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Keywords:

Social Incentives, Social Pressure, Moral Hazard in Teams, Laboratory Experiments

JEL codes:
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1. Introduction

Shirking in teams is one of the key topics addressed by economic theories of incentives (Holmström, 1982). In the absence of accurate and verifiable information regarding individual contributions, the best possible payment schemes available to a manager rely on team incentives. However, such compensation contracts provide insufficient incentives because they do not fully reward individual effort. Because team incentives are used when individual contributions cannot be contracted (see Holmström, 2017)\(^2\), any solution to shirking in teams must thus be of a non-contractual nature. Numerous solutions to shirking issues rely on social incentives (Bandiera, Barankay and Rasul, 2010; Ashraf and Bandiera, 2018). Social incentives refer to the effect of the social context on one’s motivation to complete work (see Tamir and Hugues, 2018; Corgnet, Hernan-Gonzalez, and Mateo, 2019). The social context is especially relevant in the case of teamwork because team members interact frequently (see Miller and Schuster, 1987; Ledford, Lawler and Mohrman, 1995; Hamilton, Nickerson and Owan, 2003; Lazear and Shaw, 2007; Nyberg et al., 2018).

The aim of the current paper is to compare the effectiveness and acceptability of various social incentive schemes intended to curb shirking and foster team performance. In other words, we will not only study the impact of these schemes on work effort but also measure workers’ willingness to embrace them. Our goal is thus partly to guide practitioners by identifying potential obstacles in the implementation of the various systems. We suggest that effective shirking solutions are those that promote work effort and appeal to workers at the same time.

1.1. Free Riding in Teams and Social Incentives

Social incentives typically rely on either peer pressure or social preferences. Peer pressure can be seen as the mechanism by which observing others or being observed by others affects one’s own behavior (e.g., Falk and Ichino, 2006; Mas and Moretti, 2009; Guryan, Kroft, and Notowidigdo, 2009; Corgnet, Hernan-Gonzalez, and Rassenti, 2015a; see Herbst and Mas, 2015 for a review), whereas social preferences are defined as a person’s inclination to care about others’ payoff in addition to one’s own (e.g., Fehr and Schmidt, 1999; Bolton and Ockenfels, 2000; Charness and Rabin, 2002; Fehr and Fischbacher, 2002). Either type of social incentive can influence individual behavior.

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\(^2\) Holmström comments on the difficulty of obtaining reliable information about production in his first job at Ahlström (a Finnish company) after graduation: “The integrity of the data therefore seemed questionable for technical as well as strategic reasons.” (p. 414).
Specifically, at a theoretical level, social incentives have been shown to help mitigate shirking in teams. Rotemberg (1994) and Dur and Sol (2010) suggested that the presence of altruistic motives tend to reduce shirking in teams. Because altruistic workers care about their partners’ payoffs, they refrain from shirking that they expect would hurt others’ welfare. It follows that triggering prosocial motives might be an effective solution to shirking in teams. Growing evidence suggests that prosocial concerns indeed foster cooperation (e.g., Fehr and Fischbacher, 2002; Carpenter and Seki, 2011; Chaudhuri, 2011). A challenge is to promote such prosocial concerns in organizations in which shirking is pervasive.

A mechanism that is often used to induce prosocial concerns toward team members is group identity (Akerlof and Kranton, 2000, 2005). A series of experimental papers have shown that inducing group identity could trigger prosocial and cooperative behavior (e.g., Goette, Huffman, and Meier, 2006; Charness, Rigotti, and Rustichini, 2007; Charness, Cobo-Reyes, and Jiménez, 2014). These findings corroborate the results of previous research in social psychology showing that inducing a “minimal group identity” (e.g., by grouping people according to self-reported preferences on paintings; e.g., Tajfel et al., 1971; Tajfel et al., 1979; Rabbie, Schot, and Visser, 1989; Mummendey et al., 1992; Yamagishi, Jin, and Kiyonary, 1999) could promote group cooperation. Dugar and Shahriar (2012) have also shown that group identity fostered cooperation whether it was induced using a “minimal group identity” paradigm or via real existing group identities. We thus expect organizational policies and practices that increase group identity to alleviate shirking in teams. As examples, such practices might include team-building exercises (e.g., Charness, Cobo-Reyes, and Jiménez, 2014) or communication (e.g., Chen and Li, 2009; Gioia, 2017). Because social interactions can foster group identity, they might foster altruism among group members and thus facilitate cooperation (Dur and Sol, 2010).

Social interaction can be seen as a distinctive feature of organizations versus markets (Ramalingam and Rauh, 2014). According to Ramalingam and Rauh (2014), social interactions can foster the internalization of work ethics that proscribe shirking.\(^3\) In particular, we focus on the impact of granting workers’ access to a peer chat platform as a mechanism fostering social interaction (Dawes, 1991; Chen and Li, 2009) and thus boosting team production. Communication has been found to have a large positive effect on cooperation in social dilemmas, especially in larger groups (see Sally, 1995; Balliet, 2010 for reviews). Additionally, communication fosters group identity and commitment (Kerr and Kaufman-Gilliland, 1994) as well as the development of social norms (Bicchieri, 2002).

\(^3\) This argument also relates to the study of norms in Kandel and Lazear (1992).
In addition, a vast literature on public goods games (see Ledyard, 1995; Zelmer, 2003 for an overview) has shown that the introduction of chat can increase cooperation (see e.g., Ostrom and Walker, 1991; Palfrey and Rosenthal, 1991; Ostrom, Walker, and Gardner, 1992; Davis and Holt, 1993; Gardner, Ostrom, and Walker, 1994; Sally, 1995; Bohnet and Frey, 1999; Bochet, Putterman, and Page, 2006; Bochet and Putterman, 2009). Our first conjecture is stated as follows and is formally derived in Appendix A following the models of Rotemberg (1994) and Dur and Sol (2010).

**Conjecture 1 (Chat).** Teams endowed with peer chat will exhibit higher production levels and less shirking than teams not endowed with peer chat.

Another mechanism that can foster prosocial behavior in teams is peer pressure.⁴ Peer pressure models (e.g., Kandel and Lazear, 1992; Barron and Gjerde, 1997) incorporate feelings of guilt or shame that emerge when an individual exerts less effort than team members. These feelings make shirking psychologically costly. Evidence has accumulated showing that being watched by one’s coworkers effectively deters shirking in teams by increasing the amount of shame an individual experiences. Mas and Moretti (2009) collected data on supermarket cashiers and reported positive peer effects on the number of items scanned. These positive effects emerged when cashiers were observed by highly productive workers, but not when they were observed by similarly productive workers. This suggests that the feelings of shame that emerge when others deem an individual a low producer are especially relevant in understanding peer effects. Mas and Moretti (2009) refers to mechanisms based on shameful feelings as social pressure. They emphasize that the effectiveness of social pressure in reducing shirking hinges upon people’s desire to be seen as prosocial, and thus their susceptibility to shame. This mechanism has been modeled by Kandel and Lazear (1992) as well as Bénabou and Tirole (2006) and further validated by the experimental tests in Corgnet, Hernan-Gonzalez, and Rassenti (2015a) and Corgnet, Hernan-Gonzalez, and Mateo (2019).

In addition, experimental works have reported a positive effect of being watched by others on prosocial behavior (e.g., Hoffman, McCabe, and Smith, 1996; Burnham and Hare, 2007; Andreoni and Bernheim, 2009). This effect is also stronger when more people are watching (Diener, 1980; Reyniers and Bhalla, 2013), so an organization in which more workers can monitor each other is expected to

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⁴ We abstract away from the possibility of monetary punishments toward free riders (e.g., Fehr and Gächter, 2000; Carpenter 2007a, 2007b; Nikiforakis, 2008).
outperform one in which only a few workers can monitor. Our second conjecture is stated as follows and is formally derived in Appendix A building on the model of Kandel and Lazear (1992).

**Conjecture 2 (Monitoring).** Teams endowed with peer monitoring will exhibit higher production levels and less shirking than teams not endowed with peer monitoring.

In our model in Appendix A, we assume that the effects of peer chat and peer monitoring on work effort are additive. Our model posits that organizations that use both mechanisms will outperform those that only make use of either one. This means we do not consider cases, for example, in which the effect of peer monitoring is either magnified or weakened by the presence of peer chat.

1.2. Incentives and Work Satisfaction

Our model implies that both peer monitoring and peer chat should perform well in reducing shirking in teams. The multiplicity of solutions to the shirking problem testifies to the richness of the theory, but it also puts the practitioner in the delicate situation of choosing among solutions that appear to be equally effective. How should the practitioner make this choice? Our aim is to show that organizational systems that produce similar incentive effects might, however, generate strikingly different levels of work satisfaction. Dissatisfied workers, in turn, might well be less productive or even less likely to remain in the organization.

In contract theory, the distinction between incentive effects and work satisfaction is formalized by the incentive compatibility and participation constraints (see e.g., Laffont and Martimort, 2002; Bolton and Dewatripont, 2005). Incentive compatibility constraints measure the extent to which a compensation contract fosters work effort, whereas participation constraints assess a worker’s satisfaction (measured in utility terms) compared to available alternatives.

Despite these two features of any given work arrangement, practitioners may be tempted to focus on incentive effects and downplay workers’ satisfaction because workers may find the costs of leaving their job prohibitive in the short-term. Even when workers cannot credibly leave the company immediately, however, it is crucial for managers to take into account workers’ well-being on the job (Danna & Griffin, 1999). Making sure workers are satisfied will limit workers’ resistance to organizational changes, thus tempering the counter-productive organizational behaviors which might be triggered by the new system (Niehoff and Moorman, 1993). Additionally, dissatisfied workers will exhibit low levels of motivation, thus reducing their inclination to exert effort in the absence of explicit incentives (see Frey, 1997; Fehr and Falk, 2002; Gneezy, Meier, and Rey-Biel, 2011). As the managers interviewed in Bewley’s study
(1995, p. 252) made clear: “Workers have so many opportunities to take advantage of employers that it is not wise to depend on coercion and financial incentives alone as motivators.” Because employment contracts are inherently incomplete, it is impossible to provide explicit incentives for all dimensions of a job (Holmström and Milgrom, 1991, 1994; Itoh, 1991; Maskin and Tirole, 1999). Ultimately, managers have to rely on employees’ intrinsic motivation (Deci, 1971; Frey, 1997; Deci and Ryan, 2000; Fehr and Falk, 2002) to sustain workers’ performance. Intrinsic motivation refers to a person’s inherent enjoyment of the job or task at hand (Ryan, 1982) and thus closely relates to job satisfaction (e.g., Spector, 1985; Kinicki et al., 2002).

Self-determination theory, which has been formalized in Economics by Bénabou and Tirole (2002, 2003), suggests that the three main drivers of work satisfaction are competence, autonomy and relatedness (Deci and Ryan, 1985, 2000; Ryan and Deci, 2000; Deci et al., 2001; Gagne and Deci, 2005). Thus, workers will report high levels of satisfaction on the job when they feel good at what they are doing (competence), feel they are doing it out of their own volition (autonomy), and feel socially connected to others (relatedness).

In line with self-determination theory, workplace surveillance mechanisms tend to reduce the perceived autonomy of workers, thus lowering job satisfaction (e.g., Frey, 1997; Ambrose and Alder, 2000; Stanton, 2000a,b; Ariss, 2002; Alder, Noel and Ambrose, 2006; Falk and Kosfeld, 2006). Commentators report that employees “feel degraded, stressed, and dehumanized” by a surveillance system (Ariss, 2002: 555), which “has a detrimental effect on employee morale, increases worker stress, and engenders negative job attitudes” (Alder, Noel and Ambrose, 2006, p. 895). Despite its strong incentive effects (see Mas and Moretti, 2009; Corgnet, Hernan-Gonzalez, and Rassenti, 2015a; Herbst and Mas, 2015), then, monitoring might reduce autonomy and temper workers’ satisfaction.

By contrast, peer chat is unlikely to threaten autonomy, as workers will be free to initiate or stop any conversation. In addition, peer chat will increase the perception of relatedness of workers, thus fostering work satisfaction. This is consistent with the documented benefits of participative decision-making allowing workers to provide input about organizational policies (Alge, 2001)—and also with the psychological axiom that people generally enjoy social interaction (Allport, 1924). This leads us to the following conjecture.
Conjecture 3 (Work satisfaction).

i) The effects of an organizational shirking solution on workers’ job satisfaction will diverge from the solution's incentive effects.

ii) Organizational shirking solutions involving chat will lead to greater worker satisfaction than solutions without chat.

iii) The effect of monitoring on work satisfaction will be mixed because it will increase workers’ incentives to exert effort thus boosting the revenues of the organization while at the same time making the task less enjoyable.

As we show in Appendix A, the effect of peer chat and peer monitoring on work satisfaction should not affect the magnitude of the incentive effects captured in Conjectures 1 and 2. This explains why workers’ satisfaction in an organizational solution are likely to differ from the magnitude of incentives provided by such system.

Work satisfaction should, however, ultimately foster work motivation in the long run (Westover et al., 2010) and affect work behavior in job dimensions that are not contractually incentivized. For example, a large literature has shown that satisfied workers are less likely to engage in counterproductive work behaviors that are inappropriate and harmful to the firm and their coworkers (Dalal, 2005). A satisfied workforce is also more likely to engage in organizational citizenship behavior (Niehoff and Moorman, 1993), thus going beyond the contractual definition of their job to help their coworkers and add value to the company. In addition, dissatisfied employees will be more likely to leave the company, thus generating additional turnover costs. Because workers who want to leave as a result of a change in the organizational setup might not be able to do so immediately, the negative impact of organizational changes might only be seen in the longer-run. This is why managers might be tempted to favor organizational changes that produce strong incentive effects in the short run at the risk of generating long-term costs. Our results suggest they might have to reconsider carefully the pros and cons of this approach.

1.3. Experimental Tests and Findings

To test our conjectures, we use a laboratory workplace in which workers undertake a real-effort task, while having access to the Internet for leisure purposes (see Corgnet, Hernan-Gonzalez, and Schniter, 2015). We conduct six main treatments in a 2×3 factorial between-subject design, in which we manipulate social interactions and monitoring among workers. Social interaction is manipulated at one of two levels including treatments in which workers have access to a chat platform to communicate with
other team members and treatments in which they do not have access to such platform. The monitoring dimension is manipulated at three levels. Either all workers can monitor each other’s activities, only one of the workers can monitor others, or no workers can monitor.

In our baseline treatment in which neither chat nor monitoring is present, we observe substantial shirking. Workers spend about 30% of their time on the Internet instead of working on the task. In line with Conjectures 1 and 2, shirking is substantially reduced in any of the treatments in which we introduce either peer chat, monitoring or both. In these treatments, workers spend about 10% more time on the work task and produce about 40% more than in the baseline.

To test Conjecture 3, we design an additional experiment that aims to assess participants’ willingness to work in a given organizational system. Unlike our first study, in which the organizational system was set exogenously by the experimenter, participants in this study could state their preferences for each of the six systems previously studied. The system that receives the highest average rating across team members is then implemented. Alternatively, we could have used a survey to elicit participants’ work satisfaction (e.g., Spector, 1985; Deci and Ryan, 2000) in each of the six systems. However, we wanted to employ a research design in which participants would have an incentive to truthfully reveal their preference for each organizational system. Truth-telling is encouraged in our case because workers who do not reveal their true preferences could end up working in an organizational system they dislike.

In line with Conjecture 3, we find that organizations involving peer chat without monitoring tend to be more popular than those involving monitoring without chat. In addition, organizational systems involving only monitoring rate significantly lower than the baseline whereas those involving only peer chat rate directionally (but not significantly) higher than the baseline. Unexpectedly, organizational systems involving monitoring and peer chat together are as popular as those systems involving peer chat without monitoring. This implies that workers’ negative reaction toward peer monitoring is fully offset by the presence of peer chat. This interaction effect was not part of Conjecture 3 and of our model (see Appendix A). This suggests that the negative effect of monitoring systems in terms of work satisfaction might be alleviated by fostering social interaction between workers. Workers thus seem less reluctant to be monitored by others if they can communicate with them—possibly because they can then voice their concerns regarding what could be perceived as abusive monitoring.
2. Design

The current research includes two studies that use an interactive, virtual environment to test our conjectures in a tightly-controlled fashion. Interdependent individuals perform an analytical task that also allows them to check the internet, replicating many features of a real-world work environment.

Study 1: Solutions to Shirking in Teams (Conjectures 1 & 2)

Design. To investigate the first two conjectures, we use a 2×3 between-subject factorial design in which the chat dimension is either present or absent and monitoring is either absent, given to one team member, or given to all team members (see Table 1). Each of the six treatments involves 60 different participants.

<table>
<thead>
<tr>
<th>Organizational systems</th>
<th>Chat availability</th>
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<td>Absent</td>
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<tr>
<td>Chat availability</td>
<td>Present</td>
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<tr>
<td>Absent</td>
<td>No Chat-No Monitor (Baseline)</td>
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<tr>
<td>Chat</td>
<td>Chat-No Monitor</td>
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<td>One worker</td>
<td>No Chat-One Monitor</td>
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<td>All workers</td>
<td>No Chat-All Monitor</td>
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<td>Chat-One Monitor</td>
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<td>Chat-All Monitor</td>
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The work task. The instructions indicate that participants can choose among several activities, including the work task. Adapted from previous research using summation tasks (e.g., Eriksson, Poulsen, and Villeval, 2009), the work task is a particularly long and laborious task intended to resemble the monotony that can accompany organizational life and prompt shirking at work. The task requires participants to sum up tables of 36 numbers without using a pen, scratch paper, or calculator (see Figure 1). Each table has six rows and six columns of randomly-generated integers between zero and ten. Before providing the grand total in the bottom-right cell, participants have to provide a separate subtotal for all of the 12 rows and columns. Calculating these subtotals do not directly generate earnings but could help participants compute the grand total, which generates a 40-cent profit if correct and a 20-cent penalty if incorrect. After completing a table, participants learn whether their answers are correct and how much money they earn. At the end of each period, participants learn the total amount of money generated by all ten participants’ efforts on the work task.

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</table>

Figure 1. Work Task
At any point during the experiment, all participants can switch from the work task to internet browsing. Depending on their experimental treatment, they might also have the ability to monitor or communicate with their peers. Participants can spend as much or as little time as they want on the various activities, each of which is undertaken on a separate screen. To switch activities, participants simply choose the corresponding option from a drop-down menu at the bottom-right of their screens.

**Internet.** If participants choose the internet, the work task window is replaced by an internet window (embedded in the software; see Figure 2). Within the bounds of university policy, participants can use the internet however they like, including email. Their confidentiality is assured and maintained, but the software tracks the exact amount of time spent on each activity. Although participants cannot complete the work task while browsing the internet, switching is quick and easy.

![Figure 2. Embedded Internet Screen](image)

In the ‘No Monitor’ treatments, participants cannot monitor or be monitored. In the ‘All Monitor’ treatments, all participants can choose to watch the activities of their peers. In the ‘One Monitor’ treatments, only one of the ten participants is given the ability to watch everyone else, and everyone else is aware of this ability. If participants have and select the monitoring option from the drop-down menu, they are directed to a separate window where they can choose whom to monitor (anywhere from one to all other participants) and to actually perform the monitoring. For each selected participant, a column in a table lists their activities (e.g., switched to the internet, provided a subtotal), their current earnings, and their percentage contribution to the team total. As the current research is concerned with reactions to monitoring, we do not focus on the choice to monitor but rather the experience of being monitored. Participants who are being monitored see a box indicating that “[Experiment ID of the participant] is watching you” (see Figure 3).
Peer chat. In the ‘Chat’ treatments, participants can choose to exchange instant messages with their teammates. This virtual form of communication is chosen to maintain anonymity, and because it represents a simple form of communication, bereft of potential social confounds (Gunia et al., 2012). Thus, participants who choose to communicate by selecting that option from the drop-down menu enter a chat room in which they can send a message to one or more people. Participants with whom others want to communicate see a pop-up window displaying the sender’s experiment ID and message content (see Figure 4).

Figure 4. Peer Chat

Procedures. The experiment is conducted using the Virtual Organizations software proprietarily developed by CYDeveloper LLC for the authors. The software facilitates a multi-party team task, controlled centrally by an experimenter. Upon arrival at the lab, participants are directed to private computer terminals and asked to read a set of computerized instructions. Participants have exactly 20 minutes to read the instructions, with a timer displayed on a large screen at the front of the lab. The instructions indicate that they are one of ten members of a virtual team, which undertakes a 1-hour and 40-minute task, broken up into 20-minute periods. Three minutes before the end of the instruction period, the experimenter announces the time remaining and hands out a printed summary of the instructions. At the end of the instruction round, the experimenter closes the instructions and launches the experiment from the server.

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5 The full set of instruction is available here: https://tinyurl.com/utryu2v.
Participants are 360 undergraduates (48.95% male; average age 20.12) enrolled in a subject pool at a Western U.S. university. Specifically, we conducted six sessions of ten participants for each of the six treatments. Based on previous findings using the same real-effort task (see Corgnet, Hernan-Gonzalez, and Rassenti, 2015a,b; Corgnet, Hernan-Gonzalez, and Schniter, 2015), we calculated that recruiting 60 participants for each treatment would ensure a power of 80% to detect a 20% increase in workers’ production with respect to the baseline.

Participants responded to an email offering $7 plus an unspecified amount of bonus money for participation in an experiment lasting 2.5 hours. On average, participants earned a total of $26.55, and the experiment lasted for 2.25 hours.

**Study 2: Workers’ Satisfaction (Conjecture 3)**

Study 2 uses the same task as Study 1 to investigate the same six organizational systems. It substantially extends Study 1, however, by focusing on participants’ subjective reactions to these systems (following Zweig and Webster, 2002) and by allowing them to actually experience the system that elicits the most favorable reactions. By allowing participants to choose and experience a system three times, participants are able to fine-tune their reactions if necessary.\(^6\)

**Design.** The design differs from Study 1 because participants rate each of the six organizational systems (e.g., ‘Chat-No Monitor’) before each period, and the system that receives the highest average rating across team members is announced and implemented. If multiple systems receive the highest rating (which only happened once, between two systems) a system is randomly selected. Participants are presented with a summary of the six organizational systems, and they answer the following, general question: “How much do you want to work in each of the following organizations?” (1 = not at all, 7 = very much so; see Figure 5).

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\(^6\) The full set of instruction is available here: [https://tinyurl.com/s3xpvjc](https://tinyurl.com/s3xpvjc).
Figure 5. Organization Ratings Screen

Our experimental design thus provides an incentivized elicitation of workers’ preferences for the different organizational systems. In our setup, increasing one’s own rating for an organizational system increases’ the chances that this system will be implemented. Workers who do not reveal their true preference regarding a given system might end up working in a system they dislike in the next period (see Smith, 1982). Our approach thus differs from standard survey techniques used to elicit work satisfaction (e.g., Spector, 1985) or work motivation (Deci and Ryan, 2000). Because Study 2 aims to measure work satisfaction in lieu of incentive effects, we are not interested in the production and shirking data associated with this study. This is why we use three periods instead of six and shorten the length of each period to 10 minutes. We do not use production data in Study 2 to assess the incentive effect of a given organizational system because of selection effects. Groups of workers who rate organizational systems differently are likely to differ in terms of relevant individual characteristics, such as, for example, ability on the task.

**Procedures.** Fifty undergraduate students (48% male; average age 19.71, SD = 1.69) from the same participant pool as Study 1, but who had not participated in Study 1, participated in Study 2. They responded to an email offering $7 plus an unspecified amount of bonus money for participation in an experiment lasting 1.5 hours. Five separate sessions of ten workers were conducted; on average, participants earned a total of $16.25.

3. Results

3.1. Study 1: Shirking in Teams

In line with Conjectures 1 and 2, all the organizational systems involving either chat, monitoring, or both achieved a higher level of production than the baseline organizational system in which neither chat

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7 The data that support the findings of this study are available from the corresponding author upon request.
nor monitoring was present (see left panel of Figure 6 and Table B1 in Appendix B). On average, a worker involved in any of the organizational systems endowed with either chat, monitoring or both produced 35.27% more ($7.67, SD = 5.18) than the baseline organizational system (Cohen’s $d = 0.40$). Another measure of workers’ effort is the amount of time they spent online. Indeed, for the work task used in the current setup, browsing the internet does not have any positive effect on workers’ productivity (see Corugnet, Hernan-Gonzalez, and Schniter, 2015). Browsing the web simply distracts the worker, thus reducing his or her productivity—a set of activities often called cyberloafing (Henle and Blanchard, 2008), which occurs when an employee uses the internet during the work period for non-work purposes (Lim, 2002; Wagner et al., 2012).

The time participants spent online (see right panel of Figure 6 and Table B1 in Appendix B) corresponds to 12.35% (SD = 23.29%) of the total available time in any of the organizational systems involving either chat, monitoring or both, versus 28.52% (SD = 34.80%) in the baseline (Cohen’s $d = 0.81$). The comparison of internet usage across organizational systems should, however, take into account the fact that the six systems differ in the number of activities available to workers. It follows that monitoring or chatting activities could potentially be used as substitutes for internet usage, thus mechanically lowering the time spent online in any of the organizational systems endowed with either chat, monitoring or both. To alleviate this concern, we also use the time spent on the work task screen as a measure of workers’ effort. We find that, despite having more options available, workers dedicated more time to the work task (82.38%) in the organizational systems involving chat, monitoring, or both, as compared to the baseline (71.48%) (Cohen’s $d = 0.52$) (see Figure 6, right panel). We do not observe differences in the time spent chatting across the three treatments involving chat (p-values for all three pairwise comparisons are greater than 0.1 using t-tests as in Table B2 in Appendix B). In the same vein, we do not see statistical differences in the time spent monitoring between the two ‘All Monitor’ treatments and between the two ‘One Monitor’ treatments (p-values for the two pairwise comparisons are greater than 0.1 using t-tests as in Table B2).

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8 Thus, using the internet over lunch, using it for work purposes, or using an offline application would not qualify as cyberloafing. What would qualify is any personal activity, conducted during the work period (e.g., Web browsing, email, social media; Kallman, 1993). Obvious in theory, these distinctions can blur in practice, as employees may, for example, encounter irrelevant Websites during legitimate searches or open personal emails to retrieve work-related information. Occasionally, they may also “abuse” the internet to cope with stress or to stimulate their creativity (Henle and Blanchard, 2008). While recognizing the inevitable “grey area” between use and abuse, the current research makes a rigid distinction by focusing on clear cases of abuse, which interrupt work (Jett and George, 2003) and are thus counterproductive (Henle and Blanchard, 2008).
Our analyses in Table 2 show the statistical significance of differences in production, internet usage, and time on the task between the baseline and the other organizational systems involving chat, monitoring, or both. Even though Table 2 reports the results of panel regression analyses at the period level, similar results are obtained using standard parametric and non-parametric tests that compare total workers’ production, internet usage, and time on the task across treatments (see Table B2 in Appendix B).\(^9\)

In Table 2, the coefficients associated with each organizational system dummy in regressions [1] and [5] are positive and significant except for ‘No Chat-One Monitor Dummy’ which is positive yet not significant, p-values = 0.234 and 0.133). The coefficients associated with each organizational system dummy in regression [3] are negative and significant except for ‘No Chat-One Monitor Dummy’ which is negative yet not significant, p-value = 0.116. This is consistent with our model (see Appendix A) and the work of Kandel and Lazear (1992), according to which a lower number of monitors would tend to reduce the extent of peer pressure, thus reducing the corresponding positive effect on workers’ effort.

In regressions [2], [4], and [6], we also assess the dynamics of production, internet usage, and time on the task across treatments. For all treatments involving chat, the interaction coefficients between organizational dummies and the number of periods (variable ‘period’) are positive and significant for production and time on the task (regressions [2] and [6]), while being negative and significant for internet usage (regression [4]). Thus, the positive impact of chat (and chat with monitoring) on workers’ effort

\(^9\) Similar results are also obtained when using a non-parametric test with clusters at the session level (Somers’ $d$, Somers, 1962).
tends to increase over time. This might be the case because workers need some time to get familiar with the chat feature. Alternatively, and as we argue in the conjecture section, building the necessary team identity to trigger workers’ prosocial concerns may require time.

In Table B3 (see Appendix B), we focus on the content of the conversations in the organizational systems involving chat. Two of the authors independently read and inductively extracted categories, resolving disagreements through discussion. Two graduate student coders who were unaware of the hypotheses then independently assigned each of 354 messages to one of the 13 final categories. The coders agreed on the categorization for 69.50% of the messages, for an acceptable Cohen’s Kappa of 0.65. A sizable proportion of messages (17.50%) were social in nature (e.g., introductions or jokes), thus possibly triggering team identity, as in standard greeting procedures used in the literature (e.g., Chen and Li, 2009; Gioia, 2017). In addition, a large proportion of messages (61.60%) contained content that could be considered normative (e.g., asking or informing about performance, encouraging performance). This suggests that chat might also have induced norms of cooperation (as in Kandel and Lazear, 1992), thus fostering workers’ effort. These norms of cooperation might be especially salient when workers have had the chance to get to know each other via chat. In the end, chat would tend to promote prosocial concerns either by enhancing altruism toward coworkers who share a common team identity (as in Dur and Sol, 2010) or by promoting norms of cooperation across workers who do not necessarily feel altruistic toward each other (Kandel and Lazear, 1992). These two possible mechanisms are likely both present at the same time, and our setup does not seek to isolate them.

By contrast with chat, the positive effect of monitoring on workers’ production (regression [2]) does not increase over time (the coefficients for ‘No Chat-One Monitor Dum.×Period’ and ‘No Chat-All Monitor Dum.×Period’ are not significant). In addition, the difference between the coefficients ‘Chat-All Monitor’ and ‘No Chat-All Monitor×period’ is significant (p-value = 0.001), as is the difference between the coefficients ‘Chat-One Monitor×period’ and ‘No Chat-One Monitor×period’ (p-value = 0.054). This means production is more likely to increase over time when chat is present than when it is absent, given a particular level of monitoring (‘One Monitor’ or ‘All Monitor’). This might occur because the effect of chat relies partly on building team identity or fostering cooperative norms, both of which may require time.

Regarding internet usage or time on the task, we observe an effect that increases over time for the treatment ‘No Chat-All Monitor Dummy’, which could be due to workers’ learning how to use the
monitoring features over time. This effect is not significant for the treatment ‘No Chat-One Monitor Dummy’, however.

**TABLE 2—LINEAR PANEL REGRESSION WITH RANDOM EFFECTS FOR WORKERS’ PRODUCTION (IN CENTS), INTERNET USAGE AND TIME ON THE TASK (IN SECOND)**

<table>
<thead>
<tr>
<th>Dependent variable</th>
<th>Production (in cents)</th>
<th>Internet Usage (in seconds)</th>
<th>Time on the Task (in seconds)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intercept</td>
<td>59.733***</td>
<td>85.667***</td>
<td>238.892***</td>
</tr>
<tr>
<td></td>
<td>(10.655)</td>
<td>(6.586)</td>
<td>(52.020)</td>
</tr>
<tr>
<td>Chat-No Monitor Dummy</td>
<td>55.933**</td>
<td>18.433</td>
<td>-188.746***</td>
</tr>
<tr>
<td></td>
<td>(25.606)</td>
<td>(14.222)</td>
<td>(60.877)</td>
</tr>
<tr>
<td>Chat-One Monitor Dummy</td>
<td>38.667**</td>
<td>-6.733</td>
<td>-255.424***</td>
</tr>
<tr>
<td>Chat-All Monitor Dummy</td>
<td>30.800**</td>
<td>-13.400</td>
<td>-232.191***</td>
</tr>
<tr>
<td></td>
<td>(14.342)</td>
<td>(16.188)</td>
<td>(55.780)</td>
</tr>
<tr>
<td>No Chat-One Monitor Dummy</td>
<td>22.733</td>
<td>2.933</td>
<td>-108.941</td>
</tr>
<tr>
<td></td>
<td>(19.117)</td>
<td>(20.513)</td>
<td>(69.360)</td>
</tr>
<tr>
<td>No Chat-All Monitor Dummy</td>
<td>51.600***</td>
<td>42.900***</td>
<td>-184.701***</td>
</tr>
<tr>
<td></td>
<td>(20.456)</td>
<td>(15.167)</td>
<td>(58.092)</td>
</tr>
<tr>
<td>Period</td>
<td>17.911***</td>
<td>9.267***</td>
<td>34.447***</td>
</tr>
<tr>
<td></td>
<td>(1.713)</td>
<td>(3.520)</td>
<td>(4.999)</td>
</tr>
<tr>
<td></td>
<td>(5.098)</td>
<td>(10.612)</td>
<td>(10.549)</td>
</tr>
<tr>
<td>Chat-All Monitor Dum.× Period</td>
<td>-14.733***</td>
<td>-55.959***</td>
<td>71.149***</td>
</tr>
<tr>
<td>No Chat-All Monitor Dum.× Period</td>
<td>-2.900</td>
<td>-34.135***</td>
<td>29.369**</td>
</tr>
</tbody>
</table>

**P-values (coefficient comparisons)**

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
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<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>0.532 [0.675]</td>
<td>0.056 [0.061]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
<td>0.335 [0.218]</td>
</tr>
<tr>
<td>Observations (organizations)</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
<td>1800</td>
</tr>
<tr>
<td>Prob &gt; χ²</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>R²</td>
<td>0.063</td>
<td>0.068</td>
<td>0.098</td>
<td>0.105</td>
<td>0.042</td>
<td>0.054</td>
<td>0.054</td>
<td>0.054</td>
<td>0.054</td>
<td>0.054</td>
<td>0.054</td>
<td>0.054</td>
<td>0.054</td>
</tr>
</tbody>
</table>

**Notes:** Estimation output using robust standard errors clustered at the organization level (in parentheses). Similar results are obtained using bootstrapping techniques for standard errors.

*** Significant at the 1 percent level, ** Significant at the 5 percent level, * Significant at the 10 percent level.

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10 Note that when conducting a linear panel regression (as in Table 2) with the time spent monitoring by workers as a function of the number of periods in the ‘All Monitor’ treatments we do not observe a significant increase of the amount of time spent watching over time. So the dynamics of the monitoring activity would not seem to explain this pattern.
In Table 2 (lower panel), the pairwise comparisons of coefficients in regressions [1] and [5] stress that the organizational systems involving chat, monitoring, or both do not significantly differ in terms of production and time dedicated to the task. Differences in coefficients across treatments regarding internet usage (regression [3]) might thus be due to the number of different activities available across treatments.

This absence of significant differences in workers’ production levels across organizational systems endowed with chat, monitoring, or both are not inconsistent with Conjectures 1 and 2, which only specify a significant effect with respect to the baseline. However, our model (Appendix A) suggests that the effect of chat and monitoring should be additive such that the ‘Chat-All Monitor’ (‘Chat-One Monitor’) treatment should outperform ‘No Chat-All Monitor’, ‘Chat-No Monitor’ or ‘Chat-One Monitor’ (‘No Chat-One Monitor’ or ‘Chat-No Monitor’). A possible explanation for this lack of statistical differences could be a ceiling effect, by which the level of performance achieved using only chat or monitoring is close to the maximum level of performance of a team. To assess the validity of this claim, we use the data on workers’ performance on the same task under individual incentives and in the absence of either chat or monitoring (see Corgnet, Hernan-Gonzalez, and Rassenti, 2015a). In line with the ceiling effect argument, we report that when comparing the performance of workers under individual incentives with the five treatments of the current study involving either monitoring, chat, or both, we obtain p-values that are greater than 0.1 in all cases (p-values = 0.12, 0.90, 0.95, 0.42 and 0.12) using panel regressions similar to the ones in Table 1.11 The absence of significant differences between each of these organizational systems and the case of individual incentives suggests that workers’ performance is already at a high (possibly maximum) level when chat or monitoring alone is present.

In line with Conjectures 1 and 2, we show that the organizational solutions to shirking in teams are effective. Although no significant differences are observed across these solutions in terms of workers’ performance and effort levels, Conjecture 3 suggests that organizational systems involving chat would tend to be especially valued by workers. By contrast, organizational solutions involving monitoring might not be as popular among workers. We test these claims in Study 2.

3.2. Study 2: Organizational Systems Ratings

In Study 2, workers rated each organizational system on a 1 to 7 Likert-type scale on three occasions. In Figure 7, we display the average ratings across the ten organizational members for each period. The ordering of organizational systems is the same whether the first or last rating is used, and no statistically

---

11 The comparison of individual incentives with the baseline treatment yields a p-value < 0.001.
significant differences are observed between the two ratings except for ‘Chat-One Monitor’ and ‘Chat-All Monitor’, which became less popular over time (see Table B4 in Appendix B). Even though the popularity of ‘Chat-All Monitor’ went down, it was still selected in three out of the five teams in the last period. All teams tried this organizational system in the first period, but two decided to switch to either the baseline organizational system or ‘Chat-One Monitor’. One explanation for this reduced popularity over time is that chatting requires time to effectively boost workers’ production and thus increase workers’ revenues, as is shown in our dynamic analysis of production in regression [2] of Table 2 in Study 1.12

Regardless of the dynamics of ratings, we find that, in line with Conjecture 3, adding chat to a given organizational system tends to increase its popularity. This effect is statistically significant when we consider the first rating, which is not influenced by workers’ experience with a given organizational system (see the statistical analyses in Tables B5, B6 and B7). That is, the treatments ‘Chat-No Monitor’ (‘Chat-One Monitor’) ['Chat-All Monitor’] led to significantly higher first ratings than ‘No Chat-No Monitor’ (‘No Chat-One Monitor’) ['No Chat-All Monitor’]. These findings also hold when considering the last rating and average ratings, except that the difference between ‘Chat-No Monitor’ and ‘No Chat-No Monitor’ is not statistically significant in that case. This follows from the fact that workers reduced their ratings for the “Chat-No Monitor” system over time.

By contrast, adding monitoring to an organizational system without monitoring does not increase workers’ ratings. Actually, the effect is systematically negative and, in most cases, statistically significant. That is, the treatment ‘No Chat-No Monitor’ (‘Chat-No Monitor’) led to higher ratings than ‘No Chat-One Monitor’ and ‘No Chat-All Monitor’ (‘Chat-One Monitor’ and ‘Chat-All Monitor’). These differences are always significant except for the comparison between ‘Chat-All Monitor’ and ‘Chat-No Monitor’ for the first and average ratings, and for the comparison between ‘No Chat-All Monitor’ and ‘No Chat-No Monitor’ for the first ratings. Thus, the negative effect of monitoring on workers’ enjoyment of the task appears to more than offset the positive effect of monitoring associated with higher production levels and higher workers’ revenues. All in all, workers are less willing to join a team when monitoring is present.

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12 In Study 2, the experiment was substantially shorter than in Study 1 because there were two periods less and each period was half shorter.
Figure 7. Average Ratings (with 95% confidence intervals bars)
Across Organizational Systems for First, Second and Last Periods

Form a practitioner standpoint, it thus follows that, among the organizational solutions to shirking under investigation, promoting chat among peers may be preferred. Indeed, none of the organizational solutions in which chat was absent generated higher ratings than the baseline. This means that organizational solutions relying only on monitoring will reduce workers’ satisfaction despite leading to higher organizational performance and higher workers’ revenues. In Figure 8, we show that workers’ satisfaction ratings are not aligned with organizational performance. In particular, the organizational system that received the highest ratings (‘Chat-All Monitor’) was ranked fourth out of the five solutions in terms of organizational performance (using performance data from Study 1). By contrast, the organizational system involving peer monitoring but no chat (‘No Chat-All Monitor’) received the second-lowest ratings while leading to the second-highest organizational performance. The solutions to shirking that should probably be favored are in the top right corner (shaded area) in Figure 8. These are organizational systems that produce ratings at least as high as the baseline (Study 2), while increasing workers’ performance substantially (Study 1).
Figure 8. Average Production (Study 1) and Average Ratings (Study 2) Across Organizational Systems

It is also noteworthy that organizational systems involving both monitoring and chat produced ratings similar to those involving chat only. This suggests the negative effect of monitoring is offset by the presence of chat. Although this result was not predicted by our model, it might be understood a posteriori as a positive interaction effect between chat and monitoring on workers’ intrinsic motivation (Deci, 1971; Frey, 1997; Deci and Ryan, 2000; Fehr and Falk, 2002). Granting workers the possibility to voice their concerns about intrusive monitoring or otherwise build a positive social relationship might offset the excessive control (lack of autonomy) associated with monitoring (see e.g., Wagner et al., 1997; Corgnet, Hernan-Gonzalez, and McCarter, 2015). Practitioners who are already using monitoring solutions that would be costly to dismantle might thus foster communication between peers about the monitoring system, as a means of restoring workers’ motivation. Concretely, organizations might foster peer communication by promoting user-friendly chat platforms at work and encouraging employees to use them, particularly when communicating about issues of monitoring and performance. The time for such solutions seems ripe, as a multitude of professional instant messaging platforms, conference technologies, chat rooms, blogs, and billboards have become available, many of which can dramatically increase the ease and lower the costs of communication. In addition, enterprise social networks, which are internal private social networks (e.g., Socialcast) that facilitate communication among employees, have boomed in recent years (e.g., Mishra, Walker and Mishra, 2014). Of course, organizations could also encourage peer communication in “old-fashioned” ways like task meetings or open-door policies.
4. Discussion

Shirking in teams is a major incentive-related issue in economics for which many solutions, often based on monitoring technologies, have been proposed. For example, monitoring technologies are a popular solution to curb cyberloafing in firms, which is a modern manifestation of the shirking problem in teams (Blanchard and Henle, 2008). Indeed, a large majority of organizations have implemented systems to monitor their employees’ internet use (Alge, 2001), creating an internet monitoring industry now valued at more than $300 million (Alder, Noel and Ambrose, 2006). The trend seems unlikely to reverse, although the efficacy of these systems remains unclear (Niehoff and Moorman, 1993; Stanton and Weiss, 2000).

This paper highlights both the positive effect of monitoring systems on workers’ performance and their negative impact on workers’ satisfaction. In addition, we show that other organizational solutions fostering team identity and promoting prosocial concerns can achieve the same level of worker performance without putting work satisfaction at risk. In particular, we show that organizational systems promoting peer communication are more popular solutions to Cyberloafing, and possibly to other forms of shirking, than monitoring.

Several organizations have already recognized the potential side effects of monitoring their employees excessively and the need for alternative solutions. A General Motors executive, for example, said: “The company’s philosophy is that the workplace is an environment of mutual trust and respect. This precludes a policy of accessing employee email” (Agarwal and Rodhain, 2002, p. 3). Our research supports this position. Yet, most organizations continue to monitor workers extensively (Alder, Noel and Ambrose, 2006), downplaying the long-term consequences of a dissatisfied and unmotivated workforce. Our work might motivate employers to reconsider these policies, or at least consider them carefully.

5. References

The data that support the findings of this study are available from the corresponding author upon request.


Corgnet, B., Hernán-Gonzalez, R., & Mateo, R. (2019). Rac(g)e Against the Machine?: Social Incentives When Humans Meet Robots. GATE WP.


Appendix A. Theoretical model

We rely on previous social preferences and social pressure models to study the effect of monitoring and peer chat on effort provision (see Kandel and Lazear, 1992; Rotemberg 1994; Fehr and Schmidt, 1999; Rey-Biel, 2008; Bartling and von Siemens, 2010; Dur and Sol, 2010; Englmaier and Wambach, 2010). We derive our hypotheses using the moral-hazard in teams’ model introduced by Holmström (1982). We consider \( n \) workers producing a total output \( f := f(e_1, e_2, \ldots, e_n) \) which depends on each worker’s effort \( e_i \geq 0 \) where \( i \in \{1, \ldots, n\} \). We assume that \( f(\cdot) \) is linear and separable in workers’ efforts, \( f := \sum_{i=1}^{n} a_i e_i \), where \( a_i > 0 \) is the marginal product of effort of worker \( i \). By assuming separability in workers’ effort, our production function allows us to identify each worker’s individual contribution. This is the type of production function we use in our experimental design. The cost of effort is represented by \( C(e_i) \) where \( C'(e_i) \geq 0 \) and \( C''(e_i) \leq 0 \). Each worker in the team is paid according to team incentives thus collecting a share \( \frac{1}{n} \) of total production. The utility function of worker \( i \) is thus:

\[
v_i := \frac{f}{n} - C(e_i)
\]  

**Conjecture 1. (Chat)**

We extend the utility function of worker \( i \) in [1] to account for the effect of the presence of chat and monitoring. Following Dur and Sol (2010), we assume that chat will foster social interaction between workers thus promoting altruistic motives. We capture worker \( i \)’s altruism with a parameter \( \xi_i \geq 0 \). An altruistic person \((\xi_i > 0)\) values other workers’ pay positively so that under peer chat a worker’s utility function becomes:

\[
u_i := v_i + \frac{(n-1)f}{n} \xi_i
\]

We derive our first conjecture by relying upon the fact that peer chat will induce stronger altruistic motives among team partners, in line with the model of Dur and Sol (2010). Our first conjecture abstracts away from participation constraint so that we are going to assume that workers have already accepted to work under certain organizational conditions, in this case peer chat. We thus focus on the incentive effect associated to peer chat looking into the incentive compatibility constraint of workers, which follows directly from workers’ utility maximization. In the case of an altruistic worker, the first order condition (see [3]) is such that an increase in altruistic concerns \((\xi_i)\) which follows from peer chat will lead to an increase in the level of effort exerted by workers given the assumptions on the cost of effort function:

\[
C'(e_i) = \frac{1+(n-1)\xi_i}{n} a_i
\]
This completes our proof of Conjecture 1. In addition to altruism, concerns regarding inequality aversion have been shown to be prevalent (see Charness and Rabin, 1993; Fehr and Schmidt, 1999; Bolton and Ockenfels, 2000). However, in the context of team incentives, all workers are paid the same so that any consideration regarding inequality in payoffs induced by peer chat would not have any effect.

**Conjecture 2. (Monitoring)**

Following Kandel and Lazear (1992) modeling of peer pressure in team production setups, we consider that being observed by other team workers will affect their incentive to exert effort. At the empirical level, Mas and Moretti (2009) and Corgnet, Hernan-Gonzalez, and Rassenti (2015a) have shown that workers exert higher effort when observed by other team members. We can think of a variety of reasons why workers would produce more when being observed by others. The first possibility relates to audience effects à la Andreoni and Bernheim (2009) in which case people wants to be seen as fair. In particular, the authors put forward that people are inclined to split outcomes equally when seen by others as they want to be perceived as egalitarian. In the context of team incentives, all workers are paid the same so that fairness concerns regarding strict pay equality do not apply to our setup. However, workers might still be motivated to work hard because they want to be seen as complying with a social norm of effort and production (Kandel and Lazear, 1992). The work of Corgnet, Hernan-Gonzalez, and Rassenti (2015a) shows that peer pressure in teams is effective for both low and high producers suggesting that being observed by others does not foster a common production norm although it can certainly induce a norm of high effort. One way to model the emergence of a high-effort norm under peer monitoring is to consider that team members feel shame whenever they slack off because this directly hurts others’ payoffs by reducing team production. We thus model social pressure as workers’ willingness not to hurt the payoffs of their team members. We capture this effect with the parameter $\chi_{i,j} \geq 0$ which measures the peer pressure worker $i$ suffers from worker $j$ so that the utility function of a worker who is subject to peer monitoring can be written as:

$$w_i := v_i + \frac{f}{n} \sum_j \chi_{i,j}$$

where $j \in M$ and $M$ stands for the set of workers who observe worker $i$’s performance.

Our peer pressure model is such that a worker who is observed will feel shame and thus value the payoff of other workers positively. Peer pressure thus triggers shame leading team members to behave as if they were altruistic. But, the difference between altruism and social pressure is that altruistic workers [see 3] will exert higher effort when working in a team regardless of whether they are observed or not by their team members. As for Conjecture 1, we derived our conjecture regarding peer monitoring
using the first order conditions of the worker’s utility function. The first order condition below shows that peer monitoring, by enlarging the set of monitors $M$, will make the term $\sum_j \chi_{i,j}$ larger thus boosting workers’ effort.

$$C'(e_i) = \frac{1+\sum_j \chi_{i,j}}{n} a_i$$  \[5\]

This completes our proof of Conjecture 2.

**Conjecture 3. (Work satisfaction)**

So far, we have assumed that the participation constraint was satisfied so that only incentive effects were studied. However, as we argue in our conjecture section, peer chat and peer monitoring induce different participation constraints. Peer chat increases work motivation by provide a social context to workers, which has been shown to be a crucial element of well-being at work. By contrast, peer monitoring by inducing further control and restricting autonomy will have the opposite effect. We can thus write the participation constraint of worker $i$ as follows:

$$\frac{f}{n} + \frac{f}{n} ((n-1)\xi_i + \sum_j \chi_{i,j}) - C(e_i) + pc_i - pm_i \geq v_0$$  \[6\]

where $v_0$ is the utility level obtained by a worker in the next-best alternative, and $pc_i$ represents the utility gain of worker $i$ from being in a team which can engage in peer chat and $-pm_i$ represents the utility loss of worker $i$ from being in a team in which peer monitoring is present. From [6], it directly follows that peer chat will induce greater work satisfaction (left-hand side) than a baseline treatment in which there is neither chat nor monitoring and in which the participation constraint would be such that:

$$\frac{f}{n} - C(e_i) \geq v_0.$$  

This follows from the fact that peer chat is positively valued by workers ($pc_i > 0$) and it fosters altruism ($\xi_i > 0$). The effect of peer monitoring on the participation constraint is mixed because it increases workers’ revenues ($\frac{f}{n} \sum_j \chi_{i,j} > 0$) while being negatively value by workers ($-pm_i < 0$). This leads to Conjecture 3.

It is important to note that the effect of peer chat and peer monitoring on the participation constraint do not affect the magnitude of incentive effect. However, we believe they are crucial because they might affect work behavior and in particular promote counter productive work behavior such as theft or absenteeism. It is also the case that lower work satisfaction, by decreasing the left-hand side of the participation constraint, will push workers to leave the firm thus creating additional turnover costs. Workers who want to leave because the participation constraint is not met as a result of a change in organizational design (such as the introduction of peer monitoring) might not be able to do so
immediately in which case the negative impact of a poorly accepted organizational change will be seen only in the longer run. This is why managers may sometimes bypass the participation constraint and focus on incentive effect which will produce positive effect in the short run.

**Appendix B. Additional tables**

*Table B1. Study 1 Descriptive Statistics*

<table>
<thead>
<tr>
<th>Mean Worker’s production per period (in cents)</th>
<th>Percentage of time spent on each activity:</th>
</tr>
</thead>
<tbody>
<tr>
<td>(Standard deviation)</td>
<td>Work Task</td>
</tr>
<tr>
<td>No Chat-No Monitor (Baseline)</td>
<td>113.44 (92.48)</td>
</tr>
<tr>
<td>Chat-No Monitor</td>
<td>169.44 (105.68)</td>
</tr>
<tr>
<td>Chat-All Monitor</td>
<td>144.24 (75.12)</td>
</tr>
<tr>
<td>Chat-One Monitor</td>
<td>164.24 (90.64)</td>
</tr>
<tr>
<td>No Chat-One Monitor</td>
<td>136.24 (123.28)</td>
</tr>
<tr>
<td>No Chat-All Monitor</td>
<td>165.04 (116.5)</td>
</tr>
</tbody>
</table>
Table B2. Study 1 Pairwise Comparisons Between Treatments

RANK SUM TESTS AND T-TESTS FOR TOTAL PRODUCTION
[INTERNET USAGE] (TIME ON TASK) PER WORKER

<table>
<thead>
<tr>
<th>Organizational system</th>
<th>No Chat-No Monitor (Baseline)</th>
<th>Chat-No Monitor</th>
<th>Chat-One Monitor</th>
<th>Chat-All Monitor</th>
<th>No Chat-One Monitor</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chat-No Monitor</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Rank-Sum test</td>
<td>0.005</td>
<td>&lt;0.001</td>
<td>(0.129)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>T-test</td>
<td>0.003</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-One Monitor</td>
<td></td>
<td>0.039</td>
<td>0.416</td>
<td>0.014</td>
<td></td>
</tr>
<tr>
<td>Rank-Sum test</td>
<td>&lt;0.001</td>
<td>0.273</td>
<td>(0.453)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>T-test</td>
<td>&lt;0.001</td>
<td>0.042</td>
<td>(0.215)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-All Monitor</td>
<td></td>
<td>0.069</td>
<td>0.159</td>
<td>0.783</td>
<td>0.018</td>
</tr>
<tr>
<td>Rank-Sum test</td>
<td>&lt;0.001</td>
<td>0.648</td>
<td>0.314</td>
<td>(0.443)</td>
<td>0.871</td>
</tr>
<tr>
<td>T-test</td>
<td>&lt;0.001</td>
<td>0.200</td>
<td>0.418</td>
<td>(0.018)</td>
<td>0.144</td>
</tr>
<tr>
<td>No Chat-One Monitor</td>
<td></td>
<td>0.476</td>
<td>0.055</td>
<td>0.255</td>
<td>0.379</td>
</tr>
<tr>
<td>Rank-Sum test</td>
<td>&lt;0.001</td>
<td>0.182</td>
<td>0.013</td>
<td>(0.026)</td>
<td>0.466</td>
</tr>
<tr>
<td>T-test</td>
<td>0.053</td>
<td>0.079</td>
<td>&lt;0.001</td>
<td>(0.062)</td>
<td>0.116</td>
</tr>
<tr>
<td>No Chat-All Monitor</td>
<td></td>
<td>0.015</td>
<td>0.811</td>
<td>0.616</td>
<td>0.300</td>
</tr>
<tr>
<td>Rank-Sum test</td>
<td>&lt;0.001</td>
<td>0.492</td>
<td>0.733</td>
<td>(0.043)</td>
<td>0.498</td>
</tr>
<tr>
<td>T-test</td>
<td>0.015</td>
<td>0.922</td>
<td>0.058</td>
<td>(0.015)</td>
<td>0.334</td>
</tr>
</tbody>
</table>

*p-values<.10, ** p-values<.05, and *** p-values<.01.
Table B3. Study 1 - Communication Categories.

<table>
<thead>
<tr>
<th>Master categories</th>
<th>Category Number</th>
<th>% of messages</th>
<th>Category description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Social interaction</td>
<td>1</td>
<td>2.80%</td>
<td>Greetings (Hello/Goodbye)</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>3.40%</td>
<td>Distracting others (jokes, stories)</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>11.30%</td>
<td>Personal chat (talking about likes and dislikes)</td>
</tr>
<tr>
<td></td>
<td>All</td>
<td>17.50%</td>
<td></td>
</tr>
<tr>
<td>Encouragement and help</td>
<td>4</td>
<td>13.30%</td>
<td>Encouraging others to produce</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>3.40%</td>
<td>Thanking other for their cooperative behavior</td>
</tr>
<tr>
<td></td>
<td>6</td>
<td>26.80%</td>
<td>Help others complete the task</td>
</tr>
<tr>
<td></td>
<td>7</td>
<td>11.00%</td>
<td>Ask others for help and hints to complete the task</td>
</tr>
<tr>
<td></td>
<td>All</td>
<td>54.50%</td>
<td></td>
</tr>
<tr>
<td>Performance assessment and comparisons</td>
<td>10</td>
<td>1.40%</td>
<td>Ask others’ performance on the task</td>
</tr>
<tr>
<td></td>
<td>11</td>
<td>5.70%</td>
<td>State your own performance</td>
</tr>
<tr>
<td></td>
<td>All</td>
<td>7.10%</td>
<td></td>
</tr>
<tr>
<td>Discouragement</td>
<td>8</td>
<td>0.50%</td>
<td>Discouraging others to produce</td>
</tr>
<tr>
<td></td>
<td>9</td>
<td>2.00%</td>
<td>Asking others what is the point of producing anything</td>
</tr>
<tr>
<td></td>
<td>All</td>
<td>2.50%</td>
<td></td>
</tr>
<tr>
<td>Non-strategic comments on the experiment</td>
<td>12</td>
<td>14.7%</td>
<td>General comments about the experiment and its goals</td>
</tr>
<tr>
<td></td>
<td>13</td>
<td>3.7%</td>
<td>Other specific comments on the experiment</td>
</tr>
<tr>
<td></td>
<td>All</td>
<td>18.40%</td>
<td></td>
</tr>
</tbody>
</table>

Table B4. Study 2 Pairwise Comparisons between first and last ratings for each organizational system

P-values for t-tests (sign rank tests)

<table>
<thead>
<tr>
<th>Organizational system</th>
<th>No Chat-No Monitor (Baseline)</th>
<th>Chat-No Monitor</th>
<th>Chat-One Monitor</th>
<th>Chat-All Monitor</th>
<th>No Chat-One Monitor</th>
<th>No Chat-All Monitor</th>
</tr>
</thead>
<tbody>
<tr>
<td>P-values</td>
<td>0.414 (0.352)</td>
<td>0.127 (0.136)</td>
<td>0.040 (0.012)</td>
<td>&lt;0.001 (0.001)</td>
<td>0.229 (0.663)</td>
<td>0.175 (0.091)</td>
</tr>
</tbody>
</table>
### Table B5. Study 2 Pairwise Comparisons between treatments for first ratings

P-values for t-tests (sign rank tests)\(^\text{13}\)

<table>
<thead>
<tr>
<th>Organizational system</th>
<th>No Chat-No Monitor (Baseline)</th>
<th>Chat-No Monitor</th>
<th>Chat-One Monitor</th>
<th>Chat-All Monitor</th>
<th>No Chat-One Monitor</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chat-No Monitor</td>
<td><strong>0.003</strong> (0.003)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-One Monitor</td>
<td>0.620 (0.591)</td>
<td><strong>0.025</strong> (0.037)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-All Monitor</td>
<td>&lt;0.001 (&lt;0.001)</td>
<td><strong>0.101</strong> (0.131)</td>
<td>&lt;0.001 (&lt;0.001)</td>
<td></td>
<td>&lt;0.001 (0.002)</td>
</tr>
<tr>
<td>No Chat-One Monitor</td>
<td>&lt;0.001 (&lt;0.001)</td>
<td>&lt;0.001 (&lt;0.001)</td>
<td><strong>&lt;0.001</strong> (&lt;0.001)</td>
<td>&lt;0.001 (&lt;0.001)</td>
<td>&lt;0.001 (0.008)</td>
</tr>
<tr>
<td>No Chat-All Monitor</td>
<td><strong>0.191</strong> (0.183)</td>
<td>&lt;0.001 (&lt;0.001)</td>
<td>0.109 (&lt;0.107)</td>
<td><strong>&lt;0.001</strong> (&lt;0.001)</td>
<td>0.008 (0.012)</td>
</tr>
</tbody>
</table>

\*p-value<.10, **p-value<.05, and ***p-value<.01.

\(^{13}\) Similar results are obtained when using a non-parametric test with clusters at the session level (Somers’ \(d\), Somers, 1962).
### Table B6. Study 2 Pairwise Comparisons between treatments for last ratings

<table>
<thead>
<tr>
<th>Organizational system</th>
<th>No Chat-No Monitor (Baseline)</th>
<th>Chat-No Monitor</th>
<th>Chat-One Monitor</th>
<th>Chat-All Monitor</th>
<th>No Chat-One Monitor</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chat-No Monitor</td>
<td>0.401 (0.398)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-One Monitor</td>
<td>0.182 (0.207)</td>
<td>0.003 (0.007)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-All Monitor</td>
<td>0.755 (0.338)</td>
<td>0.006 (0.936)</td>
<td>0.182 (0.007)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>No Chat-One Monitor</td>
<td>0.002 (0.014)</td>
<td>0.001 (0.001)</td>
<td>0.164 (0.008)</td>
<td>0.348</td>
<td></td>
</tr>
<tr>
<td>No Chat-All Monitor</td>
<td>0.016 (0.012)</td>
<td>0.003 (0.003)</td>
<td>0.394 (&lt;0.001)</td>
<td>&lt;0.001</td>
<td>0.754</td>
</tr>
</tbody>
</table>

*p-value<.10, **p-value<.05, and ***p-value<.01.

---

14 Similar results are obtained when using a non-parametric test with clusters at the session level (Somers’ *d*, Somers, 1962).
<table>
<thead>
<tr>
<th>Organizational system</th>
<th>No Chat-No Monitor (Baseline)</th>
<th>Chat-No Monitor</th>
<th>Chat-One Monitor</th>
<th>Chat-All Monitor</th>
<th>No Chat-One Monitor</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chat-No Monitor</td>
<td><strong>0.195</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.104)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-One Monitor</td>
<td>0.216</td>
<td>0.003</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.238)</td>
<td>(0.012)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chat-All Monitor</td>
<td>0.098</td>
<td>0.279</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.090)</td>
<td>(0.448)</td>
<td>(&lt;0.001)</td>
<td>(&lt;0.001)</td>
<td></td>
</tr>
<tr>
<td>No Chat-One Monitor</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
<td><strong>0.004</strong></td>
<td>&lt;0.001</td>
<td></td>
</tr>
<tr>
<td></td>
<td>(&lt;0.001)</td>
<td>(&lt;0.001)</td>
<td>(&lt;<strong>0.001</strong>)</td>
<td>(&lt;0.001)</td>
<td></td>
</tr>
<tr>
<td>No Chat-All Monitor</td>
<td>0.018</td>
<td>&lt;0.001</td>
<td>0.359</td>
<td>&lt;0.001</td>
<td>0.154</td>
</tr>
<tr>
<td></td>
<td>(0.026)</td>
<td>(&lt;0.001)</td>
<td>(0.278)</td>
<td>(&lt;<strong>0.001</strong>)</td>
<td>(0.098)</td>
</tr>
</tbody>
</table>

*p*-value<.10, **p*-value<.05, and ***p*-value<.01.

Similar results are obtained when using a non-parametric test with clusters at the session level (Somers’ $d$, Somers, 1962).