



**Equitable STEM+CS learning experiences for girls of color:
Nurturing an independent learning approach via a learning
ecosystem**

Journal:	<i>Journal for Multicultural Education</i>
Manuscript ID	JME-01-2023-0004.R1
Manuscript Type:	Research Paper
Keywords:	Equity, Black girls and girls of color, middle school, STEM+CS, learning ecosystem, independent learning

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Equitable STEM+CS learning experiences for girls of color: Nurturing an independent learning approach via a learning ecosystem

Purpose: There is a critical need to understand how to attract Black girls and other girls of color to the STEM+CS field. This study looked at the design and implementation of a CS learning ecosystem that supports girls of color in acquiring critical CS skills starting in middle school.

Design, methodology, and approach: This mixed-method case study included 53 girls, between the ages of 11 and 13, in four United States middle schools. Study methods included the analysis of a pre-program student survey, longitudinal interviews and focus groups, weekly observations, and computing artifacts.

Findings: Program participants were interested in CS, were confident in their ability to learn CS, had prior coding and CS experience, and had parents and teachers who encouraged them to learn CS. But some students showed dependent learning behaviors while engaging in CS activities. These included relying on instructors and being reticent to make mistakes—behaviors that limit learning. The CS learning ecosystem supported students as they shifted from applying dependent learning approaches to applying independent learning approaches. Instructors sustained a growth mindset and supported productive struggle as students learned CS skills.

Originality: A CS learning system supported equitable learning experiences and helped students develop independent learning behaviors that led to deeper engagement in CS.

Keywords: Black girls, computer science, computational thinking, dependent learning, equity, independent learning, learning behaviors, learning ecosystem, middle school girls, STEM, STEM+CS

Introduction

In the United States, Black students and their parents report higher interest and support for computer science (CS) than White and Hispanic students and their families (Google Inc. and Gallup Inc., 2015, 2016). Black students also report seeing more CS role models in the media than White and Hispanic students (Google Inc. and Gallup Inc., 2016). But Black students—Black girls in particular—have the lowest participation rates in CS throughout the education and employment pipeline (National Science Board, 2019). There is a clear disconnect between student and parent interest in CS and the makeup of CS pipelines.

To address this disconnect, we studied the design and implementation of a CS learning ecosystem that supports middle school girls – between the ages of 11 and 13 – and Black girls in particular in acquiring computer science and computational thinking (CS/CT) skills. The ecosystem is a comprehensive yearlong CS learning experience that includes culturally responsive CS curriculum materials with career exploration that highlights the achievements of Black women in technology; Black women computer scientists serving as instructors and mentors who were trained in culturally responsive and equitable instructional practices; weekly CS experiences at school; and two-weeks of summer camp at a local university.

Our guiding research question focuses on the CS learning ecosystem, and asks: How does the experience of a CS learning ecosystem (re)shape students' perceptions of and behaviors towards CS engagement, learning, and persistence? To explore this question, 53 middle school girls from four urban middle schools in the eastern United States were engaged in the study. Pre-program data on student demographics, perceptions of math and reading, and experiences in CS were captured and analyzed quantitatively. During and post-program data from observations, interviews, and

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3 computing artifacts were coded and analyzed qualitatively to understand students' CS
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5 engagement, learning, and persistence.
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10 **Literature Review**

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12 Within the context of adolescent development, students experience dynamic
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14 physical, social, and cognitive changes (National Academies of Science Engineering
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16 and Medicine, 2019), their schooling experience also changes as coursework becomes
17
18 more rigorous, teachers' expectations change, class schedules become more intense, and
19
20 parent role expectations change from more to less involved (Eccles et al., 1993).
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24 Girls of color are met with additional identity formation challenges as well.
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26 Their development includes a "racial puberty" (Kendi, 2019; Umana-Taylor et al.,
27
28 2014) which involves making sense of their racial and gender identities and navigating
29
30 through experiences of stereotype threat (Steele & Aronson, 1995), imposter syndrome
31
32 (Clance, 1985; Clance & Imes, 1978), and explicit and implicit biases (Epstein, Blake,
33
34 & Gonzalez, 2017; Young, Young, & Ford, 2017). Yet there is a further challenge for
35
36 girls of color, they are underrepresented in advanced and gifted classes at school
37
38 (Grissom & Redding, 2016; Shores, Kim, & Still, 2020; Young, Young, & Ford, 2017)
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40 and in STEM+CS focused activities (Hill, Corbett, & St. Rose, 2010; McGee, 2013;
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42 Ong et al., 2011; Young, Young, & Paufler, 2017).
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47 To better support the engagement, learning, and persistence in STEM+CS of
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49 girls of color, culturally responsive learning environments should actively develop
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51 belonging and STEM identities, support girls through the productive struggle of
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53 learning, and to nurture independent learning behaviors and approaches.
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3 *Developing belonging and STEM identities.*
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5 Past research has shown the importance of improving K-12 STEM learning
6 experiences for girls of color by creating culturally responsive learning environments
7 that foster deep engagement, learning, and persistence (Calabrese Barton, 2019; Farinde
8 & Lewis, 2012; Kang *et al.*, 2019; King, 2022; King & Pringle, 2019). Deep
9 engagement, learning, and persistence are rooted in developing a sense of belonging and
10 building STEM identities in formal and informal spaces (Kang *et al.*, 2019; King, 2022;
11 King & Pringle, 2019; Ireland *et al.*, 2018; Ong *et al.*, 2011).
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21 STEM disaffiliation, particularly for girls of color, can be seen as early as
22 elementary school with Black girls not wanting to be identified as a “smart science
23 student” (Carlone, Haun-Frank, & Webb, 2011). However, proactively developing
24 culturally affirming STEM identities serves as a foundation for STEM interest and
25 career exploration (Calabrese Barton *et al.*, 2013; Calabrese Barton *et al.*, 2008; King &
26 Pringle, 2019).
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36 *Supporting productive struggle of learning.*
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38 Providing culturally responsive learning environments involves helping students
39 navigate learning challenges and resolving productive struggle (Nottingham, 2017;
40 Nottingham and Larsson, 2018). Students must go through productive struggle to
41 construct new knowledge, that is, learning (Nottingham, 2017). However, students of
42 color not only have to resolve the cognitive conflict of sense-making (Nottingham,
43 2017; Nottingham and Larsson, 2018), but they must also contend with personal (e.g.,
44 stereotype threat) and systemic (e.g., structural racism) challenges to equitable learning
45 (Aronson *et al.*, 2009; Darling-Hammond, 2001; Delpit, 2006; Hammond, 2015;
46 Pollock, 2017).
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3 Carter Andrews and her colleagues noted (2019) that schools often become
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5 inhospitable environments wherein students of color receive mixed messages and are
6
7 held to unreasonable standards. Consistent with survey data (Epstein *et al.*, 2017) and
8
9 literature on stereotype threat (Aronson *et al.*, 2009), in-depth interviews of 70 high
10
11 school girls showed that Black girls felt it necessary to offset perceptions of being
12
13 rowdy, disruptive, unintelligent, unmotivated, or some combination of these while
14
15 feeling marginalized by their White peers and school personnel (Carter Andrews *et al.*,
16
17 2019). Black girls and other marginalized students show significant social and
18
19 emotional skills, including resilience (Rosen *et al.*, 2010) and grit (Duckworth *et al.*,
20
21 2007), when navigating inequitable spaces. However, culturally responsive learning
22
23 environments create equitable, safe, and supportive spaces for students to practice
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25 productive struggle for deep engagement and learning to take place (Darling-Hammond,
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27 2001; Hammond, 2015; Love, 2019).
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35 *Nurturing independent learning behaviors and approaches.*

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37 Delpit (2006, 2012) and Hammond (2015), grounded by the works of Ladson-
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39 Billings (1995, 2017), have noted that “good teaching” is more than just relaying
40
41 academic content. Culturally responsive learning environments are about explicitly and
42
43 proactively nurturing the learning capacity of students (academic, cognitive, social, and
44
45 emotional capacity) (Delpit, 2006, 2012; Hammond, 2015; Ladson-Billings, 1995).
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47 While younger students are dependent on their teachers to learn developmentally,
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49 adolescence is a time when effective pedagogy includes developing students’
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51 independent learning behaviors. Adolescent students are expected to become less
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53 dependent on the teacher and more independent as they learn to scaffold learning
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55 (Hammond, 2015).
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3 Students of color and other marginalized students are often not taught skills
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5 needed to move from being dependent to independent learners (Allington and McGill-
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7 Franzen, 1989; Darling-Hammond, 2001; Delpit, 2012; Hammond, 2015). Thus, there is
8
9 a critical need to systemically, consistently, and proactively support girls of color to
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11 develop culturally affirming STEM identities, successfully engage in productive
12
13 struggle, and practice independent learning behaviors and approaches to engage, learn,
14
15 and persist in CS (Scott, Sheridan, & Clark, 2015).
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19 ***BRIGHT-CS theoretical framework: A computer science learning ecosystem***
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21 The Building Student Retention through Individuated Guided coHort Training in
22
23 Computer Science (BRIGHT-CS) student program created a CS learning ecosystem for
24
25 middle school girls of color to experience CS in an environment that supports their
26
27 STEM+CS identities, helps them navigate through productive struggle, and helps them
28
29 learn skills necessary to develop learning independence. The theoretical framework of
30
31 BRIGHT-CS aligns with the social and emotional needs of middle school students
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33 (Delpit, 2006; Eccles *et al.*, 1993) and started with a cohort of girls of color from the
34
35 same middle school (grades 6-8 and ages 11-13). Across four key components of the
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37 program, the authors developed CS curricular materials and training materials focused
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39 on culturally responsive instructional practices, pedagogy, and collaborative activities
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41 (Hammond, 2015).
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54 For student outcomes, we defined CS/CT based on the K-12 Computer Science
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56 Standards (2017). Student activities and learning targets were aligned with the concepts
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58 of algorithms and programming across CS practices (Computer Science Teachers
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3 Association, 2017). An overview of the BRIGHT-CS components, CS standards, and
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5 student activities is shown in table I.
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10 Insert table I
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14 **Methods**

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16
17 Following standards suggested by Levitt and her colleagues (2018), this section
18
19 provides an overview of the research design, study participants and recruitment, data
20
21 collection, and analysis.
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23 *Research Design Overview*

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26 The research team chose a mixed-method case study design focused on
27
28 BRIGHT-CS participants across two cohorts (Baxter and Jack, 2008; Noor, 2008; Yin,
29
30 2009). We implemented a mixed-method design to better understand what works in
31
32 terms of instruction and academic supports through the lens of lived student experience
33
34 (Safir & Dugan, 2021). Quantitative data were collected before the start of the program
35
36 via a student survey focused on computer science attitudes and compared with a
37
38 national benchmark of students (Google Inc. and Gallup Inc., 2016).
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43 Qualitative data were collected during and after program implementation via
44
45 weekly observations, interviews, and computing artifacts to explore the context,
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47 phenomena, process, and potential causal explanations or theory (Denzin and Lincoln,
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49 2018; Eisner and Peshkin, 1990; Garson, 2002; Leavy, 2015; Maxwell, 1998; Strauss
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51 and Corbin, 1990).
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Study Participants

Researcher description

The first and second authors are social scientists; they collected and analyzed qualitative and quantitative study data. The third and fourth authors are computer scientists; they reviewed the CS artifacts and validated the CS codes and themes used in the qualitative analysis.

The first author is Asian. The second author is White Latina. The third and fourth authors are Black. All four authors are women of color; thus, they bring first-hand knowledge of the subtle inequities in the learning system. The intersectional lens and identities of the research team supported candor and openness during interviews with students, parents, and other stakeholders. All authors also have experience teaching students at the middle school, high school, or undergraduate levels, bringing an additional educator's perspective to data collection, analysis, and interpretation.

Participant recruitment and selection

The first, second, and fourth authors began the recruitment and selection process for BRIGHT-CS by identifying a school sponsor at each of the four middle schools included in the study. School sponsors recruited students, prioritizing recruiting Black girls, and coordinated with parents to complete the program application and obtain active consent to take part in the study. All students who applied were accepted into the program. Ergo, one White male was admitted due to parent appeal to the principal; data from this participant was not included in the sample or the analysis since the study focuses on girls.

Participant description

The study sample is composed of 53 girls from four urban middle schools in the eastern United States. The students participated in the program during the 2018-2019 and 2019-2020 school years. The 53 girls account for 77% of the 69 students who applied, participated in the program, and gave active parental consent and student assent to take part in the study.

The majority of the study sample were Black girls (66%), with 13% Hispanic, 11% White, and 11% Asian. Forty percent of the students (40%) self-reported that they spoke another language at home. The students spoke fluent English at school but other languages at home including Amharic, Bengali, Farsi, Haitian Creole, Hausa, Somali, Spanish, Tigrinya, Twi, and Urdu.

Data Collection

Quantitative data were collected on all 53 participants during the student application process, pre-program. The student survey captured student demographics, perceptions of math and reading (NCES ELS, 2001), and experiences in CS (Google Inc. and Gallup Inc., 2016).

Qualitative data were collected during and post-program and included interviews with students (N = 53), parents (N = 3), community mentors (N = 7), school staff sponsors (N = 4), school principals (N = 4), program instructors (N = 3), and summer program instructors (N = 3) at multiple time points during and after the program. In addition to the interviews, we collected program documents, computing artifacts, weekly observations of after-school sessions, and daily observations of summer sessions. Observational and interview data were prioritized during and post-program in order to capture behaviors rather than attitudes of student engagement, learning, and persistence in CS.

Analysis

Quantitative data were analyzed using descriptive statistics. Qualitative data were digitized, uploaded to DeDoose (a cloud-based qualitative analysis platform), and analyzed using a "start list" (Garson, 2002; Hill *et al.*, 1997; Leavy, 2015; Saldana, 2009). The start list was based on the main components and theoretical frameworks of the project, with initial codes for student engagement, learning, and persistence behaviors during the in-school, summer, and mentoring experiences; student experiences with peers; and instructor instructional approaches and rapport building with students.

Codes were further refined via triangulation with pre-program survey data; computing artifacts; and instructor, parent, and stakeholder interviews. Codes were finalized into core ideas, themes, and cross-analysis with validation from the program instructors (Garson, 2002; Huberman and Miles, 1994; Leavy, 2015; Strauss and Corbin, 1990), with 100% agreement on data coding across cohorts and respondents.

Findings

By analyzing the pre-program survey data collected from participants, we found that students in the BRIGHT-CS program had higher interest (36%) and higher confidence in their ability to learn CS (68%) than a national sample of 771 girls (16% and 48% respectively) (Google Inc. and Gallup Inc., 2016). Almost 2 in 3 students in the program were told by their teachers and parents that they would be good at computer science (60% for both teachers and parents), compared to about 1 in 4 girls in the national sample. Differences are not surprising given that, unlike the national benchmark sample, students in the program self-selected to participate in a CS program.

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3 Students indicated that they joined the program to learn something new (43%),
4 because they have a passion and love of computer science (30%), and they see CS as a
5 useful skill for future careers including non-STEM careers (26%). However, high
6 interest, confidence, and motivation were not enough for students to fully engage in and
7 learn CS.
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11 We observed two types of learning behavior approaches during the
12 implementation of BRIGHT-CS–dependent learning approaches (exhibited by students
13 dependent on the guidance of the instructor at each step of the aforementioned learning
14 process) and independent learning approaches (exhibited by students moving
15 independently through the learning process, without extensive guidance or support from
16 the instructors) (Delpit, 2012; Hammond, 2015). We identified 10 girls who—despite
17 their self-reported confidence—began the program applying dependent learning
18 approaches and another 10 who began with independent learning approaches. We found
19 that the CS learning ecosystem supported student engagement, learning, and persistence
20 by igniting and strengthening independent learning approaches. To highlight the core
21 themes of how students engaged in and learned CS, we will focus on the experience of
22 Sara (a pseudonym), a 6th grader in the program.
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42 ***The Story of Sara***

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44 Sara’s story demonstrates how a culturally responsive learning environment
45 (i.e., BRIGHT-CS herein), supports deeper engagement, learning, and persistence.
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47 Sara’s growth and adoption of independent learning approaches embodied the growth
48 seen in other students as well. Sara persisted by completing and presenting her
49 computing artifacts. She also continued to participate in the second year of the program
50 and recruited other girls to join her. Sara’s experience (i.e., her adoption of independent
51 learning approaches that include freely engaging in open-ended discussion and
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3 resolving challenges with minimal guidance) underpins the broader themes of becoming
4 more comfortable participating in productive struggle, iterating to maximize outcomes
5 even when it requires learning new skills, and creating a strategic plan to ‘debug’ errors
6 while programming and designing.
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12 Like many students in the BRIGHT-CS program, Sara rated herself as having
13 “high interest” and “high confidence” in learning CS. Sara had prior exposure to CS
14 through elementary school coding experiences (i.e., after-school clubs and summer
15 camps), watching her older brother code at home, and encouragement from her parents
16 to become a computer scientist in the future. Sara stated:
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24 My mom sort of started getting me into coding and computer
25 science and stuff. She was like, “In your future you could
26 become a computer science person and code robots!” At the
27 beginning, I felt like I had to do [CS] because my mom was
28 telling me to. But then when I got to know coding and stuff,
29 I was like, “It’s pretty interesting and fun.”
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38 Despite Sara’s interest, confidence, and family support for CS, at the beginning
39 of her year in BRIGHT-CS, she was anxious to avoid drawing attention to her CS skills
40 (or perceived lack thereof). This resulted in behaviors counterproductive to Sara’s
41 engagement and learning. She barely spoke to her instructor and peers, giving halting
42 answers or putting her head down when asked even the most basic questions. After a
43 group brainstorm of ideas for an internet application development project, she remained
44 silent when the instructor asked her to share her ideas, even when the other girls
45 encouraged her. Often, she sat inactive next to her partner during independent work
46 time until the instructor intervened to help them with the task.
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3 By the end of the first semester, she spoke positively about the experience of
4 working with partners whom she'd gotten to know during the program. However, she
5 remained reluctant to answer questions in front of the whole group and still spoke
6 tentatively about her CS skills. "I'm new to the things we're learning here," she said,
7 "but I'm gradually learning, I guess." Yet she re-enrolled for a second semester, even
8 convincing her twin sister to join her and recruiting additional girls into the program.
9 Eventually, she began volunteering ideas during group brainstorming and answered
10 instructor questions without prompting.
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21 By the last session, she volunteered answers to questions even after answering
22 one wrong in front of the whole group. Earlier in the program, she stated that she would
23 only raise her hand when, "I know for sure that this is the answer." Now she was willing
24 to "take chances." Once, while coding independently, she said aloud to herself, "That
25 won't work," and—when the instructor asked her why not—promptly diagnosed her error
26 in front of the whole group. She also stopped relying on the instructor to troubleshoot
27 problems and started resolving challenges on her own. She even began assisting her
28 peers when they encountered roadblocks in their own work.
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40 When asked, Sara attributed her changing comfort and engagement levels to the
41 fact that the second semester curriculum reinforced concepts and skills she'd learned in
42 the first semester. She was more confident in her knowledge because, "I've done this
43 stuff before." She also said the instructor made her feel "comfortable" speaking even
44 when she didn't know the answer. The instructor built trust with Sara by asking her
45 questions privately, "...while other people are working, not paying attention... So, like
46 everybody wasn't focusing on me." If the one-to-one conversation surfaced a
47 misconception, the instructor would normalize it by clarifying the concept for the whole
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3 class. Eventually, not knowing seemed like a part of the process, rather than something
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5 for which Sara might be judged.
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8 The instructor stated that Sara became more comfortable when the program
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10 curriculum shifted toward building skills for a long-term capstone project (i.e.,
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12 developing a web application). This relieved pressure on students to acquire new skills
13
14 quickly. With the longer focus, students weren't told, "You need to learn this by this
15
16 time; you need to learn that by that time." Students avoided becoming "overwhelmed at
17
18 having to learn so much," in a short period of time. Sara now had the time and
19
20 flexibility to productively struggle with the CS content. The instructor also reassured
21
22 students that struggle was a normal part of the learning process. This reassurance
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24 reduced Sara's anxiety around potential public humiliation and promoted the
25
26 development of independent learning behaviors.
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31 By the end of the program, Sara exhibited leadership behaviors. In a parent
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33 interview at the end of the program, her mother stated, "I'm glad you can see the Sara
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35 we see at home." She began assisting other students who were struggling, worked with
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37 the instructor to show other students how to code or debug a program, and helped
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39 recruit other girls of color to participate in the program. As she reflected on the
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41 program, she stated how she used to watch her brother code at home. Now, she knows
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43 that she can do it too.
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48 ***Key Themes***

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50 The story of Sara and her growth toward exhibiting independent learning
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52 behaviors highlights the time, patience, and care it takes to support student learning. In
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54 this case, the year-long CS learning ecosystem involved over 130 hours of the following
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56 instructional features to create a culturally responsive learning environment that
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58 supported engagement in CS/CT via independent learning behaviors and approaches.
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3 *Theme 1: Problem solving – “Give us space to figure stuff out for ourselves”*
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5 A major component of CS is framing problems in a way that a computer or other
6
7 technologies can be used to solve them (CSTA, 2017). We found that students who
8
9 primarily applied independent learning approaches enjoyed problem solving, especially
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11 when given the time and safe space needed for productive struggle (e.g., trial and error).
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13 During program observations conducted by the researchers, Sara said to herself, “That
14
15 won’t work” and continued to problem solve.
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19 As the capstone project came into focus, most of the students primarily applying
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21 independent learning approaches indicated they were eager to jump in and learn by
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23 “figuring it out” on their own (n = 8). One student who used an independent learning
24
25 approach said, “They give us a lot of space to figure stuff out for ourselves, which is
26
27 really nice. [It’s] unlike school where you watch videos of people doing it and you don’t
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29 really get to experience doing it. Here, we can actually try stuff instead of just looking
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31 at [teachers] doing it.”
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35 In contrast, students who applied dependent learning tactics were less inclined to
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37 pursue new skills or knowledge on their own. They only engaged if asked to reproduce
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39 instructor-modeled skills or knowledge without deep cognitive effort. Students who
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41 primarily applied a dependent learning approach would often answer simple recall
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43 questions or follow an instructor’s step-by-step procedures (n = 5), but they were less
44
45 likely than students with independent learning tendencies to continue working on their
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47 products when doing so required independently honing new skills. Hammond (2015)
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49 indicates that this “learned helplessness” is rooted in students’ beliefs that they do not
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51 have the capacity to improve.
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55 Students’ beliefs about what it means to master new skills and knowledge (i.e.,
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57 “to understand”) differed according to their preference for “figuring it out” or
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3 reproducing instructor procedures. Students with independent learning tendencies were
4 much more likely than their peers to associate understanding with knowing why or how
5 something “works” (n = 8). Students who primarily applied a dependent learning
6 approach were more likely to associate understanding with being able to name and
7 remember the steps for completing a task (n = 3).
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15 One student with dependent learning behavior tendencies said that math was her
16 favorite subject in school “because you just memorize stuff” and another said that “math
17 is pretty easy to me”, because “once I understand the method, it’s pretty easy to move
18 on.” This stands in contrast to students who were primarily independent learners who
19 complained about feeling like “I don’t understand” in math class because “I just know
20 [the steps] to do it. I don’t know why [the steps work].” A few students who were
21 primarily dependent learners did express frustration with the limits of procedural
22 learning, if not an awareness that the methods they favored for learning could be to
23 blame.
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36 Two students who were primarily dependent learners—who dutifully engaged
37 when BRIGHT-CS instructors invited them to follow along with a coding
38 demonstration—admitted in interviews to feeling that their understanding remained low.
39 One such student said, “[I’m] typing random words that don’t make sense.” Another
40 stated, “I might change my mind if I actually understood how to code, but right now I’m
41 just seeing a bunch of words.” This suggests that students who primarily exhibit
42 dependent learning behaviors may desire the deeper learning that comes from “figuring
43 it out” themselves, but they gravitate towards procedural learning. This could be an
44 indicator that they have yet to develop the cognitive strategies or social-emotional skills
45 to embrace opportunities for trial-and-error and other strategies that students who
46 primarily apply an independent learning approach prefer to use in STEM+CS.
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3 *Theme 2: Iterations – “I decided to improve it”*
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5 Another important skill in CS is iterating, being able to update and improve
6 products (CSTA, 2017). We found that students who were primarily independent
7 learners were never “done” with their projects, almost to the annoyance of their
8 dependent-learning peers during collaboration. Independent learning students sought
9 new ways to improve their design, program, or computing artifact. Sara was even
10 known to bring computing artifacts home to continue working on improvements.
11

12 In BRIGHT-CS, projects were often designed to allow for some choice in the
13 content, functionality, and visual presentation of the final product. Students who were
14 primarily dependent learners engaged in those aspects of their product where they
15 already had skill, such as in perfecting the visual presentation of their product, rather
16 than working on aspects of the project that challenged them to develop new skills, e.g.,
17 coding (n = 4). When working on a project to design and program a robot, one student
18 explained that she chose a “simplistic” robot because, “I didn’t want to drive myself
19 crazy with coding.” Once the robot had basic functionality, she was eager to move on
20 from learning CS to applying her artistic skills to “make [the robot] pretty.” When
21 working in groups, students who primarily apply a dependent learning approach
22 sometimes even avoided intellectual labor around the computing aspects of their
23 product entirely, forcing others to carry the cognitive load.
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46 By contrast, students who primarily apply an independent learning approach
47 were more likely to spend their time developing new skills in order to make
48 improvements to their product (n = 6). One student who designed a functioning robot to
49 turn her bedroom light switch on and off said, “Now I’m trying to figure out how to not
50 attach [it] to the computer... so I don’t have to carry the computer with me... To add it
51 to mobile devices had a lot of complicated stuff, but it’s going to help me.”
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3 Another difference between independent and dependent learning approaches
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5 involved how students decided their products were “finished” and, subsequently, that it
6
7 was time to stop working. Because students determined the features and functionality of
8
9 their project, the decision as to when their product was “finished” was largely theirs to
10
11 make. Both dependent and independent learners found that integrating additional
12
13 functionality into their product would require CS skills or knowledge beyond the basics
14
15 that the instructor had taught the whole class. In instances like these, dependent learners
16
17 were more likely to declare their product “finished” when it had yet to perform all of its
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19 intended functions, rather than work independently to develop the additional skills and
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21 knowledge they would need (n = 3) to complete their project.
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28 *Theme 3: Debugging – “It’s actually kind of fun”*
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30 A third component of CS skills is debugging or identifying and fixing errors in a
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32 program (CSTA, 2017). While we found that both independent and dependent learners
33
34 ran into challenges while working on their projects, the strategies they used to get
35
36 themselves “unstuck” varied significantly. Sara provided a good example of how
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38 debugging could initially be anxiety-provoking. At one point she stated, “At first, I
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40 thought [Python] was really hard, and it was complicated. I would just try to think about
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42 the easy stuff and then try to do that so that I could encourage myself”. However,
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44 towards the end of the program, she would “take chances” and openly present her
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46 mistakes and the various strategies she used to fix her code.
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50 Students who exhibited dependent learning behaviors were more likely to
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52 request help from instructors as their primary debugging strategy (n = 4). One
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54 dependent learner described the frustrating experience of losing her prior day’s coding
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56 because, “It didn’t save like I wanted [and] I don’t remember what to do.” Rather than
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58 try to recover her code using her own strategies, such as working together with
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3 classmates who'd also participated in the prior day's coding lesson, she relied on the
4 instructor to "show the whole class how to do it again."
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8 Independent learners also asked instructors for help, but they were much more
9 likely to tackle challenges on their own by creating a plan of attack, rereading
10 instructor-provided materials, conducting their own research, focusing on the easy parts
11 first, or using trial and error (n = 10). One independent learner said, "Help wasn't
12 always available. I had to figure out what I did wrong, how to fix it, and not make the
13 same mistakes again—through trial and error and applying what I already knew. That's
14 something I just kind of do [in all my classes]." Some independent learners even said
15 that solving their own problems using trial and error was ultimately more instructive or
16 satisfying than calling on instructors (n = 3). "I just kept trying new things, even though
17 sometimes I knew it wasn't going to work. I still wanted to try. Because it's a trial-and-
18 error process and it's actually kind of fun."
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35 **Discussion**

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37 The study looked at the effects of a CS learning ecosystem on 53 girls (grades 6-
38 8) and their engagement, learning, and persistence in CS. The study found two main
39 results. First, for students to engage, learn, and persist, students need to practice
40 independent learning behaviors; that is, moving independently through the learning
41 process without extensive guidance or support from the instructors by focusing on key
42 CS principles such as problem solving, iterating, and debugging.
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51 Second, for students to practice such behaviors, educators need to actively foster
52 culturally responsive learning environments that create hybrid spaces for students to
53 develop STEM+CS identities, support girls through the productive struggle of learning,
54 and to nurture independent learning behaviors and approaches.
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Prior research has shown the importance of persistence through grit (Duckworth *et al.*, 2007), growth mindset (Dweck, 2006; Haimovitz and Dweck, 2017), and self-affirmation (Binning *et al.*, 2019; Harackiewicz *et al.*, 2016) to student learning. While students reported that they were highly interested and very confident in learning CS, our results highlight the importance of purposefully and explicitly developing students' independent learning behaviors. To bridge the gap between high student interest in CS and low representation and participation in CS pipelines, students need a sustained focus on growth mindset as they go through the learning process, with educators actively developing and nurturing independent learning behaviors.

We found that moving from dependent to independent learning behaviors does not occur in a vacuum or naturally. For equitable learning experiences, it is necessary for instructors and educators to explicitly and intentionally model, teach, and coach students within a safe and patient learning environment (Darling-Hammond, 2001; Hammond, 2015; Oakes, 2005).

Implications for future research

While the study focused on a CS learning ecosystem (i.e., BRIGHT-CS) and CS engagement, learning, and persistence, future research should include applying the ecosystem approach to other subject areas with a disconnect between student interest and student outcomes. For example, Black students had the highest percentage of reporting confidence in their ability to excel on mathematics tests (70%) and certainty that they can master mathematics skills (77%), higher than Asian (68% and 74%), Hispanic (65% and 71%), and white (65% and 68%) students respectively (National Science Board, National Science Foundation, 2019). Yet, Black students had the lowest rates of taking AP math or science courses and exams, ranging from 2-7%, compared to White (45-56%), Asian (13-34%), and Hispanic (9-22%) students (National Science

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3 Board, National Science Foundation, 2019). This disconnect between student
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5 confidence and interest with student access and opportunity is indicative of inequitable
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7 systems, not deficits in students. More nuanced research with critical interrogation of
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9 assumptions and interpretation of data is needed to identify and rectify gaps in the
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11 system, not deficits in students.
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15 Future research should also explore how the independent learning behaviors and
16
17 approaches can be applied to other learning experiences and across disciplines. More
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19 research is needed to understand whether the independent learning approach is specific
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21 to CS, with learning and applying computational thinking principles such as iteration,
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23 debugging, and abstraction, or whether this type of learning approach can be taught and
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25 supported in other STEM fields.
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30 ***Limitations to the study***

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32 There were three main limitations to the study. First, the study focused on those
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34 students who were a part of the program and did not utilize a comparison group of girls
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36 who did not participate in the program. As a result, we cannot compare differences in
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38 student outcomes between a treatment and comparison group. Second, the study
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40 collected data across two school years (2018-2019 and 2019-2020) wherein
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42 implementation of the program not only varied across the four middle schools, but the
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44 2019-20 cohort of students was not able to finish the program when schools shifted to
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46 online learning in March 2020 due to the COVID-19 global pandemic. Finally,
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48 quantitative survey data of interest and attitudes towards CS was only captured at
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50 baseline (pre-program) because the study design prioritized measuring observable CS
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52 behaviors via qualitative coding; as a result, quantitative analysis of changes in CS
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54 attitudes were not prioritized in the study design.
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References

- Aronson, J., Cohen, G., McColskey, W., Montrosse, B., Lewis, K., & Mooney, K. (2009). Reducing stereotype threat in classrooms: A review of social-psychological intervention studies on improving the achievement of Black students (Issues & Answers Report, REL 2009–No. 076). Washington, DC: U.S. Department of Education, Institute of Education Sciences, National Center for Education Evaluation and Regional Assistance, Regional Educational Laboratory Southeast.
- Binning, K. R., Cook, J. E., Purdie-Greenaway, V., Garcia, J., Chen, S., Apfel, N., & Cohen, G. L. (2019). Bolstering trust and reducing discipline incidents at a diverse middle school: How self-affirmation affects behavioral conduct during the transition to adolescence. *Journal of School Psychology, 75*(August), 74-88. doi: 10.1016/j.jsp.2019.07.007.
- Calabrese Barton, A., Kang, H., Tan, E., O'Neill, T. B., Bautista-Guerra, J., & Brecklin, C. (2013). Crafting a future in science: Tracing middle school girls' identity work over time and space. *American educational research journal, 50*(1), 37–75.
- Calabrese Barton, A., Tan, E., & Rivet, A. (2008). Creating hybrid spaces for engaging school science among urban middle school girls. *American Educational Research Journal, 45*(1), 68-103. doi: 10.3102/000283120730864
- Carlone, H. B., Haun-Frank, J., & Webb, A. (2011). Assessing equity beyond knowledge and skills-based outcomes: A comparative ethnography of two fourth-grade reform-based science classrooms. *Journal of Research in Science Teaching, 48*, 459–485. doi:10.1002/tea.20413
- Carter Andrews, D. J., Brown, T., Castro, E., & Id-Deen, E. (2019). The impossibility of being "white and perfect": Black girls' racialized and gendered schooling

1
2
3 experiences. *American Educational Research Journal*, 56(6), 2531-2572. doi:
4
5 10.3102/0002831219849392.
6

7
8 Computer Science Teachers Association (2017). *CSTA K-12 Computer Science*
9
10 *Standards, Revised 2017*. doi: 10.1145/2593249.
11

12 Darling-Hammond, L. (2001). Inequality in Teaching and Schooling: How Opportunity
13
14 Is Rationed to Students of Color in America. In *The Right Thing to Do, The*
15
16 *Smart Thing to Do: Enhancing Diversity in the Health Professions -- Summary*
17
18 *of the Symposium on Diversity in Health Professions in Honor of Herbert W.*
19
20 *Nickens, M.D. Institute of Medicine 2001. Washington, DC: The National*
21
22 *Academies Press.* <https://doi.org/10.17226/10186>.
23
24

25
26 Delpit, L. (2006). *Other People's Children: Cultural Conflict in the Classroom*. New
27
28 *York, NY: The New Press.*
29

30
31 Delpit, L. (2012). *Multiplication is for White People: Raising Expectations for Other*
32
33 *People's Children*. New York, NY: The New Press.
34

35
36 Denzin, N. K. & Lincoln, Y. S. (Eds.) (2018). *The Sage Handbook of Qualitative*
37
38 *Research*. Thousand Oaks, CA: SAGE.
39

40
41 Duckworth, A. L., Peterson, C., Matthews, M. D., & Kelly, D. R. (2007). Grit:
42
43 *Perseverance and passion for long-term goals*. *Journal of Personality and Social*
44
45 *Psychology*, 92(6), 1087-1101. doi: 10.1037/0022-3514.92.6.1087.
46

47
48 Dweck, C. S. (2006). *Mindset: The New Psychology of Success*. New York: Random
49
50 *House.*

51
52 Farinde, A. A., & Lewis, C. W. (2012). The underrepresentation of African American
53
54 *female students in STEM fields: Implications for classroom teachers.* *US-China*
55
56 *Education Review*, 421-430.
57
58
59
60

- 1
2
3 Eccles, J. S., Midgley, C., Wigfield, A., Buchanan, C. M., Reuman, D., Flanagan, C., &
4
5 Mac Iver, D. (1993). Development during adolescence: The impact of stage-
6
7 environment fit on young adolescents' experiences in schools and in families.
8
9 *American Psychologist*, 48(2), 90-101. doi: 10.1037//0003-066x.48.2.90.
10
11
12 Eisner, E. W., & Peshkin, A. (Eds.). (1990). *Qualitative Inquiry in Education: The*
13
14 *Continuing Debate*. New York, NY: Teachers College Press.
15
16
17 Epstein, R., Blake, J., & Gonzalez, T. (2017). *Girlhood Interrupted: The Erasure of*
18
19 *Black Girls' Childhood*. Washington DC: Georgetown University, Georgetown
20
21 Law Center on Poverty and Inequality. doi: 10.2139/ssrn.3000695.
22
23
24 Garson, G. D. (2002). Case Study Research in Public Administration and Public Policy:
25
26 Standards and Strategies. *Journal of Public Affairs Education*, 8(3), 209-216.
27
28 doi: 10.1080/15236803.2002.12023551.
29
30
31 Google Inc., & Gallup Inc. (2015). *Images of Computer Science: Perceptions among*
32
33 *Students, Parents, and Educators in the U.S.* Mountain View, CA: Google Inc.
34
35
36 Google Inc., & Gallup Inc. (2016). *Diversity Gaps in Computer Science: Exploring the*
37
38 *Underrepresentation of Girls, Blacks, and Hispanics*. Mountain View, CA:
39
40 Google Inc.
41
42
43 Grissom, J. A., & Redding, C. (2016). Discretion and disproportionality: Explaining the
44
45 underrepresentation of high-achieving students of color in gifted programs.
46
47 *AERA Open*, 2(1), 1-25. doi: 10.1177/2f2332858415622175.
48
49
50 Haimovitz, K., & Dweck, C. S. (2017). The origins of children's growth and fixed
51
52 mindsets: New research and a new proposal. *Child Development*, 88(6), 1849-
53
54 1859. doi: 10.1111/cdev.12955.
55
56
57 Hammond, Z. (2015). *Culturally Responsive Teaching and the Brain*. Thousand Oaks,
58
59 CA: Corwin.
60

- 1
2
3 Harackiewicz, J. M., Canning, E., A., Tibbetts, Y., Priniski, S. J., & Hyde, J. S. (2016).
4
5 Closing achievement gaps with a utility-value intervention: Disentangling race
6
7 and social class. *Journal of Personality and Social Psychology*, 111(5), 745-765.
8
9 doi: 10.1037/pspp0000075.
10
11
12 Hill, C., Corbett, C., & St. Rose, A. (2010). *Why So Few? Women in Science,*
13
14 *Technology, Engineering, and Mathematics*. Washington DC: American
15
16 Association of University Women.
17
18
19 Hill, C. E., Thompson, B. J., & Williams, E. N. (1997). A guide to conducting
20
21 consensual qualitative research. *The Counseling Psychologist*, 25(4), 517-572.
22
23 doi: 10.1177/2f0011000097254001.
24
25
26 Huberman, A. M., & Miles, M. B. (1994). Data Management and Analysis Methods. In
27
28 N. K. Denzin & Y. S. Lincoln (Eds.), *Handbook of Qualitative Research* (pp.
29
30 428-443). Thousand Oaks, CA: Sage.
31
32
33 Ireland, D.T., Freeman, K.E., Winston-Proctor, C.E., DeLaine, K.D., McDonald Lowe,
34
35 S., & Woodson, K.M. (2018). (Un)hidden figures: A synthesis of research
36
37 examining the intersectional experiences of Black women and girls in STEM
38
39 education. *Review of Research in Education*, 42(1), 226-254.
40
41
42 Kang, H., Calabrese Barton, A., Tan, E., D Simpkins, S., Rhee, H. Y., & Turner, C.
43
44 (2019). How do middle school girls of color develop STEM identities? Middle
45
46 school girls' participation in science activities and identification with STEM
47
48 careers. *Science Education*, 103(2), 418-439. doi: 10.1002/sce.21492
49
50
51 King, N. S. (2022). Black girls matter: A critical analysis of educational spaces and call
52
53 for community-based programs. *Cultural Studies of Science Education* 17, 53-
54
55 61. doi: 10.1007/s11422-022-10113-8
56
57
58
59
60

- 1
2
3 King, N. S., & Pringle, R. M. (2019). Black girls speak STEM: Counterstories of
4
5 informal and formal learning experiences. *Journal of Research in Science*
6
7 *Teaching*, 56(5), 539-569. doi: 10.1002/tea.21513
8
9
10 Leavy, P. L. (Ed.) (2015). *The Oxford Handbook of Qualitative Research*. Oxford:
11
12 Oxford University Press.
13
14 Levitt, H. M., Bamberg, M., Creswell, J. W., Frost, D. M., Josselson, R., & Suarez-
15
16 Orozco, C. (2018). Journal article reporting standards for qualitative primary,
17
18 qualitative meta-analytic, and mixed methods research in psychology: The APA
19
20 publications and communication board task force report. *American Psychologist*,
21
22 73(1), 26-46. doi: 10.1037/amp0000151.
23
24
25
26 Love, B. (2019). *We Want to Do More Than Survive: Abolitionist Teaching and the*
27
28 *Pursuit of Educational Freedom*. Boston, MA: Beacon Press.
29
30
31 Maxwell, J. A. (1998). Designing a Qualitative Study. In L. Bickman & D. Rog (Eds.),
32
33 *Handbook of Applied Social Research Methods* (pp. 69-100). Thousand Oaks:
34
35 Sage.
36
37
38 McGee, E. O. (2013). High-achieving black students, biculturalism, and out-of-school
39
40 STEM learning experiences: Exploring some unintended consequences. *Journal*
41
42 *of Urban Mathematics Education*, 6(2), 20-41. doi: 10.21423/jume-v6i2a178
43
44
45 McGee, E. O., & Stovall, D. (2021). *Black, Brown, Bruised: How Racialized STEM*
46
47 *Education Stifles Innovation*. Cambridge, MA: Harvard Education Press.
48
49
50 National Science Board, National Science Foundation (2019) *Elementary and*
51
52 *Secondary Mathematics and Science Education. Science and Engineering*
53
54 *Indicators 2020 (NSB-2019-6)*. Alexandria, VA.
55
56
57
58
59
60

- 1
2
3 Noor, K. B. M. (2008). Case study: A strategic research methodology. *American*
4
5 *Journal of Applied Sciences*, 5(11), 1602-1604. doi:
6
7 10.3844/ajassp.2008.1602.1604
8
9
10 Nottingham, J. A. (2017). *The Learning Challenge: How to Guide Your Students*
11
12 *Through the Learning Pit to Achieve Deeper Understanding*. Thousand Oaks,
13
14 CA: Corwin.
15
16
17 Nottingham, J. A., & Larsson, B. (2018). *Challenging Mindset: Why a Growth Mindset*
18
19 *Makes a Difference in Learning--And What to Do When It Doesn't*. Thousand
20
21 Oaks, CA: Corwin.
22
23
24 Oakes, J. (2005). *Keeping Track: How Schools Structure Inequality*. New Haven, CT:
25
26 Yale University.
27
28
29 Ong, M., Wright, C., Espinosa, L., & Orfield, G. (2011). Inside the double bind: A
30
31 synthesis of empirical research on undergraduate and graduate women of color
32
33 in science, technology, engineering, and mathematics. *Harvard Educational*
34
35 *Review*, 81(2), 172-209.
36
37
38 Pollock, M. (2017). *Schooltalk: Rethinking What We Say About--And To--Students*
39
40 *Every Day*. New York, NY: The New Press.
41
42
43 Rosen, J. A., Glennie, E. J., Dalton, B. W., Lennon, J. M., & Bozick, R. N. (2010).
44
45 *Noncognitive Skills in the Classroom: New Perspectives on Education Research*.
46
47 Research Triangle Park, NC: RTI Press.
48
49
50 Safir, S., & Dugan, J. (2021). *Street Data: A Next Generation Model for Equity,*
51
52 *Pedagogy, and School Transformation*. Thousand Oaks, CA: Corwin Press.
53
54
55 Saldana, J. (2009). *The Coding Manual for Qualitative Researchers*. Thousand Oaks,
56
57 CA: Sage Publications.
58
59
60

1
2
3 Shores, K., Kim, H. E., & Still, M. (2020). Categorical inequality in Black and White:
4

5 Linking disproportionality across multiple educational outcomes. *American*
6

7 *Educational Research Journal*, 57(5), 2089-2131.
8

9 doi:10.3102/0002831219900128.
10

11
12 Strauss, A., & Corbin, J. (1990). Basics of Qualitative Research: Grounded Theory
13

14 Procedures and Techniques. Newbury Park: Sage.
15

16
17 Yin, R. K. (2009). Case Study Research: Design and Methods. Thousand Oaks: Sage.
18

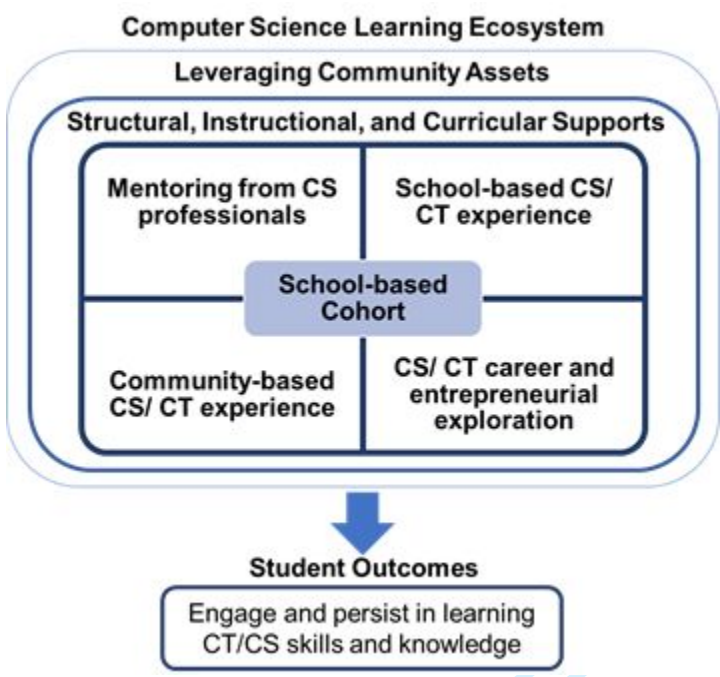
19 Young, J. L., Young, J. R., & Ford, D. Y. (2017). Standing in the gaps: Examining the
20

21 effects of early gifted education on black girl achievement in STEM. *Journal for*
22

23 *Multicultural Education*, 28(4), 290-312. doi: 10.1177/1932202x17730549.
24
25
26
27
28
29
30
31
32
33
34
35
36
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Figure 1: BRIGHT-CS program theoretical framework



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Table I: Crosswalk of BRIGHT-CS program components, CS standards, and student experiences

BRIGHT-CS components	K-12 CSTA CS Standards (2017)	Student experiences
(1) School-based CS/CT experience: 9-month weekly after-school club	- Collaborating around computing - Recognizing and defining computational problems	- Collaborated through small learning teams - Identified problems to solve using technology
(2) Community-based CS/CT experience: 2-week summer camp	- Creating computational artifacts - Testing and refining computational artifact	- Learned different coding languages to create and test shopping apps, games, art projects, and robots
(3) CT/CS career and entrepreneurial exploration	- Fostering inclusive community - Communicating about computing	- Presented computing artifacts to school leaders, parents, and mentors
(4) Mentoring from CS professionals		- Connected with mentors about CS careers