## **Supplementary Materials**

Health news sharing is reflected in distributed reward-related brain activity Doré, BP, Scholz, C, Baek, EC, & Falk, EB

## Fitting models that use ventral striatum and vmPFC as separate predictor

**variables.** Extending analyses reported in Scholz et al. (2018), we fit additional models that used ventral striatum and vmPFC as separate predictor variables of population level news sharing. The two variables (ventral striatum and vmPFC) were highly correlated at both the trial-to-trial level, r = .73, and the article-to-article level, r = .83. In separate bivariate multilevel models, we saw relationships with article sharing at the within-person level for both ventral striatum,  $\beta$  = .08, 95%CI[.06, .11] and vmPFC,  $\beta$  = .09, 95%CI[.06, .12], as well as at the article-to-article level for both ventral striatum,  $\beta$  = .38, 95%CI[.18, .59] and vmPFC,  $\beta$  = .37, 95%CI[.16, .57]. Thus, these coefficients are highly similar. Future work that tests the conditions under which different parts of the reward value system provide similar or distinct information will be informative.

Additional model R<sup>2</sup> metrics for models linking pattern expression and population level sharing. We conducted a series of additional model comparisons to better understand the relative value of the different kinds of predictor variables (self reports, brain regions of interest, brain patterns of interest). First, we estimated R<sup>2</sup> for the valuerelated ROI alone for both within-person models, R<sup>2</sup> = .009 and the between-article models, R<sup>2</sup> = .13. Second, we estimated R<sup>2</sup> for the value-related pattern alone for both within-person models, R<sup>2</sup> = .01, and the between-article models, R<sup>2</sup> = .16. Third, we estimated  $R^2$  for the value-related pattern together with ratings for both within-person models,  $R^2 = .02$ , and the between-article models,  $R^2 = .23$ .

Comparison of bivariate models with only value-related ROI versus only valuerelated pattern. In a follow-up analysis, we fit and compared two models with only a single predictor variable – the value-related ROI, and the value-related pattern. In these bivariate models, the value-related ROI was related to population sharing,  $\beta = .12$ , 95%CI[.05, .19], R<sup>2</sup>=.01. The value-related pattern was also related to population sharing,  $\beta = .14$ , 95%CI[.05, .22], R<sup>2</sup>=.02. A comparison of these models indicated that

the model using the reward-related pattern was higher in expected out-of-sample

predictive accuracy,  $\triangle LOOIC = -1.9$ , SE = 4.5.

Fitting models with comparison patterns indexing memory- and vision-related brain processes. In another follow-up analysis, we fit models that used expression of brain patterns related to processes other than reward in order to ask if these also showed relationships with population article sharing. Specifically, we used meta-analytic maps from Neurosynth for the terms 'memory' and 'vision'. Results from these models indicated that neither expression of the memory-related pattern,  $\beta = -.04$ , 95%CI[-.11, .04], R<sup>2</sup>=.002, or the vision-related pattern,  $\beta = .00$ , 95%CI[-.04, .05], R<sup>2</sup>=.00, was clearly related to population-level article sharing.

**Comparing self-reports of reading intentions and self-reports of sharing intentions.** In our analyses of study 1, we use participants' self-reports of reading intentions as a self-report rating variable, and in study 2 we use participants' self-reports of sharing intentions. In these data, we see that both kinds of self-reports show similar relationships with population-level article sharing. For example, at the article-to-article level, we see similar magnitude relationships with population article sharing for both reading intentions in study 1, r = .39, and sharing intentions in study 2, r = .34. Further, since we used the same articles in study 1 and study 2, it is possible to also show that at the article-to-article level, these two self-reports are correlated -- articles that tend to receive high reading intentions also tend to receive high sharing intentions, r = .42. Therefore, despite the fact that these two types of intentions likely reflect some shared and some distinct psychological characteristics, in an effort to make use of all of our data, in models that combine study 1 and study 2, we use both as indices of selfreported article value. **Table S1.** One predictor multilevel model in study 1. Using pattern of interest (POI) as a predictor of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(POI) + (scale(POI) | subj) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000 Group-Level Effects: ~subj (Number of levels: 39) Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sd(Intercept) 0.07 0.05 0.00 0.18 1.00 794 1129 sd(scalePOI) 0.10 0.06 0.00 0.24 1.01 589 704 cor(Intercept,scalePOI) 0.04 0.55 -0.94 0.94 1.01 587 967 Population-Level Effects: Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS Intercept 0.01 0.04 -0.08 0.09 1.00 2131 1275 scalePOI 0.10 0.04 0.03 0.18 1.00 1858 1405

Family Specific Parameters:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.99 0.03 0.94 1.04 1.00 3416 1592

*Notes.* Estimate = mean of the posterior distribution; Est. Error = standard deviation of the posterior distribution; I-95%CI and u-95%CI = lower and upper bound of the 95% credibility interval; Rhat = Gelman-Rubin convergence diagnostic; Bulk\_ESS = effective sample size obtained by the MCMC algorithm; Tail ESS = minimum of the effective MCMC sample size at the 5% and 95% quantiles;

## **Table S2.** One predictor multilevel model in study 2. Using pattern of interest (POI) as a predictor of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(POI) + (scale(POI) | subj) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000 Group-Level Effects: ~subj (Number of levels: 38) Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sd(Intercept) 0.06 0.04 0.00 0.15 1.00 570 1104 sd(scalePOI) 0.11 0.05 0.02 0.20 1.01 524 699 cor(Intercept,scalePOI) 0.22 0.52 -0.87 0.96 1.01 321 409 **Population-Level Effects:** Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS

 Intercept
 0.00
 0.03
 -0.06
 0.06
 1.00
 1936
 1534

 scalePOI
 0.15
 0.04
 0.06
 0.24
 1.00
 1593
 1128

Family Specific Parameters:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.99 0.02 0.95 1.02 1.00 2554 1182

**Table S3.** One predictor article-level model in study 1. Using pattern of interest (POI) as a predictor of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(POI) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000

Population-Level Effects:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS Intercept -0.00 0.10 -0.21 0.19 1.00 1922 1538 scalePOI 0.33 0.10 0.12 0.54 1.00 1753 1229

Family Specific Parameters:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.93 0.08 0.79 1.10 1.00 1838 1447

**Table S4.** One predictor article-level model in study 2. Using pattern of interest (POI) as a predictor of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(POI) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000

Population-Level Effects:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS Intercept 0.00 0.11 -0.21 0.21 1.00 1929 1398 scalePOI 0.41 0.10 0.21 0.62 1.00 1867 1205

Family Specific Parameters:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.96 0.08 0.82 1.13 1.00 1813 1470

**Table S5.** Three predictor multilevel model in study 1. Using pattern of interest (POI), region of interest (ROI) and rating as predictors of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(rating) + scale(ROI) + scale(POI) + (scale(rating) + scale(ROI) + scale(POI) | subj) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000 Group-Level Effects: ~subj (Number of levels: 39) Estimate Est.Error I-95% CI u-95% CI Rhat Bulk ESS Tail ESS sd(Intercept) 0.06 0.04 0.00 0.14 1.00 805 1031 sd(scalerating) 0.03 0.02 0.00 0.08 1.00 1119 976 1047 780 sd(scaleROI) 0.03 0.02 0.00 0.09 1.00 sd(scalePOI) 0.10 0.05 0.01 0.20 1.00 638 781 cor(Intercept,scalerating) -0.07 0.44 -0.83 0.79 1.00 2405 1319 cor(Intercept,scaleROI) 0.04 0.46 -0.81 0.84 1.00 1838 1394 cor(scalerating,scaleROI) 1474 0.00 0.45 -0.79 0.82 1.00 1519 cor(Intercept,scalePOI) 0.13 0.42 -0.70 0.85 1.00 759 1252 cor(scalerating,scalePOI) 0.45 -0.11 -0.87 0.80 1.01 711 1136 cor(scaleROI,scalePOI) 0.02 0.44 -0.78 0.82 1.00 1122 1644 Population-Level Effects: Estimate Est. Error I-95% CI u-95% CI Rhat Bulk ESS Tail ESS 0.00 0.03 -0.05 0.06 1.00 2684 1549 Intercept scalerating 0.18 0.03 0.13 0.23 1.00 3011 1543 scaleROI 0.07 0.03 0.02 0.13 1.00 2529 1503 scalePOI 0.08 0.03 0.01 0.14 1.00 1787 1272 Family Specific Parameters:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.97 0.02 0.94 1.01 1.01 2865 1552

**Table S6.** Three predictor multilevel model in study 2. Using pattern of interest (POI), region of interest (ROI) and rating as predictors of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(rating) + scale(ROI) + scale(POI) + (scale(rating) + scale(ROI) + scale(POI) | subj) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000 Group-Level Effects: ~subj (Number of levels: 38) Estimate Est. Error I-95% CI u-95% CI Rhat Bulk ESS Tail ESS sd(Intercept) 805 0.09 0.06 0.01 0.21 1.01 852 sd(scalerating) 0.07 0.05 0.00 0.18 1.01 744 946 sd(scaleROI) 0.06 0.05 0.00 0.16 1.00 1015 1107 684 sd(scalePOI) 0.09 0.06 0.00 0.23 1.00 1057 cor(Intercept,scalerating) -0.05 0.45 -0.83 2732 0.77 1.00 1415 cor(Intercept,scaleROI) 0.05 0.44 -0.76 0.82 1.00 2696 1675 cor(scalerating,scaleROI) 0.79 1.00 -0.07 0.46 -0.84 2032 1400 cor(Intercept,scalePOI) 0.10 0.45 -0.77 0.84 1.00 1657 1551 cor(scalerating,scalePOI) -0.00 0.44 -0.79 0.80 1.00 1724 1548 cor(scaleROI,scalePOI) 0.02 0.45 -0.79 0.82 1.00 1741 1894 Population-Level Effects: Estimate Est. Error I-95% CI u-95% CI Rhat Bulk ESS Tail ESS 0.00 0.04 -0.08 0.08 1.00 2581 1740 Intercept 1473 scalerating 0.19 0.04 0.11 0.27 1.00 3006 scaleROI 0.09 0.04 0.02 0.17 1.00 3503 1512 scalePOI 0.12 0.04 0.03 0.20 1.00 2977 1486 Family Specific Parameters: Estimate Est. Error I-95% CI u-95% CI Rhat Bulk ESS Tail ESS

sigma 0.96 0.03 0.92 1.02 1.00 2838 1226

**Table S7.** Three predictor article-level model in study 1. Using pattern of interest (POI), region of interest (ROI) and rating as predictors of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(rating) + scale(ROI) + scale(POI) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000

Population-Level Effects:

Esti	mate Est	Error I	-95% CI	u-95% CI Rł	nat Bulk_	_ESS Tai	I_ESS
Intercept	0.00	0.10	-0.19	0.19 1.00	2399	1430	
scalerating	0.30	0.10	0.10	0.51 1.00	2322	1538	
scaleROI	0.14	0.10	-0.06	0.34 1.00	2144	1375	
scalePOI	0.32	0.10	0.12	0.52 1.00	2237	1582	

Family Specific Parameters:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.86 0.07 0.72 1.02 1.00 2223 1308

**Table S8.** Three predictor article-level model in study 2. Using pattern of interest (POI), region of interest (ROI) and rating as predictors of population sharing.

Family: gaussian Links: mu = identity; sigma = identity Formula: scale(population\_sharing) ~ scale(rating) + scale(ROI) + scale(POI) Samples: 4 chains, each with iter = 1000; warmup = 500; thin = 1; total post-warmup samples = 2000

Population-Level Effects:

 Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS

 Intercept
 0.00
 0.10
 -0.21
 0.21 1.00
 2225
 1591

 scalerating
 0.10
 0.12
 -0.12
 0.32 1.00
 2088
 1542

 scaleROI
 0.22
 0.12
 -0.02
 0.45 1.00
 1608
 1557

 scalePOI
 0.26
 0.12
 0.03
 0.49 1.00
 1610
 1211

Family Specific Parameters:

Estimate Est.Error I-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.95 0.08 0.82 1.12 1.00 2040 1284

## **Table S9.** Cluster table of Neurosynth 'reward' association test map, thresholded at z > 2.56, k=50.

Cluster Table of map: "Neurosynth - Reward, Association Test"

x y z | k | max z | mean z | Talaraich label

13	6	-2	4782   28.1   5.1   RH Lentiform Nucleus (Lateral Globus Pallidus)
3	-26	30	72   5.3   3.6   RH Cingulate Gyrus (Brodmann area 23)
-50	13	26	457   -4.6   -3.1   LH Inferior Frontal Gyrus (Brodmann area 9)
-50	-46	9	273   -4.2   -3.0   LH Superior Temporal Gyrus (Brodmann area 22)
24	-67	46	150   -4.1   -3.0   RH Superior Parietal Lobule (Brodmann area 7)
48	-66	9	102   -4.1   -3.0   RH Middle Temporal Gyrus (Brodmann area 37)
-45	-63	-1	194   -4.1   -3.0   LH Middle Temporal Gyrus (Brodmann area 37)
-24	-59	53	214   -4.1   -3.0   LH Precuneus (Brodmann area 7)
61	-27	20	72   -4.0   -3.0   RH Postcentral Gyrus (Brodmann area 40)
-8	8	52	72   -3.8   -3.0   LH Medial Frontal Gyrus (Brodmann area 6)
62	17	14	52   -3.6   -3.0   RH Inferior Frontal Gyrus (Brodmann area 44)

**Table S9.** Surface rendering of Neurosynth 'reward' association test map, thresholded at z > 2.56, k=50.

