



# Reward networks in the brain as captured by connectivity measures

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An assortment of human behaviors is thought to be driven by rewards including reinforcement learning, novelty processing, learning, decision making, economic choice, incentive motivation, and addiction. In each case the ventral tegmental area/ventral striatum (nucleus accumbens) (VTA–VS) system has been implicated as a key structure by functional imaging studies, mostly on the basis of standard, univariate analyses. Here we propose that standard functional magnetic resonance imaging analysis needs to be complemented by methods that take into account the differential connectivity of the VTA–VS system in the different behavioral contexts in order to describe reward based processes more appropriately. We first consider the wider network for reward processing as it emerged from animal experimentation. Subsequently, an example for a method to assess functional connectivity is given. Finally, we illustrate the usefulness of such analyses by examples regarding reward valuation, reward expectation and the role of reward in addiction.

**Keywords:** reward, connectivity, learning, addiction, functional magnetic resonance imaging

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## INTRODUCTION

The investigation of the behavioral and neural consequences of rewarding (reinforcing) and punishing events has a long standing history, dating back to the early investigations of operant conditioning. Recent neurobiological research has suggested that important aspects of reward processing are coded by dopaminergic neurons arising from the ventral tegmental area (VTA) and projecting to the ventral striatum (VS) via the mesolimbic pathway. Interestingly, the VTA–VS dopamine system has been found to be of eminent importance in a variety of motivated behaviors and cognition. For example, it has been implicated in reinforcement learning (Schultz, 1998), action monitoring (Holroyd and Coles, 2002; Kramer et al., 2007), novelty processing and learning (Schott et al., 2004b),

decision making and economic choice (McClure et al., 2004), incentive motivation (Berridge and Kringelbach, 2008), and addiction (Heinz et al., 2004; Reuter et al., 2005). In turn, all of these processes have been linked to reward processing. This is surprising, because the VTA is a comparatively small assembly of cells (with about 400000 cells in the adult human). The question thus arises as to how the VTA–VS system can modulate the wide spectrum of behaviors mentioned above. One possibility which we will elaborate on in this article is that the VTA–VS system is part of a wider network of brain structures. Depending on the specific context, activity in the VTA–VS system may interact with different other subcortical and cortical brain areas which could be the basis of the flexibility of this system.

## Rewards

Rewards are events and objects that modulate behavior: a behavior leading to a reward is more likely to occur again, whereas negative consequences lead to the opposite effects. Rewards are therefore potent learning signals.

The identification of a particular brain region with a specific cognitive function has been a central topic in neuroscience. Indeed, the major goal of functional magnetic resonance imaging (fMRI) analyses is to capture the blood oxygenation level-dependent (BOLD) signal associated with a particular task-related neural activity. In many fMRI experiments, multiple areas are found to be coactivated during a given task. However, standard univariate MRI analysis is not able to capture the task-related dynamics within a network of brain areas. A more complete understanding of the brain processes associated to a specific process requires both regionally specific activations and regionally specific interactions.

In the following, we will first review the key structures that are involved in reward-related behavior, including its relation to the learning and motivational circuits. We will then give an example for a method that can be used to assess interregional connectivity in conjunction with fMRI. The reward-related networks as evident for reward valuation, reward anticipation and addiction will be briefly discussed.

## KEY STRUCTURES MEDIATING REWARD-RELATED BEHAVIOR

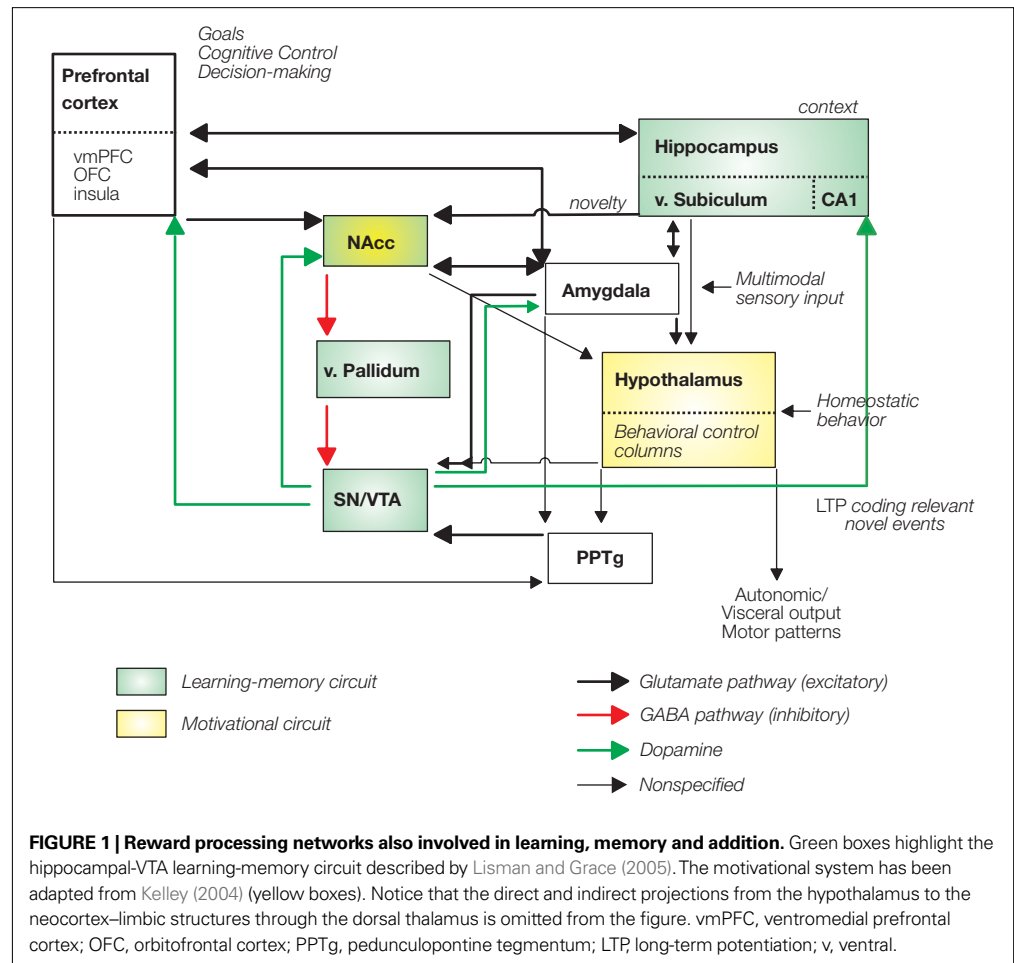
The desire to maximize rewards and to minimize negative possible outcomes is an important drive of human behavior. Because of this, humans are motivated to identify and seek possible cues in the environment which might predict the possible appearance of rewards or negative outcomes, as well as behaviors which could cause the appearance of these outcomes. The association of an event with a reward or a punishment therefore constitutes a powerful learning signal. In addition, we use information from the feedback signals elicited by our actions to influence our future decisions. In ambiguous contexts and situations in which different outcomes are probable or when feedback information is not available, humans might need to make decisions which can be considered risky, erratic or impulsive, sometimes irrationally pursuing short-term pleasures without considering that these actions could lead to negative after-effects in the future. Recent work in experimental economics (Glimcher et al., 2005) and decision making science (Schall, 2005) suggests that there are large interindividual differences with regard to the way we deal with rewards and punishments of different magnitude in certain situations. Interestingly, the cognitive processes required for successful adaptation in these situations might require the elicitation of affective responses (emotional valuation), the ability to associate neutral events to the

appearance of an emotional-charged outcome (learning) and the ability to store this information in order to make predictions (memory). This intersection between affective processes, learning and memory is a core aspect of reward processing, motivated behavior and decision making in humans.

A primary challenge in affective neuroscience is to understand to which degree these processes are subserved by specific brain regions or by common, partially overlapping networks. Indeed the ultimate aim would be to describe the specific role of each brain region and how the specific information computed in each brain region is assembled by larger brain midbrain-limbic-(sub)cortical networks in a process-specific way. A main problem encountered by the standard functional imaging approach to reward processing is that a large number of activations are usually seen. Reward processing thus consistently increases the BOLD response in a common set of regions comprising the VS (the nucleus accumbens, NAcc), the amygdala, the prefrontal cortex (orbitofrontal cortex, OFC), and the insula (Delgado et al., 2000, 2003; Breiter et al., 2001; Knutson et al., 2001, 2003b; McClure et al., 2004; Yacubian et al., 2006; Tom et al., 2007). Several studies have also identified activations in the midbrain regions (see for a review Duzel et al., 2009) as well as the ventromedial prefrontal cortex (vmPFC) or anterior cingulate cortex (ACC), although less consistently (Knutson et al., 2003b; Sanfey et al., 2003; Ullsperger and von Cramon, 2003).

The NAcc and VTA are placed prominently within a network that is not only implicated in the immediate processing of rewards but also in learning and motivation (Figure 1, see also Figure 1 in Münte et al., 2008).

The learning hippocampal-VTA (HP-VTA) loop (Figure 1, green boxes) has been adapted from Lisman and Grace (2005) who have proposed that hippocampal novelty signals might be conveyed to the midbrain (SN/VTA) through the NAcc and the ventral pallidum. The role of the ventral pallidum as an essential region involved in liking sensations has also been highlighted (Berridge and Kringelbach, 2008). This loop is important for encoding predictions based on stimulus-novelty. Novelty detected by the hippocampus might be sent through the subiculum, NAcc and ventral pallidum to the dopaminergic midbrain regions. Phasic activity in these midbrain neurons in primates have been observed to change according to the delivery of and expectation for salient and rewarding events (Schultz, 1998). Specifically, increases of DA cell firing have been associated to positive outcomes,



171 whereas choices that did not lead to a reward  
 172 evoked dips in the firing rate below baseline  
 173 (Schultz, 2002). This phasic firing might result in  
 174 release of dopamine in the hippocampus where  
 175 it might enhance long-term potentiation, and as  
 176 a consequence, memory storage and learning.  
 177 Notice also, that midbrain dopaminergic system  
 178 projects to and thus modulates other striatal-  
 179 orbitofrontal and prefrontal regions involved  
 180 in reward processing through the mesocortical  
 181 and mesolimbic pathways (Apicella et al., 1991;  
 182 Hikosaka and Watanabe, 2000; Wise, 2002). The  
 183 mesocortical pathway projects primarily to the  
 184 vmPFC, ACC and entorhinal cortex. The meso-  
 185 limbic pathway directly innervates the NAcc,  
 186 septum, olfactory tubercle, amygdala and piri-  
 187 form cortex. As a confirmation of the importance  
 188 of this HP-VTA loop in learning and memory,  
 189 the activation of the substantia nigra/VTA and the  
 190 hippocampus has been recently associated with  
 191 novelty processing and facilitation of memory  
 192 formation (Schott et al., 2004a, 2006; Wittmann  
 193 et al., 2008). Similarly, in a reward-motivated  
 194 memory formation task (Adcock et al., 2006),

195 high-reward cues preceding remembered but not  
 196 forgotten scenes activated VTA, the NAcc and the  
 197 hippocampus.

198 The second "motivational" circuit (Kelley  
 199 et al., 2005) allows the organism to seek specific  
 200 stimuli needed for survival by producing spon-  
 201 taneous locomotor behavior and exploration,  
 202 ingestive, defensive and reproductive behaviors.  
 203 These systems have been recently integrated in  
 204 what is termed the "behavioral control columns"  
 205 (Swanson, 2000), which are defined as a set of  
 206 highly interconnected nuclei in the hypothalamus  
 207 and its brainstem extensions devoted to the  
 208 elicitation and control of specific behaviors nec-  
 209 essary for survival (see Figure 1, yellow boxes).  
 210 These motivational systems might be activated  
 211 by specific environmental (internal or external)  
 212 stimuli and are amplified and energized by affect  
 213 or emotion. During evolution, these hard-wired  
 214 hypothalamic-brainstem circuits have been pro-  
 215 gressively interconnected with phylogenetically  
 216 more recent structures such as the PFC, striatum  
 217 and limbic regions, allowing the implementation  
 218 of cognitive control and more flexible motivated

### Functional Connectivity

Functional connectivity is defined as the statistical association or dependency among two or more neurophysiological time-series recorded in spatially remote areas. Correlational approaches are used and the direction of information flow can not be determined.

219 behavior. Massive direct and indirect afferents  
220 from the hippocampus, amygdala, VS and PFC  
221 project to the behavioral control columns, allow-  
222 ing the implementation of highly complex cogni-  
223 tive processes. For example, the amygdala, which is  
224 considered a key structure in emotional valuation,  
225 projects to the lateral hypothalamus and removal  
226 of this amygdalo-hypothalamic pathway does not  
227 abolish food intake *per se* but it alters the assess-  
228 ment of the comparative value of the food based  
229 on learning (Petrovich et al., 2005). Importantly  
230 these hypothalamic structures project to the mid-  
231 brain dopaminergic neurons which in the case of  
232 expectation and consumption of primary and sec-  
233 ondary rewards might elicit the activation of the  
234 NAcc and the PFC. Importantly, the hypothalamic  
235 behavioral control subsystems project massively  
236 back to the cerebral cortex via the dorsal thalamus  
237 (not shown in **Figure 1**). These feedforward pro-  
238 jections provide higher order cortical centers with  
239 access to internal motivational states.

240 Notice, that in both circuits the NAcc is a key  
241 integrative region weighting the different inputs  
242 coming from cortical areas (OFC, vmPFC, DLPFC,  
243 insula), limbic regions (amygdala, hippocampus;  
244 Groenewegen et al., 1999) and midbrain (SN/  
245 VTA) and therefore modulating the selection of  
246 appropriate responses and goal-directed behavior  
247 (Berridge and Robinson, 1998; Goto and Grace,  
248 2005; Kelley et al., 2005). Moreover, the interac-  
249 tions of the medial prefrontal cortex (ACC) and  
250 the VS (both receiving DA input from the mid-  
251 brain) in the adjustment of behavior have been  
252 highlighted (Holroyd and Coles, 2002).

### FUNCTIONAL CONNECTIVITY MEASURES

253 Obviously, **Figure 1** presents the basic network in  
254 which the VTA–VS reward system exerts its influ-  
255 ence on different behaviors. The key question is,  
256 how the different elements within this network  
257 work together in different behavioral contexts.  
258 This question might be answered by studying  
259 connectivity patterns and indeed cognitive neu-  
260 rosience has increasingly acknowledged the need  
261 for a network approach (Rykhlevskaia et al., 2008).  
262 Accordingly, a growing number of neuroimaging  
263 studies have shifted the focus from standard uni-  
264 variate to connectivity analyses. **Functional con-**  
265 **nectivity** is defined as the statistical association  
266 or dependency among two or more neurophysi-  
267 ological time-series recorded in spatially remote  
268 areas (Friston, 1994; Horwitz, 2003). Initial func-  
269 tional connectivity studies (PET and fMRI) used  
270 correlation analysis between a small number of  
271 pre-selected regions or between voxels of interest  
272 in order to study functional connectivity (Biswal  
273 et al., 1995). As fMRI uses an indirect measure of

275 neurophysiological functioning (BOLD signals),  
276 inter-regional dependencies can be investigated  
277 using correlation of BOLD signals between  
278 remote areas. Those regions showing large cor-  
279 relations are considered functionally connected.  
280 Correlation between two regions might exist even  
281 in the absence of a direct connection, therefore  
282 mediated by a third region. Partial correlation  
283 measures could be used in this particular case  
284 for removing the contributions of pair-wise cor-  
285 relations that might arise due to global or third-  
286 party effects (Hampson et al., 2002; Sun et al.,  
287 2004; Salvador et al., 2005).

288 When two regions are active roughly at the  
289 same time, then the two BOLD time-series might  
290 be highly correlated. In this particular case, imme-  
291 diate instantaneous or zero-order correlation  
292 measures between the two time series will capture  
293 the relationship between these signals (Hampson  
294 et al., 2002). However, when one signal is delayed  
295 from the other but showing a similar fluctuation,  
296 a time-shifted or lagged cross-correlation analysis  
297 is needed in order to capture the possible linear  
298 but delayed relationship between these regions.  
299 Notice that because the characteristics of the  
300 BOLD response, the correlations are based on  
301 low-frequency fluctuations. In a functional con-  
302 nectivity correlational study Cordes et al. (2001)  
303 showed that over 90% of their connectivity were  
304 due to low-frequency (below 0.1 Hz) fluctua-  
305 tions in a block-design paradigm. One important  
306 caveat of simple correlation analyses is that this  
307 measure is highly sensitive to the shape of the  
308 hemodynamic response function, such as onset-  
309 delay, time-to-peak, and width, which are region-  
310 specific due to differences in vascular properties  
311 across regions (Buckner et al., 1996; Bandettini  
312 and Cox, 2000). Because of that, this method is  
313 mostly appropriate to block-design analysis in  
314 which the shape of the hemodynamic response  
315 function shows less variability.

316 In contrast, coherence-related measures are  
317 less prone to the shape of the hemodynamic  
318 response function, as they are equivalent to the  
319 cross-correlation-related approaches, but using  
320 information from the frequency-domain. Cross-  
321 coherence measures have been shown to be very  
322 useful in investigating functional connectivity  
323 across brain regions (Leopold et al., 2003; Sun  
324 et al., 2004; Salvador et al., 2005). Because the  
325 analysis is performed in the frequency domain,  
326 this measure is blind to the possible lags of one  
327 region when compared to another one. In this  
328 sense, if the frequency content of one series is  
329 similar to another one, then the spectral coher-  
330 ence will be large indicating strong connectivity  
331 between two regions.

### Effective Connectivity

Effective connectivity approaches seek to describe causal influences of one brain area over another. Thus, the direction of information flow is determined. Such approaches need specified models with only a limited number of nodes.

332 Because the restriction of cross-correlation  
333 connectivity analyses to block designs, a new  
334 methodological approaches has been introduced  
335 to characterize functionally interacting regions  
336 using event-related fMRI designs (Rissman et al.,  
337 2004). This approach is based on the parameter  
338 estimates obtained in the context of the general  
339 linear model. Within this approach, a series of  
340 parameter estimates is extracted from a seed region  
341 and correlated with voxels from the whole-brain.  
342 Using this method, it is possible to identify specific  
343 functionally related brain networks. Similar solu-  
344 tions have been proposed by other authors (e.g.,  
345 Siegle et al., 2007; Aizenstein et al., 2009).

346 In recent years, many improvements have been  
347 made also in the description and localization  
348 of functional patterns (for a review, see Rogers  
349 et al., 2007). Some concerned the reduction of  
350 the number of regions involved in the correla-  
351 tion analysis. As the number of regions that are of  
352 interest increases, the covariance matrix becomes  
353 increasingly larger and thereby computations  
354 become more complex and more difficult to  
355 interpret. Indeed, different statistical multivariate  
356 approaches have been used to simplify the  
357 model, such as multidimensional scaling, principal  
358 component analysis, independent component  
359 analysis, and principal least squares, among  
360 others. These methods are very attractive in the  
361 sense that they do not require any prior hypoth-  
362 esis about the connectivity links of interest.

363 When studying functional connectivity it is  
364 also worth to consider the possible presence of  
365 spontaneous correlations between different brain  
366 regions. For example, Biswal et al. (1995) showed  
367 consistent correlations between different parts of  
368 the brain (bilateral primary motor and supple-  
369 mentary motor regions) during resting states (i.e.,  
370 when a participant is not performing any particu-  
371 lar task; see similar results in Xiong et al., 1999;  
372 Cordes et al., 2000; Lowe et al., 2000). In a subse-  
373 quent study, Hampson et al. (2002) investigated  
374 the changes in functional connectivity induced by  
375 a task (listening to continuous speech) when com-  
376 pared to a resting condition. Interestingly, higher  
377 correlations were observed between Broca's and  
378 Wernicke's regions when participants were actively  
379 listening to speech. The description of a consistent  
380 "default-mode" network in the resting brain in  
381 the absence of any stimulus (Raichle and Snyder,  
382 2007) implies that such networks have to be taken  
383 into account when evaluating regional BOLD cor-  
384 relations during task conditions (Hampson et al.,  
385 2002). It is possible that these coherent spontane-  
386 ous oscillations might account for a fraction of  
387 the trial-to-trial variability in BOLD event-related  
388 responses (Fox et al., 2006).

389 Finally, it is important to remind oneself that  
390 the presence of a structural connection makes a  
391 functional connection biologically meaningful and  
392 more likely to occur. Therefore, analyses of anatom-  
393 ical connectivity open a useful tool for restricting  
394 the number of functional connections to be ana-  
395 lyzed. Until recently, most connectivity approaches  
396 did not take into account the details of anatomical  
397 connectivity. However, if two regions are not  
398 anatomically connected, a functional connection is  
399 biologically implausible. Moreover, it is reasonable  
400 to expect that the strength of anatomical connec-  
401 tions might modulate the corresponding functional  
402 connections. Traditionally, anatomical connectivity  
403 maps have been restricted to animal invasive his-  
404 tological experimentation (Beaulieu, 2002) but the  
405 advent of diffusion tensor imaging combined with  
406 the development of new analysis tools opens up new  
407 opportunities. DTI based tractography provides  
408 detailed information of the structural connections  
409 (Hagmann et al., 2007). To use functionally defined  
410 seed points for fiber tracking algorithms appears  
411 very promising to investigate the direct relation  
412 between brain function and structure (Staempfli  
413 et al., 2008). Indeed, two recent studies that com-  
414 bined structural (DTI) and functional connectivity  
415 measures have shown a high degree of similarity  
416 between both connectivity estimates (Skudlarski  
417 et al., 2008; Honey et al., 2009).

418 The functional connectivity measures dis-  
419 cussed thus far are uninformative about the cau-  
420 sality of directionality of the influence between  
421 the different brain regions, i.e., what it is known  
422 as "**effective connectivity**" (Friston, 1994).  
423 Causality is taken into account by another set of  
424 methods such as structural equation modeling,  
425 dynamic causal modeling and psychophysiological  
426 interaction analyses Friston et al. (1997). These  
427 approaches constrain the connectivity analysis  
428 to a limited number of regions, based on prior  
429 knowledge (model based) about anatomical con-  
430 nections or functional systems (for a review on  
431 these methods, see Horwitz et al., 2005).

432 In sum, connectivity measures may greatly  
433 enhance our ability to map brain activations to  
434 behavior. However, several important limitations  
435 have to be considered. First, the BOLD response is  
436 a rather indirect measure of the brain at work. In  
437 particular, the question arises as to which aspect  
438 of the neural activity of the reward area is reflected  
439 in the BOLD response. With regard to the BOLD  
440 signal in the NAcc, a thoughtful review Knutson  
441 and Gibbs (2007) have convincingly suggested that  
442 it is modulated by dopamine signals arising from  
443 the VTA. Dopamine is released in the NAcc and  
444 shows a tendency to diffuse over wide areas (Garris  
445 et al., 1994) stimulating presynaptic D2-type

446 autoreceptors D1/D2-type postsynaptic receptors. 503  
 447 Animal experiments have shown that the onset of 504  
 448 changes in the membrane potential of postsynap- 505  
 449 tic neurons is around 200 ms and lasts for about 506  
 450 1000 ms after a single neural impulse. Knutson 507  
 451 and Gibbs (2007) further argue that the average 508  
 452 firing rate of five impulses per second should 509  
 453 lead to changes in extracellular dopamine levels 510  
 454 on a second-to-second basis. That these changes 511  
 455 in extracellular dopamine influence the BOLD 512  
 456 signal has been further substantiated by animal 513  
 457 experiments which showed that NAcc extracellular 514  
 458 dopamine and the BOLD signal had a similar tem- 515  
 459 poral profile and that lesioning of dopaminergic 516  
 460 neurons also abolished the NAcc BOLD response 517  
 461 (Chen et al., 1997). With regard to humans, it 518  
 462 has been demonstrated by many studies that the 519  
 463 BOLD signal to rewards or reward cues peaks at 520  
 464 about 4–6 s (Knutson et al., 2003a; Camara et al., 521  
 465 2008; Riba et al., 2008). Knutson and Gibbs (2007) 522  
 466 suggest a time-line connecting dopamine release 523  
 467 and fMRI BOLD response as follows: (a) dopamine 524  
 468 is released and activates postsynaptic D1 and D2 525  
 469 receptors 0–2 s after firing; (b) this changes the 526  
 470 postsynaptic membrane potential (0–2 s after fir- 527  
 471 ing) which (c) requires energy and oxygen from 528  
 472 nearby capillaries, which (d) is followed by an 529  
 473 increase of the BOLD signal 4–6 s after firing. 530  
 474 Thus, we can be reasonably sure that the BOLD 531  
 475 response in the NAcc tracks changes in dopamine 532  
 476 level over time. Simultaneous recordings of elec- 533  
 477 trophysiological signals and the BOLD response 534  
 478 in animals have suggested that in cortical areas the 535  
 479 BOLD response is related to local field potentials 536  
 480 rather than multi-unit activity (Logothetis et al., 537  
 481 2001; Logothetis and Wandell, 2004). This suggests 538  
 482 that connectivity between the NAcc and cortical 539  
 483 areas may reflect the dopaminergically modulated 540  
 484 influence of the NAcc on these areas. 541

485 A second problem is the slowness of the BOLD 542  
 486 signal and the fact that data-points are obtained 543  
 487 approximately every 1.5–2 s. Therefore, it might 544  
 488 well be that fMRI-based connectivity measures 545  
 489 underestimate the degree of interregional exchange 546  
 490 in the brain. A promising complementary line of 547  
 491 research to elucidate the mechanisms that sustain 548  
 492 connectivity in the brain is the investigation of 549  
 493 neurophysiological oscillations in different brain 550  
 494 regions. Synchronous oscillatory activity in dis- 551  
 495 tant regions might be a mechanism that sustains 552  
 496 functional connectivity. For example, synchronous 553  
 497 oscillatory activity within subcortical and cortical 554  
 498 networks have been related to learning and decision 555  
 499 making (Paz et al., 2006; Pesaran et al., 2008). In a 556  
 500 recent study, Popescu et al. (2009) showed learn- 557  
 501 ing-related increases in gamma coherence between 558  
 502 the basolateral amygdala and the ventral putamen,

using local field potentials recorded in cats per- 503  
 forming an appetitive learning task. Analysis of 504  
 electrophysiological activity has also demonstrated 505  
 that communication between distant brain regions 506  
 may also be established by phase-locking in differ- 507  
 ent frequency bands. Note, that such synchroniza- 508  
 tion processes might not necessarily be associated 509  
 with an increase in metabolism and a change in the 510  
 BOLD signal, and in this sense functional connec- 511  
 tivity fMRI methods might be limited to investigate 512  
 these issues. Animal experimentation is therefore 513  
 needed to explore the limits of connectivity assess- 514  
 ment using fMRI. 515

Human electrophysiological investigations may 516  
 provide interesting insights on interregional com- 517  
 munication in particular with regard to reward. 518  
 For example, several studies have already shown 519  
 oscillatory activity in the theta, beta and gamma 520  
 bands in humans related to reward processing 521  
 using non-invasive measurements (Cohen et al., 522  
 2007; Marco-Pallares et al., 2008). There is also a 523  
 small but growing number of studies that have 524  
 used simultaneous recordings from intra cerebral 525  
 electrodes in the NAcc and surface electrodes 526  
 (Münte et al., 2007; Cohen et al., 2009a,b,c; Heinze 527  
 et al., 2009) which took advantage of possibility 528  
 to assess correlations between depth and surface 529  
 electrodes. Such investigations, while limited in 530  
 the areas that can be reached with intracerebral 531  
 electrodes by clinical considerations, may pro- 532  
 vide crucial timing information. For example, 533  
 in a recent intracranial study in awake humans, 534  
 Cohen et al. (2009c) showed increased theta activ- 535  
 ity in the NAcc in monetary loss feedbacks trials 536  
 but not in gain trials in a reversal learning task. In 537  
 these “loss” trials, participants had to adjust their 538  
 behavioral strategy in order to gain more money. 539  
 This study provided compelling evidence about 540  
 the role of the VS in behavioral adjustment as it is 541  
 clearly responsive to negative feedback that signals 542  
 the need of such a readjustment. 543

## EMPIRICAL DATA: THE BRAIN'S REWARD 544 SYSTEM IN DIFFERENT CONTEXTS 545

### A NETWORK SUPPORTING REWARD VALUATION 546

The valuation of monetary gains and losses 547  
 activates a similar fronto-subcortical-limbic 548  
 network, but to a different degree. Specifically, 549  
 large activations have been reported in the VS, the 550  
 cingulate cortex, the superior frontal cortex, the 551  
 inferior parietal lobule, the insular cortex, hip- 552  
 pocampal regions, the thalamus, and the cau- 553  
 date nuclei (Delgado et al., 2000, 2003; Gottfried 554  
 et al., 2003; May et al., 2004; Nieuwenhuis et al., 555  
 2005; Dreher, 2007; Marco-Pallares et al., 2007; 556  
 Tom et al., 2007; Camara et al., 2008). However, 557  
 it is still controversial to which degree the neural 558

### Motivational Network

This network of brain areas supports behaviors needed for survival. It produces spontaneous locomotor activity and explorative behavior in order to seek specific stimuli as well as ingestive, defensive and reproductive behaviors.

### Structural Connectivity

Structural connectivity describes anatomical connections between remote brain areas. Recent advances in neuroimaging (DTI-based tractography) allow to image white matter connections between areas.

mechanisms underlying reward and punishment valuation recruit different brain regions. For example, Frank et al. (2004) proposed a differential modulation of excitatory and inhibitory pathways in the VS by positive and negative outcomes. Similarly, Wrase et al. (2007) have reported differences in adjustment of motor responses after the delivery of rewards compared to punishments. In this sense, the examination of brain connectivity patterns might help to differentiate the networks engaged in the processing reward and losses.

We recently conducted a study to address this issue. Importantly, functional connectivity results showed a different mesolimbic network than previous classical univariate analyses (Camara et al., 2008). In this study a gambling task required participants to bet on one of two sums of possible money which could be gained or lost (win or loss feedback appeared after the participant's decision). Occasionally, unexpectedly large sums were won or lost, which were five times larger in magnitude than the standard wins and losses but occurred in only 10% of the trials. Functional connectivity analyses showed an extensive network of regions supporting similar responses to reward and punishment valuation including the insular cortex and OFC, the amygdala, the hippocampus and the SN/VTA mid-brain regions. Notice, that this network clearly engaged the HP-VTA learning circuit proposed in **Figure 1** (Lisman and Grace, 2005) (see also **Figure 2A,B**). These regions correlated with the activity observed in the VS seed region (NAcc), which was the region which was selected as a seed point in order to perform the functional connectivity analysis. For losses stronger correlations were found between the VS and the medial OFC, indicating a relatively stronger relationship between these structures in the valuation of punishments. Moreover, there was a tendency for a greater involvement of the amygdala in the network elicited by losses (see **Figure 2B**).

These results complement a previous connectivity study (Cohen et al., 2008) in which microstructural properties of white matter tracts were predictive of functional connectivity after reward delivery. Importantly, the projections connecting the amygdala with the hippocampus, the OFC, and the VS not only predicted connectivity derived from fMRI time series but also participants' behavior following both positive and negative feedback in a reversal learning task (Cohen et al., 2008). One important aspect is that these results highlight the involvement of the VS (NAcc) as a key region in the **motivational network** developed on the basis of animal research (Kelley et al., 2005) (**Figure 2**). The different patterns obtained using the classical

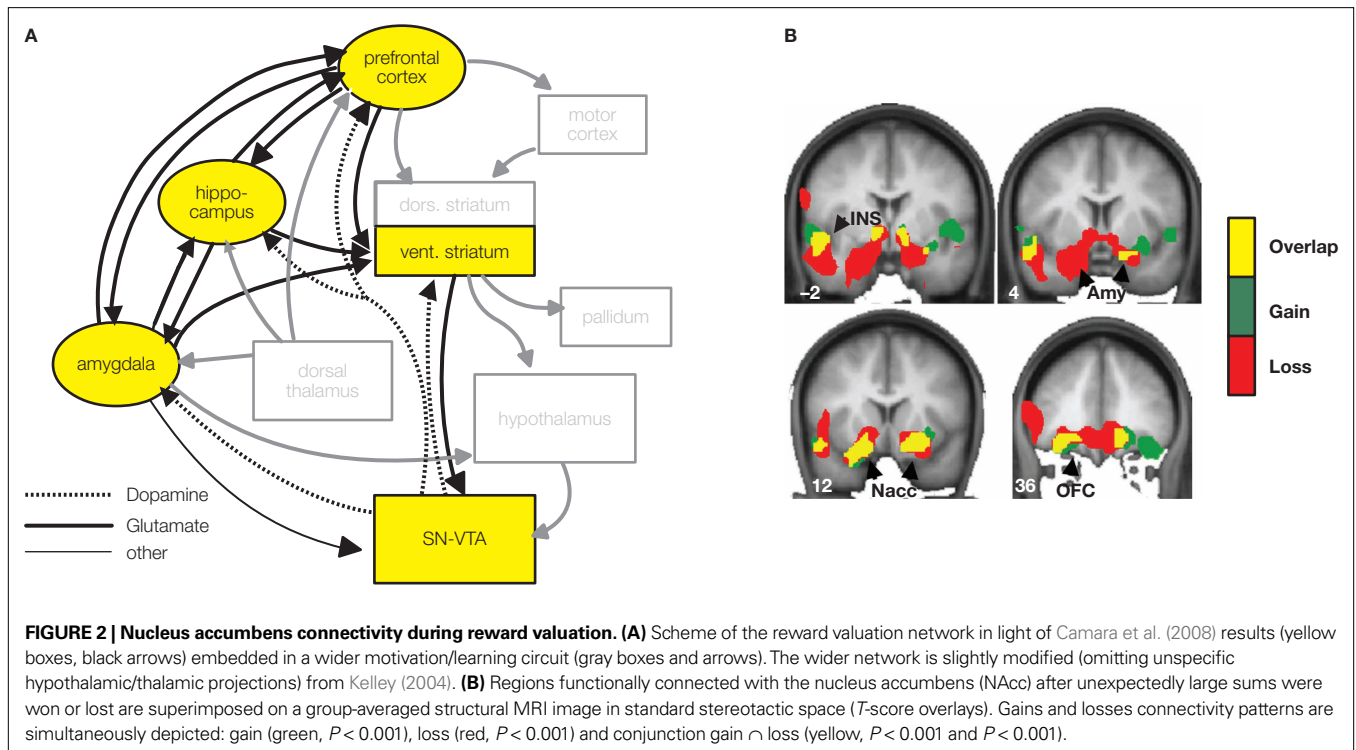
univariate analysis and connectivity analysis suggest that different information is retrieved using these two methods and stresses the importance of using functional connectivity as a complementary tool (Gazzaley et al., 2004; Rissman et al., 2004; Buchsbaum et al., 2005; Ranganath et al., 2005; Fiebach and Schubotz, 2006). An important aspect which was not analyzed in our previous study is in which degree other networks could have been identified if a different seed region would have been chosen. For example, the BOLD response in the vmPFC cortex did not correlate with the NAcc activation, which might suggest that a different network could be identified and related to a different functional role.

### A NETWORK SUPPORTING REWARD EXPECTATION

As well as the processing of reward outcomes, the expectation of primary (O'Doherty et al., 2002), monetary (Knutson et al., 2000, 2001), and social rewards (Izuma et al., 2008, 2009; Spreckelmeyer et al., 2009) is supported by similar fronto-subcortical-limbic networks, including the VS (NAcc) and the PFC, including the insular cortex (see Fehr and Camerer, 2007; Knutson and Greer, 2008 for reviews). Among these regions, the NAcc plays a primary role and is more activated for cues signaling potential rewards than cues signaling no reward. This anticipation effect has been linked to dopamine transmission in the NAcc (Knutson and Greer, 2008; Schott et al., 2008) and can be modulated by altering dopamine reuptake (Scheres et al., 2007; Strohle et al., 2008), changing dopamine breakdown (Yacubian et al., 2007), or using dopamine receptor agonists or antagonists (Abler et al., 2007). The PFC is assumed to control impulsive behaviors, being important for emotion regulation during decision making (McClure et al., 2004). It has been reported that individuals who tend to continue previously rewarded behaviors (rather than impulsive behaviors) show stronger **structural connectivity** between the striatum and the prefrontal cortex (Cohen et al., 2008). The insular cortex, on the other hand, is mainly associated with emotional processes and interacts with the VS during reward delivery (Camara et al., 2008).

In a recent fMRI study using functional connectivity (Ye et al., 2009) we further showed that these regions interact as a network during reward expectation. Moreover, this network can be distorted by dopamine receptor agonists such as pramipexole<sup>1</sup>, which is widely prescribed to treat

<sup>1</sup>Pramipexole is characterized mostly as a D2/D3 agonist. It also has some effects on other receptors, however (e.g., noradrenergic alpha-2 receptors, serotonergic 5HT1 receptors) (Millan et al., 2002).

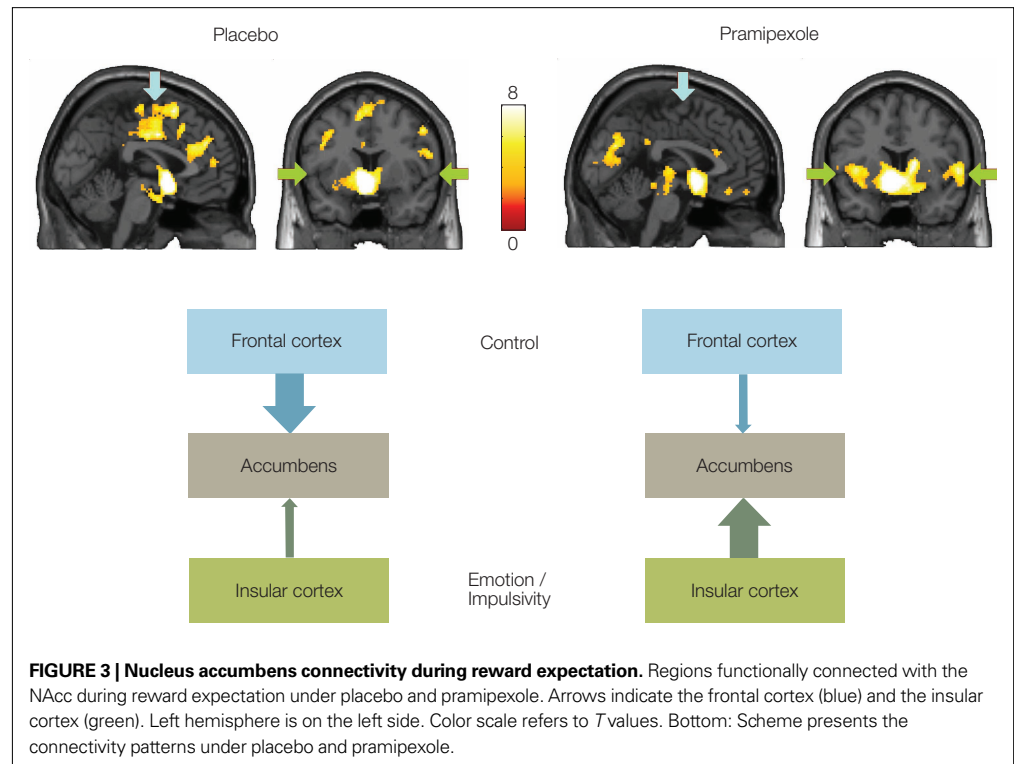


666 Parkinson's disease but has been reported to cause  
 667 pathological gambling as well as other impulse  
 668 control disorders (Dodd et al., 2005; Weintraub  
 669 et al., 2006). More specifically, intensive func-  
 670 tional connectivity was observed between the  
 671 NAcc and the PFC during the anticipation of  
 672 monetary rewards (see **Figure 3**, placebo con-  
 673 dition). This prefrontal–striatal connectivity,  
 674 however, is reduced by the administration of  
 675 pramipexole. Instead, the connection between  
 676 the insular cortex and the VS is enhanced (see  
 677 **Figure 3**, pramipexole condition). The weakened  
 678 connectivity between the VS and the prefrontal  
 679 cortex may lead to an impaired top-down execu-  
 680 tive control of impulsive behaviors, while the  
 681 enhanced connectivity between the VS and the  
 682 insular cortex may amplify the emotional influ-  
 683 ences on decision making (see **Figure 2**, schemes).  
 684 Indeed, the role of the vmPFC cortex in emotion  
 685 regulation is well established, projecting directly  
 686 to the amygdala and most probably providing  
 687 some inhibitory input (Quirk and Beer, 2006).  
 688 This shift in connectivity patterns may give rise  
 689 to an overestimation of potential rewards but to  
 690 an underestimation of possible risks. The imbal-  
 691 ance between the prefrontal–striatal circuitry  
 692 and the insula–striatal circuitry may explain  
 693 why pramipexole treated patients tend to develop  
 694 pathological gambling and other impulse control  
 695 disorders. The Ye et al. results are consistent with

696 predictions that follow from the tonic–phasic  
 697 dopamine hypothesis proposed by Grace and  
 698 colleagues (Grace, 1991; Bilder et al., 2004). This  
 699 hypothesis assumes that dopamine dynamics in  
 700 the striatum are driven by the interactions of  
 701 phasic and tonic dopamine release. Pramipexole  
 702 may reduce phasic dopamine release by activating  
 703 dopamine autoreceptors D2/D3 and at the same  
 704 time change tonic dopamine release by affect-  
 705 ing prefrontal–striatum glutamatergic projec-  
 706 tions. It has been reported that the stimulation  
 707 of cortical dopamine D2 receptors may directly  
 708 inhibit the activity of glutamate neurons in the  
 709 prefrontal cortex and subsequently the activity of  
 710 dopamine neurons in the NAcc, eventually lead-  
 711 ing to a decrease in extracellular dopamine level  
 712 (Beyer and Steketee, 2000; Del Arco and Mora,  
 713 2005). To compensate the change in dopamine  
 714 receptor stimulation, the amplitude of dopamine  
 715 efflux is increased. The effect of pramipexole on  
 716 phasic processes may be overridden by the effect  
 717 of pramipexole on tonic processes, resulting  
 718 in the increased NAcc activity during reward  
 719 anticipation. The increased NAcc activity may  
 720 reflect exaggerated incentive responses to possi-  
 721 ble rewards, and could be followed by impulsive  
 722 behaviors and suboptimal choices (Kuhnen and  
 723 Knutson, 2005).

724 It is interesting to note that an imbalanced  
 725 network of reward expectation may also account





726 for the tendency of adolescents to conduct  
 727 risky behaviors and to make suboptimal deci-  
 728 sions (Galvan et al., 2006; Casey et al., 2008; Van  
 729 Leijenhorst et al., 2009). Adolescents are endowed  
 730 with a functionally mature limbic system which is  
 731 sensitive to incoming rewards, but it is well docu-  
 732 mented that the prefrontal cortex continues to  
 733 develop into early adulthood. Consequently, as  
 734 compared to young adults, adolescents demon-  
 735 strated more activations in the VS and the insular  
 736 cortex (Van Leijenhorst et al., 2009), but less acti-  
 737 vation in the prefrontal cortex during the antici-  
 738 pation of monetary rewards (Galvan et al., 2006).  
 739 In other words, reward evaluation in adolescents  
 740 is biased by the limbic system rather than the pre-  
 741 frontal system. The engagement of two systems has  
 742 also been proposed to underlie decisions involv-  
 743 ing tradeoffs among benefits according to their  
 744 expected delays (intertemporal choice). Choices  
 745 of immediately available rewards are mediated  
 746 by the activity of limbic regions, while choices of  
 747 long-term rewards are supported by the activity  
 748 of prefrontal regions (McClure et al., 2004).

#### 749 REWARD AND ADDICTIVE BEHAVIOR

750 As pointed out above, a key question with regard  
 751 to the role of the core reward areas VTA and  
 752 VS in different behavioral contexts is how they  
 753 might interface with different parts of the wider  
 754 system shown in Figure 1 in these contexts. In

755 the case of addiction, besides the expectation  
 756 and delivery of the drug, we can also distinguish  
 757 craving states, which induce active drug seeking  
 758 and that can be elicited by drug-related cues.  
 759 The investigation of connectivity patterns in  
 760 addiction might be especially promising as it has  
 761 been proposed on the basis of animal studies that  
 762 there is a profound change in the way the ven-  
 763 tral and dorsal striatum interact. Whereas drug-  
 764 seeking behavior in the early phases of addiction  
 765 is a goal-directed behavior with the drug being  
 766 ingested because of its rewarding effects (simi-  
 767 lar to reward expectation described in Section  
 768 “A Network Supporting Reward Expectation”),  
 769 its behavior is maintained by drug-associated  
 770 cues in the sense of a stimulus-response habit  
 771 (Everitt et al., 2001; Redish, 2004; Everitt and  
 772 Robbins, 2005; Volkow et al., 2006). In the initial  
 773 phases drug seeking is thought to be control-  
 774 led by the VS. Subsequently, control is gradu-  
 775 ally shifted to the dorsal striatum. This shift may  
 776 be realized by serial “spiralling” connections  
 777 between the NAcc and the dorsal striatum via  
 778 midbrain dopamine neurons. In a recent lesion  
 779 experiment Belin and Everitt (2008) used an  
 780 intra-striatal disconnection procedure to disrupt  
 781 this striato-midbrain-striatal connectivity bilat-  
 782 erally and found a greatly decreased drug-seek-  
 783 ing behavior in rats addicted to cocaine. The shift  
 784 from ventral to dorsal striatum in drug seeking

785 behavior is consistent with evidence from MRI  
786 studies in addicted but drug-free humans in  
787 whom cue-elicited craving activates mainly  
788 the dorsal striatum as well as the amygdala and  
789 limbic prefrontal cortical areas (Grant et al.,  
790 1996; Childress et al., 1999; Garavan et al., 2000;  
791 Volkow et al., 2002) but not the VS.

792 These animal studies provide an interesting  
793 hypothesis regarding the connectivity of ventral  
794 and dorsal striatal regions in different stages of  
795 addiction. Thus, far there are only few studies  
796 that have used connectivity measures in relation to  
797 addiction. Recently, Filbey et al. (2008) used alco-  
798 holic tastes that were delivered to heavy drinking  
799 volunteers. A region of interest (ROI) approach  
800 was used and connectivity was studied by per-  
801 forming correlations between the different ROIs.  
802 Unfortunately, the authors defined a large ROI  
803 encompassing both, ventral and dorsal striatum,  
804 in addition to ROIs encompassing the SN/VTA, the  
805 mPFC and the OFC. Significant correlations were  
806 reported between these regions when comparing  
807 alcohol-related cues vs. rest. Obviously, to test the  
808 question of a shift in connectivity from early to late  
809 phases of addiction would require the investiga-  
810 tion of carefully selected participants and moreover  
811 the application of connectivity methods that are  
812 sensitive to the direction of information flow (see  
813 Section “Functional Connectivity Measures”).

814 There have been suggestions that obesity and  
815 the associated food intake behavior has strong  
816 parallels with drug addiction (Volkow and Wise,  
817 2005; Volkow et al., 2008). For example, damage  
818 to the VTA–VS dopamine system suppresses free  
819 feeding and the willingness to press a lever for  
820 food rewards in rats. The same procedure also  
821 attenuates the reward effects of drugs (Wise and  
822 Rompre, 1989). Against this background, Stoeckel  
823 et al. (2009) recently investigated effective con-  
824 nectivity within a “reward-network” in obese and  
825 normal weight women who were exposed to pic-  
826 tures depicting high and low calorie food. Based  
827 on prior hypotheses they selected the NAcc, the  
828 amygdala, and the OFC and performed structural  
829 equation modeling. Compared to the normal  
830 weight women, the obese group showed less amy-  
831 gdala-modulated activation of the OFC and the  
832 NAcc. On the other hand overweight participants  
833 showed an increased influence of the OFC on the  
834 activation of the NAcc. These findings suggest that  
835 obese women not only show an overall greater  
836 activation of the reward system to food stimuli  
837 (as demonstrated in Stoeckel et al., 2008, using  
838 univariate analyses) but also showed differences  
839 in the interregional interaction in the studied net-  
840 work. While the analysis provided information  
841 regarding the direction of information flow

842 between the brain areas, it has to be acknowledged  
843 that the model used in Stoeckel et al. (2009) was  
844 very simple due to the inherent limitations of  
845 the statistical analysis. For example, key areas  
846 involved in the motivational circuit outlined in  
847 **Figure 1** and involved in the homeostatic regula-  
848 tion of food intake (e.g., the hypothalamus) were  
849 not included in the model. Another recent study  
850 used psychophysiological interaction analysis to  
851 investigate the differences in connectivity between  
852 appetizing and bland food stimuli (pictures)  
853 (Passamonti et al., 2009). Moreover, in a second  
854 step, the authors investigated to what extent these  
855 interactions were modulated by the external food  
856 sensitivity of the participants, i.e., their prone-  
857 ness to react to appetizing food. High external  
858 food sensitivity was associated with reduced dif-  
859 ferential connectivity in the network comprising  
860 the NAcc, amygdala, mPFC, and premotor corti-  
861 cal areas. This network has been suggested to be  
862 involved when controlling feeding behavior in  
863 animals (Kelley et al., 2005).

## 864 CONCLUDING REMARKS

865 The human brain roughly features  $10^{11}$  neurons  
866 and  $10^{14}$  synapses. The very architecture of the  
867 brain readily suggests that neurons act in coordi-  
868 nated concert on a microscopic level, e.g., within  
869 nuclei and cortical columns, and on a macroscopic  
870 regions, i.e., between distant brain regions. This  
871 fact has been largely neglected during the first two  
872 decades of brain imaging mostly for the lack of  
873 appropriate techniques. In the present paper we  
874 illustrated the advantages of connectivity analyses  
875 for the investigation of reward processing. What  
876 emerges at this point is still a very sketchy pic-  
877 ture, however. What is needed is a more system-  
878 atic assessment of the connectivity of the VTA–VS  
879 reward system in different contexts using the same  
880 methods. As illustrated by several examples in this  
881 review, altered connectivity of the VTA–VS system  
882 with other brain area may underlie behavioral and  
883 brain imaging changes observed during develop-  
884 ment and in pathological conditions. Such analy-  
885 ses will therefore contribute to our understanding  
886 of the pathophysiology of such states.

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## REFERENCES

- 897  
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955  
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962  
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964
- Abler, B., Erk, S., Herwig, U., and Walter, H. (2007). Anticipation of aversive stimuli activates extended amygdala in unipolar depression. *J. Psychiatr. Res.* 41, 511–522.
- Adcock, R. A., Thangavel, A., Whitfield-Gabrieli, S., Knutson, B., and Gabrieli, J. D. (2006). Reward-motivated learning: mesolimbic activation precedes memory formation. *Neuron* 50, 507–517.
- Aizenstein, H. J., Butters, M. A., Wu, M., Mazurkewicz, L. M., Stenger, V. A., Gianaros, P. J., Becker, J. T., Reynolds, C. F. III, and Carter, C. S. (2009). Altered functioning of the executive control circuit in late-life depression: episodic and persistent phenomena. *Am. J. Geriatr. Psychiatry* 17, 30–42.
- Apicella, P., Ljungberg, T., Scarnati, E., and Schultz, W. (1991). Responses to reward in monkey dorsal and ventral striatum. *Exp. Brain Res.* 85, 491–500.
- Bandettini, P. A., and Cox, R. W. (2000). Event-related fMRI contrast when using constant interstimulus interval: theory and experiment. *Magn. Reson. Med.* 43, 540–548.
- Beaulieu, C. (2002). The basis of anisotropic water diffusion in the nervous system – a technical review. *NMR Biomed.* 15, 435–455.
- Belin, D., and Everitt, B. J. (2008). Cocaine seeking habits depend upon dopamine-dependent serial connectivity linking the ventral with the dorsal striatum. *Neuron* 57, 432–441.
- Berridge, K. C., and Kringelbach, M. L. (2008). Affective neuroscience of pleasure: reward in humans and animals. *Psychopharmacology (Berl.)* 199, 457–480.
- Berridge, K. C., and Robinson, T. E. (1998). What is the role of dopamine in reward: hedonic impact, reward learning, or incentive salience? *Brain Res. Brain Res. Rev.* 28, 309–369.
- Beyer, C. E., and Steketee, J. D. (2000). Intramedial prefrontal cortex injection of quinpirole, but not SKF 38393, blocks the acute motor-stimulant response to cocaine in the rat. *Psychopharmacology (Berl.)* 151, 211–218.
- Bilder, R. M., Volavka, J., Lachman, H. M., and Grace, A. A. (2004). The catechol-O-methyltransferase polymorphism: relations to the tonic-phasic dopamine hypothesis and neuropsychiatric phenotypes. *Neuropsychopharmacology* 29, 1943–1961.
- Biswal, B., Yetkin, F. Z., Haughton, V. M., and Hyde, J. S. (1995). Functional connectivity in the motor cortex of resting human brain using echo-planar MRI. *Magn Reson. Med.* 34, 537–541.
- Breiter, H. C., Aharon, I., Kahneman, D., Dale, A., and Shizgal, P. (2001). Functional imaging of neural responses to expectancy and experience of monetary gains and losses. *Neuron* 30, 619–639.
- Buchsbaum, B., Pickell, B., Love, T., Hatrak, M., Bellugi, U., and Hickok, G. (2005). Neural substrates for verbal working memory in deaf signers: fMRI study and lesion case report. *Brain Lang.* 95, 265–272.
- Buckner, R. L., Bandettini, P. A., O’Craven, K. M., Savoy, R. L., Petersen, S. E., Raichle, M. E., and Rosen, B. R. (1996). Detection of cortical activation during averaged single trials of a cognitive task using functional magnetic resonance imaging. *Proc. Natl. Acad. Sci. U.S.A.* 93, 14878–14883.
- Camara, E., Rodriguez-Fornells, A., and Münte, T. F. (2008). Functional connectivity of reward processing in the brain. *Front. Hum. Neurosci.* 2, 19. doi:10.3389/neuro.09.019.2008
- Casey, B. J., Getz, S., and Galvan, A. (2008). The adolescent brain. *Dev. Rev.* 28, 62–77.
- Chen, Y. C. I., Galpern, W. R., Brownell, A. L., Matthews, R. T., Bogdanov, M., Isacson, O., Keltner, J. R., Beal, M. F., Rosen, B. R., and Jenkins, B. G. (1997). Detection of dopaminergic neurotransmitter activity using pharmacologic MRI: Correlation with PET, microdialysis, and behavioral data. *Magn. Reson. Med.* 38, 389–398.
- Childress, A. R., Mozley, P. D., McElgin, W., Fitzgerald, J., Reivich, M., and O’Brien, C. P. (1999). Limbic activation during cue-induced cocaine craving. *Am. J. Psychiatry* 156, 11–18.
- Cohen, M. X., Axmacher, N., Lenartz, D., Elger, C. E., Sturm, V., and Schlaepfer, T. E. (2009a). Good vibrations: cross-frequency coupling in the human nucleus accumbens during reward processing. *J. Cogn. Neurosci.* 21, 875–889.
- Cohen, M. X., Axmacher, N., Lenartz, D., Elger, C. E., Sturm, V., and Schlaepfer, T. E. (2009b). Neuroelectric signatures of reward learning and decision-making in the human nucleus accumbens. *Neuropsychopharmacology* 34, 1649–1658.
- Cohen, M. X., Axmacher, N., Lenartz, D., Elger, C. E., Sturm, V., and Schlaepfer, T. E. (2009c). Nuclei accumbens phase synchrony predicts decision-making reversals following negative feedback. *J. Neurosci.* 29, 7591–7598.
- Cohen, M. X., Elger, C. E., and Ranganath, C. (2007). Reward expectation modulates feedback-related negativity and EEG spectra. *Neuroimage* 35, 968–978.
- Cohen, M. X., Elger, C. E., and Weber, B. (2008). Amygdala tractography predicts functional connectivity and learning during feedback-guided decision-making. *Neuroimage* 39, 1396–1407.
- Cordes, D., Haughton, V. M., Arfanakis, K., Carew, J. D., Turski, P. A., Moritz, C. H., Quigley, M. A., and Meyerand, M. E. (2001). Frequencies contributing to functional connectivity in the cerebral cortex in “resting-state” data. *AJNR Am. J. Neuroradiol.* 22, 1326–1333.
- Cordes, D., Haughton, V. M., Arfanakis, K., Wendt, G. J., Turski, P. A., Moritz, C. H., Quigley, M. A., and Meyerand, M. E. (2000). Mapping functionally related regions of brain with functional connectivity MR imaging. *AJNR Am. J. Neuroradiol.* 21, 1636–1644.
- Del Arco, A., and Mora, F. (2005). Glutamate-dopamine in vivo interaction in the prefrontal cortex modulates the release of dopamine and acetylcholine in the nucleus accumbens of the awake rat. *J. Neural. Transm.* 112, 97–109.
- Delgado, M. R., Locke, H. M., Stenger, V. A., and Fiez, J. A. (2003). Dorsal striatum responses to reward and punishment: effects of valence and magnitude manipulations. *Cogn. Affect. Behav. Neurosci.* 3, 27–38.
- Delgado, M. R., Nystrom, L. E., Fissell, C., Noll, D. C., and Fiez, J. A. (2000). Tracking the hemodynamic responses to reward and punishment in the striatum. *J. Neurophysiol.* 84, 3072–3077.
- Dodd, M. L., Klos, K. J., Bower, J. H., Geda, Y. E., Josephs, K. A., and Ahlskog, J. E. (2005). Pathological gambling caused by drugs used to treat Parkinson disease. *Arch. Neurol.* 62, 1377–1381.
- Dreher, J. C. (2007). Sensitivity of the brain to loss aversion during risky gambles. *Trends Cogn. Sci.* 11, 270–272.
- Duzel, E., Bunzeck, N., Guitart-Masip, M., Wittmann, B., Schott, B. H., and Tobler, P. N. (2009). Functional imaging of the human dopaminergic mid-brain. *Trends Neurosci.* 32, 321–328.
- Everitt, B. J., Dickinson, A., and Robbins, T. W. (2001). The neuropsychological basis of addictive behaviour. *Brain Res. Rev.* 36, 129–138.
- Everitt, B. J., and Robbins, T. W. (2005). Neural systems of reinforcement for drug addiction: from actions to habits to compulsion. *Nat. Neurosci.* 8, 1481–1489.
- Fehr, E., and Camerer, C. F. (2007). Social neuroeconomics: the neural circuitry of social preferences. *Trends Cogn. Sci.* 11, 419–427.
- Fiebach, C. J., and Schubotz, R. I. (2006). Dynamic anticipatory processing of hierarchical sequential events: a common role for Broca’s area and ventral premotor cortex across domains? *Cortex* 42, 499–502.
- Filbey, F. M., Claus, E., Audette, A. R., Niculescu, M., Banich, M. T., Tanabe, J., Du, Y. P., and Hutchison, K. E. (2008). Exposure to the taste of alcohol elicits activation of the mesocorticolimbic neurocircuitry. *Neuropsychopharmacology* 33, 1391–1401.
- Fox, M. D., Snyder, A. Z., Zacks, J. M., and Raichle, M. E. (2006). Coherent spontaneous activity accounts for trial-to-trial variability in human evoked brain responses. *Nat. Neurosci.* 9, 23–25.
- Frank, M. J., Seeberger, L. C., and O’Reilly, R. C. (2004). By carrot or by stick: cognitive reinforcement learning in parkinsonism. *Science* 306, 1940–1943.
- Friston, K. (1994). Functional and effective connectivity in neuroimaging: a synthesis. *Hum. Brain Mapp.* 2, 56–78.
- Friston, K. J., Buechel, C., Fink, G. R., Morris, J., Rolls, E., and Dolan, R. J. (1997). Psychophysiological and modulatory interactions in neuroimaging. *Neuroimage* 6, 218–229.
- Galvan, A., Hare, T. A., Parra, C. E., Penn, J., Voss, H., Glover, G., and Casey, B. J. (2006). Earlier development of the accumbens relative to orbitofrontal cortex might underlie risk-taking behavior in adolescents. *J. Neurosci.* 26, 6885–6892.
- Garavan, H., Pankiewicz, J., Bloom, A., Cho, J. K., Sperry, L., Ross, T. J., Salmeron, B. J., Risinger, R., Kelley, D., and Stein, E. A. (2000). Cue-induced cocaine craving: neuroanatomical specificity for drug users and drug stimuli. *Am. J. Psychiatry* 157, 1789–1798.
- Garris, P. A., Ciolkowski, E. L., Pastore, P., and Wightman, R. M. (1994). Efflux of dopamine from the synaptic cleft in the nucleus accumbens of the rat brain. *J. Neurosci.* 14, 6084–6093.
- Gazzaley, A., Rissman, J., and Desposito, M. (2004). Functional connectivity during working memory maintenance. *Cogn. Affect. Behav. Neurosci.* 4, 580–599.
- Glimcher, P. W., Dorris, M. C., and Bayer, H. M. (2005). Physiological utility theory and the neuroeconomics of choice. *Games Econ. Behav.* 52, 213–256.
- Goto, Y., and Grace, A. A. (2005). Dopaminergic modulation of limbic and cortical drive of nucleus accumbens in goal-directed behavior. *Nat. Neurosci.* 8, 805–812.
- Gottfried, J. A., O’Doherty, J., and Dolan, R. J. (2003). Encoding predictive reward value in human amygdala and orbitofrontal cortex. *Science* 301, 1104–1107.

- 965 Grace, A. A. (1991). Phasic versus tonic  
966 dopamine release and the modulation  
967 of dopamine system responsivity: a  
968 hypothesis for the etiology of schizo-  
969 phrenia. *Neuroscience* 41, 1–24.
- 970 Grant, S., London, E. D., Newlin, D. B.,  
971 Villemaigne, V. L., Liu, X.,  
972 Contoreggi, C., Phillips, R. L.,  
973 Kimes, A. S., and Margolin, A. (1996).  
974 Activation of memory circuits during  
975 cue-elicited cocaine craving. *Proc. Natl.  
976 Acad. Sci. U.S.A.* 93, 12040–12045. 
- 977 Groenewegen, H. J., Wright, C. I.,  
978 Beijer, A. V., and Voorn, P. (1999).  
979 Convergence and segregation of ven-  
980 tral striatal inputs and outputs. *Ann.  
981 N. Y. Acad. Sci.* 877, 49–63.
- 982 Hagmann, P., Kurant, M., Gigandet, X.,  
983 Thiran, P., Wedeen, V. J., Meuli, R., and  
984 Thiran, J. P. (2007). Mapping human  
985 whole-brain structural networks with  
986 diffusion MRI. *PLoS ONE* 2, e597.  
987 doi:10.1371/journal.pone.0000597
- 988 Hampson, M., Peterson, B. S.,  
989 Skudlarski, P., Gatenby, J. C., and  
990 Gore, J. C. (2002). Detection of func-  
991 tional connectivity using temporal  
992 correlations in MR images. *Hum.  
993 Brain Mapp.* 15, 247–262.
- 994 Heinz, A., Siessmeier, T., Wrase, J.,  
995 Hermann, D., Klein, S., Grusser, S. M.,  
996 Flor, H., Braus, D. F., Buchholz, H. G.,  
997 Grunder, G., Schreckenberger, M.,  
998 Smolka, M. N., Rosch, F., Mann, K.,  
999 and Bartenstein, P. (2004). Correlation  
1000 between dopamine D(2) receptors in  
1001 the ventral striatum and central  
1002 processing of alcohol cues and craving.  
1003 *Am. J. Psychiatry* 161, 1783–1789.
-  1004 Heinze, H. J., Heldmann, M., Voges, J.,  
1005 Hinrichs, H., Marco-Pallares, J.,  
1006 Hopf, J. M., Müller, U. J., Galazky, I.,  
1007 Sturm, V., Bogerts, B., and Münte, T. F.  
1008 (2009). Counteracting incentive sensi-  
1009 tization in severe alcohol dependence  
1010 using deep brain stimulation of the  
1011 nucleus accumbens: clinical and basic  
1012 science aspects. *Front. Hum. Neurosci.*  
1013 (in press).
- 1014 Hikosaka, K., and Watanabe, M. (2000).  
1015 Delay activity of orbital and lateral  
1016 prefrontal neurons of the monkey  
1017 varying with different rewards. *Cereb.  
1018 Cortex* 10, 263–271.
- 1019 Holroyd, C. B., and Coles, M. G. (2002).  
1020 The neural basis of human error  
1021 processing: reinforcement learning,  
1022 dopamine, and the error-related nega-  
1023 tivity. *Psychol. Rev.* 109, 679–709.
- 1024 Honey, C. J., Sporns, O., Cammoun, L.,  
1025 Gigandet, X., Thiran, J. P., Meuli, R.,  
1026 and Hagmann, P. (2009). Predicting  
1027 human resting-state functional con-  
1028 nectivity from structural connectiv-  
1029 ity. *Proc. Natl. Acad. Sci. U.S.A.* 106,  
1030 2035–2040.
- 1031 Horwitz, B. (2003). The elusive concept  
1032 of brain connectivity. *Neuroimage* 19,  
1033 466–470.
- Horwitz, B., Warner, B., Fitzer, J.,  
Tagamets, M. A., Husain, F. T., and  
Long, T. W. (2005). Investigating the  
neural basis for functional and effec-  
tive connectivity. Application to fMRI.  
*Philos. Trans. R. Soc. Lond., B, Biol. Sci.*  
360, 1093–1108.
- Izuma, K., Saito, D. N., and Sadato, N.  
(2008). Processing of social and mon-  
etary rewards in the human striatum.  
*Neuron* 58, 284–294.
-  Izuma, K., Saito, D. N., and Sadato, N.  
(2009). Processing of the incentive for  
social approval in the ventral striatum  
during charitable donation. *J. Cogn.  
Neurosci.*
- Kelley, A. E. (2004). Memory and addic-  
tion: shared neural circuitry and  
molecular mechanisms. *Neuron* 44,  
161–179.
- Kelley, A. E., Baldo, B. A., Pratt, W. E., and  
Will, M. J. (2005). Corticostriatal-  
hypothalamic circuitry and food motiva-  
tion: integration of energy, action and  
reward. *Physiol. Behav.* 86, 773–795.
- Knutson, B., Adams, C. M., Fong, G. W.,  
and Hommer, D. (2001). Anticipation  
of increasing monetary reward selec-  
tively recruits nucleus accumbens.  
*J. Neurosci.* 21, RC159.
- Knutson, B., Fong, G. W., Bennett, S. M.,  
Adams, C. M., and Hommer, D.  
(2003a). A region of mesial prefrontal  
cortex tracks monetarily reward-  
ing outcomes: characterization with  
rapid event-related fMRI. *Neuroimage*  
18, 263–272.
- Knutson, B., Fong, G. W., Bennett, S. M.,  
Adams, C. M., and Hommer, D.  
(2003b). A region of mesial prefrontal  
cortex tracks monetarily reward-  
ing outcomes: characterization with  
rapid event-related fMRI. *Neuroimage*  
18, 263–272.
- Knutson, B., and Gibbs, S. E. (2007).  
Linking nucleus accumbens  
dopamine and blood oxygenation.  
*Psychopharmacology (Berl.)* 191,  
813–822.
- Knutson, B., and Greer, S. M. (2008).  
Anticipatory affect: neural correlates  
and consequences for choice. *Philos.  
Trans. R. Soc. Lond., B, Biol. Sci.* 363,  
3771–3786.
- Knutson, B., Westdorp, A., Kaiser, E., and  
Hommer, D. (2000). fMRI visualiza-  
tion of brain activity during a mon-  
etary incentive delay task. *Neuroimage*  
12, 20–27.
- Kramer, U. M., Cunillera, T., Camara, E.,  
Marco-Pallares, J., Cucurell, D.,  
Nager, W., Bauer, P., Schule, R.,  
Schols, L., Rodriguez-Fornells, A.,  
and Münte, T. F. (2007). The impact  
of catechol-O-methyltransferase and  
dopamine D4 receptor genotypes on  
neurophysiological markers of  
performance monitoring. *J. Neurosci.*  
27, 14190–14198.
- Kuhnen, C. M., and Knutson, B. (2005).  
The neural basis of financial risk tak-  
ing. *Neuron* 47, 763–770.
- Leopold, D. A., Murayama, Y., and  
Logothetis, N. K. (2003). Very slow  
activity fluctuations in monkey  
visual cortex: implications for func-  
tional brain imaging. *Cereb. Cortex*  
13, 422–433.
- Lisman, J. E., and Grace, A. A. (2005). The  
hippocampal-VTA loop: controlling  
the entry of information into long-  
term memory. *Neuron* 46, 703–713.
- Logothetis, N. K., Pauls, J., Augath, M.,  
Trinath, T., and Oeltermann, A. (2001).  
Neurophysiological investigation of  
the basis of the fMRI signal. *Nature*  
412, 150–157.
- Logothetis, N. K., and Wandell, B. A.  
(2004). Interpreting the BOLD signal.  
*Annu. Rev. Physiol.* 66, 735–769.
- Lowe, M. J., Dzemidzic, M., Lurito, J. T.,  
Mathews, V. P., and Phillips, M. D.  
(2000). Correlations in low-frequency  
BOLD fluctuations reflect cortico-  
cortical connections. *Neuroimage* 12,  
582–587.
- Marco-Pallares, J., Cucurell, D.,  
Cunillera, T., Garcia, R., ndres-  
Pueyo, A., Münte, T. F., and Rodriguez-  
Fornells, A. (2008). Human oscillatory  
activity associated to reward processing  
in a gambling task. *Neuropsychologia*  
46, 241–248.
- Marco-Pallares, J., Muller, S. V., and  
Münte, T. F. (2007). Learning by  
doing: an fMRI study of feedback-  
related brain activations. *Neuroreport*  
18, 1423–1426.
- May, J. C., Delgado, M. R., Dahl, R. E.,  
Stenger, V. A., Ryan, N. D., Fiez, J. A.,  
and Carter, C. S. (2004). Event-related  
 functional magnetic resonance imag-  
ing of reward-related brain circuitry  
in children and adolescents. *Biol.  
Psychiatry* 55, 359–366.
- McClure, S. M., York, M. K., and  
Montague, P. R. (2004). The neural  
substrates of reward processing in  
humans: the modern role of fMRI.  
*Neuroscientist* 10, 260–268.
- Millan, M. J., Maiorini, L., Cussac, D.,  
Audinot, V., Boutin, J. A., and  
Newman-Tancredi, A. (2002).  
Differential actions of antiparkinson  
agents at multiple classes of monoam-  
inergic receptor. I. A multivariate  
analysis of the binding profiles of 14  
drugs at 21 native and cloned human  
receptor subtypes. *J. Pharmacol. Exp.  
Ther.* 303, 791–804.
- Münte, T. F., Heldmann, M., Hinrichs, H.,  
Marco-Pallares, J., Kramer, U. M.,  
Sturm, V., and Heinze, H. J. (2007).  
Nucleus accumbens is involved in  
human action monitoring: evidence  
from invasive electrophysiological  
recordings. *Front. Hum. Neurosci.* 1,  
11. doi:10.3389/neuro.09.011.2007
- Münte, T. F., Heldmann, M., Hinrichs, H.,  
Marco-Pallares, J., Kramer, U. M.,  
Sturm, V., and Heinze, H. J. (2008).  
Contribution of subcortical structures to  
cognition assessed with invasive electro-  
physiology in humans. *Front. Neurosci.* 2,  
72–78. doi:10.3389/neuro.01.006.2008
- Nieuwenhuis, S., Heslenfeld, D. J.,  
von Geusau, N. J., Mars, R. B.,  
Holroyd, C. B., and Yeung, N. (2005).  
Activity in human reward-sensitive  
brain areas is strongly context depend-  
ent. *Neuroimage* 25, 1302–1309.
- O’Doherty, J. P., Deichmann, R.,  
Critchley, H. D., and Dolan, R. J.  
(2002). Neural responses during  
anticipation of a primary taste reward.  
*Neuron* 33, 815–826.
- Passamonti, L., Rowe, J. B.,  
Schwarzbauer, C., Ewbank, M. P.,  
Von Dem Hagen, E., and Calder, A. J.  
(2009). Personality predicts the brain’s  
response to viewing appetizing foods:  
the neural basis of a risk factor for  
overeating. *J. Neurosci.* 29, 43–51.
- Paz, R., Pelletier, J. G., Bauer, E. P., and  
Pare, D. (2006). Emotional enhance-  
ment of memory via amygdala-driven  
facilitation of rhinal interactions. *Nat.  
Neurosci.* 9, 1321–1329.
- Pesaran, B., Nelson, M. J., and  
Andersen, R. A. (2008). Free choice  
activates a decision circuit between  
frontal and parietal cortex. *Nature* 453,  
406–409.
- Petrovich, G. D., Holland, P. C., and  
Gallagher, M. (2005). Amygdalar  
and prefrontal pathways to the later-  
al hypothalamus are activated by  
a learned cue that stimulates eating.  
*J. Neurosci.* 25, 8295–8302.
-  Popescu, A. T., Popa, D., and Pare, D.  
(2009). Coherent gamma oscillations  
couple the amygdala and striatum  
during learning. *Nat. Neurosci.*
- Quirk, G. J., and Beer, J. S. (2006).  
Prefrontal involvement in the regu-  
lation of emotion: convergence of  
rat and human studies. *Curr. Opin.  
Neurobiol.* 16, 723–727.
- Raichle, M. E., and Snyder, A. Z. (2007).  
A default mode of brain function:  
a brief history of an evolving idea.  
*Neuroimage* 37, 1083–1090.
- Ranganath, C., Heller, A., Cohen, M. X.,  
Brozinsky, C. J., and Rissman, J.  
(2005). Functional connectivity with  
the hippocampus during successful  
memory formation. *Hippocampus*  
15, 997–1005.
- Redish, A. D. (2004). Addiction as a com-  
putational process gone awry. *Science*  
306, 1944–1947.
- Reuter, J., Raedler, T., Rose, M., Hand, I.,  
Glascher, J., and Büchel, C. (2005).  
Pathological gambling is linked to  
reduced activation of the mesolimbic  
reward system. *Nat. Neurosci.* 8,  
147–148.

- Riba, J., Kramer, U. M., Heldmann, M., Richter, S., and Münte, T. F. (2008). Dopamine agonist increases risk taking but blunts reward-related brain activity. *PLoS ONE* 3, e2479. doi: 10.1371/journal.pone.0002479
- Rissman, J., Gazzaley, A., and D'Esposito, M. (2004). Measuring functional connectivity during distinct stages of a cognitive task. *Neuroimage* 23, 752–763.
- Rogers, B. P., Morgan, V. L., Newton, A. T., and Gore, J. C. (2007). Assessing functional connectivity in the human brain by fMRI. *Magn. Reson. Imaging*
- Rykhlevskaia, E., Gratton, G., and Fabiani, M. (2008). Combining structural and functional neuroimaging data for studying brain connectivity: a review. *Psychophysiology* 45, 173–187.
- Salvador, R., Suckling, J., Schwarzbauer, C., and Bullmore, E. (2005). Undirected graphs of frequency-dependent functional connectivity in whole brain networks. *Philos. Trans. R. Soc. Lond., B, Biol. Sci.* 360, 937–946.
- Sanfey, A. G., Hastie, R., Colvin, M. K., and Grafman, J. (2003). Phineas gauged: decision-making and the human prefrontal cortex. *Neuropsychologia* 41, 1218–1229.
- Schall, J. D. (2005). Decision making. *Curr. Biol.* 15, R9–R11.
- Scheres, A., Milham, M. P., Knutson, B., and Castellanos, F. X. (2007). Ventral striatal hypo-responsiveness during reward anticipation in attention-deficit/hyperactivity disorder. *Biol. Psychiatry* 61, 720–724.
- Schott, B. H., Minuzzi, L., Krebs, R. M., Elmenhorst, D., Lang, M., Winz, O. H., Seidenbecher, C. I., Coenen, H. H., Heinze, H. J., Zilles, K., Düzel, E., and Bauer, A. (2008). Mesolimbic functional magnetic resonance imaging activations during reward anticipation correlate with reward-related ventral striatal dopamine release. *J. Neurosci.* 28, 14311–14319.
- Schott, B. H., Seidenbecher, C. I., Fenker, D. B., Lauer, C. J., Bunzeck, N., Bernstein, H. G., Tischmeyer, W., Gundelfinger, E. D., Heinze, H. J., and Düzel, E. (2006). The dopaminergic midbrain participates in human episodic memory formation: evidence from genetic imaging. *J. Neurosci.* 26, 1407–1417.
- Schott, B. H., Sellner, D. B., Lauer, C. J., Habib, R., Frey, J. U., Guderian, S., Heinze, H. J., and Düzel, E. (2004a). Activation of midbrain structures by associative novelty and the formation of explicit memory in humans. *Learn. Mem.* 11, 383–387.
- Schott, B. H., Sellner, D. B., Lauer, C. J., Habib, R., Frey, J. U., Guderian, S., Heinze, H. J., and Düzel, E. (2004b). Activation of midbrain structures by associative novelty and the formation of explicit memory in humans. *Learn. Mem.* 11, 383–387.
- Schultz, W. (1998). Predictive reward signal of dopamine neurons. *J. Neurophysiol.* 80, 1–27.
- Schultz, W. (2002). Getting formal with dopamine and reward. *Neuron* 36, 241–263.
- Siegle, G. J., Thompson, W., Carter, C. S., Steinhauer, S. R., and Thase, M. E. (2007). Increased amygdala and decreased dorsolateral prefrontal BOLD responses in unipolar depression: related and independent features. *Biol. Psychiatry* 61, 198–209.
- Skudlarski, P., Jagannathan, K., Calhoun, V. D., Hampson, M., Van Leijenhorst, L., Zanolie, K., Van Meel, C. S., Westenberg, P. M., Rombouts, S. A., and Crone, E. A. (2008). Measuring brain connectivity: diffusion tensor imaging validates resting state temporal correlations. *Neuroimage* 43, 554–561.
- Spreckelmeyer, K. N., Krach, S., Kohls, G., Rademacher, L., Irmak, A., Konrad, K., Kircher, T., and Gruner, G. (2009). Anticipation of monetary and social reward differently activates mesolimbic brain structures in men and women. *Soc. Cogn. Affect. Neurosci.* 4, 158–165.
- Staempfli, P., Reischauer, C., Jaermann, T., Valavanis, A., Kollias, S., and Boesiger, P. (2008). Combining fMRI and DTI: a framework for exploring the limits of fMRI-guided DTI fiber tracking and for verifying DTI-based fiber tractography results. *Neuroimage* 39, 119–126.
- Stoeckel, L. E., Kim, J., Weller, R. E., Cox, J. E., Cook, E. W. III, and Horwitz, B. (2009). Effective connectivity of a reward network in obese women. *Brain Res. Bull.* 79, 388–395.
- Stoeckel, L. E., Weller, R. E., Cook, E. W. III, Twieg, D. B., Knowlton, R. C., and Cox, J. E. (2008). Widespread reward-system activation in obese women in response to pictures of high-calorie foods. *Neuroimage* 41, 636–647.
- Strohle, A., Stoy, M., Wrase, J., Schwarzer, S., Schlagenhauf, F., Huss, M., Hein, J., Nedderhut, A., Neumann, B., Gregor, A., Juckel, G., Knutson, B., Lehmkuhl, U., Bauer, M., and Heinz, A. (2008). Reward anticipation and outcomes in adult males with attention-deficit/hyperactivity disorder. *Neuroimage* 39, 966–972.
- Sun, F. T., Miller, L. M., and D'Esposito, M. (2004). Measuring interregional functional connectivity using coherence and partial coherence analyses of fMRI data. *Neuroimage* 21, 647–658.
- Swanson, L. W. (2000). Cerebral hemisphere regulation of motivated behavior. *Brain Res.* 886, 113–164.
- Tom, S. M., Fox, C. R., Trepel, C., and Poldrack, R. A. (2007). The neural basis of loss aversion in decision-making under risk. *Science* 315, 515–518.
- Ullsperger, M., and von Cramon, D. Y. (2003). Error monitoring using external feedback: Specific roles of the habenular complex, the reward system, and the cingulate motor area revealed by functional magnetic resonance imaging. *J. Neurosci.* 23, 4308–4314.
- Volkow, N. D., Fowler, J. S., Wang, G. J., and Goldstein, R. Z. (2002). Role of dopamine, the frontal cortex and memory circuits in drug addiction: insight from imaging studies. *Neurobiol. Learn. Mem.* 78, 610–624.
- Volkow, N. D., Wang, G. J., Fowler, J. S., and Telang, F. (2008). Overlapping neuronal circuits in addiction and obesity: evidence of systems pathology. *Philos. Trans. R. Soc. Lond., B, Biol. Sci.* 363, 3191–3200.
- Volkow, N. D., Wang, G. J., Telang, F., Fowler, J. S., Logan, J., Childress, A. R., Jayne, M., Ma, Y., and Wong, C. (2006). Cocaine cues and dopamine in dorsal striatum: mechanism of craving in cocaine addiction. *J. Neurosci.* 26, 6583–6588.
- Volkow, N. D., and Wise, R. A. (2005). How can drug addiction help us understand obesity? *Nat. Neurosci.* 8, 555–560.
- Weintraub, D., Siderowf, A. D., Potenza, M. N., Goveas, J., Morales, K. H., Duda, J. E., Moberg, P. J., and Stern, M. B. (2006). Association of dopamine agonist use with impulse control disorders in Parkinson disease. *Arch. Neurol.* 63, 969–973.
- Wise, R. A. (2002). Brain reward circuitry: insights from unsensed incentives. *Neuron* 36, 229–240.
- Wise, R. A., and Rompre, P. P. (1989). Brain dopamine and reward. *Annu. Rev. Psychol.* 40, 191–225.
- Wittmann, B. C., Schiltz, K., Boehler, C. N., and Düzel, E. (2008). Mesolimbic interaction of emotional valence and reward improves memory formation. *Neuropsychologia* 46, 1000–1008.
- Wrase, J., Kahnt, T., Schlagenhauf, F., Beck, A., Cohen, M. X., Knutson, B., and Heinz, A. (2007). Different neural systems adjust motor behavior in response to reward and punishment. *Neuroimage* 36, 1253–1262.
- Xiong, J., Parsons, L. M., Gao, J. H., and Fox, P. T. (1999). Interregional connectivity to primary motor cortex revealed using MRI resting state images. *Hum. Brain Mapp.* 8, 151–156.
- Yacubian, J., Glascher, J., Schroeder, K., Sommer, T., Braus, D. F., and Buchel, C. (2006). Dissociable systems for gain- and loss-related value predictions and errors of prediction in the human brain. *J. Neurosci.* 26, 9530–9537.
- Yacubian, J., Sommer, T., Schroeder, K., Glascher, J., Kalisch, R., Leuenberger, B., Braus, D. F., and Buchel, C. (2007). Gene-gene interaction associated with neural reward sensitivity. *Proc. Natl. Acad. Sci. U.S.A.* 104, 8125–8130.
- Ye, Z., Hammer, A., Camara, E., and Münte, T. F. (2009). Pramipexole modulates the neural network of reward anticipation. *Brain* (submitted).

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