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"I think you just got mixed up": confident peer tutors hedge to support partners' face needs

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Abstract During collaborative learning, computer-supported or otherwise, students balance 10 task-oriented goals with the interpersonal goals of relationship-building. In peer tutoring, some 11 pedagogically beneficial behaviors may be avoided by peer tutors due to their likelihood to get 12in the way of relationship-building. In this paper, we explore how the interpersonal closeness 13 between students in a peer tutoring dyad and the peer tutors' instructional self-efficacy impacts 14 those tutors' delivery style of various tutoring moves, and explore the impact those tutoring 15move delivery styles have on their partners' learning. We found that tutors with lower social 16closeness with their tutees provide more positive feedback to their tutee and use more indirect 17instructions and comprehension-monitoring, but only for tutors with greater tutoring self-18 efficacy. And in fact, those tutees solved more problems and learned more when their tutors 19 hedged instructions and comprehension-monitoring, respectively. We found no effect of 20hedging for dyads with greater social closeness, on the other hand, suggesting that interper-21sonal closeness may reduce the face-threat of direct instructions and comprehension-monitor-22ing, and hence reduce the need for indirectness, while tutors' instructional self-efficacy allows 23tutors to use those moves without feeling threatened themselves. These results emphasize that 24designers of CSCL systems should understand the nature of how the interpersonal closeness 25between collaborating students intersects with those students' self-efficacy to impact the use 26and delivery of their learning behaviors, in order to best support them in collaborating 27effectively. 28

Keywords Peer tutoring \cdot Hedging \cdot Rapport \cdot Self-efficacy \cdot Social bonds \cdot Indirectness 29

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Introduction

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In collaborative learning interactions, whether computer-mediated or face-to-face, students 32balance task-oriented goals with the interpersonal goals of relationship-building (Tract and 33 03 Coupland, 1990). In some forms of collaborative learning, such as peer tutoring, students may 34 offer each other advice, instructions, or feedback. Such pedagogical behaviors, while 35supporting the task goal of helping their partner learn, may also conflict with the interpersonal 36 goal of relationship-building. That conflict may arise from the potential for such behaviors to 37 be pedagogically helpful, while at the same time, potentially threatening for their partner's 38 "positive face", or desire to be approved of by others (Brown and Levinson 1987). 39

To mitigate the relational consequences of pedagogical behaviors that are likely to threaten 40tutees' interpersonal needs, such as feedback and comprehension-monitoring, peer tutors 41 without sufficient interpersonal closeness with their tutee might avoid providing the necessary 42tutoring move altogether (Person et al. 1995). If, however, they are more skilled at attending to 43 interpersonal needs, they might phrase their words in an indirect, or "hedged", manner, 44 reducing the implicit threat to their partners' "face". Some computer supports for learning, 45such as some forms of intelligent tutoring systems (ITS), indiscriminately apply this indirect 46 style to the feedback and instructions provided to students, to reduce the threat of feedback and 47instructional directives (Johnson and Rizzo 2004). An overuse of such indirect instructional 48 moves, however, may have a negative impact on student learning, due to the inherent 49ambiguity of indirectness (Person et al. 1995). 50

Therefore, if computer-supported collaborative learning (CSCL) systems were to simply 51prompt all collaborating students to always use indirectness, as in the ITS example, such a 52recommendation may not be the most effective or socially-appropriate way for peer tutors to 53deliver feedback or instructions. As Carmien et al. (2007) argue, students bring their own 54internal scripts to bear in collaborative learning interactions, which may conflict with the 55scripts provided by a CSCL system. In order to design CSCL systems that can support 56students' productive collaborative discourse (as in Tegos et al. 2016), we must first understand 57whether and how students' interpersonal closeness impacts with the resources they bring to 58bear (here, tutoring self-efficacy and prior knowledge) to impact their use and delivery style of 59various tutoring strategies. Will effective tutoring moves be avoided due to concerns about 60 their potential face-threat? Will peer tutors modify the delivery style of such moves to mitigate 61that potential face-threat? 62

In this paper, we first investigate how peer tutors' interpersonal closeness with their tutees 63 impacts their use and delivery style of potentially face-threatening tutoring moves like (1) 64instructional directives, (2) feedback, and (3) explicit reflections on their partners' compre-65 hension. We include in this analysis two potentially mediating factors: tutors' prior domain 66 knowledge and their tutoring self-efficacy (the belief that one is a capable tutor), We then 67 investigate the relationship between peer tutors' delivery style, their domain knowledge and 68 tutoring self-efficacy, and their interpersonal closeness with their tutee, on tutees' problem-69 solving and learning. 70

Results support the importance of a process-oriented approach to understanding the 71 delivery style of tutoring behaviors, and the importance of bringing factors other than simply 72 friendship to bear in studying the impact of social influences on learning. We find that while 73 peer tutors with a less strong relationship with their tutee can support their partners' learning 74 behaviors, problem-solving, and learning gains by hedging some of the more face-threatening 75 tutoring moves, not all peer tutors are equally as likely to hedge such moves. We find that peer 76

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tutors with greater self-efficacy for their ability to tutor are in fact *more* likely to hedge their 77 face-threatening tutoring moves, suggesting that a greater tutoring self-efficacy might allow 78 peer tutors to hedge, potentially appearing to be uncertain, in order to save their partners' face 79when needed. These findings can help inform the design of CSCL systems that might detect 80 students' interpersonal closeness or tutoring self-efficacy and suggest different ways for 81 students to deliver instructions, feedback, and comprehension-monitoring to mitigate face-82 threat, or lead to a system with virtual agents that could intervene to provide that support itself 83 when necessary. 84

Related work

Reciprocal peer tutoring is a form of collaborative learning where same-age students work 86 together by taking turns teaching one another, despite neither of them being an expert 87 04 (Palinscar and Brown 1984). Prior work has shown that it can be an improvement over 88 individuals learning alone, but the differences between novice peer tutors and expert tutors 89 in both content knowledge and pedagogical knowledge may have significant consequences for 90 both the process and outcomes of tutoring (Palinscar and Brown 1984). To better understand 91whether and how interpersonal closeness between peer tutors and their tutees intersects with 92tutors' domain knowledge and tutoring self-efficacy to impact their use of indirectness while 93 tutoring, we draw on a number of prior theories. First, we describe the role that "face", or, 94desire to be approved of by others (Goffman, 1955; Brown and Levinson 1987) may play in the tutoring process, for both tutor and tutee. We then discuss prior approaches to face-threat 96 mitigation in learning, such as through indirectness in tutoring. We then discuss other potential 97 interpersonal goals that tutors' indirectness might serve instead of face-threat mitigation, such 98as to demonstrate tutors' own uncertainty or lack of confidence in their own ability to tutor. 99 Finally, we discuss the nature of interpersonal closeness and how the relationship or rapport 100between tutor and tutee might impact the ways that tutors pursue the interpersonal goal of face-101threat mitigation. 102

Impact of face-threat in peer tutoring

First, prior work has argued that the provision of instructional feedback, directions, or 104unsolicited advice is a socially mediated process impacted, in part, by the interpersonal 105closeness between tutor and tutee (Wichmann and Rummel 2013; Feng and Magen 2015). 106The experience of being tutored by a peer may be a highly threatening experience for tutees, 107and as such, effective tutoring moves may be avoided by peer tutors when their interpersonal 108goals of building a relationship with their partner conflict with the interactional goals of 109tutoring (Person et al. 1995). 110

To understand the impact of those potentially threatening pedagogical moves on peer 111 tutoring, in particular the delivery of feedback, instructions, and tutors' explicit reflections 112on their partners' comprehension, we draw on theories of face management and politeness 113(Goffman, 1955; Brown and Levinson 1987). According to Brown and Levinson, social actors 114are motivated by their desire for what is referred to as positive face, or the desire to be 115approved of by others, and negative face, which is the desire to be autonomous and unimpeded 116by others (Brown and Levinson 1987). According to Goffman, interlocutors (more so in some 117 cultures, but to some extent in all cultures) are careful to avoid threatening their conversational 118

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partners' face – by approving of the partner to uphold positive face, and by allowing the 119 partner more autonomy, to protect the partner's negative face. If face threat is unavoidable, 120 such as when a tutor must correct a tutee's answer, Brown and Levinson claims that the 121 speaker will attempt to *mitigate the threat* by speaking indirectly or obliquely, by being polite, 122 or by simply avoiding the provision of that type of response entirely. 123

Tutors' instructions or directions, if they take the form of demands, may thus threaten 124students' negative face, and tutors' instructional feedback and comprehension-monitoring, if 125given in a blunt manner, may threaten their tutees' positive face (Brown and Levinson 1987; 126Johnson and Rizzo 2004; Roscoe and Chi 2008). Prior work has argued that, in response to 127such face-threat, untrained peer tutors use fewer instances of comprehension monitoring in part 128due to the social pressure to avoid what might be seen as a threatening comparison between 129tutor and tutee (Ray et al., 2013). Thus, peer tutors' desire to communicate agreeable, face-13006 boosting information may distort or impinge on the quality of the collaborative learning 131between a peer tutor and tutee (Dame and Tynan 2005; Person et al. 1995). 132

Mitigating face-threat in learning

More skilled peer tutors, however, may be able to mitigate the face threat of such instructional 134moves to their students, perhaps by delivering those moves in an indirect or polite way (Person 135et al. 1995). As we described above, indirectness is one of the verbal conversational strategies 136that play a role in face management (some others are praise and acknowledgement). However, 137as Person et al. (1995) have argued, while indirectness and politeness may reduce face-threat, 138they may also introduce ambiguity and vagueness when used in an instructional or tutoring 139context (Person et al. 1995). Particularly for what Person et al. (1995) refer to as "closed-140world" domains, such as in algebra, where there is a definitive answer to problems, the 141 repeated use of indirectness over time from the tutor may lead the tutees to distrust the tutors' 142competence. 143

In a computer-supported collaborative learning (CSCL) context, the medium of the inter-144action is likely to impact the ways in which collaborating students attempt to mitigate the face-145threat of their instructional moves. For instance, Mottet and Beebe (2002), Kerssen-Griep et al. 14607 (2008) and many others have identified a set of nonverbal behaviors that can help mitigate 147face-threat in classroom instruction. They identified that interpersonally skilled instructors use 148the nonverbal immediacy behaviors of establishing eye contact, smiling, and body orientation 149to indicate their connection to the students as a way of reducing potential face-threat (Mottet 150and Beebe 2002; Kerssen-Griep et al. 2008). For CSCL systems, however, the medium of 151interaction may not allow for such nonverbal behaviors. If the CSCL system is purely text-152based, then the students no longer have the ability to use nonverbal immediacy to mitigate 153face-threat, and must instead rely on verbal strategies such as indirectness or avoiding the 154potentially threatening move entirely (Morand and Ocker 2003). 15508

Impact of domain knowledge and instructional self-efficacy on indirectness

While face-threat mitigation may be one role played by indirectness in instructional moves157from peer tutors, it may instead be the case that hedging is used an indicator of the uncertainty158of the peer tutor. Coates (1987) has argued that hedging is used as part of socio-cognitive159Q9processes to fulfill the conversational strategies of politeness, uncertainty, or indirectness160(Coates, 1987). Hedges, and other markers of indirectness such as "subjectivizers" like "I161

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think" or "I guess", can thus be viewed as what Prince et al. (1982) calls "shields", to create a 162distance between the speaker and their proposition (Prince et al. 1982). Rowland (2007), in his 163Q10 analysis of indirectness in the math classroom, describes the linguistic role that hedging plays 164in middle school students' verbalization of mathematic predictions (Rowland 2007). Accord-165ing to Rowland, students use hedges, subjectivizers, and what he refers to as "approximators" 166or "vague category extenders" ("and stuff", "or something", etc) for much the same shielding 167function, using uncertainty to save them from the risk of embarrassment if they are wrong in 168their predictions of the answer (Rowland 2007). However, it is not clear whether and in what 169situations peer tutors use indirectness to indicate their own uncertainty and save their own face, 170or to allow for the possibility of being wrong in order to save their tutees' face. 171

Although expert tutors and teachers may be able to effectively mitigate the face-threat of a 172tutoring move through the strategic use of indirectness, politeness, or a self-effacing disclosure, 173174untrained peer tutors may not be as deft in their face-work (Kerssen-Griep et al. 2008). Prior work has shown that in addition to domain knowledge, teachers' instructional self-efficacy is 175176likely to impact their ability to attend to the *interpersonal* goals of teaching as well as the instructional goals (Gibson and Dembo 1984; Mojavezi and Tamiz 2012; Saklofske et al. 177011 1988). Teachers' instructional self-efficacy, or their beliefs about their ability to impact student 178outcomes and the confidence that they can do so, have been shown to impact teachers' use of 179different types of feedback (Gibson and Dembo 1984) as well as impacting their students' 180motivation and achievement (Mojavezi and Tamiz 2012). In addition, though we did not 181 measure tutoring ability, we used the tutors' score on a pre-test as a proxy for their prior 182domain knowledge, following Rowan et al.'s (1997) findings that teacher prior domain 183knowledge was predictive of their students' performance, and following the intuition that 184tutors' prior domain knowledge may impact their own certainty in their responses. However, 185those prior findings have been for adult teachers, and thus it is not clear whether and how 186instructional self-efficacy impacts peer tutors' ability to balance between the interpersonal and 187 instructional goals of tutoring. That prior work also does not take into account the interpersonal 188closeness of the tutor and the tutee, and is thus unable to identify whether or how that closeness 189impacts the tutors' use of face-threat mitigation in tutoring. 190

Impact of interpersonal closeness on face-threat mitigation

In some cases, it is possible that the potential face-threat of the three types of tutoring moves 192described above (feedback, instructions, and comprehension monitoring) may not need to be 193explicitly mitigated by a peer tutor at all. Instead of delivering potentially face-threatening 194moves indirectly, the interpersonal closeness between students may allow for behaviors that 195might otherwise be perceived as face-threatening to instead be permissible (Brown and 196Levinson 1987). This follows theories of rapport-building, such as from Spencer-Oatey 197 (2005), which suggest that a greater rapport, or interpersonal closeness, between interlocutors 198allows for speech acts which would otherwise be perceived as face-threatening. 199

To operationalize the development of interpersonal closeness and its impact on facemanagement, we draw on Tickle-Degnen and Rosenthal (1990)'s work on interpersonal 201 rapport as well as Spencer-Oatey's (2005) model of face-management, as integrated into a 202Q12 theory of rapport-building that incorporates face-management, mutual attentiveness, and 203 coordination between members of an interacting dyad, by Zhao et al. (2014). As rapport, or 204 short-term interpersonal closeness, begins to develop, partners convey their mutual attention to 205 each other, both nonverbally, as well as through verbal behaviors that index that attention, such 206

as referencing shared interests and experiences (Zhao et al. 2014). Initially, partners may need 207to expend more effort in managing the face-threat to their partner, perhaps through what 208Tickle-Degnen and Rosenthal (1990) have described as nonverbally displaying positivity to 209the other person. This may also take the form of face-boosting behaviors like praise or self-210effacing negative self-disclosure, such as "I suck at these kinds of problems too" (Zhao et al. 2112014). These theories argue that we would expect the relative importance of face management 212to decrease as the relationship or rapport between tutor and tutee develops (Spencer-Oatey 2132005; Tickle-Degnen and Rosenthal 1990). 214

In tutoring, prior work has found, using friendship as a proxy for long-term rapport, that 215tutoring dyads of friends engage in more violations of social norms, such as playful teasing and 216social challenges, and that these are correlated with learning gains in friends. Those same 217behaviors, however, led to decreased learning among strangers (Ogan et al. 2012). This further 218suggests that a social relationship between tutor and tutee allows them to playfully challenge the 219other while tutoring, in what might be seen as face-threatening acts if done between strangers. 220However, this prior work did not look at the impact of such social relationships (friendship or 221rapport) on indirect delivery of tutoring moves as one way to mitigate the potential face-threat 222involved in learning, and how tutors' self-efficacy predicted their use of this indirectness, as 223described above. 224

In sum, while some have argued that indirectness may be used by peer tutors to mitigate 225potential face-threat to tutees, others have argued that indirectness is instead used as a shield to 226save the speaker's face when they are uncertain. In addition, some prior work has argued that 227indirectness is beneficial for mitigating face-threat to students, while others have argued that it 228might be harmful due to its ambiguity, or that it may be simply unnecessary for dyads of students 229with sufficient interpersonal closeness. However, it remains unclear (1) whether and to what 230extent untrained peer tutors' indirectness is used to soften the blow of potentially face-threatening 231tutoring moves or to indicate their own uncertainty; (2) how that use of indirectness is impacted by 232their interpersonal closeness with their tutee, their instructional self-efficacy, and domain knowl-233edge; and (3) how that indirectness impacts their tutees' subsequent problem-solving and learning 234gains. To help address this gap, we propose the following research questions. 235

Research questions

RQ1

How does a tutoring dyad's interpersonal closeness impact tutors' use of potentially face- 238 threatening tutoring moves? 239

RQ2

How do tutors' self-efficacy and interpersonal closeness impact their use of indirectness while 241 delivering those tutoring moves? 242

RQ3

How does a tutor's use of indirect feedback, instructions, and comprehension-monitoring 244 impact tutees' learning behaviors and outcomes? 245

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Methods

We seek to investigate how the interpersonal closeness between peer tutors and their tutees 247impacts the tutors' use of indirectness with instructions, feedback, and comprehension mon-248itoring, and how those moves in turn impact tutees' learning. We will first describe the peer 249tutoring data we collected, including the two ways we operationalize the interpersonal 250closeness between members of a peer tutoring dyad (i.e. their relationship and their rapport). 251We then describe the set of tutoring moves and indirectness dialogue markers that we 252annotated our dialogue corpus for, and, finally, we describe the measures we use to 253operationalize tutors' self-efficacy and tutees' learning. 254

Dialogue corpus

The dialogue corpus described here was collected as part of a larger study on the effects of 256rapport-building on reciprocal peer tutoring. The participants were assigned to 12 dyads that 257alternated tutoring one another in linear algebra equation solving for 5 weekly hour-long 258sessions, for a total corpus of ~60 h of face-to-face interactions. Each session was structured 259such that the students engaged in brief social chitchat in the beginning, then had one tutoring 260period of 20 min with one of the students randomly assigned to the role of tutor. They then 261engaged in another social period, and concluded with a second tutoring period where the other 262student was assigned the role of tutor. This process was repeated for five sessions over five 263weeks. As each student was randomly assigned to be the tutor for half of the tutoring periods, 264they were not expected to have any greater prior knowledge than their partner for the problems 265they were tutoring them on. 266

All students were supported with a set of instructions on how to teach the particular 267problems for which they were assigned the role of tutor. These instructions include procedural 268instructions for problems of a similar form as the ones the tutees were solving. The students 269took a pre-test before the first session and a post-test after the fifth session to assess their 270learning gains. The participants (mean age = 13.5, min = 12, max = 15) came to a lab on an 271American university campus in a mid-sized city for the study. Half were male and half were 272female, assigned to same-gender dyads, so that, in other work with this corpus, gender 273differences in the social, rapport-building behaviors of the participants could be identified. 274No gender differences were found here. To investigate how the use of various tutoring 275behaviors differs between dyads with varying degrees of interpersonal closeness, we used 276friendship as a proxy for long-term closeness and asked half of the participants to bring a 277same-age, same-gender friend to the session with them, and for the other half of the dyads, we 278paired them with a stranger. Audio and video data were recorded, transcribed, and segmented 279for clause-level dialogue annotation, following Chi (1997). 280

Rapport rating

The rapport, or short-term interpersonal closeness between the participants, was evaluated 282 using a "thin-slice" approach following Ambady and Rosenthal (1992). They found that 283 rapidly-made judgments of interpersonal interactions were highly accurate assessments of 284 those interpersonal dynamics (Ambady and Rosenthal 1992). Following this, we divided our 285

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corpus into 30-s video slices, and provided naive raters with a simple definition of rapport, as286well as provided them with those clips in a randomized order, so they would rate each slice's287rapport, and not the delta across multiple slices. Because, in the thin-slice methodology, the288raters are intended to be naïve observers, we did not use a train-retrain approach as is common289in dialogue annotations (Chi 1997; Ambady and Rosenthal 1992).290

Instead, three raters rated the rapport present in each slice in our corpus on a Likert scale 291from 1 to 7, so that those ratings could provide ground truth for future analyses of the rapport 292dynamics. To account for each rater's overuse or underuse of the Likert scale categories, we 293used a weighted majority vote approach, following Sinha and Cassell (2015), and Kruger et al. 294(2014). We weighted each rater's vote for the slice's rating by the inverse of that rater's 295frequency of use for that rating category, so that each rater's vote was weighted to account for 296their overall overuse or underuse of a particular rating. The final single rating was then chosen 297for each slice using that inverse bias-corrected weighted majority vote approach. 298

While this is useful for obtaining the rapport between participants at any given moment of 299the interaction, it does not provide a summary measure with which we can understand the 300 relationship between the interpersonal closeness of the dyad, the tutors' use of indirectness 301with their tutoring, and the tutees' learning. Therefore, from the roughly 120 thirty-second 302 slices in each hour-long session, we calculated a summary rapport score for each session 303 following Sinha (2016). Prior work has shown that statistical summaries such as a measure of 304 central tendency or proportion of high and low ratings of rapport collapse the temporal 305dimension and are not as robust as more stochastic-based models which capture the evolution 306 of rapport over time (Sinha 2016). Sinha (2016) found a significant relationship between a 307 stochastic-based measure of rapport and students' learning. This was more predictive of 308student learning than statistical summaries such as the simple average, and thus, we adopt 309 Sinha's approach for generating one such stochastic measure of rapport, or "utopy". The 310"utopy" is, intuitively, the likelihood of the rapport to be increasing, weighted by the size of the 311 increase; this measure can thus capture the temporal dynamics of interpersonal closeness 312 development. 313

To obtain this measure for each session, we fit a Markov chain of order 1 to the sequence of 314 120 rapport ratings in each session, to generate the transition probability matrix for the 315likelihood of that dyad to transition from one rapport level to another. We then compute the 316"utopy" by summing the transition probabilities of each transition from one rapport level to 317another (e.g. rapport 2 to 4), weighting each of the transition probabilities by their distance 318from the diagonal, so that larger changes in rapport were given more weight. This provides us 319with a measure of the "utopy" for a given session, or the likelihood that rapport will be 320 transitioning to higher states for that dyad in that particular session (Dillenbourg 2015; Sinha 3212016). In other work, we have found a significant association between utopy and students' 322problem-solving and learning gains (reference removed). In this paper, we build off of that 323 work by using the utopy measure to investigate the relationship of rapport dynamics with 324learning process behaviors, here, indirectness with peer tutoring moves. 325

Dialogue annotation

As part of a larger study on the relationship between rapport-building and peer tutoring, this 327 corpus was annotated by a set of four trained annotators for a set of pedagogical, tutoring-328 related behaviors from both the tutor and tutee, as well as a set of social, rapport-building 329

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verbal conversational strategies and nonverbal behaviors (not discussed in this article). In 330 Table 1, we describe the tutoring strategies included in the analyses in this paper. This set of 331 tutoring behaviors includes feedback from tutors on their partner's correctness, step-level 332 procedural instructions (also called "knowledge-telling"), and explicit comprehension moni-333 toring on the part of the tutor (following Madaio et al. 2016). We also annotated for learners' 334step-level verbalizations of their problem-solving procedures, to understand how their self-335 explanations were impacted by their partners' tutoring strategies. The Krippendorff's alpha for 336 all codes was over 0.7. 337

To understand the ways that tutors in dyads with differing levels of interpersonal closeness 338 (both friendship status and rapport) modified the delivery style of their tutoring instructions 339 and feedback, we coded our corpus for four types of indirectness. These indirectness markers 340 were coded independently of the tutoring moves. Thus the indirectness markers may have been 341 either used alone or co-occurring with the tutoring moves (e.g. instructions, advice, or 342feedback). We annotated for: apologizing, hedging or qualifying, the use of vague category 343 extenders, and "subjectivizing" (Zhang 2013; Neary-Sundquist 2013; Fraser 2010), as de-344scribed in more detail in Table 1. The Krippendorff's alpha for all four codes was over 0.7. 345

To understand how tutoring moves were delivered indirectly, we analyzed the co-346 occurrence of the annotated indirectness markers with the annotated tutoring strategies for 347 each given clause. We identified clauses as indirect feedback if a clause had an annotation for 348 indirectness and an annotation for feedback, either positive or negative. Typical examples of 349indirect feedback include: "I think you got it." "I guess that's what it is." "Oh, no, actually, it's 350not." "No, it's just nineteen." We similarly identified each clause as indirect instructions if it 351had an annotation for indirectness and an annotation for procedural instructions. Typical 352examples of indirect instructions are: "Actually, just add five here." "I think you're gonna 353 divide it by a fraction or something." "I would probably subtract the sixteen." Finally, we 354labeled each clause as *indirect comprehension-monitoring* if it had an annotation for indirect-355ness and an annotation for comprehension monitoring. Some typical examples include: "Yeah, 356 you just got mixed up between the terms." "It just seems like you roam a lot." "Oh. I guess 357 you're not confused." All other instances of these three tutoring strategies without co-358 occurrence with an indirectness marker were thus identified as "direct" feedback, instructions, 359or comprehension-monitoring. 360

Finally, we provided the participants with a questionnaire following the study with a set of 361 items for constructs relevant to rapport-building and the tutoring process. To evaluate their 362 self-efficacy for tutoring, we used a 7-item scale indexing whether the participants believed 363 that they were able to be effective in positively impacting their tutees, following Gibson and 364 Dembo's construct of "personal teaching efficacy" (Gibson and Dembo 1984). We use a 365 median-split on those survey results to categorize tutors as high or low self-efficacy, relative to 366 the rest of the participants. 367

Results

In the following sections, we investigate our research questions about the impact that interpersonal closeness and tutors' prior knowledge and tutoring self-efficacy have on peer tutors' 370 tutoring strategies. We do this first by analyzing the base rates of three types of tutoring 371 behaviors and the impact of the aforementioned factors on tutors' use of those behaviors. We 372 then describe tutors' base rates of indirectness markers (qualifiers, subjectivizers, etc) and 373

| | | Example | "Sorry, it's negative 2." "You just add 5 to both sides." | "You have to multiply or something." | "I guess you divide by 3 here." "Yen. that's it!" | "No, that's wrong." "Multiply it by the fraction." | "I don't think you understood this." | "Ok, so I'm gonna add 3 first." | |
|------------|---|------------|---|--|--|--|--------------------------------------|--|-----|
| | 5 | Definition | Apologies used to soften direct speech acts. Qualifying words to reduce intensity or | certaining on uncrainces. Words used to indicate uncertainty by referring | to vague caregories. Words that reduce intensity or certainty. Explicit evaluations of correct answers or stens. | Explicit evaluations of incorrect answers of steps. Explicit evaluations of incorrect answers or steps. | the future. | reneering on mens or men partners knowledge state. Tutees verbalizing their own problem-solving steps | ROC |
| JAC | Table 1 Annotations of tutoring moves and indirect language | Code | Apology Hedges | Extenders | Subjectivizer Positive feedback | Negative feedback Procedural instructions | Comprehension- | monuoring Procedural verbalizations | |
| | Table 1 Annotations o | Code Type | Indirectness Indirectness | Indirectness | Indirectness Tutoring strategy | Tutoring strategy | Tutoring strategy | Learning strategy | |
| 🖄 Springer | t1.1 | t1.2 | t1.3 t1.4 | t1.5 | t1.6 t1.7 | t1.8 t1.9 | t1.10 | t1.11 | |

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investigate the impact that interpersonal closeness, tutors' prior knowledge and tutoring selfefficacy have on tutors' co-occurring usage of indirectness with the three tutoring moves, or, their *indirect tutoring strategies*. Finally, we investigate the impact that these indirect tutoring strategies, and their direct counterparts have on tutees' responses and learning outcomes. For the following analyses, all results are significant after correcting for multiple hypothesis testing using the Benjamini-Hochberg post-hoc correction (Benjamini and Hochberg 1995). 374

Peer tutors' use of face-threatening tutoring moves

First, we investigated our research question about whether tutors with lower interpersonal 381 closeness with their tutees used fewer instances of negative feedback and comprehension 382monitoring (RQ1). Due to the potential for those two types of tutoring moves in peer tutoring 383 dyads with lower interpersonal closeness to be perceived as face-threatening, we hypothesized 384that tutors in dyads of strangers and tutors in low-rapport dyads would provide fewer instances 385 of negative feedback (Person et al. 1995) and fewer instances of explicit comprehension 386 monitoring (Roscoe and Chi 2008). Conversely, we hypothesized that friends and high-387 rapport dyads, who may have less need for face-threat mitigation (Spencer-Oatey 2005), 388 would thus provide negative feedback more often (Person et al. 1995), as well as providing 389 more comprehension monitoring (Roscoe and Chi 2008). In this section, we analyze all 390 occurrences of those two types of tutoring moves (feedback and comprehension monitoring), 391regardless of whether they were delivered in a direct or indirect manner. We normalized the 392aggregate frequencies for those two tutoring moves by the total number of "on-task" utter-393 ances, or the total number of annotated tutoring strategies (e.g. explanations, feedback, 394comprehension monitoring) given by each speaker, in each session, following Madaio et al. 395(2016).396

To investigate this hypothesis about the influence of interpersonal closeness and self-397 efficacy on tutoring strategies, we first ran an omnibus repeated measures MANOVA on the 398 normalized frequency of tutors' negative feedback, positive feedback, and comprehension-399 monitoring. We crossed the between-subjects factors of relationship (friend/stranger), rapport 400 (high/low), prior knowledge, and tutoring self-efficacy with the within-subject factor of session 401 402 and tutoring period, using each dyad's tutoring period and session number as error terms. The rapport and relationship factors are intended to capture the phenomena of interpersonal 403404 closeness in the short-term and long-term, respectively. In this dataset there was no correlation between the rapport and the relationship (i.e. there are both high- and low-rapport dyads of 405friends and strangers), and so we include both factors in our model. This MANOVA revealed a 406 significant multivariate main effect of relationship (F (3, 68) = 5.23, p < 0.01) on the three 407 outcome variables (negative feedback, positive feedback, comprehension-monitoring). Given 408the significance of the overall omnibus test, univariate tests were conducted to identify the 409differential impact of those effects on the three outcome variables. 410

Surprisingly, for the univariate model for *negative feedback*, there was no statistically 411 significant effect of any of the factors. For the univariate model for *positive feedback*, however, 412 we found a highly significant univariate main effect of relationship on the amount that tutors 413 used positive feedback (F(1,64) = 12.8, p < .001). To find the direction of that difference, we 414 ran a t-test, and found that stranger tutors were significantly more likely (t(71) = 3.77, p < .001)415to use positive feedback (m = .17, sd = .12) than friend tutors (m = .08, sd = .08). We hypoth-416 esized that perhaps stranger tutors were using more positive feedback because their tutees were 417 solving more problems correctly. Therefore, we conducted a t-test, which showed that stranger 418

tutees did not solve significantly more problems than friend tutees. This suggests that this 419positive feedback was serving an interpersonal function, rather than the interactional function 420of indicating correctness. Finally, for the univariate model for *comprehension-monitoring*, we 421 found a significant univariate main effect of relationship on the amount that tutors provide 422 comprehension monitoring (F(1,70) = 5.9, p < .05). Friend tutors used significantly more (m = 423 0.05, sd = 0.04) comprehension monitoring than stranger tutors (m = 0.03, sd = 0.02), at 424 (t(60.9) = 2.43, p = .01), confirming our hypothesis. 425

Peer tutors' overall use of indirectness

Before we investigated our research questions about the co-occurrence of indirectness markers 427 with tutoring strategies, we first inspected the base rate of the four types of indirectness we 428 annotated, used in any utterance in our corpus (both on-task and off-task). Because our 429annotators coded indirectness in any utterance in the corpus, we normalized the frequency 430of these codes by the total number of utterances from that speaker, in that session. By far the 431most frequently used marker of indirectness in our dataset was the use of qualifiers or hedges 432 (e.g. "just", "actually", etc.) with normalized mean = .05 and standard deviation = .04, follow-433 ed by subjectivizers (e.g. "I think", "I guess", etc) (m = .02, sd = .03), and apologies (m = .01, sd = .03)434sd = .01), and with vague category extenders by far the most infrequent (e.g. "and stuff", "or 435something", etc.) (m = .002, sd = .01). This distribution aligns with findings from Rowland's 436(2007) work studying the use of hedges, subjectivizers (what he calls "shields"), and extenders 437 in student-teacher mathematics lessons. See Fig. 1 for a boxplot showing the distribution of the 438frequency of each of the four types of indirectness annotated for, as normalized by the total 439number of utterances from that speaker, in that session. 440

Peer tutors' use of indirectness with tutoring moves

We then wanted to investigate the factors impacting peer tutors' use of each of these four types 442 of indirectness when used with their procedural instructions, feedback, and comprehension 443 monitoring (RO2). From prior literature on the use of hedges and subjectivizers to convey 444

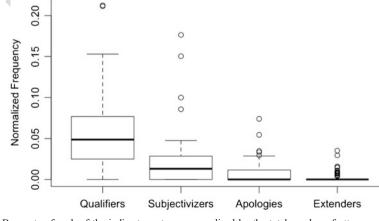


Fig. 1 Base rate of each of the indirectness types, normalized by the total number of utterances for a given speaker in a given session

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uncertainty (Rowland 2007), we hypothesized that tutors with lower tutoring self-efficacy 445 would use more indirect language to indicate their uncertainty. However, an alternative 446 hypothesis is that tutors with greater tutoring self-efficacy would attend more to their tutees' 447 needs for face-management and would thus use more indirect language to mitigate the 448 potential face-threat of tutoring moves (Brown and Levinson 1987; Kerssen-Griep et al. 4492008; Saklofske et al. 1988). We additionally hypothesized that tutors with low interpersonal 450013 closeness with their tutees (stranger tutors and tutors in low-rapport dyads), would use a more 451indirect style when delivering tutoring moves that may be potentially face-threatening 452(feedback, procedural instructions, and comprehension-monitoring) than tutors with greater 453interpersonal closeness (Brown and Levinson 1987; Johnson and Rizzo 2004). 454

To investigate this hypothesis, we first ran an omnibus repeated measures MANOVA on the 455normalized frequency of tutors' indirect feedback, indirect instructions, and indirect compre-456hension-monitoring. To do this, we first computed the aggregated frequency of the three 457annotated tutoring moves that co-occurred with an annotation of an indirectness marker in 458 the same utterance, as described in the methods section. We then normalized each of these 459aggregated frequencies by the total number of occurrences of that tutoring move by that 460 speaker, in that session, to control for the opportunities for a given tutoring move to be delivered 461 indirectly. As in RO1, we crossed the between-subjects factors of relationship (friend/stranger), 462rapport (high/low), prior knowledge, and tutoring self-efficacy with the within-subject factor of 463 session and tutoring period, using each dyad's tutoring period and session number as error 464terms. This MANOVA revealed a significant multivariate interaction effect of relationship with 465tutoring self-efficacy (F (3, 33) = 3.00, p < 0.05) on the three outcome variables (indirect 466 feedback, indirect instructions, and indirect comprehension-monitoring). Given the significance 467 of the overall omnibus test, univariate tests were conducted to identify the differential impact of 468 those effects on the three outcome variables. For the univariate model for *indirect feedback*, 469there was no statistically significant effect of any of the factors. 470

For the univariate model for tutors' use of *indirect instructions*, we found a highly significant 471 main effect of relationship (F(1,70) = 7.4, p < .01). To find the direction of that difference, we ran a 472Mann-Whitney-Wilcoxon test, for comparing means of non-normal distributions. Tutors who 473paired with a stranger were significantly more likely (U=3749, p < .05) to use indirect instruc-474 tions (m = .02, sd = .03) than tutors paired with a friend (m = .01, sd = .02), which aligns with our 475hypothesis about interpersonal closeness. In this univariate model, there was also a significant 476interaction effect of rapport and self-efficacy on tutors' use of *indirect instructions* (F(1,70) = 4.5, 477 p < .05), regardless of the relationship with the tutee. High-self-efficacy tutors with low rapport 478 with their tutees were significantly more likely (U=4052, p<.001) to use more indirect 479instructions (m = .03, sd = .02) than high self-efficacy tutors with high rapport with their tutees 480 (m = .01, sd = .01).481

This lends further support to our hypothesis that tutors with lower interpersonal closeness 482(here, rapport) with their tutees would use more indirectness with potentially face-threatening 483tutoring moves than those with greater interpersonal closeness. However, it is primarily the 484high self-efficacy tutors with low rapport with their tutees that appear to use this strategy, as 485they were marginally more likely (U=928, p=.07) to use more indirect instructions (m = .03, 486sd = .02) than low self-efficacy tutors with low rapport with their tutees (m = .01, sd = .01). 487 This also lends support to the hypothesis that greater self-efficacy for tutoring may allow the 488tutors to strategically use indirectness to fulfill interpersonal goals (i.e. mitigating face-threat). 489See Fig. 2 for the interaction effect between rapport and self-efficacy on tutors' indirect 490**Q14** instructions. 491

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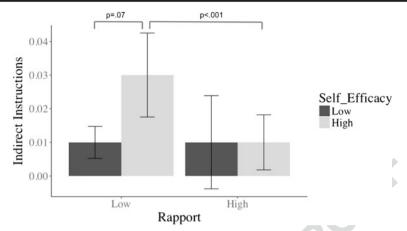


Fig. 2 Interaction effect of rapport and self-efficacy on tutors' use of indirect Instructions

We then ran the same univariate test on tutors' use of *indirect comprehension-monitoring*, 492 finding a highly significant interaction effect of relationship and self-efficacy on the tutors' use 493 of indirect comprehension-monitoring (F(1,35) = 8.86, p < .01). Much like the effect of rapport 494 on indirect instructions, tutors who are strangers are more likely to deliver their 495 comprehension-monitoring indirectly when they have high tutoring self-efficacy. However, 496 no pairwise comparisons were significant. 497

Impact of peer tutors' indirect tutoring moves on tutees' learning

Finally, we investigated our research question about the effect of tutors' use of indirect 499instructions and comprehension-monitoring on tutees' learning process and outcomes 500(RQ3). From prior literature on the motivational benefits of face-threat mitigation, we hypoth-501esize that there will be an interaction between a dyad's interpersonal closeness and their use of 502indirect tutoring language on learning outcomes. Specifically, we hypothesize that in dyads 503with low interpersonal closeness (stranger dyads and low-rapport dyads (both friend and 504stranger)), when tutors use more indirect tutoring moves, their tutees will attempt and solve 505more problems and will learn more from pre- to post-test (Kerssen-Griep et al. 2008; Roscoe 506and Chi 2008). An alternative hypothesis is that more *direct* feedback, instructions, and 507comprehension monitoring is associated with improved problem solving and learning, follow-508ing Person et al. (1995)'s findings that indirectness may lead to ambiguity in closed-world 509domains like algebra. 510

We thus ran a linear mixed-effect model using the tutees' *percent of problems solved* in 511 each tutoring period as the dependent variable, and using the tutors' normalized frequency of 512

| Parameter | Estimate | Standard error | df | t value | <i>p</i> val |
|-----------------------------------|----------|----------------|-------|---------|--------------|
| Indirect comprehension-monitoring | -0.01 | 0.37 | 8.78 | -0.23 | 0.98 |
| Direct comprehension-monitoring | -0.07 | 0.26 | 9.05 | -0.27 | 0.79 |
| Indirect instructions | 0.64 | 0.32 | 22.24 | 1.98 | 0.04 |
| Direct instructions | -0.08 | 0.19 | 20.00 | -0.43 | 0.68 |
| Tutees' self-explanations | 0.05 | 0.19 | 23.95 | 0.27 | 0.78 |

t2.1 **Table 2** Model details of mixed effect model for indirect and direct tutoring moves on problems solved

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| 3.1 | Table 3 | Model | details o | of mixed | effect model | for indirect | and direct | tutoring mo | oves on r | problems attem | pted |
|-----|---------|-------|-----------|----------|--------------|--------------|------------|-------------|-----------|----------------|------|
| | | | | | | | | | | | |

| t3.2 | Parameter | Estimate | Standard error | df | t value | p value |
|------|-----------------------------------|----------|----------------|-------|---------|---------|
| t3.3 | Indirect comprehension-monitoring | -0.002 | 0.36 | 9.11 | -0.007 | 0.99 |
| t3.4 | Direct comprehension-monitoring | -0.07 | 0.25 | 9.09 | -0.29 | 0.77 |
| t3.5 | Indirect instructions | 0.84 | 0.27 | 22.32 | 3.08 | 0.005 |
| t3.6 | Direct instructions | -0.23 | 0.18 | 11.62 | -1.26 | 0.23 |
| t3.7 | Tutees' self-explanations | 0.19 | 0.16 | 23.14 | 1.25 | 0.22 |

indirect instructions and indirect comprehension-monitoring as fixed effects, along with 513interaction terms for each of the above with Relationship and Rapport, with random effects 514for Dyad and Session. We also included normalized frequency of the tutees' self-explanations 515as a fixed effect, following the findings of Madaio et al. (2016) that tutees' self-explanations 516were a significant predictor of their learning. We also included the frequency of *direct* 517instructions and comprehension-monitoring (in addition to the indirect versions of those 518moves) as fixed effects to identify the impact of that directness on tutee learning. As detailed 519in the methods, the direct instructions and comprehension-monitoring were the remainder of 520the tutoring moves of those types without a co-occurring annotation of indirectness. In this 521model, stranger tutors' use of *indirect* instructions was positively predictive ($\beta = .64, p < .05$) 522of their tutees' problem-solving. No other factors were significant. All model parameters for 523the problems solved model are reported in Table 2. 524

We further investigated whether the use of indirectness with instructions and feedback 525 might serve a motivational role, leading to an increased amount of problems attempted for the 526 tutee. We thus ran the same mixed-effects model, but with the tutees' *percent of problems* 527 *attempted* as the dependent variable. In this model, stranger tutors' use of *indirect instructions* 528 was also significantly positively predictive ($\beta = .84$, p < .01) of their tutees' amount of 520 problems attempted. No other factors were significant. All model parameters for the problems 530 attempted model are reported in Table 3. 531

In addition to the shorter-term benefits of problem-solving during the tutoring interaction, 532we also wanted to investigate whether all of this hedging was beneficial for the tutees' learning 533gains from pre- to post-test. We thus ran a linear mixed effect model with tutees' overall 534learning gains as the dependent variable, and with the same fixed effects (tutors' indirect and 535direct tutoring moves, and tutees' self-explanation) and random effects (Dyad and Session) as 536described above. In this model, stranger tutors' use of *indirect comprehension-monitoring* on 537their partners' knowledge was positively predictive ($\beta = .29$, p = .05) of their tutees' overall 538learning gains, with the only other effect being a marginal negative association ($\beta = -.43$, 539p = .09) of tutors' *direct* instructions with their tutees' learning gains. All model parameters for 540the learning gains model are reported in Table 4. 541

| Parameter | Estimate | Standard Error | df | t value | <i>p</i> 7 |
|-----------------------------------|----------|----------------|-------|---------|------------|
| Indirect comprehension-monitoring | 0.29 | 0.15 | 28.96 | 1.95 | 0.0 |
| Direct comprehension-monitoring | 0.21 | 0.14 | 28.19 | 1.47 | 0. |
| Indirect instruction | 0.17 | 0.15 | 28.69 | 1.17 | 0. |
| Direct instructions | -0.43 | 0.25 | 28.69 | -1.70 | 0. |
| Tutees' self-explanations | 0.03 | 0.23 | 15.58 | 0.14 | 0. |

t4.1 **Table 4** Model details of mixed effect model for indirect and direct tutoring moves on learning gains

Tutees' responses to tutors' hedged instructions

These results show potential for hedged instructions and comprehension-monitoring to improve learning for some students in peer tutoring (perhaps due to the mitigation of face-threat).543We then wanted to look deeper to understand how the tutees *responded* to these hedged545dialogue moves, and how those responses differ by the interpersonal closeness of the dyad, to546better understand how the hedged instructions impact the tutees' problem-solving process. We547therefore used an adjacency pair approach, following Boyer et al. (2010) to identify the most548549

While a thorough analysis of all of the possible tutee responses to tutors' moves is beyond the 550scope of this article, we will discuss here the adjacency pairs that included tutees' responses to the 551indirect and direct instructions used by their partners. We extracted all of the tutees' responses to 552their tutors' use of *indirect instructions* as well as *direct instructions*, to identify differences in the 553way tutees with greater interpersonal closeness responded to the same type of tutoring move 554delivered directly and indirectly. We normalized the frequency of these adjacency pairs by the 555total number of moves included in the pair (e.g. tutors' direct instructions and tutees' self-556explanations) similar to our approach in RQ2. This was due to the large variance in the distribution 557of the tutoring and learning behavior types included in the adjacency pairs. 558

For peer tutoring dyads with low rapport, tutees are significantly more likely (t(21.4) = 2.3,559p = .03) to respond to their tutors' *indirect instructions* with *self-explanations* than to respond to 560*direct instructions* with self-explanations. Specifically, tutees in low-rapport dyads respond to 561indirect instructions with their own verbalized self-explanations three times as often (m = .007) as 562they respond to tutors' direct instructions with self-explanations (m = .002). This provides support 563for the hypothesis that indirect instructions are beneficial for tutees' problem-solving in low 564rapport dyads. Crucially, however, there were no significant differences in the ways tutees 565responded to indirect instructions in high-rapport dyads. That is, while tutors' hedged instructions 566led to increases in tute self-explanations, this benefit *only* accrued for low-rapport dyads. 567

Discussion

In this article, we investigated the extent to which peer tutors' instructional self-efficacy and 569domain knowledge intersect with their interpersonal closeness (both short- and long-term) with 570their tutee to impact the process and outcomes of peer tutoring, through peer tutors' hedging of 571potentially face-threatening tutoring moves. Critical components of the collaborative learning 572process, such as providing procedural instructions, feedback, and monitoring their partners' 573comprehension, have been postulated to be more likely to be avoided by peers due to the 574potential for those moves to threaten their partners' "face", particularly for peers with a more 575distant relationship with their partner (Brown and Levinson 1987; Person et al. 1995). 576

Here, the differences we found in tutors' use of indirectness with various tutoring behaviors 577 suggests that interpersonal, relational aspects of collaborative learning interactions like reciprocal 578peer tutoring are likely to impact the process by which students pursue their pedagogical task 579goals. For instance, although stranger tutors used more positive feedback than friends, their tutees 580were not solving significantly more problems correctly than friend-only dyads. This suggests that 581those tutors may be using that positive feedback to boost their partners' face rather than accurately 582diagnosing the correctness of their partners' problem-solving. This indicates that, while that 583positive feedback may serve a relational, face-boosting role, it may not serve a pedagogically 584

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useful role, and may even, as Person et al. (1995) pointed out, lead to ambiguity about the correct 585 procedures or answers, possibly eroding the tutee's trust in their tutor over time. 586

However, some peer tutors, particularly those with greater self-efficacy in their own 587 tutoring abilities, may be able to modulate their delivery of those tutoring moves to mitigate 588their potential face-threat, through hedging, qualifying, subjectivizing, or other forms of 589indirect delivery. It was not the case, as we hypothesized, that tutors with lower domain 590knowledge and instructional self-efficacy hedged more. In fact, we found instead that peer 591tutors with greater self-efficacy were *more* likely to hedge their face-threatening tutoring 592moves, but only when they had lower interpersonal closeness with their tutee (both 593relationship and rapport). This suggests that that hedging was serving an interpersonal 594function, rather than indicating tutors' uncertainty about the tutoring strategies they were 595using. We also found that only tutees with low interpersonal closeness with their tutors 596benefited from such hedging of instructions and comprehension-monitoring. This suggests 597 that peer tutors with greater instructional self-efficacy have a greater ability to attend to 598interpersonal as well as instructional goals, and that this self-efficacy may allow them to 599engage in beneficial interpersonal tactics, such as face-threat mitigation through indirect-600 ness, that might otherwise be avoided if their confidence in their tutoring abilities were 601 lower. More broadly, this work contributes to a more robust understanding of the ways in 602 which interpersonal closeness and instructional self-efficacy intersect to impact the collab-603 orative learning process, by way of reciprocal peer tutors' use of an indirect delivery style 604with tutoring strategies. 605

Researchers and designers of computer-supported collaborative learning (CSCL) systems 606 should thus be aware of how the interactional goals of tutoring may be impacted by the 607 interpersonal goal of face-threat mitigation: specifically, through peer tutors' overuse of 608 positive feedback or through their strategic use of hedging when delivering instructions and 609 explicit comprehension-monitoring. As Carmien et al. (2007) pointed out, while CSCL 610 systems may provide external scripts for students to follow, these scripts may conflict with 611 the internal scripts that students bring to bear on the interaction. As we found here, students' 612 interpersonal closeness with their partners provides one influencing factor on their interactional 613 behaviors. Thus, a CSCL system that does not take into account the interpersonal closeness 614 615 between collaborating students may find that the interactional support it provides to students conflicts with their interpersonal goals (i.e. mitigating face-threat). That is, a CSCL system that 616 recommends that students explicitly reflect on each other's knowledge or comprehension, 617 perhaps similar to Weinberger et al.'s work on argumentative discourse (Weinberger et al. 618 2005) may find that students are hesitant to provide such reflection, depending on the 619 directness with which it's phrased. Some peer tutors, in addition, such as the lower self-620 efficacy tutors we saw here, may need more scaffolding and support from a CSCL system to 621 deliver their tutoring moves in more interpersonally sensitive ways. 622

The selection, frequency of use, and delivery style of pedagogical behaviors used by 623collaborating students may differ depending on whether the collaborating students are 624 friends or strangers, or have a greater or lower rapport, and those same behaviors may have 625 different impacts on student learning, depending on that interpersonal closeness. Designers 626 of collaborative systems, such as Olsen et al.'s (2014) collaborative intelligent tutoring 627 system or Walker et al.'s (2011) adaptive collaborative learning system, might therefore 628 build in awareness of the interpersonal closeness between students. In addition to cognitive 629 instructional supports, such systems might provide social, interpersonal supports, such as 630 recommending students phrase their instructions or comprehension-monitoring to each 631

other more indirectly when interpersonal closeness is lower, particularly for tutors with632lower self-reported self-efficacy. However, these recommendations run the risk of633overscripting (Dillenbourg 2002), and should thus be used judiciously.634

635 These findings can also inform the design of collaborative dialogue systems, with conversational agents that could support collaborative learning by modeling different ways of 636 delivering instructions, feedback, comprehension-monitoring, or other potentially face-637 threatening instructional moves. Such agents have been used in prior CSCL work as in Tegos 638 et al. (2016) and Wang et al. (2017) to promote students' academically productive talk and 639 transactive talk, respectively. Those conversational agents might detect the interpersonal 640 dynamics among the students and between students and the conversational agent and recom-641 mend interpersonal moves (such as indirectness) to fulfill interpersonal goals in addition to the 642 interactional goals of learning. One such rapport-building system is the "Socially-Aware Robot 643 Assistant", or S.A.R.A., system (Matsuyama et al., 2016; Zhao et al. 2014; Sinha and Cassell 644016 2015), which uses a set of social conversational strategies to build a deeper rapport with its 645 users over time. To that end, as one way to detect the interpersonal closeness described here, 646 Yu et al. (2013) developed a method for the automatic prediction of friendship, which we 647 found (the lack of) here to be a significant predictor of indirect instructions and comprehen-648 sion-monitoring. For the shorter-term closeness of rapport, which we found to be a significant 649 predictor of indirect instructions, Zhao et al. (2016) have developed a method for the automatic 650 detection of rapport based on temporal association rules between multimodal data such as 651students' social conversational moves and nonverbal behaviors and the subsequent change in 652their rapport, which Madaio et al. (2016) extended to include the tutoring and learning 653**018** behaviors of a peer tutoring dyad to detect their rapport. 654

Limitations and future work

This work is part of a larger research program to understand the ways in which interpersonal 656 rapport may impact teaching and learning, and it is already being used to inform the design of 657 conversational agents that simulate a peer tutor as the front end of an intelligent tutoring system. 658 Such a virtual agent could collaborate on teams with students in the ways that a peer would, while 659 playing the role of a peer tutor. This goal is furthered by studying the process of rapport 660 development in peer tutoring, implementing that model in the agent's dialogue management, 661 and, perhaps, by reducing the face-threat of particular instructional moves when necessary by 662 delivering those tutoring strategies indirectly, in socially appropriate ways. 663

One of the limitations of the study reported in this article is the small sample size, particularly 664 for dyads of strangers. While a path analysis may have elucidated possible mediation effects from 665 interpersonal closeness to tutoring moves to tutee learning, we did not find such effects, perhaps 666 due to the low power of our small sample size. We have thus recently finished conducting a 667 similar study with 22 dyads of strangers to better understand how the rapport-building process 668 develops within dyads starting from the same interpersonal baseline, and how that rapport impacts 669 their teaching and learning processes and outcomes. Another limitation of this work is the 670 culturally dependent nature of what may be perceived as face-threatening or indirect by the 671 interlocutors (Spencer-Oatey 2005). While we did not code for face-threat here, future work may 672 provide an operationalization of the face-threat of each utterance (following Cassell and Bickmore 673 2003) to investigate the putative mechanism by which directness and indirectness may impact 674 student learning. Thus, future work exploring this face-threat should take into account how the 675 Intern. J. Comput.-Support. Collab. Learn.

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culture of the participants impacts perceptions of face-threat and indirectness (and how the culture 676 of the annotators may impact their annotation). Ogan et al., (2015), among others, have already 677**Q19** begun to explore how the collaboration process differs from culture to culture, studying how 678 students collaborate while using intelligent tutoring systems in Chile and the United States, among 679 other countries. 680

We are also currently involved in investigating other potentially face-threatening pedagog-681 ical behaviors, to understand whether and how high-rapport dyads engage in, for instance, 682 cognitive conflict, help-seeking, help-offering, and others. In this article, while we used the 683 normalized aggregate frequency of a particular set of annotated behaviors to understand 684 differences between groups of dyads, some of the most beneficial tutoring behaviors may 685 occur infrequently or may have their benefits impacted by the contingent patterns of use and 686 response from their partner (Ohlsson et al. 2007). Therefore, an analysis that does not take this 687 contingent, temporal pattern of use into account may miss important effects. In this article, we 688 have begun to analyze these contingent response patterns by using an analysis of the adjacency 689 pairs between tutor and tutee. We are currently building on this approach by using temporal 690association and sequence mining approaches to identify the core sequences of pedagogical and 691 social behaviors that contribute to greater rapport and learning. 692

We intend this article to contribute to the body of knowledge on the impact of social bonds 693 on the process of collaborative learning, as well as contributing to the design of socially-aware 694 computer-supported collaborative learning systems, which can more appropriately respond to 695learners' social bonds in pedagogically beneficial ways, and vice-versa. 696

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