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Supporting classroom orchestration with real-time feedback: A role for teacher dashboards and real-time agents

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Abstract

This paper investigates the role of the physical classroom environment, coupled with a 13technology environment that includes real-time agents and data analytics, to support the 14 orchestration of complex collaborative inquiry designs in a high school physics class-15room. This design-based research contributes to the wider domain of scripting and 16orchestration (e.g., Dillenbourg 2012; Dimitriadis 2012; Fischer et al. 2013). Guided by 17 a theoretical perspective of learning in knowledge communities (Authors, in press), we partnered with a physics teacher to co-design curricular activities and assessments that 19engaged students in collectively solving, tagging and evaluating physics problems, 20creating a knowledge base of student-contributed examples, and using those examples 21as a resource in collaborative inquiry challenges. To support the teacher in orchestrating 22such a complex curricular design, we developed a tablet application that allowed the 23teacher see the state of the class in real-time, control the flow of activities and helped him 24know when and where he was needed within the flow of class activities. The tablet 25leveraged a set of specially designed real-time software agents to process student 26interactions in real time, allowing dynamic orchestration of student groups, material 27allocation, and teacher notifications. The paper begins with a review of recent literature 28on scripting and orchestration, drawing connection to the theoretical perspective of 29knowledge communities. We then describe our theoretical model, the design-based 30 method, and details of our curriculum and technology environment. The paper concludes 31with a summary of how the teacher tablet and the real-time software agents helped 32support the teacher's real-time facilitation and orchestration. 33

Keywords Learning · Collaboration · Orchestration · Teacher dashboards

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Introduction

Many scholars have advocated for learning designs that build on socioscientific issues, support37twenty-first century learning, and connect learners across formal and informal learning contexts.38New media and technologies open the door to the design of powerful social forms of interaction,39including Web 2.0 aggregation of user-contributed content, social tagging, voting, and collabora-40tive editing (e.g., wikis). Theoretical work in CSCL has also proceeded, defining "collaboration41scripts" (Fischer et al. 2013), and the orchestration of such scripts in technology rich environments42(Dillenbourg and Jermann 2007; Schwarz et al. 2018; Tchounikine 2016).43

The research described here builds on the notion of learning communities, in which all students 44 in a classroom work collectively to develop a knowledge base that can serve as a resource for their 45further inquiry. We apply The Knowledge Community and Inquiry (KCI) model to specify a 46 complex collaboration script where students are assigned to a progressive sequence of groups (e.g., 47jigsaw), with context-sensitive materials, real-time collaboration amongst students (e.g., co-editing 48 documents, jointly voting, and tagging), and dynamic, "emergent" representations of student ideas 49and resources. These elements can be presented and orchestrated across a wide range of devices 50(laptops, tablets), displays (surfaces, walls, tables) and other interactive media. The presence, for 51example, of large projected displays can serve as a vital reference for teacher-led discourse about 52class progress (Tissenbaum and Slotta 2019). 53

We applied the KCI model to develop a high school physics curriculum, leveraging a range of technologies and learning analytic approaches to orchestrate students in their assignments to groups, allocation of materials and activities, and collection and aggregation of resources. Forming a co-design partnership with the teacher (Roschelle et al. 2006) we designed a semester-length course in which students developed a sophisticated web of user-contributed content that was socially and semantically tagged, serving as a source of materials for subsequent inquiry activities and informing the large, dynamic displays of their emergent knowledge.

We begin with a review of the literature surrounding collective inquiry and learning 61communities, including the role of scripting and orchestration, and identify a possible role 62for real-time software agents as a means of orchestrational support. We follow this with the 63 description of our curriculum, focusing on the co-designed culminating smart classroom 64activity, and the technology framework we developed to support its enactment, called SAIL 65Smart Space (S3). We then analyze S3 and its software architecture in terms of its ability to 66 support the enactment and orchestration of real-time inquiry activities, with a focus on the 67 tablet-based real-time teacher dashboard. We conclude with a discussion of the role played by 68 real-time agents and other data-driven orchestration supports within our knowledge commu-69 nity and inquiry curriculum. 70

Collective inquiry and learning communities

One promising approach to the design of active learning is to consider the entire classroom as a 72learning community (i.e., as opposed to each student learning independently). In its most 73 simple form, this occurs whenever an instructor asks for a show of hands, or uses a clicker-74system to show students how their opinions on some problems may be distributed within the 75community. A more elaborate application of this approach involves user-contributed content, 76where the whole class is asked to contribute resources to form a collective knowledge base, 77 such as a Pinterest board, a wiki, or a Google Doc. In this approach, each student feels as if 78they are contributing something to a larger corpus that will be consequential for the 79

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community's progress. Engaging students in contributing, curating and applying their own 80 content to inquiry projects is a daunting challenge for educators – even those who are 81 experienced in inquiry-oriented methods. 82

In a learning community approach, students bring their diverse interests and expertise to some 83 common goal. They must all hold a shared understanding that their learning activities will align to 84 advance the community's cause while at the same time helping individuals learn, and allowing 85 everyone to benefit from the community's resources (Bielaczyc et al. 2006). In a review of learning 86 community models, Slotta and Najafi (2013) articulated three common characteristics: (1) an 87 epistemic commitment to collective advancement, (2) a shared community knowledge base, and 88 (3) common modes of discourse. Several scholars have observed that it is challenging for teachers 89 or researchers to coordinate a learning community approach (van Aalst and Chan 2007). As 90 observed by Kling and Courtright (2003, p. 221), "developing a group into a community is a major 91 accomplishment that requires special processes and practices, and the experience is often both 92frustrating and satisfying for the participants". The limited success or uptake of this approach has 93 been due to the pragmatic and epistemic challenges of shifting from a didactic mode of "knowl-94edge transmission" into one of collective inquiry. But it is also due to the lack of explicit models to 95guide the design of curriculum where students are interconnected in a progression of individual, 96 small group and whole class activities, creating and consuming materials from a community 97 knowledge base (Slotta and Peters 2008). 98

The Knowledge Community and Inquiry (KCI) model (Fig. 1) guides the design of science 99 curricula in which the whole class (or even multiple class sections) work together, with all 100students held accountable for content learning gains (Slotta and Peters 2008). The model 101includes principled requirements for (1) a knowledge base that is indexed to the targeted 102science domain (2) collective, collaborative and individual inquiry activities in which students 103co-construct the knowledge base and then use it as a resource for further inquiry, and (3) 104assessable learning outcomes that allow teachers to evaluate student progress. The teacher has 105a scripted role within a KCI design, but also plays a general orchestration role, with aid from a 106technology environment that coordinates group assignments, material allocation, aggregation 107of content into "emerging learning objects," and real-time processing of student interactions. 108

Within KCI curriculum, students are typically engaged in computer-supported inquiry activi-109ties, including note taking, observations, brainstorms, problem solving, modeling and simulation, 110design and argumentation. Prior KCI research has developed sophisticated server software known 111 as the Scalable Architecture for Interactive Learning (SAIL) that captures student contributions 112(i.e., the knowledge base), and client applications for students and teachers that support the 113collection, distribution, curation and application of that content. This software infrastructure, 114collectively known as SAIL Smart Space (S3) provides a flexible foundation for collective inquiry, 115and was extended and applied in the proposed work, supporting (1) the development of real-time 116agents that influence student grouping and the distribution of materials; (2) the application of large, 117dynamic displays (e.g., projectors or smart boards) of the community's emergent knowledge in 118 influencing discourse; and (3) teacher orchestration tools, including representations of the state of 119the class and flow-control applications. 120

Scripting in CSCL activities

Curricular designs that engage students as a learning community and integrate rich inquiry and 122 technology environments are likely to be more complex and dynamic then in previous 123 generations of CSCL. Increasingly, designs will need to include the configuration (and 124

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Fig. 1 The KCI Model

potentially dynamic reconfiguration based on emergent patterns) of student groups and 125 activities, teacher roles, and technologies. In order to deal with this complexity, many 126 researchers have advocated for the development of pedagogical "scripts" that can help guide 127 students through complex inquiry tasks by segmenting the learning into more cognitively 128 manageable phases and providing instruction on the formation of groups, distribution of roles, 129 phases of work, and expected deliverables (Kirschner et al. 2004; Dillenbourg 2002). 130

Inquiry-based scripts often span multiple class sessions, consuming weeks or even months 131of curriculum time, and thus need to accommodate multiple scales of time, student configu-132ration, and contexts (Lemke 2000; White 2018). In order to respond to the varying granular-133ities of a curriculum (i.e., across space and time, as well as other variables), designers need to 134think both in terms of the macro script, which describes the overall goals and timing of 135individual activities (e.g., a field-trip, or a homework task), as well as the finer grain scripts, 136which specify the individual homework items, student work groups, materials, tools, and 137scaffolds (Tissenbaum and Slotta 2019). 138

Of particular interest within and across such scripts, are the granularities of student 139 collaboration. Dillenbourg and Jermann (2007) define five general grain sizes of activity: the 140 individual phase; the group phase; the class phase; the community phase (influencing peers 141 outside one's classroom); and the world phase (contributions to the wider public). In designing 142

curricular scripts, it is critical to ensure the granularity of the task (i.e., the work to be done) 143 matches the granularity of the group size, as a poor match can significantly hinder student 144 learning (Lemke 2000; White 2018). 145

Within a script, group configuration serves to formalize student roles. Students may 146alternate in roles such as presenter, discussion leader, moderator, or devil's advocate (Soller 147 2001; Palincsar and Herrenkohl 2002). Scripting group configurations can allow different 148materials to be distributed amongst group members, reducing the need for every student to 149have all the knowledge "in their own skulls" (Hollan et al. 2000), or to require collaboration in 150order for them to complete tasks (similar to Jigsaw groups - Aronson, 1978). For instance, in 151Q3 Alien Contact (Dunleavy et al. 2009), each student in a group was assigned a role (Chemist, 152Cryptologist, Computer Hacker, and FBI Agent). Depending on their role, each member 153received different data on their handheld device about an alien artifact (e.g., a spaceship wing), 154which they had to share in order to determine its significance. This distribution of roles and 155information fostered positive interdependence and cross-disciplinary knowledge sharing and 156higher-order thinking skills amongst group members. 157

Some researchers argue that there is a need to understand the varying strengths, weak-158nesses, background knowledge, and interests of students when configuring groups in order to 159ensure productive outcomes (O'Donnell and Dansereau 1992). At the outset of many inquiry 160curricula, such detailed information on individual group members may not be easily available, 161 only coming to light during the curriculum's enactment. It may therefore be necessary to 162capture and process information on individual and group performance (either by the teacher or 163the system itself) to enable adaptive group or material assignments. Technology can play a 164vital role in this regard, as requiring teachers to process large amounts of interactional data 165(e.g., responses to assessments, preferences, or patterns of engagement) on their own would be 166 prohibitive (Tissenbaum et al. 2012). 167

Orchestration of scripted activities

As described above, complex inquiry scripts – especially those involving technology environ-169ments and real-time or adaptive conditions – can place a heavy load on teachers, requiring 170them to simultaneously organize materials, assign student roles and groups, and track individ-171ual, group, and whole class progression through activities (Dimitriadis 2012). Several scholars 172have advanced the notion of Orchestration to define the enactment of such scripts in both the 173short- and long-term, across multiple contexts and social levels (e.g., Dillenbourg et al. 2009). 174Whereas scripting deals with the structuring of activities *before they are enacted*, orchestration 175is concerned with the regulation of an activity once it has begun (Soller 2001). Orchestration 176introduces a level of flexibility to the execution of a script, allowing for the "re-scripting" of 177groups, student roles, materials presented, and even which steps come next. This is especially 178important in inquiry-based curricula, which often require the ability to adapt in response to 179emergent class patterns, community voices, or new and interesting avenues for investigation. 180

Orchestration places the teacher at the center of the learning process as a "conductor," 181 orchestrating a broad range of activities (Kollar et al. 2011). Rather than as a knowledge 182 provider, the teacher is responsible for making timely and context-relevant adjustments to the 183 script based on assessments of individual and whole class progress, collaboration and growth 184 of ideas (Sharples 2013). While this support of activity progression and resource distribution 185 could theoretically be done without technology supports, many have argued that technology 186 environments can make the process "smoother" and reduce the teacher's "orchestrational load"

particularly in scripts that require the tracking of every student in the class and their
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 individual resource needs (Dillenbourg 2012; Nussbaum et al. 2009).

Technologies that work in support of the orchestration of classroom activities generally fall 190under one of two complementary forms: Orchestration Technologies and Orchestrable Tech-191nologies (Tchounikine 2013). Orchestration technologies directly support the teacher in 192managing curricular activities (Dillenbourg et al., 2011). In Edunova (Roschelle et al. 2010), 19304 for example, students are sent fractions problems on their handheld devices to collaboratively 194solve in small groups. As groups submit answers, the teacher (on his or her personal device) 195sees a color-coded matrix letting him know which students got the answer right on the first try 196(green), within a specified number of tries (yellow), or if they failed to get the answer right 197within a specified number of tries (red). Given this detailed information the teacher can enact 198formative assessments and adapt his actions in response to specific student needs. A similar 199approach can be seen with Texas Instruments Nspire Navigator system, which allows the 200teacher to control the flow of the class through actions such as beaming a student's screen to 201the front of the class or sending quizzes to students' calculators (Clark-Wilson 2010). In this 202case the teacher can use the system to generate formative assessment of the class' knowledge 203and adapt the orchestration of the classroom activities. 204

Orchestrable technologies are those whose precise function can be determined or adapted 205both before and during an activity. In some cases, orchestrable technologies can add a layer of 206flexibility to the script by allowing for fine-tuning or real-time adapting of the script by 207teachers, students, or the system itself. For instance, in EvoRoom (Lui and Slotta 2014), 208students are immersed in a simulated rainforest as they conduct investigations about flora and 209fauna. The teacher is equipped with an "orchestration tablet" that allows her to advance or 210retreat the date of the simulation across millions of years, depending on the kinds of habitat and 211ecology she wants the students to investigate. In this way, the teacher can adapt the conditions 212of the classroom in response to emergent class patterns, questions, or inquiry needs. 213

What is critically important in the examples above is that they provide specific insight into 214the state of the class, without requiring that the teacher (or TA) take any specific action. Rather, 215the technologies simply provided information to help them make decisions. Other orchestra-216tion technologies may have a more direct role in controlling the flow of activities. Cognitive 217Tutors, for example, (e.g., Anderson et al., 1995) employ student models to provide timely 21805 prompts and progress students through activities based on their past work, freeing up teachers 219to help those students most in need. However, such fully automated systems have been shown 220to be prone to "gaming the system" and other off-task behaviors (Baker et al. 2004). 221

Software agents

With the ability to capture and process data from students' interactions within technology 223environments in real time, important patterns or insights can be made invisible that would 224otherwise be too time consuming for teachers to compile on their own. One form of 225technology that can serve an orchestrating role includes "software agents" – small, active 226software elements that respond to pre-specified contexts or conditions, process the actions or 227interactions of participants, performing a kind of real-time data mining (Serenko and Detlor 228 2002), and operating on semantic metadata (Brusilovsky, 2001). In addition to their use in 22906 education (Serenko and Detlor 2002; Yau et al. 2003), software agents have seen significant 230growth in recent years across multiple sectors including business and e-commerce (Papazoglou 2312001; Jennings 2001), health (Abowd and Mynatt 2004; Cook and Das 2007), air traffic 232

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control (Wooldridge and Jennings 1995), and video games (Stanley et al. 2005). What 233234separates agents from traditional software is that agents are capable of responding to the state of their environment and conducting *flexible* autonomous actions in order to meet their design 235objectives (Jennings & Wooldridge, 1998).

O'Driscoll et al. (2008) state that for educational settings, agents need to be particularly 237aware of the *context* in which the learning takes place, the identities of nearby people and 238objects, the social setting (individual, small group, or whole class settings), the specific activity 239being performed, and that an agent must ideally be able to adapt according to its context, 240including to any changes that might happen to these various factors over time. Within these 241 contexts, agents can capture individual and whole class learning traces, generated artifacts, and 242emergent metadata to provide new insights and supports for student learning (Roschelle et al. 2432013). 244

Software agents thus hold promise for the design of scaffolding environments to support 245inquiry learning, in part because they allow orchestration of scripts that are deliberately *ill* 246determined (i.e., scripts where it is not known, in advance of the enactment, what outcomes or 247conditions will emerge from the products of student interactions). The use of agents allows for 248open-ended designs, enabling the script to evolve in relation to student interactions. For 249example, students might be engaged as a learning community to understand environmental 250conditions in their neighborhoods. Agents could identify two students who independently 251looked up CO₂ sensors and then suggest they share notes or work together to advance their 252understanding. Agents could then dynamically re-group these students with peers they hadn't 253worked with previously to combine their ideas with those of the larger class. The core idea here 254is that agents can respond to a wide spectrum of conditions as they emerge – most of which 255would be operationally impossible for a teacher to do on his own. 256

Supporting the teacher as a facilitator

The goal of smart classrooms and agent driven orchestration should be to engage teachers as 258active co-participants and facilitators of student learning, rather than relegating them to "guides 259on the side" (Pea and Maldonado 2006). The idea of the teacher as a "wandering facilitator" 260has been advanced by Hmelo-Silver's (2000) as a paradigm for supporting learning in student-261driven inquiry designs. In the wandering facilitator model, the facilitator rotates from group to 262group, adjusting the time spent with each of the groups in the classroom according to their 263needs (Hmelo-Silver 2004). 264

However, supporting a teacher as a wandering facilitator is a persistent challenge (Hmelo-265Silver 2004), as it requires the teacher to be aware of each group's state within the flow of 266activities and where she is most critically needed. Adding additional informational cues (Alavi 267et al. 2009) and real-time agents, can reduce the orchestrational load placed on teachers and 268help them make better informed decisions about where they are needed in real-time. For 269instance, Schwarz et al. (2018) showed how the use of machine learning could provide 270teachers alerts when small groups were engaged in *critical moments* during a collaborative 271geometry class. Making these critical moments visible for the teacher provided insight into 272where and when they were needed, helping them orchestrate the class' conceptual learning. 273

Given the rapidly growing spectrum of data that *can* be provided to teachers, we need to 274make sure the information we provide is useful, timely, and actionable. Providing extraneous 275information or "noise" can actually increase orchestrational load and, in fact, become a 276hindrance to effective teacher facilitation. 277

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Research questions

In response to these ideas, with a focus on the design and implementation of a tablet to support 279 teacher classroom orchestration, we investigate two central research questions: (1) How can the teacher tablet leverage real-time agents to support the orchestration of collective inquiry, 281 including context sensitive material assignment, appropriate student grouping, and coordina-282 tion of inquiry activities? And, (2) How can the real-time information provided on the tablet 283 help a teacher orchestrate class activities? 284

In answering these questions, it is important to note that this work focuses on how the teacher tablet supported the orchestration of the overall curriculum, rather than student learning and interactions. With this said, the ability of the environment to successfully enact the kinds of complex pedagogy described below and support the teacher in orchestrating classroom activities, has been described as a grand challenge within the CSCL community. As such, this paper plays an important role in advancing research into this area. 290

In the following section, we outline our design-based research method, in which a twelveweek KCI curricular intervention was developed for two Grade 11 physics classes. We describe a smart classroom framework developed to support our orchestration, including the role real-time agents, and our analytic approach for evaluating our design in term of how those features supported the orchestration of our curriculum. 295

Material and methods

Co-design and design-based research

A design-based research approach (DBRC 2003) was employed for this study, which 298built upon several earlier design cycles (see Tissenbaum and Slotta 2019). Rather than 299validating a particular curriculum, the central goal of design-based research is to 300 advance a set of theories on learning that transcend any particular design or enactment 301(Barab and Squire 2004). To this end, the primary outcome of this research is the 302 design and evaluation of the technological and orchestrational infrastructures them-303 selves (rather than any particular student outcomes), with the aim of understanding 304their role in supporting complex collective inquiry activities. 305

Even when well designed, technology-enhanced learning environments can be quite 306 challenging for a teacher to integrate into her everyday classroom practice (Slotta and Linn 307 2009). Success can be heavily dependent on how well the teacher perceives the "fit" between 308 the intervention and his or her goals for students, teaching strategies, and expectations for 309student learning (Roschelle et al. 2006). As the complexity of the learning design increases 310 (e.g., in a KCI learning community approach, which can entail substantive commitment to 311collective and collaborative inquiry designs), the teacher will be increasingly challenged to 312 integrate all the elements successfully – even if she was an active participant in the curriculum 313 design. We employed a co-design methodology (Penuel, Roschelle & Shechtman, 2006), in 31408 which the teacher was engaged as an active participant in the curriculum and technology 315designs to ensure that our innovations fit within his content expectations and goals for student 316learning. The current design builds on several earlier iterations within the same classroom, 317 which together have addressed the notion of a "smart classroom" infrastructure for supporting 318 collective inquiry (Tissenbaum and Slotta 2019). 319

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Participants

This study involved two grade-eleven physics classes (n = 22, n = 23), in a fee-based, high 321 achieving urban high school in a major metropolitan city. The same teacher (a science teacher 322 with over 10 years of teaching experience) taught both classes and was the co-design partner 323 from earlier smart classroom studies spanning the previous two years (for a detailed 324 description of the iterative technology and room development see Tissenbaum and Slotta 3252019). While this setting may not reflect all the circumstances of public-school settings, it is an 326 appropriate context for our research, which entails complex designs with many different 327 technologies and a high level of autonomous inquiry from students. We do anticipate extend-328 ing these approaches to support a wider range of contexts, which is addressed in our 329discussion. 330

Technology infrastructure: SAIL smart space (S3)

Our designs required a flexible and adaptive technology infrastructure that could support the 332 orchestration of collaborative activities including spatial, social, and semantic dependencies. In 333 response, we developed SAIL Smart Space (S3 - Fig. 2), an open source framework that can 334capture the products of student inquiry (e.g., notes, votes, or tags), the coordination of complex 335



Fig. 2 SAIL Smart Space (S3) systems architecture, showing the use of direct WebSocket messaging to enable communications amongst any element of the environment, a persistent, non-relational (no SQL) database (MongoDB) and software agents

S3 Software Architecture

v 2.0 - Apr 25 2013

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pedagogical sequences, including dynamic sorting and grouping of students, and the delivery 336 of materials from the knowledge base based on emergent semantic connections. 337

An important goal in developing S3 was to allow the physical space of the learning 338 environment to play a meaningful role within the learning design – either through locational 339 mapping of pedagogical elements (e.g., different locations in the room are scripted to focus 340 student interactions of different elements or topics of inquiry) or through orchestrational 341 support (e.g., physical elements of the space, like projected displays, help guide or coordinate 342 student movement, collaboration, or activities). 343

We also added a layer of intelligence to our learning environment through the addition of real-time data mining and computation performed by software agents. Because of the complex nature of our KCI designs and the high demand they place on teachers (as described above), we felt that such agents could play an important role in support of student inquiry and in reduce the teacher's orchestrational load, by automating some tasks and helping the teacher make more informed orchestrational decisions. 346 347 348 348 349

Our design improvements to S3 also included ambient displays that were coupled with the 350 community's emerging knowledge and in-the-moment activities such that they provide a 351 source of peripheral information for students and teachers alike (i.e., about time remaining 352 on tasks, or progress in the knowledge base). 353

S3 comprised a suite of five core technologies: (1) a portal for student accounts and 354software application management; (2) and software agent framework for data mining and 355tracking of interaction in real-time; (3) a central database that houses the designed curriculum 356 and products of student interactions; (4) a visualization layer that controls how materials are 357 presented to students across a range of devices and displays (e.g., tablets, laptops, interactive 358tabletops and large format displays); and (5) a communication framework for connecting 359materials in the knowledge base (e.g., student notes, class polls, or multi-media) and tangible 360 and physical inputs (e.g, through Arduino micro-controllers) in real-time. 361

Real-time software agents

As described above, an important new component of S3 was the development of real-time 363 agents to support the orchestration of inquiry activities that included real-time allocation of 364 materials, assignment to groups, or feedback to the teacher. We included four distinct types of 365 agents, as outlined in Table 1. 366

An important feature of the S3 agents is that they work in concert with each other to create 367 ecologies of orchestration (i.e., nested conditions that feed into each other to allow for 368 interdependent decisions and orchestrational moves). As part of our description of the curricular intervention, we outline several of these ecologies. We follow this with an evaluation of 370 their support for classroom orchestration. 371

Developing an inquiry script – PLACE

In order for the smart classroom activity to be more than just supplemental in nature, we 373 needed to develop a complete curriculum in which the smart classroom was one of several 374 learning contexts, integrated within a broader progression of activities across classroom and 375 home settings. In order to investigate how the smart classroom could leverage studentcontributed content for purposes of authentic inquiry activities, we required a script in which 377 students produced artifacts that would be meaningfully reused in successive activities. We

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| Student Sorting Agent | Sorts students both into groups and around the room |
|------------------------|--|
| | •Sorts can be designed in two ways: |
| | •Pre-set by the instructor or researcher |
| | Emergent based on individual, small group, or whole class actions |
| Consensus | •Monitors groups of students where activities require achieving consensus |
| Agent | Students cannot move to the next step until consensus is achieved |
| | Also used as an orchestration tool to alert teacher to review student consensus when necessary |
| Bucket | •Coordinates the distribution of materials to students in two possible ways: |
| Agent | •Ensure that all members within a group had an equal but unique subset of materials from a given set (i.e. a series or problems or equations) |
| | or |
| | Distributed materials to all members to ensure reduce the variance |
| | between members completing a task (quicker students may receive |
| | more items to work on than slower students) |
| Student Progress Agent | Tracks individual, small group, and whole class progress |
| | Sends updates to other devices (i.e. ambient display, teacher tablet) |
| | Can aid both teacher and students in knowing if students are falling behind the rest of the class |
| | •Coordinates the timing and delivery of materials |

therefore designed a KCI-based physics curriculum in which smart classroom technologies379supported collaborative and collective forms of inquiry for students, and supported critical380reflection and formative interventions for the teacher.381

The teacher shared two main goals for his course: first, to help students recognize "physics 382 in their everyday lives" and bring this view of physics back into traditional classroom settings; 383 second, for students to develop a *coherent understanding* of the underlying principles of the 384course, including the connections amongst those physics principles (i.e., to "see that all the 385 principles are tied together"). These goals aligned with the regional curricular guidelines for 386 grade-11 physics: (1) use the appropriate scientific models to explain and predict the behavior 387 of natural phenomena; (2) analyze and synthesize information for the purpose of identifying 388 problems for inquiry, and solve the problems using a variety of problem solving skills; and (3) 389locate, select, analyze, and integrate information on topics under study, working both inde-390pendently and as part of a team (Ontario Ministry of Education 2008). 391

In response to the second goal, we began by generating set of fourteen "core" principles 392 (Table 2) that the teacher felt were of core relevance to the course. 393

We then co-designed a 12-week curriculum called PLACE (Physics Learning Across 394 Contexts and Environments), which engaged students in capturing examples of physics in 395 the world around them (through pictures, videos, or open narratives), and then using those 396 examples as a source of inquiry – generating problems, applying conceptual tags, and using 397 them as examples. The products of these various inquiry activities became a dynamic 398 "community knowledge base" (one of the central features of KCI) that evolved from one unit 399

| t2.1 Table 2 | Grade 11 | Fundamental | Principles |
|--------------|----------|-------------|------------|
|--------------|----------|-------------|------------|

| 2 Newton's First Law Newton's Second Law Newton's Third Law | Acceleration Uniform Motion Kinetic Friction Static Friction | Fnet = 0 Fnet = Constant (non-zero) Fnet = non-constant Vectors | Kinetic Energy Potential Energy Conservation of Energy |
|---|---|--|--|
|---|---|--|--|

to the next. This knowledge base, called PLACE.web, served as a resource for the culminating400smart classroom activity, in which students applied what they had learned across all three units401to solve ill-structured physics problems relating to scenes from popular Hollywood movies.402The smart classroom served as the technology enhanced environment in which we address the403research questions articulated above (i.e., within the Smart classroom environment).404

Culminating smart classroom activity

As a culminating activity in the PLACE design, we created a challenging task in which students analyzed the physics contained within several popular Hollywood movie clips, in order to test their validity of the scenes. This culminating activity centered around the Smart Classroom, involved three short scripts that spanned home, a traditional class setting, and a smart classroom, and relied heavily on S3 agents to coordinate the distribution of materials, roles, and tasks.

At home activity At home, students were tasked with looking at a collection of the412problems they had been assigned during the preceding 12-weeks (including their own413contributed challenge problems and new problems developed by the teacher), verify-414ing their tagging of relevant physics principles, and adding equations that might be415used to solve the problems.416

Classroom activity In-class, students worked in small groups, using tablet computers to reach 417 consensus on a refined "final set" of the tags and equations for each problem. The goal of this 418 activity was for students to achieve consensus about the principles and equations that had been 419420 assigned to each problem in the corpus. The group was assigned one of the problems, with each student seeing the problem and its various tags on his or her tablet (from the individual at-421 home activity), and asked to agree or disagree. The group was required to reach consensus on 422 all of the principles and equations before they could move to the next problem. Achieving 423consensus is an important task for students, as it provided opportunities for student to clarify 424 concepts and understanding, towards gradually improving their knowledge through sharing 425and discussion (Purba and Hwang 2017). Students could see the work of their group members 426 in real-time, reflected on their own tablets, which helped facilitate face-to-face discussions. 427 The resulting set of problems, tagged with principles and equations, was then stored in the 428 knowledge base for use within the final smart classroom activity. 429

Main activity: In the smart classroom For the third and final stage of the culminating script, 430we developed a set of tools that took advantage of the physical and collaborative affordances 431of the smart classroom, including large projected displays accompanying each station, and 432individual tablet computers to support students as they performed activities. Both classes were 433split into two smaller sections of 11 or 12 students, with each section engaging in the smart 434 classroom activity on a different day (i.e., 4 days in total). Upon entering the smart classroom, 435students were engaged in solving a series of ill-structured physics problems using Hollywood 436movie clips as the domain for their investigations (e.g., could Iron Man Survive a fall to earth, 437 as depicted in the movie?). Four videos were presented to the students, each at a distinct 438physical location within the room. The students were engaged collectively, working as a whole 439group of 10-12, as well as collaboratively, in various small group configurations as directed by 440 the S3 real-time agents. 441

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The smart room script was broken up into four different steps as shown in Fig. 3: (1) Principle442Tagging; (2) Principle Negotiation and Problem Assignment; (3) Equation Assignment, and443Assumption and Variable Development; and (4) Solving and Recording. In each step, students444moved from one video to another, completing a set of collective and collaborative tasks that built445upon the emerging knowledge base, using tablets and large format interactive displays.446



Fig. 3 The smart classroom "Hollywood Physics" script involved four distinct steps. The dark blue boxes indicate actions mediated by real-time software agents. The red box indicates the point in the script where the real-time software agents alerted the teacher to review individual groups' work for approval or to have them go back and refine their thinking

The large-format interactive displays aggregated the products of individual student work 447 from their individual tablet inputs and helped facilitate group discussion (Fig. 4). S3 software 448 agents provided students with context specific tasks and materials, facilitated the dynamic 449grouping of students, and ensured student consensus on final products was reached on all 450collaborative tasks. We also developed a set of ambient displays that showed real-time 451information on the state of class activities and an orchestration tablet that provided the teacher 452with additional procedural information and control over the progression of class activities. 453Below, we outline our rationale for each of these technologies with a specific focus on their 454roles in supporting real-time teacher facilitation and orchestration. 455

> 456 457

Designs of teacher orchestration supports

In order to support the teacher as a wandering facilitator and to know where and when he was 458most needed, we developed a specialized teacher orchestration tablet (Fig. 5). The orchestra-459tion tablet, iterating on observations and feedback from previous designs (Tissenbaum and 460Slotta 2019), moved from a device showing student work post hoc (which the teacher was 461 unable to act upon in real-time), to an orchestrational tool that allowed the teacher to more 462directly orchestrate the flow of activities. The goal of the orchestration tablet was to give the 463 teacher control of class progression at both whole class and small group levels, and to inform 464him when he was needed at key moments in the script. The tablet showed him which tasks 465each group had completed (in contrast to the information at the grain size of the individual, 466 available on the large ambient display), alerted him when he needed to review a group's work, 467 and allowed him to easily progress the whole class to the next step in the activity (pressing a 468 button on the tablet would send a signal to the S3 system, which then managed the student 469groupings, location assignments, and material distribution). 470



Fig. 4 The smart classroom setting with (1) An interactive collaborative display that orients students towards a specific Hollywood scenario, aggregates student contributions specific to that video and facilitates idea negotiation; (2) A second board with a different scenario facilitates similar but thematically distinct student interactions (two other boards are similarly placed on the opposite wall; (3) Individual tablets provide students task instructions, allow them to access the knowledge base, and contribute ideas to the shared display; and (4) An ambient display that shows where students are in the room, their completed tasks and the time left in the activity

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| l | NEOplace - | Hollywood F | hysics |
|----------------|--|--|--|
| | Ste Students tag equations & wri | p 3 of 4 te their va | riables and assumptions |
| | First, tap the 'Re-sort' button to a | ssign the st | idents to new video boards. |
| | | | 1920 - |
| | Only students have signed in, | tap 'Start' to | begin equation tagging activity. |
| | | 0 | |
| | | Start | |
| When Vhen s | students are done tagging their equation students are done writing their assumptio button chang You will then need go to that video boar | is, you will s ns and varia e colour and d to review t | ee the appropriate button's colour change ibles, you will see the appropriate approval light up. he ir work b efore tapping 'Approve'. |
| When When s | students are done tagging their equation students are done writing their assumptio button chang You will then need go to that video boar BOARD A | is, you will s ns and varia e colour and d to review t | ee the appropriate button's colour change ables, you will see the appropriate approval light up. heir work before tapping 'Approve'. BOARD B |
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| When s | students are done tagging their equation students are done writing their assumption button change You will then need go to that video board BOARD A Done tagging equations Approve assumptions & variables | is, you will s ins and variis colour and d to review t | ee the appropriate button's colour change ables, you will see the appropriate approval ight up. BOARD B Done tagging equations Approve assumptions & varial |
| When s | students are done tagging their equation students are done writing their assumptio but ton chang You will then need go to that video board BOARD A Done tagging equations Approve assumptions & variables BOARD C | is, you will s ins and varie e colour and d to review t | ee the appropriate button's colour change ables, you will see the appropriate approval ight up. beir work before tapping 'Approve'. BOARD B Done tagging equations Approve assumptions & varial |
| When s | students are done tagging their equation students are done writing their assumption buttom change You will then need go to that video board BOARD A Done tagging equations hoprove assumptions & variables BOARD C Done tagging equations | s, you will s e colour and d to review t | ee the appropriate button's colour change ables, you will see the appropriate approval ight up. beir work before tapping 'Approve'. BOARD B Done tagging equations Approve assumptions & varia BOARD D Done tagging equations |
| When s | students are done tagging their equation students are done writing their assumption button change You will then need go to that video board BOARD A Done tagging equations Opprove assumptions & variables BOARD C Done tagging equations Approve assumptions & variables | is, you will s ins and varie colour and d to review t | ee the appropriate button's colour change ables, you will see the appropriate approval. BOARD B Done tagging equations Approve assumptions & varial BOARD D Done tagging equations |
| When s | students are done tagging their equation button change You will then need go to that video board BOARD A Done tagging equations Approve assumptions & variables BOARD C Done tagging equations Approve assumptions & variables | is, you will s ins and varies colour and d to review t | ee the appropriate button's colour change ables, you will see the appropriate approval ight up. BOARD B Done tagging equations Approve assumptions & varial BOARD D Done tagging equations Approve assumptions & variables |
| When s | students are done tagging their equation buttom change You will then need go to that video board BOARD A Done tagging equations Approve assumptions & variables BOARD C Done tagging equations Approve assumptions & variables Tap 'Goto Step 4' only after you have | is, you will s ins and varies colour and d to review t | ee the appropriate button's colour change ables, you will see the appropriate approval ight up. .teir work before tapping 'Approve'. BOARD B Done tagging equations Approve assumptions & variat Done tagging equations Approve assumptions & variables all the students assumptions. |

Fig. 5 The Teacher Orchestration Tablet. The tablet (1) Enabled the teacher to start a stage for the whole class; (2) Showed each group's progression through the activity; (3) Alerted the teacher when a group reached a point for intervention (pre-defined by the teacher); and (4) Let the teacher advance the class to the next Step

Design of software agents for curriculum orchestration

In order to address our first research question – how the orchestration tablet could leverage 472 real-time agents to support the teacher's real-time orchestration – we designed specific tasks 473 for the real-time software agents to enact during the activity. Below, we outline these tasks 474 along with the specific agents we developed (described in Table 1 above). 475

Sorting students based on emergent classroom conditions

Grouping and re-grouping students is a persistent challenge in live classroom settings. This 477 challenge becomes compounded when the conditions for the sorting must emerge during the 478enactment of the activity itself (and therefore cannot be known a priori). During the smart 479classroom activity, we wanted students to be sorted based on a set of predefined conditions set 480 in co-design with the teacher: 1) After Step 1, students were sorted based on the frequency of 481 their tags at each of the scenarios in the room, as we felt this might show a particular affinity 482 towards that topic by the student; and 2) After Step 2, as we wanted students to work with 483 students they hadn't worked with in the previous step. Since we didn't know which scenarios 484

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each student would tag with which principles beforehand these sorts could only happen in real-485 time (once the teacher pressed "sort" on his tablet). Once the teacher pressed "sort" on his 486tablet, the large ambient display at the front of the class showed the students where each of 487 them was to go in the room. In this case only one agent was used: 488

Sorting Agent: The sorting agent created tables of students' activities (i.e., tags assigned to 489 each scenario, and who each student had worked with previously), and based on these tables, 490placed students around the room. 491

Supporting the teacher in just-in-time orchestration

Helping a teacher function as a wandering facilitator within a complex real-time activity takes 493more than simply making student work visible to them. In order for the teacher to truly react to 494 the just-in-time needs of the students in the class we needed to enable him to know both when 495and where he was needed. To understand how the S3 architecture supported the teacher's 496 ability to respond, we had two agents work in concert to alert the teacher when a group reached 497 the end of Step 3 (see Fig. 3 above): 498

Consensus agent Ensured that the group had sorted all of the variables and assumptions 499submitted to the negotiation area on the collaborative display. Once completed the consensus 500agent sent a message to the Progress Agent. 501

Progress agent Once the progress agent received a notification that a group had reached 502consensus on their variables and assumptions, it then sent a message to the teacher's orches-503tration tablet, alerting his to review student work - either approving it (allowing them to 504progress) or have them go back and work on it some more. By using the underlying agent 505infrastructure and messaging protocol in S3, the orchestrational load placed on the teacher to 506know (at least on some level) when and where he was needed was reduced by his awareness 507that he would be alerted on his tablet. This allowed the teacher to more freely roam the room 508engaging with students based on group needs. 509

Measures and analytic approach

While it is important to situate the culminating activity within the context of the larger 512curriculum (in order to show its significance as more than a stand-alone activity), the analysis 513below will not evaluate the parts of the curriculum that preceded it (for analysis on the 514preceding activities see Tissenbaum and Slotta 2015). Rather, we restrict our analysis to the 515enactment and orchestration of the culminating Smart Classroom activity. 516

As stated above, the main goal of this research was to evaluate the efficacy of S3 teacher 517tablet and agents to support the teacher's classroom orchestration (rather than evaluating the 518particulars of student learning). To this end, our measures and analytic approaches focus 519primarily on evaluating the design in terms of its ability to support the enactment of the 520designed curriculum, and the role agents played in this enactment. 521

In order to evaluate and understand the enacted design, we used a mixed-method approach 522that included multiple data sources to triangulate data and gain a more complete picture of the 523study (Greene 2006; Mason 2006, Johnson et al., 2007). The use of multiple data sources is 524Q10 particularly relevant in design-based approaches, due to the complexity and innovative nature 525

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of their design and enactment (DBRC 2003). Data sources included pre- and post-interviews526with teachers and students, sever logs, user contributed artifacts, and audio and video527recordings. All user data, generated artifacts, and interaction with the system was collected528using S3's data collection infrastructure. Below, we use this data to illustrate key examples of529the role the teacher tablet and the real-time agents played in supporting classroom530orchestration.531

During the culminating activity, six video cameras were used (one at each of four "zones" 532 in the room; one camera with a fixed view of the whole room; and one wandering video 533 camera) to capture student and teacher interactions. To capture student and teacher discourse, 534 voice recorders were place at each of the four zones and the teacher had his own lanyard 535 microphone. Video and sound recordings were synchronized and analyzed using Ingscribe, a 536 popular coding software platform. 537

Results and discussion

Sorting students based on emergent classroom conditions

We wanted to understand the efficacy of the agents for sorting student based on emergent class 540patterns. To this end, we examined the server logs to see how the agents sorted the students 541once the teacher pressed "sort" on the orchestration tablet. In all four class sections, after the 542teacher pressed "sort", the Sorting Agent successfully sorted students based on their earlier 543actions in the room. Table 3 (initially reported in Tissenbaum and Slotta 2015) provides data 544from one section's sorting. The agents used a cascading approach to assign one student to 545Board A based on their frequency of principles, then one to Boards B, C, and D in order, 546before repeating this process until all student were sorted. For instance, Jason was assigned to 547board B and not A, C, or D because the agent had already placed Alice at Board A, and Jason 548had the most tags when the agent went looking for a Board B student (i.e., for the second 549assignment by the agent's algorithm). 550

In our design, the agents used a simple table-based system to decide how to sort students, 551and we recognize that other approaches could allow for more complex approaches in making 552such real-time grouping decisions. However, this method was sufficient to demonstrate that the 553underlying agents were able to track these conditions in real time and make the appropriate 554decisions as laid out by the teacher. Video analysis of the student sorting noted that the average 555time from when the teacher pressed "sort" on his tablet to when students were in their new 556groups, ready to start the next task, was under 20 s. This is noteworthy, compared to what 557might be achieved in a low-tech classroom setting, where re-grouping based on evidence from 558a previous activity would take time for the teacher to compute, followed by more time to 559convey the grouping to students and get them to move around the room. In our case, the 560automated tracking and assigning of students within groups allowed the teacher to focus on 561helping the students and not the logistical aspects of the group sorting. This point was 562reinforced by the teacher's comment in the exit interview: 563

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Each [agent sort] was a different ensemble, using physics pedagogy and other schemes565to figure out where kids should go. During transitions when you're a teacher getting kids567up, moving them to different seats – you waste so much class time doing that. Even a568

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| # of Tag | gs by studer | ıt | | | | | | |
|----------|--------------|---------|---------|---------|------------------------------|-------------------------------|----------------------|--------------------------------|
| Student | Board A | Board B | Board C | Board D | First sort: sent to board | Second sort: Sent to board | Sorted to new board? | Sorted with new members? |
| Alice | 4 | 3 | 3 | 4 | A | В | Y | Y |
| Pearl | 3 | 0 | 3 | 2 | А | С | Y | Y |
| Jason | 4 | 3 | 4 | 4 | В | С | Y | Y |
| Rob | 0 | 3 | 3 | 1 | В | D | Y | Υ |
| Desi | 3 | 2 | 3 | 0 | С | D | Y | Y |
| Raffi | 0 | 2 | 2 | 2 | С | А | Y | Y |
| Becky | 2 | 2 | 3 | 3 | D | А | Y | Y |
| Sun | 2 | 2 | 0 | 2 | D | В | Y | Y |

common group, cooperative learning scenario, like a games theory thing, where kids are569really learning from each other, just getting the kids to move around the classroom570adequately for that, I find cumbersome – I just kind of dread moving the kids around the571class and organizing that, rather than doing the activities themselves, and so I just loved572the logistical assistance that [the S3 agents] offered.573

Supporting the teacher in just-in-time orchestration

Across all four class sections of the culminating activity, the S3 agent and messaging 576 framework successfully notified the teacher whenever a group required the teacher's review 577 and approval. For each individual group, the *Consensus Agent* was able to pick up when they 578 completed their assumption and variable negotiation (at the end of Step 3), and sent a 579 notification over S3's messaging service. The *Progress Agent* was able to interpret the event 580 as one that required teacher response, and sent the appropriate message to his tablet (Fig. 6). 581

In total, the teacher was sent 23 alerts to review students work (Table 4). It is worth noting that across the four sections there were only sixteen groups (four in each section). The reason for the seven extra alerts was that the teacher asked six groups to refine their thinking more and resubmit it for review (with one group being asked to re-submit twice). This is important in several ways. The first is that it shows the flexibility of the agents to respond to multiple similar events with the same group, which allowed for a more flexible (rather than a strict linear) progression through the seven the seven the seven through the seven as the seven extra alerts was that the teacher asked six groups to refine their thinking more and resubmit it several ways. The seven extra alerts was that the teacher asked to re-submit twice) and the seven extra alerts with the same seven extra alerts was that the teacher asked to re-submit twice and the seven extra alerts was that the teacher asked to re-submit twice and the seven extra alerts was that the teacher asked to re-submit twice and the seven extra alerts was that the teacher asked to re-submit twice and the seven as the sev



Fig. 6 An example of the event messages handled and sent by the real-time software agents from a group's collective display to the teacher tablet. A *Consensus Agent* would monitor the group's work and wouldn't allow them to submit their work until all the items were sorted. Once the items were sorted and the group pressed submit, a *Progress Agent* would pick up the message and send an alert to the teacher on their tablet

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t4.1 **Table 4** Agent-orchestrated alerts for review of student work sent to the teacher's orchestration tablet. Across all four sections the teacher was successfully alerted and reviewed every groups' negotiated set of assumptions and variables

| | Zone A | Zone B | Zone C | Zone D |
|-------------------------|--|--|--|--|
| Section 1 Review Alerts | 1 | 2 | 2 | 1 |
| Section 2 Review Alerts | 1 | 1 | 1 | 1 |
| Section 3 Review Alerts | 1 | 2 | 1 | 2 |
| Section 4 Review Alerts | 3 | 1 | 1 | 2 |
| | Section 1 Review Alerts Section 2 Review Alerts Section 3 Review Alerts Section 4 Review Alerts | Zone ASection 1 Review Alerts1Section 2 Review Alerts1Section 3 Review Alerts1Section 4 Review Alerts3 | Zone AZone BSection 1 Review Alerts12Section 2 Review Alerts11Section 3 Review Alerts12Section 4 Review Alerts31 | Zone AZone BZone CSection 1 Review Alerts122Section 2 Review Alerts111Section 3 Review Alerts121Section 4 Review Alerts311 |

activity. In addition to the orchestrational flexibility these alerts provided the teacher, this approach 588 shows how an orchestration tablet (powered by the software agents) may allow teachers to offload 589 the need to constantly monitor the state of student work, instead focusing his attention where they are most needed (as prompted by the tablet). 590

Understanding the effect of the teacher's just-in-time orchestration

To understand the effect of the teacher alerts and teacher follow-ups on student outcomes, we 593 evaluated the student generated products from that stage of the activity. First, to evaluate the 594 quality of each groups' final constructed set of assumption and variables across all four 595 sections, the teacher (post hoc) scored them using a four-point scale that rated them based on their completeness for setting up the problem (Table 5). Across all four sections, groups 597 averaged 2.6 (out of maximum 3) with no group scoring below a 2, indicating an overall high 598 quality of problem setup. 599

Next, to understand the effect of the teacher's reviewing and approving of groups' work on 600 their final completeness score, we rescored the original assumptions and variables of the 601 groups the teacher had asked to resubmit (i.e., before their edits). Figure 7 shows changes in 602groups' completion scores. A paired *t*-test showed that increases in completion scores were 603 significant (p < .05) when comparing scores prior to the teacher's intervention (M = 2.17, 604 SD = 0.41) and after (M = 2.83, SD = 0.41; t = 3.1623, p = 0.025). While the sample size is 605 small (n=6), de Winter (2013) has shown that small-sample *t*-tests are acceptable when 606 assessing changes in student outcomes. 607

When we examined video of the teacher's interactions with the groups, we found that the 608 teacher largely focused on "teasing out" how the groups came up with their variables and 609 assumptions. For instance, during Session 3, the following exchange shows an interaction in 610 which the teacher asked the group in Zone B to further refine their variables and assumption: 611

| Score | Level | Description |
|-------|-------------------------------------|--|
| 0 | No correct assumptions or variables | The group failed to provide any assumptions or variables that could be used to solve the video |
| 1 | One assumption or Variable | The group were able to successfully identify at least one variable or assumption that they needed to solve the video |
| 2 | Partially Complete set | The group was able to assign several assumptions and variables to the video but did not identify all of them |
| 3 | Complete Set | The group successfully provided all of the necessary assumptions and variables needed to solve the video |

t5.1 **Table 5** Rubric for scoring group assumption and variable construction during Step 3 of the culminating activity

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Fig. 7 Variable and Assumption scores for groups before and after the teacher requested the group go over their negotiated set again

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| Teacher: How did you come up with the initial height? | 613 |
|---|-----|
| Student 1: I didn't come up with that, I mean | 615 |
| Teacher: You made it up? | 618 |
| Students: [laugh] | 629 |
| | 021 |

The teacher then had the group think more deeply about their decision-making process and 622 justifications. When the students asked the teacher to review their work again, he examined 623 their assumptions and variables, and in response to one of their variables, suggested they "have 624the tank shoot at 90 degrees every time, it isn't really, but it's close". This overall exchange 625 shows that the teacher encouraged students to work out their reasoning themselves, and, in the 626 second case, a slight refinement of their thinking, rather than giving them the answer outright. 627 Combined with the average increase in the group's final completion scores, this seems to 628 indicate that the teacher's orchestration was effective in helping students think deeper, rather 629 than giving them the right answer. 630

It is worth noting that of Day 1, Zone B, the score was already 3/3 and no additional 631 elements were added which may indicate that the teacher simply asked them to think about it 632 some more, but they did not have to make any changes. On Day 3, Zone B, the group did add 633 another element that was considered significant by the teacher, but they still missed one 634 preventing them from achieving a perfect score. Taken as a whole, the significant changes 635 in groups' completeness scores highlights that the teacher knowing when and where they are 636 needed can have a significant impact on students' knowledge construction and provide 637 important orchestrational support at key moments in students' learning. 638

Conclusions

This study introduces a new approach to supporting the orchestration of real-time inquiry 640 activities, in which the design of the physical space and the accompanying technologies are 641

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carefully considered in parallel with the curricular content. In particular, this study provides 642 evidence for the important role that a real-time teacher tablet, supported by data mining and 643 software agents, can play in reducing the teacher's orchestrational load and supporting him as a 644 wandering facilitator. 645

The teacher's real-time orchestration tablet became the conduit through which much of the 646 orchestration flowed. By providing detailed information about where students were in the 647 activity at multiple granularities (i.e., small groups and the whole class), the teacher was able to 648 make critical decisions on where he was needed and when the class was ready to progress to 649 the next step in the activity. In particular, the alert that let the teacher know when he needed 650 and the resulting increase in student completion scores, highlights the utility of such a tool to 651support productive student outcomes. Recent CSCL research has shown how timely teacher 652support can be critical in student problem solving and idea negotiation. The work of Ingulfsen 653 et al. (2018) and Furberg (2016) showed that students often struggle to make connections 654between relevant data, and require timely teacher intervention – such as, conceptual support 655 and probing for elaboration – in order to make successful progress. However, without supports 656 to make these moments visible, teachers may miss critical moments, and students may need to 657 compete with peers to get the teacher's attention (Alavi et al. 2009). As such, the design and 658 development of these feedback and visualization tools requires careful consideration. De-659 signers need to understand exactly what teacher needs to see to make better informed 660 decisions, and what elements can be effectively "hidden" to run autonomously. 661

The ability of this study's orchestration tablet to effectively sort students into groups and 662 place them around the room based on emergent conditions, is an example of how information 663 can be hidden while still supporting classroom orchestration. The teacher did not need to know 664which students were going to be placed at which spot in the room when he pressed the "sort" 665 button on his tablet (Table 3) – it was enough for him to know that it would be done. 666 Removing this load from the teacher allowed him to focus on the students rather than these 667 managerial tasks. Perhaps the most encouraging feedback on the efficacy these orchestration 668 supports was the teachers comments on the ability of the tablet and agents to reduce his 669 orchestrational load: 670

It was such a sort of shifting paradigm kind of lesson, with the pacing and, I don't know, just the kinetics and the motion in the room and kids moving around was a lot to follow, [but] I didn't need to worry about it, it was just taken care of by the various technologies.

Students also noticed the efficacy of S3 in freeing the teacher from many of the managerial677tasks in the class, noting that they did not "need the teacher for that any more... he could just678focus more on going around and talking to the groups" (student, Jen).679

As classroom interventions become increasingly infused with digital technologies to 680 support collaboration and knowledge construction, the real-time state of a student within the 681 class (i.e., their knowledge, interests, or where they are within a particular activity) is 682 increasingly hidden "behind a screen" (Sharples 2013). However, the ability to track the 683 complex connections between students and their peers, the emergent knowledge, and the 684 teacher's goals for learning, offers new support for orchestration that previously would have 685 been too difficult to process manually, especially in real-time. The introduction of real-time 686 software agents can help process this stream of data and connect it to desired learning patterns 687 and teacher needs. Well-designed agents allow researchers, learning designers, and teachers to 688 establish a priori the events they wish the learning environment to respond to, without the 689

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explicit need to know *who will fill those conditions* or *when they will fulfill them* prior to the 690 activity's enactment. When done well, these technologies can reduce the teacher's orchestrational load, freeing him or her to do the important tasks of working with students and helping 692 them overcome challenges and refining their thinking. 693

While this study points to the potential efficacy of an agent-supported tablet system to 694support classroom orchestration, it admittedly does so in an activity with a fairly linear script. 695 This raises questions about how similar approaches would work in more open-ended scenarios 696 where the script is less-structured (Dillenbourg et al. 2009). Similar to Dillenbourg's (2009) 697 concerns of over-scripting, these kinds of learning spaces may end up over-orchestrating the 698 activity, with students and teachers feeling that the activity is "on rails". The challenge of 699 balancing flexible orchestration while providing the correct level of guidance and regulation is 700 not new (Dillenbourg et al. 2009; Kirschner et al. 2004). As shown by the teacher being able to 701 ask students to revisit work before progressing, we attempted to find this balance. During the 702 exit interview, the teacher noted that it would have been nice for the students to be able to 703 revisit their work if they realized they needed more data from a previous step - a level of 704 flexibility not afforded by this particular classroom script. Based this feedback and our own 705observations, we have since developed scripts that engage in shorter cycles of discussion and 706 problem solving that allow the class to engage in discussion about next steps and revisit and 707 refine their thinking before going through another cycle (Moher et al. 2015). 708

A similar challenge concerns how to design orchestration systems that are flexible and 709 robust enough to still function if or when the agents or the system make mistakes (which is 710likely to happen in any system over time!). Similar to the issues with availability, partition 711 tolerance, and consistency in distributed systems (Kleppmann, 2015), designers of real-time 712 orchestration dashboards will need to consider what happens when issues occur such as 713 dropped data, missed messages, or devices temporarily disconnecting from the system. In 714 the design discussed in this paper, certain orchestration functions could still be conducted by 715the teacher if the system failed. The teacher could still act as a wandering facilitator going to 716 groups and examining their work, even if he did not receive an alert. However, this would 717 require students to spend time trying to get his attention rather than working (an orchestration 718 challenge similar to Alavi et al. 2009). The teacher could also advance groups to the next step, 719even if his tablet indicated that not all students were done. On the other hand, problems could 720 arise if students were not put into groups or the content from the database was not properly sent 721 to students' tablets. Designers will need to carefully consider what effect a failure would have 722 on the overall ability of the system to function. 723

Another possible limitation to this study is that the teacher was well versed in the 724 pedagogical approach, having worked with the research team for several years as a co-725 designer. Getting teachers acquainted with novel technological and pedagogical approaches 726 is a persistent challenge in CSCL research (e.g., Koh and Hong 2017). However, our goal was 727 to test the capabilities and feasibility of our design, rather than aim for broad applicability. As 728such, working with an experienced co-design teacher allowed us to focus on the design and 729implementation. As part of the co-design team, the teacher was well acquainted with how the 730 script was expected to unfold. However, prior to running the class activity, he had not seen the 731 tablet in action. As such, he was responding to the tablet for the first time live. His ability to 732 successfully use the tablet to help his orchestrate classroom activities, points to the efficacy of 733 the tablet's design. Part of this stems from the intentional simplicity of its design. Rather than 734providing the teacher with everything we could from the live data, we only provided him 735 things that were determined to be immediately and timely actionable (e.g., forming groups, 736

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checking student work). This responded to earlier challenges we had with previous versions of 737 a real-time teacher tablet (see Tissenbaum and Slotta 2019). 738

One final limitation of this work is that it only ran for one session for each class. 739 Developing an orchestrational framework that leverages data over longer scales of time and 740a diversity of activities (compared to the four steps and two sorts in this activity) would likely 741 require more complex tracking of student interactions and trajectories. For example, some 742 intelligent tutoring systems have seen early success with more complex and longitudinal data 743 analytics (Rubio-Fernandez et al., 2019). However, we believe there is a place for both of these 744 approaches to coexist. In many cases, there is limited or non-complex data available for 745sorting. In the case of the study we present here, the teacher wanted a very specific kind of 746 sort - having the students work with peers they had not worked with before in a modified 747 jigsaw. This would not require the level of complexity, and increased variability, of approaches 748 like k-means clustering. Educational designers need to deeply understand and consider these 749kinds of trade-offs when designing real-time agent-based orchestration tools. 750

It is worth noting here that smart classroom setups such as these are generally rare. They 751require a significant commitment to a physical-technical architecture that is at odds with many 752traditional classroom configurations. However, many, if not all, of these approaches can be 753achieved in similar lower-cost ways. Active Learning Classrooms (Dori and Belcher 2005) 754approach classroom design with similar clusters of students working around large, often 755interactive, shared displays (Charles and Whittaker 2015). While these have been primarily 756 situated in post-secondary classrooms, we are seeing growing adoption of them in K-12 757 settings (Hod et al. 2016). We are also seeing carts of tablets becoming more common 758throughout K-12 schools, opening up the opportunity for increased mobility of both teachers 759and students. What is important from this work is less the particular technologies used, but the 760kinds of learning, collaboration, and orchestration it supports. Just as the early work with Palm 761 Pilots (Roschelle and Pea 2002) and multi-user computer screens (Szewkis et al. 2011), 762provided important evidence for future research and classroom implementations, our work 763 work aims to provide a set of generalizable exemplars grounded within the learning sciences 764for the future research of others. 765

766 Understanding the potentially powerful role that agents can play in reducing the timeconsuming tasks of sorting students into groups, and providing them timely and context-767 sensitive materials is something that we believe can have a lasting impact on classrooms 768 broadly. By taking these administrative tasks out of teachers' hands and automating them, we 769 can free the teacher up to spend more time with students, providing more time for classroom 770 learning and collaboration (instead of waiting around for the teacher to make groups and 771 distribute materials manually). In our own design, the teacher noted that the lesson seemed to 772 "gain time" as it progressed, allowing more learning and collaboration to be packed into the 773 class period than he normally expected (Tissenbaum and Slotta 2019). Another key element of 774 this design that we feel can be generalized to other contexts, is to understand what information 775 can help teachers make real-time decisions quickly, and what information might simply 776 increase the teacher's orchestrational load and would be better left for post hoc reflection. In 777 our design, there was a lot of processing going on "under the hood", and yet, we kept the 778 design simple - a limited set of alerts letting the teacher know where students were in the 779activity and where and when he was needed. This allowed the teacher to keep a heads-up view 780of the class and did not require him to make complicated assessments of the whole class' 781 learning. This complements the work by Schwarz et al. (2018), which showed how a clear and 782uncluttered real-time display of small group work can help a teacher intervene at *critical* 783 momentsin students' problem solving. Their simplified alerts (changing the color of the
boarders around each group's work to indicate a specific state), allowed the teacher assess784where they most needed in the moment with minimal additional orchestrational load. While
this may be valuable in some cases, designers need to carefully assess the load this places on
teachers and the resulting trade-offs.784785787

Our design study showed the potential for a teacher tablet that leverages the emergent real-789time data in a classroom to help offload much of the monitoring and management tasks to the 790underlying system. We feel there is considerable potential for technology approaches that free 791 teachers to focus more on the students and act as informed wandering facilitators. The 792 underlying S3 agent architecture played a key role in our work, monitoring student interactions 793 at the individual, small group, and whole class levels. This collection of loosely coupled 794software agents provides a pedagogically driven blueprint that others can follow within their 795 own CSCL designs. Rather than developing large monolithic monitoring tools, more flexible 796 agents such as the ones in this study, offer the potential for designs that approach orchestration 797 as an ecology, in which agents can work in concert or individually, responding to emergent 798classroom patterns. As mentioned above, as new tools are developed to harness the huge 799 amounts of data generated in CSCL environments, researchers will need to make decisions 800 concerning their orchestrational flexibility. Key questions moving forward will include under-801 standing what is gained, and critically, what is lost, when we automate some class activities, 802 thus reducing the ability of the teacher to orchestrate elements the class on their own. Similarly, 803 we will need to deeply consider how much is too much data. This work aimed to find a 804 reasonable balance between giving the teacher a lens into the class, while hiding other 805 potentially distracting information away. Moving forward, designers will need to carefully 806 consider how information provided to teachers will be actionable, and more importantly, what 807 the learning outcomes of these teacher interventions will be. 808

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