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Conflicted Emotions Following Trust-based Interaction

Comments

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Conflicted Emotions Following Trust-based Interaction

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Abstract

We investigated whether 20 emotional states, reported by 170 participants after participating in a Trust game, were experienced in a patterned way predicted by the “Recalibrational Model” or Valence Models. According to the Recalibrational Model, new information about trust-based interaction outcomes triggers specific sets of emotions. Unlike Valence Models that predict reports of large sets of either positive or negative emotional states, the Recalibrational Model predicts the possibility of conflicted (concurrent positive and negative) emotional states. Consistent with the Recalibrational Model, we observed reports of conflicted emotional states activated after interactions where trust was demonstrated but trustworthiness was not. We discuss the implications of having conflicted goals and conflicted emotional states for both scientific and well-being pursuits.

JEL Classifications: C73, C91, D87

Keywords: emotion, affect valence, recalibrational theory, Trust game, experiment

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1. Introduction

We have investigated whether 20 emotions, reported by 170 participants after completing a Trust game and learning of its outcome, were experienced in a patterned way that conforms to predictions of the “Recalibrational Model” or predictions of Valence Models (e.g., Lang et al. 1993, 1994; Clore, Ortony, and Foss 1987; Ortony, Clore, and Collins 1988; Russell and Carroll 1999; Kuppens et al. 2012). The Recalibrational Model predicts the activation (or arousal) of emotions according to several dimensions (short-sighted, long-sighted, positive, negative, interpersonal, and intrapersonal) while Valence Models often predict the activation of emotions according to a positive-negative affect dimension alone. The Recalibrational Model is also distinguished from the Valence Models in that it predicts the possibility of conflicted emotional states.

The first part of the Recalibrational Model (1.1 in Figure 1), untested by this study, describes how the relative calibration of “short-sighted” (*V*), and “long-sighted” (*U*) programs ultimately determines Investor and Trustee behavior propensity when individuals are confronted with a Trust game choice dilemma. The relative power of these programs determines the extent to which an individual’s behavior in a trust-based interaction trades off the short-sighted goal (opportunism) for the long-sighted goal (developing a trust-based exchange relationship).

The second part of the model (1.2 in Figure 1), in which the integration of new information (from trust-based decisions and interaction outcomes) triggers emotions, is the focus of this study. According to the Recalibrational Model, emotions acting jointly in five sets to recalibrate the operation of short-sighted and long-sighted programs in both self and others are triggered by new information about Trust game outcomes. These sets of emotions computationally identify and respond to the presence of specific adaptive problems based on Trust game decisions and outcomes.

A third part of this model (1.3 in Figure 1), untested with this research, specifies the kinds of targeted recalibration effects (i.e., “positive” upregulation and/or “negative” downregulation of short-sighted and long-sighted programs) that we expect activated emotions to encourage, resulting in changes to behaviors in self and others.

Consistent with the recalibrational functions proposed by our model and the premise of competing programs in humans’ conflicted minds, we observed participants frequently reporting conflicted emotional states. In particular, participants reported experiencing simultaneously

activated positive and negative emotion sets after interactions where trust was extended but trustworthiness not demonstrated. For example, investors reported experiencing simultaneous activation of emotions in Set 5 (especially guiltiness) and Set 1 (especially contentment) while trustees reported simultaneous activation of emotions in Set 4 (especially anger) and Set 2 (especially pride). States of conflicted emotion activation are not predicted by simpler models still *de riguer* today, such as the bipolar affect Valence Model where activated positive or negative emotions are experienced as interdependent negatively correlated opposites (e.g., see Lang et al. 1993; Ortony, Clore, and Collins 1988; Russell and Carroll 1999), or the two dimensional V-shaped “arousal=effect” models¹ where it is assumed that arousal reflects either the intensity of pleasure or displeasure but never both (e.g., Clore, Ortony, and Foss 1987; Lang, 1994; Kuppens et al. 2012).

Below, we report the results of factor analysis and structural equation modeling, providing evidence that the multivariate Recalibrational Model significantly outperforms the Valence Model when describing the patterned experience of emotions reported after a Trust game. These results support the theory that sets of recalibrational emotions are triggered in patterned response to the adaptive problems produced by trust-based interactions.

1.1. Trust-Based Decision Dilemmas and Behavior Regulation by Short-Sighted and Long-Sighted Programs

When one is confronted with a dilemma, there is an internal conflict over how to pursue alternative desired outcomes that cannot be simultaneously fulfilled at their maxima. We study such a dilemma modeled by Berg, Dickhaut, and McCabe (1995), which we refer to as the Trust game. In the Trust game, an Investor first decides how much of a \$10 endowment to send a paired Trustee, with the amount sent tripled, and then the Trustee decides how much of the tripled investment, or income, to return to the Investor.

¹ Thayer (1989) proposed that two different types of arousal existed, one positive and the other negative. Similarly, the PANAS was designed with some items that contribute to an intended PA scale (designed to assess the combination of positive valence and high arousal) and other items that contribute to an intended NA scale (designed to assess the combination of negative valence and high arousal). As such, each valenced scale is intended to activate according to “arousal=valence”. We appropriated the PANAS approach of surveying emotions but test predictions that deviate from expectations of the original assumptions that activation of arousal on one scale is exclusive to activation of arousal on the others scale.

The Trust game provides both short-sighted opportunity for gaining available resources and the possibility of developing the foundations for a trust-based exchange relationship – a long-sighted security against the income risks associated with endowment asymmetry (such as resulting from the 50% chance of being Investor in this kind of experiment). Short-sighted programs evolved to solve the adaptive problem of competition for limited resources with fleeting availability by encouraging capture of all resources present before they are depleted, foregone, or the possibility of seizing them becomes less certain or riskier. Reliable trust based exchange relationships are important securities that buffer against resource shortages and times of scarcity associated with risky income (e.g., from hunting, where ‘lucky’ individuals with food share with ‘unlucky’ individuals without food, with the expectation of reciprocity when roles are reversed). Indeed, laboratory studies have demonstrated that, in response to unsynchronized resource availability among individuals in a common environment, people act pre-disposed to engage in asynchronous trading relationships (Kaplan et al. 2012).

We propose that these adaptive problems, modeled by the Trust game, are regulated by short-sighted and long-sighted programs (e.g., see Carrillo 1998; Kurzban 2010) in conflict with one another (Livnat and Pippenger 2006). The relative calibrations of an individual’s short- and long-sighted programs (determined by their unique histories, emotional capital, and present demands) regulate individuals’ behavior propensity² in dilemmas such as the Trust game (see Figure 1). According to this dual program perspective, the Investor decision trades off his short-sighted “opportunistic” goal (achieved with earnings from a kept endowment and a maximally profitable investment) with his long-sighted “cooperative” goal (achieved by developing an exchange relationship in which both trust and trustworthiness are maximally demonstrated). Likewise, the Trustee, having received a trust-based multiplied transfer of funds from the Investor, must decide whether to pursue her short-sighted program’s goal (by keeping this income), or else pursue her long-sighted program’s goal of developing a trust-based exchange relationship by returning an amount equal to or greater than what the Investor originally sent and thereby demonstrating her trustworthiness.

² While we expect individual differences in *degree* (i.e., variance in relative strengths of regulatory programs or emotions), we do not expect differences in *kind* (i.e., direction of calibrational effects), since we take the existence of these programs to be species-typical adaptations.

1.2. Recalibrational Functions of Emotions and Prediction of Emotional Experience

Based on our review of the emotion literature and the functional features of recalibrational emotions proposed by Schniter and Shields (2013), we consider 20 emotions that cluster into five categories (see Table 1) based on constellations of their shared functional features. We chose to classify and predict the twenty emotional states studied because they are frequently used in versions of the one-dimensional Positive and Negative Affect Scale (PANAS) developed by Watson et al. (1988), and predicted by the Valence Model that we compare to the Recalibrational Model.

Nesse (2004, p.1138) states that, while emotions have been selected for because of their ability to solve specific adaptive problems, “...there is no one to one correspondence between an emotion and a function. One emotion can serve multiple functions, and one function may be served by several different emotions.” Consistent with Nesse, our functional classification of twenty emotions yields five unique sets containing multiple emotions that we expect to be triggered in concert for common functional purposes (i.e., to facilitate achievement of short- and long-sighted program goals). We characterize these functions as *positive* and *negative* recalibrations, *intra* and *interpersonally* targeting *short-* and *long-*sighted programs.

Generally, an adaptationist and functional perspective of emotions (e.g., Tooby and Cosmides 1990; Buck 1999; Cosmides and Tooby 2000) argues that emotions facilitate behavioral regulation by recruiting the assistance of a number of psychological, physiological, and behavioral processes that provide either positive or negative feedback (pleasant and unpleasant experience) used in updating the calibration of conflicting internal regulatory variables. Pleasant experiences are rewarding and can incentivize approach behavior and continuation of the prior behavior or interaction that triggered them (Watson et al. 1999; Carver and Scheier 1990). Unpleasant experiences are costly and motivate a change, whether through behavior reduction, avoidance, or aggression (Gray 1971). Of the set of twenty emotional states, we conjecture that nine [*appreciative, happy, content, cheerful, triumphant, inspired, secure, proud, believable*] are experienced as positive, one [*surprise*] could be either positive OR negative (forming the unique Set 3), and ten [*disgusted, jealous, aggravated, frustrated, angry, depressed, sad, embarrassed, ashamed, guilty*] are negative.

Emotions are ultimately designed to deal with adaptive problems requiring program orchestration (Tooby and Cosmides 1990, Cosmides and Tooby 2000). The optimal calibration

of programs managing an individual's behavior in a particular interaction depends in part on the calibration of other's behavior regulation programs. To solve this adaptive problem, emotions facilitate the achievement of program goals via intrapersonal and interpersonal behavior regulation (e.g., see Levenson 1999; Van Kleef et al. 2004; Butt, Choi, and Jaeger 2005). Below we discuss emotions with intrapersonal effects (reinforcing, maintaining, or changing one's own behavior), and interpersonal effects (reinforcing, maintaining, or changing another's behavior).

When one's prior actions did not succeed in achieving an adaptive goal, intrapersonal negative emotions are triggered to recalibrate one's own regulatory programs (Carver and Scheier 1990; Gómez-Miñambres and Schniter 2013). For example, guilt, an intrapersonal emotion triggered exclusively in response to a failure of the long-sighted program, recalibrates (i.e., downregulating) one's short-sighted program, decreasing the value of immediate opportunistic payoffs. As a result, valuation of the long-sighted goal increases, affecting the ability to "commit" to its pursuit (Frank 1988). On the other hand, when one's prior actions have succeeded in achieving an adaptive goal, positive emotions are triggered and recalibrate regulatory programs in the self to ensure further achievements. For example, the experience of feeling believable and proud occurs when positive emotions are triggered by the decision to engage in cooperative behavior. These positive intrapersonal emotions upregulate the long-sighted program (relative to the short-sighted program) so as to further encourage the behavior that led to successful cooperation. We conjecture that, of the twenty emotional states studied, seven [*triumphant, inspired, secure, proud, believable, surprised, guilty*] are exclusively intrapersonal.

Another way that emotions are designed to function is interpersonally: by regulating others' programs in an effort to affect interaction behaviors with one's self. For example, consider the gratitude emotion. Discovery that another has foregone short-term rewards in the pursuit of a long-term exchange relationship with one's self, for example by providing resource or assistance, presents a fortunate relationship building opportunity for the recipient. Gratitude and appreciation can signal one's favorable valuation of the other and propensity to cooperate with them (Tooby and Cosmides 2008), encouraging future trust much in the way that "promises" do (Schniter, Sheremeta, and Sznycer 2013). Experimental evidence supports this functional account of grateful and appreciative feelings (Tesser, Gatewood, and Driver 1968; Algoe, Haidt, and Gable 2008; McCullough et al. 2001).

We conjecture that most of the emotions studied (thirteen of twenty), function both intrapersonally and interpersonally. By initiating self-imposed recalibrations with the functional equivalence of recalibrations that the offended party might otherwise impose, shame and embarrassment downregulate the other's non-cooperative inclination so as to preempt targeted inter-personal recalibrations of one's self. For example, shame could preempt another's anger or disgust reaction if it preemptively led to self-punishment and distancing one's self from the other. Likewise, the appeasement function of embarrassment may act remedially; by effectively allowing rule violators to hedonically punish themselves, the angry and aggressive responses of offended parties are preempted (Keltner, Young, and Buswell 1997; de Jong 1999).

While we conjecture that fifteen of the twenty emotional states studied may facilitate the achievement of either short- or long-sighted programs' goals, we consider five emotional states to exclusively facilitate achievement of the long-sighted program's goal. Of these we derive two unique sets: a positive and exclusively intrapersonal Set 2 [*proud, believable*] and a negative Set 5 [*embarrassed, ashamed, guilty*]. The positive emotional states that facilitate both short-sighted and long-sighted programs [*appreciative, happy, content, cheerful, triumphant, inspired, secure*] form the unique Set 1. The negative emotional states that facilitate both short-sighted and long-sighted programs [*disgusted, jealous, aggravated, frustrated, angry, depressed, sad*] form the unique Set 4. We next explain how, according to the recalibrational theory, emotions are triggered by computationally assessments of successes and failures in the Trust game.

Our recalibrational theory of emotions is built around conflicting short-sighted and long-sighted behavior regulation programs, which determine an individual's choices when faced with decision dilemmas, such as in the Trust game. We propose that the emotions facilitating achievement of these programs' goals computationally assess game outcomes for the purpose of identifying and reacting to successes and failures of the short-sighted and long-sighted programs (in self and other). According to our model, emotions are "triggered" when they integrate information about the Trust game outcome and computationally identify specific successes and failures. We label these computational triggers *L* and *S*, for the long-sighted program's goal achievement and the short-sighted program's goal achievement, respectively.

We define the following variables observed in our Trust game: endowment (= *e*), amount sent by Investor (= *s*), amount returned by Trustee (= *r*). We calculated success (with a maximum

of 1 and minimum of 0) of the short-sighted program achieving its goal (S) according to competing perspectives of the Investor (I) and Trustee (T):

$$S_I = (e - s + r)/(e + 2s)$$

$$S_T = (3s - r)/3s \text{ if } s > 0, \text{ else } 0.$$

We calculated success (with a maximum of 1 and minimum of 0) of the long-sighted program achieving its goal (L), based on the mutual perspective shared by Investor and Trustee:

$$L = \text{Trust} * \text{Trustworthiness},$$

$$\text{where Trust} = s/e, \text{ and Trustworthiness} = \min\{r/s, 1\} \text{ if } s > 0, \text{ else } 0.$$

S evaluates the short-sighted program's goal achievement after Investor and Trustee decisions have been made. In addition to valuing any portion of the endowment kept, an Investor's short-sighted program values maximally recouping profitable returns on any investment made. Thus, to reasonably evaluate opportunity captured by an Investor we consider how much of the endowment was kept and how much of the multiplied investment was recouped by calculating $(e - s + r)/(e + 2s)$. Accordingly, an Investor's S is maximized ($= 1$) when all endowment was kept (in which case $s = 0$ and $r = 0$), or if – in addition to any endowment kept – the maximum possible profitable return from the investment was recouped (i.e., $r = 3s$). A Trustee's S is maximized ($= 1$) when $s > 0$ and $r = 0$ and is minimized ($= 0$) when either $s = 0$ or when $s > 0$ and $r = 3s$.

The consummation of a cooperative trust-based relationship requires that both trust and trustworthiness be demonstrated. Trust is demonstrated by the invested amount of endowment at risk. Trustworthiness is demonstrated by proportion of investment voluntarily reciprocated to the Investor. A cooperative trust-based relationship fails to be established when either the Investor or the Trustee has pursued maximum opportunism. As such, $L = 0$ when $s = 0$ or when $r = 0$.

The relationship between s and r should also be predictive of emotion activation. Notice that L increases with s . For a fixed s , L also increases with r . S_I decreases with s , and increases with r for a fixed s , while S_T increases with s and decreases with r for a fixed s . As detailed in Figure 1, L affects Sets 1, 2, 4, and 5, while S_I and S_T affect Sets 1 and 4. Let us hold s constant and assume the loadings on L and S are approximately equal. If we compare scenarios where $r < s$ to scenarios where $r > s$, the aforementioned comparative statistics would predict that (i) there is higher emotion activation when $r > s$, and (ii) there is more conflicted emotion activation when $s > r$. The first prediction is due to upregulation of Sets 1 and 2 and downregulation of Sets 4

and 5. The second prediction is due to downregulation of all sets, creating “conflicted” emotion activation of both negative and positive sets, albeit lesser positive emotion activation than where $r > s$. The prediction of conflicted emotions holds true even if the loadings on L are much smaller than on S_I or S_T .

Prior research has found that when s is relatively large (e.g., greater than half of the endowment), r tends to exceed s , whereas when s is relatively small, r tends to be equal or less than s (e.g., see Ostrom and Walker, 2003). Given we find this same distribution, the above predictions should hold: L becomes larger as s and r increase. S_I is largest when s is large and $r > s$. The effect of S_T is ambiguous, as it increases with s but decreases with r .

In summary, emotions are triggered by computational assessments of short- and long-sighted programs’ successes and failures. Positive emotional states are maximally experienced when trigger values are largest (= 1) and negative emotional states are maximally experienced when trigger values are smallest (= 0). According to their design functions, triggered emotions either contribute to the reinforcement of successes or the reduction of failures by upregulating or downregulating specific programs in self and others. We tested whether constellations of specific antecedents (the L and S triggers produced by Trust game interaction) reliably predict specific sets of emotional experiences.

2. Design, Predictions and Procedures

2.1. Natural Experiment Design

We conducted a Trust game in which the Investor received an endowment of \$10 and could send any portion of it to the Trustee, with the amount sent tripled (see Appendix A instructions). The Trustee then decided how much of the tripled investment, or income, to return (or else keep). Following the Trust game we administered a 20-item emotional status survey³

³ To avoid experimenter demand effects that might result by soliciting reports on only a few select emotional states commonly ascribed to failed trust-based interactions (i.e., anger and guilt) and identified in the literature (e.g., Ketelaar and Au 2003), we constructed a survey of a large array of emotional states, based on the Positive and Negative Affect Scale (PANAS), a self-report measure of positively and negatively valenced affect state activations developed by Watson et al. (1988) that has been demonstrated across large non-clinical samples to be a reliable and valid measure of these states (Crawford and Henry 2004). Consistent with the moderately high reliability of internal consistency reported previously by Watson et al. (1988) and others (e.g. (Jolly et al., 1994; Mehrabian, 1998; Roesch, 1998) we found the Cronbach alpha coefficient was 0.909 For the Positive Affect Scale and 0.874 for the Negative Affect Scale of our 20 item emotion survey.

(see the Appendix) in which participants reported how much they felt activation of each of 20 emotional states (on a five point scale labeled (1) very slightly or not at all, (2) a little, (3) moderately, (4) quite a bit, (5) extremely) as a consequence of their recent game interactions and outcomes. The computer software presented all emotional states on one screen in a randomized order, different from that suggested in Table 1 (see Appendix B emotion survey). Using this laboratory implementation of the Trust game that engaged participants in one-shot anonymous economic interactions, followed by a well-established emotional status survey, we investigated whether emotional experiences were reported in a patterned and predicted way as a consequence of game outcomes.

2.2. Predictions

Inspired by functional theories of social emotions (Trivers 1981; Cosmides and Tooby 1989), Nesse (1990, p.275; 1999, p.458) made predictions (not yet tested) about how specific emotions mediating reciprocity would be triggered by the four types of interaction patterns produced by a repeated Prisoner's Dilemma game. While our study differs in that it examines a wider set of emotions and single anonymous interactions, we contend that emotions should still be predictable independent of so called "one-shot" cues. Our evolved psychology errs to caution by processing information about one-shot interactions with uncertain resource asymmetries under the premise that they may in fact be repeated in the future (e.g., see Delton et al. 2011). We also suspect that Investors who make trust-based choices discover the consequent effects on their payoffs and extend this information when constructing generalizable models about the trustworthiness of Trustees in the population (e.g., the experimental subject pool). Working with these assumptions, this study takes a natural experiment approach, examining the relationship between participants' endogenous generation of Trust game outcomes and consequential reports of multiple emotions. Like the Prisoner's Dilemma, the Trust game is another model for transactions that require trust, albeit a model of asynchronous (rather than synchronous) trust-based exchange.

With the set of predictions below, generated by Valence Models and the Recalibrational Model, we tested assumptions and compared how well different models of emotional experience predict the emotions reported by participants who had just completed Trust games.

Valence Models

P1: In all three versions of the Valence Model predicted below (**P1.1-P1.3**), emotions show positive correlation within the Positive Affect (PA) and Negative Affect (NA) set.

P1.1: PA and NA sets are *independent*: no correlation ($= 0$) is expected between them.

P1.2: PA and NA sets are strictly *interdependent* with negative ($= -1$) correlation between them. Consistent with a purely “bipolar” model of valence, reports of simultaneously experienced strong positive emotion and strong negative emotion are unexpected.

P1.3: Interdependence is *unrestricted* between emotions in the PA and NA sets. While negative correlation is expected between sets, positive correlation between items in PA and NA sets can also occur.

Recalibrational Model

P2: Conflicted minds produce conflicted emotions: at times involving positive correlation between (simultaneously experienced) positive and negative emotions.

P3: Emotion experiences are reported in a patterned way according to a multivariate set of shared recalibrational features (i.e., *positive* and/or *negative* recalibration effects, targeting *short-* and/or *long-sighted* programs, *intra* and/or *interpersonally*).

P4: *L* and *S* (variables evaluating outcomes from Trust game interactions) predict the patterned experience of emotions for the 4 testable sets of the Recalibration Model better than for the 2 sets of the Valence Models.

2.3. Experimental Procedures

The experiment, programmed using z-Tree (Fischbacher 2007), was conducted at Chapman University’s Economic Science Institute. Participants were recruited from a campus-wide subject pool consisting primarily of undergraduate students.

There were eight experimental sessions, each lasting approximately thirty-five minutes. No participant participated more than once. Each session had between 18 and 24 participants, seated in individual cubicles, and was conducted as follows. An experimenter read the instructions aloud explaining experimental procedures and payoffs while every participant followed along with their own copy of the instructions. After finishing the instructions, participants were given five minutes to privately write down their answers to several quiz

questions. After participants completed the quiz, the experimenter distributed a printed copy of the correct quiz answers. To ensure understanding, any remaining questions were answered privately.

Participants, randomly assigned to one of two roles: “person 1” (Investor) or “person 2” (Trustee), interacted anonymously in the Trust game over a local computer network, then completed the 20 item survey in which they reported the intensity of various emotional states consequent on their decisions, game interactions, and resulting outcomes. Earnings from the Trust game plus \$7 for arriving to the experiment on time and participating were paid out privately at the end of the experiment.

3. Results

In this section, we report general results of the Trust game and the emotional status survey. In section 3.1 we investigate whether 20 emotions were experienced in a patterned way that conforms to predictions of the Recalibrational Model or predictions of Valence Models. In section 3.2 we examine the full models of emotional experiences according to four triggers (based on computation of adaptive problems consequent of economic decisions and interactions), comparing the fit of the unrestricted Valence Model and the Recalibrational Model.

We found no significant differences between the eight sessions and report the joint results of all 170 participants. Figure 2 displays the scatter plot of the amount sent and the amount returned. There was substantial variability in individual behavior. On average, Investors sent \$6.01 (SD = 3.64) and Trustees returned \$6.16 (SD = 5.92), resulting in profits of \$10.14 (SD = 3.72) and \$11.88 (SD = 7.12), respectively. These results are consistent with previous findings of Berg et al. (1995). Likewise, there was substantial variability in individual reports of emotional experience. The average reported emotional state (as a result of Trust game interactions) had a mean of 2.20 (median = 1, SD = 1.45), near 2 (“a little”). Ratings on every emotional state ranged from 1 (“very slightly or not at all”) to 5 (“extremely”). While the modal report for most (17/20) emotional states was 1 (“very slightly or not at all”) modes were also seen at 3 for *happy* and 5 for *content* and *appreciative*. Reports of 1 were more frequent for emotional states in the negative set than for the positive set (1218/1700 versus 527/1700, respectively), contributing to significantly lower intensity of reported negative states (M = 1.61, SD = .77) than positive states

($M = 2.80$, $SD = 1.08$) according to Wilcoxon matched-pairs tests ($Z = 7.605$, $p < .001$). This pattern of significantly lower reported negative states was observed in both Investors ($Z = 5.853$, $p < .001$) and Trustees ($Z = 4.888$, $p < .001$). We constructed an “activation” score based on individuals’ average reports across all emotions. Upon exploration of the data distribution of activation scores with regards to game outcomes (see Figure 2) we discovered that where trust was extended ($s > 0$) and trustworthiness demonstrated ($s < r$), both Investors and Trustees (Investor Mdn = 2.40, Trustee Mdn = 2.55, 34 pairs) experienced more activation than did the Investors and Trustees (Investor Mdn = 2.08, Trustee Mdn = 2.12, 40 pairs) where trustworthiness was not demonstrated ($0 < s$ and $s \geq r$). A Wilcoxon rank-sum (Mann-Whitney) test was used to compare average activation (Investor: $Z = 2.834$, $p = .005$; Trustee: $Z = 3.821$, $p < .001$).

3.1 Shared Features of Emotions

Valence Models assume two factors: one comprised of a standard set of positive emotional states that positively correlate with one another [*appreciative, happy, content, cheerful, triumphant, inspired, secure, proud, believable, and surprised*], and the other comprised of a standard set of negative emotional states that positively correlate with one another [*disgusted, jealous, aggravated, frustrated, angry, depressed, sad, embarrassed, ashamed, and guilty*]⁴. Using exploratory factor analysis, we rejected that a two-factor model fit the data best, as the Bayesian information criterion (BIC) was inferior to models with three, four, five, six, seven, and eight factors⁵. Consistent with **P1**, item analysis indicated that not all (43 of 45) correlations were significantly positive between positive states, nor between all (36 of 45) negative states.

Consistent with **P1.3** and **P2**, cross tabulation indicated occurrences of simultaneously activated positive and negative emotions. We observed 57 cases from 13 (7.64% of) respondents reporting activation of positively valenced (P) and negatively valenced (N) emotions that were both felt “extremely” (= 5); 231 cases from 34 (20% of) respondents reporting activation of P and N emotions that were both felt in the range from “quite a bit” to “extremely” (≥ 4); 973 cases from 69 (40.59% of) respondents reporting activation of P and N emotions that were both felt in

⁴ These “standard” sets were based on the PANAS (e.g., Watson et al. 1988).

⁵ Results are available upon request.

the range from “moderately” to “extremely” (≥ 3); and 2653 cases from 114 (67.06%) respondents reporting positive and negative states that were both felt in the range from “a little” to “extremely” (≥ 2). We also examined simultaneous activation of the 9 positively valenced (P) emotions (not including surprise) and the 10 negatively valenced (N) emotions and constructed a “conflicted” score (valued 1-5) based on the maximum level at which any pair of P and N emotions were both equally activated (i.e., the $\min\{\max P, \max N\}$). We find that where trust was extended but trustworthiness not demonstrated ($0 < s$ and $s \geq r$), both Investors and Trustees experienced more conflicted emotions (Investor Mdn =2.00, Trustee Mdn =3.00, 40 pairs) than did the Investors and Trustees (Investor Mdn =1.00, Mdn =1.00, N=34 pairs) where trustworthiness was demonstrated ($0 < s < r$). According to the two-sample Wilcoxon rank-sum (Mann-Whitney) test the medians of conflicted scores are significantly different (Investor: $Z = 3.204, p = .001$; Trustee: $Z = 3.347, p < .001$).⁶

We used Confirmatory Factor Analysis (CFA)⁷ to measure how well our data on reported emotional states fit the Recalibrational Model and variants of the Valence Models. Each variant of the Valence Model shares the assumption that positive correlations exist among individuals’ reporting positive states and positive correlations exist among individuals’ reporting negative states. Therefore, in all Valence Models we constrained each emotion to load onto only one of the two factors. However, because each variant of the Valence Model has a different assumption concerning relationships that might exist among simultaneously experienced of positive and negative states, they differ only in the constraints they impose on the positive and negative factor correlations. Model 1 constrains the factors to have a zero correlation (where positive states bear no relationship with negative states), Model 2 constrains the factors to a correlation of negative one (as would be appropriate if the experience of emotional states was only possible on a bipolar continuum), and Model 3 imposes no restrictions on the factors’ correlation.

⁶ These significantly greater conflicted scores are not simply reflecting greater activation, however, as they are found among the otherwise lower activation set where extended trust failed to elicit trustworthiness, rather than among the set where extended trust successfully elicited trustworthiness and significantly greater activation was found (see Figure 2).

⁷ We used Stata version 12.1 and the ‘SEM’ procedure (Structural Equations Model) finding the fit for maximum likelihood. One participant was dropped who reported the same value for all emotional states. Additionally, we omitted surprise from both the Recalibrational Model’s and Valence models’ fit tests because, being the only emotional state included in the Recalibrational Model Set 3, it would have produced an automatic significant loading within the Recalibrational Model, unfairly biasing results in its favor.

Summaries of CFA results for the three Valence models and the Recalibrational model are shown in Table 2. The lesser Bayesian information criterion (BIC), lesser root mean square error of approximation (RMSEA), greater comparative fit index (CFI), and greater log-likelihood (LL) made it apparent that Model 1 fit better than Model 2, and that Model 3 fit better than Model 1.⁸ The difference between Model 3 and Model 1 was statistically significant ($X^2(1) = 48.01, p < .001$). Consistent with **P1.3**, Model 3's correlation between positive and negative factors was $-.70$ and significantly different from zero and from -1 ($p < .001, 95\% \text{ CI } [-.787, -.614]$).

In Table 2 we describe the derived CFA fit from the Recalibrational Model as "Model R". Model R predicted the patterned experience of emotions according to the four factors corresponding to Set 1, Set 2, Set 4, and Set 5 of the Recalibrational Model (see Table 1). Consistent with **P3**, all emotional states loaded positively and significantly (at a 1% level) onto the predicted latent factors of Model R, but not the predicted latent factors of Valence Model 3.⁹ With a greater LL, greater CFI, lesser RMSEA and lesser BIC, Model R provided a better fit than the unrestricted Valence Model 3 (according to guidelines set forth by Gefen et al. 2011).

3.2. Comparison of Structural Equation Model Fit

We used Structural Equation modeling to compare the fit of the Recalibrational Model to the fit of the Unrestricted Valence Model. To compare these models we tested both with triggers *S* and *L*, as computed from game interactions. Results of both exercises are shown in Table 3. Given that we did not find support for either perfect independence or interdependence in section 3.1, we did not restrict correlations and allowed all latent factors to freely correlate in both models.¹⁰ The Recalibrational Model has more factors than the Valence Model, which could arguable lead to "overfitting" – having a better fit by describing more error instead of predicted

⁸ While neither model has good approximate model fit according to guidelines proposed by Hu and Bentler (1999), we caution readers that there is substantial disagreement in the literature about interpreting such guidelines (Marsh, Hau, and Wen 2004; Beauducel and Wittmann 2005; Fan and Sivo 2005; Yuan 2005; Tomarken and Waller 2005; Barrett 2007).

⁹ Results available upon request.

¹⁰ Comparative measures penalized for additional variables. Given that all between factor correlations were significant for the unrestricted Valence Model but not for the Recalibrational Model, this choice should bias against the Recalibrational Model's relative fit superiority.

relationship. To avoid overfitting, we report the BIC, which penalizes for added variables. Finally, we report the difference between models, assessing whether the better fit was statistically significant despite the difference in factors seen in Table 3.

As with CFA, we found superior results via SEM for the Recalibrational Model, consistent with **P4**. Despite penalizing for additional fitted variables, the difference was significant. In the Recalibration Model, all of the latent DV triggers' coefficients were significant (below 5%) with the predicted sign (see Table 4), whereas in the Unrestricted Valence Model not all (17 of 19) of the trigger's coefficients were significant (see Table 4). Finally, we found the overall equation level goodness of fit higher for the Recalibration Model.

4. Discussion

Using confirmatory factor analysis to assess latent sets, and structural equation models to assess triggers on latent sets we demonstrated that the Recalibrational Model predicts the experience of predicted emotional states (specifically, five latent sets of these) following the Trust game, strongly and significantly outperforming the Valence Models.

The important take-away from results of our model comparisons is that we provide an improvement over the widely accepted Valence Models used in predicting emotional reports, and that we can additionally predict the activation of conflicted emotional states which have received relatively little attention in the emotion literature (though see Lerner and Keltner 2000; Fong 2006). In addition to better fit, our Recalibrational Model is interpretable because it is derived from principles of recalibrational theory. Below we discuss potential sources of unexplained variance, consider future directions for further exploring the predicted effects of recalibrational emotions on trust-based interactions, and suggest some implications of having conflicted minds for both scientific and well-being pursuits.

We consider two potential classes of explanations for currently unexplained variation in how strongly emotional experiences are rated by participants: first, that participants either have imperfect access to their emotional states, differing interpretations of the emotion labels, or the fidelity of their reports is compromised, and second, that there is heterogeneity in the experience of certain emotions (e.g., guilt).

People who are asked to rate single emotions may not be able to accurately describe their emotional states (Ellsworth and Tong 2006) if emotion experiences are more often and accurately described with multiple words (Izard 1977), or with different words among different people. While we acknowledge that language could present problems for this research, the success of previous research on self-reported emotions in conjunction with experimental games (Ketelaar and Au 2003) gave us encouragement in pursuing measures of self-reported emotions following an economic game. Nevertheless, analysis of emotion reports revealed a “floor effect” that might have resulted from a problem with the instrument used, untruthful results, a problem interpreting emotion labels, and difficulty identifying and reporting emotional states.

Nemanick and Munz (1994) suggested that the PANAS scale’s lower anchor may not be properly constructed to form a true lower pole. The PANAS response option “1”, the lowest response possible on the five point Likert, is labeled with a combination of two state levels: “not at all” and “very slightly”. By combining both state levels into a single response option a larger proportion of response types from the possible spectrum (not at all to always) may accumulate at that value. Future research should consider restructuring the response options and testing whether a different distribution of responses results.

Data quality may also have been affected if participants made untruthful reports. Experimental economists are particularly concerned that participants “will not ‘tell the truth’ unless incentives make truth telling compatible with maximizing utility” (Lopes 1994, p.218). According to a meta-review by Camerer and Hogarth (1999) there is no clear evidence that additional financial incentives would improve the quality of responses in a simple survey task like ours. In fact, it has been noted that for short tasks like PANAS surveys that people are known to voluntarily complete without problem (because they have sufficient intrinsic motivation to do so), an attempt at increasing participation via financial incentives often “backfires” with counter-intentional effects (e.g., see Mellstrom and Johannesson 2008). Nevertheless, wary of the possibility that participants may have been incentivized to use efficiency tactics to expediently complete the survey (such as marking all responses with the same), we reviewed our data and found only one apparent case¹¹ of such behavior (< 1% of sample).

¹¹ This individual reported 3s on all emotions.

A battery of 20 emotional states (like in the PANAS) may be too broad for the purpose of studies like ours. Self-access to some of the emotional states studied may be limited and interpretation of labels may not be uniform. We chose this battery because it is comparable among widely used measures of multiple emotional states, however future studies on trust and emotions would benefit from a select set more appropriate to the problems studied. In selecting a refined set, universally interpretable and recognizable emotional states should be considered. We suggest appreciative, happy, proud, frustrated, angry, and guilty: a selected set with balanced valance that is representative of the functional categories covered by our model. Additionally, we observed that the emotional states represented in this selected set were individually predicted well (see Table 4) and demonstrated less of a floor effect than the other emotional states we studied (total 1s reported were 479/1020 or 47% versus 1266/2380 or 53.2%, respectively).

By triangulating with more objective neurological, physiological, and behavioral measures of emotional states some of the discussed limitations of itemized self-reports could be overcome. While untruthful reporting and reports by abnormal types in the population may occur and should be of consider by future research, we do not expect that these factors account for a large portion of observed variance.

While we have developed theory of emotions' ultimate function (i.e., what the mechanisms were selected to do) and derived our predictions of antecedents from it, this study only tests the emotions' proximate functioning (i.e., how and when the mechanisms are triggered). Future studies can take the Recalibrational Model one step further and test for ultimate functions by examining whether the future actions of those individuals who report emotions under the predicted conditions are affected as theory predicts. Three studies which we know of have taken this approach to testing ultimate functions already: Ketelaar and Au (2003) who demonstrated how the experience of guilt leads to choices of cooperation over opportunism, Fehr and Gächter (2002) who demonstrated that angry individuals are more likely to engage in costly punishment, and Dunn and Schweitzer (2005) who have shown that Investor happiness and gratitude increase trust in the Trust game.

The experience of conflicted emotions has been discussed by others (Cacioppo, Gardner, and Berntson 1999) and evoked with wins and losses in the laboratory (Larsen, McGraw, and Cacioppo 2001; Larsen et al. 2004), as well as by trust-based interaction in our study. However, conflicted emotions are not well appreciated as a core trait of human nature, and may appear as

flaws of human nature that interfere with rationality (Sherer 1984; Elster 1995). While some researchers have long moved past the bipolar affect models, instead recognizing that positive and negative affect are at times independent dimensions (e.g., see Watson, Clark, and Tellegen 1988), psychophysicologists (Driscoll, Tranel, and Anderson 2009; Lang, Greenwald, Bradley, and Hamm 1993) neuroscientists (Proverbio, Zani, and Adorni 2008; Screenivas, Boehm, and Linden 2012) behavioral economists (Morretti and di Pellegrino 2010; Brandts, Riedl, and van Winden 2009; Van den Berg, Dewitte, and Warlop 2008; Morris 1995) and decision scientists (Hogarth, Portell, Cuxart, and Kolev 2011; Reid and Gonzalez-Vallejo 2009; Schlosser, Dunning, and Fetchenhauer 2013) continue to use bipolar affect scales (for example, the Self-Assessment-Manikin valence scale developed by Lang (1980)). Our study cautions against assuming that the explanatory power provided by the Valence Model is sufficient for understanding relationships between trust-based behavior and emotions. We suggest that more complex multivariate models, such as the Recalibrational Model, better track the triggered experience of conflicted emotions and subsequent behaviors.

In light of our model, we have identified the mechanics that could produce an emotional equilibrium between partners: a condition in which the emotional impact of partners' behavior on each other and on themselves is one in which all engaged programs are kept in the same relative state (before and after action). An important implication of this stable equilibrium is that even under such conditions, the mind is expected to remain conflicted. The recognition of a conflicted mind and the experience of conflicted emotions challenges our intuitions of a "self's" singular internal interests and the consonance of a non-contradictory self-representation generally attributed to a sane mind. While the implications of a modular recalibrational theory of emotions might be existentially and even epistemologically difficult to grapple with, we extend the following practical implications for mental health professionals and laypeople alike.

Psychotherapists often treat patients who complain of and suffer from emotional states – and it is not uncommon for patients that patronize these professionals to seek an escape from unwelcomed emotions (e.g., Nesse 1991, 2000). Treatment of these emotions, whether through behavioral intervention or psycho-pharmaceutical treatment, may benefit from the degree to which psychologists, psychiatrists, and counselors are informed of (1) the functional uniqueness and similarities distinguishing the emotions and their taxonomic classifications, and (2) the choice dilemmas and post-decision situations from which emotional states precipitate. The

Recalibrational Model suggests that there are distinct situations following trust-based interactions that not only lead to the activation of “positive” feelings or “negative feelings”, but that also might activated conflicted feelings. We expect that the Recalibrational Model also applies to social and moral dilemmas. Dewitte and De Cremer (2010) have identified that the inherent cooperation vs. opportunism tradeoff structures of social dilemmas share common features with self-control problems. Similarly, Gomez-Minambres and Schniter (2013) propose an emotionally regulated dual-program model to explain activation of recalibrational emotions as a consequence of decisions made in self-control dilemmas.

Personal and social life entail suffering and happiness because both positive and negative emotional states serve needed functions, recalibrating our own inner-workings as well as the inner-workings of those that we interact with. Given the uncertain future, we need to constantly engage in recalibration of ourselves and others to make the most of opportunities given our needs.

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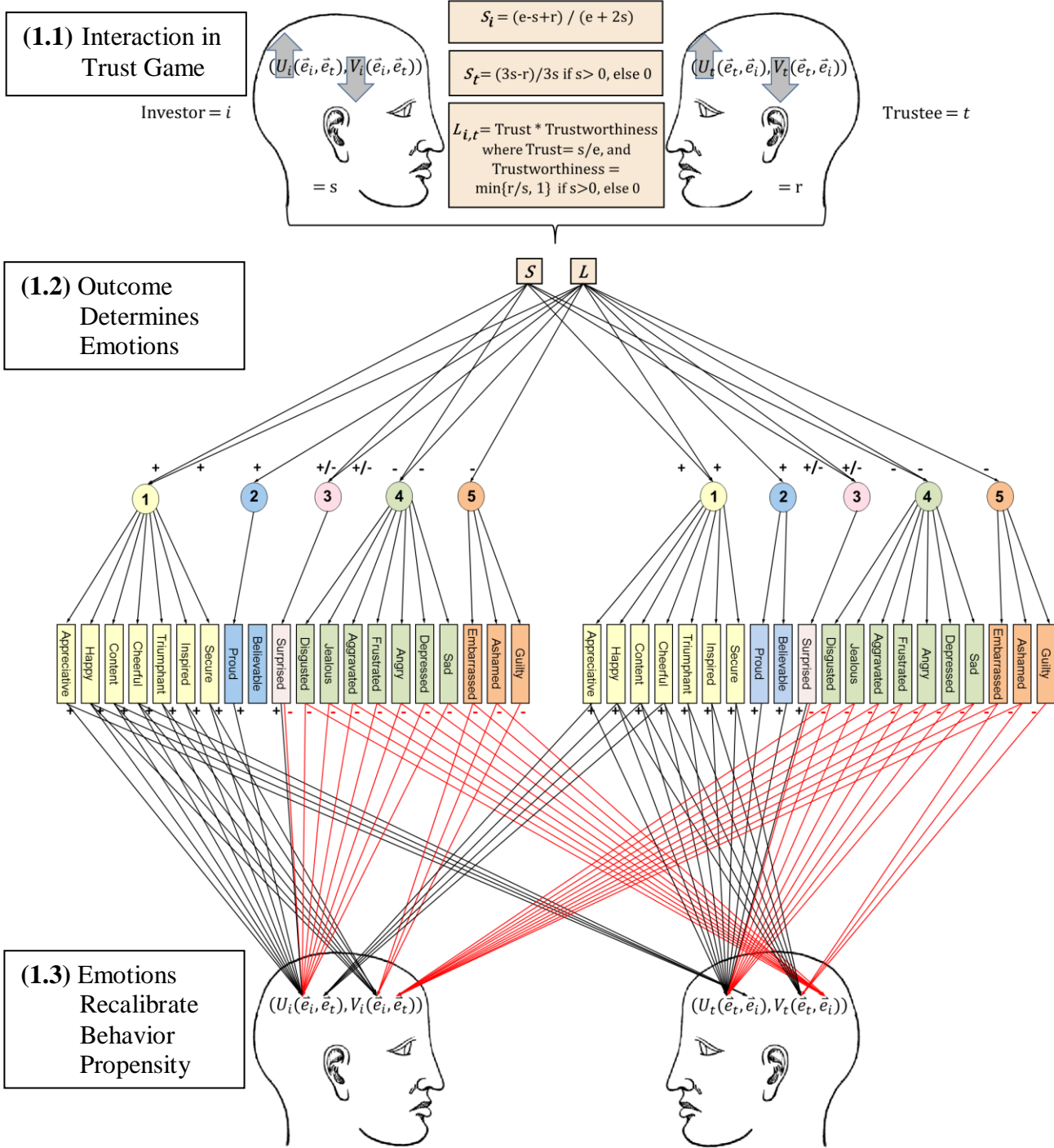
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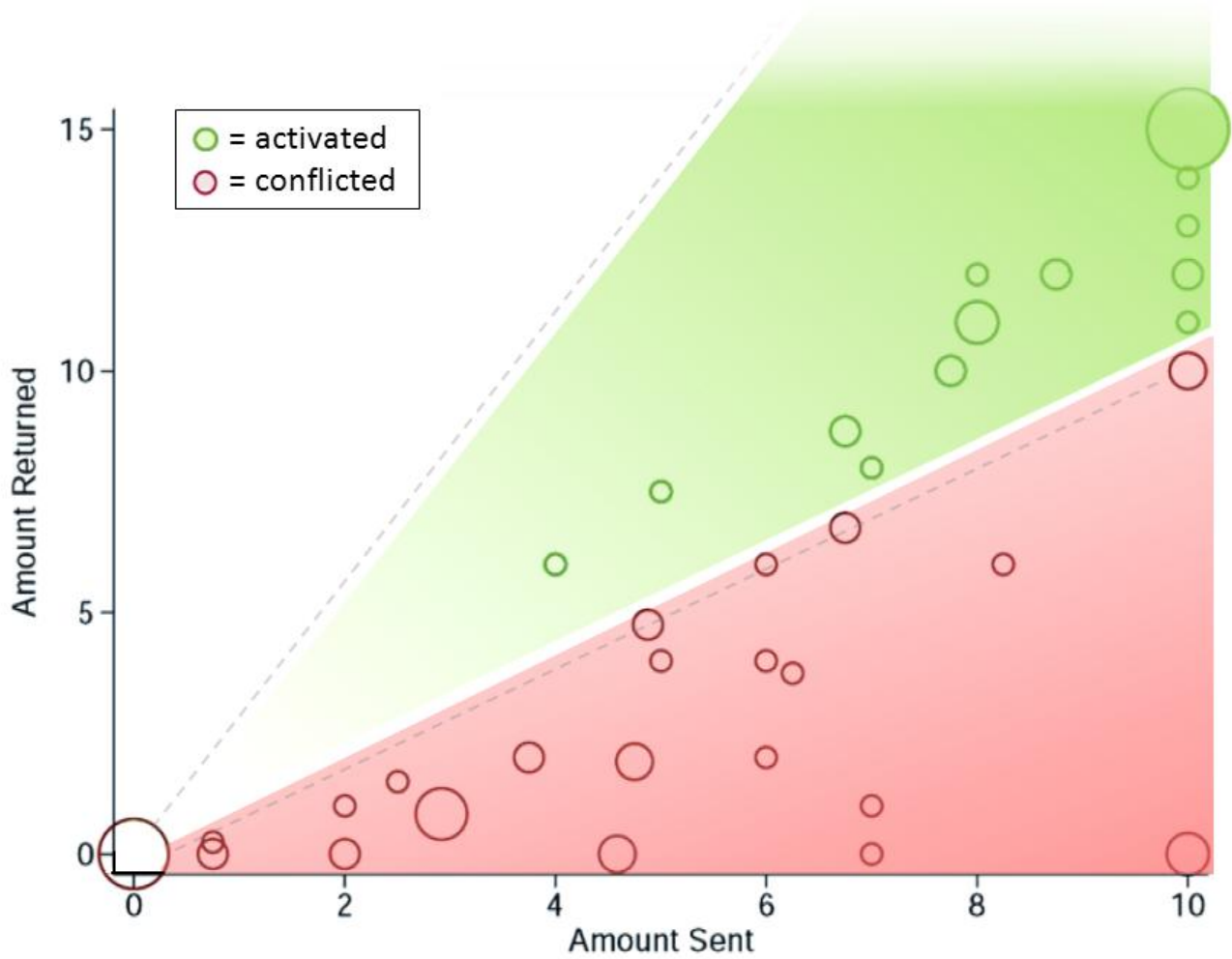
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Figure 1: Complete Recalibrational Model



Note: **(1.1)** Investor (i) and Trustee (t) each have two weights U and V , where $U \geq 0$ and $V \geq 0$, determining the relative power of conflicting long-sighted and short-sighted programs, respectively. The balance of weights determines behavior propensity via decision function, where U weakly increases and V weakly decreases amount sent ($= s$) or amount returned ($= r$), for the Investor (i) and Trustee (t), respectively. **(1.2)** Emotions compute trigger values S and L resulting from game outcomes. Positive emotional states are maximally experienced when their trigger values are largest ($= 1$) and negative emotional states are maximally experienced when their trigger values are smallest ($= 0$). **(1.3)** The weights are up- and down-regulated by arrays of positive and negative emotions (e_i, e_t) produced by self and other. Thus the weights are dynamically updated after being targeted by recalibrational emotions.

Figure 2: Bubble Plot of the Amount Sent and the Amount Returned



Note: Observations were plotted with bubbles, where the relative size indicates the number of observations. The smallest bubble plotted represents one observation and the largest bubble plotted represents eight observations. Colored regions indicate observations that are significantly ($p < .001$) more “conflicted” (where $0 < s \geq r$) and more “activated” (where $0 < s < r$) than among all other observations according to a Wilcoxon rank-sum (Mann-Whitney) test.

Table 1: Specific Classifying Features of Emotional States with Intrapersonal and Interpersonal Targets

Set	Emotional State	Functional Features					
		Facilitating Adaptive Goal(s)		Recalibrational Effect		Recalibrational Target	
		<i>U: Long-sighted</i>	<i>V: Short-sighted</i>	Positive	Negative	Intrapersonal	Interpersonal
1	<i>Appreciative</i>	X	X	X		U	U
	<i>Happy</i>	X	X	X		U,V	U
	<i>Content</i>	X	X	X		U,V	U
	<i>Cheerful</i>	X	X	X		U,V	U
	<i>Triumphant</i>	X	X	X		U,V	
	<i>Inspired</i>	X	X	X		U,V	
	<i>Secure</i>	X	X	X		U,V	
2	<i>Proud</i>	X		X		U	
	<i>Believable</i>	X		X		U†	
3	<i>Surprised</i>	X	X	X	X	U‡	
4	<i>Disgusted</i>	X	X		X	U	V
	<i>Jealous</i>	X	X		X	U	V
	<i>Aggravated</i>	X	X		X	U	V
	<i>Frustrated</i>	X	X		X	U	V
	<i>Angry</i>	X	X		X	U	V
	<i>Depressed</i>	X	X		X	U	V
	<i>Sad</i>	X	X		X	U	V
5	<i>Embarrassed</i>	X			X	V	V
	<i>Ashamed</i>	X			X	V	V
	<i>Guilty</i>	X			X	V	

Note: X's indicate classifying features of emotional states. Cells populated under "Intrapersonal" and "Interpersonal" specify the targets (U = long-sighted program, and V = short-sighted program) in self and others that those emotional states' recalibrations were designed to affect.

† *Believable* intrapersonally targets the long-sighted program only for the Trustee.

‡ *Surprise* may either positively or negatively target the long-sighted program in one's self according to discovery that one has under- or over-expected achievement of the long-sighted program's goal.

Table 2: Details of Confirmatory Factor Analyses

Model	Specification	N	DF	Log-likelihood	Root Mean Square of Approximation [90% CI]	Comparative Fit Index	Bayesian Information Criteria	Difference in fit compared to Model R
1	Independent Valence Model: Zero (0) PA/NA Correlation	169	57	-4,419.15	.144 [.134,.156]	.767	9,130.71	Chi ² (6)= 147.47 p < .001
2	Bipolar Valence Model: Negative (-1) PA/NA Correlation	169	57	-4,371.49	.131 [.120,.142]	.808	9,035.39	Chi ² (6)= 99.81 p < .001
3	Unrestricted Valence Model: Unrestricted PA/NA Correlation	169	58	-4,371.14	.131 [.120,.142]	.808	9,039.82	Chi ² (5)= 99.46 p < .001
R	Recalibrational Model: 5 Factors	169	63	-4,271.68	.099 [.088,.111]	.892	8,866.55	

Note: One observation, where the participant reported the same value for all emotions was dropped.

Table 3: Details of Structural Equation Fit Analyses

Model	N	DF	Log-likelihood	Root Mean Square of Approximation [90% CI]	Comparative Fit Index	Bayesian Information Criteria	Difference in fit compared to Unrestricted Valence
Unrestricted Valence Model	169	62	-4,784.99	.125 [.130,.149]	.799	9,888.03	
Recalibrational Model	169	69	-4,681.98	.099 [.088,.109]	.880	9,717.92	Chi ² (7)=103.01 p < .001

Table 4: Structural Loadings and Equation Level Goodness of Fit for Valence and Recalibrational Models

		<i>Valence Model</i>			<i>Recalibration Model</i>					
<i>Structural</i>		Beta	SE	R ²		Beta	SE	R ²		
PA	S	.202 ***	(.063)	.439	L1	S	.180 ***	(.056)	.427	
	L	.702 ***	(.043)			L	.691 ***	(.044)		
NA	S	-.335 ***	(.065)	.328	L2	L	.644 ***	(.068)	.415	
	L	-.590 ***	(.054)			L4	S	-.348 ***		(.064)
							L	-.589 ***	(.054)	
							L5	L	-.199 **	(.082)
<i>Measurement</i>		Beta/ Intercept	SE	R ²		Beta/ Intercept	SE	R ²		
Appreciative	PA	.835 ***	(.025)	.697	L1	.833 ***	(.025)	.693		
	Constant	1.105				Constant	1.135			
Happy	PA	.919 ***	(.015)	.845	L1	.921 ***	(.015)	.848		
	Constant	1.408				Constant	1.439			
Content	PA	.831 ***	(.026)	.690	L1	.832 ***	(.026)	.693		
	Constant	1.586				Constant	1.614			
Cheerful	PA	.876 ***	(.020)	.767	L1	.873 ***	(.021)	.761		
	Constant	1.023				Constant	1.055			
Triumphant	PA	.795 ***	(.030)	.632	L1	.793 ***	(.030)	.629		
	Constant	1.044				Constant	1.071			
Inspired	PA	.665 ***	(.044)	.443	L1	.657 ***	(.045)	.432		
	Constant	.866				Constant	.893			
Secure	PA	.639 ***	(.047)	.408	L1	.640 ***	(.047)	.409		
	Constant	1.441				Constant	1.461			
Believable	PA	.507 ***	(.059)	.257	L2	.573 ***	(.062)	.328		
	Constant	1.373				Constant	1.418			
Proud	PA	.658 ***	(.045)	.433	L2	.743 ***	(.057)	.553		
	Constant	1.256				Constant	1.314			
Disgusted	NA	.808 ***	(.029)	.654	L4	.800 ***	(.030)	.640		
	Constant	2.107				Constant	2.104			
Jealous	NA	.507 ***	(.059)	.258	L4	.510 ***	(.059)	.260		
	Constant	1.875				Constant	1.880			
Aggravated	NA	.901 ***	(.018)	.812	L4	.907 ***	(.017)	.823		
	Constant	2.188				Constant	2.198			
Frustrated	NA	.885 ***	(.020)	.782	L4	.889 ***	(.019)	.791		
	Constant	2.121				Constant	2.129			
Angry	NA	.870 ***	(.021)	.758	L4	.873 ***	(.021)	.763		
	Constant	2.179				Constant	2.185			
Depressed	NA	.670 ***	(.045)	.449	L4	.660 ***	(.046)	.436		
	Constant	2.129				Constant	2.124			
Sad	NA	.744 ***	(.037)	.554	L4	.737 ***	(.037)	.543		
	Constant	2.114				Constant	2.111			
Embarrassed	NA	.430 ***	(.066)	.185	L5	.679 ***	(.049)	.461		
	Constant	1.743				Constant	1.526			
Ashamed	NA	.115	(.079)	.013	L5	.902 ***	(.035)	.814		
	Constant	1.610				Constant	1.714			
Guilty	NA	.095	(.079)	.009	L5	.816 ***	(.038)	.665		
	Constant	1.496				Constant	1.598			

Note: Standardized beta reported. Equation level R² reported for each dependent variable.
 * indicates statistical significance at $p < .10$, ** significant at $p < .05$, and *** at $p < .01$.