

Journal of Experimental Social Psychology
Computational and Motivational Mechanisms of Human Social Decision Making
Involving Close Others
--Manuscript Draft--

Manuscript Number:	JESP-D-20-00062R2
Article Type:	Full Length Article
Keywords:	social decision-making; close others; motivation; Computational Modeling; discounting
Corresponding Author:	Joao Guassi Moreira, M.A. University of California Los Angeles UNITED STATES
First Author:	Joao Guassi Moreira, M.A.
Order of Authors:	Joao Guassi Moreira, M.A. Sarah M Tashjian, Ph.D., J.D. Adriana Galvan, Ph.D. Jennifer A Silvers, Ph.D.
Abstract:	Every day, human beings make decisions with social consequences. These social consequences matter most when they impact those closest to us. Recent research has shown that humans exhibit reliable preferences when deciding between conflicting outcomes involving close others – for example, prioritizing the interests of one's family member over one's friend. However, virtually nothing is known about the mechanisms that drive these preferences. We conducted a pre-registered study in a large (maximum N =375) sample to quantify the computational and motivational mechanisms of human social decision-making preferences involving close others. By pairing assessment techniques from behavioral economics and psychological science with computational modeling and random coefficient regression, we show that value-based cognitive computations (e.g., risk and loss aversion) drive social decision-making preferences involving financial outcomes, whereas socioemotional motivations (e.g., relationship quality) underlie preferences involving social outcomes. These results imply mechanistic heterogeneity, underscoring a need for greater attention to contextual specificity in social decision-making.
Suggested Reviewers:	Matthew Apps, PhD matthew.apps@psy.ox.ac.uk Luke Chang, PhD luke.j.chang@dartmouth.edu Rob Chavez, PhD rchavez@uoregon.edu Molly Crockett, PhD molly.crockett@yale.edu Patricia Lockwood patricia.l.lockwood@gmail.com Leah Somerville, PhD somerville@fas.harvard.edu
Response to Reviewers:	

UCLA College
Life Sciences
Psychology

November 11th, 2020

Roger Giner-Sorolla, Ph.D.
Editor, *Journal of Experimental Social Psychology*

Dear Dr. Giner-Sorolla,

We are pleased to submit a revised version of our manuscript entitled, "Computational and Motivational Mechanisms of Human Social Decision-making Involving Close Others", for consideration in the *Journal of Experimental Social Psychology*. We appreciate the time and effort that the editor and reviewers put into evaluating our work. We have made every effort to respond as thoughtfully and carefully as possible to their comments in hopes of bettering the manuscript and making it more competitive for publication. We hope that the revisions now render the manuscript suitable for publication, though we are more than willing to make further edits if the reviewers and editor believe they will strengthen the manuscript. Please do not hesitate to contact us with additional questions should they arise.

Thank you,
Joao Guassi Moreira, C.Phil and Jennifer Silvers, PhD

University of California, Los Angeles
Department of Psychology
Franz Hall – 502 Portola Plaza
Los Angeles, CA 90095
jguassimoreira@ucla.edu
silvers@ucla.edu

Revision of JESP-D-20-00062R1

Computational and Motivational Mechanisms of Human Social Decision Making Involving Close Others

General Comments

We appreciate the thoughtful feedback of the reviewers and the editor, and are thrilled to have the opportunity to revise our manuscript. We have made every effort to respond to the comments of the editor and all three reviewers.

We tried to be especially diligent in (i) responding to concerns about sample size, (ii) the interpretation of monetary v social outcomes, and (iii) the benefits of computational modeling, among the other issues raised by reviewers.

The following is a quoted list of concerns from the reviewers and our response to each. Though we tried to be as responsive as possible, we are happy to make additional changes in the editor and reviewers believe it is in the best of interest of producing the strongest possible manuscript.

Overall, we believe our response to the comments below have greatly strengthened the paper and we are confident it is more suitable for publication in the *Journal of Experimental Social Psychology*.

Reviewer 1

1. “The sample sizes for the discounting task and the gambling task separately seem adequate to me, but I am concerned about the limited overlap between them. N = 75 would not be ideal for assessing even a bivariate association. I fear that it is substantially underpowered for testing multiple moderation effects. I think that the authors need to at least acknowledge this as a limitation in the discussion, and refrain from strongly interpreting null results in the corresponding analyses.”

We agree with the reviewer that very large samples are critical for performing statistical inference in individual differences research. However, we note that our use of multilevel logistic regression in modeling decision-level behavior on the discounting tasks affords us more power than may be initially apparent. While we only have N = 75 individuals (level-2 observations), we also had 98 decisions per run of each task (level-1 observations). This is important to note because hypothesis tests regarding individual differences (e.g., the extent to which relationship quality or computational parameters moderated choice behavior on the discounting tasks) involve *cross-level interactions* (e.g., Condition (level-1) x Individual Difference (level-2)). That we have nearly 100 level 1 observations means our power for cross-level interactions receive quite a boost. Simulation studies from quantitative psychology show, generally, 80% power can be achieved with at least 70 level-2 units and 75 level-1 units (Astivia, Gadermann, & Guhn, 2019; Mathieu, Aguinis, Culpepper, & Chen, 2012).

However, for good measure and out of caution, we have included the following statement about power in the *Limitations & Future Directions* section of the manuscript.

Pg. 26-27:

“Additional follow-up work could also focus on directly replicating our individual differences findings with larger sample sizes. While we note that our individual difference analyses were well-powered on the strength of having nearly 100 decision-level (level 1) observations (Astivia, Gadermann, & Guhn, 2019; Mathieu, Aguinis, Culpepper, & Chen, 2012), future work with larger sample sizes would lend increased confidence in the present results by further reducing noise in estimates of random slopes and effect sizes and potentially improving power.”

2. “*In table 1/figure 2, the reward ratio coefficient is significantly positive in the delay-monetary case. Does this mean that in this version of the game, people actually prefer smaller, later amounts of money to sooner, larger amounts? If so, what are we to make of this?*”

We thank the reviewer for their careful attention to our output and for spotting this. There was a clerical error when entering the relevant value into Table 1. This has since been corrected, showing that while positive, the parameter for this effect is not significant. We have verified the rest of the table for accuracy. Figure 2 remains accurate.

3. “*It is difficult to know how to interpret the monetary vs. social results. The authors - perhaps wisely - do not attempt to directly contrast them. However, it is difficult to avoid some comparison, if only informal, rather than statistical. Were the social results weaker because people really think differently about shared time than money? Or was the magnitude of the manipulation stronger in the monetary condition? Although I realize that it is difficult to equate these qualitatively, it might be possible to get some traction on this problem by looking at the point of subjective equivalence between shared time with friends/family, and those same close others receiving money? That is, how much money would you have to give my friend for me to feel like that was worth 20 minutes of our time together? I don't suppose that the authors have any additional data in which participants asked such a question? I'm not sure it would make sense to try to collect new data on this right now (even online) since people' preferences may well have shifted due to the results of covid19. However, if some data along these lines does already exist, it might be possible to use it get a better sense of how to compare time and money.*”

The reviewer is correct in noting that we were hoping to avoid too much comparison between the pattern of results with social and nonsocial outcomes for exactly the reasons described. That said, we realize that we still have an obligation to make thoughtful and careful remarks about results in the social and nonsocial domains and appreciate the reviewer’s thoughts on this matter.

To address this comment, we first ran additional analyses on the discounting tasks as suggested by the reviewer. For a given discounting task (e.g., probabilistic discounting, social outcomes; delay discounting, monetary outcomes; etc.), we fit traditional value-based hyperbolic models on

each condition of the data: trials on which the parent received the discounting outcome/friend received the non-discounting outcome, and trials on which the friend received the discounting outcome/parent received the non-discounting outcome. This allowed us to compare discounting rates between the two conditions.

Afterwards, we took a difference score [parent discounting rate – friend discounting rate] for each discounting task and statistically compared difference scores from social and monetary outcomes (e.g., compared discounting rate difference score in delay discounting with monetary outcomes to discounting rate difference score in delay discounting with social outcomes). These analyses revealed that there were greater differences in delay discounting rates when monetary outcomes were at stake, compared to social outcomes. That is, individuals discounted larger long-term monetary rewards for friends in favor of smaller short-term monetary rewards for parents at a greater rate than the opposite. By contrast, delay discounting rates between the conditions were more equivocal when social outcomes were at stake. There were no significant differences in probabilistic discounting behavior. The manuscript has been updated to reflect these data. The details are presented in the supplement (*Follow-Up Analysis: Comparing Discounting Decision Preferences with Monetary and Social Outcomes*), and the results are summarized in the results section (Pg. 21, **Post-Hoc Analysis: Directly Comparing Behavior when Monetary vs Social Outcomes are at Stake**) of the main document.

To further address this comment, we have added a new paragraph to the discussion where we explicitly compare the monetary rewards to the social rewards. The aforementioned results from additional analyses are referenced there, in addition to other points made by the reviewer. Pg. 23:

“A post-hoc and preliminary statistical comparison provides some evidence to suggest that observed differences between reward types were not due to chance—that parent-oriented preferences with monetary rewards are indeed stronger than parent-oriented preferences with social rewards when making delay discounting decisions (but not probabilistic discounting). This could be due to a number of reasons, ranging from the salience of the social rewards we offered, to task demands related to computing value for money compared to social rewards (it may be more difficult to compute value calculations for ‘time spent’ than dollars), and to the ease with which individuals implicitly compute probabilities when making discounting decisions (potentially explaining the null finding when contrasting decision preferences for monetary and social outcomes in probabilistic discounting). Additional manipulations in future studies can help unpack these effects, perhaps with the help of cross-model reward decisions (e.g., asking individuals to assign a monetary value to social outcomes).”

Last, we respond to the reviewer’s question about additional data existing to further disambiguate our pattern of findings with social vs monetary outcomes by noting that no such data exist. We agree with the reviewer that it is inappropriate to collect more data on this matter since large-scale changes in social dynamics as a result of the COVID-19 pandemic may play a confounding role in swaying social decision preferences.

4. “*The IPPA seems like a reasonable measure of relationship quality, but its items do not seem*

to tap into the notion of obligation or family duty. Many people and cultures view parents and children as owing each other certain obligations, regardless of their quality of their relationship. This is especially true in financial matters, where family are often expected to support one another. This sense of duty might account for some of participants' preference for their parents in the monetary discounting task, and their desire to avoid occasioning them with financial loss gambling task. The authors touch on this on page 22-23. Do any of the other individual difference measures that the authors collected have items that could be used to explore this possibility?"

We agree with the reviewer that, in theory, notions of obligation to parents and friends likely play a role in shaping social decision preferences. In fact, we have pursued this notion in prior studies that also examined social decision-making behavior involving conflicting outcomes for parents and friends. Analyses in said prior studies have yielded inconclusive findings regarding the role that familial obligation plays in shaping social decision preferences (authors, 2018). In our view, the main challenge that lies behind this research endeavor lies in methodology. First, we have yet to find a satisfactory measure of obligation to friends (whereas a valid and reliable analog readily exists for family/parents), making it very difficult to compare the role that relationship obligation plays when making decisions on behalf of multiple close others. Second, and perhaps this is the reason behind the first issue, it is difficult to equate obligation to parents and friends because of intrinsic differences in how cultures construe relationships with kin and friends. One possible albeit indirect way of examining these issues in future work would be to systematically examine whether cultural identity moderates decisions regarding parents and friends. We have added an additional passage to the discussion to convey these thoughts to readers.

Pg. 23-24: "If these differences are replicated, future work may wish to identify why they exist in the first place. We have speculated that obligation to parent and friend relationships most likely plays a large role (authors, 2018), but were unable to find compelling evidence (and no study measures collected here can speak to this notion). Additional work in this vein will need to carefully consider what obligation means in the context of parent and friend relationships, and thoughtfully select (or even devise) methods to sensibly equate the two for quantitative analyses. This difficulty highlights the possibility that qualitative analysis may be useful in resolving this issue while the field attempts to find an appropriate quantitative articulation. Finally, given prior work demonstrating the role that culture plays in feelings of familial obligation (Fuligni & Pedersen, 2002; Tsai, Telzer, Gonzales, & Fuligni, 2015), future work ought to formally examine how cultural differences shape social decision making for parents and friends."

5. "Although I personally am well-disposed towards taking a more computational approach to these questions, I think it might be worth describing their benefits in this context more. That is, what were the benefits of using economic games and computational modeling over simply asking people whether they would prefer to give money (or time) to their parents or to their friends? (Or to gamble with their parents' or friends' money?). A few extra sentences on this point might help better justify the methods to more skeptical readers."

We agree that this could be useful and have added additional justification for the use of computational models to the introduction. Pg. 6:

“We note that our use of behavioral economics tasks in conjunction with computational modeling is a strength of this study for two major reasons. First, models make precise predictions about phenomena in ways that are easily verifiable and generative (i.e., probabilistic). Second, they can do not require introspection, and thus provide complementary insights to information gathered by self-report (Nisbett & Wilson, 1977).”

Reviewer #2:

1. *“I think it might be worth including in the discussion some thoughts about how these results should be interpreted given the demographics in the study. Specifically, my understanding is that there are meaningful gender differences in both value-based decision making and socioemotional attachment. Given that this study had less than 20% male participants, is there reason to be cautious about assuming these results would generalize to men as well as women?”*

We agree with the reviewer that we should have included a note about generalizability given the gender breakdown of the current sample. Despite the fact this study—and others from our group—did not find gender differences, it is nevertheless important to make readers aware of prior work that *has* observed gender differences and to raise the possibility that these results may not generalize. The following has been added to the discussion section Pg. 26:

“In further considering generalizability, it is worth noting that our sample demographics (mostly white and Asian-American, mostly female, mostly late adolescent/young adult) is certainly not representative of all adults in the United States, let alone the world. Future work should strive to examine these effects in other populations, and we caution readers to bear these constraints on generalizability in mind when evaluating our results. This is particularly important to consider given prior work linking gender to differences in value-based decision making (Cardoos et al., 2017; Fancis, Hasan, Park, & Wu, 2015; Zachry, Johnson, & Calipari, 2019), though we observed no such gender differences in the present study.”

Running head: COMPUTATIONS AND MOTIVATIONS OF SOCIAL DECISION-MAKING

Computational and Motivational Mechanisms of Human Social Decision Making Involving
Close Others

João F. Guassi Moreira*, Sarah M. Tashjian, Adriana Galván, & Jennifer A. Silvers*

Department of Psychology, University of California, Los Angeles

***corresponding authors**

A191 Franz Hall
502 Portola Plaza
Los Angeles, CA, 90095-1563
310-794-4953
jguassimoreira@ucla.edu
silvers@ucla.edu

Author Contributions

J. F. Guassi Moreira and J. A. Silvers developed the study concept with input from S. M. Tashjian. J. F. Guassi Moreira, and J. A. Silvers designed the study. Data were collected by J. F. Guassi Moreira. Data were prepared and analyzed by J. F. Guassi Moreira under the supervision of J. A. Silvers. All authors interpreted the results. J. F. Guassi Moreira drafted the first version of the manuscript with extensive input from J. A. Silvers. S. M. Tashjian, A. Galván, and J. A. Silvers all provided critical revisions. All the authors approved the final manuscript for submission.

1 Computations and Motivations of Social Decision-Making

Running head: COMPUTATIONS AND MOTIVATIONS OF SOCIAL DECISION-MAKING

Computational and Motivational Mechanisms of Human Social Decision Making Involving
Close Others

2 Computations and Motivations of Social Decision-Making

Abstract

Every day, human beings make decisions with social consequences. These social consequences matter most when they impact those closest to us. Recent research has shown that humans exhibit reliable preferences when deciding between conflicting outcomes involving close others – for example, prioritizing the interests of one’s family member over one’s friend. However, virtually nothing is known about the mechanisms that drive these preferences. We conducted a pre-registered study in a large (maximum $N=375$) sample to quantify the computational and motivational mechanisms of human social decision-making preferences involving close others. By pairing assessment techniques from behavioral economics and psychological science with computational modeling and random coefficient regression, we show that value-based cognitive computations (e.g., risk and loss aversion) drive social decision-making preferences involving financial outcomes, whereas socioemotional motivations (e.g., relationship quality) underlie preferences involving social outcomes. These results imply mechanistic heterogeneity, underscoring a need for greater attention to contextual specificity in social decision-making.

Key words: social decision-making; close others; motivation; computational modeling; discounting

3 Computations and Motivations of Social Decision-Making

Computational and Motivational Mechanisms of Human Social Decision Making Involving Close Others

As members of a highly social species, humans frequently make decisions that have consequences for other individuals. In real life, our decisions most commonly impact those closest to us, including our friends and family (Henrich, 2002; Mathews, 1987). Indeed, observational research suggests that close others can strongly influence decisions regarding education, employment, health, and finances (Fareri, Smith, & Delgado, 2020; Johnson, 2009; Kirchler, Holzl, Rodler, & Meier, 2001; Lavner, Weiss, Miller, & Karney, 2018). However, the vast majority of social decision-making experiments to date have relied upon paradigms wherein participants make decisions involving a single unfamiliar confederate (Camerer, 2011; Feldmanhall & Chang, 2018; Rilling & Sanfey, 2011). Drawing broad inferences about social decision-making from paradigms with anonymous others is problematic because close relationship status, and even general familiarity, can profoundly influence social behavior (Ahmed & Chloe, 2014; Dunham, 2018; Fareri, Chang, & Delgado, 2015; Guassi Moreira & Telzer, 2018; McKelvey & Kerr, 1988; Meyer & Anderson, 2000; Telzer, Masten, Berkman, Lieberman, & Fuligni, 2010). For example, individuals are inclined to be highly generous to close others, sometimes even at a cost to oneself, but less so to strangers (Lockwood et al., 2017; Powers et al., 2018; Telzer et al., 2010). Consistent with this, when decision makers are forced to choose whose interests to prioritize, they rely heavily on the identity of those involved – for example, young adults are more likely to allocate resources to a parent at the expense of a friend than the contrary (Guassi Moreira, Tashjian, Galván, & Silvers, 2018, 2019). This suggests that social decision preferences may differ depending on whom they affect.

4 Computations and Motivations of Social Decision-Making

At least two issues have yet to be addressed, with regard to how humans make decisions impacting close others. First, it is unknown whether or not individuals consistently prioritize the same close others when different outcomes are at stake (e.g., do I spend both money and time on my parent instead of my friend?). This knowledge gap stems from the fact that prior social decision-making work involving close others has primarily examined decision-making preferences when monetary outcomes are at stake (Fareri et al., 2020; Guassi Moreira et al., 2018, 2019; Powers et al., 2018). Determining how stable social preferences are across different contexts (e.g., social, financial) is a necessary first step in forming comprehensive and unifying theories of social decision-making, which could promote integration between different research lines, generate future scientific predictions, and inspire applications outside of the laboratory.

Second, extant research has failed to tell us the '*why*' behind social decision-making preferences. We do not know why an individual might prioritize one close other over another. Recent research suggests that value-based psychological systems are involved in tracking important social information (e.g., social status) and close-other status strongly modulates the subjective value of rewards (Feng, Zhao, & Donnay, 2013; Hackel, Doll, & Amodio, 2015; Morelli, Chang, Carlson, Kullar, & Zaki, 2018; Parkinson, Kleinbaum, & Wheatley, 2017; Wang, Cheng, Lee, & Chuang, 2019; Zerubavel, Bearman, Weber, & Ochsner, 2015; Zhao, Feng, & Kazinka, 2014). Coupled with the fact that prior studies on social decision-making with close others have found evidence of consistent preferences with monetary outcomes (Guassi Moreira et al., 2019), it is likely that a value-based psychological architecture may be driving these preferences while also leaving open the possibility that non-value-based architectures may drive preferences involving other outcomes (e.g., social motivations).

5 Computations and Motivations of Social Decision-Making

In light of the aforementioned gaps in the literature, our study had two major aims. First, we examined whether parent-over-friend preferences previously observed in the monetary domain would generalize to decision-making contexts that involved social consequences (time spent with close other). We used two forms of discounting (delay: overweighting of smaller, immediate rewards over larger, delayed rewards; probabilistic: overweighting more certain rewards with a smaller expected value over less certain rewards with larger expected values) as our model decision-making paradigm largely due to its ability to accommodate both social and monetary rewards and amenability to fitting models that can account for both within and between subject effects. If prior research generalizes to discounting decisions, we would expect that individuals would temporally and probabilistically discount on trials when a parent benefitted and refrain from doing so when a parent did not. Moreover, if social preferences are domain-general, we expected parents would be prioritized both in decisions with financial and social outcomes.

Our second aim was focused on testing two candidate mechanisms that could potentially drive parent-versus-friend preferences. The first candidate mechanism was value-based computations, the implicit or explicit cognitions that support the subjective appraisal of value. Several frameworks have been proposed to formally articulate value-based computations thought to underlie decision-making (Feldmanhall & Chang, 2018; Glimcher, 2004; Niv & Chan, 2011). While each framework can provide uniquely meaningful information, we elected to adopt prospect theory—a dominant decision-making framework from behavioral economics (Tversky & Kahneman, 1992)—due to its success in characterizing decision-making behavior and ubiquity in the field. Prospect theory suggests computations of value are fueled by attitudes towards loss and risk. Briefly, loss aversion is the extent to which individuals overweight losses relative to

6 Computations and Motivations of Social Decision-Making

gains. Risk aversion is the extent to which diminishing sensitivity to marginal rewards reduces one's subjective valuation and engenders “safe”, risk averse behavior. Our focus on these two parameters was motivated by the facts that they both inform subjective computation of value and are sensitive to social cognitive demands such that they can be flexibly manipulated to support broader goals (Sokol-Hessner et al., 2009). In the present study we suspected that differences in social decision-making behaviors among close others—at both the group and individual level—could be mechanistically driven by differences in how averse individuals are to loss and risk for different close others, in turn influencing the subjective value computations for each close other. We note that our use of behavioral economics tasks in conjunction with computational modeling is a strength of this study for two major reasons. First, models make precise predictions about phenomena in ways that are easily verifiable and generative (i.e., probabilistic). Second, they can do not require introspection, and thus provide complementary insights to information gathered by self-report (Nisbett & Wilson, 1977).

The second candidate mechanism tested was socioemotional motivations. Socioemotional motivations were operationalized as reported relationship quality with each close other, under the assumption that individuals are motivated to maintain high quality relationships (Gable & Impett, 2012). Support for this mechanism would suggest that social preferences are not driven by differences in appraisals of value, but instead rely on the fulfillment of socioemotional goals. Finally, we also pursued an exploratory aim that tied the former two mechanisms together: assessing whether individual differences in candidate mechanisms tracked with individual differences in social decision-making preferences.

Methods

Study Overview & Hypotheses

In the current study, 225 participants completed discounting tasks involving two close others, a nominated parent and close friend, to probe social decision-making preferences. Decisions were compared under four conditions, a crossed design with two types of discounting—probabilistic, delay—and two types of outcomes—social, monetary. We employed two types of discounting and two types of outcomes to enhance the generalizability of our findings. Value-based computational (or simply, ‘computational’) mechanisms (risk and loss aversion, as defined by prospect theory) were derived from behavior on an independent gambling task that a set of partially overlapping (overlap: $N = 75$) participants completed on behalf of a parent and friend (separate runs for each, $N = 225$). All participants ($N = 375$) additionally completed a self-report measure of relationship quality to assess socioemotional motivational mechanisms of social decision-making preferences (Inventory of Parent and Peer Attachment; IPPA (Armsden & Greenberg, 1987)). Hypotheses, methods, and analyses were pre-registered on the Open Science Framework prior to the beginning of the study. See the Methods section for more details. Our hypotheses follow ($H1$ and $H2$ were pre-registered, $H3$ was not).

H1. Based on prior research (Guassi Moreira et al., 2018), participants will favor parents over friends during decision-making that involves discounting across two outcome domains (social and monetary). As noted in the methods, we operationalize a preference for a given close other as discounting when it benefits that individual and refraining from discounting when it does not benefit the individual.

H2. If decision-making preferences (e.g., prioritizing parents over friends) are informed by value-based computations, then participants will show group-level differences in loss and risk

8 Computations and Motivations of Social Decision-Making

aversion parameters when making decisions on behalf of a parent and close friend (i.e., participants will have different loss and/or risk aversion values for parents compared to friends).

H3. If socioemotional motivations are consequential for social decision-making preferences, then participants will show group-level differences in self-reported relationship quality (our measure of said mechanism) with a parent and close friend (i.e., participants will report different levels of relationship quality for parents compared to friends).

We had additional, exploratory aims to relate individual differences in computational and motivational mechanisms to decision-making preferences (both aims were pre-registered).

A1. Determine whether individual differences in computational mechanisms (loss aversion, risk aversion) track with individual differences in decision-making preferences.

A2. Determine whether individual differences in motivational mechanisms (relationship quality) track with individual differences in decision-making preferences.

Participants and Sampling Strategy

Participants were recruited from the metropolitan area of a large university in the Western United States via the undergraduate psychology subject pool. This study was part of a broader data collection effort aimed at understanding social decision-making processes involving close others—more information is given our pre-registration document (https://osf.io/6278m/?view_only=3264b80da15b44cca983e3c45e1f8e6b). For this study, we selected a subset of participants who either completed one of two computerized tasks, or both of them. N = 225 participants completed a task meant to index social decision-making preferences; N = 225 completed a task aimed at capturing computational mechanisms underlying social decision-making preferences; N = 75 completed both tasks. Thus, our sample size for the study

9 Computations and Motivations of Social Decision-Making

was set at 375 total participants. The sample size for these cells were determined *a priori*. In setting our sample size, our goal was to recruit enough participants for well-powered tests for each task while also having a minimum number to examine cross-task, exploratory correlations (see preregistration document for further details). Two participants were excluded from analyses for non-compliance and one for being unable to nominate an appropriate close other. Thus, our final total sample was comprised of $N = 372$ (71 males, Mean age = 20.32 years, SD = 1.57, range = 18-29). Ethnically, 76 participants identified as Hispanic/Latinx whereas the remainder of the sample did not. Racially, 153 participants identified as Asians (41.1%), 119 identified as Caucasian (32%), 11 identified as African American (3%), 0 identified as Pacific Islander (0%), 1 identified as Native American (0.3%), 68 considered themselves to be another race, or mixed race (18.3%), and the remaining participants declined to respond about their race. Participants provided written consent in accordance with the policies of the local Institutional Review Board, and were compensated with course credit for their time. Data, code, and materials are publicly available on the Open Science Framework (OSF; https://osf.io/d42ar/?view_only=3264b80da15b44cca983e3c45e1f8e6b).

Procedure

Upon arriving to the laboratory participants provided informed consent, nominated a parent and close friend of their choice, underwent a manipulation to increase the salience of their nominated parent and friend, took a survey containing several self-report measures of interest, and then completed one of nine possible pairings of two computerized tasks. Though all nine pairings were part of a broader data collection effort concerning social decision-making, we only report data from two tasks here (other tasks have yielded results published elsewhere (blinded, 2019), or have not yet been analyzed). An experimenter extensively trained participants on how

10 Computations and Motivations of Social Decision-Making

to complete each task, and proceeded with the experiment only after the experimenter judged the participant adequately understood the instructions. The experimenter then unobtrusively observed each participant as they completed the experiment to ensure compliance. Key measures involved in the current report are described in detail below. Full measures are disclosed in the supplement and available on our OSF page (https://osf.io/6278m/?view_only=3264b80da15b44cca983e3c45e1f8e6b).

Parent—Friend Nomination & Salience Manipulation.

At the beginning of the study, participants were told they would be completing hypothetical decisions that would affect a parent and close friend, and that they would be required to choose the parent and close friend discussed in the scenarios. Participants were told to nominate any parent and any close friend that satisfied the following conditions: both individuals were still alive, the friend was not a current romantic partner, the friend was not a family member (e.g., sibling, cousin), and the friend was still an active friend (e.g., a friend who they no longer kept in touch with was not allowed). Afterwards, we asked participants to write down basic information about each person (e.g., name, age, sex), a memory they had with each person, and a handful of words and phrases describing each person. The memory manipulation was used to enhance the likelihood that close others would be salient in participants minds while making hypothetical decisions about them (Guassi Moreira et al., 2018).

Experimental Tasks and Measures

Overview.

At least one of the following two tasks was administered to all participants in the current study. A subset of participants completed both sets of tasks. Participants who did not complete

11 Computations and Motivations of Social Decision-Making

both tasks described here completed additional tasks, but those data have been reported elsewhere (the full list of tasks is available online in our pre-registration document). All tasks were programmed in PsychoPy (v1.90.3) and run on PCs running Windows 10; all surveys were administered via Qualtrics. Though all rewards (monetary or social) were hypothetical, participants were carefully and repeatedly instructed to complete the tasks as if the rewards were truly real. The use of hypothetical rewards here was virtually necessary for social rewards, as we could not force participants to actually spend time with a parent or friend of their choice

Discounting Task. Social decision-making preferences were assessed using a suite of discounting decisions. We chose to focus on discounting for several reasons. First, we hoped to expand prior work demonstrating a preference for parents over friends in risk taking and probabilistic learning contexts (Guassi Moreira et al., 2018, 2019) by assessing whether this preference generalized to an entirely different kind of decision-making context. Second, discounting decisions are thought to be heavily consequential for wellbeing and adjustment outcomes (e.g., 12) since they simulate value-based trade-offs that individuals often encounter in everyday life. Last, discounting tasks are flexible and allow for modeling behavior across many different reward outcomes (Seaman et al., 2016).

Participants completed discounting decisions in a fully crossed 2 X 2 design that yielded four conditions: one dimension varied in terms of the type of discounting (delay and probabilistic) and the other varied in terms of the type of hypothetical reward (monetary and social). Regardless of condition, participants were presented with two scenarios and were told one would affect their parent and the other would affect their friend. One scenario involved a more immediate or certain reward, and another involved a more delayed or uncertain reward. Half of the trials in each run involved the parent receiving the former and the friend receiving the

12 Computations and Motivations of Social Decision-Making

latter, while the other half involved the opposite. In the delay discounting conditions, participants could choose between a relatively immediate reward or a larger delayed reward—the time delays could take the value of zero (i.e., ‘Today’), 2 weeks, 4 weeks, and 6 weeks. In the probabilistic discounting conditions, participants could choose between a relatively more certain reward or a larger, more uncertain reward—the certainties could take the value of 100% (i.e., no uncertainty), 75%, 50%, and 25%. Reward values ranged from 2 to 30 in all tasks. In the monetary rewards condition, reward values were treated as being equivalent to US Dollars; in the social reward condition, rewards were treated as time spent with either close other (e.g., 16 minutes spent with one of the nominated close others). There is an inherent asymmetry to this design (monetary rewards are *won for* a close other, social rewards are *shared with* them), but attempts to equate the two could introduce other, more serious confounds (e.g., participants could have made social decisions that involved allowing parent/friend to spend time with another loved one, but that is unlikely to represent a meaningful preference). For example, on the delay discounting task involving monetary rewards, one might choose between earning \$6 for their parent now or earning \$14 for their friend in four weeks. In this particular case, choosing \$6 for a parent, thereby discounting the \$14 reward for a friend, would be consistent with a parent preference. Further, someone with such a preference would make the opposite choice when the parent-friend labels were reversed (e.g., choosing \$14 for a parent in four weeks over \$6 for a friend now). We assume the same for the social condition—a parent preference is indicated by selectively discounting when the parent benefits, and refraining from discounting when it also benefits the parent (the opposite pattern would be indicative of a friend pattern). The exact values for each trial were adapted from an open resource (osf.io/bths8/). Participants completed 98 trials per run and the runs were self-paced (although participants were told not to dwell on any

13 Computations and Motivations of Social Decision-Making

individual choice for too long). Figure 1 (left panel) depicts a schematic of the task. Although all rewards were hypothetical, it was heavily stressed to participants they were to complete the task as if the rewards were real.

Gambling Task. In order to understand the computational mechanisms of social decision-making preferences (i.e., an individual's tendency to prioritize their parent or friend), we employed a simple binary gambling task in which participants played for hypothetical monetary rewards. Data from this task are widely used in the behavioral economics literature as a means to quantify subjects' loss and risk aversion (Sokol-Hessner et al., 2009; Sokol-Hessner, Raio, Gottesman, Lackovic, & Phelps, 2016). During the task, participants are required to make 150 binary decisions between a certain guaranteed reward and a gamble with two potential outcomes (50% chance each). Most trials (120) involved a gamble whose two potential outcomes were a positive amount and a negative amount (amounts varied across trials), compared to a guaranteed option of zero dollars. A subset of trials (30) involved a gamble whose two potential outcomes were a positive amount and zero dollars, compared to a guaranteed option of a smaller positive amount. See Figure 1 (right panel) for a schematic. The inclusion of these two trial types in the task allowed us to tease apart attitudes towards risk and loss¹, each of which involve unique underlying computations. Probabilities were made explicit to participants during training. Trial order was randomly selected across participants. Participants were given unlimited time to make their decision on each trial (but were encouraged to not dwell too long on any individual choice), and decision outcomes were presented for 1500ms. Monetary values for each trial were used from a prior study with open data (Sokol-Hessner et al., 2015; osf.io/i5knh/). Participants

¹ Loss and risk aversion tend to be highly correlated. As a result, gain and loss trials are necessary to estimate each parameter in the same dataset (See Sokol-Hessner et al., 2009).

14 Computations and Motivations of Social Decision-Making

completed two runs of the task, one in which they were instructed to make decisions as if their nominated friend were to be affected and another as if their nominated parent was to be affected. Although decisions involved hypothetical monetary outcomes, it was heavily stressed that participants were asked to complete the task as if their parent and friend stood to actually gain or lose based on the outcomes of the task.

Inventory of Parent & Peer Attachment (IPPA). We operationalized social motivations using a measure of relationship quality with parents and friends. We specifically used relationship quality to index motivational mechanisms contingent upon the notion that individuals are motivated to maintain high quality relationships. We hypothesized that relationship quality might serve as a prospective motivational mechanism behind social decision-making. As we had done in prior, related studies, relationship quality with parents and friends was assessed using the Inventory of Parent and Peer Attachment (IPPA; Armsden & Greenberg, 1987). The IPPA has been widely used to assess relationship quality (e.g., Fanti, Henrich, Brookmeyer, & Kuperminc, 2008) and was initially validated in college-age samples. Using a 5-point Likert scale (1 = *almost never or never*, 5 = *almost always or always*), participants answered 28 items about their relationships with their nominated parent (example item: “My parent respects my feelings”), and 25 items about their relationship with their nominated friend (example item: “When we discuss things, my friend considers my point of view”). Responses for parents and friends were reverse scored as needed and averaged to yield a single mean score of parent and friend relationship quality, respectively. Model based reliability statistics show that our administration of both parent and friend scales are internally consistent and that composites appropriately capture multidimensionality (Parent: $\omega = .96$, ω -hierarchical = .76; Friend: $\omega = .96$,

15 Computations and Motivations of Social Decision-Making

ω -hierarchical = .71). A list of additional self-report measures that were collected can be found online in our pre-registration document as well as the Supplement.

Modeling

Discounting. Data from all four runs of the discounting task (one for each condition, delay discounting with monetary rewards, delay discounting with social rewards, probabilistic discounting with monetary rewards, and probabilistic discounting with social rewards) were analyzed in four separate random coefficient regression models using the HLM for Windows computer program (Raudenbush & Byrk, 2002). We modeled trial-level decisions nested within participants using the following equation.

$$\text{Logit}(\text{Decision}_{it}) = \pi_{0i} + \pi_{1i}(\text{Condition}_{it}) + \pi_{2i}(\text{RewardRatio}_{it}) + \varepsilon_{it}$$

Here we modeled the log odds of the t -th decision from the i -th participant (1 = discount, = 0 non-discount) as a function of condition (0 = Friend associated with discounting choice, parent associated with non-discounting choice; 1 = parent associated with discounting choice, friend associated with non-discounting choice) (note that this condition parameter is different from the conditions of the discounting tasks, e.g., delay monetary), and a grand-mean centered reward ratio (a ratio of the non-discounting option over the discounting option—greater values indicated a greater difference between the more delayed/more uncertain reward and the more immediate/more uncertain reward). The π_{0i} parameter represents the intercept (i.e., when expected log-odds of making a discounting choice when condition = 0 and when RewardRatio is at the grand mean), the π_{1i} parameter is the adjusted logit(odds ratio) of choosing to discount when a parent's outcome is associated with the discounting option (and friend's outcome is associated with non-discounting option) compared to the opposite, and the π_{2i} parameter

16 Computations and Motivations of Social Decision-Making

represents the expected change in the log odds of choosing to discount given a 1 unit increase in the RewardRatio (non-discounted reward option divided by the discounted reward option), over and above the effect of condition. The Condition variable is coded in such a way that a positive value indicates that individuals favor parents over friends (e.g., individuals are more likely to discount when a parent benefits at the expense of a friend, less likely to discount when a friend benefits at the expense of a parent), a negative value indicates individuals favor friends over parents, and a value of zero indicates no preference. Inclusion of the RewardRatio parameter is meaningful, because it (i) controls for lower level features of the task and (ii) can indicate whether individuals are generally paying attention to the task (i.e., greater RewardRatio should be related to decreased discounting behavior, consistent with previous work (Ludwig et al., 2019)). Notably, these parameters were allowed to vary randomly across participants (i.e., the level 2 units). Results in this modeling framework are from the population average solution with robust standard errors (modeled over dispersion).

We ran additional models that included between subject predictors, namely variables aimed at capturing underlying computational and motivational mechanisms of decision-making preferences. More information is described in the ‘Analysis Plan’ section.

Gambling. Computational mechanisms of social decision-making were modeled from the gambling task data by fitting Sokol-Hessner and colleague’s models derived from prospect theory (Sokol-Hessner et al., 2009, 2015, 2016; Tversky & Kahneman, 1992). The subjective utility ($u(x)$) of objective rewards during the task was calculated using the following piecewise functions.

$$u(x^+) = p(x^+) * (x^+)^{\rho}$$

17 Computations and Motivations of Social Decision-Making

$$u(x) = -\lambda * p(x) * (-x)^\rho$$

The two equations model the subjective utilities of gains and losses, respectively. The objective monetary amount is represented by x , $p(x)$ represents the probability of the objective monetary amount (1 for the guaranteed option, .5 for the gamble), ρ is the risk aversion parameter (1 indicates risk neutrality, < 1 indicates risk aversion for gains, risk seeking for losses, > 1 indicates risk seeking for losses, risk aversion for gains), and λ is the loss aversion parameter (1 indicates gain-loss neutrality, < 1 indicates gain seeking, > 1 indicates loss aversion). Subjective utility values were translated into choice probabilities using the SoftMax function.

$$p(\text{gamble}) = (1 + e^{-\mu * [u(\text{gamble}) - u(\text{guaranteed})]})^{-1}$$

Where e is Euler's number and μ is a choice consistency parameter. The latter term describes whether the relationship between subjective value results choice behavior is stochastic or deterministic (greater parameter values indicate more deterministic decision-making). This equation was adapted into a likelihood function in order to perform maximum likelihood estimation. The `optim()` function in R was used to perform constrained optimization (Byrd, Lu, Nocedal, & Zhu, 1995) on the likelihood function using starting parameters obtained via grid search. Loss and risk aversion values were constrained between 0 and 10, whereas choice consistency parameters were constrained to fall between 0 and 35.

Results

Analysis Plan

Every step of the analysis plan was decided *a priori* unless noted otherwise (i.e., ‘post-hoc’). We first ran four random coefficient regression models on the discounting data to conceptually replicate prior work demonstrating that young adults prioritize parents over friends. Specifically, because prior work only examined decision-making with monetary consequences, we sought to test in the present study whether young adults would also prioritize parents over friends when social outcomes were at stake. Afterwards, we examined possible computational mechanisms underlying these social preferences. To do this, we tested whether loss and risk aversion parameters, as well as gambling decisions (post-hoc), differed when playing for a parent or a close friend. We then examined socioemotional motivational mechanisms underlying social decision-making preferences by testing whether relationship quality was greater for a parent or friend. For our exploratory aims, we tested whether individual differences in loss and risk aversion parameters (and number of gambling decisions, a post-hoc analysis), in addition to individual differences in relationship quality, could predict decision-making preferences. Finally, we also conducted a post-hoc test that formally compared discounting preferences between social and monetary outcomes.

Results Related to H1: Social Decision-making Preferences for Monetary and Social Consequences during Decision-making. The condition parameter (variable describing who was affected by a discounting decision, see methods; π_{Ii}) was positive and significant for the probabilistic and delay discounting runs involving monetary rewards, indicating that participants were more likely to favor their parent over a friend. Table 1 lists the coefficient values and associated standard errors for each discounting condition’s random coefficient regression model. Participants were 34.6% more likely to discount (versus not discounting) when a parent benefited from the relatively certain monetary option and a friend benefited from the relatively uncertain

19 Computations and Motivations of Social Decision-Making

monetary option compared to the opposite; participants were 35.5% more likely to discount (versus not discounting) when a parent benefited from the relatively immediate monetary option and a friend benefited from the relatively delayed monetary option compared to the opposite. The condition parameter was marginally significant for the probabilistic discounting run with social rewards (17.0% effect size), and was not significant for the delay discounting run with social rewards (13.7% effect size). These results showed that individuals were on average more likely to favor parents over friends, and these trends were significant for the probabilistic and delay discounting runs with monetary rewards, and marginally significant with the probabilistic discounting run with social rewards. Figure 2 visualizes these coefficients and their respective standard errors. These results conceptually replicate prior work indicating that young adults prioritize parents over friends in monetary decisions and partially support hypothesis *H1*. Specifically, they support *H1* as related to decisions involving monetary outcomes, but not necessarily when they involve social outcomes.

Results Related to H2: Computational Underpinnings of Social Decision-making Preferences.

Participants were more likely to choose to gamble when playing for their friend than when playing on behalf of a parent (Parent $M(SD)$: 55.32 (25.94), Friend $M(SD)$: 60.06 (27.43); $t(221) = -3.319, p = .001$, Cohen's $d = 0.223$, Pearson's $r = 0.684$). In terms of risk aversion (intolerance of risk arising due to diminishing sensitivity to marginal rewards), participants exhibited comparable levels of risk aversion when playing for parents and friends (Parent $M(SD)$: 1.34 (1.38), Friend $M(SD)$: 1.44 (1.65); $t(221) = -0.849, p > .250$, Cohen's $d = 0.053$, Pearson's $r = 0.223$). By contrast, participants tended to be more loss averse (overweighting of losses relative to gains) when their decisions affected a parent compared to when their decisions affected a friend (Parent $M(SD)$: 2.64 (2.51), Friend $M(SD)$: 2.40 (2.35); $t(221) = 1.798, p =$

.074, Cohen's $d = 0.117$, Pearson's $r = .646$). These results garner partial, modest support for hypothesis $H2$ in that they suggest a difference in loss aversion, but not risk aversion, explains parent-over-friend preferences in decision-making.

Results Related to H3: Motivational Underpinnings of Social Decision-Making Preferences.

Though self-reported relationship quality with parents and friends was high overall, participants reported relatively higher relationship quality with their nominated friend ($M(SD)$: 4.32 (0.44)) compared to their nominated parent ($M(SD)$: 3.92 (0.66); $t(371) = 11.488$, $p < .001$, Cohen's $d = .596$, Pearson's $r = 0.318$). These results strongly support hypothesis $H3$, since $H3$ predicted a difference in relationship quality between parents and friends. However, we note the direction of the effect is the opposite of the direction the significant effects related to $H1$ and $H2$ (and is consistent with prior related work (Guassi Moreira et al., 2018)).

Results Related to Exploratory A1: Individual Differences in Computational Biases Shape Decision-making with Monetary Outcomes. As shown in Table 2A, we observed that a greater propensity to gamble with hypothetical rewards for friends was related to an increased propensity to favor *parents* during both discounting conditions for monetary, but not social rewards (aim $A1$). Effects in a similar direction emerged for parents, but did not exceed our threshold for statistical significance. Afterwards, we found that individuals who were less risk averse for parents (for gains) tended to be more likely to favor them in the monetary discounting conditions. A similar trend emerged for friends (less risk aversion for friends was related to a greater propensity to favor them) during probabilistic discounting for monetary rewards. There was no relationship between metrics of loss aversion and social decision-making preferences (see supplement). Supplementary Figure 1 (top) visualizes these results.

Results Related to Exploratory A2: Individual Differences in Motivational Biases Decision-making with Social Outcomes. Parent and friend relationship quality scores from the IPPA were used as between-person predictors in random coefficient regression models of discounting (grand mean centered, aim A2). Parent relationship quality predicted trial-by-trial decision-making behavior for the discounting runs involving social rewards – participants with greater relationship quality with their parents were even more likely to favor a parent over a friend, and those with reduced relationship quality were less relatively less likely to favor a parent over a friend (Table 3). A similar pattern was present across all discounting tasks with friend relationship quality. However, despite non-trivial effect sizes, these friend relationship quality results did not reach statistical significance. Supplementary Figure 1 (bottom) visualizes these results.

Post-Hoc Analysis: Directly Comparing Behavior when Monetary vs Social Outcomes are at Stake. After observing the results described above, we opted to directly comparing social decision preferences between parents and friends when different types of outcomes were at stake. A supplementary, follow-up analysis employing traditional computational models of discounting (i.e., hyperbolic discounting rates; Burns et al., 2020; Seaman et al., 2018) found some evidence to indicate that parent-over-friend discounting preferences were modestly stronger when monetary, compared to social rewards, were at stake for delay discounting decisions (Cohen's $d = .12$). No such effect was observed for probabilistic discounting decisions (Cohen's $d = .06$). Analytic details and full statistical output is provided in the Supplement.

Discussion

22 Computations and Motivations of Social Decision-Making

Every day humans make decisions that affect close others. The present study examined how individuals make these decisions and characterized the mechanisms that drive social decisions about close others. Consistent with prior work (Guassi Moreira et al., 2018, 2019), the present study found that young adults favor their parents over friends when decisions have financial consequences. However, these same participants were more equivocal when deciding whether to prioritize parents or friends when making decisions with social consequences. In evaluating potential mechanisms, computational models revealed that individuals exhibited more loss aversion for parents compared to friends. Individuals who demonstrated greater risk aversion for their parent, however, were more likely to prioritize parents in decisions with monetary consequences, suggesting value-based computations drive social decisions about financial resources. Conversely, participants reported greater relationship quality with friends than parents when social motivations were assessed, and the magnitude of this motivational bias tracked with one's likelihood of prioritizing a friend in decisions with social outcomes. Importantly, computational biases did not track with preferences when social outcomes were at stake and vice versa, underscoring the specificity of each mechanism.

The present findings suggest that value-based computations and socioemotional motivations guide different types of social decisions. This pattern of results suggests social goals are context-dependent and can dynamically shift decision-making cognitions and behaviors. Parents and friends serve different functions in the lives of young adults (Hopmeyer & Medovoy, 2017; Roisman, Masten, Coatsworth, & Tellegen, 2004) and social obligations are likely to vary as a result. Preferences for parents in situations with financial outcomes may thus be driven by a need to contribute or reciprocate (Fuligni, 2018). By contrast, the apparent lack of a parent or friend preference in decisions with social outcomes (i.e., time spent with close other) may be due

23 Computations and Motivations of Social Decision-Making

to the fact that some individuals prefer spending time with their parent and others with their friend, leading to a net zero preference at the group level. It is also important to consider that time spent with a close other is not necessarily synonymous with the concept of a social preference. It is possible that stronger social preferences would emerge if different social rewards were considered (i.e., engaging in specific social activities with different close others), or if social preferences were evaluated at different stages of development (e.g., children or middle-aged adults).

A post-hoc and preliminary statistical comparison provides some evidence to suggest that observed differences between reward types were not due to chance—that parent-oriented preferences with monetary rewards are indeed stronger than parent-oriented preferences with social rewards when making delay discounting decisions (but not probabilistic discounting). This could be due to a number of reasons, ranging from the salience of the social rewards we offered, to task demands related to computing value for money compared to social rewards (it may be more difficult to compute value calculations for ‘time spent’ than dollars), and to the ease with which individuals implicitly compute probabilities when making discounting decisions (potentially explaining the null finding when contrasting decision preferences for monetary and social outcomes in probabilistic discounting). Additional manipulations in future studies can help unpack these effects, perhaps with the help of cross-model reward decisions (e.g., asking individuals to assign a monetary value to social outcomes). If these differences are replicated, future work may wish to identify why they exist in the first place. We have speculated that obligation to parent and friend relationships most likely plays a large role (authors, 2018), but were unable to find compelling evidence (and no study measures collected here can speak to this notion). Additional work in this vein will need to carefully consider what obligation means in the

context of parent and friend relationships, and thoughtfully select (or even devise) methods to sensibly equate the two for quantitative analyses. This difficulty highlights the possibility that qualitative analysis may be useful in resolving this issue while the field attempts to find an appropriate quantitative articulation. Finally, given prior work demonstrating the role that culture plays in feelings of familial obligation (Fuligni & Pedersen, 2002; Tsai, Telzer, Gonzales, & Fuligni, 2015), future work ought to formally examine how cultural differences shape social decision making for parents and friends.

Group-level results suggest that young adults exhibit value-based computational biases for their parents and social motivational biases for their friends, indicating at least two mechanistic pathways by which social decisions are generated. Exploratory analyses indicated that individual differences in computational and motivational variables predict decision-making preferences, thereby revealing the plausibility and subsequent significance of each mechanism.

In terms of computational mechanisms, individuals on average exhibited slightly greater loss aversion for parents versus friends and individual differences in risk aversion predicted parent-over-friend preferences for monetary rewards. This shows that an individual could be presented with a choice containing the *same* objective return and uncertainty for a parent and friend, yet subjectively value the choice differently. As such, this indicates higher-order social preferences are built upon basic psychological computations, such as risk assessment, that vary as a function of social context (Tamir & Thornton, 2018). That individual differences in patterns of subjective valuation were systematically related to social decision-making preferences suggests that social goals act as a prism—the same input (i.e., contextual decision features) may shine in, and multiple actionable value judgments emerge (i.e., computed subjective value), each driving a different pattern of behavior. Crucially, this implies that a *nested* and *unified*

25 Computations and Motivations of Social Decision-Making

framework for social decision-making may exist, where behavior is motivated by the arbitration of several systems that are differentially weighted depending on numerous features of the situation (Sims, 2018). Indeed, this may partially explain why we observed a main effect of loss aversion while risk aversion was systematically related to individual differences in decision preferences (the decision-making context in the discounting tasks emphasized features salient to risk aversion, whereas other scenarios might have underscored individual differences in loss aversion). While we cannot yet formalize such a framework, these results suggest that context-dependent loss and risk calculations underlie social decision making.

With regard to motivational mechanisms, we found that individuals on average reported greater relationship quality with friends compared to parents, while individual differences in relationship quality predicted decision-making preferences when social, but not monetary, outcomes were at stake. The pattern of preferences observed in decisions with social outcomes is a departure from what we observed in this study and others with monetary outcomes (Guassi Moreira et al., 2019). That decision preferences were sensitive to contextual features indicates that social goals do not manifest uniformly across all contexts. An apparent need to contribute or reciprocate may manifest as favoring a parent over a friend when monetary, but not social, rewards are at stake. Prior work indicates that friendships are predicated upon the need to feel understood by someone or having someone with whom to share thoughts (Arnett, 2000; Hopmeyer & Medovoy, 2017; Kaniušonytė & Žukauskienė, 2018). As such, choosing to prioritize a friend over a parent during decisions involving social rewards may represent the pursuit of these social goals. This possibility is further supported by our individual differences finding that relationship quality with one's friend moderated decision-making behavior regarding social rewards. Together, these results show that decision-making involving social outcomes is

swayed by motivational, but not computational, biases, and possibly a different set of social goals than decisions with monetary outcomes. Given that a vast range of social rewards exist in the real world, it is possible – and ought to be addressed in future work – that a different pattern might have emerged with different kinds of social rewards (e.g., social capital).

Limitations and Future Directions. The present study illustrates how social preferences are shaped by contextual features of decision-making scenarios, as well as individual differences in computational and motivational biases. Though these notions represent exciting avenues for future social decision-making research, direct and conceptual replications are needed to further enhance confidence in these conclusions. To fully test the boundary conditions of our findings, future work ought to examine decisions about different close others (e.g., romantic partners), using other decision-making tasks, and in different developmental and cultural populations. These efforts would help identify the generalizability of the present results and flesh out how social goals influence lower level psychological processes to influence decision-making behavior. In further considering generalizability, it is worth noting that our sample demographics (mostly white and Asian-American, mostly female, mostly late adolescent/young adult) is certainly not representative of all adults in the United States, let alone the world. Future work should strive to examine these effects in other populations, and we caution readers to bear these constraints on generalizability in mind when evaluating our results. This is particularly important to consider given prior work linking gender to differences in value-based decision making (Cardoos et al., 2017; Fancis, Hasan, Park, & Wu, 2015; Zachry, Johnson, & Calipari, 2019), though we observed no such gender differences in the present study. Additional follow-up work could also focus on directly replicating our individual differences findings with larger sample sizes. While we note that our individual difference analyses were well-powered on the strength

27 Computations and Motivations of Social Decision-Making

of having nearly 100 decision-level (level 1) observations (Astivia, Gadermann, & Guhn, 2019; Mathieu, Aguinis, Culpepper, & Chen, 2012), future work with larger sample sizes would lend increased confidence in the present results by further reducing noise in estimates of random slopes and effect sizes and potentially improving power. Replication in larger samples would lend increased confidence to the current individual differences results by reducing the likelihood that factors tangentially related to sample size (e.g., variability of random slopes, skew of level 2 variables, etc.) were driving the results here.

Concluding Remarks. The present results showed that social decision preferences are not necessarily conserved when different types of outcomes (monetary versus social) are at stake. This was highlighted by the facts that parent relationships tended to be favored over friend relationships when monetary outcomes were involved, whereas decision preferences were more equivocal when social outcomes are at stake. The present results also shed light on the computational and socioemotional underpinnings of social decision preferences, with the former influencing individual differences when monetary outcomes were involved and the latter influencing individual differences when social outcomes were at stake. These results emphasize the importance of delineating social decision behavior across multiple contexts, reinforcing the notion that there is no ‘one-size-fits-all’ heuristic for social decision preferences while hopefully paving the way for unifying theories of decision-making behavior.

Open Practices

All data and materials for this study, in addition to its pre-registration, can be accessed at the Open Science Framework
(https://osf.io/6278m/?view_only=3264b80da15b44cca983e3c45e1f8e6b &
https://osf.io/d42ar/?view_only=3264b80da15b44cca983e3c45e1f8e6b).

References

- Ahmed, T., & Chloe, C. (2014). One foot out of the nest: How parents and friends influence social perceptions in emerging adulthood. *Journal of Adult Development*, 147–158.
<https://doi.org/10.1007/s10804-014-9187-9>
- Armsden, G. C., & Greenberg, M. T. (1987). The inventory of parent and peer attachment: individual differences and their relationship to psychological well-being in adolescence. *Journal of Youth & Adolescence*, 16(5), 427–454. <https://doi.org/10.1007/BF02202939>
- Arnett, J. J. (2000). Emerging adulthood: A theory of development from the late teens through the twenties. *American Psychologist*, 55(5), 469–480. <https://doi.org/10.1037/0003-066X.55.5.469>
- Burns, P., Fay, O., McCafferty, M.-F., McKeever, V., Atance, C., & McCormack, T. (2020). Examining children's ability to delay reward: Is the delay discounting task a suitable measure? *Journal of Behavioral Decision Making*, 33, 208–219.
<https://doi.org/10.1002/bdm.2154>
- Byrd, R. H., Lu, P., Nocedal, J., & Zhu, C. (1995). A limited memory algorithm for bound constrained optimization. *SIAM Journal on Scientific Computing*, 16(5), 1190–1208.
- Camerer, C. F. (2011). *Behavioral game theory: Experiments in strategic interaction*. Princeton, NJ: Princeton University Press.
- Cardoos, S. L., Suleiman, A. B., Johnson, M., van den Bos, W., Hinshaw, S. P., & Dahl, R. E. (2017). Social status strategy in early adolescent girls: Testosterone and value-based decision making. *Psychoneuroendocrinology*, 81, 14–21.

- Dunham, Y. (2018). Mere Membership. *Trends in Cognitive Sciences*, 22(9), 780–793.
- Fancis, B., Hasan, I., Park, J. C., & Wu, Q. (2015). Gender differences in financial reporting decision making: Evidence from accounting conservatism. *Contemporary Accounting Research*, 32(3), 1285–1318.
- Fanti, K. A., Henrich, C. C., Brookmeyer, K. A., & Kuperminc, G. P. (2008). Toward a Transactional Model of Parent-Adolescent Relationship Quality and Adolescent Psychological Adjustment. *Journal of Early Adolescence*, 28(2), 252–276.
<https://doi.org/10.1177/0272431607312766>
- Fareri, D. S., Chang, L. J., & Delgado, M. R. (2015). Computational substrates of social value in interpersonal collaboration. *Journal of Neuroscience*, 35(21), 8170–8180.
- Fareri, D. S., Smith, D. V., & Delgado, M. R. (2020). The influence of relationship closeness on default-mode network connectivity during social interactions. *Social Cognitive and Affective Neuroscience*, *in press*, 1–16.
- Feldmanhall, O., & Chang, L. J. (2018). Social Learning: Emotions Aid in Optimizing Goal-Directed Social Behavior. In R. Morris, A. Bornstein, & A. Shenhav (Eds.), *Goal-Directed Decision-Making: Computations and Circuits* (pp. 309–330). London, United Kingdom: Academic Press. <https://doi.org/10.1016/B978-0-12-812098-9.00014-0> ©
- Feng, T., Zhao, W., & Donnay, G. F. (2013). The endowment effect can extend from self to mother: Evidence from an fMRI study. *Behavioural Brain Research*, 248, 74–79.
- Fuligni, A. J. (2018). The need to contribute during adolescence. *Perspectives on Psychological Science*, 1–13. <https://doi.org/10.1177/1745691618805437>

31 Computations and Motivations of Social Decision-Making

Fuligni, A. J., & Pedersen, S. (2002). Family obligation and the transition to young adulthood. *Developmental Psychology, 38*(5), 856–868. <https://doi.org/10.1037/0012-1649.38.5.856>

Gable, S. L., & Impett, E. A. (2012). Approach and avoidance motives and close relationships. *Social and Personality Psychology Compass, 6*(1), 95–108.

Glimcher, P. W. (2004). *Decisions, uncertainty, and the brain: The science of neuroeconomics*. MIT Press.

Guassi Moreira, J. F., Tashjian, S. M., Galván, A., & Silvers, J. A. (2018). Parents versus peers: Assessing the impact of social agents on decision making in young adults. *Psychological Science, 29*(9), 1526–1539. <https://doi.org/10.1177/0956797618778497>

Guassi Moreira, J. F., Tashjian, S. M., Galván, A., & Silvers, J. A. (2019). Is social decision making for close others consistent across domains and within individuals? *Journal of Experimental Psychology: General, (Advance online publication)*, 1–18.

Guassi Moreira, J. F., & Telzer, E. H. (2018). Mother still knows best: Maternal influence uniquely modulates adolescent reward sensitivity during risk taking. *Developmental Science, 21*(1), 1–11. <https://doi.org/10.1111/desc.12484>

Hackel, L. M., Doll, B. B., & Amodio, D. M. (2015). Instrumental learning of traits versus rewards : dissociable neural correlates and effects on choice. *Nature Neuroscience, 18*(9), 1233–1235. <https://doi.org/10.1038/nn.4080>

Henrich, J. (2002). Decision-making, cultural transmission and adaptation in economic anthropology. In J. Ensminger (Ed.), *Theory in Economic Anthropology* (pp. 251–295). Plymouth, UK: AltaMira Press.

Hopmeyer, A., & Medovoy, T. (2017). Emerging adults' self-identified peer crowd affiliations and college adjustment. *Emerging Adulthood*, 5(2), 143–148.

<https://doi.org/10.1007/s11218-017-9390-1>

Johnson, R. W. (2009). Family, public policy, and retirement decisions: Introduction to the special issue. *Research on Aging*, 31(2), 139–152.

Kaniušonytė, G., & Žukauskienė, R. (2018). Relationships with parents, identity styles, and positive youth development during the transition from adolescence to emerging adulthood. *Emerging Adulthood*, 6(1), 42–52. <https://doi.org/10.1177/2167696817690978>

Kirchler, E., Holzl, E., Rodler, C., & Meier, K. (2001). *Conflict and decision-making in close relationship: Love, money, and daily routines*. Psychology Press.

Lavner, J. A., Weiss, B. W., Miller, J. D., & Karney, B. R. (2018). Personality change among newlyweds: Patterns, predictors, and associations with marital satisfaction over time. *Developmental Psychology*, 54(6), 1172.

Lockwood, P. L., Hamonet, M., Zhang, S. H., Ratnavel, A., Salmony, F. U., & Husain, M. (2017). Prosocial apathy for helping others when effort is required. *Nature Human Behaviour*, 1(0131), 1–10. <https://doi.org/10.1038/s41562-017-0131>

Ludwig, R. M., Flournoy, J. C., & Berkman, E. T. (2019). Inequality in personality and temporal discounting across socioeconomic status? Assessing the evidence. *Journal of Research in Personality*, 81, 79–87. <https://doi.org/10.1016/j.jrp.2019.05.003>

Mathews, H. F. (1987). Predicting decision outcomes: Have we put the cart before the horse in anthropological studies of decision making? *Human Organization*, 46(1), 54–61.

33 Computations and Motivations of Social Decision-Making

McKelvey, W., & Kerr, N. H. (1988). Differences in conformity among friends and strangers.

Psychological Reports, 62, 759–762.

Meyer, D. J. C., & Anderson, H. C. (2000). Preadolescents and apparel purchasing: Conformity to parents and peers in the consumer socialization process. *Journal of Social Behavior and Personality*, 15(2), 243–258.

Morelli, S. A., Chang, Y., Carlson, R. W., Kullar, M., & Zaki, J. (2018). Neural detection of socially valued community members. *Proceedings of the National Academy of Sciences*, 115(32), 8149–8154. <https://doi.org/10.1073/pnas.1712811115>

Nisbett, R. E., & Wilson, T. D. (1977). Telling more than we can know: Verbal reports on mental processes. *Psychological Review*, 84(3), 231–259.

Niv, Y., & Chan, S. (2011). On the value of information and other rewards. *Nature Neuroscience*, 14(9), 1095–1097. <https://doi.org/10.1038/nn.2918>

Parkinson, C., Kleinbaum, A. M., & Wheatley, T. (2017). Spontaneous neural encoding of social network position. *Nature Human Behaviour*, 1(5), 1–7. <https://doi.org/10.1038/s41562-017-0072>

Powers, K. E., Yaffe, G., Hartley, C. A., Davidow, J. Y., Kober, H., & Somerville, L. H. (2018). Consequences for peers differentially bias computations about risk across development. *Journal of Experimental Psychology: General*, 147(5), 671. <https://doi.org/10.1037/xge0000389>

Rilling, J. K., & Sanfey, A. G. (2011). The neuroscience of social decision-making. *Annual Review of Psychology*, 62, 23–48. <https://doi.org/10.1080/1068316X.2017.1414817>

- Roisman, G. I., Masten, A. S., Coatsworth, J. D., & Tellegen, A. (2004). Salient and emerging developmental tasks in the transition to adulthood. *Child Development*, 75(1), 123–133.
- Seaman, K. L., Brooks, N., Karrer, T. M., Castrellon, J. J., Perkins, S. F., Dang, L. C., ... Samanez-larkin, G. R. (2018). Subjective value representations during effort, probability and time discounting across adulthood. *Social Cognitive and Affective Neuroscience*, 13(5), 449–459. <https://doi.org/10.1093/scan/nsy021>
- Seaman, K. L., Gorlick, M. A., Vekaria, K. M., Hsu, M., Zald, D. H., & Samanez-larkin, G. R. (2016). Adult age differences in decision making across domains: Increased discounting of social and health-related rewards. *Psychology and Aging*, 31(7), 737–746.
- Sims, C. R. (2018). Efficient coding explains the universal law of generalization in human perception. *Science*, 656, 652–656.
- Sokol-Hessner, P., Hsu, M., Curley, N. G., Delgado, M. R., Camerer, C. F., & Phelps, E. A. (2009). Thinking like a trader selectively reduces individuals' loss aversion. *Proceedings of the National Academy of Sciences*, 106(13), 5035–5040.
<https://doi.org/10.1073/pnas.0806761106>
- Sokol-Hessner, P., Lackovic, S. F., Tobe, R. H., Camerer, C. F., Leventhal, B. L., & Phelps, E. A. (2015). Determinants of propranolol's selective effect on loss aversion. *Psychological Science*, 26(7), 1123–1130. <https://doi.org/10.1177/0956797615582026>
- Sokol-Hessner, P., Raio, C. M., Gottesman, S. P., Lackovic, S. F., & Phelps, E. A. (2016). Acute stress does not affect risky monetary decision-making. *Neurobiology of Stress*, 5, 19–25.
<https://doi.org/10.1016/j.ynstr.2016.10.003>

Tamir, D. I., & Thornton, M. A. (2018). Modeling the predictive social mind. *Trends in Cognitive Sciences*, 22(3), 201–212.

Telzer, E. H., Masten, C. L., Berkman, E. T., Lieberman, M. D., & Fuligni, A. J. (2010). Gaining while giving: an fMRI study of the rewards of family assistance among white and Latino youth. *Social Neuroscience*, 5(5–6), 508–518. <https://doi.org/10.1080/17470911003687913>

Tsai, K. M., Telzer, E. H., Gonzales, N. A., & Fuligni, A. J. (2015). Parental cultural socialization of mexican-american adolescents' family obligation values and behaviors. *Child Development*, 86(4), 1241–1252. <https://doi.org/10.1111/cdev.12358>

Tversky, A., & Kahneman, D. (1992). Advances in prospect theory: Cumulative representation of uncertainty. *Journal of Risk and Uncertainty*, 5, 297–323.

Wang, S.-M., Cheng, Y.-H., Lee, C.-F., & Chuang, S.-C. (2019). Endowment effect: Trading for oneself versus trading for others. *Psychological Reports*, 122(6), 2298–2319.
<https://doi.org/10.1177/0033294118802555>

Zachry, J. E., Johnson, A. R., & Calipari, E. S. (2019). Sex differences in value-based decision making underlie substance use disorders in females. *Alcohol and Alcoholism*, 54(4), 339–341.

Zerubavel, N., Bearman, P. S., Weber, J., & Ochsner, K. N. (2015). Neural mechanisms tracking popularity in real-world social networks. *Proceedings of the National Academy of Sciences*, 112(49), 15072–15077. <https://doi.org/10.1073/pnas.1511477112>

Zhao, W., Feng, T., & Kazinka, R. (2014). The extensibility of the endowment effect to others is mediated by degree of intimacy. *Asian Journal of Social Psychology*, 17, 296–301.

36 Computations and Motivations of Social Decision-Making

<https://doi.org/10.1111/ajsp.12061>

Table 1. Social Decision-making Preferences Across Four Discounting Conditions

Predictor	Probabilistic—Monetary	Probabilistic—Social	Delay—Monetary	Delay—Social
Intercept	-0.240 (.064)***	-0.230 (.068)**	-0.162 (.038)***	-0.297 (.074)***
Condition	0.297 (.074)***	0.157 (.082)°°	0.304 (.075)***	0.128 (.088)°
Reward Ratio	-0.150 (.007)***	-0.138 (.007)***	0.008 (.005)°°	-0.108 (.006)***

Note. *** $p < .001$, ** $p < .01$, * $p < .05$, °° $p < .10$, ° $p < .250$. Coefficients are on a logit scale.

Intercept represents the log likelihood of discounting when condition = 0 and reward ratio is at its grand mean. Condition was coded such that 1 = discounting benefits parent, 0 = discounting benefits friend (positive values indicate a parent-over-friend preference, negative values indicate a friend-over-parent preference). Reward ratio reflects the division of the non-discounting option over the discounting option (grand mean centered), thus negative values indicate individuals were less likely to discount when the discounting reward was relatively smaller than the non-discounting reward. Results reflect robust standard errors from the population-average solution.

Table 2.

A. Social Decision-making Preferences Across Four Discounting Conditions, Moderated by Number of Gamble Decisions During the Gambling Task

Predictor	Probabilistic—Monetary	Probabilistic—Social	Delay—Monetary	Delay—Social
Intercept	0.041 (.297)	-0.121 (.265)	-1.495 (.208)***	-1.160 (.278)***
Sex	-0.055 (.326)	0.001 (.301)	0.596 (.244)*	0.567 (.313)°°
Parent GD	0.006 (.007)	-0.013 (.006)*	-0.004 (.005)	-0.003 (.006)
Friend GD	-0.014 (.006)*	0.001 (.006)	0.003 (.004)	0.002 (.005)
Condition				
Sex	-0.010 (.366)	0.172 (.230)	0.259 (.138)°°	0.307 (.319)
Parent GD	0.223 (.390)	-0.249 (.279)	-0.106 (.212)	-0.523 (.364)°
Friend GD	-0.012 (.007)°	0.008 (.006)°	-0.009 (.008)	0.004 (.007)
Reward Ratio	-0.200 (.036)***	-0.164 (.024)***	-0.105 (.032)**	-0.099 (.023)***
Sex	0.058 (.038)°	0.018 (.027)	-0.006 (.034)	-0.008 (.027)
Parent GD	-0.001 (.001)°	-0.001 (.001)	-0.001 (.001)	-0.001 (.000)*
Friend GD	0.001 (.001)°	0.001 (.001)°	-0.001 (.000)°	0.000 (.000)

B. Social Decision-making Preferences Across Four Discounting Conditions, Moderated by Risk Aversion Parameters from the Gambling Task

Predictor	Probabilistic—Monetary	Probabilistic—Social	Delay—Monetary	Delay—Social
Intercept	0.074 (.310)	-0.049 (.260)	-1.436 (.220)***	-1.140 (.285)***
Sex	-0.102 (.337)	0.088 (.294)	0.523 (.253)*	0.540 (.318)°°
Parent ρ	-0.046 (.066)	-0.027 (.066)	-0.066 (.069)	-0.074 (.041)
Friend ρ	-0.073 (.042)°°	-0.051 (.070)	-0.029 (.062)	0.020 (.082)
Condition				
Sex	0.083 (.379)	0.145 (.259)	0.287 (.139)*	0.361 (.227)
Parent ρ	0.118 (.403)	-0.215 (.293)	-0.145 (.199)	-0.595 (.244)°
Friend ρ	-0.183 (.087)*	-0.007 (.056)	-0.149 (.078)°°	-0.002 (.091)
Reward Ratio	-0.158 (.018)***	-0.152 (.024)***	-0.101 (.033)**	-0.092 (.024)**
Sex	0.009 (.019)	0.004 (.027)	-0.011 (.035)	-0.018 (.028)
Parent ρ	-0.011 (.010)	-0.012 (.005)*	0.008 (.007)	-0.009 (.008)
Friend ρ	-0.002 (.016)	-0.005 (.008)	-0.011 (.006)°°	-0.003 (.006)

Note. *** $p < .001$, ** $p < .01$, * $p < .05$, ° $p < .10$, °° $p < .250$. Coefficients are on a logit scale.

Condition was coded such that 1 = discounting benefits parent, 0 = discounting benefits friend (positive values indicate a parent-over-friend preference, negative values indicate a friend-over-parent preference). Sex was coded such that 0 = Male, 1 = Female. Reward ratio reflects the division of the non-discounting option over the discounting option (grand mean centered), thus

39 Computations and Motivations of Social Decision-Making

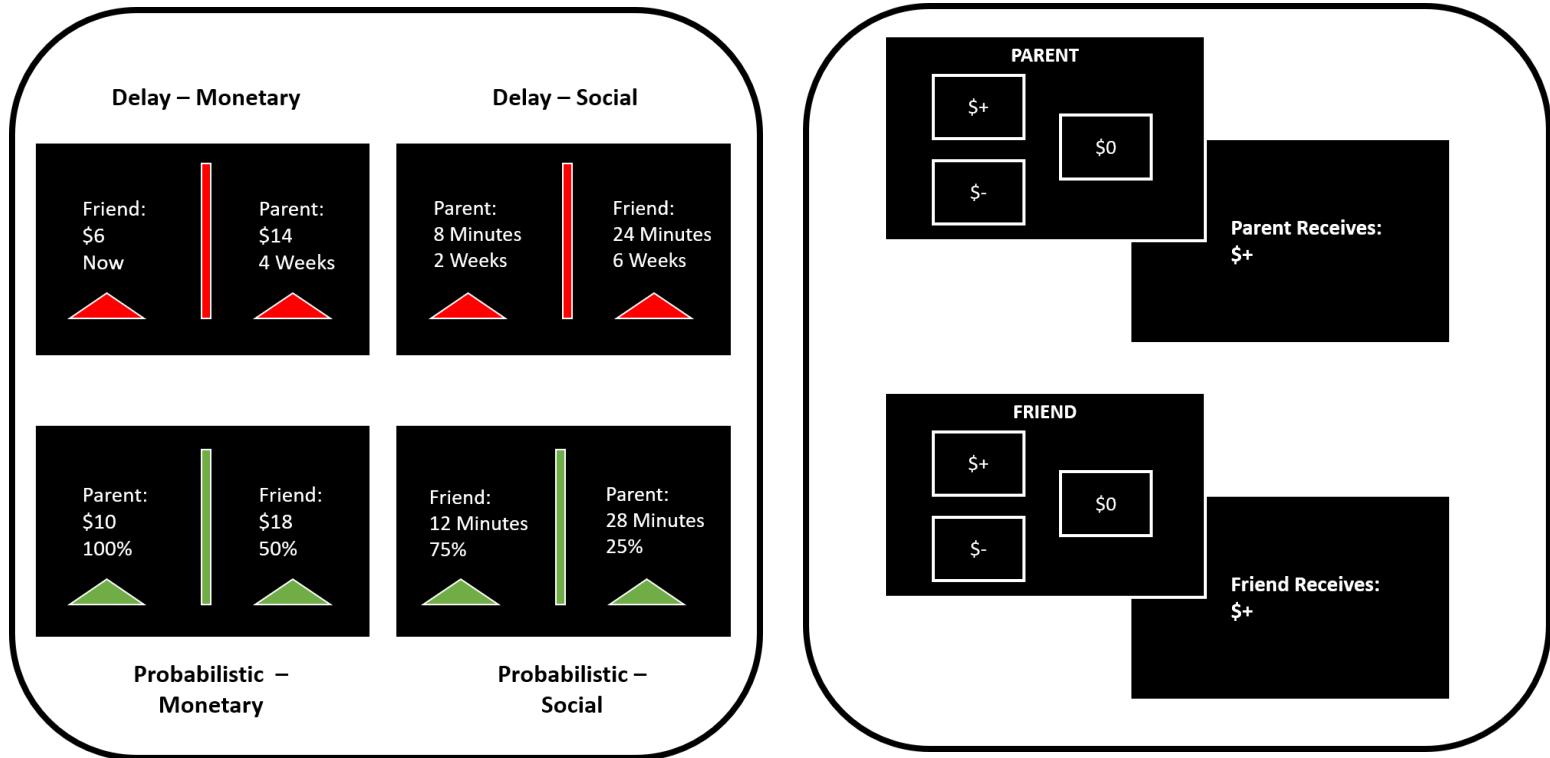
negative values indicate individuals were less likely to discount when the discounting reward was relatively smaller than the non-discounting reward. Results reflect robust standard errors from the population-average solution. GD refers to the number of ‘gamble’ decisions made when completing the task for a parent and for a friend. ρ refers to risk aversion (a potential computational mechanism) when completing the task for a parent and for a friend.

Table 3. Social Decision-making Preferences Across Four Discounting Conditions, Moderated by Relationship Quality

Predictor	Probabilistic—Monetary	Probabilistic—Social	Delay—Monetary	Delay—Social
Intercept	-0.218 (.064)°	-0.300 (.182)°	-0.130 (.104)°	-0.520 (.187)**
Sex	-0.022 (.182)	0.083 (.196)	-0.040 (.112)	0.255 (.203)°
Parent RQ	-0.025 (.121)	-0.291 (.132)*	-0.040 (.056)	-0.431 (.130)**
Friend RQ	0.017 (.176)	0.081 (.172)	0.175 (.107)°	-0.071 (.186)
Condition				
Sex	0.068 (.239)	0.222 (.218)	0.317 (.191)°°	0.141 (.227)
Parent RQ	0.274 (.252)	-0.080 (.232)	-0.009 (.208)	-0.011 (.244)
Friend RQ	0.143 (.112)°	0.815 (.168)***	0.065 (.108)	0.825 (.180)***
Reward Ratio	-0.158 (.018)***	-0.143 (.014)***	0.005 (.004)	-0.099 (.013)***
Sex	0.009 (.019)	0.006 (.016)	0.005 (.005)	-0.010 (.015)
Parent RQ	-0.011 (.010)	-0.040 (.010)***	0.005 (.004)°	-0.031 (.009)**
Friend RQ	-0.002 (.016)	-0.006 (.017)	-0.002 (.005)	0.019 (.016)°

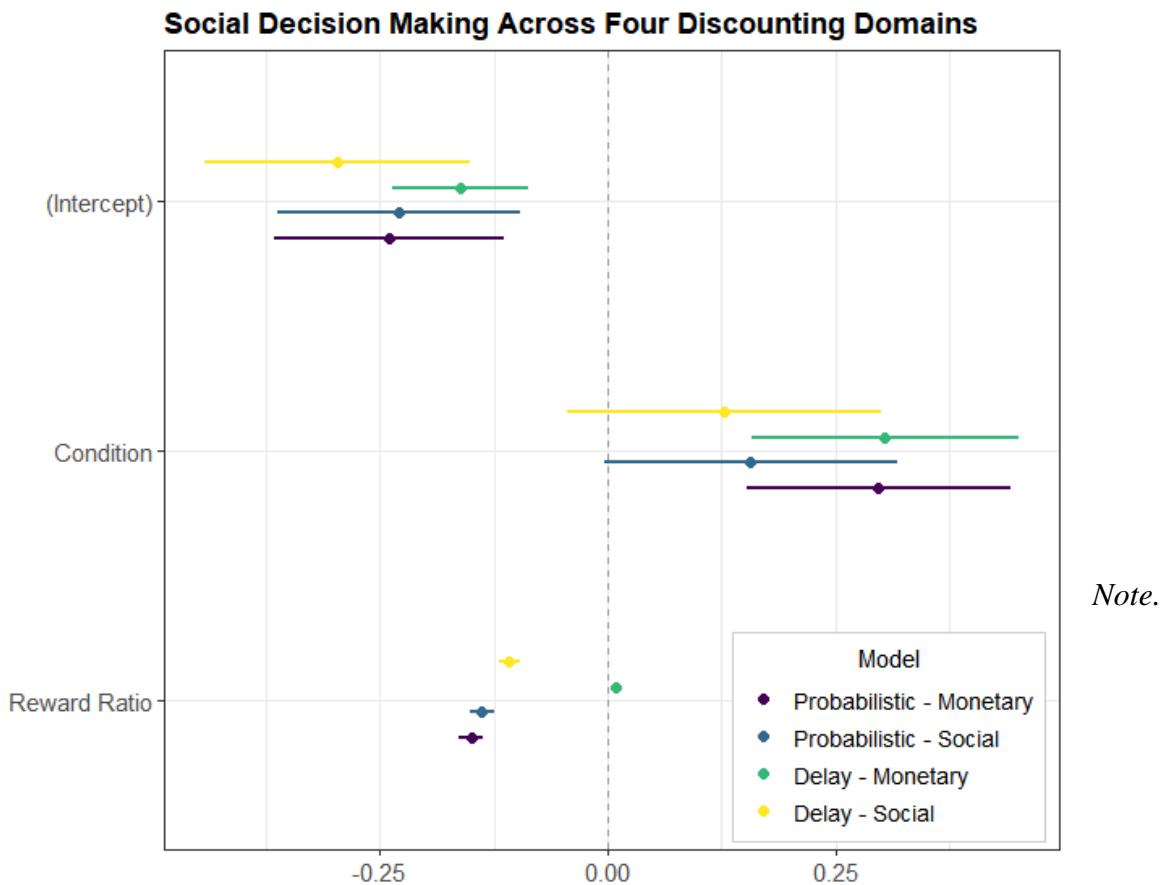
Note. *** $p < .001$, ** $p < .01$, * $p < .05$, °° $p < .10$, ° $p < .250$. Coefficients are on a logit scale.

Condition was coded such that 1 = discounting benefits parent, 0 = discounting benefits friend (positive values indicate a parent-over-friend preference, negative values indicate a friend-over-parent preference). Sex was coded such 0 = Male, 1 = Female. Reward ratio reflects the division of the non-discounting option over the discounting option (grand mean centered), thus negative values indicate individuals were less likely to discount when the discounting reward was relatively smaller than the non-discounting reward. Results reflect robust standard errors from the population-average solution. RQ refers to relationship quality for parent and friend, respectively.

Figure 1. Schematic overviews of the discounting (left) and gambling tasks (right).

Note. All rewards were hypothetical. Outcomes for the gambling task depended on participant choices (e.g., gamble or safe). Reward values in figure above are blank, but varied across trials. The risky option on some trials contained a gain and zero, whereas others contained a gain and loss (depicted above). See the methods for more details. The discounting tasks sought to model social decision-making preferences between parents and friends; the gambling tasks involved completing separate runs for parent and friend, respectively, and helped model computational processes that might support social decision-making preferences.

Figure 2. Visualizing Social Decision-making Preferences (positive Condition values indicate preference toward parents)



Coefficients are on a logit scale. Intercept represents the log likelihood of discounting when condition = 0 (Friend benefits from discounting option, parent benefits from non-discounting option) and reward ratio (defined below) is at its grand mean. Condition was coded such that 1 = discounting benefits parent, 0 = discounting benefits friend. Reward ratio reflects the division of the non-discounting option over the discounting option (grand mean centered), such that larger values indicate greater reward magnitude for the non-discounting option over the discounted option. 95% confidence intervals are depicted, computed using robust standard errors from the population-average solution.

Supplementary Materials

Other Measures

The data in this report come from a broader data collection effort aimed at understanding social decision making. The following measures were collected as part of this effort. Notably, this effort encompassed several pre-registered hypotheses, a subset of which have already been published in a prior report.

Computerized Tasks

- *Columbia Card Task (results published in [BLINDED])
- *Social Gambling Task (results published in [BLINDED])
- *Lexical Trait Judgment Task (results published in [BLINDED])
- *Vanishing Bandit Task (unpublished)

Self-Report Surveys

- *Domain Specific Risk Taking Scale (unpublished)
- *Brief Sensation Seeking Scale (unpublished)
- *Emotion Regulation Questionnaire (unpublished)
- *Emotion Regulation Questionnaire, Capacity (unpublished)
- *Substance use scale (unpublished)

Supplementary Table 1.

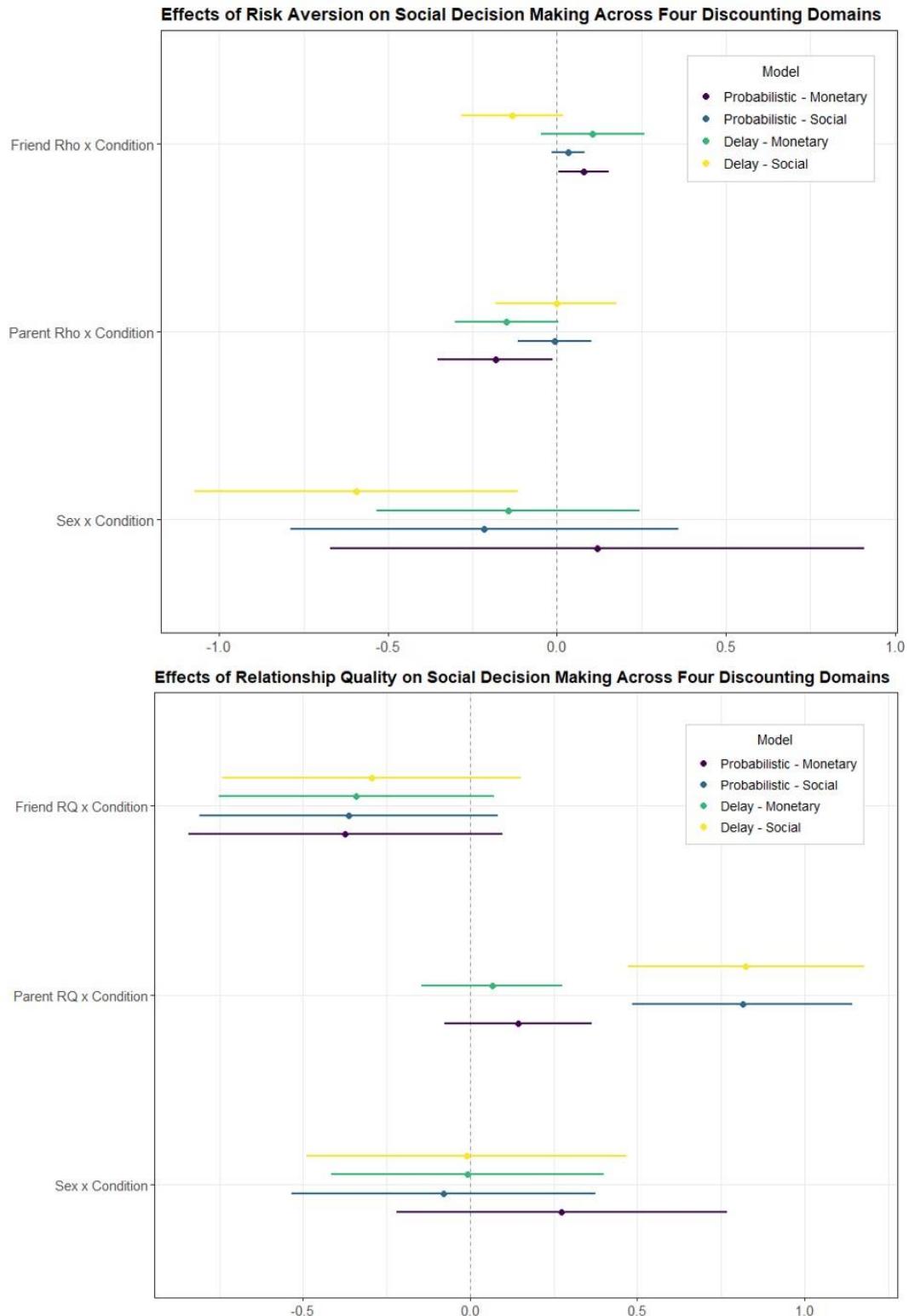
A. Social Decision Making Preferences Across Four Discounting Conditions, Moderated by Number of Loss Aversion During the Gambling Task

Predictor	Probabilistic—Monetary	Probabilistic—Social	Delay—Monetary	Delay—Social
Intercept	-0.008 (.313)	-0.087 (.295)	-1.509 (.223)***	-1.152 (.302)***
Sex	-0.003 (.342)	-0.040 (.323)	0.612 (.259)*	0.559 (.329)°°
Parent λ	-0.021 (.061)	0.050 (.064)	-0.050 (.055)	0.001 (.076)
Friend λ	0.004 (.070)	0.052 (.064)	0.029 (.083)	0.053 (.077)
Condition				
Sex	0.025 (.360)	0.146 (.245)	0.362 (.138)**	0.292 (.330)
Parent λ	0.187 (.385)	-0.219 (.284)	-0.219 (.205)	-0.512 (.365)°
Friend λ	-0.026 (.033)	-0.080 (.092)	0.086 (.056)°	-0.034 (.091)
	0.034 (.055)	-0.069 (.092)	-0.053 (.111)	-0.136 (.088)°
Reward Ratio	-0.194 (.034)***	-0.159 (.022)***	-0.104 (.031)**	-0.093 (.022)***
Sex	0.051 (.036)°	0.012 (.026)	-0.008 (.033)	-0.014 (.025)
Parent λ	0.005 (.006)	-0.002 (.006)	0.007 (.005)°	0.007 (.007)
Friend λ	-0.004 (.006)	-0.002 (.006)	0.005 (.005)	0.001 (.006)

Note. *** $p < .001$, ** $p < .01$, * $p < .05$, ° $p < .10$, °° $p < .250$. Coefficients are on a logit scale.

Condition was coded such that 1 = discounting benefits parent, 0 = discounting benefits friend (positive values indicate a parent-over-friend preference, negative values indicate a friend-over-parent preference). Sex was coded such that 0 = Male, 1 = Female. Reward ratio reflects the division of the non-discounting option over the discounting option (grand mean centered), thus negative values indicate individuals were less likely to discount when the discounting reward was relatively smaller than the non-discounting reward. Values in parentheses reflect robust standard errors from the population-average solution. λ refers to loss aversion (a potential computational mechanism) when completing the task for a parent and for a friend.

Supplementary Figure 1. Visualizing Computational and Motivational Effects on Social Decision Making Preferences



Note. Coefficients are on a logit scale. Condition was coded such that 1 = discounting benefits parent, 0 = discounting benefits friend. 95% confidence intervals are depicted, computed using robust standard errors from the population-average solution. Sex was included as a covariate (coded 1 = female, 0 = male); Condition was coded such that 1 = discounting benefits parent, 0 = discounting benefits friend. Top panel: ‘Rho’ refers to the risk aversion parameter (ρ) and captures attitudes about risk towards parent and friend. Bottom panel: RQ refers to relationship quality and describe motivational influences tied to parents and friends.



Click here to access/download
Supplementary Material

G-MTGS_PvP_CC_JESP_111120_SM_submitted.docx