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Trait Reward Sensitivity Modulates Connectivity with the Temporoparietal Junction and Anterior Insula during Strategic Decision Making

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Conflict of interest statement

The authors declare no conflicts of interest.

27 **Data and code availability**

28 Analysis code related to this project can be found on GitHub: (<https://github.com/DVS-Lab/istart->
29 [ugd](https://github.com/DVS-Lab/istart-)). Thresholded and unthresholded statistical maps are located on
30 <https://neurovault.org/collections/15045/>. In addition, all raw data is made available on
31 OpenNeuro (<https://openneuro.org/datasets/ds004920/versions/1.1.1>).

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40

41 **Abstract**

42

43 Many decisions happen in social contexts such as negotiations, yet little is understood about
44 how people balance fairness versus selfishness. Past investigations found that activation in
45 brain areas involved in executive function and reward processing was associated with people
46 offering less with no threat of rejection from their partner, compared to offering more when there
47 was a threat of rejection. However, it remains unclear how trait reward sensitivity may modulate
48 activation and connectivity patterns in these situations. To address this gap, we used task-
49 based fMRI to examine the relation between reward sensitivity and the neural correlates of
50 bargaining choices. Participants (N = 54) completed the Sensitivity to Punishment
51 (SP)/Sensitivity to Reward (SR) Questionnaire and the Behavioral Inhibition System/Behavioral
52 Activation System scales. Participants performed the Ultimatum and Dictator Games as
53 proposers and exhibited strategic decisions by being fair when there was a threat of rejection,
54 but being selfish when there was not a threat of rejection. We found that strategic decisions
55 evoked activation in the Inferior Frontal Gyrus (IFG) and the Anterior Insula (AI). Next, we found
56 elevated IFG connectivity with the Temporoparietal junction (TPJ) during strategic decisions.
57 Finally, we explored whether trait reward sensitivity modulated brain responses while making
58 strategic decisions. We found that people who scored lower in reward sensitivity made less
59 strategic choices when they exhibited higher AI-Angular Gyrus connectivity. Taken together, our
60 results demonstrate how trait reward sensitivity modulates neural responses to strategic
61 decisions, potentially underscoring the importance of this factor within social and decision
62 neuroscience.

63

64 Key Words: Reward Sensitivity, Strategic Behavior, Ultimatum Game, Dictator Game,
65 Connectivity

66 **Introduction**

67 Social situations such as negotiations often require people to strategically consider social norms
68 while minimizing the threat of being rejected. It is understood that people act fairly when they
69 could be rejected in the Ultimatum Game (UG; Güth et al., 1982; Wells & Rand, 2013) and
70 selfishly when there is not a threat of rejection in the Dictator Game (DG; Engel, 2011;
71 Kahneman et al., 1986). Thus, people exhibit strategic behavior by making smaller contributions
72 in the DG than in the UG (Charness & Gneezy, 2008). Past investigations suggested there are
73 relations between strategic behavior and measures of social functioning such as emotional
74 intelligence (Kench et al., 2007) and Machiavellianism (Spitzer et al., 2007). A possible
75 explanation for strategic behavior is the social heuristics hypothesis, which suggests people
76 share more or less intuitively based on self-interest, and greater deliberation yields more
77 strategic choices (Rand, 2016; Rand et al., 2016).

78
79 Strategic decisions as defined by making larger contributions in UG compared to DG have also
80 been associated with brain activation in the ventral striatum (VS), dorsal lateral prefrontal cortex
81 (dlPFC), and lateral orbitofrontal cortex (OFC) (Spitzer et al., 2007). Other work has implicated
82 dorsal anterior cingulate cortex (dACC) and the posterior cingulate cortex (PCC) in strategic
83 decision making (Weiland et al., 2012). Decisions made in social contexts reliably elicit
84 activation in the right temporoparietal junction (rTPJ) (Behrens et al., 2008; Carter et al., 2012;
85 Dennison et al., 2022), and higher rTPJ activation is associated with greater contributions in the
86 DG (Gianotti et al., 2018; Morishima et al., 2012). Further, stimulation of the right dlPFC is
87 associated with proposing greater contributions in UG and less in DG (Knoch et al., 2006; Ruff
88 et al., 2013; Strang et al., 2015). Finally, people make lower contributions in the DG after
89 stimulation the right dlPFC (Zinchenko et al., 2021). In sum, there is evidence that brain
90 activation can distinguish between some strategic decision making in social contexts.

91
92 Relatively less is known, however, about how strategic decisions in bargaining situations are
93 modulated by task-dependent changes in connectivity across neural circuits supporting reward
94 related decision-making and social cognition (Friston et al., 1997). Past research suggests that
95 signals related to the receipt of rewards are encoded through corticostriatal connectivity (D. V.
96 Smith, Rigney, et al., 2016). Moreover, VS-TPJ connectivity (Park et al., 2017) and dorsal
97 striatum-lateral PFC connectivity (Crockett et al., 2017) were modulated by contributions
98 proposed in DG. Since past findings suggest that anticipating the intentions of another person in
99 an investment game (Zhu et al., 2012) and greater contributions in UG versus DG (Spitzer et al.,
100 2007) were associated with elevated VS responses, it is possible that corticostriatal connectivity
101 may be modulated by social contexts in bargaining situations.

102
103 Additionally, individual differences in trait reward sensitivity may affect how people make social
104 valuations, possibly moderating neural connectivity in social contexts. Reward sensitivity has
105 been studied in clinical contexts (Alloy et al., 2016; Carver & White, 1994; Nusslock & Alloy,
106 2017), revealing that people who are hyper and hyposensitive to rewards are at risk for
107 substance use and bipolar or depressive disorders (Bart et al., 2021). However, little is known
108 about how corticostriatal connectivity is modulated by reward sensitivity (Sazhin et al., 2020).
109 For instance, people who are more sensitive to rewards may overvalue their initial endowment
110 in UG and DG contexts and may be loath to share it with a stranger.

111
112 Since reward sensitivity is associated with risky behavior (Scott-Parker & Weston, 2017), higher
113 Machiavellianism (Birkás et al., 2015), and with more strategic behavior (Scheres & Sanfey,
114 2006), it is plausible that mechanisms underlying strategic decision making may be modulated
115 by reward sensitivity through VS activation or elevated task-based connectivity with the VS.
116 Evidence supporting this interpretation would suggest that strategic decisions may be

117 mechanistically driven by reward processing and that reward sensitivity is a reflection of bottom-
118 up reward responses. Alternatively, strategic decisions may evoke cognitive processes involved
119 in attention and social decision making from brain regions such as the TPJ. Evidence supporting
120 this interpretation would suggest that strategic decisions are driven by top-down cognitive
121 processes and may be modulated by trait reward sensitivity. Overall, examining the role of
122 reward sensitivity and brain responses during strategic decisions could unpack reward,
123 attentional, or value-based decision-making mechanisms that facilitate overcoming social
124 heuristics to act on self-interest.

125

126 Since the VS is sensitive to social valuation (Chen et al., 2011; Fareri & Delgado, 2014), it is
127 plausible that trait reward sensitivity may modulate VS response to social contexts. Testing
128 these relations could help unravel how aberrant reward processing promotes maladaptive
129 decisions that contribute to substance use (Dalley & Robbins, 2017), or possibly diminishes
130 strategic behavior in social situations. Thus, our aims in this investigation were to assess how
131 brain activity and connectivity are modulated by one's strategic decisions, and the extent to
132 which these relations vary by trait reward sensitivity. Using functional magnetic resonance
133 imaging (fMRI), we administered Ultimatum and Dictator Games to participants to investigate
134 associations between strategic behavior, reward sensitivity, and brain connectivity. The study
135 examined activation patterns during both endowment and decision phases, corticostriatal
136 connectivity during the decision phase, and how these patterns were modulated by strategic
137 behavior and reward sensitivity.

138

139 To examine these questions, we assessed several pre-registered hypotheses
140 (<https://aspredicted.org/55gd8.pdf>). Participants proposed offers in DG and UG (eg: DG-P and
141 UG-P) and received offers as a recipient (UG-R). We expected greater activation of the VS and
142 vmPFC during the endowment of money, and that reward sensitivity would potentiate activation

143 in the VS and vmPFC. Such activation during endowment would suggest that reward receipt is
144 modulated by reward sensitivity. Next, we investigated activation within the dIPFC, ACC, SPL,
145 IPS, vmPFC, VS, and TPJ during each task condition and specifically in response to strategic
146 decisions (UG-P > DG-P). We hypothesized that the dIPFC would exhibit stronger activation in
147 response to strategic decisions. These findings during the decision phase would suggest that
148 changes in social context, with respect to norm compliance, evoke differential activation in the
149 brain. Finally, we expected to find elevated ventral striatal responses to strategic behavior (UG-
150 P > DG-P) during the decision phase to be associated with enhanced task-dependent changes
151 in connectivity in regions modulated by social information (e.g., vmPFC, mPFC, and TPJ). In
152 addition, we hypothesized that these neural effects would be enhanced in individuals with higher
153 level of self-reported reward sensitivity. Such findings would suggest that reward sensitivity is an
154 important dimension of understanding brain responses associated with strategic behavior.

155

156 Our analyses focus on two key questions. First, how do strategic decisions in social situations
157 modulate brain activation and connectivity? Second, how does trait reward sensitivity modulate
158 brain connectivity while making strategic decisions? Assessing neural connectivity during
159 strategic decision making and how reward sensitivity modulates these processes would 1)
160 improve our understanding of the mechanisms of how people cooperate and defect in social
161 situations, and 2) help determine how aberrant patterns of reward sensitivity may be a risk
162 factor for maladaptive social decision making.

163 **Materials and Methods**

164 *Participants*

165

166 Although in our pre-registration (<https://aspredicted.org/55gd8.pdf>) we specified that imaging
167 data would be collected from 100 participants (ages 18-22) (Sazhin et al., 2020), we ultimately
168 recruited 59 participants (D. V. Smith et al., 2024) due to constraints imposed by the COVID-19
169 pandemic. Five participants were excluded from our neuroimaging analyses based on our pre-
170 registered criteria and missing data. Specifically, three participants were excluded due to failure
171 to respond during behavioral tasks, where there were greater than 20% missing responses on a
172 given run. One participant was excluded due to incomplete behavioral data. One participant was
173 excluded due to issues with data collection. Three of the 54 participants had one of the two task
174 runs excluded due to excessive head motion. Our final neuroimaging sample resulted in 54
175 participants (mean age: 20.95 years, SD: 1.78 years; 24.1% male). Our final sample size ($N =$
176 54) would enable us to detect medium effects strategic behavior or reward sensitivity ($f^2 =$
177 0.15) or medium to large interaction effects ($f^2 = 0.19$) with 80% power and an alpha of 5%.

178
179 Several behavioral analyses related to social functioning had a more limited sample due to
180 missing data. Specifically, 9 participants were missing behavioral data related to social
181 functioning, resulting in a sample of 45 participants (mean age: 20.74 years, SD: 1.54 years;
182 24.4% male) for several behavioral analyses. All participants were compensated at a rate of \$25
183 per hour inside the scanner and \$15 per hour outside the scanner, and received bonuses based
184 on their decisions, resulting in a total payment ranging from \$140 to \$155. Participants were
185 recruited using Facebook advertisements and fliers posted around the Temple University
186 campuses. We verified that participants were eligible to be scanned using fMRI by the following
187 criteria: a) not being pregnant, b) free of major psychiatric or neurologic illness, and c) not under
188 the influence of substances as evidenced by a breathalyzer test and urine drug screen. All the
189 participants provided written informed consent as approved by the Institutional Review Board of
190 Temple University (protocol number: 24452). Data was acquired using a 3T Siemens PRISMA
191 MRI scanner at Temple University using the Ultimatum and Dictator Games.

192 ***Procedure***

193 Potential participants were identified based on their responses to an online screener
194 questionnaire using the SONA research platform that assessed reward sensitivity using the
195 Behavioral Activation Subscale (BAS; Carver & White, 1994) and the Sensitivity to Reward
196 subscale (SR; Torrubia et al., 2001). Using methods consistent with our prior work (e.g., Alloy,
197 Bender, et al., 2009), we compared results between both SR and BAS to ensure that
198 participants were responding consistently and truthfully by excluding participants with scores
199 that were less than +/-1 quintile on both subscales. Participants also were called on the phone
200 and asked to abstain from alcohol or drug usage for 24 hours prior to the scan. Participants
201 were excluded if they reported that they took any psychoactive medications. Participants
202 attended two appointments, consisting of a battery of psychometric surveys, and a mock scan,
203 followed by a second appointment consisting of the fMRI scan and behavioral tasks.

204 ***Individual Difference Measures***

205

206 **Reward Sensitivity.** To measure reward sensitivity, we used the Behavioral Activation Scale
207 (BAS; Carver & White, 1994) and the Sensitivity to Punishment/Sensitivity to Reward
208 Questionnaire Reward subscale (SPSRWD; Torrubia et al., 2001)). The BAS is a 20-item self-
209 report questionnaire that measures sensitivity to appetitive motives. The SPSRWD is a 24-item
210 self-report measure that assesses how people feel in response to rewarding stimuli.

211

212 **Substance Use.** Given the relation between reward sensitivity and substance use (Bart et al.,
213 2021), it was important to control for alcohol and drug use disorders in all analyses that include
214 reward sensitivity. To measure substance use, we used the Alcohol Use Disorders Identification
215 Test (AUDIT; Babor et al., 1992) and the Drug Use Identification Test (DUDIT; A. Berman et al.,

216 2003; A. H. Berman et al., 2005). The AUDIT is a 10-item self-report measure that assesses
217 frequency of usage over the past year and the self-reported extent to which alcohol use affects
218 the person's life. The DUDIT scale is an 11-item self-report measure counterpart of the AUDIT
219 that assesses frequency and disruptiveness of non-alcohol related substance use. DUDIT
220 contains references to a wide array of substances, including marijuana, cocaine, and others.

221
222 **Social Functioning.** To measure social functioning, we measured trait emotional intelligence
223 and attitudes toward rejection. The trait Emotional Intelligence (EI) questionnaire (TEIQe) is a
224 30-item self-report measure that assesses individual differences in trait empathy, emotion
225 regulation and perspective taking in emotional contexts (Petrides, 2009). Attitudes toward
226 reciprocity were investigated through the 9-item punishment sub-scale of the Personal Norms of
227 Reciprocity (PNR) measure (Perugini et al., 2003).

228 *Experimental Design*

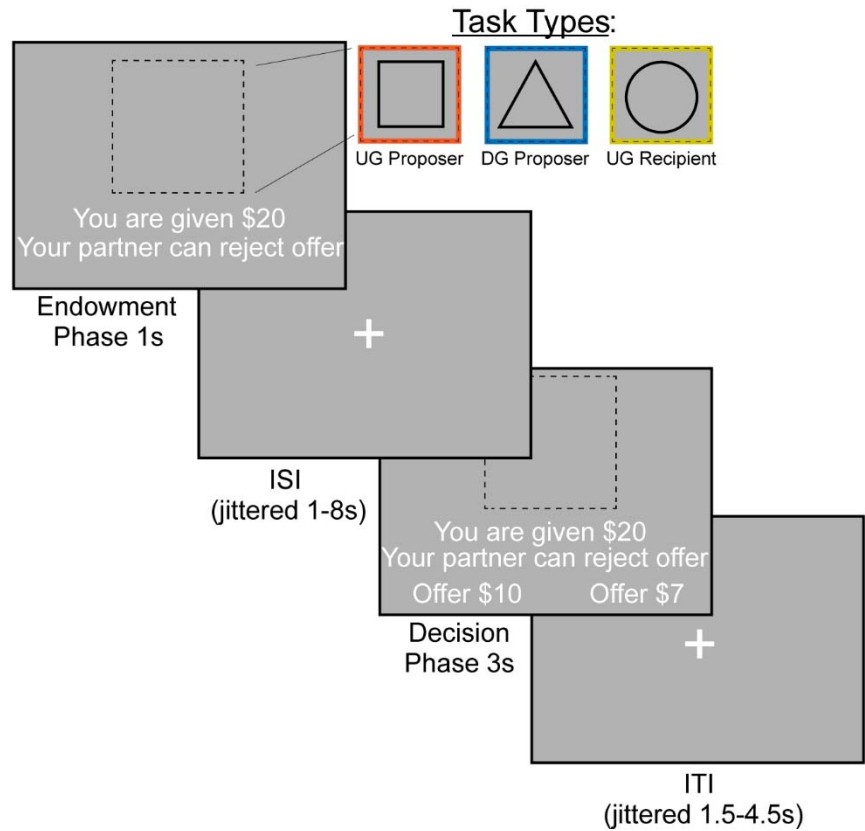
229 We examined bargaining behavior using the Ultimatum (Figure 1) (Güth et al., 1982) and
230 Dictator Games (Figure 1) (Kahneman et al., 1986) (~15 min, counterbalanced across
231 participants). In the Dictator Game (DG), the participant decided how much of an endowed sum
232 (\$15-25) to share with their partner. To ensure that participants were deceived into believing
233 that their decisions had a social impact, the participant was told their partner was represented
234 by decisions made by past participants in the study, and that their decisions would be used with
235 future participants. In addition, each decision was made by a different partner, resulting in each
236 trial being a one-shot game. This design is used to minimize the concern for reciprocity,
237 reputation or other motives beyond social preferences for fairness while making each choice
238 (Yamagishi et al., 2012). In the Ultimatum Game (UG), participants acted as the proposer in
239 some trials and the responder in other trials. As the proposer, participants chose a split of their
240 endowment; however, they were aware that their counterpart could reject their offer. As a

241 recipient in the UG, participants were presented offers from partners that they could choose to
242 accept or reject. If they chose to reject the offer, neither they nor the proposer made any money
243 for that trial. Although our hypotheses and analyses were not focused on the recipient decisions,
244 we included this condition to make the task more believable by making participants think that
245 their unfair proposals could be rejected. We characterize strategic behavior as behavior that
246 offers lower amounts in DG and generally higher amounts in UG, as this strategy would
247 maximize earnings and minimize the threat of rejection.

248

249 The experiment consisted of three conditions (Dictator Game- Proposer (DG-P), Ultimatum
250 Game- Proposer (UG-P), Ultimatum Game- Recipient (UG-R)) that were presented in a
251 counterbalanced order. The tasks were administered using PsychoPy (Peirce et al., 2019)
252 across two 7:30 minute runs. Each run consisted of 36 trials, with 12 trials in each condition. On
253 each trial, the participant was endowed with a sum of money between \$15–\$25 and was
254 presented with the type of trial the participant is playing through a cue. If they were acting as the
255 proposer in the DG, they were presented with a triangle. If they were acting as a proposer in the
256 UG, they were presented with a square. Finally, if they were acting as a recipient in the UG they
257 were presented with a circle. Subsequently, the participant experienced an interstimulus interval
258 (ISI) of 1.5-8 seconds, $M = 2.7s$. During the decision phase as proposer, participants are
259 presented with the option to select a More or Less split. During the decision phase as a
260 recipient, participants have the choice whether to accept or reject the offer. If a participant
261 missed a trial, the screen indicated that they were too slow and recorded a missed trial in the
262 log. Subsequent to each trial, there was a variable duration intertrial interval of 1-4.5 seconds; M
263 = 2.42s.

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269 **Figure 1. FMRI-based Bargaining tasks to Measure Strategic Behavior Using the Dictator and Ultimatum**

270 **Games.** We operationalized strategic behavior as offering more in the Ultimatum Game and less in the Dictator

271 Game, as this strategy would maximize earnings. During the Endowment phase, the participant learned how much

272 money they were given and which task they would complete. A square indicated that the participant would be acting

273 as the Proposer in the Ultimatum Game or deciding how much money to split with a counterpart. A triangle indicated

274 that the participant would act as the Proposer in the Dictator Game. Finally, a circle indicates that the participant

275 would be the Recipient in the Ultimatum Game, which allowed them to decide whether they would accept or reject an

276 offer given to them. We included the Recipient condition so that participants buy into the manipulation of the threat of

277 punishment during the Ultimatum Game as a proposer. During the Decision Phase, the participant as a proposer

278 decided to offer More or Less to their counterpart. As a recipient, whether to accept or reject the offer.

279 ***Behavioral Data Analysis***

280 Strategic behavior was identified for each participant by calculating how much each person
281 chose to share when there was a threat of punishment versus when there was not a threat of
282 punishment. Specifically, for each participant, we calculated the average proportion of the
283 endowment proposed in UG minus DG. Proportions closer to 0 reflected participants who
284 generally proposed a more even split, whereas proportions closer to 0.5 reflected participants
285 who proposed more unfair offers in DG versus UG. We used this method of measuring strategic
286 behavior rather than pooling hypothetical total earnings (see deviations from pre-registration) as
287 it avoids inferring earnings and simply used the participants' decisions.

288
289 To examine whether participants acted strategically through offering more as a Proposer in the
290 Ultimatum Game condition versus the Dictator Game condition, we used a mixed effect linear
291 model. The regressors included the task (UG-P or DG-P), trial endowment, and the proportion
292 of the endowment the participant offered. While we included the recipient condition (UG-R) so
293 that participants experience offers to understand the threat of punishment as proposers, our
294 main questions do not assess recipient behavior. Nonetheless, as a manipulation check to
295 assess whether participants rejected unfair offers more frequently (i.e., offers with a proportion
296 substantially less than half of the endowment) in the Ultimatum Game as a recipient, we
297 regressed participants' choices to accept or reject an offer on partner endowment and the
298 proportion offered. Next, we assessed whether there were associations between decisions and
299 measures of social functioning, reward sensitivity, and substance use. Given that both hyper-
300 and hypo-sensitivity to rewards have been linked to substance use (Alloy et al., 2009; Bart et al.,
301 2021; Franken & Muris, 2006), we control for levels of substance use in our data while
302 assessing reward sensitivity. We used correlations between measures (i.e., social functioning,
303 reward sensitivity, and substance use) with the proportions offered in the UG versus DG (i.e.,

304 Spitzer et al., 2007). This method of measuring strategic behavior was used rather than pooling
305 hypothetical total earnings (see deviations from pre-registration) as this method avoided
306 inferring earnings and simply used the participants' decisions. We also conducted exploratory
307 analyses to 1) assess whether there are associations between strategic behavior and reward
308 sensitivity and substance use, and 2) whether there are associations between the individual
309 difference measures and individual conditions (DG-P, UG-P, and UG-R).

310

311 We conducted analyses on the included self-report measures to ensure that they were correctly
312 operationalized for further analyses. Since the BAS and SR subscale of the SPSRWD were
313 highly correlated $r(52) = .71, p < .001$, we combined them into a single composite measure of
314 reward sensitivity using their combined z-scores. Reward sensitivity scores were binned into
315 deciles to produce an even distribution for subsequent analysis. Finally, because both hyper-
316 and hypo-sensitivity to rewards have been linked to substance use (e.g., Alloy et al., 2009; Bart
317 et al., 2021; Franken & Muris, 2006), we squared the binned composite reward sensitivity
318 scores to create an additional, quadratic measure of aberrant reward sensitivity. In other words,
319 aberrant reward sensitivity explores whether there are consistent patterns across people who
320 are either high or low in reward sensitivity. Next, we found that AUDIT and DUDIT also were
321 correlated $r(52) = .32, p = .02$. As a result, we operationalized problematic substance use
322 through z-scoring the responses between the measures and combining them into a single
323 composite z-score of problematic substance use using the same method as described for
324 reward sensitivity. Behavioral data analyses were completed using MATLAB (MATLAB, 2022),
325 R (R Core Team, 2022), and Python (Van Rossum & Drake, 2009).

326 ***Neuroimaging Data Analyses***

327 Functional images were acquired using a 3T Siemens PRISMA MRI scanner at Temple
328 University. Neuroimaging data were converted to the Brain Imaging Data Structure (BIDS) using

329 HeuDiConv (Halchenko et al., 2024). We applied spatial smoothing with a 5mm full-width at
330 half-maximum (FWHM) Gaussian kernel using FEAT (FMRI Expert Analysis Tool) Version 6.00,
331 part of FSL (FMRIB's Software Library, www.fmrib.ox.ac.uk/fsl). See the *Supplemental*
332 *Information* for the full neuroimaging data acquisition and preprocessing pipeline.

333

334 Neuroimaging analyses used FSL version 6.0.4 (Jenkinson et al., 2012; S. M. Smith et al.,
335 2004). We used two general linear models with local autocorrelation (Woolrich et al., 2001).
336 Both models included a common set of confound regressors consisting of the six motion
337 parameters (rotations and translations), the first six aCompCor components explaining the most
338 variance, non-steady state volumes, and the framewise displacement (FD) across time. Finally,
339 we used high-pass filtering (128s cut-off) through a set of discrete cosine basis functions.

340

341 The first model tested task-based brain activation elicited during the endowment (duration =
342 1,000 ms) and decision (duration = 3,000 ms) phases and the effect of strategic behavior on
343 brain function during these phases. To do this, we included 6 task-specific regressors
344 (endowment: DG-P, UG-P, and UG-R; decision: DG-P, UG-P, and UG-R), and the same 6 task-
345 specific regressors that we parametrically modulated by the proportion of the offer proposed by
346 the participant. In other words, the parametric modulator measured brain responses to the
347 fairness of the offer proposed. A thirteenth regressor modelled missed trials. By including both
348 parametrically modulated and non-modulated task-based regressors, we were able to
349 investigate the parametric effects while properly controlling for changes in activation across UG
350 and DG conditions.

351

352 The second type of model focused on the task-dependent connectivity using the ventral striatum
353 as a seed and areas related to social processing as target regions. To estimate the changes in
354 connectivity resulting from strategic behavior, we used psychophysiological interaction (PPI)

355 analysis (Friston et al., 1997; O'Reilly et al., 2012). Meta-analyses have demonstrated that PPI
356 is able to reliably reveal specific patterns of task-dependent connectivity (D. V. Smith, Gseir, et
357 al., 2016; D. V. Smith, Rigney, et al., 2016; D. V. Smith & Delgado, 2017). Our PPI analysis
358 focused on task-dependent changes in connectivity using the ventral striatum (VS; Oxford-GSK-
359 Imanova atlas) as a seed. Additionally, we used seeds derived from whole-brain analyses (e.g.,
360 Inferior Frontal Gyrus and Anterior Insula) to find non-pre-registered target regions in secondary
361 analyses (O'Reilly et al., 2012). The average time course of activation from this seed region was
362 extracted and used as an additional fourteenth regressor. To construct the PPI model, we used
363 the same model described above and added 14 additional regressors (1 regressor for the VS
364 region and 13 regressors for the interaction between the VS region and the task-based
365 regressors), yielding a total of 25 regressors in each seed-based PPI model. Both activation and
366 connectivity models were then run through a fixed effects second level analysis that combined
367 the first and second runs. For participants with missing data, or for runs that were excluded due
368 to head motion, we used a participant's one good level one run in the group level analyses. For
369 all participants and their combined runs, we used a fixed-effects model.

370

371 Group-level analysis focused on activation and connectivity patterns and their associations
372 between bargaining behavior, substance use and BOLD responses, independent of reward
373 sensitivity. The analyses were carried out using FLAME (FMRIB's Local Analysis of Mixed
374 Effects) Stage 1 (Beckmann et al., 2003; Woolrich et al., 2004). Our group-level model focused
375 on comparisons between the Dictator and Ultimatum Games as a Proposer; these comparisons
376 included covariates to account for reward sensitivity, the second-order polynomial expansion of
377 reward sensitivity (which captures effects tied to aberrant reward sensitivity), substance use,
378 strategic behavior, temporal signal to noise ratio (tSNR) and mean framewise displacement (fd
379 mean). Strategic behavior as a covariate in the group model was identified based on the
380 average proportion offered in UG minus DG for each individual participant. In other words, a

381 participant that was more strategic would have exhibited a larger difference in contributions
382 compared to someone who was less strategic. We also applied two additional models that
383 explored interaction effects. The first interaction model included additional regressors of
384 substance use and reward sensitivity and substance use and aberrant reward sensitivity. The
385 second interaction model included additional regressors of the interaction of strategic behavior
386 and reward sensitivity, and main effects of strategic behavior and aberrant reward sensitivity.
387 We controlled for multiple comparisons through identifying pre-registered regions of interest and
388 by correcting for multiple comparisons across the whole brain using Z-statistic images that were
389 thresholded parametrically (Gaussian Random Field Theory) using clusters determined by
390 $Z > 3.1$ and a (corrected) cluster significance threshold of $P = 0.05$ (Flandin & Friston, 2019;
391 Nichols & Hayasaka, 2003; see *Supplemental Information* for more details).

392 ***Deviations from Pre-Registration***

393 Once data collection and analyses began, we made several adjustments based on four issues
394 that were unspecified in our pre-registration. First, we initially specified that we would use the
395 parametric effect of endowment, but not for decisions. For decisions, we expected to use the
396 actual offers selected (High, Low) in our analyses. However, since many participants selected
397 High more often in the UG condition and Low in the DG condition, these regressors had fewer
398 events for comparison. To address this issue, we modeled strategic decisions as parametric
399 effects of offer amount through the difference in the proportions of the endowments offered
400 between DG-P and UG-P. Second, we adjusted the covariates in our group level models due to
401 missing data. Although we originally planned to study Machiavellianism, due to an error in data
402 collection, this survey was not completed by our participants. Next, whereas substance use
403 analyses were not mentioned in the pre-registration, we intended to complete them in
404 accordance with the broader aims and hypotheses of the grant, which are also described in the
405 grant report (Sazhin et al., 2020). Third, we used the (Clithero & Rangel, 2014) (-2, 28, -18)

406 meta-analysis vmPFC coordinates for our mask rather than the mask specified in the pre-
407 registration (Delgado et al., 2016) for greater spatial specificity in our analyses. Fourth, we
408 explored group level models that included the interaction of reward sensitivity, substance use
409 and strategic behavior despite not being initially pre-registered. Taken together, these
410 adjustments from the pre-registration have allowed us to analyze the data more robustly. Our
411 results and discussion take care to differentiate between confirmatory and exploratory results,
412 especially emphasizing differences in our group level models.

413 **Results**

414 Below, we report results from behavioral analyses, task-based neural activation and connectivity
415 analyses. We begin by presenting results of the behavioral tasks, assessing whether
416 participants made choices as expected, and if their choices relate to self-reported levels of
417 emotional intelligence, attitudes toward rejection, reward sensitivity, and substance use. Next,
418 we examined pre-registered hypotheses examining strategic choices between the dictator and
419 ultimatum games within reward-related and social neural systems (see *Supplemental*
420 *Information*). Although our pre-registered ROI-based analyses did not support our hypotheses
421 (see *Supplemental Information*), these analyses were followed with a whole-brain analyses that
422 examined activation and connectivity in response to strategic decisions, revealing that elevated
423 IFG and AI activation is associated with strategic decisions. Subsequently, we investigated task-
424 dependent connectivity using the IFG and AI as seeds for potential target regions. These
425 analyses found that IFG-pTPJ connectivity is modulated by elevated strategic decisions. Finally,
426 we present exploratory results that investigate associations between attitudes toward fairness,
427 reward sensitivity, and brain connectivity.

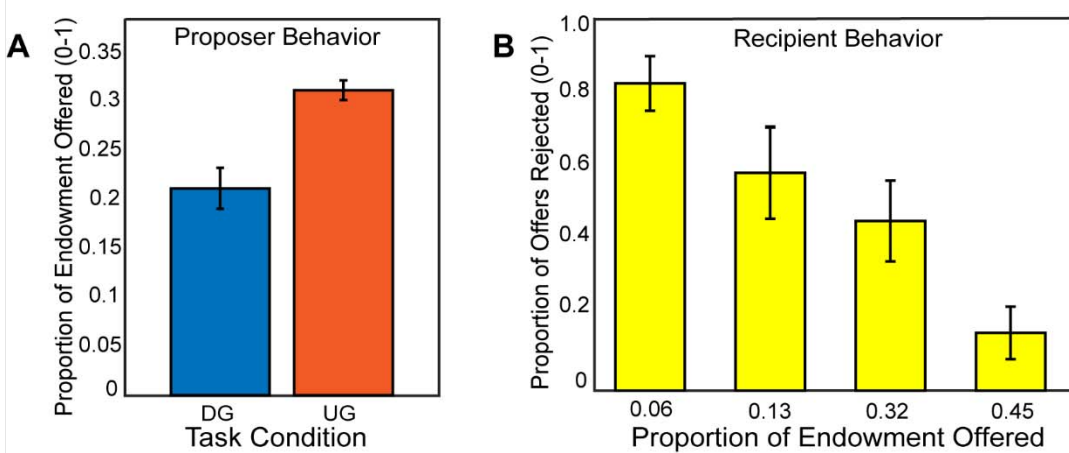
428 ***Strategic Behavior***

429 If participants made higher offers in the Ultimatum Game compared to the Dictator Game, this
430 would indicate that participants were acting most consistently toward maximizing their earnings,
431 thereby exhibiting strategic behavior. Consistent with our expectations, using a mixed effects
432 model for a random intercept, we found that participants (N=54) made more selfish offers in the
433 DG vs. the UG conditions, (B = -0.43, SE = 0.015, $t(2550) = -28.09$, $p < .001$), see *Figure 2*), with
434 the overall model reporting an adjusted R^2 of 0.19. As a manipulation check, we investigated
435 whether participants rejected unfair offers in the recipient condition. A binary logistic regression
436 indicated that participants reject more often with lower offers, (B = 1.72, SE = 0.095, $t(1252) =$
437 18.06 , $p < .001$), with the overall model reporting an adjusted R^2 of 0.50. Next, we explored
438 whether there was a relation between strategic behavior and rejection rate as a function of offer
439 amount as a recipient, finding no significant association, $r(52) = -.19$, $p = .16$. Given that there
440 was no relationship of recipient choices to strategic decisions as proposers, we excluded these
441 measures from subsequent analyses.

442
443 Next, we assessed whether measures of social functioning (N=45) were related to strategic
444 decisions. Several participants had missing questionnaire data, resulting in a smaller dataset for
445 these analyses. Consistent with our hypotheses, individuals scoring higher on the Emotional
446 Intelligence (EI) scale made higher offers as a proposer in the Ultimatum Game, $r(43) = .35$, $p =$
447 $.02$. Contrary to our hypotheses, we did not find associations between strategic behavior,
448 emotional intelligence, or attitudes toward rejection that met a p -value of less than $p = .05$.
449 Inasmuch as there was no effect of strategic behavior and our measures of social functioning as
450 we hypothesized, these measures were excluded from further analyses and used the full
451 dataset of 54 participants for further analyses.

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456 **Figure 2. Participants make strategic decisions by offering lower in the Dictator Game versus the Ultimatum**

457 **Game.** In Panel A, we find that participants made higher offers in the Ultimatum Game as a proposer compared to

458 the Dictator Game. In Panel B, we show that participants rejected unfair offers more frequently when they acted as a

459 recipient in the Ultimatum Game. Overall, these behavior results are consistent with our hypotheses and past

460 literature.

461

462 Although we did not expect relations between strategic behavior and measures of reward

463 sensitivity and substance use, we explored whether there were such associations to

464 contextualize any brain relations we may have found with these respective individual difference

465 measures. We did not find any significant associations between reward sensitivity and

466 substance use, and strategic behavior or individual task conditions (DG-P, UG-P, UG-R) that

467 met a threshold of $p < .05$.

468 **Neural Responses while Making Strategic Decisions**

469

470 To examine how people make strategic decisions in bargaining situations, we investigated how

471 people propose offers in the Ultimatum Game (UG) versus the Dictator Game (DG). First, we

472 assessed whether there were activation differences between UG and DG conditions, failing to
473 find any significant activation that met $p = .05$ or lower. Next, we assessed if brain responses
474 tracked the fairness of the offers proposed differently between DG and UG. In other words, do
475 participants have differing brain activation when proposing higher proportions of the endowment
476 or lower proportions of the endowment when there is a threat of punishment versus when there
477 is not a threat of punishment?

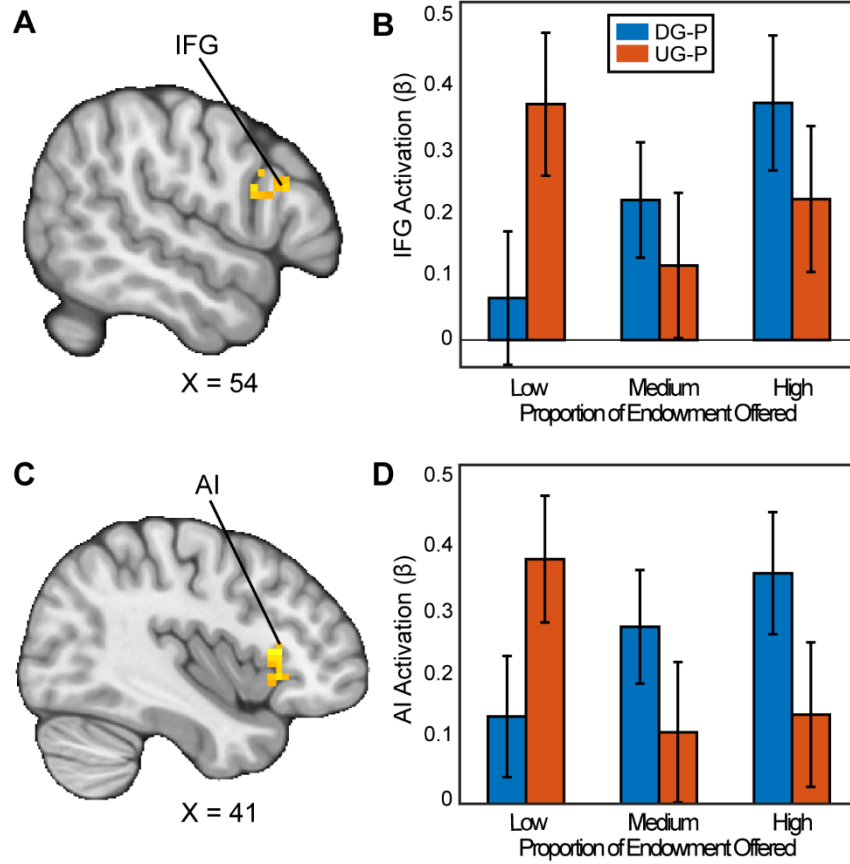
478

479 Our results indicated that when participants chose to be selfish versus fair in the contrast
480 between the DG and UG as a proposer, there were significant clusters in the Inferior Frontal
481 Gyrus (IFG) (MNIxyz = 51, 24, 24; cluster = 20 voxels, $p=.035$) and a cluster spanning the
482 Anterior Insula (AI), extending into the Orbitofrontal Cortex (OFC) (MNIxyz = 33, 27, -4; cluster =
483 54 voxels, $p<.001$). We did not find significant activation in the vIPFC or the VS. In the contrast
484 between UG and DG (i.e., choosing to be fair versus unfair), we found a significant cluster in
485 cerebellum (MNIxyz = 30,-82, -36; cluster = 37, $p<.001$). In sum, some of our results
486 successfully replicated past investigations of strategic behavior.

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Figure 3. Activation associated with strategic thinking. We conducted a whole-brain analysis to investigate whether there were regions in the brain that differentially responded when acting as a proposer in the DG versus UG conditions. When assessing the parametric effect associated with acting more strategically, Panels A and C reflect regions (Inferior Frontal Gyrus (IFG) (MNIxyz = 52, 16, 22; cluster = 20 voxels, $p=0.035$, and Anterior Insula (AI) extending into the Orbitofrontal Cortex (OFC) (MNIxyz = 37, 23, 2; cluster = 54 voxels, $p<.001$ respectively) with significant activation. That is to say, as participants made fairer offers in the DG condition, they experienced stronger activation compared to when they made fairer offers in the UG condition. (Thresholded: <https://neurovault.org/images/803473/>; Unthresholded: <https://neurovault.org/images/803474/>). For illustrative purposes, Panels B and D shows the extracted parameter estimates within each region. We note that Z statistic images were thresholded parametrically (Gaussian Random Field Theory) using clusters determined by $Z>3.1$ and a (corrected) cluster significance threshold of $p=.05$.

506 *Table 1: We incorporated several group level models assessing strategic behavior and reward sensitivity while*
 507 *controlling for substance use. We assessed the interactions of reward sensitivity and strategic behavior and*
 508 *substance use respectively. If there were no interaction effects, we interpreted main effects using the no interaction*
 509 *model. We completed these group level analyses across both activation and PPI models. The PPI model used a pre-*
 510 *registered VS seed, and IFG and AI seeds as derived from our secondary results. The initial group level models were*
 511 *derived from initial hypotheses, though the interaction of reward sensitivity and strategic behavior was an exploratory*
 512 *model driven by our results. Thresholded and unthresholded images are available on Neurovault:*
 513 <https://neurovault.org/collections/15045/>

514

<u>Model Type</u>	<u>Confirmatory/Exploratory</u>	<u>Covariates</u>
No Interactions	Confirmatory	Strategic Behavior, Substance Use, Reward Sensitivity, Aberrant Reward Sensitivity
Reward Sensitivity x Substance use	Confirmatory	No Interaction model plus substance use x reward sensitivity, substance use x aberrant reward sensitivity
Reward Sensitivity x Strategic Behavior	Exploratory	No Interaction model plus strategic behavior x reward sensitivity, strategic behavior x aberrant reward sensitivity

515

516 ***Strategic Behavior and Neural Connectivity***

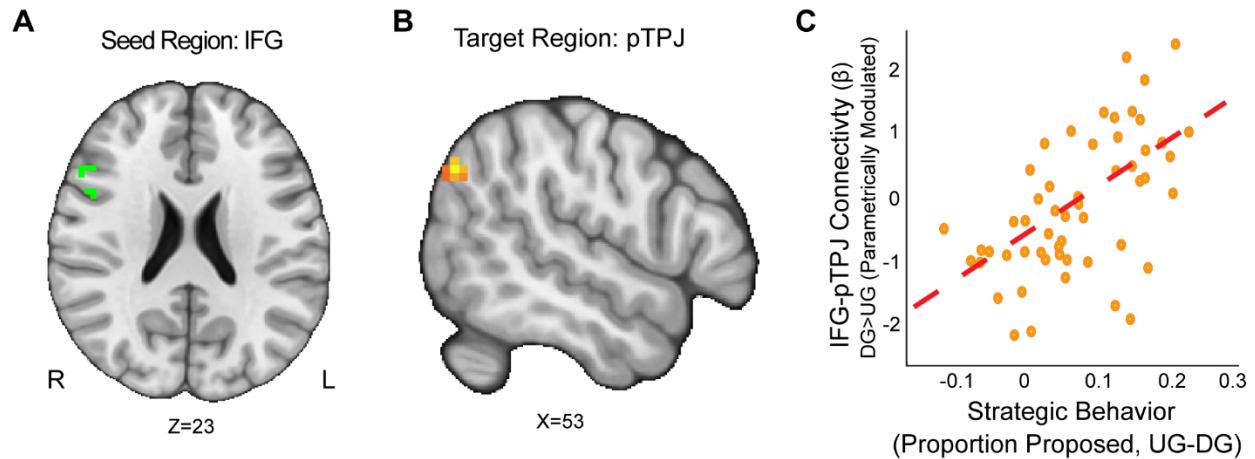
517 Beyond activation patterns, we studied whether task-dependent connectivity patterns related to
 518 reward sensitivity and strategic decisions made in the Dictator and Ultimatum games. We
 519 included the IFG and AI as seeds because they were derived from the activation of DG versus
 520 UG in response to the level of proportion offered. Our group level analyses employed several
 521 covariates, including motion-based nuisance regressors, reward sensitivity, substance use, and

522 strategic behavior. We also explored two additional models that investigated the interactions of
523 reward sensitivity, strategic behavior, and substance use respectively.

524

525 First, we wanted to examine if strategic behavior as measured by the choices our participants
526 made was associated with brain connectivity. Using the IFG as a seed (MNIxyz = 52, 16, 22),
527 we found that enhanced connectivity with a left rpTPJ target region (Schurz et al., 2017)
528 extending into the SMG (MNIxyz = 50, -68, 35; cluster = 22 voxels, $p = .008$) was modulated by
529 strategic behavior in the Dictator versus Ultimatum Game (see Figure 4). That is to say, selfish
530 participants (i.e.: by making lower proposals in the DG versus UG conditions) experienced
531 enhanced IFG-rpTPJ connectivity contingent on whether or not there was a threat of rejection.
532 Our results suggest that enhanced IFG-rpTPJ connectivity may facilitate the social processing
533 associated with strategic decisions in social contexts. We also examined if connectivity from an
534 AI seed related to strategic situations was modulated by strategic behavior. Using the AI seed
535 (MNIxyz = 33, 27, -4), we found that attenuated connectivity with the neighboring insular cortex
536 (MNIxyz = 50, 6, -1; cluster = 26 voxels, $p = .003$) was modulated by strategic behavior in UG
537 versus DG condition. That is to say, participants who were more selfish when there was no
538 threat of rejection exhibited lower AI-Insula connectivity. Our results suggest that attenuated co-
539 activation of the insular cortex may contribute to making more selfish choices in social contexts.

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543 **Figure 4. IFG-rpTPJ Connectivity is Modulated by Strategic Behavior.** We found that connectivity between an

544 inferior frontal gyrus (IFG) seed (Panel A), and a right pTPJ target (Panel B) was related to elevated strategic

545 behavior (Panel C) (DG > UG) (MN_{xyz} = 50, -68, 35; cluster = 22 voxels, $p = .008$). (Thresholded:

546 <https://neurovault.org/images/803475/> Unthresholded <https://neurovault.org/images/803476/>). These results suggest

547 that IFG- right pTPJ connectivity may modulate strategic behavior contingent on whether there is a threat of rejection

548 or not. Participants who experienced elevated IFG-right pTPJ connectivity were also those who were more selfish in

549 DG and offered closer to even splits in UG. For illustrative purposes, we extracted the parameter estimates within

550 each region (Panel C). We note that Z statistic images were thresholded parametrically (Gaussian Random Field

551 Theory) using clusters determined by $Z > 3.1$ and a (corrected) cluster significance threshold of $p = .05$ and the images

552 are plotted using radiological view.

553 **Exploratory Analyses: Anterior Insula-Angular Gyrus Connectivity, Trait**

554 **Reward Sensitivity, and Strategic Behavior**

555 Next, we explored how the interaction of reward sensitivity and substance use may modulate

556 brain connectivity patterns associated with strategic thinking in bargaining situations.

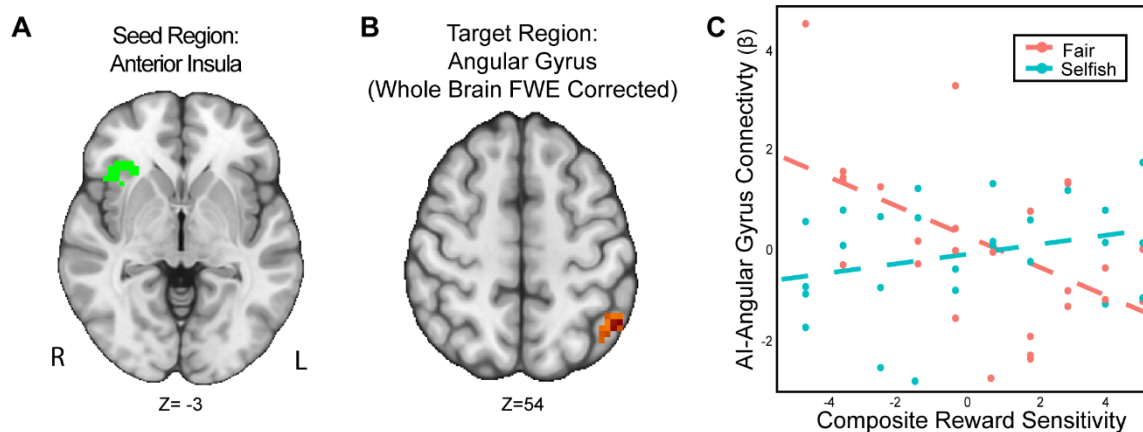
557 Investigating how a trait like reward sensitivity may modulate brain responses can reveal an

558 important factor for understanding both behavior and brain relationships. Specifically, we used a

559 model that included interaction covariates of strategic thinking with reward sensitivity and

560 aberrant reward sensitivity. The model also controlled the main effects of strategic behavior,

561 reward sensitivity, aberrant reward sensitivity, and substance use. We included substance use
562 as a controlling variable due to its known relationships with reward sensitivity in
563 psychopathology (Joyner et al., 2019).
564
565 We found that the interaction of reward sensitivity and strategic behavior modulated AI-Angular
566 Gyrus connectivity in the UG versus DG condition (Figure 5). That is to say, participants with
567 higher reward sensitivity and attenuated AI-Angular Gyrus connectivity tended to make more
568 strategic choices when there was a threat of rejection relative to when there was not. Moreover,
569 participants with lower reward sensitivity *and* enhanced AI-Angular Gyrus connectivity tended to
570 make more strategic choices when there was a threat of rejection compared to when there was
571 not. These exploratory results suggest that the combination of strategic decisions and a
572 person's trait reward sensitivity together may modulate connectivity patterns in social situations
573 requiring strategic thinking.
574



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576
577 **Figure 5. The interaction of reward sensitivity and strategic behavior modulated AI – Angular Gyrus**
578 **connectivity in social situations requiring strategic thinking.** We conducted a whole-brain analysis exploring the
579 interaction of trait reward sensitivity and strategic behavior. We found that higher reward sensitivity is associated with
580 1) more strategic behavior and 2) elevated task-dependent changes in connectivity between AI (Panel A) and the
581 Angular Gyrus (MNI; xyz = -47, -56, 54; cluster = 23 voxels, $p = .005$). Conversely, for participants with low reward

582 *sensitivity, we found that their AI-Angular connectivity is lower as they exhibit strategic behavior. For illustrative*
583 *purposes (Panel C), we used a median split to indicate the relation of reward sensitivity and strategic behavior. Next,*
584 *we extracted the parameter estimates within each region (Panel C). We note that Z statistic images were thresholded*
585 *parametrically (Gaussian Random Field Theory) using clusters determined by $Z > 3.1$ and a (corrected) cluster*
586 *significance threshold of $p=.05$ and the images are plotted using radiological view. See images here:(Thresholded:*
587 *<https://neurovault.org/images/803477/>; Unthresholded: <https://neurovault.org/images/803482/>).*

588 **Discussion**

589 This study investigated how relations between strategic behavior in bargaining situations and
590 reward responses correspond to patterns of brain activation and connectivity. First, **the**
591 behavioral results are consistent with past work suggesting that participants act strategically in
592 bargaining situations through acting fairly when there is a threat of rejection (e.g., Ultimatum
593 Game; UG) while keeping more for themselves when there is not a threat of rejection (Dictator
594 Game; DG) (Charness & Gneezy, 2008). Second, the neuroimaging analyses revealed that
595 strategic behavior in the Dictator versus Ultimatum Games evoked activation in the Inferior
596 Frontal Gyrus (IFG) and Anterior Insula (AI), results that were consistent with past investigations
597 (i.e., Spitzer et al., 2007). Our analyses also indicated that elevated IFG-rTPJ connectivity was
598 related to enhanced strategic behavior and attenuated AI-Angular Gyrus connectivity was
599 modulated by the interaction of reward sensitivity and strategic behavior. Taken together,
600 whether people choose to be fair or selfish in bargaining situations is associated with pattern of
601 neural connectivity, which in turn may depend on a person's trait reward sensitivity.

602

603 This work fits in with past literature suggesting that norm compliance is regulated by cortical
604 activation. Although we did not find activation during UG versus DG in the pre-registered
605 regions of interest, whole brain analyses revealed activation in the right IFG and AI as
606 participants made strategic decisions, replicating previous work (Spitzer et al., 2007; Zheng &
607 Zhu, 2013). Next, both IFG and AI activation has been observed in other decision-making

608 contexts. For example, FeldmanHall and colleagues reported AI activation during moral decision
609 making (FeldmanHall et al., 2014). In addition, other work has shown that increased activation
610 in the anterior insula in a trust task is associated with inequity aversion (van Baar et al., 2019;
611 FeldmanHall et al., 2014). Further, our results are consistent with stimulation-based research
612 that found elevated right dlPFC area activation corresponded to more strategic behavior (Knoch
613 et al., 2006; Ruff et al., 2013; Strang et al., 2015) and inhibition of dlPFC activity diminished
614 strategic choices (Müller-Leinß et al., 2018; Zinchenko et al., 2021). In sum, our findings are
615 consistent with the IFG and AI being involved in norm compliance decisions.

616
617 Additionally, the results extend on past literature through investigating how reward processes
618 and cortical connectivity interact with strategic behavior. The results indicate that elevated IFG-
619 rpTPJ connectivity is associated with increased strategic behavior, whereas attenuated AI-
620 Angular Gyrus connectivity is modulated by the interaction of reward sensitivity and strategic
621 behavior. Although recent work has shown that the dlPFC and rpTPJ regulate norm compliance
622 in the UG and DG (Gianotti et al., 2018), and that the right TPJ does not necessarily yield
623 greater generosity (Brethel-Haurwitz et al., 2022), the results indicate that strategic decision
624 making in social situations modulates the connectivity between the dlPFC and TPJ.

625 Understanding how connectivity modulates strategic decisions is a critical component of
626 characterizing how the TPJ and dlPFC may be regulated during decision making, with the TPJ
627 potentially orienting the IFG toward changes to social context and thus greater opportunities to
628 be strategic. Additionally, when including reward sensitivity as a covariate, the results indicated
629 that people with varying levels of trait reward sensitivity respond to strategic decisions
630 differently. Specifically, people with low reward sensitivity are more strategic with decreasing AI-
631 Angular Gyrus connectivity, whereas people with higher reward sensitivity are more strategic
632 with increasing AI-Angular Gyrus connectivity.

633

634 It has been previously found that reward sensitivity is associated with risky behavior (Scott-
635 Parker & Weston, 2017), higher Machiavellianism (Birkás et al., 2015), and more strategic
636 behavior (Scheres & Sanfey, 2006). This yields an interpretation that reward sensitivity could be
637 a factor in guiding norm compliance in social situations as people with higher reward sensitivity
638 may be more motivated to maximize their own rewards. Specifically, reward sensitivity may
639 modulate strategic decisions by increasing the degree people are self-oriented, and their
640 willingness to take risk even at the potential of being rejected in a bargaining situation. Thus, AI-
641 Angular Gyrus connectivity may modulate how people experience opportunities to cooperate
642 and defect, which could affect how people employ social heuristics in bargaining situations
643

644 We speculate that among self-interested people who aim to maximize earnings, reward
645 sensitivity may modulate strategic decisions through increasing attentional processes to
646 changes in social context through AI-Angular Gyrus connectivity. Specifically, connectivity
647 between the AI-Angular Gyrus may serve as a mechanism for overriding fairness norms to
648 share evenly with their partner by orienting people to changes in social context. This process
649 could be driven through bottom-up attention, or through top-down cognitive mechanisms.
650 Specifically, the angular gyrus is implicated in bottom-up attentional processing (Cabeza et al.,
651 2012; Seghier, 2013), interpreting contextual information (Carter & Huettel, 2013; Ramanan et
652 al., 2018), and social cognition (Numssen et al., 2021). The AI integrates fairness, empathy, and
653 cooperation (Cheng et al., 2017; Lamm & Singer, 2010). Given this, it is plausible that AI-
654 Angular Gyrus connectivity could help bottom-up orientation of changes in context affecting
655 social norms. Alternatively, AI engagement could reflect differences in top-down cognitive
656 control among participants (Sridharan et al., 2008), and AI-angular gyrus connectivity may
657 reflect top-down orientation to the changes in social context. Additionally, AI-Angular Gyrus
658 connectivity may be modulated by reward sensitivity. Reward sensitivity is associated with risky
659 behavior (Scott-Parker & Weston, 2017), higher Machiavellianism (Birkás et al., 2015), and

660 more strategic behavior (Scheres & Sanfey, 2006). Thus, AI-Angular Gyrus connectivity may
661 modulate how people experience opportunities to cooperate and defect, which could affect how
662 people employ social heuristics in bargaining situations.

663

664 One interpretation is that people with higher reward sensitivity may be more motivated in the
665 task and may be more likely to defect in bargaining tasks. Increased deliberation may, in turn,
666 override default fairness norms. This deliberative process may modulate bottom-up attention or
667 contextual orienting in the Angular Gyrus, or top-down cognitive processing in the AI. Our
668 results suggest a nuanced view of AI-Angular Gyrus and IFG-TPJ coupling (Lockwood et al.,
669 2020), indicating that these brain regions do not necessarily reflect altruistic choice on their own
670 (Hutcherson et al., 2015), but may modulate cognitive reward processes while making social
671 decisions. Additionally, we speculate that our results reflect that downregulation of bilateral TPJ
672 activation and AI deactivation (FeldmanHall et al., 2014) interacts with trait reward sensitivity.
673 Specifically, our findings may provide insight into how people with aberrant levels of reward
674 sensitivity respond to strategic decisions in bargaining situations. The results indicated that
675 people with lower reward sensitivity had higher AI-Angular Gyrus connectivity when being less
676 strategic, whereas people with higher reward sensitivity had higher connectivity when being
677 more strategic. If higher AI-Angular Gyrus connectivity is a reflection of increased motivation
678 among participants, the results suggest that trait reward sensitivity may inform strategic
679 behavior and how people employ social heuristics to be fair or selfish. Specifically, people who
680 have higher reward sensitivity may need to have greater AI-Angular gyrus connectivity to be
681 more strategic compared to people who have lower reward sensitivity. Additionally, since
682 aberrant reward sensitivity is a predictor for elevated substance use, investigating how reward
683 sensitivity modulates brain processes in social contexts could provide insight into how people
684 make decisions resulting in substance use (Bart et al., 2021; Heilig et al., 2016; Wyngaarden et
685 al., 2023).”

686

687 Although our work has found that strategic behavior is modulated by both AI-Angular Gyrus and
688 IFG connectivity with the TPJ, and reward sensitivity, we acknowledge that our study has
689 several limitations that merit discussion. First, although the results suggest bilateral TPJ
690 connectivity and strategic behavior, we do not infer specificity in lateralization. Past
691 investigations suggest mixed findings (Carter et al., 2012; Coricelli & Nagel, 2009; Saxe &
692 Kanwisher, 2003) as to the roles of the right and left TPJ, and we judged that exploring these
693 results further was beyond the scope of this paper. Additionally, we acknowledge that
694 connectivity analyses are not causal or directional with respect to inference despite identifying
695 the IFG and AI as seeds and the temporoparietal junction as target. Further, since strategic
696 behavior as a proposer was not related to recipient choices, we judged that these results are
697 beyond the scope of this investigation. A possible future direction includes evaluating AI-Angular
698 Gyrus and IFG-TPJ connectivity patterns, associations with reward sensitivity, and their
699 relations with recipient decisions in the Ultimatum Game.

700

701 Second, we acknowledge that fMRI analysis techniques carry elevated risk of Type II errors.
702 The results reported in the manuscript are a product of whole-brain analyses which were
703 conservatively thresholded to control for multiple comparisons whereas our confirmatory ROI-
704 based analyses registered null results. In line with recommended practices (Gentili et al., 2021),
705 we pre-registered and conducted ROI-based analyses to increase power to detect activation
706 and connectivity by limiting multiple comparisons (Poldrack et al., 2007). Secondary whole-brain
707 analyses naturally follow ROI analyses if appropriately thresholded (Poldrack, 2007; Szycik et
708 al., 2009) and were reported accordingly. Nonetheless, conducting brain-wide association tests
709 with individual difference measures may be underpowered (Marek et al., 2022). Thus, while the
710 sample included people with high and low reward sensitivity and conducted rigorous test-retest
711 with SR and BIS/BAS to ensure that participants were consistent across these measures, we

712 acknowledge that relations with reward sensitivity should be considered exploratory in nature.

713 Future analyses could examine how reward sensitivity modulates brain responses using

714 multivariate methods to improve effect size estimation (Reddan et al., 2017) with canonical

715 correlation analysis (Zhuang et al., 2020), multivariate pattern analysis (Kragel & LaBar, 2015)

716 or machine learning algorithms to assess neural signatures (Wager et al., 2013) of bargaining.

717

718 Third, we note that relations with social context, reward sensitivity, and brain connectivity could

719 be studied more extensively in a clinical population to assess how these relations are modulated

720 by substance use and manic-depressive symptoms. Whereas we were able to control for levels

721 of substance use to account for reward sensitivity effects (Joyner et al., 2019), the sample had

722 too limited variability in substance use to make inferences about how substance use may

723 contribute to maladaptive strategic decisions. Additionally, while we assessed strategic

724 behavior, we did not assess it in a dynamic context. As social contexts increase exploration and

725 obtained rewards (Plate et al., 2023), a fruitful future direction could investigate how brain

726 responses to changes over time reflect social decisions.

727

728 A final notable limitation was that we did not find evidence that suggests ventral striatal

729 activation or connectivity is related to strategic behavior. Past investigations suggested that

730 higher VS activation was associated with more strategic behavior (Spitzer et al., 2007), with

731 more unfair offers in UG being associated with higher VS activation (Chen et al., 2017). Thus, it

732 is possible that the lack of VS activation was due to participants not finding the differences in

733 offers sufficiently salient, or not being sufficiently incentivized by the small differences in

734 rewards between UG and DG, or potentially that we were underpowered to detect these effects.

735 Alternatively, some individuals may have increased VS activation that may be responding to

736 prosociality, when giving more money to their partner. Across the DG, studies have found

737 increased VS activation for keeping more for oneself (Tricomi et al., 2010), and for sharing with

738 others (Moll et al., 2006). In the UG, the VS tracked inequity in both prosocial and individualistic
739 people (Haruno et al., 2014). Thus, it is possible that in our sample we had individuals that had
740 higher VS activation and acted least strategically toward maximizing their own earnings. Future
741 studies may be able to investigate if there is higher VS activation between people who maximize
742 earnings for themselves or for others across the UG and DG tasks.

743
744 Despite the limitations, our findings indicate that strategic decisions in social contexts are
745 associated with elevated IFG-TPJ connectivity and that AI-Angular Gyrus connectivity while
746 making strategic decisions is modulated by trait reward sensitivity. These results provide greater
747 insights into how reward processes interact with social decisions, involving brain processes that
748 appraise the roles of other people while making choices. Since aberrant reward sensitivity is a
749 major mechanism in substance use and depressive and bipolar disorders, investigating how
750 reward sensitivity modulates brain processes during social contexts could provide considerably
751 more understanding into how people make maladaptive decisions resulting in substance use
752 (Bart et al., 2021; Heilig et al., 2016; Wyngaarden et al., 2024). Such work could help identify
753 people at risk for substance use disorders and help develop interventions for people with
754 aberrant reward patterns.

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