score differential on the science FSA between Black and White students was **smaller** in classrooms where teachers reported **higher levels of resourcefulness and coping** (influential factor) ( $\beta = 10.45, p < 0.05$ ).

SES: Students receiving FRL scored significantly lower on Achieve3000 Lexile scores ( $\beta = -31.03, p < 0.05$ ) and the science FSA ( $\beta = -6.58, p < 0.01$ ). Students receiving FRL scored significantly lower on the ELA FSA as well, and this main effect was qualified by a significant interaction, which indicated that this score differential on ELA FSA between students receiving FRL and those who do not was **larger** when teachers **completed more additional (non-Code.org) CS activities** (structural component) ( $\beta = -6.41, p < 0.01$ ) and reported **higher levels of resourcefulness and coping** (influential factor) ( $\beta = -5.23, p < 0.01$ ). In addition, students receiving FRL scored significantly lower on the mathematics FSA. This main effect was qualified by a significant interaction, which indicated that this score differential on mathematics FSA between students receiving FL and those who do not was **smaller** when teachers reported **higher levels of facilitation of group work** (interactional component) ( $\beta = 4.84, p < 0.05$ ).

ELL Proficiency: ELLs ( $\beta = -8.78, p < 0.01$ ) scored significantly lower on the science FSA than native English speakers. ELLs scored significantly lower on the ELA FSA than native English speakers. This main effect was qualified by a significant interaction, which indicated that this score differential on ELA FSA between ELLs and native English speakers was **larger** when teachers reported **higher levels of facilitation of group work** (interactional component) ( $\beta = -9.75, p < 0.01$ ). However, this score differential on ELA FSA was **smaller** when teachers reported **higher levels of resourcefulness and coping** (influential factor) ( $\beta = 5.92, p < 0.05$ ). Furthermore, ELLs scored significantly lower on the mathematics FSA. This main effect was qualified by a significant interaction, which indicated that this score differential on mathematics FSA between ELLs and native English speakers was **smaller** when teachers **reported completing a higher percentage of non-grade-level Code.org CS lessons** ( $\beta = 30.13, p < 0.01$ ) (structural component), **facilitating higher levels of student intellectual risk taking** (interactional component) ( $\beta = 9.98, p < 0.01$ ), **higher levels of resourcefulness and coping** (influential factor) ( $\beta = 6.60, p < 0.05$ ), **and more years of teaching CS** (influential factor) ( $\beta = 1.77, p < 0.05$ ). On the other hand, this score differential on mathematics FSA was **larger** when teachers reported higher levels of **facilitation of group work** (interactional component) ( $\beta = -12.48, p < 0.01$ ).

### 5.2 Attitudinal Outcomes

Gender: No significant differences were observed for students' attitudinal outcomes.

Race: Asian students ( $\beta = 0.28, p < 0.05$ ) reported significantly higher CS identity attitudes than White students.

SES: Students receiving FRL reported more positive CS affinity attitudes. This main effect was qualified by significant interactions, which indicated that this difference in CS affinity attitudes between students receiving FRL and those that do not was **larger** in classrooms where teachers **taught the Time4CS Modules** (whole intervention) ( $\beta = 0.47, p < 0.05$ ), and when teachers reported **greater use of interdisciplinary teaching practices** (interactional components) ( $\beta = 0.46, p < 0.01$ ) and **more years of teaching CS** (influential factor) ( $\beta = 0.09, p < 0.01$ ). In addition, students receiving FRL reported more positive CS identity attitudes ( $\beta = 0.16, p < 0.05$ ).

Table 6. Results from the two-level, random-intercept and coefficient-as-outcome model examining students' scores on the science FSA

Level 2 Variables	Level 1 Variables				
	Asian Students	African American Students	Female Students	Students receiving FRL	ELLs
<i>Non-grade level assigned lessons</i>	71.00 (68.79)	26.26 (31.39)	-3.11 (17.56)	13.86 (23.5)	24.55 (41.29)
<i>Assigned grade-level computer science lessons</i>	-170.22 (147.96)	-22.82 (45.16)	1.66 (30.15)	-56.46 (42.28)	-35.28 (95.93)
<i>Additional computer science</i>	40.93 (32.41)	13.52 (8.87)	-10.36 (6.17)	-10.11 (8.56)	-2.1 (13.49)
<i>Interdisciplinary teaching practices</i>	55.60 (25.19)*	-5.86 (6.91)	-5.06 (4.18)	7.02 (4.94)	-2.62 (7.16)
<i>Group work</i>	-24.68 (29.07)	-6.16 (6.57)	-3.96 (5.59)	-4.23 (6.67)	-2.38 (8.99)
<i>Cognitively demanding work</i>	19.88 (24.42)	1.38 (6.61)	0.29 (4.91)	2.13 (6.10)	-4.14 (8.34)
<i>Student intellectual risk-taking</i>	12.49 (14.72)	4.69 (7.46)	-1.78 (4.53)	-5.63 (5.88)	3.75 (8.21)
<i>Treatment condition</i>	24.07 (18.85)	7.57 (8.5)	2.87 (5.49)	7.62 (7.10)	-0.38 (12.9)
<i>Innovativeness</i>	-16.00 (13.37)	-0.57 (5.13)	0.41 (3.69)	2.00 (4.65)	-8.14 (6.93)
<i>Resourcefulness and coping</i>	21.37 (12.96)	10.45 (4.87)*	0.73 (4.18)	0.75 (5.04)	12.87 (7.83)
<i>Years of teaching CS</i>	-1.03 (3.50)	-2.46 (1.89)	0.53 (0.93)	1.15 (1.28)	3.56 (2.45)

Note. \*  $p < .05$ . Standardized beta coefficients and standard error (SE) terms are presented in the table with SE reported in parentheses. We controlled for students' ethnicity (i.e., Hispanic and non-Hispanic identity) in all models. No significant group differences emerged for ethnicity; therefore, it was not included in any interaction terms at Level-2. ELL = English-language learner.

Table 7. Results from the two-level, random-intercept and coefficient-as-outcome model examining students' CS affinity attitudes

Level-2 Variables	Level-1 Variables				
	Asian Students	African American Students	Female Students	Students receiving FRL	ELLs
<i>Non-grade level assigned lessons</i>	–	–	–	-0.25 (0.67)	–
<i>Assigned grade-level computer science lessons</i>	–	–	–	-1.55 (1.53)	–
<i>Additional computer science</i>	–	–	–	0.05 (0.21)	–
<i>Interdisciplinary teaching practices</i>	–	–	–	0.46 (0.13)*	–
<i>Group work</i>	–	–	–	0.10 (0.15)	–
<i>Cognitively demanding work</i>	–	–	–	-0.13 (0.17)	–
<i>Student intellectual risk-taking</i>	–	–	–	-0.22 (0.15)	–
<i>Treatment condition</i>	–	–	–	0.47 (0.20)*	–
<i>Innovativeness</i>	–	–	–	-0.09 (0.15)	–
<i>Resourcefulness and coping</i>	–	–	–	0.21 (0.12)	–
<i>Years of teaching CS</i>	–	–	–	0.09 (0.03)*	–

*Note.* \*  $p < .05$ . Standardized beta coefficients and standard error (SE) terms are presented in the table with SE reported in parentheses. Group differences were only observed for students' socio-economic status. We controlled for students' ethnicity (i.e., Hispanic and non-Hispanic identity) in all models. No significant group differences emerged for ethnicity; therefore, it was not included in any interaction terms at Level-2. ELL = English-language learner.

## 6. Discussion

### 6.1 Academic Achievement Gaps Persist

Considerable efforts have been made to address inequities in achievement outcomes for diverse populations of students; however, gaps persist. Unsurprisingly, and consistent with previous research, findings from the current study indicate that Black students, students from economically disadvantaged backgrounds (i.e., students receiving FRL), students acquiring English as a second language, and female students scored lower on some achievement outcomes compared to their reference groups (Hadden et al., 2020; Xu, 2019). This study addresses calls from prior research (e.g., Lubienski, 2008) by moving beyond identifying group differences in outcomes to offering potential reasons *why* such gaps persist. Employing CBR in this study enabled us to identify the ways different structural and interactional *components* of the Time4CS Modules influenced the degree of group difference (i.e., amplifying versus reducing it).

### 6.2 Structural Components of the Time4CS Modules Contributed to Larger or Smaller Group Differences in Outcomes

This study found that CS-related components (e.g., implementation of grade-level assigned Code.org lessons; implementation of non-grade level assigned Code.org lessons; implementation of additional non-Code.org lessons) influenced whether the group difference was amplified or reduced. Our earlier analysis (Century, Ferris, & Zuo, 2020) showed no associations between the percentage of grade-level assigned Code.org lessons and FSA outcomes. However, by disentangling the effect of student group, we observed group differences between Black and White students for mathematics FSA outcomes, and this difference was amplified in classrooms where teachers reported teaching more grade-level assigned CS lessons. In BCPS, similar to other large, urban public-school districts, it is possible that disproportionate numbers of Black students reside in low-income areas and attend under-resourced schools with fewer educational supports. When teachers dedicated more time within the literacy block to grade-level assigned CS lessons, this in turn, may have reduced the time Black students had available for instruction in FSA tested areas (e.g., mathematics). This finding suggests Black students face compounded inequities; attending under-resourced schools limits the instructional support present at the outset so when opportunities to engage in CS are presented, outcomes in these other subject areas may suffer.

We also observed through use of CBR that teaching “additional CS activities” was associated with lower ELA FSA scores for students receiving FRL. Because this group by definition come from families with lower incomes and/or parental educational attainment, they also reside in, and attend, more under-resourced neighborhoods and schools. It is possible that the presence of additional CS activities took time away from the particular kinds of ELA instructional supports these learners needed to score highly on the state-administered examination.

Moreover, our results about non-grade level assigned Code.org CS lessons underscore additional equity-related complexities. Our earlier analysis (Century, Ferris, & Zuo, 2020) illustrated that non-grade level assigned Code.org CS lessons were positively associated with students’ ELA and mathematics FSA scores. However, when examining specific student groups, the presence of more non-grade level assigned Code.org CS lessons reduced the group difference in mathematics FSA between ELLs and their native English speakers. This group difference may have been reduced for several reasons. For instance, teachers may have selected non-grade-level assigned Code.org CS lessons that complemented and enhanced the curricula they were already teaching, in turn allowing them to provide more appropriate scaffolding for language learners. Or, they may have selected non-grade-level Code.org CS lessons to support their own, and their students’, varied experiences, and interests. Emphasizing students’ diverse experiences contributes to the creation of an inclusive classroom environment where teaching reflects the diversity of needs of all students and teachers consider how to include *rather than* exclude learners who experience difficulty (i.e., inclusive practice; Abels, 2019), which is central to cultivating positive learning experiences with ELLs (de Jon et al., 2013). Teachers may also have felt that their students were capable of completing these optional lessons, which at times covered more-advanced CS content, because they felt that their students had the ability to understand more-challenging CS concepts. Thus, more non-grade level Code.org CS lessons may have been conducive to their learning, engagement, and growth as students, overall, as well as reducing potential outcome differences based on language proficiency. Conversely, teachers may have carried out more of these lessons covering less-advanced content (i.e., below grade level: fourth grade students completing non-grade level assigned lessons intended for third graders) because they believed they were more-closely aligned with their students’ abilities. Additional studies focusing

specifically on academic achievement and STEM and CS attitudes are needed with ELLs in the future to better understand how to create equitable learning opportunities for this growing number of students and foster teachers' inclusive practice in their classrooms.

Significant attitudinal effects only emerged in the presence of the structural component related to whether teachers taught the Time4CS Module. Specifically, FRL students demonstrated higher CS affinity attitudes, and this difference was larger in classes where the Time4CS Module was taught. This finding suggests that although we found no differences between treatment and comparison groups as a whole (Century, Ferris, & Zuo, 2020), it appears that for students receiving FRL, engaging with the Time4CS Module (e.g., exposure to CS topics presented in an interesting manner) was beneficial for their CS affinity.

### *6.3 Interactional Components of the Time4CS Modules Contributed to Larger and Smaller Group Differences in Outcomes*

We also observed interaction effects with some of the interactional Time4CS Module components. For example, for mathematics FSA outcomes, teachers' facilitation of student intellectual risk-taking contributed to reducing group differences in outcomes between ELL students and native-English speakers. These findings suggest that ELLs may benefit from opportunities to take risks during instruction, and that those opportunities may positively contribute to better academic achievement. Accordingly, teachers who create intellectually safe and supportive learning environments when teaching CS may positively shape students' overall outlook on learning and motivation to persevere when faced with uncomfortable learning challenges (Rabe, 2018). As such, risk-taking may be especially important for ELLs who need extra support to combat stereotypes about their abilities and what they are truly capable of accomplishing in the classroom (Diamond et al., 2016).

Findings related to teachers' facilitation of group work were mixed. For example, when teachers reported more facilitation of group work, the group difference in mathematics FSA scores between low and high SES students was reduced. This finding supports prior research suggesting that group work promotes academic achievement for disadvantaged students (Bailey & Bradbury-Bailey, 2006). Through group work, these students may receive extra peer-to-peer support when working with their classmates as well as increased opportunities to take ownership of work aligning to their skillsets (Slavin, 2014). However, when teachers reported more facilitation of group work, the difference widened between ELLs and Native-English speakers. This effect also emerged when examining ELA FSA scores. These findings may have resulted because group work requires greater communication and collaboration, which are areas where ELL students may be at a disadvantage due to their more limited English-language proficiency.

In addition, findings on interdisciplinary teaching practices further disentangled effects observed for all students in our previous paper (Century, Ferris, & Zuo, 2020). Our earlier analysis indicated that the presence of interdisciplinary teaching practices was positively associated with ELA FSA outcomes for the whole student population (Century, Ferris, & Zuo, 2020), whereas the current analysis revealed that the presence of these practices contributed to two instances of amplified group differences. Specifically, on the science FSA, the group difference between Asian and White students was larger in classrooms where teachers' reported using more interdisciplinary teaching practices. Some research (Farris, 2015) has suggested a link between interdisciplinary teaching practices, multiculturalism, and culturally-responsive instructional strategies, which may be beneficial for Asian students' academic achievement, in particular (Change & Le, 2010). Additionally, the group difference in CS attitudes between FRL students and their counterparts was larger when more interdisciplinary teaching practices were present. The reasons behind these findings are unclear, but we recommend that future researchers use CBR to investigate these effects more specifically to gain a better understanding of the role of interdisciplinary instructional practices.

#### *6.4 Influential Factors Also Contribute to Larger or Smaller Group Differences in Outcomes*

Two teacher characteristics contributed to changes in group differences in outcomes. First, the group difference in science FSA scores was reduced between Black and White students in classrooms where teachers reported having more resourcefulness and coping skills. Similarly, when teachers reported more resourcefulness and coping skills, the group difference between ELL and native-English speakers on mathematics FSA was reduced. On the other hand, the group difference on the ELA FSA between FRL students and their counterparts was larger in the presence of more teacher resourcefulness and coping. One can speculate that teachers who perceive themselves as being resourceful and having coping abilities might be more adept at inclusive practice (Bentley-Williams, Grima-Farrell, & Laws, 2017), in turn providing Black and ELL students with the opportunities and resources they need to increase their academic achievement. Seeing the findings for FRL students; however, suggests that these benefits are not universal, for reasons this study did not reveal and that should be the focus of future investigations.

More commonly measured teacher characteristics – years of teaching CS – were also associated with reductions in group differences. We observed smaller group differences between ELL students and their native-English speaking counterparts for mathematics FSA and ELA FSA scores. While neither of these characteristics was revealed to have direct effects on student outcomes (Century, Ferris, & Zuo, 2020), one can speculate that for ELL students in particular, having teachers with more experience with CS content is beneficial.

### **7. Conclusion**

The current study explored differences in students' attitudinal and academic achievement outcomes based on teachers' implementation of the Time4CS Modules in their classrooms. We employed CBR to disentangle associations between *specific components* of the Time4CS Modules and students' attitudes and academic achievement. Our results assess beyond whether the Time4CS Modules worked and indicate that academic achievement and CS attitudinal gaps persist. Utilizing CBR allowed us to identify potential reasons why some *components* of the Time4CS Modules (i.e., intervention) were beneficial for certain groups of students, but not others. Use of CBR also underscored areas of continued improvement for researchers and educators striving to provide equitable CS learning experiences for young learners. By pinpointing how students' socio-demographic characteristics interact with aspects of their learning context (e.g., teachers' instructional practices), researchers and practitioners can better understand the contexts and conditions needed to promote desired student outcomes related to CS. Therefore, CBR is a valuable approach for moving beyond "gap gazing" to gain insights about what parts of an innovation worked, for which students, and under what conditions. Despite these promising findings and their implications for researchers and educational professionals dedicated to creating more-equitable CS learning opportunities, this project represents merely a starting point. Future research must address the current study's limitations to generate insights that can inform the continued development of CS educational experiences for today's diverse learners.

#### *7.1 Limitations and Future Directions*

Our findings must be considered in light of several limitations. First, the team was unable to control for the BCPS academic calendar (e.g., standardized testing schedule), which likely impacted implementation of the Time4CS Modules given that some teachers taught the modules before FSAs were administered, whereas others used the module after testing took place. Greater attention must be placed on when teachers are implementing the intervention in future studies so accurate effects of the intervention on outcomes can be generated. Second, the teacher and student questionnaire data is subject to self-report bias and socially-desirable reporting, which may contribute to inflated (e.g., more positive) response patterns in the data (Van de Mortel, 2008). Future research should adopt multi-method data collection strategies (Bulsara, 2015), including interviews or focus groups with teachers and students. Classroom observations could also be used to capture real-time accounts of teachers' use of different components to better understand the reasons particular components may be associated with larger or smaller group differences (Archer et al., 2016). Third, the Time4CS Modules used were in the early stage of development during the study, and they may have required revisions to improve usability and to ease teachers' learning curve when implementing them.

We also want to acknowledge two considerations for future examinations focused on equity. First, we did not use methods that accounted for intersectionality. Although our results underscore the importance of CBR when disentangling the effect of structural and relational components of a curriculum designed to promote more equitable student outcomes, follow-up investigations are also needed to better understand how different student characteristics intersect (e.g., race, gender). An intersectional conceptual and analytical approach (Else-Quest & Hyde, 2016a; Else-Quest & Hyde, 2016b) was not the primary focus of this investigation. We recognize that the use of additive analytic approaches (Cole, 2009) limits the conclusions that can be drawn regarding students' dual and overlapping identities (e.g., being Asian *and* female; being Black *and* male; being multiethnic; being ELL and a FRL recipient). Nonetheless, our findings emphasize the need to look more deeply at group differences and recognize that instructional practices may be beneficial to some groups, but detrimental to others. As a result, we encourage future researchers to employ intersectional conceptual and analytical approaches to simultaneously examine student characteristics, attitudes, and academic outcomes as such investigations have the potential to better inform the creation of strategies that can promote educational equity.

Second, we want to acknowledge the implicit *and* explicit biases associated with our academic outcome measures, standardized tests (Knoester & Au, 2017). For instance, some studies indicate that standardized tests are inequitable by design (Au, 2010; Morgan, 2016; Well et al., 2019). Thus, focusing on standardized testing outcomes when striving for educational equity across diverse student populations overlooks how the test design itself contributes to group differences. Students' standardized test scores may be misleading and not accurately depict student learning (Morgan, 2016). Future research needs to acknowledge these shortcomings when referencing empirical literature, use more equitable assessments, and investigate a wide variety of educational outcomes, including students' attitudes toward different subject areas, motivations in the classroom, and interests.

### *7.2 The Call for Research that Will Contribute to More Accurate and Useful Findings about Equitable CS Learning Opportunities*

In light of these limitations, the findings of this study communicate a powerful message: CBR enables researchers to examine the impact of specific *aspects* of an intervention for different groups of learners. When studies of CS interventions show positive or negative main effects, it is essential to look deeper through such approaches. Examining the *components* of educational interventions and their effect on *particular* student groups will enable us to learn more about how to develop more equitable CS learning opportunities and teachers' inclusive practice. Although CS interventions vary widely, some of them share common components with others. Using a CBR approach can enable the field to accumulate knowledge about these shared components. This will pave the way for researchers and practitioners to develop robust knowledge about the best structures and interactions to use in CS instruction to meet the learning needs of all students.

### **Contributors**

Drs. Ferris, Century, and Zuo are researchers at Outlier Research & Evaluation, UChicago STEM Education at the University of Chicago. Outlier Research & Evaluation is a group dedicated to change-focused work through research, evaluation, and strategic planning. We dare to make education equitable.

### **Acknowledgements**

The authors would like to acknowledge the contributions of their practitioner colleagues at Broward County Public Schools, specifically Dr. Lisa Milenkovic, Debra Thomas, and Annmargareth Marousky. Their support allowed us to bring Time4CS to BCPS' elementary teachers and students.



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## Appendix A

### Time4CS Teacher Implementation Questionnaire Items (Module 2, Grade 4)

#### School Information

1. What is the name of your school?
2. This year, what grade STEM+C Integrated Literacy Block materials did you use?

#### Module Completion

As part of the STEM+C Integrated Literacy Block project, you were provided with activities and lessons to use during a daily 180-minute integrated literacy block. There were two components of this 180-minute block:

- 1) The 90-minute **Reading Block** using your normal ELA curriculum and/or suggested texts and center activities provided as part of the STEM+C materials; and
- 2) The 90-minute interdisciplinary **STEM+C Module**, which includes the 5E lessons that integrate science, social studies, and computer science.

Throughout the questionnaire, we will use "**Integrated Literacy Block**" to refer to the entire 180-minutes. We will use "**Reading Block**" and "**STEM+C Module**" to refer to the two **90-minute** components of the 180-minute integrated Literacy Block.

1. First, we'd like to ask you to think back to **last semester** (Fall 2016 - Winter 2016). Did you teach the STEM+C Integrated Literacy Block **Module 1**?
  - 1) Yes
  - 2) No
2. Now we'd like you to think about the most **recent semester** (Winter 2016 - Spring 2017). Did you teach the STEM+C Integrated Literacy Block **Module 2**?
  - 1) Yes
  - 2) No

#### Time

3. Now we're interested in learning more about how you used the Integrated Literacy Block materials.

Of the **total 180 minutes** designated each day for the Integrated Literacy Block, how many minutes a day, on average, did your students spend doing the following? **The times you select across all questions must add up to 180 minutes.**

- \_\_\_\_\_ Doing STEM+C Module 5E activities  
 \_\_\_\_\_ Reading ELA **texts** recommended in the STEM+C Module materials  
 \_\_\_\_\_ Doing ELA **center activities** recommended in the STEM+C Module  
 \_\_\_\_\_ Doing other activities

4. You said students engaged in **other** activities during the 180-minute Integrated Literacy Block. What types of other activities did they do? Please select all that apply.

- 1) **Literacy** activities not outlined in the STEM+C Module materials (including Achieve3000 articles)
- 2) **Science** activities not outlined in the STEM+C Module materials
- 3) **Social studies** activities not outlined in the STEM+C Module materials
- 4) **Computer science** activities not outlined in the STEM+C Module materials
- 5) Other (please explain):

Now we would like you to **just think about the STEM+C Module** component of the 180-minute Integrated Literacy Block. These are the 5E lesson activities on science, social studies, and computer science.

5. The STEM+C Module lessons include “**core**” learning resources that you were expected to use as well as “non-core” or **supplemental** elements that were optional. Please consider the entire STEM+C Module (**both the core AND supplemental activities of the 5E lessons**) when answering the following questions, unless otherwise stated.

Approximately how many weeks did you spend doing the **entire STEM+C Module**?

▼ 1 week (1) ... 12 weeks

Each STEM+C Module lesson is intended to take 90 minutes, but we know there will be variations in how much time each teacher actually spends.

6. Do you think 90 minutes was sufficient to complete the STEM+C Module 5E lesson activities as outlined in the Module?
  - 1) Yes
  - 2) No
7. About how much more time (than the 90 minutes) each day do you feel you needed to complete all of the STEM+C Module 5E lesson activities?
  - 1) Less than 15 minutes
  - 2) 15-30 minutes
  - 3) 31-45 minutes
  - 4) 46-60 minutes
  - 5) More than 60 minutes

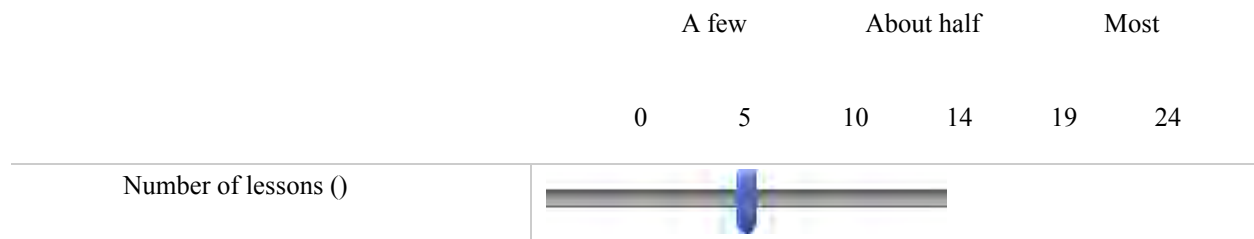
We also know that some teachers worked with media specialists and other “specials” teachers (e.g., art, music) to implement the STEM+C Module 5E lesson activities. For example, some teachers had their media specialist do the online computer science activities that were part of the STEM+C Module.

8. Did you have a media specialist or other specials teacher do any part of the STEM+C Module 5E lessons with your students?
  - 1) Yes
  - 2) No

You said you had a media specialist or other specials teacher do parts of the STEM+C Module lessons with your students. Please take those activities into account and answer as best as you can for yourself and on behalf of the specials teacher(s) when answering the following questions.

#### Lesson Omission

9. Did you omit any of the STEM+C Module lessons entirely? By “*entirely*,” we mean you skipped the entire lesson and did not do any parts of it.
  - 1) Yes
  - 2) No
10. About how many STEM+C Module lessons did you omit entirely for any reason?

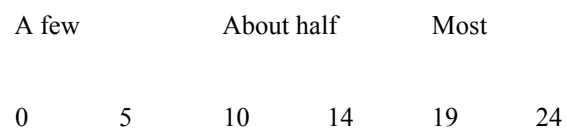


11. In all the lessons you taught, did you ever omit one or more of the 5E components (e.g., "Engage," "Explain") for any reason? Do not count 5E components that a specials teacher taught on your behalf as an omission.
- 1) Yes
  - 2) No
12. About how often did you omit each of the following 5E components for any reason? (Scale: Never, Rarely, Sometimes, Often, Always)
- 1) Engage
  - 2) Explore
  - 3) Explain
  - 4) Elaborate/Extend
  - 5) Evaluate
13. You said you omitted some portion of the STEM+C Module. From the list below, please select the top three reasons you did so.
- 1) The activities were not identified as core parts of the lesson.
  - 2) I didn't have access to the necessary technology (e.g., computers, Internet) to complete the activities.
  - 3) I had access to the necessary technology (e.g., computers, Internet) but it wasn't working.
  - 4) The lesson parts were too long.
  - 5) The lesson parts required too much preparation.
  - 6) The lesson parts weren't written clearly enough to teach them.
  - 7) I didn't understand the content of the lesson parts well enough to teach them.
  - 8) The lesson parts weren't relevant to my students.
  - 9) The lesson parts weren't as important to teach compared to others I needed to teach.
  - 10) Other: (please explain) \_\_\_\_\_

### Lesson Modification

Sometimes teachers modify lessons by departing from what is written in the curriculum materials. This includes using different materials to complete the lesson and/or teaching the lessons in a different way than written (e.g., doing the lesson as a whole group activity instead of partner work). Modification does not include omitting or skipping portions of the STEM+C Module.

14. Of the STEM+C Module lessons that you taught; did you modify them in any way?
- 1) Yes
  - 2) No
15. About how many of the STEM+C Module lessons did you modify?





### Lesson Supplementation

Sometimes teachers **add to** curricular materials like the STEM+C Module with their own activities or lessons.

16. Did you add any **new activities** to the STEM+C Module? *This includes anything you added that wasn't part of the STEM+C Module core and non-core "supplemental" activities.*

- 1) Yes
- 2) No

17. In about how many STEM+C Module lessons did you add activities?

A few                      About half                      Most

0            5            10            14            19            24



### Culminating Project

18. Did your students complete a culminating project?

- 1) Yes
- 2) No

19. Which version of the culminating project did your students complete?

- 1) Unplugged
- 2) Plugged
- 3) Both

20. Did your students complete any additional culminating projects that were not identified as part of the STEM+C Module?

- 1) Yes
- 2) No

21. Please describe the additional culminating project that your students completed.

### Student Journaling

22. As part of the STEM+C Module, did your students use journals?

- 1) Yes
- 2) No

23. From the options below, please select the one that best represents how your students used journals.

- 1) My students used a separate journal for each subject (e.g., one journal for math, one for science, one for literacy, etc.).
- 2) My students used one journal for all subjects.



**ACHIEVE3000**

As part of the STEM+C Integrated Literacy Block project, you were asked to administer Achieve3000 diagnostic assessments to students. We'd like to know about your students' use of other (non-assessment) Achieve3000 materials and activities (e.g., nonfiction texts from Achieve3000).

24. On average, since the beginning of the school year, how many times **per week** did you have your students use Achieve 3000 (not-assessment) materials and activities?

0      1      2      3      4      5



25. When students used Achieve3000 for non-assessment materials and activities, approximately how much time did they spend each time they used it? Please include time spent at school and when assigned for homework.

0      15      30      45      60      75      90

**Computer Science Materials**

Now think back to both last semester (Module 1) and the most recent semester (Module 2).

26. Which of the following **Code.org**, **Course 3** and/or **Scratch** activities did you use with your students? Please select all that apply.

*Please include any lessons taught by a media specialist or other specials teacher on your behalf.*

	Module 1	Module 2
Course 3, Lesson 1: Computational Thinking (unplugged)		
Course 3, Lesson 2: Maze (plugged)		
Course 3, Lesson 3: Artist (plugged)		
Course 3, Lesson 4: Functional Suncatchers (unplugged)		
Course 3, Lesson 5: Artist Functions (plugged)		
Course 3, Lesson 6: Bee Functions (plugged)		
Course 3, Lesson 7: Bee Conditionals (plugged)		
Course 3, Lesson 8: Maze Conditionals (plugged)		
Course 3, Lesson 9: Songwriting (unplugged)		
Course 3, Lesson 10: Dice Race (unplugged)		
Course 3, Lesson 11: Artist Nested Loops (plugged)		
Course 3, Lesson 12: Farmer While Loops (plugged)		
Course 3, Lesson 13: Bee Nested Loops (plugged)		
Course 3, Lesson 14: Bee Debugging (plugged)		
Course 3, Lesson 15: Bounce (plugged)		
Course 3, Lesson 16: Play Lab Create a Story (plugged)		
Course 3, Lesson 17: Play Lab Create a Game (plugged)		
Course 3, Lesson 18: Internet (unplugged)		
Course 3, Lesson 19: Crowdsourcing (unplugged)		
Course 3, Lesson 20: Digital Citizenship (unplugged)		
Course 3, Lesson 21: Artist Patterns (plugged)		
Scratch Lesson 2: Moving Sprites		
Scratch Lesson 3: Broadcasting a Message		
I did not use any of these Code.org or Scratch Lessons (25)		

27. When using the Code.org, Course 3 or Scratch lessons, approximately what proportion of time did your students spend in each of the following configurations? All percentages must add up to 100%.

*If your students worked with a media specialist or other specials teacher, please answer on his/her behalf to the best of your ability.*

Whole Class : \_\_\_\_\_ Small Group : \_\_\_\_\_ Partner : \_\_\_\_\_  
Independent : \_\_\_\_\_ Total : \_\_\_\_\_

28. On average, what proportion of your students did all of the Code.org, Course 3 or Scratch activities you used?

*If your students worked with a media specialist or specials teacher, please answer on his/her behalf to the best of your ability.*

- 1) None
- 2) Some
- 3) About half
- 4) Most
- 5) All

Now we would like to know about any computer science activities you used with your students **during the entire 180-minute Integrated Literacy Block** that were **not identified** in the STEM+C Module.

29. Which additional computer science materials, if any, did you use? Please select all that apply.

*If your students worked with a media specialist or specials teacher, please answer on his/her behalf to the best of your ability.*

- 1) Code.org, Course 1
- 2) Code.org, Course 2
- 3) Code.org, Course 4
- 4) Accelerated Course
- 5) Code Studio Hour of Code activities
- 6) Barefoot Computing
- 7) Khan Academy computer science courses
- 8) Videos on computational thinking and computer science
- 9) Google's computational thinking resources
- 10) Programs that use Blockly programming language
- 11) Computer science vocabulary
- 12) Scratch
- 13) Other (please explain) \_\_\_\_\_
- 14) I did not use any additional computer science materials

30. About how many Code.org **Course 1** lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 18

31. About how many Code.org Course 2 lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 19

32. About how many Code.org Course 4 lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 22

### Implementation - Instructional Practices (Treatment)

For the following questions, please only think of your teaching during the STEM+C Module lessons.

33. When you taught **STEM+C Module lessons**, about how often did you **explicitly**...

- 1) Refer to the overarching Module problem
- 2) Explicitly make lesson content relevant to students (e.g., ask about past experiences, apply content to students' daily lives)
- 3) Cover standards from multiple subject/content areas in the same lesson
- 4) Make connections across content areas during discussions with students

34. When you taught STEM+C Module lessons, approximately what proportion of time did your students spend in each of the following configurations? Proportions must add up to 100%

Whole Class: \_\_\_\_\_ Small Group : \_\_\_\_\_ Partner : \_\_\_\_\_  
Independent : \_\_\_\_\_ Total : \_\_\_\_\_

35. How often did you intentionally and explicitly do the following when students were working in groups or pairs during STEM+C Module lessons? (Scale: Never, Rarely, Sometimes, Often, Always)

- 1) Encourage group members to work to solve problems together
- 2) Encourage group members to share ideas and strategies respectfully
- 3) Encourage group members to listen to what others have to say

36. During STEM+C Module lessons, about how often did you explicitly ask students to do the following?

(Scale: Never, Rarely, Sometimes, Often, Always)

- 1) Organize, process, manipulate, and/or evaluate data
- 2) Explain how they solved a problem
- 3) Consider other students' ideas in comparison to their own
- 4) Demonstrate reasoning (e.g., noting relationships between concepts, making comparisons)
- 5) Consider the merits of using different approaches or problem-solving strategies
- 6) Solve problems in more than one way
- 7) Communicate their thought process to others

37. During **STEM+C Module lessons**, about how often did you **explicitly ask students** to do the following?

(Scale: Never, Rarely, Sometimes, Often, Always)

- 1) Answer a question even if they were unsure
- 2) Try new things even if they might make mistakes
- 3) Share their ideas even if they were different from others

### Support – Implementation

38. Apart from the professional development workshops at the beginning of the school year, did you receive additional support in implementing the STEM+C integrated module? Please select all that apply.

- 1) Broward STEM+C Module contacts (i.e., Annmargareth Marousky, Kelly Thomas, and/or Lisa Milenkovic)
- 2) Your school's STEM+C Module designee
- 3) Media specialist(s)
- 4) Other teachers **at your school** implementing STEM+C Module materials
- 5) Other teachers **from different schools** implementing STEM+C Module materials
- 6) Other teachers **not implementing** STEM+C Module materials
- 7) PLC facilitator(s)
- 8) In-service facilitator(s)
- 9) School leaders (e.g., principal, assistant principal)
- 10) Canvas website
- 11) Online discussion boards or forums
- 12) Other online resources
- 13) Other (please explain) \_\_\_\_\_
- 14) I did not receive any additional support.

### **Attitudes Toward Interdisciplinary Teaching Practices (Treatment)**

Now we're interested in learning about your attitudes toward interdisciplinary teaching practices. The STEM+C Module is an example of interdisciplinary teaching that integrates literacy, science, computer science, and social studies.

39. Prior to teaching the STEM+C Module, how much experience did you have teaching interdisciplinary modules?
- 1) None
  - 2) A little bit
  - 3) A fair amount
  - 4) Quite a bit
  - 5) A lot

For the following questions, please think about interdisciplinary teaching practices **in general**, not just about the STEM+C Module.

40. How much do you agree or disagree with the following statements about your interdisciplinary teaching practices in general? (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) I have sufficient content knowledge across multiple subject areas to be very effective at teaching interdisciplinary modules.
  - 2) I have nearly every skill I need to teach interdisciplinary modules well at the elementary school level.
  - 3) I have nearly every skill I need to teach interdisciplinary modules well at the elementary school level.
  - 4) I am really good at teaching interdisciplinary modules at the elementary school level.
  - 5) I am effective at making interdisciplinary modules relevant to my students.
  - 6) I have sufficient access to the hardware and/or software I need to teach interdisciplinary modules effectively

41. How much do you agree or disagree with the following statements about interdisciplinary teaching practices **in general**? Interdisciplinary teaching... (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) Is more valuable for elementary students' content understanding than teaching each subject separately.
  - 2) Is more engaging for elementary students than teaching each subject separately.
  - 3) Helps students understand the relevance and applicability of the subjects they are learning.
42. How much do you agree or disagree with the following statement? (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- Teaching interdisciplinary modules is a feasible instructional model for elementary classrooms.
43. You said you disagree that teaching with interdisciplinary modules is a feasible instructional model for elementary classrooms. What are the **top three reasons** why you disagreed?
- 1) I don't have enough time.
  - 2) I lack content knowledge across multiple subject areas.
  - 3) I don't have sufficient access to adequate hardware and/or software.
  - 4) I am not confident teaching with technology.
  - 5) I have difficulty managing my students.
  - 6) I don't have adequate materials.
  - 7) Other (please explain): \_\_\_\_\_

#### **Leadership Attitudes Toward CS (Treatment)**

We're interested in your perception of your school leaders' attitudes towards computer science. Your answers are **confidential** and will not be shared with anyone outside of the research team.

44. How much do you agree or disagree with the following statements?  
My school leaders... (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) Think it's important for all students to learn computer science in elementary school.
  - 2) Believe that learning computer science in elementary school will help students get a good job someday.
  - 3) Communicate the value of computer science to students, parents, teachers, and staff.

#### **Teacher Attitudes Survey (All)**

45. How much do you agree or disagree with the following statements about **your teaching in general?**  
 (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) I have nearly every skill I need to teach elementary school well.
  - 2) I am really good at teaching elementary school.
  - 3) I am effective at making schoolwork relevant to my students.
46. How much do you agree or disagree with the following statements about your teaching in general?  
 (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) I experiment with new practices all the time.
  - 2) I am always looking for new ways of doing things in my teaching.
  - 3) I am constantly the first to try new things in my school.
47. How much do you agree or disagree with the following statements about your teaching in general?  
 (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) I am able to manage the pressure and stress at my school well.
  - 2) I see difficult tasks through to the end.
  - 3) I find ways to accomplish my goals.
  - 4) When planning for my work, I prepare for potential challenges.
  - 5) I am able to manage my work even when there are unexpected changes and constraints.

After this last set of questions about computer science, you will be done!

48. On a scale of 0-5, how familiar are you with the term "computer science?"
- |          |   |                   |   |            |
|----------|---|-------------------|---|------------|
| familiar |   | somewhat familiar |   | / familiar |
| 0        | 1 | 3                 | 4 | 5          |

Level of Familiarity ()	
-------------------------	--



49. If someone asked you what computer science is, what would you say?
50. If you had to pick, which one of the following would you choose to explain what computer science is?
- 1) Keyboarding and typing.
  - 2) Designing websites, apps, and video games.
  - 3) Studying how computers work.
  - 4) Fixing computers when they break.
  - 5) Building computers and other hardware.
  - 6) Making spreadsheets and graphs.
  - 7) Using computers to solve complex problems.
  - 8) Doing science on a computer.
  - 9) Analyzing data.
  - 10) Robotics.
  - 11) Other (please explain):

Thanks for telling us what you think computer science is. For this survey, when we say computer science, we mean:

Studying and using computers to help solve problems and to create technology, like video games and apps.

51. How much do you agree or disagree with the following statements about your teaching of computer science? (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) I understand computer science concepts well enough to be a very effective teacher of computer science at the elementary school level.
  - 2) I have nearly every skill I need to teach computer science well at the elementary school level.
  - 3) I am really good at teaching computer science at the elementary school level.
  - 4) I am effective at making computer science relevant to my students.
52. How much do you agree or disagree with the following statements about computer science? (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) I enjoy computer science
  - 2) I am interested in computer science
  - 3) I like doing computer science.
53. What part of computer science do you like the most? Please select only one answer.
- 1) Designing websites, apps, and/or video games.
  - 2) Computer programming and coding
  - 3) Data analysis
  - 4) Robotics
  - 5) Fixing computer hardware problems
  - 6) Fixing computer software problems
  - 7) Studying how computers work
  - 8) Building computers and other hardware
  - 9) Learning new computer programming languages (e.g. Python, C++, Java)
  - 10) Other (please explain):
54. How much do you agree or disagree with the following statements regarding learning computer science in elementary school? (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)
- 1) Is just as important for students to learn as non-core subjects, like art, music, and foreign language.

- 2) Is just as important for students to learn as core-subjects, like math, science, and English/Language Arts.
- 3) Is not developmentally-appropriate for elementary school children.
- 4) Will ensure students are prepared for future academic success.
- 5) Will ensure students get good jobs in the future.

55. How much do you agree or disagree with the following statements regarding learning computer science in elementary school? (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)

Learning computer science in elementary school can help students learn how to...

- 1) Consider alternative approaches to their work.
- 2) Analyze (organize, process, manipulate, evaluate) data.
- 3) Problem solve when something doesn't work the way they want it to work.

56. How much do you agree or disagree with the following statements regarding learning computer science in elementary school? (Scale: Strongly Disagree, Disagree, Somewhat Disagree, Somewhat Agree, Agree, Strongly Agree)

Learning computer science in elementary school can help students learn how to...

- 1) Explain the logic and reasoning supporting their solutions.
- 2) Explain why they agree or disagree with the work of other students.
- 3) Communicate their thought processes to others.

**School Grade (Control)**

57. What grade(s) do you teach? Please select all that apply.
- 1) 3rd grade
  - 2) 4th grade
  - 3) 5th grade

**Essential Program Elements (Control)**

First, we'd like to ask you some questions about Achieve3000.

As part of the STEM+C Integrated Literacy Block project, you were asked to administer Achieve3000 diagnostic assessments to students. We'd like to know about your students' use of other (**non-assessment**) Achieve3000 materials and activities (e.g., nonfiction texts from Achieve3000).

58. About how many times did you have your students use Achieve3000 (non-assessment) materials and activities?

0 5 10 15 20 25 30 35 40 45 50



59. When students used Achieve3000 (non-assessment) materials and activities, approximately how much time did they spend each time they used it? Please include time spent at school and when assigned for homework.

0 15 30 45 60 75 90



Now we are interested in learning about your use of Code.org's computer science activities. This includes the Computer Science Fundamentals Courses 1-4, the Accelerated Course, and the eight Code Studio Hour of Code activities (e.g., Play Lab, Artist, Frozen).

60. Since the start of the school year, which Code.org materials, if any, have you used with your students? Please select all that apply.
- 1) Course 1
  - 2) Course 2
  - 3) Course 3
  - 4) Course 4
  - 5) Accelerated Course
  - 6) Hour of Code activities
  - 7) I did not use any Code.org materials with my students.

A Code.org “lesson” may be either an unplugged lesson, or a “stage” (that is composed of a series of puzzles).

61. About how many Code.org **Course 1** lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 18 (18)

62. About how many Code.org Course 2 lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 19 (19)

63. About how many Code.org Course 3 lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 21 (21)

64. Q112 About how many Code.org Course 4 lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 22 (22)

65. About how many Code.org Accelerated Course lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 20 (20)

66. About how many Code.org Hour of Code lessons (unplugged or stages) did most/all of your students complete?

▼ 1 (1) ... 8 (8)

67. Now we would like a little more specific information on Course 2 lessons only. Which Course 2 lessons did all/most of your students complete?

- 1) Course 2, Lesson 1: Graph Paper Programming (unplugged)
- 2) Course 2, Lesson 2: Real-Life Algorithms (unplugged)
- 3) Course 2, Lesson 3: Maze Sequence (plugged)
- 4) Course 2, Lesson 4: Artist Sequence (plugged)
- 5) Course 2, Lesson 5: Getting Loopy (unplugged)
- 6) Course 2, Lesson 6: Maze Loops (plugged)
- 7) Course 2, Lesson 7: Artist Loops (plugged)
- 8) Course 2, Lesson 8: Bee Loops (plugged)
- 9) Course 2, Lesson 9: Relay Programming (unplugged)
- 10) Course 2, Lesson 12: Conditionals with Cards (unplugged)

- 11) Course 2, Lesson 13: Bee Conditionals (plugged)
  - 12) Course 2, Lesson 14: Binary Bracelets (unplugged)
  - 13) Course 2, Lesson 15: The Big Event (unplugged)
  - 14) Course 2, Lesson 16: Flappy (plugged)
  - 15) Course 2, Lesson 17: Play Lab: Create a Story (plugged)
  - 16) Course 2, Lesson 18: Your Digital Footprint (unplugged)
  - 17) Course 2, Lesson 19: Artist: Nested Loops (plugged)
68. Please select any other (not from Code.org) computer science activities that you used in the last semester.
- 1) Barefoot Computing
  - 2) Khan Academy computer science courses
  - 3) Videos on computational thinking and computer science
  - 4) Google's computational thinking resources
  - 5) Programs that use Blockly programming language
  - 6) Scratch
  - 7) Other (please explain): \_\_\_\_\_
  - 8) I did not use any other computer science-related activities.

### Implementation - Instructional Practices (Control)

69. For the following questions, please **only think** of your teaching from the **past 12 weeks**.
- Over the **past 12 weeks**, about how often did you... (Scale: Never, Rarely, Sometimes, Often, Always)
- 1) Explicitly make lesson content relevant to students (e.g., ask about past experiences, apply content to students' daily lives)
  - 2) Cover standards from multiple subject/content areas in the same lesson
  - 3) Make connections across content areas during discussions with students
70. Over the **past 12 weeks**, approximately what proportion of time did your students spend in each of the following configurations? All proportions must add up to 100%.  
 Whole Class: \_\_\_\_\_ Small Group: \_\_\_\_\_ Partner: \_\_\_\_\_  
 Independent: \_\_\_\_\_ Total: \_\_\_\_\_
71. Over the past 12 weeks, about how often did you intentionally and explicitly do the following when students were working in groups or pairs? (Scale: Never, Rarely, Sometimes, Often, Always)
- 1) Encourage group members to work to solve problems together
  - 2) Encourage group members to share ideas and strategies respectfully
  - 3) Encourage group members to listen to what others have to say
72. Over the **past 12 weeks**, about how often did you **explicitly ask students** to do the following? (Scale: Never, Rarely, Sometimes, Often, Always)
- 1) Organize, process, manipulate, and/or evaluate data
  - 2) Explain how they solved a problem
  - 3) Consider other students' ideas in comparison to their own
  - 4) Demonstrate reasoning (e.g., noting relationships between concepts, making comparisons)
  - 5) Consider the merits of using different approaches or problem-solving strategies
  - 6) Solve problems in more than one way
  - 7) Communicate their thought process to others

73. Over the **past 12 weeks**, about how often did you **explicitly ask students** to do the following? (Scale: Never, Rarely, Sometimes, Often, Always)

- 1) Answer a question even if they were unsure
- 2) Try new things even if they might make mistakes
- 3) Share their ideas even if they were different from others'

### **Attitudes Toward Interdisciplinary Teaching Practices (Control)**

Now we're interested in learning about your attitudes toward interdisciplinary teaching practices. Here is an example of what we mean by "interdisciplinary teaching practices":

You teach a 7-week module on North American Regions by:

- integrating reading, researching, writing, and presenting about a North American region (ELA);
- studying the geography and natural/man-made landmarks of a region (social studies);
- researching and studying the climate, vegetation, and natural resources of the region (science);
- and using models/simulations to create a map of the region (computer science).

74. How much experience do you have teaching interdisciplinary modules like the one described in the example above?

- 1) None
- 2) A little bit
- 3) A fair amount
- 4) Quite a bit
- 5) A lot

75. How much do you agree or disagree with the following statements about interdisciplinary teaching practices?

- 1) I have enough time to teach interdisciplinary modules.
- 2) I have sufficient content knowledge across multiple subject areas to be very effective at teaching interdisciplinary modules.
- 3) I have nearly every skill I need to teach interdisciplinary modules well at the elementary school level.
- 4) I am really good at teaching interdisciplinary modules at the elementary school level.
- 5) I am effective at making interdisciplinary modules relevant to my students.

76. How much do you agree or disagree with the following statements about interdisciplinary teaching practices?

- 1) I have sufficient access to the hardware and/or software I need to teach interdisciplinary modules effectively.

77. How much do you agree or disagree with the following statements about interdisciplinary teaching practices? Interdisciplinary teaching...

- 1) Is more valuable for elementary students' content understanding than teaching each subject separately.
- 2) Is more engaging for elementary students than teaching each subject separately.
- 3) Helps students understand the relevance and applicability of the subjects they are learning.

78. How much do you agree or disagree with the following statement?

- 1) Teaching interdisciplinary modules is a feasible instructional model for elementary classrooms.

79. You said you disagree that teaching with interdisciplinary modules is a feasible instructional model for elementary classrooms. What is the main reason you disagreed? Please select only one.
- 1) I don't have enough time.
  - 2) I lack content knowledge across multiple subject areas.
  - 3) I don't have sufficient access to adequate hardware and/or software.
  - 4) I am not confident teaching with technology.
  - 5) I have difficulty managing my students.
  - 6) I don't have adequate materials.
  - 7) Other (please explain):

**Leadership Attitudes Toward CS (Control)**

80. We're interested in your perception of your school leaders' attitudes toward computer science. Remember, **your answers are confidential** and will not be shared with anyone outside of the research team. How much do you agree or disagree with the following statements? My school leaders...
- 1) Think it's important for all students to learn computer science in elementary school.
  - 2) Believe that learning computer science in elementary school will help students get a good job someday.
  - 3) Communicate the value of computer science to students, parents, teachers, and staff.

## Appendix B

### Time4CS Student Questionnaire Items

#### School Background Questions

1. What school do you go to? [drop-down menu listing schools]
2. What is your teacher's last name? [drop-down menu listing teachers' names and a free-response option if students did not see their teacher's name]

#### General School Affinity

How much do you agree or disagree with the following statements about school?



1. I like schoolwork.
2. My schoolwork is interesting.
3. I enjoy doing my schoolwork.

#### General School Self-Efficacy

How much do you agree or disagree with the following statements about school?



1. I have the ability to do my schoolwork.
2. I am better at schoolwork than most of the other kids at my school.
3. I am very good at doing my schoolwork.
4. I can figure out how to solve the most difficult problems in school if I try.

### Knowledge of Computer Science

1. Have you ever heard of “computer science” before? [Yes/No]
2. If someone asked you what computer science is, what would you say? [Free-response]
3. Which one of the following would you choose to explain what computer science is? [Single response choice]
  - 3.1. Computer science is...
    - 3.1.1. Being good at computers.
    - 3.1.2. Doing science on a computer.
    - 3.1.3. Engineering.
    - 3.1.4. What is inside a computer?
    - 3.1.5. Learning how computers work.
    - 3.1.6. Building computers.
    - 3.1.7. Making websites, apps, and video games.
    - 3.1.8. Studying computers.
    - 3.1.9. Using computers to solve problems.
    - 3.1.10. Computer programing and coding.
    - 3.1.11. Typing.
    - 3.1.12. Robotics.
    - 3.1.13. Other (please explain) [Free-response]
  - 3.2. Prompt: Thank you for telling us what you think computer science is. For this survey, when we say computer science, we mean: Studying and using computers to help solve problems and to create technology, like video games and apps.

### Computer Science Affinity

How much do you agree or disagree with the following statements about school?



1. I like computer science.
2. I think computer science is interesting.
3. I want to learn more computer science.
4. I like doing computer science activities.



**Computer Science Self-Efficacy**

How much do you agree or disagree with the following statements about school?



1. I have the ability to learn computer science.
2. I am better at computer science than most of the other kids at my school.
3. I am very good at computer science.
4. I can figure out how to solve hard problems in computer science if I try.

**Computer Science Identity**

How much do you agree or disagree with the following statements about school?



1. Kids like me do computer science.
2. I like computer science more than other kids at my school.
3. I think I could become a computer scientist one day.
4. I do computer science in my free time.

**Computer Science Utility**

How much do you agree or disagree with the following statements about school?



1. It's important for me to learn computer science.
2. Computer science is useful for me to learn.
3. Learning computer science can help me get a job someday.

**Prior Computer Science Experience**

1. Have you ever done robotics before? [Yes/No]
2. Have you done activities on Code.org before [Yes/No]

### **Demographics**

Now we'd like to ask you some questions about you?

1. Are you a boy or a girl? [Boy/Girl]
2. Are you Hispanic/Latino? [Yes/No]
3. What is your ethnicity? [Select all that apply]
  - 3.1. African American
  - 3.2. Asiana
  - 3.3. White
  - 3.4. Other
4. How old are you? [Drop-down menu listing age choices]

*Note.*

1. Researchers should contact the authors prior to use of items from this questionnaire or the questionnaire in its entirety. All requests can be sent to Jeanne Century at outlieruchicago@gmail.com
2. Researchers planning to utilize items from this questionnaire, or this questionnaire in its entirety, should use the following citation in their reference list.

Century, J., Ferris, K. A., & Zuo, H. (2020). Finding time for computer science in the elementary school day: a quasi-experimental study of a transdisciplinary problem-based learning approach. *International Journal of STEM Education*, 7, 1-16.