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1	Neural predictors of evaluative attitudes towards celebrities
2	
3	Abbreviated Title: Decoding social attitude
4	
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1 ABSTRACT

2 Our attitudes towards others influence a wide range of everyday behaviors, and have been the most extensively studied concept in the history of social psychology. Yet 3 4 they remain difficult to measure reliably and objectively, since both explicit and 5 implicit measures are typically confounded by other psychological processes. We here address the feasibility of decoding incidental attitudes based on brain activations. 6 7 Participants were presented with pictures of members of a Japanese idol group inside 8 an fMRI scanner while performing an unrelated detection task, and subsequently 9 (outside the scanner) performed an incentive-compatible choice task that revealed 10 their attitude toward each celebrity. We used a real-world election scheme that exists 11 for this idol group, which confirmed both strongly negative and strongly positive 12 attitudes towards specific individuals. Whole-brain multivariate analyses (searchlight-13 based support vector regression) showed that activation patterns in the anterior 14 striatum predicted each participant's revealed attitudes (choice behavior) using leave-15 one-out (as well as 4-fold) cross-validation across participants. By contrast, attitude 16 extremity (unsigned magnitude) could be decoded from a distinct region in the 17 posterior striatum. The findings demonstrate dissociable striatal representations of 18 valenced attitude and attitude extremity, and constitute a first step toward an objective 19 and process-pure neural measure of attitudes.

20

21 Key Words: attitude, attitude extremity, preference, fMRI, MVPA, striatum

1	Attitudes comprise our evaluations of an object, a place, an idea, another person,
2	or oneself: which are good or bad, which do we want to approach or avoid? Attitudes
3	influence a wide range of human behaviors ranging from everyday social interactions
4	with other people, health behavior, and political behavior to international relations.
5	Although attitude is one of the most extensively studied concepts in the entire history
6	of social psychology (Petty et al., 2009), it has remained a challenge how best to
7	objectively and accurately measure people's attitudes. Problems with self-report
8	measures have been well-documented in the past (in particular, social desirability
9	bias) (DeMaio, 1984; Podsakoff et al., 2003). While implicit measures of attitudes
10	(Wittenbrink and Schwarz, 2007), such as the implicit association test (IAT;
11	Greenwald et al., 1998), have offered partial solutions to this problem, their validity is
12	also debated (e.g., Bosson et al., 2000; Karpinski and Hilton, 2001; Fazio and Olson,
13	2003; Arkes and Tetlock, 2004; Conrey et al., 2005; Blanton et al., 2006; Fiedler et
14	al., 2006; Sherman, 2009). For example, scores on the IAT are influenced by multiple
15	cognitive processes, not only implicit attitude (e.g., Conrey et al., 2005; Sherman,
16	2009).

17 In the present study, we aimed to test the feasibility of measuring people's 18 attitudes toward other familiar people based on the power of their brain activations to 19 predict choice behavior, and without requiring any explicit or implicit task in the 20 scanner. A neural measure of incidental attitudes would have significant potential to 21 provide a process-pure metric, and avoid the contamination with many other 22 processes that limits currently available explicit as well as implicit measures of 23 attitude (Conrey et al., 2005; Sherman, 2009). Our approach used multi-voxel pattern 24 analysis (MVPA) together with support vector regression on functional magnetic 25 resonance imaging (fMRI) signals. While conventional univariate fMRI analysis

1	compares the strength of activations in each single voxel independently, MVPA
2	classifies the distributed patterns of activations across multiple voxels in a high-
3	dimensional space, and can be more sensitive for detecting and distinguishing
4	different psychological states (e.g., Vickery et al., 2011; Jimura and Poldrack, 2012).
5	In the present study, we particularly focus on people's attitudes toward social
6	objects, familiar people. Although many past social psychological studies as well as
7	social neuroscience studies on attitudes have focused on racial attitudes (e.g., Phelps
8	et al., 2000; Stanley et al., 2011; for reviews, Ito and Bartholow, 2009; Kubota et al.,
9	2012; Amodio, 2014), studies of familiar famous people offer some advantages:
10	unlike racial attitude, the relationship between self-reported attitude and behavior can
11	be more straightforward (e.g., less susceptible to social desirability bias) and thus the
12	relationship among neurally measured attitude, self-report attitude, and behavior
13	(revealed attitude) is more easily interpretable.
	Furthermore, in addition to testing the feasibility of inferring people's attitudes
14	
14 15	based on brain activations (a question of psychological interest regardless of the
14 15 16	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural
14 15 16 17	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural representations of attitudes and attitude extremity (i.e., how extreme attitude is
14 15 16 17 18	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural representations of attitudes and attitude extremity (i.e., how extreme attitude is regardless of its valence). While neuroeconomics studies have extensively
14 15 16 17 18 19	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural representations of attitudes and attitude extremity (i.e., how extreme attitude is regardless of its valence). While neuroeconomics studies have extensively investigated neural representations of attitude toward non-social objects (e.g., foods,
14 15 16 17 18 19 20	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural representations of attitudes and attitude extremity (i.e., how extreme attitude is regardless of its valence). While neuroeconomics studies have extensively investigated neural representations of attitude toward non-social objects (e.g., foods, DVDs, etc.) (for review, Levy and Glimcher, 2012; Clithero and Rangel, 2014), only
14 15 16 17 18 19 20 21	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural representations of attitudes and attitude extremity (i.e., how extreme attitude is regardless of its valence). While neuroeconomics studies have extensively investigated neural representations of attitude toward non-social objects (e.g., foods, DVDs, etc.) (for review, Levy and Glimcher, 2012; Clithero and Rangel, 2014), only few studies (e.g., Cunningham et al., 2003; Knutson et al., 2006; Cunningham et al.,
 14 15 16 17 18 19 20 21 22 	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural representations of attitudes and attitude extremity (i.e., how extreme attitude is regardless of its valence). While neuroeconomics studies have extensively investigated neural representations of attitude toward non-social objects (e.g., foods, DVDs, etc.) (for review, Levy and Glimcher, 2012; Clithero and Rangel, 2014), only few studies (e.g., Cunningham et al., 2003; Knutson et al., 2006; Cunningham et al., 2008; Tusche et al., 2013) investigated the neural representation of attitude toward
 14 15 16 17 18 19 20 21 22 23 	based on brain activations (a question of psychological interest regardless of the neuroanatomical details), the present study also investigates the regional neural representations of attitudes and attitude extremity (i.e., how extreme attitude is regardless of its valence). While neuroeconomics studies have extensively investigated neural representations of attitude toward non-social objects (e.g., foods, DVDs, etc.) (for review, Levy and Glimcher, 2012; Clithero and Rangel, 2014), only few studies (e.g., Cunningham et al., 2003; Knutson et al., 2006; Cunningham et al., 2008; Tusche et al., 2013) investigated the neural representation of attitude toward other familiar people, despite the high relevance of this topic to our everyday social

25 attitude properties (Petty & Krosnick, 1995). For example, attitude extremity affects

an individual's information processing (e.g., Powell and Fazio, 1984; Van Boven et
al., 2012) and modulates the relationship between explicit and implicit attitudes
(Karpinski et al., 2005). However, its neural mechanisms also remain largely
unexplored (see Cunningham et al., 2008; Luttrell et al., 2016, for a notable
exception).

6 In the fMRI scanner, participants were presented with 10 members of a Japanese 7 female idol group (the Japanese music performance group "AKB48"). After the 8 scanning session, they performed an incentive-compatible choice task, which 9 behaviorally quantified each participant's attitude toward each of the 10 members 10 (with a real-world outcome). MVPA was applied to these data to identify those brain 11 regions that could predict a participant's choice behavior toward each member. Since 12 different idol group members were in fact associated with idiosyncratic preferences 13 amongst our participants, the neural data should uniquely encode the attitude for a 14 member that predict the preference choice, unconfounded by the perceptual 15 appearance of the member. We expected to find neural signatures of social attitude 16 within those brain regions previously associated with value representations in general, 17 including striatum, ventromedial prefrontal cortex (vmPFC), insula, amygdala, and 18 anterior cingulate cortex (ACC) (e.g., Buchel et al., 1998; LaBar et al., 1998; Delgado 19 et al., 2000; Knutson et al., 2000; O'Doherty et al., 2001; Knight et al., 2005; Hare et 20 al., 2008; Izuma et al., 2008; for meta-analyses, Etkin et al., 2011; Bartra et al., 2013; 21 Sescousse et al., 2013). Nonetheless, in order to obtain a data-driven set of results, our 22 primary analysis used a whole-brain approach.

23

24 MATERIALS and METHODS

2 Participants

3	A total of 23 college students participated in the study. One participant was
4	excluded from the analysis due to excessive head motion, and the remaining 22
5	participants were included in the analyses (11 female; mean age = 19.7 , SD = 1.35).
6	All participants were pre-screened so that all of them knew at least 20 members of the
7	idol group and had at least one highly liked member and one highly disliked member.
8	The participants were all right-handed with no history of neurological or psychiatric
9	illness. All participants gave written informed consent for participation, and the study
10	was approved by the Institutional Review Board of Tamagawa University.
11	
12	Stimuli
13	Photographs of the faces of individual members of the Japanese female idol group
14	"AKB48" (https://en.wikipedia.org/wiki/AKB48) were used in the present study. All
15	stimuli were obtained from the Internet. AKB48 consists of more than 100 members
16	and is popular especially among young Japanese people. We decided to use this idol
17	group for our source of experimental stimuli for two reasons: first, and most
18	importantly, their unique annual election contest allows us to have an incentive-
19	compatible choice task with real-world validity (see below for more detail on the
20	choice task). Every year, the idol group has a unique annual election contest as a
21	marketing strategy. Before the election contest, their new album is released, which
22	includes a voting code. Each fan can vote for his/her favorite member using the code,
23	and this election contest has a significant influence on each member's media
24	exposure. The more votes a member receives, the more heavily she will be promoted.

This election system allows each fan to vote more than once by purchasing many albums. Second, related to the first point, because of the competitive relationship among members of the group created by the election contest, each fan (participant) typically has a positive attitude toward some members and a negative attitude toward other members. This large variance in attitudes provides an ideal real-world background for the purpose of the current study.

7 During the fMRI scanning, each participant was presented with pictures of 10 8 different members. Since we were interested in the neural representation of already 9 established social attitudes (as opposed to first impressions toward people never seen 10 before), we ensured that each participant was familiar with 10 members presented 11 during the fMRI scanning with the following procedure. Before an fMRI experiment, 12 each participant was asked to provide the experimenter with names of his/her most 13 and least favorite members (at least one for each) by email. Based on the names each 14 participant gave, the experimenter picked 10 members for the participant including 15 his/her favorite and least favorite members. If participants named fewer than 10 16 members, the experimenter selected other members based on general popularity. On 17 the day of the fMRI experiment, before they entered an fMRI scanner, each 18 participant was shown pictures of the 10 members and asked if they could identify all 19 of them. If there were any members that could not be identified, participants were 20 asked to pick other members they knew from a list of all members' names and 21 pictures. Accordingly, picture stimuli used in the fMRI experiment were different for 22 each participant, although there was overlap. More specifically, a total of 43 different 23 members of the idol group were used in the experiment, and each of them was 24 presented to at least one participant. 15 out of the 43 members (34.9%) were 25 presented to only one participant. The most consistently used member was presented

to 15 different participants. However, these 15 participants had idiosyncratic attitudes
toward the member ranging from revealed attitude scores of 3 to 17 (possible range =
0 - 18; see below for more information on the revealed attitude score). Thus, the
results of the across-participant MVPA reported below are highly unlikely to be
explained by decoding of person identity or face features (e.g., hair length).

6

7 Experimental tasks

8 In each trial, a picture of a member of the idol group was presented for 4 seconds, 9 and participants were asked to perform a simple button press task during the fMRI 10 session. At a random point between 1.5 to 3 seconds after the onset of the picture 11 presentation, the picture became darker for 0.2 sec (Fig.1a). Participants were asked 12 to press the button as soon as they detected the luminance change. The inter-trial 13 interval (ITI) was set to 4, 6 or 8 seconds (pseudo-randomly determined). In each 14 fMRI run, each of 10 members was presented 3 times (30 trials). Each run lasted 5 15 min, and there were 8 fMRI runs in total. Importantly, before the scanning, 16 participants were not told that the experiment was about attitudes, and they were not 17 explicitly asked to think about their attitude toward each member during the scanning.



Figure 1. Experimental tasks. (a) A single trial during the fMRI scanning. Each participant was asked to press a key when the luminance of a picture changed. (b) A single trial of the choice task. After the fMRI session, each participant performed the choice task. In each trial, two members of the idol group were presented on the screen, and participants were asked to choose the member they want to vote for at the next election event. Note that due to copyright restrictions, the two individuals depicted in the pictures in this figure are not actual members of the idol group.

8

24

9 After the fMRI session, participants were asked to perform a choice task outside 10 of the scanner (they were not told before the scanning that there would be the choice 11 task). In each trial, two members from the 10 members used in the fMRI task were 12 presented on the computer screen (Fig.1b), and participants were asked to select the 13 one they want to vote for at the next election event by pressing one of two keys on the 14 keyboard. There were 45 unique choice pairs, and each pair was presented twice (with 15 switched positions) so that participants made 90 binary choices in total. During the 16 choice task, each of the 10 members was presented 18 times. Therefore, each member 17 could be selected a minimum of 0 times and maximum of 18 times. Since any attitude 18 measure is considered to be good so long as it can predict relevant behaviors, this 19 "revealed attitude score" for each member was used as labels in the subsequent 20 MVPA analysis. Thus, in this project, we aim to predict individual's choice behavior 21 (i.e., revealed attitude score) based on brain activations that would encode attitude, 22 but without relying on any psychological measures of attitude (e.g., self-report). 23 Importantly, before they started the choice task, all participants were instructed

vote for the member the participant selected in that trial. All participants were shown

that after the choice task, one trial would be selected randomly, and we would actually

26 a new music album including a voting code, and using the code, the experimenter

actually voted for a member selected by the participant in a randomly-selected trial.

Finally, participants were asked to rate each of 10 members on attitude (how much do you like this member?) and attractiveness (how attractive do you think this member is) using a 9-point scale. At the end of the experiment, all participants received the album in addition to the monetary compensation (8,000 Japanese yen) for their participation.

6

7 fMRI data acquisition

8 All fMRI data were acquired using a Siemens 3.0 Tesla Trio scanner with a 32 channel phased array headcoil. For functional imaging, interleaved T2*- weighted 9 10 gradient-echo echo-planar imaging (EPI) sequences were used to produce 34 11 contiguous 3.5-mm-thick trans-axial slices covering nearly the entire cerebrum 12 (repetition time [TR] = 2,000 ms; echo time [TE] = 25 ms; flip angle $[FA] = 90^{\circ}$; field 13 of view [FOV] = 192 mm; 64×64 matrix; voxel dimensions = $3.0 \times 3.0 \times 3.5$ mm). A 14 high-resolution anatomical T1-weighted image (1 mm isotropic resolution) was also 15 acquired for each participant.

16

17 fMRI data preprocessing

The fMRI data were analyzed using SPM8 (Wellcome Department of Imaging Neuroscience) implemented in MATLAB (MathWorks). Before data processing and statistical analysis, we discarded the first four volumes to allow for T1 equilibration. After correcting for differences in slice timing within each image volume, head motion was corrected. Following motion correction, the volumes were normalized to MNI space using a transformation matrix obtained from the normalization of the first EPI image of each individual participant to the EPI template (resliced to a voxel size

1 of $3.0 \times 3.0 \times 3.5$ mm). These normalized data were used for the MVPA data

2 analyses. For the univariate analysis, the normalized fMRI data were spatially

3 smoothed with an isotropic Gaussian kernel of 4 mm (full-width at half-maximum).

4

5 fMRI data analysis: searchlight MVPA

6 In the across-participant MVPA analysis, we attempt to predict attitudes toward 7 each of 10 idol group members based on the data obtained from all other participants. 8 We first ran a conventional general linear model (GLM) analysis. In the GLM, each of 9 10 members was separately modeled (duration = 4 sec). Button presses (duration = 010 sec) and head motions were also included in the model as nuisance regressors. 11 Contrast images for each member were created by using the data from all of the 8 12 fMRI runs. These contrast images were used as input, and revealed attitude scores 13 (how many times each member was chosen during the choice task; 0-18) were used as 14 labels in the MVPA analysis. 15 MVPA was performed by using custom-made MATLAB in combination with

16 LIBSVM (http://www.csie.ntu.edu.tw/~cjlin/libsvm/). In order to predict the

17 parametric variable of revealed attitude scores, we employed support vector

18 regression (SVR; Drucker et al., 1997), as implemented in LIBSVM, with a linear

19 kernel and a cost parameter of c = 0.01. This cost parameter was selected a priori

20 following a previous study investigating value and salience signals in the brain (Kahnt

et al., 2014). The regression MVPA was performed using a whole brain searchlight

- 22 procedure (Kriegeskorte et al., 2006) with a radius of 3 voxels (maximum of 123
- 23 voxels, and less at the boundaries of the brain). In each searchlight, accuracy at
- 24 predicting revealed attitude scores for the 10 members was computed using leave-

1 one-participant-out cross validation. In each cross-validation, one participant was left 2 out, and the SVR was performed using the data from all other participants and then 3 tested on the left-out participant. This procedure was repeated for each participant (a 4 total of 22 times; to test the robustness of our findings, we also ran a 4-fold crossvalidation which replicated the main findings [see Supplementary Results and Table 5 6 S2]). In each searchlight analysis, Spearman's rank-order correlations were computed between a participant's revealed attitude scores and their predicted attitudes; this 7 8 correlation value was then assigned to the center voxel of the searchlight, resulting in 9 an anatomical correlation map for each participant. The correlation values were Fisher 10 z-transformed, spatially smoothed with an isotropic Gaussian kernel of 4 mm (full-11 width at half-maximum), and then submitted to the second level analysis (i.e., one 12 sample t-tests across all participants).

In order to identify the neural representations of attitude extremity, we also ran the same across-participant MVPA analysis using the attitude extremity score, which is computed by calculating the absolute value of the difference between a revealed attitude score for each member and the midpoint of the revealed attitude score (i.e., 9). We further ran three control MVPA analyses using 1) self-report attractiveness ratings, 2) self-report attitude ratings, and 3) reaction times (RTs) (see Supplementary Information).

20

21 fMRI data analysis: Univariate analysis

We also ran a standard univariate fMRI analysis to see whether univariate activations might be correlated with revealed attitude score. The GLM included three main regressors; 1) all idol group member presentations (duration = 4 sec), 2) member

1	presentations parametrically modulated by participant's revealed attitude score, and 3)
2	member presentations parametrically modulated by participant's attractiveness rating.
3	Button presses (duration = 0 sec) and head motions were also included in the model
4	as nuisance regressors.
5	For both the MVPA and univariate analyses, the statistical threshold was set at p $\!<\!$
6	0.001 voxelwise (uncorrected) and cluster $p < 0.05$ (FWE corrected for multiple
7	comparisons).
8	
9	RESULTS
10	Behavioral results
11	Not surprisingly, revealed attitude scores (choice behavior) were highly
12	correlated with self-reported attitudes (average $r = 0.91$, $t(21) = 18.11$, $p < 0.001$).
13	They were also correlated with attractiveness ratings (average $r = 0.76$, $t(21) = 10.80$,

14 p < 0.001). The revealed attitude scores were more strongly correlated with the 15 attitude ratings than the attractiveness ratings (t(21) = 3.33, p = 0.003). Attitude 16 ratings and attractiveness ratings were also correlated with each other (average r = 17 0.76, t(21) = 10.69, p < 0.001).

Inside the fMRI scanner, participants were asked to press the button as soon as a picture gets dimmed, and their performance for this simple button press task was nearly perfect (98.6%), and average RT across participants was 297 ms (SD = 52), indicating that participants paid attention to each picture stimulus. The analyses also revealed that RTs were significantly negatively correlated with participants' revealed attitude scores (average r = -0.20, t(21) = -3.14, p = 0.005) and attitude ratings (average r = -0.16, t(21) = 2.27, p = 0.034). Thus, the more favorable their attitudes

1 were toward members, the faster (smaller) the RT. This result may suggest that 2 pictures of their favorite members captured attention relative to less favorite 3 members, which in turn enhanced their reaction times. RTs were not correlated with 4 the attractiveness ratings (average r = -0.08, t(21) = 1.24, p = 0.23, n.s.) or the attitude extremity scores (average r = -0.12, t(21) = 1.30, p = 0.21, n.s.). 5 6 After the scanning, participants were asked to make binary choices between two 7 members of the idol group for whom they wanted to vote in the next election event. 8 They were presented with the same choice pair twice, and their choices were largely

9 consistent across two presentations of the same pair (choice consistency = 86.7%).

10 Inconsistent choices were more likely to happen when the difference in attitude

11 ratings between two members was small (e.g., when two persons were similarly

12 liked). When the within-pair ratings difference was equal to or less than 2, choice

13 consistency was 79.3%, while it was 94.0% when the difference was more than 2.

14 This difference between the two choice consistency values was significant (t(21) =

15 5.32, p < 0.001). Not surprisingly, participants' choices were highly accurately

16 predicted by their self-report attitudes (mean choice prediction accuracy = 88.5%),

17 and the accuracy was significantly higher than chance (50%) (t(21) = 20.1, p < 0.001;

18 note that choice accuracy was computed after excluding inconsistent choice pairs).

19

20 fMRI results: searchlight MVPA

The searchlight MVPA analysis revealed that spatial activation patterns in the anterior part of the right striatum significantly predicted participants' revealed attitude scores (Fig. 2a). The average Spearman's correlation coefficient in the peak of this anterior striatum cluster was 0.26 (t(21) = 7.24, p < 0.001). Activation patterns in

right inferior frontal gyrus (IFG) also significantly predicted participants' attitudes (Spearman's Rho = 0.23, t(21) = 6.49, p < 0.001; Fig. 2a, Table 1). No other region significantly predicted participants' revealed attitude scores. The average correlation of 0.26 found in the anterior striatum is equivalent to 59.6% in terms of accuracy for predicting participants' binary choices (see Supplementary Results for more details about the choice prediction accuracy results and the direct comparison of the choice prediction accuracy between neural and self-report measures).

8 The MVPA analysis with the attitude extremity score revealed that activation 9 patterns in a more posterior part of the right striatum significantly predicted attitude 10 extremity (Fig. 2a). Activation patterns in other brain regions including inferior 11 orbitofrontal cortex (OFC), supplementary motor area (SMA), posterior insula, 12 precentral gyrus, inferior parietal lobule and occipital pole also significantly predicted 13 attitude extremity (Table 1).

14 To further quantify the functional dissociation within the striatum, we extracted 15 each participant's correlation coefficients for both the revealed attitude score and the 16 attitude extremity score from the voxels within a 4mm sphere surrounding the peaks 17 of these two striatum sub-regions (note that the following statistical analysis is of 18 course not statistically independent of our above discovery; it is intended merely to 19 provide further detail for the dissociation). After Fisher's z transforming these 20 correlation coefficients, we performed a 2 (anterior vs. posterior striatum) x 2 21 (revealed attitude score vs. attitude extremity score) repeated-measure analysis-of-22 variance (ANOVA). It revealed a significant interaction (F(1, 21) = 26.4, p < 0.001) 23 (Fig. 2b). We further performed a paired t-test within each of the anterior and 24 posterior striatum regions. In the anterior striatum, prediction performance (mean 25 correlation) was significantly higher for revealed attitude compared to attitude

1	extremity (t(21) = 4.24, $p < 0.001$). In contrast, in the posterior striatum, prediction
2	performance was significantly higher for attitude extremity than for revealed attitude
3	(t(21) = 3.48, p = 0.002) (Fig. 2b). Furthermore, correlations were not significantly
4	different from zero for attitude extremity in the anterior striatum and for revealed
5	attitude in the posterior striatum. These results indicate a clear functional dissociation
6	within the striatum such that the anterior striatum represents an individual's
7	evaluation of each celebrity, whereas the posterior striatum represents how extreme
8	his/her attitude is regardless of valence. Main effects of region and attitude property
9	were both not significant (ps $>$ 0.50). Finally, a mixed ANOVA with gender as an
10	additional between-subject factor revealed no main or interaction effect involving
11	gender (all $ps > 0.35$).
12	Our control analyses further confirmed that the MVPA results reported above
13	cannot be explained by perceived attractiveness of faces or reaction times (i.e.,
14	attention) (see Supplemental Results, Figure S1 and Table S1). Furthermore,
15	additional searchlight MVPA analysis restricted to an anatomical mask of the striatum
16	confirmed that activation patterns within the striatum are responsible for the findings
17	reported above, and could not be ascribed to partial volume effects from nearby
18	regions (see Supplemental Results).



Figure 2: fMRI results. (a) Activation patterns in the anterior striatum significantly predicted participants' revealed attitude scores, whereas activation patterns in the posterior striatum (blue) predicted attitude extremity. (b) Mean within-participant Spearman's rank correlation between predicted values and actual values (revealed attitude or attitude extremity) from the anterior and posterior striatum. ** p < 0.01,

7 *** p < 0.001. Error bars represent standard error of mean (SEM).

1 Table 1. Regions where activity patterns significantly predicted revealed attitude

		MNI coordinates				Number	
Location	Side	Х	У	Z	z value	of Voxel	
Revealed attitude score							
Anterior striatum	R	18	26	6	5.07	37	
Inferior frontal gyrus (IFG)	R	54	14	20	4.76	56	
Attitude extremity score							
posterior striatum	R	6	5	10	4.56	31	
inferior orbitofrontal cortex (OFC)	R	30	38	-4	3.90	25	
supplementary motor area (SMA)	R	15	2	66	4.82	26	
posterior insula	L	-48	-7	2	4.00	46	
precentral gyrus	L	-33	-19	62	5.07	63	
inferior parietal lobule	L	-51	-31	44	4.38	48	
occipital pole	R	30	-95	-15	4.48	45	

2 score and attitude extremity score.

3 Note that the size of a voxel is $3 \times 3 \times 3.5$ mm

4

5 fMRI results: Univariate analysis

6	The results of the univariate fMRI data analysis revealed that the activity only in
7	the left posterior fusiform gyrus (x = -27, y = -91, z = -15; 60 voxels) was
8	significantly positively correlated with the revealed attitude scores. Only when the
9	threshold was lowered to $p < 0.005$, did we find activations in the ventromedial
10	prefrontal cortex (vmPFC; $x = -6$, $y = 53$, $z = -12$), one of the areas commonly
11	associated with preference and valuation (see Bartra et al., 2013; Clithero and Rangel,
12	2014). We did not find any activation in the striatum even with this lowered threshold.
13	In addition, no area was significantly negatively correlated with the revealed attitude
14	scores. Furthermore, no area was significantly correlated (either positively or
15	negatively) with the attitude extremity scores.

2 **DISCUSSION**

3 The present study investigated the possibility of measuring people's social attitude 4 toward familiar others based on multivariate neural activations from fMRI. The 5 across-participant MVPA revealed that activation patterns in the anterior striatum can 6 significantly predict the choices made based on one's attitude toward members of an 7 idol group. This result indicates that spatial patterns of activations in the anterior 8 striatum contain reliable information about an individual's attitudes, and these neural 9 representations of attitudes in the anterior striatum are sufficiently similar across 10 different individuals so that it is possible to infer attitudes of an individual based on 11 the association between revealed attitudes and brain activation patterns found in other 12 individuals. In contrast, our univariate analysis failed to find any significant 13 activations related to participants' attitudes (and attitude extremity) in reward-related 14 areas, indicating that average amplitude of activity is an insufficiently sensitive 15 measure to represent attitudes in the present study. Although previous studies have 16 demonstrated that univariate activations in reward-related brain regions such as 17 vmPFC and striatum are correlated with people's attitudes or preference for various items, especially when individuals were asked to report their preference for each 18 19 stimulus inside the scanner (Izuma et al., 2010; Lebreton et al., 2009; for review, see 20 Bartra et al., 2013), such univariate (mean) activation does not appear to robustly 21 track preference for stimuli during passive viewing (but see also Levy et al., 2011 and 22 Tusche et al., 2013). In general, our results add to a growing body of evidence 23 showing the higher sensitivity of MVPA compared with univariate analyses (e.g., 24 Jimura & Poldrack, 2012; Kohler et al., 2013).

1	Although we found that activation patterns in the anterior striatum and right IFG
2	can predict participants' attitudes revealed in the choice task (i.e., behavior)
3	significantly better than would be expected by chance, prediction accuracy was
4	considerably lower than their self-reported attitude ratings. Thus, at this point, self-
5	report measure outperforms neural measures of attitudes. This result is not surprising
6	because self-report attitude toward celebrities should be much less susceptible to
7	social desirability bias compared to attitudes toward socially sensitive issues (e.g.,
8	racial prejudice), and there is no apparent reason for participants to regulate their
9	answers during attitude ratings of the celebrities. Nonetheless, our present study
10	provides an important reference point to which future social neuroscience studies can
11	be compared. For example, it might well be the case that neurally measured attitude
12	can outperform self-report or implicit measures of attitudes in predicting racially
13	discriminatory behavior for which prediction accuracy of behavioral measures (both
14	implicit and explicit measures) are known to be limited (Greenwald et al., 2009;
15	Oswald et al., 2013), an important direction for future studies.
16	The greatest potential of a neural measure of attitude comes from the fact that we
17	could successfully predict participants' choice behavior (i.e., revealed attitude score)
18	from their brain activations alone without asking them to engage in any attitude-
19	related task during scanning, and without incorporating information about attitude
20	judgments into our analysis of the fMRI data. While implicit attitude measures
21	circumvent some of the problems with self-reports, they cannot be "process-pure", so
22	that a score on an implicit attitude measure generally reflects some factors other than
23	implicit attitude toward a stimulus (Conrey et al., 2005; Sherman, 2009). In contrast,
24	in the present study, participants were never asked to report their attitudes (i.e., self-
25	report) or make any response based on attributes of attitude objects (i.e., implicit

1 measure) during the scanning – yet the activation patterns could robustly predict later 2 choices based on attitudes. This suggests that it may be possible to infer people's 3 spontaneous attitudes using incidental brain imaging methods. As neurally measured 4 incidental attitudes are unlikely to be influenced by any automatic or controlled 5 processes, there is no translational gap between the construct (i.e., attitude) and the 6 way it is measured (Sherman, 2009). Our findings suggest that it may be possible to 7 measure people's attitudes even when they are unable and/or unwilling to report them 8 truthfully (although ethical concerns for such an approach need to be carefully 9 considered). Thus, a neural measure of incidental attitudes has the potential to be 10 highly, if not completely, process-pure as well as relatively effort-free, offering some 11 distinct advantages over any implicit (and explicit) behavioral measures.

12 It should be noted that while the neural measure was outperformed by self-report 13 in predicting individual's choice behaviors, the choice prediction accuracy found in 14 the present study (59.6%) is comparable to the three past neuroeconomics studies 15 (Tusche et al., 2010; Levy et al., 2011; Smith et al., 2014), which attempted to predict 16 individual's choices from neural responses while passively viewing stimuli (Levy et al., 2011; Smith et al., 2014) or actively engaging an demanding attention task 17 18 (Tusche et al., 2010). In contrast to the present study, all of the three studies focused 19 on within-participant predictions (i.e., predicting a participant's choice between two 20 items based on his/her brain activations in response to other items) and reported the 21 choice prediction accuracy ranging from 56% to 83%. Smith et al. (2014) also tested 22 across-group predictions, which is conceptually similar to our across-participant 23 predictions and reported the prediction accuracy of 61.2%. Taken together with the 24 findings of Smith et al. (2014), the present results indicate that neural activation 25 patterns associated with attitudes are sufficiently similar across different individuals,

1 suggesting a great potential for an objective measure of attitudes using a

2 neuroimaging method.

3 We also found that attitude extremity can be predicted by activation patterns in 4 more posterior parts of the striatum among other regions. This result indicates that 5 both highly liked and highly disliked members induce similar patterns of activations 6 in this region. While the striatum is widely implicated in reward processing (Delgado, 7 2007), it is also known to respond to saliency (e.g., Zink et al., 2004; Jensen et al., 8 2007). As stimuli with high attitude extremity scores in the present study are also 9 likely to be highly salient, our results are consistent with the role of the striatum in 10 processing saliency. In addition, the pattern of results we obtained (Fig. 2b) clearly 11 indicates a functional dissociation within the striatum; whereas the anterior striatum 12 represent attitudes, the posterior striatum represents saliency or attitude extremity. 13 Functional dissociation within the striatum found in the present study is further 14 supported by past studies which identified several sub-regions within the striatum 15 based on intrinsic functional connectivity (Choi et al., 2012) and patterns of 16 coactivation with other cortical areas (Pauli et al., 2016). Although the reports on how 17 the striatum is organized differ slightly across studies, it seems clear that the two 18 striatal regions found in the present study lie in different sub-regions within the 19 striatum. Furthermore, the anterior striatum's role in representing attitudes is 20 consistent with the findings from large-scale coactivation data that this sub-region is 21 the most strongly involved in representing stimulus value (Pauli et al., 2016) (note 22 that the anterior striatum cluster found in the present study seems to be the closest to 23 the cluster labeled "ventral striatum" in Pauli et al., 2016). In contrast, however, 24 posterior striatum identified in the present study seems to be the striatal sub-region 25 most strongly associated with executive function (Pauli et al., 2016), which is not

necessarily consistent with our finding that this region is involved in attitude
 extremity or salience.

3 Nonetheless, our findings may be explained by two types of dopamine neurons 4 systematically located in the midbrain and projections from the midbrain to the 5 striatum. Matsumoto and Hikosaka (2009) found two types of dopamine neurons in 6 the monkey midbrain; those that are excited by positive stimuli (juice) and inhibited 7 by negative stimuli (airpuff to the eyes), and those that are excited by both the 8 positive and negative stimuli (Matsumoto and Hikosaka, 2009). Consistently across 9 two monkeys, those neurons which respond to both positive and negative stimuli (i.e., 10 motivationally salient stimuli) are located in the dorsolateral substantia nigra pars 11 compacta which projects mainly to the dorsal striatum, whereas those neurons which 12 predominantly respond to positive stimuli are located in the ventromedial substantia 13 nigra pars compacta (also ventral tegmental area) which send projections mainly to 14 the ventral striatum (Haber and Knutson, 2010). Thus, striatal functional dissociation 15 found in the present study might reflect the activations of these two types of neurons 16 in the midbrain. Thus, although speculative, the across-participant MVPA results may 17 suggest that different populations of neurons encode reward (attitude) and saliency 18 (attitude extremity), each of which is localized in a different sub-region of the 19 striatum and receives projection from dopamine neurons in a different sub-region of 20 the midbrain.

21

22 CONCLUSION

The present study investigated the potential of a neuroimaging method to predict
people's incidental social attitudes toward others. We found that patterns of

1 activations in the anterior striatum can reliably predict an individual's attitudes toward 2 famous people, suggesting the feasibility of such an approach. Although the 3 prediction accuracy was higher for the self-report measure than the neural measure, 4 the present study represents an essential first step toward neural measures of social 5 attitudes and demonstrated that we could successfully predict attitudes without asking 6 participants to engage in any attitude-related task. While we focused on explicit 7 attitudes toward familiar people, it will be important in future research to see how 8 accurately similar neural measures might predict implicit attitudes such as prejudice 9 toward racial or social outgroups.

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