Exploring variability in basal ganglia connectivity with functional MRI in healthy aging

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Abstract
Changes in functional connectivity (FC) measured using resting state fMRI within the basal ganglia network (BGN) have been observed in pathologies with altered neurotransmitter systems and conditions involving motor control and dopaminergic processes. However, less is known about non-disease factors affecting FC in the BGN. The aim of this study was to examine associations of FC within the BGN with dopaminergic processes in healthy older adults. We explored the relationship between FC in the BGN and variables related to demographics, impulsive behavior, self-paced tasks, mood, and motor correlates in 486 participants in the Whitehall-II imaging sub-study using both region-of-interest- and voxel-based approaches. Age was the only correlate of FC in the BGN that was consistently significant with both analyses. The observed adverse effect of aging on FC may relate to alterations of the dopaminergic system, but no unique dopamine-related function seemed to have a link with FC beyond those detectable in and linearly correlated with healthy aging.

Keywords Functional connectivity · Basal ganglia · Resting state fMRI · Healthy aging · Dopamine · Parkinson’s disease

Introduction
It is well established that in absence of specific tasks, the brain is organised into largely independent resting state networks (RSN) (Smith et al. 2009), detectable using resting state fMRI (rfMRI). The basal ganglia resting state network (BGN), although often not reported as one of the main RSNs, is identifiable, reproducible (across subjects, resting state conditions and imaging parameters), and corresponds with the motor control circuit, opening a new way to investigate the functional connectivity of the basal ganglia (Robinson et al. 2009; Di Martino et al. 2008).

Aberrant functional connectivity (FC) within the BGN has been observed in pathologies with altered neurotransmitter systems and conditions involving motor control and dopaminergic processes in general. In particular, reduced FC has been observed in patients with early Parkinson’s disease (PD) relative to healthy controls (HC) (Szweczyk-Krolkowski et al. 2014; Tan et al. 2015). This alteration appears not to be related to Alzheimer’s disease (Rolinski et al. 2015) but is present in individuals at risk of developing PD (patients with REM sleep behavior disorder) (Rolinski et al. 2016), suggesting that FC determined by rfMRI may be
a promising biomarker for PD. Changes in FC of the BGN have also been observed in other dopamine-related conditions such as depression (Hwang et al. 2016), schizophrenia (Duan et al. 2015), and impulsive behaviors (Schmidt et al. 2002). Therefore, a better characterization of variability of FC in healthy subjects might provide insight into the biological basis of the difference in FC found in pathological conditions. Moreover, identifying factors that account for this variability can also help to increase the specificity of potential biomarkers based on resting state FC. Statistically controlling for factors that are related to normal aging should increase power to detect disease-specific changes.

To this aim we used data from 486 healthy individuals aged 60 to 82 years participating in the Whitehall II imaging (WHII) sub-study (Filippini et al. 2014), which included brain MRI, demographic and health data, and tests of cognitive and motor performance. These data provided sufficient statistical power to investigate the relationship between FC in the BGN and a wide range of variables related to demographics, impulsive behavior, self-paced tasks, mood and motor correlates. To cross-validate findings, we used both a region-of-interest (ROI) and a voxel-wise approach.

Methods

Brain MRI data from 486 participants in the WHII sub-study were analysed (age 69.45 ± 5.23 years, range 60–82 years, M/F = 388/98, education 14.00 ± 3.12 years). Details of the study, the MRI acquisition protocol, and pre-processing steps are described elsewhere (Filippini et al. 2014). Briefly, rfMRI data pre-processing included motion correction, brain extraction, high-pass temporal filtering (cut-off 100 s), field-map correction, artefact removal (FSL-FIX), spatial smoothing (FWHM 6 mm), registration to the individual structural (high-resolution T1-weighted) scans and to standard space (FNIRT, optimized using BBR approach). The template of RSNs used in this study was derived from a separate set of 45 age-matched elderly healthy controls, using group ICA (FNIRT, optimized using BBR approach). The template of high-resolution T1-weighted scans and to standard space (FWHM 6 mm), registration to the individual structural was used to generate subject-specific maps of parameter estimates (PE) for the 50 components (29 RSNs and 21 artefacts) from pre-processed rfMRI data of the 486 subjects. Subject-specific BGN maps entered subsequent FC analyses. Structural T1 images of the individual subjects were segmented using FIRST (Faltenbacher et al. 2011) and the volume of BG structures (caudate nucleus, putamen and globus pallidus) was calculated to be included as covariate.

We tested the potential links between FC in the BGN and different domains using both a region-of-interest (ROI) and voxel-wise approach. Characteristics of the subjects are summarised in Table 1 (see (Filippini et al. 2014) for more details about the tests). The variables included in each domain and the tests performed are summarised in Table 2.

For the ROI analysis, the average PE of the BGN was extracted for each subject within a mask including caudate nucleus, putamen, and globus pallidus, from the Harvard-Oxford probabilistic atlas (threshold 30%) as a measure of FC. The analyses were then performed with SPSS (version 24.0) to evaluate the variance in FC explained by the variables explored in our study. All independent variables entered into the regression models at the same time (enter method or forced entry). Further analyses were performed on the significant predictors on the single BG structures (with Bonferroni correction across structures).

Whole-brain voxel-wise analyses of the BGN maps were performed to investigate the possible relationships between the explanatory variables and FC of the BGN with no restriction to the basal ganglia, to allow exploring possible associations also in cortical areas. Significance of associations were tested using a non-parametric permutation test (randomise, part of FSL (Winkler et al. 2014)) and results were considered significant for p < 0.05 after correction for multiple comparisons using the threshold-free cluster enhancement (TFCE) approach.

Results

In the ROI analysis, the only analysis that led to significant results was the multiple linear regression that included demographic variables. The model explained 4.8% of the variance in the data when looking at all the structures together (adjusted $R^2 = 0.48$, p < 0.001). Age (beta $= -0.212$, p < 0.001) and sex (M > F, beta $= 0.117$, p = 0.014) were significant predictors. Figure 1 illustrates the negative correlation between FC connectivity in the BGN and age. Further investigation of correlations between FC within the single structures and age in the full sample showed significant negative correlation (p < 0.01, Bonferroni corrected across six structures) between age and bilateral caudate (Spearmann’s rho left $= -0.256$, right $= -0.232$) and bilateral putamen (Spearmann’s rho left $= -0.182$, right $= -0.187$). No significant differences (Bonferroni corrected across structures) between men and women were observed (all p > 0.05). When splitting the subjects by gender, the negative
correlation between FC and age remained significant for men (Spearmann’s rho = −0.228, p < 0.001), but not for women (Spearmann’s rho = −0.149, p = 0.144) (Fig. 1). However, the difference between the two correlation coefficients tested using Fisher r-to-z transformation (Myers and Sirois 2006) was not statistically significant (z = −0.72, p = 0.472).

### Table 1 Characteristics of the subjects

| Variable                                      | N (subjects with available data) | Mean     | Std. deviation |
|-----------------------------------------------|----------------------------------|----------|----------------|
| **Demographics**                              |                                  |          |                |
| Age (years)                                   | 486                              | 69.45    | 5.23           |
| Handedness (scale from −24 (left) to 24 = (right) handed) | 482                              | 17.12    | 12.79          |
| Sex (M/F)                                     | 486                              | 388 / 98 |                |
| **Impulsive behavior**                        |                                  |          |                |
| Current smoking (non-smoker / occasional / smoker) | 486                              | 465/3/18 |                |
| Alcohol consumption (units per week)          | 477                              | 15.34    | 15.06          |
| Body mass index                               | 486                              | 26.19    | 4.15           |
| **Self-paced tasks**                          |                                  |          |                |
| Letter fluency (average N)                    | 486                              | 15.69    | 4.57           |
| Categorical fluency (average N)               | 486                              | 22.09    | 5.68           |
| **Mood**                                      |                                  |          |                |
| CES-D                                         | 485                              | 5.22     | 6.21           |
| **Motor performance and sleep**               |                                  |          |                |
| CANTAB RTI MT (average simple Movement Time, msec) | 480                              | 273.52   | 88.42          |
| CANTAB RTI RT (average simple Reaction Time, msec) | 480                              | 317.07   | 79.21          |
| Pegboard assembly task (average N)            | 272                              | 26.06    | 6.13           |
| Pegboard both hands (average N)               | 272                              | 9.80     | 1.78           |
| Pegboard left hand (average N)                | 272                              | 12.11    | 2.07           |
| Pegboard right hand (average N)               | 275                              | 12.29    | 2.00           |
| PSQI 11c (0/1/2/3)§                           | 478                              | 390/48/23/17 |            |

Legend: CES-D = Centre for Epidemiological Studies Depression Scale; CANTAB RTI = Cambridge Neuropsychological Test Automated Battery Reaction Time (CANTAB eclipse 5.0; Cambridge Cognition Ltd. http://www.camcog.com) touchscreen version (MT = movement time for correct responses; RT = mean simple reaction time for correct responses); Pegboard = Purdue pegboard task; § PSQI = Pittsburgh Sleep Quality Index. Sub-item 11c: “How often in the past month have you had legs twitching or jerking while you sleep?” 0 = not during the past month, 1 = less than once a week, 2 = once or twice a week, 3 = three or more times a week.

### Table 2 Details of variables and tests

| Domain                                      | Explanatory variables of interesta (predictors) | Analysis performed | N (subjects with available data) |
|---------------------------------------------|-------------------------------------------------|--------------------|----------------------------------|
| **Demographics**                            | Age, sex, handedness                            | Linear regression  | 482                              |
| **Impulsive behavior**                      | Alcohol consumption, current smoking, Body Mass Index (BMI) | Linear regression | 477                              |
| **Self-paced tasks**                        | Verbal fluency, semantic fluency                | Linear regression  | 486                              |
| **Mood**                                    | CES-D                                           | Partial correlation| 485                              |
| **Motor performance and sleep**             | CANTAB RTI (MT, RT), Pegboard (left, right, both, assembly), PSQI_11c | Linear regression | 270                              |

Legend: CES-D = Centre for Epidemiological Studies Depression Scale; CANTAB RTI = Cambridge Neuropsychological Test Automated Battery Reaction Time (CANTAB eclipse 5.0; Cambridge Cognition Ltd. http://www.camcog.com) touchscreen version (RT = mean simple reaction time for correct responses, MT = movement time for correct responses); Pegboard = Purdue pegboard task; PSQI = Pittsburgh Sleep Quality Index. $a$Volume of BG structures was included as covariate in all (ROI and voxel-wise) analyses.
Similar to the ROI analyses, the voxel-wise analyses showed significant results only when testing demographic variables. Age, the only significant correlate, showed a significant negative correlation with FC in the BGN in the bilateral putamen (clusters’ peak MNI coordinates (x,y,z): left=(−24,12,−4); right=(28,6,−6)), extending into bilateral caudate nucleus, bilateral thalamus, and bilateral amygdala (Fig. 2). No other significant associations were found with the remaining variables/domains (all statistical maps are available at: https://neurovault.org/collections/2681).

Discussion

The aim of this exploratory study was to better understand the variability of functional connectivity within the basal ganglia resting state network in healthy older adults, with particular focus on potential links to dopamine-related function. We investigated the association between FC and variables related to demographics, impulsive behavior, self-paced tasks, mood, and motor variables in a large cohort of 486 participants in the Whitehall-II imaging sub-study. We found that age was the strongest correlate of FC in the BGN. In particular, a significant negative correlation between age and FC was observed in several basal ganglia structures in both the ROI and voxel-wise analyses.

It is known that aging affects FC in general (Biswal et al. 2010) and that FC in the default mode network decreases in older age (Andrews-Hanna et al. 2007; Damoiseaux et al. 2008). Regarding the BGN, the
The relationship between FC and age described by (Sole-Padulles et al. 2016) and (Allen et al. 2011) was positive rather than negative. However, the age ranges in these two studies (7–18 and 12–71 years respectively) were very different from ours, which focused on elderly subjects (60–82 years). In this framework, our results add complementary information to the current literature, suggesting an inverted U-shaped pattern of FC of the BGN across the life span.

The effect of age on the BG has been previously shown with PET and SPECT. The review by (Reeves et al. 2002) showed an association between age and loss of striatal dopamine transporters (DATs) in the caudate and putamen in both hemispheres (Wong et al. 1984; van Dyck et al. 2002). Those findings suggest that the changes underlying the negative effect of aging on the FC within the BGN may be related to biological alterations of the dopaminergic system, like those involved in PD. In fact, previous studies using SPECT (Ba and Martin 2015) and fMRI (Szewczyk-Krolikowski et al. 2014; Rolinski et al. 2015) found differences in idiopathic PD compared to controls within the BG in terms of DATs and FC. These areas are the same that show a negative correlation with age in our sample of healthy aging participants.

It needs to be acknowledged that the subjects are from a cohort of civil servants recruited in 1985 (Marmot and Brunner 2005) and therefore not entirely representative of the general population. In particular, the observed correlation with age that seems to be driven by males, would need further investigation on a more balanced sample, due to the gender bias in this cohort.

Regarding the relationship between FC and other dopamine-related behavioral data, none of the tested domains showed significant correlations with FC in the BGN. This could be due to multiple reasons: on one hand, the variables we tested were selected because they can be influenced by dopamine, but, since they reflect also other functions, may not be strictly related to the dopaminergic changes in the BG. On the other hand, the variability we observed in FC in the BG is possibly a sum of dopamine-related and dopamine-unrelated processes and therefore not strictly related to the dopamine-related behavioral data. Aging is a process that incorporates multiple domains, including dopamine-related changes, and might better explain the FC variability observed in the BGN with fMRI.

In light of these results, our findings might have implications for the development of imaging biomarkers, for example for the detection of PD, which must have age-norms to be maximally useful. In fact, since age accounts for some of the spread in FC of healthy subjects, statistically controlling for its effect might increase the specificity of a biomarker based on BGN functional connectivity.

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Compliance with ethical standards

Conflict of interest The authors declare that they have no conflict of interest.

Ethical standards All procedures followed were in accordance with the ethical standards of the responsible committee on human experimentation (institutional and national) and with the Helsinki Declaration of 1975, and the applicable revisions at the time of the investigation.

Informed consent Informed consent was obtained from all individual participants included in the study.

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