Executive functions and prefrontal cortex: a matter of persistence?

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INTRODUCTION

High-level cognitive functions have traditionally been localized to the anterior frontal association cortex, commonly referred to as the prefrontal cortex (PFC). Numerous studies have expanded upon this static view of executive localization, identifying functional networks activated by complex executive processes and served by extensive reciprocal connections between the PFC, posterior parietal cortex and various cortical and limbic regions ( Selemon networks activated by complex executive processes and served this static view of executive localization, identifying functional prefrontal cortex (PFC). Numerous studies have expanded upon the anterior frontal association cortex, commonly referred to as the associative prefrontal cortex. High-level cognitive functions have traditionally been localized to the anterior frontal association cortex, commonly referred to as the prefrontal cortex (PFC). Numerous studies have expanded upon this static view of executive localization, identifying functional networks activated by complex executive processes and served by extensive reciprocal connections between the PFC, posterior parietal cortex and various cortical and limbic regions ( Selemon networks activated by complex executive processes and served this static view of executive localization, identifying functional prefrontal cortex (PFC). Numerous studies have expanded upon the anterior frontal association cortex, commonly referred to as the associative prefrontal cortex.

Executive function is thought to originate from the dynamics of frontal cortical networks. We examined the dynamic properties of the blood oxygen level dependent time-series measured with functional MRI (fMRI) within the prefrontal cortex (PFC) to test the hypothesis that temporally persistent neural activity underlies performance in three tasks of executive function. A numerical estimate of signal persistence, the Hurst exponent, postulated to represent the coherent firing of cortical networks, was determined and correlated with task performance. Increasing persistence in the lateral PFC was shown to correlate with improved performance during an n-back task. Conversely, we observed a correlation between persistence and increasing commission error – indicating a failure to inhibit a prepotent response – during a Go/No-Go task. We propose that persistence within the PFC reflects dynamic network formation and these findings underline the importance of frequency analysis of fMRI time-series in the study of executive functions.

Key words: executive function, prefrontal cortex, persistence, BOLD, Hurst exponent, networks, functional MRI

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values of $H$ and persistence have come to indicate more complex and coordinated dynamics in the underlying network activities and degradation of fractal complexity has come to indicate the desynchronization of the biological system and reflect malfunction, aging or disease (Lipsitz and Goldberger, 1992; Goldberger et al., 2002; DeKosky and Marek, 2003; Salkovic-Petrisic et al., 2009).

The frequency analysis of fMRI time-series has reported interesting although sometimes discordant results. Wink et al. (2006) demonstrated a direct association between aging and scopolamine administration with the $H$ in resting fMRI but subsequently demonstrated an association between higher persistence and faster response in a prior encoding task (Wink et al., 2008). The paradoxical results were explained by a complex multifractal analysis of the previous data, although the original result of increased $H$, equivalent to the persistence in the time-series, was confirmed. Aging, disease and pharmacological agents, however, alter brain vascular reactivity and perfusion (Estrada et al., 1983; Fukuyama et al., 1996; Farkas and Luiten, 2001) that will act as confounder in the BOLD frequency response.

The general aim of this work was to build on the evidence described above and examine in detail the association between coherence in neuronal firing in the frontal cortices, as reflected in the average frequency content of fMRI signal and direct behavioral measures of executive function tasks in normal volunteers.

Our primary hypothesis was that estimates of the $H$ from BOLD activity, obtained using standard wavelet scalograms (Bullmore et al., 2004) from within the lateral PFC were associated with performance in three tasks of executive function ($n$-back, Go/No-Go, and Tower of London). Additionally, we investigated persistence in a number of associated cortical regions, commonly activated as part of an “executive” network during these tasks.

**MATERIALS AND METHODS**

**PARTICIPANTS**

Forty-six healthy, right-handed participants aged 19–59 (18 male; mean age = 35.1 years, SD = 11.6) were included. All participants were free of physical illness and had no history of psychiatric or neurological disorder as assessed using a physical examination, medical history and a structured interview (First and Pincus, 2002). The study was conducted with ethical approval from Hammersmith Hospital Research Ethics Committee, and all participants gave written, informed consent.

**TASK DESIGN**

The tasks performed in the fMRI scanner were all programmed with E-Prime v1.1 (Psychology Software Tools, Inc., Pittsburgh, PA, USA), presented using the Integrated Functional Imaging System (IFIS-SA, Invivo, Orlando, FL, USA), and completed during the same fMRI session, in a counter-balanced order.

$n$-back

During this working memory task, a dot was presented in one of four spatial locations, corresponding to a particular button on a response pad and arranged horizontally across the screen (Gevins and Cutillo, 1993). Participants were required to either press the button that corresponded to where the dot was currently appearing (0-back), where the dot had appeared one trial previously (1-back) or where the dot had appeared two trials previously (2-back). Dots were presented for 500 ms followed by a blank screen for 1500 ms, with 12 dots per block. Three 0-back blocks, three 1-back blocks and three 2-back blocks, each lasting 24 s, were presented in a pseudo-random order, interspersed with two blocks of a low-level rest condition (fixation). Instructions were presented for 3 s prior to the start of each block. Each subject underwent two sets of $n$-back blocks during the imaging session to test the robustness of any findings. One subject failed to complete the second set of blocks and was included in the analysis using data from the first set only. Response accuracy (proportion of correct responses) and response time were used as measures of performance.

Go/No-Go

This task of response inhibition consisted of ten 36-s blocks of two conditions: “Go,” requiring a response to every stimulus and “No-Go,” requiring a response to all but the “stop” stimulus. Five Go blocks and five No-Go blocks were presented in a pseudo-random order, interspersed with two blocks of a rest condition (fixation) after the third and seventh blocks. Go stimuli (the letters F, H, K, P, and S) were presented every 3 s for 500 ms during the blocks. The No-Go stimulus (the letter V) was substituted for a Go stimulus in 45% of trials in the No-Go blocks. Number of commission (responses to a No-Go signal) and omission (no response to a Go signal) errors and response time to Go signals were used as a measure of performance.

Tower of London

This planning task required participants to determine the minimum number of moves to re-arrange three colored balls from a start position to a target position (Owen et al., 1996). Participants were presented with a display containing an upper “target” section that indicated the final position of the colored balls, and a lower “start” section that indicated the start position of the colored balls and were then asked to indicate the minimum number of moves via a single finger keypad response. During the “Control” block participants were requested to count the number of balls in both the start and target pockets. Four “Easy” blocks (comprising two and three move problems), four “Difficult” blocks (comprising four to six moves), and four “Control” blocks were presented in a pseudo-random order, interspersed with three blocks of a rest condition (fixation). Each block lasted 30 s and each trial was self-paced. Instructions were presented for 3 s prior to commencement of each block. Response accuracy (proportion of correct responses) and response time to correctly solved problems were used as measures of performance.

**IMAGING PARAMETERS**

Functional MRI was performed with a Philips Intera 3 Tesla MRI scanner. Functional T2*-weighted images were acquired using gradient-echo echoplanar imaging, with an automated higher order shim procedure (SENSE factor 2; TE 30 ms; TR 3000 ms; flip angle 90°; FOV 280 mm; voxel dimensions 2.2 mm × 2.2 mm × 2.75 mm). Images were acquired in 48 contiguous 2.75 mm axial slices per brain volume. The total number of brain volumes acquired for each of these tasks was 218, 154, and...
The first five volumes of each scan were discarded to account for T1 equilibration effects. Functional images were acquired during two runs of 5 min 27 s (n-back), a single run of 7 min 42 s (Go/No-Go) and one of 8 min 27 s (Tower of London). A high-resolution T1-weighted TFE structural scan was also acquired for each participant for subsequent high-resolution image registration (TE 4.6 ms; TR 9.7 ms; Flip angle 8°; FOV 240 mm; voxel dimensions 0.94 mm × 0.94 mm × 1.2 mm).

**IMAGE ANALYSIS**

Initial processing of functional data sets was carried out using FSL (FMRIB's Software Library, http://www.fmrib.ox.ac.uk/fsl; Smith et al., 2004). Each 4D functional volume was motion corrected, brain-extracted, spatially smoothed with a Gaussian kernel of FWHM 5 mm and grand-mean intensity normalized.

Given the importance of the PFC to executive function, mean BOLD time-series collected during each task were extracted from three bilateral regions of interest (ROI), defined by the Harvard–Oxford Cortical Structural Atlas and chosen to represent the lateral PFC (middle frontal gyrus; inferior frontal gyrus pars opercularis, and pars triangularis; Table 1, shown in Figure 1A) to test the primary hypothesis that executive function task performance is associated with persistence in the BOLD activity within the lateral PFC.

Many studies have described functionally connected networks comprising several common structures and engaged by executive task demands regardless of the specific task involved (Menon et al., 2001; Schall et al., 2003; Owen et al., 2005; Seeley et al., 2007; Stevens et al., 2007, 2009; Simmonds et al., 2008). In addition to the primary ROI defined above, BOLD time-series were also extracted from several ROI previously identified as part of the executive networks (anterior and posterior cingulate cortex, cuneus, precuneus, medial frontal cortex, parahippocampal gyrus, insula, and superior parietal cortex, Figure 1B and Table 1) to investigate BOLD signal persistence outside of the PFC.

Regional time-series were obtained by registering standard space ROI masks to the filtered functional datasets via individual high-resolution T1-weighted structural images with a 12 degrees of freedom affine registration using FSL's linear registration tool (FLIRT) and then averaging the pixel time-data within the native-space masks. In addition, group average BOLD activation maps were generated using FSL’s FEAT. After pre-processing, data were entered into a GLM to determine where BOLD activation occurred during each task condition (Contrasts: n-back: 0-back > Rest; 1-back > Rest; 2-back > Rest; Go/No-Go: Go > Rest; No-Go > Rest; Tower of London: Easy > Rest; Difficult > Rest) on average across the group. Z-statistic images were thresholded at Z = 2.3 and cluster-corrected across the whole brain using family-wise error (FWE) with a significance threshold of p < 0.05 (see Figure 2). To investigate region-specific changes in BOLD in ROI where persistence correlated significantly with executive task performance, the mean BOLD signal change (%) associated with parameter estimates from the modeled experimental paradigm within each ROI was estimated using FSL's FEATquery.

### Table 1 | Center of gravity (COG) in MNI coordinates of cortical regions of interest.

| Lateral prefrontal cortex | COG       |
|--------------------------|-----------|
|                          | x   y   z |
| Middle frontal gyrus     | -38.0 19.0 41.4 |
| Left                     | 38.8 | 19.2 | 42.1 |
| Right                    | -50.1 14.4 16.5 |
| Inferior frontal gyrus   | -48.1 28.4 9.0 |
| pars opercularis         | 51.3 | 15.5 | 16.5 |
| Right                    | -29.2 -48.9 57.5 |
| Inferior frontal gyrus   | 50.0 | 28.3 | 8.2 |
| pars triangularis        | -23.1 -8.9 30.2 |
| Middle frontal gyrus     | -37.7 | 3.2 | -0.6 |
| Right                    | -23.3 | -77 | -30.5 |
| Insula                   | -36.7 | 1.7 | -0.1 |
| Parahippocampal gyrus    | -23.1 -8.9 30.2 |
| Right                    | 29.2 | -476 | 58.8 |
| Superior parietal cortex | 0.7 | 19.3 | 24.4 |
| Anterior cingulate cortex| 0.8 | -371 | 30.1 |
| Posterior cingulate cortex| 1.5 | -78.2 | 27.4 |
| Cuneus                   | 0.9 | -58.7 | 38.1 |
| Precuneus                | 0.1 | 43.7 | -170 |

**Numerical Estimates of Persistence**

For some class of models, persistence (P) of a time-series can be defined as the ratio between the energy at low vs. high frequencies. By this definition, a signal that persists longer has longer memory and displays a larger numerical ratio than a signal with lower persistence. The non-linear and complex organization of the human brain and its internal dynamics are reflected in output signals at the microscopic, mesoscopic and macroscopic level that are generally fractal and exhibit decay in energy that is proportional to 1/f. For a complete review on the topic see Werner (2010).

Numerical estimates of persistence in fractal time-series can be determined using the H, a number defined in the [0–1] range where values of H = 0.5 indicate a purely random process, values in the [0–0.5] interval indicate antipersistent, mean-reverting behavior and values in the [0.5–1] range indicate persistent/long-memory processes. The estimation of H is far from trivial and requires an understanding of the generating process (see review...
non-stationary but the class of ownership of a signal must be determined so that the proper estimator can be used. This can be tested by plotting its power spectrum and measuring its rate of decay. 1/f noises with \(-1 < \beta < 1\) and \(1 < \beta < 3\) are almost identical to fGn and fBm signals, respectively (Beran, 1994). Figure 3 shows the plotted power spectrum of a typical regional fMRI BOLD time-course that demonstrates a rate of decay \((-2\)\) that assigns the signal to the fBm class. For this class, time–frequency estimators of \(H\) based on the wavelet transform are optimal (Eke et al., 2002). In this work, estimates of \(H\) were obtained using standard wavelet scalograms (Bullmore et al., 2004), in the simple implementation of Jones et al. (1996) and robust within the sampling rate available. In detail, for each time-series, the wavelet transform uses a sequential bank of filters to extract the energy of sequential frequency bands, from the high- to the low. The wavelet scalogram then plots the \(\log\) of the square of the average of the wavelet coefficients for each band against the \(\log\) of ordinal number of each band (an integer \(N = 1, 2, 3, \ldots\) indexing the bands from the highest frequency band). The \(H\) is calculated as \(H = (S – 1)/2\) where \(S\) is the slope of the scalogram. Here, the algorithm to calculate \(H\) was written in Matlab (The MathWorks Inc., Natick, MA, USA), version R2008b, where the scalogram was obtained using wavelet packets and the Haar wavelet as described in Jones et al. (1996). Six wavelet bands were used covering the 0.1667–0.005 Hz frequency range. An example of wavelet scalogram for the same regional time-series is illustrated in Figure 3.

**Results**

**n-back Behavior**

Forty-six participants completed the first \(n\)-back set of blocks and scored 89.7 ± 8.9, 88.3 ± 10.1 and 70.4 ± 20.3% correct for the 0-, 1-, and 2-back tasks respectively. Increasing task difficulty had a significant effect on response accuracy (ANOVA, \(p < 0.001\) and response latency (\(p < 0.05\)). Mean response time was 640 ± 157, 532.9 ± 232, and 542 ± 254 ms for each difficulty level. Forty-five participants successfully completed a second set of \(n\)-back blocks, response accuracy (0-back: 92.4 ± 6.4%; 1-back: 86.7 ± 12.4%; 2-back: 70.7 ± 20.4%) and response time (0-back: 642 ± 158 ms; 1-back: 553 ± 253 ms; 2-back: 534 ± 300 ms) did not differ significantly between the first and second sets.

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Figure 2 | Group BOLD activation during each task of executive function. Group average BOLD activations during the difficult condition (Difficult > Rest) of the Tower of London task, the 2-back (2-back > Rest) condition of both \(n\)-back sets and the No-Go (No-Go > Rest) condition of the Go/No-Go task are shown in MNI152 standard space in radiological convention (whole-brain cluster corrected, \(Z = 2.3, p < 0.05\)).
Faster response time during the first \( n \)-back set was correlated with higher \( H \) across the lateral PFC. Repeated measures ANOVA revealed a main effect of \( H \) on response time (Table 2) in the left inferior frontal gyrus pars opercularis, pars triangularis, and middle frontal gyrus and in the right pars triangularis and middle frontal gyrus (all \( p < 0.05 \)). Response accuracy did not significantly correlate with \( H \) in the lateral PFC.

Post hoc analysis (Table 3) revealed significant correlations between \( H \) during the whole \( n \)-back set and individual response times during 0-back blocks in the left inferior frontal gyrus pars opercularis and middle frontal gyrus and the right middle frontal gyrus (all \( p < 0.05 \)), during the 1-back blocks in the left pars opercularis (\( p < 0.01 \)), pars triangularis (\( p < 0.05 \)) and middle frontal gyrus (\( p < 0.01 \)) and in the right pars triangularis (\( p < 0.05 \)) and middle frontal gyrus (\( p < 0.01 \)) and during the 2-back blocks in the left pars triangularis and middle frontal gyrus (both \( p < 0.05 \)).

During the second \( n \)-back set, the overall pattern of correlations between \( H \) and response time was similar, particularly in the left PFC (Figure 4), although fewer statistically significant correlations were observed (Table 4). A main effect of \( H \) on response time was present at a trend level in the left pars opercularis and pars triangularis (both \( p = 0.07 \)) and significant in the left middle frontal gyrus (\( p = 0.05 \)). No significant main effects were observed in the right hemisphere and no significant correlations were observed between \( H \) and response accuracy. Post hoc analysis revealed significant correlations between \( H \) in the left middle frontal gyrus and response time during the 0-back (\( t = -2.38, p = 0.022 \)) and 1-back (\( t = -2.08, p = 0.044 \)) but not 2-back blocks (Figure 4).

Blood oxygen level dependent signal change was examined where significant correlations between \( H \) and \( n \)-back performance were observed, i.e., bilaterally in the middle frontal gyrus and in the left inferior frontal gyrus pars opercularis. Group average BOLD activations during the 2-back condition (2 > Rest contrast; \( Z > 2.3, \)
Table 3 | *Post hoc* analysis of significant correlations between *H* in the lateral PFC and *n*-back response time.

| Region                  | 0-Back | 1-Back | 2-Back |
|-------------------------|--------|--------|--------|
|                         | *t*    | *p*    | *t*    | *p*    |
| Left pars opercularis   | −2.32  | 0.025* | −2.93  | 0.005**| −1.53  | 0.134  |
| Left middle frontal gyrus | −2.15  | 0.037* | −2.72  | 0.009**| −2.1   | 0.042* |
| Right middle frontal gyrus | −2.06  | 0.045* | −2.93  | 0.005**| −1.79  | 0.080  |

*p < 0.05; **significance passing correction for multiple testing at *p* < 0.05.

FWE-corrected *p* < 0.05) are shown in Figure 2. Positive significant correlations between BOLD signal change and *n*-back response time were observed in the left pars opercularis, and middle frontal gyrus during the 0-back (both *p* < 0.05) and 1-back (both *p* < 0.01) but not 2-back blocks. In the right hemisphere significant correlations were observed in the middle frontal gyrus during the 0-back blocks (p < 0.05) and the 1-back blocks (p < 0.01). Table 5 shows BOLD and response time correlations within the PFC, notably, the observed significant correlations were all opposite in direction to the correlations with *H*. During the second *n*-back set, BOLD signal change in the left middle frontal gyrus did not correlate with response time during any block (0-back: *r* = 0.28, *p* = 0.063; 1-back: *r* = 0.16, *p* = 0.302; 2-back: *r* = 0.05, *p* = 0.752).

Outside of the PFC, no correlations between *H* and response time during the first or second *n*-back sets were observed. A significant main effect of *H* on response accuracy during the first but not the second *n*-back task was seen in the posterior cingulate (*F*<sub>(1,44) = 5.09, *p* = 0.029) and bilaterally in the anterior parahippocampal gyri (Left: *F*<sub>(1,44) = 7.89, *p* = 0.007; Right: *F*<sub>(1,44) = 5.29, *p* = 0.026) but these effects did not pass correction for multiple testing.

**Go/No-Go Behavior**

Forty-three participants successfully completed the Go/No-Go task; of these, 34 made commission errors (responded to a STOP signal). Within this group, the median number of commission errors (range) was 3 (1–8). Errors of omission (failing to respond to a Go signal) were committed a median (range) of 2 (0–8) times across the whole group. Mean response time to Go signals was 447.3 ± 97.9 ms.

**Imaging**

High *H* in the lateral PFC correlated with increased commission errors made during the Go/No-Go task. Pairwise correlations between *H* in each PFC region over the course of the task and the number of commission errors committed during the No-Go blocks revealed significant associations bilaterally in the pars opercularis (Left: *p* < 0.05; Right: *p* < 0.01) and middle frontal gyrus (Left: *p* < 0.05; Right: *p* < 0.01) and in the left pars triangularis (*p* < 0.05; Table 6). Figure 5 illustrates this association in the left and right middle frontal gyri. Omission error and response time to Go signals were not significantly correlated with *H* in any region of the lateral PFC. BOLD signal change during No-Go blocks in the lateral PFC did not correlate with commission error, group activations are shown in Figure 2 (Contrast: No Go > Rest, *Z* > 2.3, FWE-corrected *p* < 0.05). Outside of the PFC, increasing commission error was only correlated with *H* in the precuneus (Spearman’s *rho* = 0.375, *p* = 0.013) but this effect did not survive multiple comparison correction.

**TOWER OF LONDON Behavior**

Forty-five participants completed the Tower of London task successfully. As expected increasing task difficulty was associated with increasing response accuracy and increasing response latency (ANOVA, *p* < 0.001). Participants scored (mean ± SD) 91.1 ± 6.3, 62.4 ± 15.4, and 29.5 ± 17.4% correct for the control, easy (two and three moves) and difficult (four to six moves) tasks, respectively. Mean response time was 1647 ± 379, 5957 ± 1868, and 11611 ± 4165 ms, respectively.

**Imaging**

No significant correlations between *H* in the lateral PFC and response time during the Tower of London task were observed. A main effect of *H* on response accuracy only approached significance in the left middle frontal gyrus (*F*<sub>(1,44) = 3.01, *p* = 0.09) and the right pars opercularis (*F*<sub>(1,44) = 3.36, *p* = 0.074).

Outside of the lateral PFC, a single significant correlation between *H* and response accuracy was observed in the right insula (*F*<sub>(1,44) = 4.79, *p* = 0.034) but this effect did not pass correction for multiple testing.

**PERSISTENCE ACROSS TASKS**

We observed that *H* within the middle frontal gyri correlated with performance during both the *n*-back and Go/No-Go tasks and achieved trend-level significance during the Tower of London task. We investigated if *H* within the middle frontal gyri remained consistent across each task.

The correlations between *H* within the middle frontal gyri during each task are shown in Tables 7 and 8. In the left hemisphere (Table 7) *H* during the Tower of London task correlated with *H* during the first (*p* < 0.01) and second (*p* < 0.05) *n*-back sets and the Go/No-Go task (*p* < 0.05). Only *H* during the first *n*-back set and the Go/No-Go task did not significantly correlate. In the right hemisphere (Table 8), significant correlations were observed between *H* during the Tower of London task and the first (*p* < 0.01) and second (*p* < 0.01) *n*-back sets and between *H* during the second *n*-back set and the Go/No-Go task (*p* < 0.01). Correlations between *H* during the first and second *n*-back sets and between the Tower of London and Go/No-Go tasks approached significance (*p* = 0.064 and 0.083, respectively).

**DISCUSSION**

Using fMRI, we have demonstrated a correlation between performance of executive function tasks and the dynamic properties of their underlying BOLD time-series within the lateral PFC. Traditionally, the study of behavioral correlates have focused on transient events, here we demonstrate that BOLD signals sustained across time are also functionally relevant. Increasing persistence, *H*, postulated to represent the translation of...
The n-back task requires sustained attention with increasing working memory load at the most difficult levels. An extensive working memory network comprising the dorsolateral and mid-ventrolateral PFC, parietal cortex, medial cerebellum and anterior, and dorsal cingulate cortex has been well documented (Owen, 1997; Owen et al., 2005). Here we confirm the presence of this network and present significant correlations between increasing n-back performance and faster response time in the middle PFC. Specifically, we observed correlations between H and temporally persistent and coherent firing of cortical neurons into a smoother BOLD signal, correlated with improved performance during n-back tasks. Conversely, and perhaps surprisingly, we also observed a correlation between H and increasing commission error – indicating a failure to inhibit a prepotent response – during a Go/No-Go task. Importantly, H was consistent across tasks suggesting that persistence may reflect a hard-wired property of neuronal networks projecting into the PFC.

**n-BACK**

The n-back task requires sustained attention with increasing working memory load at the most difficult levels. An extensive working memory network comprising the dorsolateral and mid-ventrolateral PFC, parietal cortex, medial cerebellum and anterior, and dorsal cingulate cortex has been well documented (Owen, 1997; Owen et al., 2005). Here we confirm the presence of this network and present significant correlations between increasing H and faster response time in the n-back task across the lateral PFC. Specifically, we observed correlations between H...
Table 4 | Main effects of H on response time during the second n-back set.

| Region                          | R(1,44) | p     |
|---------------------------------|---------|-------|
| Left pars opercularis           | 3.40    | 0.072 |
| Left pars triangularis          | 3.58    | 0.065 |
| Left middle frontal gyrus       | 4.06    | 0.050*|
| Right pars opercularis          | 0.01    | 0.932 |
| Right pars triangularis         | 0.47    | 0.499 |
| Right middle frontal gyrus      | 0.54    | 0.047 |

*p < 0.05.

Table 5 | Post hoc analysis of significant correlations between BOLD signal change in the lateral PFC and n-back response time.

| Region                          | 0-back r | 1-back r | 2-back r | 0-back p | 1-back p | 2-back p |
|---------------------------------|----------|----------|----------|----------|----------|----------|
| Left pars opercularis           | 0.31     | 0.46     | 0.20     | 0.038*   | 0.001*   | 0.187    |
| Left middle frontal gyrus       | 0.37     | 0.47     | 0.14     | 0.012*   | 0.001*   | 0.371    |
| Right middle frontal gyrus      | 0.31     | 0.47     | 0.11     | 0.034*   | 0.001*   | 0.452    |

*p < 0.05.

Table 6 | Correlation between number of commission errors and H in the PFC during the Go/No-Go task.

| Region                          | Correlation with commission errors |
|---------------------------------|-----------------------------------|
|                                | rho†                              |
|                                 | p                                 |
| Left pars opercularis          | 0.32                              | 0.036*|
| Left pars triangularis         | 0.35                              | 0.022**|
| Left middle frontal gyrus      | 0.35                              | 0.021**|
| Right pars opercularis         | 0.49                              | 0.001**|
| Right pars triangularis        | 0.20                              | 0.210 |
| Right middle frontal gyrus     | 0.44                              | 0.003**|

*Spearman’s rank correlation was used due to non-normally distributed behavioral data.
 apologized for multiple testing at p < 0.05.

Go/No-Go

Surprisingly, H correlated with increased commission error during the Go/No-Go task. Correlations were observed bilaterally in the middle frontal gyrus, inferior frontal gyrus pars opercularis and pars triangularis with the strongest found in the right middle frontal gyrus and right pars opercularis. The right PFC is commonly associated with response inhibition, activating as part of a right-lateralized network during No-Go trials (Simmonds et al., 2008), although a similar role has also been demonstrated in the left hemisphere (Swick et al., 2008).

Ettinger et al. (2008) recently described how the successful inhibition of a reflexive rapid eye movement (saccade) in favor of a volitional response (antisaccade) relates to the inferred “plasticity” of the cognitive process hypothesized to underlie it. Both the saccade/antisaccade and Go/No-Go task require the inhibition of an unwanted response and the PFC—predominantly the right inferior frontal gyrus—is part of a network active during response inhibition, independent of modality (Chikazoe et al., 2007; Leung and Cai, 2007). Electrophysiological evidence suggests that coherent firing of groups of neurons across both spatial and temporal domains within the PFC occurs in the gamma frequency range and this synchrony is thought to underlie performance in tasks of executive performance (Howard et al., 2003; Sederberg et al., 2003; Gregoriou et al., 2009). Our observations support this hypothesis and further evidence indicates that a transient suppression of gamma band synchronization occurs within the ventrolateral PFC during a high-level cognitive task (Lachaux et al., 2007) and more specifically, selective desynchronization in the gamma band has been observed during the No-Go condition of a Go/No-Go task (Harmony et al., 2009). While these observations may seem
have attempted to delineate structures underlying the cognitive sub-processes involved (Lazeron et al., 2000; Rowe et al., 2001; Newman et al., 2003; Schall et al., 2003). Overall, they concluded that PFC regions were primarily involved in the generating, selecting and remembering appropriate moves over the evaluation of specific solutions.

We confirm this widespread network of activation during the Tower of London task but were not able to determine if H in the PFC was associated with task performance. It is possible that we were not able to dissociate this relationship due to the complexity and self-paced nature of the task, compared to the relatively simple n-back and Go/No-Go tasks. The Tower of London is a multifaceted task and it has been suggested that classifying task difficulty based on the minimum number of moves to solve each problem is inadequate to capture the full cognitive load of doing so correctly (Berg et al., 2010). Variation in goal hierarchy, search depth and solution paths between problems of the same difficulty can potentially influence performance and task-related PFC BOLD activation during the Tower of London task, potentially confounding our observations (Unterrainer et al., 2006; McKinlay et al., 2008; Newman et al., 2009). Therefore, while the dynamics of BOLD signaling in the PFC may be a marker of the underlying processes that result in successful planning, the present study design may not have been appropriate to capture the necessary information.

Connectivity analysis of fMRI data has been used to construe and test the hypothesis that persistent neuronal activity sustaining executive programs is enacted not just by specific properties of the PFC but is an emerging property of complex brain networks (Dosenbach et al., 2007; Seeley et al., 2007). As such, persistence should be intrinsic to the brain and therefore measurable by fMRI paradigms not just in case of executive paradigms but also at rest (Dosenbach et al., 2007; Seeley et al., 2007).

Other investigators have examined the same ideas with mixed blocked/event-related methods designed to extract sustained signals (Chawla et al., 1999; Braver et al., 2003; Dosenbach et al., 2006; Cole and Schneider, 2007). In addition, meta-analytic and more theoretical approaches have also been brought to bear on
These ideas (Dosenbach et al., 2008; Cole et al., 2010; Duncan, 2010) and the idea of trade-offs between sustained and momentary executive signals has been brought to bear on such network analyses by Dosenbach et al. (2007) and Seeley et al. (2007). Here, we have demonstrated that $H$ in the lateral PFC remains relatively consistent across three tasks of executive function and may mirror a stable dynamic network conformation and represent, at least in part, an aspect of “hard-wired” cortical dynamics in individuals.

It was interesting and surprising to note that the linear correlation observed between $H$ and performance was positive in the $n$-back task and negative in the Go/No-Go. This could be interpreted as persistence benefiting tasks requiring sustained attention to the detriment of tasks requiring rapid adaptive changes in behavior. Converging electrophysiological (Funahashi et al., 1989), neurochemical (Goldman-Rakic et al., 2000), computational (Durstewitz et al., 2000a,b) and functional (Nolan et al., 2004; Winterer et al., 2006; Muller et al., 2007) evidence suggests dopamine-mediated fluctuations in PFC neuronal activity contribute to a stable or flexible network state that benefit task requiring sustained attention and adaptive changes in behavioral set, respectively (Bilder et al., 2004). Indeed, performance in both the $n$-back and Tower of London tasks is sensitive to endogenous dopamine levels (Mehta et al., 2005; Reeves et al., 2005; Apud et al., 2007). By contrast, stable network formation may prove detrimental to tasks that require flexible changes in behavioral set, such as the Go/No-Go. This apparently antagonistic nature of PFC activity on executive performance has been described elsewhere (Durstewitz and Seamans, 2008) and previous studies have demonstrated the role of dopamine in performance in tasks requiring cognitive stability at a cost to attentional set-shifting and response flexibility (Mehta et al., 2004; Nolan et al., 2004; Dreisbach et al., 2005; Stefanis et al., 2005). Additionally, recent EEG findings have demonstrated associations between opposing dynamics of frontal networks and intelligence (Thatcher et al., 2005, 2008). However, it is important to consider that these conclusions are limited to a cohort of normal controls, and to the performance range measured. It is known that in pathological conditions, poor performance in the $N$-back task is usually associated with poor performance in the Go/No-Go task (Verdejo-Garcia et al., 2006), so it is possible that optimal performance in executive tasks is associated with persistence in a non-linear relation, possibly U-shaped, where both signal content at low and high frequency combine (Arnsten, 1997; Zahrt et al., 1997; Williams and Castner, 2006). Further investigations, possibly combined with measures of the underlying receptor systems are thus warranted.

METHODOLOGICAL CONSIDERATIONS

No correlations with $H$ were seen outside of the PFC during the tasks after correction for multiple testing. Abundant connections exist between the PFC and the rest of the cortex, and delay-period neuronal firing is not unique to PFC neurons (Miller et al., 1993; Constantinidis and Steinmetz, 1996). The process is, however, most prominent and robust in lateral PFC neurons which may explain our observations outside of the PFC (Quintana and Fuster, 1999). It must also be considered that the lack of observed correlations outside of the PFC may be due to low sensitivity and anatomical specificity of the current method and not an intrinsic property of non-PFC regions.

We also tried to determine if the observed relationships between $H$ and performance were influenced by changes in BOLD activity. BOLD activation patterns can be characterized by a higher persistence and previous studies have demonstrated task-specific increases in cerebral blood flow and BOLD activation in the PFC (Beauchamp et al., 2003; Schall et al., 2003; Thurner et al., 2003). Our results demonstrate that $H$ is not directly associated with mean BOLD signal change associated with each task block. In the Go/No-Go task, the % change of BOLD signal did not predict performance in regions where $H$ appeared to do so. In the first $n$-back tasks, BOLD and $H$ in the lateral PFC predicted performance. However, while $H$ had a direct correlation with performance (i.e., a negative correlation with response time), BOLD signal change was inversely correlated with performance (i.e., a positive correlation with response time). We postulate that this aligns with previous suggestions that BOLD activity in the PFC during this task indexes neuronal tuning, with greater activity thus being associated with less efficient processing and slower or poorer performance (Mattay et al., 2006; Mier et al., 2009). However, due to fundamental differences in the nature of the two analyses, it is difficult to provide a direct comparison between them. In order to achieve parity, BOLD signal change was calculated within the atlas-defined ROI used to estimate $H$, and not from GLM-based activations. This may have resulted in an under-estimation of the BOLD response in these regions and con-founded our observations. Note, however, that this information can be captured more robustly from $H$ rather than BOLD signal change, in this instance. The results observed are also in line with the observed negative correlation of the latency of response in a fame decision/facial encoding task with the $H$ of resting-state data acquired after task performance that was reported by Wink et al. (2008).

Finally, the sensitivity of the estimator of $H$ limited the application of the analysis of large regions with potential loss of anatomical specificity. An extended effort will be undertaken in the future to code novel and more robust $H$ estimators with lower noise sensitivity to allow the pixel-by-pixel mapping of $H$.

CONCLUSION

We present correlations between $H$, postulated to represent long-memory processes of persistent, coherent neuronal firing within the lateral PFC, and performance in tasks of executive function. The finding, obtained from the frequency analysis of BOLD signals, underlines the importance of dynamical analysis of fMRI time-series in the study of brain–mind relations. Future work will consider the determination of $H$ in resting-state fMRI paradigms and extend the measurements from controls to cohorts with executive impairments.

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