Measuring extrastriatal dopamine release during a reward learning task

Elske Vrieze\textsuperscript{1}, Jenny Ceccarini\textsuperscript{2}, Diego A. Pizzagalli\textsuperscript{3}, Guy Bormans\textsuperscript{4}, Mathieu Vandenbulcke\textsuperscript{1}, Koen Demyttenaere\textsuperscript{1}, Koen Van Laere\textsuperscript{2}, and Stephan Claes\textsuperscript{1}

\textsuperscript{1}Department of Psychiatry, University Hospital Leuven, Belgium
\textsuperscript{2}Division of Nuclear Medicine, University Hospital Leuven, Belgium
\textsuperscript{3}Neuroimaging Center & Center for Depression, Anxiety, and Stress Research, McLean Hospital, Harvard Medical School, Belmont, Massachusetts, USA
\textsuperscript{4}Laboratory for Radiopharmacy, University Hospital Leuven, Belgium

Abstract

**Objectives**—Reward learning is critical for survival. Animal research emphasizes the role of dopaminergic (DA) mesocorticolimbic pathways in reward learning, but few studies have evaluated extrastriatal DA functioning in humans. The purpose of this study was to examine presynaptic DA release in extrastriatal regions of the reward circuit by measuring displacement of the high affinity D\textsubscript{2}/D\textsubscript{3} radioligand \textsuperscript{18}F\textsubscript{Fallypride} during a reward task.

**Design**—Ten healthy volunteers underwent a \textsuperscript{18}F\textsubscript{Fallypride} Positron Emission Tomography protocol while performing a reward task, allowing us to assess participants’ ability to modulate behavior as a function of reward. DA receptor ligand displacement was correlated with task performance and self-reported anhedonia.

**Observations**—Parametric t-maps revealed significant decrease in \textsuperscript{18}F\textsubscript{Fallypride} binding in the medial orbitofrontal cortex (mOFC), ventromedial prefrontal cortex (vmPFC) and dorsal anterior cingulate cortex (dACC), indicating endogenous DA release in these regions. Increasing anhedonic symptoms correlated with DA release in the left vmPFC, left dACC, and right dACC emerged (all rs > 0.65, ps < 0.05). Similarly, reduced reward learning correlated with higher DA release in left vmPFC, right vmPFC, and left dACC (all rs < −0.64, ps < 0.05). Left dACC (r = 0.66, p = 0.04) and left vmPFC (r = 0.74, p = 0.01) DA release showed a significant positive correlation with impaired tendency to modulate behaviour as a function of prior positive reinforcements.

**Conclusions**—These findings support the hypothesis that DA release in mOFC, vmPFC and dACC regions plays an important role in reinforcement learning in the human brain.

**Keywords**

Positron emission tomography; PET; Reward learning; Extrastriatal Reward Circuit; Dopamine; \textsuperscript{18}F\textsubscript{Fallypride}; Anhedonia

---

Corresponding author: Elske Vrieze, Division of Psychiatry, University Hospital, campus Leuven, Sint Rafael Hospital, Kapucijnenvoer 33, 3000 Leuven, Belgium, elske.vrieze@uc-kortenberg.be, tel: +32-495502562, fax: +32-16-332640.

All other authors reports no biomedical financial interest.
Introduction

Anhedonia – the loss of pleasure or lack of reactivity to pleasurable stimuli – is considered a promising endophenotype of Major Depressive Disorder (MDD) (Hasler et al. 2004; Vrieze and Claes, 2009). However, its neurobiological underpinnings in humans remain largely unknown. Various components of anhedonic behaviour have been linked to dysfunction in both striatal and extrastriatal mesocortical areas, which are part of the dopaminergic brain reward circuit (Dillon et al. 2008; Knutson and Wimmer, 2007; Kringselbach and Berridge, 2009; Lucas et al. 2004). Reward learning has been linked directly to DA neurotransmission (Hasler et al., 2009a). Striatal regions of the brain reward circuit (e.g., nucleus accumbens) have generally been implicated in anticipation of reward and reward-seeking behaviour (Ikemoto and Panksepp, 1999; Wrase et al. 2007). The extrastriatal areas appear to play an important role in reward-related decision, reinforcement learning, and reward consumption (Knutson et al. 2001; Rushworth and Behrens, 2008). Consistent with this notion, the orbitofrontal cortex (OFC) and ventromedial prefrontal cortex (vmPFC) regions have been implicated in encoding representations of expected value, whereas the anterior cingulate cortex (ACC) appears to utilize reinforcement histories to guide behavior (Cox et al. 2005; Jocham et al. 2011; Knutson et al. 2005; Roesch and Olson, 2004; Rushworth et al. 2007).

Recent developments of high affinity positron emission tomography (PET) D2/D3 radioligands, such as [18F]Fallypride, provide the unique opportunity to directly investigate regions with low D2/D3 receptor density in vivo, including the extrastriatal reward circuit (Aalto et al. 2005; Mukherjee et al. 2002; Riccardi et al. 2008), using both pharmacological challenges as well as functional stimulation tasks (Badgaiyan et al. 2009; Christian et al. 2006; Riccardi et al. 2006a). Extending this prior work, the purpose of this study was to examine presynaptic DA release in extrastriatal regions of the reward circuit by measuring D2/D3 radioligand [18F]Fallypride displacement in response to a probabilistic reward task involving monetary reward. Specifically, our goal was to correlate reward task performance and self-report measures of anhedonia with the extent of DA release in those areas showing a task-induced DA modulation. Based on prior findings (Keedwell et al. 2005; St. Onge et al., 2011; Santesso et al. 2008; Wacker et al. 2009), we expected to find task-induced DA release in the vmPFC, OFC and dorsal ACC and to uncover links between (1) the spatial extent of the estimated DA release within each specific region of interest and (2) subjective hedonic capacity as well as the ability to respond to reinforcement stimuli.

A single-scan session PET design was used. Changes in ligand binding were computed by applying the linear extension of the simplified reference region model (LSRRM), with reduced binding indicating dopamine release (Alpert et al. 2003). The reward task has been previously used to objectively measure reward responsiveness in healthy volunteers (Pizzagalli et al. 2005) and dysfunctional reinforcement learning in MDD (Pizzagalli et al. 2009). Moreover, a single dose of a DA agonist – hypothesized to activate DA autoreceptors and thus reduce DA release – blunted reward responsiveness (Pizzagalli et al. 2008) and altered reward-related dorsal ACC activation (Santesso et al. 2009) in healthy volunteers, suggesting that performance in this task is modulated by DA.

2. Material and Methods

2.1 Participants

The study included 10 healthy, non-smoking, right-handed volunteers (mean age ± SD, 33.3 ± 8.2 y). In light of gender differences in extrastriatal DA release (Riccardi et al. 2006b), only females were included. Participants with current neurological (e.g. head trauma, seizures) or somatic illnesses, current or past mood disorders, psychotic disorders, substance and/or alcohol abuse (as determined by the Structured Clinical Interview for DSM-IV-TR...
(SCID-I) (Spitzer et al. 1992) were excluded. None of the subjects was taking psychotropic medication. All subjects refrained from food and drinks for at least four hours before scanning. Blood and urine samples were taken prior to the assessment to exclude pregnancy, substance abuse and abnormal thyroid levels. Written informed consent was obtained from each subject. The study was approved by the local ethics committee (UZ Leuven commissie voor Medische Ethiek) and performed according to the World Medical Association Declaration of Helsinki. Participants received 150 euro for participating in the study.

2.2 Study design and procedure

Subjects meeting inclusion criteria were invited for a single day of testing. Participants started the session by completing the Snaith Hamilton Pleasure Scale (SHAPS) (Leventhal et al. 2006; Snaith et al. 1995). The SHAPS is a 14-item questionnaire probing participants’ hedonic capacity in a variety of situations. Answers are logged on a 4-point scale (strongly agree, agree, disagree or strongly disagree). A total score was computed by summing the responses to each item, with higher scores indicating lower hedonic capacity, i.e., higher anhedonia (Franken et al. 2007). Further, participants filled out the Beck Depression Inventory (BDI) (Beck et al. 1996). The BDI is a widely used and reliable measure of depressive symptoms; higher scores reflect increasing levels of depressive symptoms.

Next, a structural T2- and volumetric T1-weighted magnetic resonance imaging (MRI) head scan was obtained on a 1.5 Tesla Vision Scanner (Siemens, Germany). For the PET assessment, a single imaging protocol was initiated using a HR+ PET (Siemens, Ehrlangen, Germany) operating in three-dimensional acquisition mode. The subjects’ head and body were fixated to minimize head movement. A brief training session of the reward task was provided before scanning. The computer task was presented on a flat screen placed in a comfortable viewing position.

The PET emission was acquired in two blocks, in accordance with the PET imaging protocol of Christian and colleagues (2006). The first PET session, representing the baseline radiotracer kinetics, consisted of 35 frames (first 6 of 60 sec/frames and 120 sec/frames thereafter) and was initiated after intravenous administration of $[^{18}\text{F}]$Fallypride (mean ± SD, 179 ± 17 MBq). After a 15-min break, the second PET emission data were collected for another 70 min: the first 20 min represented an extension of the baseline scan. Then, at 100 min post-injection, the reward task was initiated, lasting on average 46 min (SD: 3 min). The scan ended 150 min post-injection, at which time all participants had completed the task. To correct for attenuation, transmission scans were acquired with a 68-germanium source before radiotracer administration and at the end of the scanning session.

2.3 Reward Task

The task was a computerized probabilistic reward task in which correct identifications of two difficult-to-discriminate stimuli were differentially rewarded (Pizzagalli et al. 2005). The task consisted of 600 trials, divided in six blocks of 100 trials, separated by five short breaks (30 sec). Each trial started with a fixation point, shown for 500 msec in the middle of the screen, which was replaced by a mouthless cartoon face. After 500 msec, a short (11.5 mm) or long (13 mm) mouth appeared for 100 msec. The participants’ task was to determine, as quickly and accurately as possible, whether the short or long mouth had been presented by pressing a corresponding key. Before the PET scan, participants received verbal and written descriptions of the reward task, and were told that the goal of the task was to win as much money as possible. Moreover, they were informed that not all correct responses would result in a monetary reward. However, it was emphasized that more correct identifications would result in more reward feedback. Participants were informed that they would receive the total amount of accumulated money at the end of the experiment.

_Hum Brain Mapp._ Author manuscript; available in PMC 2014 March 01.
In each of the six blocks, both stimuli were shown an equal number of times. In each block, a monetary reward feedback was given to approximately 40 correct answers (Fig. 1). To induce a response bias, an asymmetrical reinforcer schedule was used, such as correct responses for one mouth (referred to as the ‘rich stimulus’) were rewarded three times more frequently (30 vs. 10) than correct responses of the other mouth (referred to as the ‘lean stimulus’). Due to this unequal frequency of reward feedback, participants with high reward responsiveness were expected to develop a response bias in favor of the rich stimulus. Subjects with low reward responsiveness, such as subjects with elevated depressive (particularly anhedonic) symptoms, were expected to develop a smaller or no bias, consistent with prior findings (Pizzagalli et al. 2005, 2009).

2.4 Data reduction and statistical analysis

2.4.1 PET data—Subject images were first reconstructed using a standard three-dimensional (3D) filtered back-projection algorithm (Kinahan and Rogers, 1990) including model-based scatter as well as attenuation correction. During reconstruction, the intrinsic resolution was modeled as a 4.3-mm Gaussian (Adam et al. 1997). The image was reconstructed as a 128 × 128 × 63 voxel matrix (pixel size = 2.06 mm × 2.06 mm × 2.43 mm). Attenuation- and scatter-corrected reconstruction images were post-smoothed with a Gaussian filter with FWHM (full-width at half-maximum) equals to 4.8 mm, yielding images with a final spatial resolution of 6.5 mm.

PET frames were realigned, coregistered onto the individual MRI, and spatially normalized to the MNI template (using SPM2, Wellcome Department of Neurology).

For each individual, the mOFC (BA11), vmPFC (BA10) and dACC (BA32) were automatically defined in MNI space using sets of volume-of-interests (VOI) defined according to Brodmann areas (BA) on the basis of the Talairach Atlas (Talairach and Tournoux, 1988). VOI were constructed using the PMOD software VOI tool (PMOD Inc., Zurich, Switzerland), and then applied to each corresponding spatially normalized T1-weighted MRI image.

To measure reward-induced DA release in extrastriatal regions, we implemented a kinetic analysis of a single $^{[18F]}$Fallypride scan based on a linear extension of the simplified reference region model (LSRRM) (Alpert et al. 2003) using Matlab-based in-house software (MathWorks, Natick, MA, USA). This technique includes a baseline and an activation condition. Assuming that the steady physiological state is not maintained during the single scan session, the task-induced effects are measured by time-dependent changes in ligand binding. Since binding of $^{[18F]}$Fallypride to extrastriatal D$_2$/D$_3$ receptors is sensitive to endogenous DA levels (Mukherjee et al. 2002), a reduction in $^{[18F]}$Fallypride binding potential is assumed to be caused by direct competition of the tracer with DA at the D$_2$/D$_3$ receptor sites. The modified LSRRM approach allows for the dissociation rate of ligand ($k_{2a}$) to change over time in response to DA fluctuations, consequently to a change of binding potential. This time-dependent change of $k_{2a}$ is represented by the time-dependent parameter $k_{2a} + \gamma h(t)$, where $\gamma$ represents the amplitude of transient effects and the function $h(t)$ describes a rapid change following task onset and dissipation over time. The function $h(t)$ is an exponential decay function ($Alpert et al. 2003; Christian et al. 2006$) with the following form: $h(t) = 0$ (for $t < T$) and $h(t) = \exp^{-\tau(t-T)}$ (for $t \geq T$), where $\tau$ controls the rate at which activation effects dissipate ($\tau = 0.03 \text{ min}^{-1}$) and $T$ indicates the task initiation ($T = 100 \text{ min}$). Thus, an increased $k_{2a}$ would be reflected in a decreased binding potential for D$_2$/D$_3$ receptors due to an increased DA release and would result in a positive value of $\gamma$.

For each subject, a voxel-based analysis of the data was carried out using the LSRRM, which generated individual quantitative parametric maps of the kinetic parameters. Next, in
order to identify regions showing maximum radioligand displacement across subjects, these maps were combined in a t-statistic map of the $\gamma$ parameter ($t > 5$, $p < 0.000005$, one-tailed). This threshold corresponded to Bonferroni-corrected $p < 0.05$ (0.05/average total number of voxels analyzed per subject ($= 11,017$) = 0.0000045). Consistent with prior studies (Christian et al. 2006), the percentage of statistically significant voxels within each region of interest was calculated to identify the spatial extent of the estimated DA release during the reward task.

2.4.2 Reward Task—We evaluated overall task performance by computing response bias (RB) and discriminability. The primary variable of interest was RB, which captures participants’ ability to modulate behaviour as a function of reward. RB was computed as:

$$\log b = \frac{1}{2} \log \left( \frac{rich_{correct} \times lean_{incorrect}}{rich_{incorrect} \times lean_{correct}} \right)$$

To enable RB calculation in cases with zero in one cell of the formula, 0.5 was added to each cell in the matrix. RB is high when participants correctly classify the stimulus associated with more frequent reward (rich stimulus), and misclassify the lean stimulus. Discriminability, which captures participants’ ability to perceptually distinguish between the two stimuli and thus provides a measure of task difficulty, was computed as:

$$\log d = \frac{1}{2} \log \left( \frac{rich_{correct} \times lean_{correct}}{rich_{incorrect} \times lean_{incorrect}} \right)$$

Moreover, reaction time (RT) and hit rates (% correct responses) for the rich and lean stimulus were calculated, to confirm that the task elicited the intended behavioural effects. Outlier RTs were excluded using the 2-step procedures described in prior work (Pizzagalli et al. 2005). RB and discriminability were analyzed using one-way analysis of variance (ANOVA) entering Block (1–6) as repeated measure. For RT and hit rate scores, Stimulus (rich, lean) was entered as an additional repeated measure. The Greenhouse-Geisser correction was used, when appropriate. Significant ANOVA effects were further evaluated with simple t-tests.

To directly assess overall reward learning, a difference score between RB in block 1 and 6 was computed ["$\Delta$ response bias" = RB(Block 6) – RB(Block 1)]. In a secondary analysis, we computed the probability of rich misses as a function of which stimulus was rewarded in the immediately preceding trial. The calculated probability values allowed us to investigate the strength of the response bias as a function of the rewarded stimulus in the immediately preceding trial.

2.4.3 Correlations—Pearson correlations were computed among (1) SHAPS and BDI scores, (2) task performance, and (3) percentage of statistically significant voxels using SAS version 9.2.

3. Results

3.1 Reward task performance

Replicating prior studies (Pizzagalli et al. 2005, 2008, 2009), the one-way ANOVA on RB scores revealed a main effect of Block ($R(5,45) = 3.51, p = .041, \epsilon = 0.48$) and within-subjects analyses indicated that the linear contrast was significant ($R(1,9) = 17.01, p = .003$) due to a general increase in RB scores over blocks (Figure 2A). Follow-up paired t-tests
indicated that RB in block 4 (t(9) = 2.24, p = .052), block 5 (t(9) = 2.79, p = .021), and block 6 (t(9) = 2.68, p = .025) was greater than in block 1. Moreover, 9 of the 10 subjects had a positive Δresponse bias (binomial $p(9/10) < .01$). Critically, the one-way ANOVA on discriminability scores revealed no significant effect of Block ($F(5,45) = 0.48, p > .66, \varepsilon = 0.49$) (Figure 2B), indicating that task difficulty was stable across blocks.

Evidence for a behavioural preference in favour of the rich stimulus emerged also from analyses of hit rates and reaction time. For hit rates, significant Stimulus ($F(1,9) = 13.86, p < .005$) and Stimulus × Block ($F(1,45) = 3.54, p < .009$) effects emerged, due to significantly higher scores for the rich relative to the lean stimulus and the fact that differences become larger over the course of the blocks (Figure 2C). For reaction time scores, a main effect of Stimulus emerged, due to significantly faster response to the rich than lean stimulus ($F(1,9) = 5.32, p < .05$) (Figure 2D). Collectively, these findings indicate that participants developed a response bias in favour of the more frequently rewarded rich stimulus in the absence of fluctuations in task difficulty, confirming that the reward task produced the intended behavioural effect.

3.2 DA release during the reward task

Using LSRRM, we first performed a voxel-based estimation of the kinetic parameters for each subject. Figure 3 represents an illustrative example of a $\gamma$ parametric image for one subject, showing that the rate of ligand displacement increased during the reward task in the medial OFC, vmPFC, and dACC. In these regions, the mean $\gamma$ obtained was between 0.008 and 0.01. All mean $\gamma$ values were positive, indicating DA release. A covariance image of $\gamma$ (standard deviation, sd($\gamma$)) was calculated to generate a statistic t-map for $\gamma$ [$t = \gamma/sd(\gamma)$], which allowed us to visualize regions with significant task-induced ligand displacement. Parametric t-map across subjects (Figure 4) confirmed significant ligand displacement in the mOFC (Brodmann area (BA) 11) and vmPFC (BA10). Significant tracer displacement was also seen in the dACC (BA32), which is not shown in these slices. Table 1 lists the percentages of significant voxels with a t score that exceeded the threshold of $t > 5$ within the three activated regions of interest: mOFC, vmPFC and dACC. For all three regions, there was no significant difference between the percentage of significant voxels in the left versus right hemisphere. Each subject showed a considerable ligand displacement within the mOFC, vmPFC and dACC during the task, as indicated by the average statistical parametric t map of $\gamma$, as well as the total number of significant voxels.

3.3 Correlations between questionnaire data, task performance and ligand displacement

All ten subjects completed the SHAPS (mean ± SD, 17.9 ± 2.92) and BDI (mean ± SD, 1.7 ± 1.57). Kolmogorov-Smirnov tests indicated that Δresponse bias and self-rating scores across the 10 subjects were normally distributed (all ps > .15). Box and whisker plots did not detect outliers. Pearson correlations with Δresponse bias and BDI ($r = -0.32, p = .37$), as well as SHAPS ($r = -0.52, p = .13$) were not significant, likely due to the truncated range in this healthy sample and the small sample size. SHAPS scores were positively correlated with percentage of statistically significant voxels in left vmPFC ($r = 0.65, p = .04$), left dACC ($r = 0.74, p = .01$) and right dACC ($r = 0.66, p = .04$), indicating that lower hedonic capacity was associated with a higher number of activated voxels. Along similar lines, Δresponse bias scores were negatively correlated with the percentage of significant voxels in the left vmPFC ($r = -0.70, p = .02$), right vmPFC ($r = -0.64, p = .04$) and left dACC ($r = -0.67, p = .03$), suggesting that a lower ability to modulate behaviour as a function of the reinforcement schedule was associated with a higher number of activated voxels (Table 1). Highlighting the specificity of this finding, mean discriminability scores did not correlate with the percentage of significant voxels in these areas (all ps > 0.3). Finally, the probability of rich misses after an immediately preceding rewarded rich trial was significantly
correlated with the percentage of significant voxels in the left dACC ($r = 0.66$, $p = .04$) and left vmPFC ($r = 0.74$, $p = .01$). These correlation data are plotted in Figure 5.

Box and whisker plots of the percentage of statistically significant voxels in BA10 and BA32 showed that subject 9 behaved as an outlier in our sample. She was the youngest participant (21 years old), but there were no indications of irregularities in data collection.

When excluding her data, the correlations between (1) the percentage of statistically significant voxels and (2) SHAPS ($0.29 < r_s < 0.52$) and $\Delta$response bias ($-0.36 < r_s < -0.28$) were substantially weakened. The correlation with the percentage of significant voxels in the left vmPFC and the probability of rich misses in the reward task still showed a significant trend ($r = 0.66$, $p = .052$).

4. Discussion

The main goal of this study was to provide *in vivo* proof for the hypothesis that presynaptic DA release in specific areas of the extrastriatal reward circuit plays an important role in reward sensitivity and hedonic capacity. By measuring the high affinity $D_2/D_3$ radioligand $[^{18}F]$Fallypride binding potential in response to a monetary reward learning task, we found that the tracer was successfully displaced from $D_2/D_3$ receptors in the mOFC (BA10), vmPFC (BA11), and dACC (BA32), providing indirect evidence of endogenous DA release during the task. In addition, to capture the spatial extent of extrastriatal DA release, the percentage of statistically significant voxels within the three activated regions was computed and correlated with subjective and objective measures of anhedonia. These analyses revealed that SHAPS scores (i.e., anhedonic symptoms) correlated positively with the percentage of significant voxels in the left vmPFC and left bilateral dACC. Moreover, when considering performance in the probabilistic reward task, we found a significant inverse correlation between $\Delta$response bias and the percentage of statistically significant voxels in the left and right vmPFC and left dACC. Accordingly, increasing anhedonia and reduced increases in response bias over the course of the experiment were associated with a spatially larger DA release in the vmPFC and dACC. These findings were further extended by the observation that the probability of rich misses after a preceding rewarded rich trial was positively correlated with the percentage of significant voxels in the left dACC and left vmPFC. Thus, a relatively lower ability to integrate reinforcement history over time was associated with a larger DA release in the dACC and vmPFC. Collectively, these findings indicate that dopamine played a direct role in reward responsiveness and more specifically in the ability to modulate behaviour as a function of reinforcement history, particularly in the presence of immediate reinforcement.

These results are in line with growing evidence implicating mOFC, vmPFC and dACC in reward processing. Using catecholamine depletion, Hasler et al. (2008, 2009b) have shown a direct relationship between DA dysfunction and reward responsiveness, as well as depressive symptoms and anhedonia, within the limbic-cortical- striatal-pallidal-thalamic circuit. In addition, the role of the OFC has specifically been reported in reward-related and goal-directed decision making (Bechara et al. 2000; Hare et al. 2008). Moreover, reward learning has been associated with the vmPFC (Rudebeck et al. 2008) and it is hypothesized that the vmPFC receives information based on the expectancy of reinforcement, which is used for adaptive decision making (Gottfried et al. 2003; Schoenbaum and Roesch, 2005). Furthermore, a growing number of studies on reinforcement learning and goal-directed decision making implicate the ACC in reward-related adaptive behaviour (Amiez et al. 2006; Kennerly et al. 2006; Mansouri et al. 2009; Rogers et al. 2004), particularly in the ability to influence a current choice by means of previous action-reinforcement history (Rudebeck et al. 2008; Williams et al. 2004).
Although a significant tracer displacement was observed in the OFC, we failed to find a reliable correlation between task performance and the percentage of significant voxels in the mOFC. This might be explained by the fact that the OFC has been primarily linked to reward consumption (O’Doherty et al. 2000, 2001) and its activity has been assumed to reflect the subjective (hedonic) experience of a decision (Peters and Büchel, 2010). The current reward task was designed to assess how behaviour is modulated by reinforcement history. Although the asymmetrical reinforcement schedule was successful in inducing a behavioral response bias, it was likely too weak and short-lived to elicit a clear-cut hedonic response. Unfortunately, we did not include an independent affective rating to assess hedonic responses during the task, which is a limitation that should be addressed in future studies.

These findings echo prior results of positive correlations between vmPFC activation in response to positive stimuli and anhedonia (Harvey et al. 2007; Keedwell et al. 2005; St. Onge et al. 2011; Wacker et al. 2009). In MDD, reward-related vmPFC responses have been interpreted as reflecting cortical compensatory mechanisms due to reduced striatal responses to positive stimuli (Dunn et al. 2002). The present findings of positive correlations between spatial extent of DA release and both objective and subjective measures of anhedonia are consistent with this speculation. Alternatively, it is possible that subjects with lower reward responsiveness might activate cortical dopaminergic mechanisms to increase attention and execute the task more accurately, leading to more correct responses for the less frequently rewarded stimulus, and thus a reduced response bias. This alternative interpretation fits accounts that depressed individuals have a more accurate view of reality and are less susceptible to positivity biases (Alloy and Abramson, 1979). Clearly, even though several studies have identified areas of the PFC as important regulators of reward-related behavior, the precise mechanism of action for processing motivationally salient information and guide adaptive behavior remains subject to much debate. Moreover, interrelations between DA levels in the PFC and striatal areas remain unclear (Karreman and Moghaddam, 1996; Murase et al. 1993; Taber and Fibiger, 1993) and PET procedures allowing for the assessment of both striatal and extrastriatal DA release will be required to evaluate such relations.

The strengths of the current study warrant mention. First, using a high-affinity D2/D3 antagonist radiotracer to explore the extrastriatal DA release was an important extension of research on the reward circuit, which until recently focused on striatal changes (Bressan and Crippa, 2005; Schultz, 2000). Second, most of the findings on the function of DA modulation in the reward circuit, such as the involvement of DA release in the vmPFC and dACC reward learning, stem from animal research, which await in vivo validation in humans. The current strategy to combine a well-established reward task with a single dynamic PET protocol is novel and afforded the opportunity to directly assess reward-related DA neuromodulation in brain regions hypothesized to be critically involved in the pathophysiology of neuropsychiatric illnesses, particularly MDD (Dunlop and Nemeroff, 2007; Nestler and Carlezon, 2005). Because we identified correlations between reward processing and anhedonic symptoms, on one hand, and a measure of DA release in the vmPFC and dACC, on the other hand, and all these variables have been suggested to be impaired in MDD, especially in patients with elevated anhedonic symptoms (Price and Drevets, 2010), we provided an important support for the function of the reward circuit as a plausible biological basis for an anhedonic endophenotype in MDD. A critical next step will be to evaluate patients with MDD. This would increase our understanding of the link between abnormal DA release and the psychopathology of symptoms and course of MDD, which may have important therapeutic implications.
The limitations of the current study should be acknowledged. First, the PET sample is relatively small and the results should be considered preliminary. Accordingly, the correlational findings emerging from the current study await independent replications in larger samples. Second, this study used a novel PET imaging technique which measured the extent of DA involvement as the percentage of significant voxels. Although we believe that the percentage of significant voxels is one of the best methods to quantify DA release in a individual areas, this quantitative approach should be tested more broadly to confirm its reliability. Third, the results do not imply causality and, as stated before, many questions about the function of each area are still unclear and should be addressed in future research. Fourth, striatal regions of the brain reward circuit, such as nucleus accumbens, were not analyzed in this experiment because the pseudo-equilibrium of the $^{[18]}$F]Fallypride was not ensured. Compared to lower density D$_2$/D$_3$ receptor regions (i.e. PFC), the relatively higher D$_2$/D$_3$ receptor density in the striatal regions require a longer baseline scan duration (2–3 hours) in order to reach a similar proportion of receptors to be occupied by the DA-competing ligand $^{[18]}$F]Fallypride, and thus achieve a stable measurement of BP (Christian et al. 2000). Accordingly, because the portion of the baseline scan, and consequently the task initiation, should have been adjusted in order to investigate different brain regions, it was not feasible to use LSRRM to measure DA release concurrently in both striatal and extrastriatal regions. To corroborate our experimental design and estimate quantitatively the ability of the LSRRM model to detect DA transmission simultaneously in both extrastriatal and striatal regions, we analyzed the kinetic characteristics of $^{[18]}$F]Fallypride through a simulation study with variable stimulus intensity and timing (J. Ceccarini, E. Vrieze, M. Koole, T. Muylle, G. Bormans, S. Claes and K. Van Laere, unpublished observations). The suitability of the settings used in the current study was confirmed; however, in line with previous suggestions (Christian et al. 2000), a postponed task initiation at 120–150 min post-injection could even enhance the relative sensitivity of detecting DA release in striatal regions (Vernaleken et al. 2011).

Acknowledgments

The authors acknowledge the PET radiopharmacy for their skilled collaboration and declare no competing financial interests.

Financial Disclosure

This study was supported by the Fund for Scientific Research, Flanders, Belgium (FWO). Dr. Van Laere and Dr. Claes are both Senior Clinical Investigator of the FWO and Dr. Claes has received research support from Johnson and Johnson. Dr. Vrieze was supported by research grant OT 06/60 from the University of Leuven (KUL). Dr. Pizzagalli was supported by NIMH grant R01MH68376 and R21MH078979. Over the past three years, Dr. Pizzagalli has received research support from ANT North America (Advanced Neurotechnology), consulting fees from ANT and AstraZeneca, and honoraria from AstraZeneca for projects unrelated to the current study.

References


Beck, AT.; Steer, RA.; Brown, GK. Beck Depression Inventory Manual. 2nd ed. San Antonio, TX: The Psychological Corporation; 1996.


Dunlop BW, Nemeroff CB. The role of dopamine in the pathophysiology of depression. Arch Gen Psychiatry. 2007; 64:327–337. [PubMed: 17339521]


Hum Brain Mapp. Author manuscript; available in PMC 2014 March 01.


Pizzagalli DA, Evins AE, Schetter EC, Frank MJ, Pajtas PE, Santesso DL, Culhane M. Single dose of a dopamine agonist impairs reinforcement learning in humans: behavioural evidence from...


Figure 1.
Schematic representation of the probabilistic reward task. In each trial, the participants’ task was to decide (via key press) whether a short or long month stimulus had been presented in the mouthless face on the screen. In approximately 40% of the trials, monetary reward was presented after correct identifications.
Figure 2.
Task performance during the probabilistic reward task. Response Bias (A), discriminability (B), hit rate (C) and reaction time (in ms) (D) for the whole sample (n = 10). Error bars represent standard errors. For hit rate and reaction time, the rich condition (black bars) refers to the stimulus associated with more frequent reward, whereas the lean condition (light grey bars) refers to the stimulus associated with less frequent reward.
Figure 3.
Illustrative example of a $\gamma$ parametric image (shown in transversal (A), sagittal (B) and coronal (C) views) overlaid on a structural MRI for a single subject. Increasing $\gamma$ values correspond to greater ligand displacement.
Figure 4.
Statistical parametric $t$-map (across all subjects) showing medial OFC and vmPFC regions with significant tracer displacement during the probabilistic reward task. The color-coded $t$-values ($t > 0$) are overlaid on a MRI template. [$^{18}$F]Fallypride displacement is shown in coronal (A), sagittal (B) and transversal (C) views. L, left; R, right.
Figure 5.
Scatterplot and Pearson correlation between Δresponse bias and the percent of significant voxels in the (A) left vmPFC (BA10L), (B) right vmPFC (BA10R) and (C) left dACC (BA32L). Panel D and H show the relationship between the probability of a rich miss immediately after a rewarded rich trial and the percentage of statistically significant voxels in the left vmPFC (BA10L) and left dACC (BA32L), respectively. Panel E, F, and G show the relationship between SHAPS scores and the percentage of statistically significant voxels in the left vmPFC (BA10L), left dACC (BA32L), and right dACC (BA32R).
Table 1

Spatial extent of the estimated dopamine release induced by the reward task, represented as the percentage of significant voxels exceeding the threshold of \( t > 5 \) within each activated region.

<table>
<thead>
<tr>
<th>Subject</th>
<th>Spatial extent of the estimated dopamine release</th>
<th>Reward Task</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>BA11L</td>
<td>BA11R</td>
</tr>
<tr>
<td>1</td>
<td>46%</td>
<td>56%</td>
</tr>
<tr>
<td>2</td>
<td>29%</td>
<td>21%</td>
</tr>
<tr>
<td>3</td>
<td>8%</td>
<td>7%</td>
</tr>
<tr>
<td>4</td>
<td>8%</td>
<td>4%</td>
</tr>
<tr>
<td>5</td>
<td>12%</td>
<td>9%</td>
</tr>
<tr>
<td>6</td>
<td>57%</td>
<td>50%</td>
</tr>
<tr>
<td>7</td>
<td>70%</td>
<td>64%</td>
</tr>
<tr>
<td>8</td>
<td>37%</td>
<td>41%</td>
</tr>
<tr>
<td>9</td>
<td>26%</td>
<td>17%</td>
</tr>
<tr>
<td>10</td>
<td>8%</td>
<td>12%</td>
</tr>
</tbody>
</table>

BA11 = medial orbitofrontal cortex (mOFC); BA10 = ventromedial prefrontal cortex (vmPFC); BA32 = dorsal anterior cingulate cortex (dACC); L = left; R = right.

Right columns represent main results from the reward task, based on \( \Delta \) response bias (computed as RB(block 6) – RB(block 1)) and the probability of a rich miss immediately after a rewarded correct identification of the rich stimulus.