Speech structure links the neural and socio-behavioural correlates of psychotic disorders

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ABSTRACT

Background: A longstanding notion in the concept of psychosis is the prominence of loosened associative links in thought processes. Assessment of such subtle aspects of thought disorders has proved to be a challenging task in clinical practice and to date no surrogate markers exist that can reliably track the physiological effects of treatments that could reduce thought disorders. Recently, automated speech graph analysis has emerged as a promising means to reliably quantify structural speech disorganization. Methods: Using structural and functional imaging, we investigated the neural basis and the functional relevance of the structural connectedness of speech samples obtained from 56 patients with psychosis (22 with bipolar disorder, 34 with schizophrenia). Speech structure was assessed by non-semantic graph analysis. Results: We found a canonical correlation linking speech connectedness and i) functional as well as developmentally relevant structural brain markers (degree centrality from resting state functional imaging and cortical gyri
cation index) ii) psychometric evaluation of thought disorder iii) aspects of cognitive performance (processing speed deficits) and iv) functional outcome in patients. Of various clinical metrics, only speech connectedness was correlated with biological markers. Speech connectedness filled the dynamic range of responses better than psychometric measurements of thought disorder. Conclusions: The results provide novel evidence that speech dysconnectivity could emerge from neurodevelopmental deficits and associated dysconnectivity in psychosis.

1. Introduction

Psychotic disorders such as schizophrenia were originally conceptualized as conditions exhibiting loosening of associations in thought processes (Bleuler, 1950) and a weakening of associative-links in the brain (Lanczik and Keil, 1991). Loosening of associations in thought processes manifest as speech disturbances in patients. Neuroimaging investigations, on the other hand, provide the means to quantify brain dysconnectivity in psychosis. Understanding the relationship between these two levels of dysconnectivity is likely to provide an important lead in the pathophysiology of psychosis (Goghari et al., 2010). Although several attempts have been made to ascertain this connection (Sumner et al., 2017; Cavelti et al., 2018; Kircher et al., 2018), consistent neurobiological features underpinning thought disorder in psychosis are yet to be identified.

Assessment of the subtle aspects of thought disorders is a challenging task in clinical practice (Chapman and Chapman, 1973). Semi-structured instruments are often insufficient to detect subtle speech deficits during the course of a clinical interview. The use of speech samples (e.g. Thought and Language Index (Liddle et al., 2002a)) has improved the detection rate to some extent, making it possible to detect deviations in speech even in apparently healthy individuals (Sommer et al., 2010), though clinical judgment plays a key role in identifying such deficits. This introduces significant inter-rater variability (Liddle et al., 2002a), hindering efforts to understand the brain-symptom relationship. Recently, a novel approach using non-semantic graph...
analysis to calculate the connectedness between words has been suggested to quantify the manifest-scale dysconnectivity in speech samples (Mota et al., 2017; Mota et al., 2014; Mota et al., 2012). Specifically, it was shown that patients with schizophrenia diagnosis report their memories with lower connectedness between words, in comparison with non-psychotic individuals, or patients with psychotic symptoms diagnosed with bipolar disorder (Mota et al., 2017; Mota et al., 2014; Mota et al., 2012). One advantage of this new approach is to be naturalistic, analysing the free discourse of the patient, as manifested in the clinical interview. Second, it is completely independent of the subjective interpretation of symptoms. Third, this method allows for systematic comparison using multiple random graphs made with the same words used by the patient, thus improving the measurement precision (Mota et al., 2017). Thus, speech graph analysis might relate readily to the elusive pathophysiology of disorganization or formal thought disorder (FTD) seen in psychosis.

FTD is a feature of various psychotic disorders, including schizophrenia and bipolar disorder (Passby and Broome, 2017), with some elements of speech disturbances being observed more often in one disorder than the other (Andreasen and Grove, 1986; Hoffman et al., 1986). A valid measure of thought disorder can be expected to detect thought disturbances and its severity irrespective of diagnostic boundaries (i.e. schizophrenia and bipolar disorder with psychosis), while at the same time contributing to the clinical distinction among the disorders. Therefore, we expected to replicate previous results and find lower connectedness between words in patients with the schizophrenia diagnosis, in comparison with patients with psychotic symptoms and a diagnosis of bipolar disorder. Given the importance of FTD in defining the nature of psychosis and predicting its clinical (Demjaha et al., 2017) and global functional impact (Caveli et al., 2016), one must be able to relate such a measure to functional outcome as well as any existing scales for FTD. The measure should also be expected to relate to core cognitive deficits in psychoses such as processing speed (Dickinson et al., 2007), as FTD shows a complex interplay with cognitive impairment (Xu et al., 2014; Nagels et al., 2016). Finally, this measure can also be expected to relate to functional and structural measures of neuroanatomical measures.

Neuroimaging allows several variables reflecting brain dysconnectivity to be derived from subjects with psychosis. Disruption in connectivity during early development results in abnormalities in cortical folding and surface area. Unlike DTI that provides a current index of structural connectivity, the degree of cortical folding (gyrification) (White and Hilgetag, 2011; Dubois et al., 2008) provides a quantitative proxy of the degree of developmental brain connectivity in subjects with psychosis. Degree centrality, a measure based on the application of graph theory to brain networks, captures the number of functional connections between a single voxel and the rest of the brain (Buckner et al., 2009). Certain brain regions have a high degree centrality and act as core hubs for connectivity. We have recently reported that while the core hub architecture is largely preserved in patients with psychosis, decentralization of the core is noted with strengthening of peripheral hubs, leading to a change in the voxel-wise variance of degree centrality (Palaniyappan and Liddle, 2014). The variance of core degree centrality (VCC) provides a measure of decentralization of the core brain hubs, and has been used to quantify the connectational heterogeneity in social networks (Barocas et al., 2011; Stephenson and Zelen, 1989).

In the present study, we quantified ‘loosening of associations’ in speech processes using an unbiased speech-graph method in a sample of 56 clinically stable subjects with a psychotic disorder. We studied the differences in speech graph connectedness between schizophrenia and bipolar disorder. Then we verified whether speech connectedness was correlated with two measures of brain structure/function: VCC and gyrification. We also studied correlations of speech connectedness to the psychometric evaluation of formal thought disorder, to processing speed, working memory, and to social and occupational function in patients. This was performed within each diagnostic group as well as trans-diagnostically. Given the multiple association of FTD with biological, behavioural and functional outcomes in psychosis, our hypothesis was that the precise and objective measure of FTD provided by non-semantic graph features would be related to the psychometric evaluation of this class of symptom, to cognitive and global functioning, and – most critically – to brain structure and function.

2. Methods

2.1. Participants

The sample consisted of 34 patients satisfying DSM-IV criteria for schizophrenia, and 22 patients with DSM-IV bipolar disorder with psychotic features. Patients were recruited from the community based mental health teams (including Early Intervention in Psychosis teams) in Nottinghamshire and Leicestershire, UK. The diagnosis was made in a clinical consensus meeting in accordance with the procedure of Leckman et al. (Leckman et al., 1982), using all available information including a review of case files and a standardized clinical interview (Symptoms and Signs in Psychotic Illness (SSPI) (Liddle et al., 2002b)). All patients were in a stable phase of illness (defined as a change of no > 10 points in their Global Assessment of Function (GAF, DSM-IV) score, assessed six weeks prior and immediately prior to study participation). No patient had a change in antipsychotic, antidepressant or mood stabilizing medications in the six weeks prior to the study. Subjects with age < 18 or > 50, subjects with neurological disorders, current substance dependence, or IQ < 70 using Quick Test (Ammons and Ammons, 1962) were excluded. The median Defined Daily Dose (WHO, 2003) was calculated separately for antipsychotics (Supplementary Table 1). The study was given ethical approval by the National Research Ethics Committee, Derbyshire, UK. All volunteers gave written informed consent.

2.2. Clinical assessment

Patients were interviewed on the same day as the scan and symptom scores assigned according to the SSPI. Two subscales scores (Disorganized thought/language and Impoverished thought/language) were assessed using speech samples in line with the validated procedure for administering Thought Language Index. To generate free speech samples, 3 pictures from Thematic Apperception Test (Murray, 1943) were used in accordance with Sommer et al. (Sommer et al., 2010). Speech samples were audio recorded by two research psychiatrists (LP and VB) and transcribed by researchers (JP, SO) blind to the diagnostic status and symptom burden of the subjects. These four researchers (LP, VB, SO, JP) participated in several meetings directed by the original author of the scale to receive training in the assessment. Very good inter-rater reliability was demonstrated for 15 speech samples (intraclass correlation for total TLI score 0.83, 95% CI = 0.59 to 0.96); the remaining speech transcripts were rated by a single rater (SO) blind to diagnosis and symptom burden and neuroimaging findings. We also quantified current occupational and social dysfunction using the Social and Occupational Functioning Assessment Scale (SOFAS) (Goldman et al., 1992) and assessed speed of cognitive processing, a consistent and prominent cognitive deficit in schizophrenia, using the Digit Symbol Substitution Test (Dickinson et al., 2007). DSST was administered using a written and an oral format with a mean DSST score computed from the two formats (Palaniyappan et al., 2013a). N-back test with English alphabets was used to measure working memory (Kirchner, 1958), with performance measured using overall accuracy of responses averaged across 2 sessions (7 task-blocks with randomly presented 0-, 1- and 2-back conditions, each of 30 s duration with 10 s interval in-between (110 s/block; 7 blocks/session, 2 sessions in total). Each condition included 4 targets and 11 non-target stimuli with a 2 s inter-stimulus interval. (For more details, see (Palaniyappan et al., 2013a)).
2.3. Image acquisition

Functional MRI datasets were acquired on a 3 Tesla Philips Achieva MRI scanner (Philips, Netherlands) during 10 min of rest, with eyes open. To enhance sensitivity, dual-echo gradient-echo planar images (GE-EPI) were acquired (Posse et al., 1999), using an eight-channel SENSE head coil with SENSE factor 2 in anterior-posterior direction, TE1/TE2 25/53 ms, flip angle 85°, 255 × 255 mm field of view, with an in-plane resolution of 3 mm × 3 mm and a slice thickness of 4 mm, and TR of 2500 ms (40 contiguous axial slices in descending order, 240 time points in total). A magnetisation prepared rapid acquisition gradient echo image (MPRAGE) with 1 mm isotropic resolution, 256 × 256 × 160 matrix, TR/TE 8.1/3.7 ms, shot interval 3 s, flip angle 8°. SENSE factor 2 was also acquired for structural analysis. Out of 56 subjects who underwent clinical assessment, 2 with schizophrenia and 4 with bipolar disorder were excluded due to movement artefacts in either fMRI or structural scans, providing a final sample of 50 subjects for fMRI/structural analysis. There were no differences in the SSPI (total SSPI score mean (SD) in the included group = 10.94(7.96), excluded group = 0.4572) or TLI (total TLI score mean (SD) in the included group = 0.73(0.93), excluded group = 0.96(1.29), p = 0.7821) between patients who were included or excluded in the analysis.

2.4. fMRI analysis

The data was preprocessed using SPM8 (http://www.fil.ion.ucl.ac.uk/spm) and Data Processing Assistant for resting-state fMRI (Chao-Gan and Yu-Feng, 2010). After an initial correction for slice-timing differences, spatial realignment to the first image was carried out. We took several precautions to reduce movement induced confounds. These are detailed in Supplementary Material. A single weighted summation of the dual-echo dynamic time course was obtained for each subject (Posse et al., 1999), followed by retrospective physiological correction using RETROICOR (Glover et al., 2000). Unified segmentation based spatial normalization and smoothing using a Gaussian kernel of 8 mm Full-Width at Half Maximum was carried out. Following this, linear detrending and filtering using a band pass filter (0.01–0.08 Hz) was done to eliminate low frequency fluctuations and high frequency noise. Finally, variance accounted for by six head motion parameters, global mean signal, white-matter signal and CSF signal was removed by regression before conducting the degree centrality (DC) analysis.

Pre-processed data was analysed by deriving degree of centrality measure for every grey matter voxel using the cortical hub analysis procedure described by Buckner et al. (Buckner et al., 2009), and implemented in the REST software (Chao-Gan and Yu-Feng, 2010). For each voxel, we extracted the BOLD time course and correlated with every other voxel in the brain. For each voxel j the number of strong voxel-to-voxel correlations (defined as correlation coefficient r > 0.25) was computed to determine the DC of j. The threshold of 0.25 was chosen to avoid the inclusion of voxels that had low functional connectivity with the index voxel. For each subject, a map with DC values for every grey matter voxel was obtained. These maps were then z-transformed to enable group comparisons. Core hubs were identified using the entire sample as previously reported (Palaniyappan and Liddle, 2014) (Supplementary Tables 2 and 3; Suppl. Fig. 1). For voxels contained within the core hubs the voxelwise variance of normalized degree centrality of the core (VCC) was computed for each subject.

2.5. Surface extraction

Cortical surfaces were reconstructed from MPRAGE images using FreeSurfer version 5.1.0. The pre-processing was performed using standard procedures as described by Dale et al. (Dale et al., 1999). Following skull-stripping and intensity correction, the grey–white matter boundary for each cortical hemisphere was determined using tissue intensity and neighbourhood constraints. Total surface area (TSA = sum of right and left hemispheric area) was computed from reconstructed grey-white boundary. Vertexwise cortical folding pattern was quantified using local gyriﬁcation index, using the method advocated by Schaer et al. (Schaer et al., 2008) and described in detail in our previous study using an overlapping sample (Palaniyappan and Liddle, 2014). A global index of gyriﬁcation was obtained by computing the mean of the local the gyriﬁcation index (LGI) values from each vertex in each subject. A mean gyriﬁcation index (MGI) > 1 indicates a larger folded area (‘buried’ surface) compared to the outer circumference of the cortex (‘visible’ surface”).

2.6. Speech graph analysis

Graphs were constructed from each subject’s speech samples obtained during 3 instances of 1 min freely generated speech on the presentation of the pictures for the TLI. The speech samples were blindly transcribed by 2 researchers (JP and SO). Next, the samples were converted to graphs by a rater who was blind to diagnostic status (NBM), using an automated algorithm freely available online (SpeechGraphs, http://www.neuro.ufr.br/softwares/speechgraphs) (Mota et al., 2014). In these graphs each word corresponded to a node, and each temporal link between words corresponded to an edge (Fig. 1A). As a result, each word trajectory corresponded to an unweighted directed graph. To quantify graph connectedness, we used the SpeechGraphs software to calculate 2 connectedness speech graph attributes (Fig. 1A), which comprised connected components (LCC and LSC) (Fig. 1A). The LCC (largest connected component) is the total of nodes within the largest openly connected component of the graph, in which nodes are linked by at least one directed path. The LSC (largest strongly connected component) is the total of nodes within the largest closed component of the graph, in which any pair of nodes is linked by a direct or indirect path, and therefore mutually reachable, i.e., node “a” reaches node “b” and node “b” reaches node “a”.

In order to control for verbosity differences, connectedness attributes were calculated in sliding windows of 30 words (with 15-word overlap among consecutive windows) (Fig. 1B) and averaged over all windows from each report. To verify how close to randomness was the connectedness (LCC and LSC) of each graph from each word window, we performed 100 shuffles in the sequence of each set of words (Fig. 1C). Then we averaged the connectedness attributes of all random graphs and all word windows, producing a mean random value of LCC and LSC for each text. The original and the random values of LCC and LSC were averaged over the 3 images reports, generating one original mean value of LSC and LCC and one random value of LSC and LCC per subject. Then we calculated the ratio of original connectedness (LCC or LSC) by random connectedness (random LCC or random LSC) to verify how close to randomness this original report was (ratio closer to 1 means closer to randomness). We called these measures LCCr and LSCR. A subject, whose speech output, irrespective of number of words spoken, appears to have random links with low degree of goal direct- edness, will have LCCr value close to 1. On the other hand a subject whose speech makes no referential ties, and is simply a collection of random words temporally linked together, will have an LSCR value close to 1.

Previous studies showed that patients with schizophrenia diagnosis produced reports with lower speech connectedness (smaller LCC and LSC) reaching levels statistically similar to those of random speech connectedness (LSCR), when reporting a dream or an affective image (Mota et al., 2017), but not when reporting a spontaneous recollection of waking activities (Mota et al., 2014). Since the contents of the verbal reports in the present study were different from those mentioned above, and given the difference in the languages used (previous studies investigated native Portuguese speakers, the present study assessed native English speakers), we calculated the four measures of speech connectedness to verify whether the results could be replicated.
2.7. Statistical analysis

The graph attributes (originals LCC and LSC, and ratios LCCr and LSCr) were compared between the two groups using a Rank-sum test.

2.8. Canonical correlations

In order to correlate different sets of speech graph connectedness measures to other brain measures, psychometric evaluation of thought disorder, global functioning, and aspects of cognitive performance, we calculated canonical correlations (Das and Sen, 2014). To avoid collinearity, we performed multicollinearity diagnosis and used only sets of measures with conditioning index lower than 30. After that, we considered LCC, LSC and LSCr as the set for speech connectedness, VCC and LGI for brain structure/function, two sub-scales indexes of TLI (Disorganization and Impoverishment) for psychometric evaluation of thought disorder, GAF and SOFAS for global functioning, DSST, and n-back (0, 1 and 2-back) respectively for processing speed and working memory, two different aspects of cognitive performance. The main hypothesis was that speech connectedness should be correlated with psychometric, behavioural, functional and biological correlates of FTD; this was assessed with the canonical correlation. This analysis allowed us to verify whether there was a correlation between the sets of measures collected and how much variance we could explain from these relationships. The canonical coefficients define the relationship between the initial variable and the canonical variable (showed on Fig. 2B and C, X and Y axis). Also, we studied Spearman correlations for each pair of variables to better understand the direction of each relationship. As each of the 5 sets of assessments (speech connectedness, brain structure/function, psychometric evaluation of thought disorder, aspects of cognitive performance and global functioning) were correlated with the other 4 sets, and as each set was not expected to be independent from each other (given their associations), the canonical correlation results were considered significant after Bonferroni correction for 4 comparisons ($p < 0.0125$). This threshold aimed to minimize false positives without increasing false negatives. The canonical correlations were also performed within each diagnostic category (Schizophrenia and Bipolar Disorder). For isolated Spearman correlations we considered $p < 0.0062$, with Bonferroni correction for 8 comparisons (considering VCC, LGI, Disorganization, Impoverishment, GAF, SOFAS, DSST, n-back across two sessions).

Fig. 1. Methods and examples. A) An example text describing an image from the Apperception Test, represented as a speech graph (each word represented as a node and each temporal sequence of consecutive words represented as an edge). Light grey circle is the set of nodes in the largest connected component (LCC) and the black circle is the set of nodes in the largest strongly connected component (LSC). B) An illustrative example of the previous text analysed by the use of a 20-word slide window with 50% overlap. C) An illustrative example of the first word-window text from the previous example after shuffling of word order. D) Two representative examples of the Schizophrenia and Bipolar groups.
3. Results

The demographic and clinical features of the two groups are shown in Table 1. There was a statistical difference related to socioeconomic status between the groups, but no other measure related to age, sex, IQ, equivalent dose of medication, handedness or disease duration was found to be statistically different. There were no significant differences related to psychometric evaluation, global functioning, and aspects of cognitive performance or brain structure/function between the Schizophrenia and Bipolar groups.

3.1. Speech graph analysis

From mathematical definitions of the four speech connectedness measures (LCC, LSC, LCCr and LSCr) we understand that they measure similar aspects of what we call here “speech connectedness”. To observe
empirically on this sample the magnitude of the shared co-variance between those measures, we calculated R square (R2) and p values of each correlation pair. All different combinations showed p values smaller than 0.0031, confirming that all measures were correlated, but the level of co-variance spanned between $R^2 = 0.22$ to 0.99 ($R^2$; LCC x LSC = 0.99; LCC x LCCr = 0.22; LCC x LSCr = 0.38; LSC x LCCr = 0.23; LSC x LSCr = 0.39; LCCr x LSCr = 0.81), which means that those measures are not exactly the same. We interpreted our results after taking this point into account, with emphasis on measures corrected for randomness.

Speech connectedness indexed by LCC, LCCr and LSCr was significantly different between the 2 groups. In general, patients with schizophrenia produced less connected reports, with speech connectedness closer to chance when compared to bipolar disorder (lower LCC, LCCr and LSCr, far from 1) (Fig. 1D and 2A), indicating that in schizophrenia patients the referential ties as well as the goal directedness were reduced when compared to bipolar disorder.

3.2. Correlations with brain structure/function

We found significant canonical correlation between speech connectedness and brain structure/function, which explained nearly 25% of the variance ($R = 0.49$, $p = 0.00091$; Fig. 2B, D and Table 2). Using the Spearman correlation, VCC (during resting state) was anti-correlated with LCC and LCCr (Table 3), but there was no significant correlation with LGI, only a trend between LGI and LSCr (Table 3). No other significant canonical correlations between brain structure/function (LGI and VCC) and psychometric evaluations (formal thought disorder measured by TLI), global functioning assessments (GAF and SOFAS) or aspects of cognitive performance (DSST, N-back) were observed (Fig. 2C, D and Table 2).

### Table 2

<table>
<thead>
<tr>
<th>Brain markers</th>
<th>Cognitive performance</th>
<th>Global functioning</th>
<th>Aspects of cognitive performance</th>
<th>Thought disorder</th>
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</thead>
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<td>Speech connectedness</td>
<td>0.54</td>
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<td>Speech connectedness x Global functioning</td>
<td>0.59</td>
<td>0.0006</td>
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<td>0.0006</td>
</tr>
<tr>
<td>Speech connectedness x Thought disorder</td>
<td>0.37</td>
<td>0.0796</td>
<td>0.37</td>
<td>0.0796</td>
</tr>
</tbody>
</table>

3.3. Correlations with clinical assessment

There was a significant canonical correlation between speech connectedness and the psychometric evaluation of thought disorder, which explained 36% of the variance ($R = 0.6$, $p = 0.0003$; Fig. 2B, D and Table 2). Impoverishment and LSC showed negative Spearman correlation (Table 3): More severe symptoms were associated with largest closed loops closer to random. A significant canonical correlation occurred between speech connectedness and global functioning, which explained 34% of the variance ($R = 0.59$, $p = 0.0006$; Fig. 2B, D and Table 2), with positive Spearman correlations between GAF and LCC, LSC and LSCr (Table 3): Higher scores on GAF were associated with more nodes in the largest closed and open components, and closed components farther from what would be expected by chance. Importantly, speech connectedness showed significant canonical correlation with aspects of cognitive performance, explaining 29% of the variance ($R = 0.54$, $p = 0.0047$; Fig. 2B, D and Table 2), with positive Spearman correlations between DSST and LCC, LCCr and LSCr (Table 3). This means that better performance was correlated with larger deviations of open and closed components in relation to randomness. There were significant canonical correlations among all behavioural assessments, except for psychometric evaluation and global functioning (Table 2).

3.4. Correlations within the schizophrenia and bipolar groups

To investigate potential group specificities we separately calculated the same canonical correlations for each diagnostic group (schizophrenia and bipolar disorder) (Suppl. Fig. 2). Only for the schizophrenia group we found a significant relationship between i) brain function and speech, and ii) social function and speech. The correlations between speech connectedness and thought disorder/cognitive tests were not significant for any of the groups separately (Suppl. Fig. 2). It is important to consider that the R values of the canonical correlations obtained were very similar to those obtained when considering all subjects together, and therefore the reduction in N most likely explains the increase in p values.

4. Discussion

The results suggest that the application of graph theory to analyse the connectedness of speech samples in clinically stable patients can
help to establish the relationship of this pathological phenomenology with biological markers. Speech connectedness (the amount of nodes or words engaged in the largest open and closed components of word trajectory graphs, as well as the deviation from random word graphs made with the same set of words) is correlated to measures of cortical development and dysconnectivity. Speech connectedness explained a core cognitive deficit and functional outcome in patients, irrespective of diagnostic boundaries.

In the present study, we note that the less connected the speech, the lower the scores in global functioning (stronger results with GAF) and processing speed (digit-symbol test). The relationship between social outcome and impoverished speech has been demonstrated previously both in chronic schizophrenia (Bowie and Harvey, 2008) and in adolescents at high risk for schizophrenia (Bearden et al., 2011). This reaffirms the consistent associations shown between communication disturbances and predictors of social functioning such as social cognition and engagement (Bowie et al., 2011). In their meta-analysis of neuro-cognition in relation to clinical syndromes, Dominguez et al. (Dominguez Mde et al., 2009) noted a significant association between disorganization and processing speed deficits. Our observations suggest that this relationship is mediated by the presence of impoverished thought. While we observed a robust relationship between speech connectedness and GAF, the correlation with SOFAS was somewhat weaker, not reaching our stringent thresholds for statistical significance. As GAF factors in current psychopathology as well as social function, the observed relationship may not be specific to social functioning.

In previous work with Portuguese-speaking patients it was found that speech connectedness measured by graph analysis was strongly anti-correlated with severity in negative and cognitive symptoms measured by psychometric evaluation in chronic patients (Mota et al., 2014) and even before the formal diagnosis in recent onset psychosis patients (Mota et al., 2017), helping differential diagnosis of Schizophrenia in both situations (Mota et al., 2017; Mota et al., 2014; Mota et al., 2012). We replicate these results, also finding a more connected speech structure in patients with Bipolar Disorder diagnosis in comparison to the Schizophrenia diagnosis, even in native speakers of another language (English). This corroborates the language invariance of this method (Mota et al., 2014). These findings are in line with previous works suggesting that in schizophrenia, less syntactically complex utterances are prevalent, with a lack of coherent discourse structure (Hoffman et al., 1986; Thomas et al., 1996; Fraser et al., 1986; King et al., 1990). This supports the notion that the thought disorder has a dimensional distribution across psychotic disorder, with selected components of this dimension being more pronounced in schizophrenia than in bipolar disorder (Perlini et al., 2012).

Extending previous work (Mota et al., 2017; Mota et al., 2014), we have also identified the relationship between abnormalities in the language structure and the clinician’s perception of thought disorder. A less connected speech output, as measured by low values of connectedness speech graphs attributes, maps on to an impoverished thought/language score as rated by clinicians. Taken together, these evidences indicate that connectedness of speech, particularly in comparison to randomness, is associated to poverty in thought process expressed through language.

Besides this relationship between graph analysis of speech and psychometric evaluation, we notice that speech connectedness explains 24% of the variance in brain structure/function, but not between the latter and the psychometric evaluation of thought disorder (TLI). The naturalistic, automated method of quantifying speech connectedness indexes the variation necessary to relate thought disorder and brain structure/function. Specifically, reduced goal directedness (i.e. connectedness similar to random set of words, low LCCr) of the speech structure was associated with higher variance in the centrality. This suggests that with a reduction in the integrity of the core functional hubs (especially left inferior parietal, superior temporal, anterior and posterior cingulate and insula, see Supplement Table S2), reflected by an increase in voxel-wise variance of the centrality, speech structure becomes disorganized, approaching randomness. To our knowledge, this is the first time that a relationship between a measure reflecting the decentralization of brain’s functional network architecture and thought disorder has been reported. This highlights the potential usefulness of the combined clinical and automated graphical assessment of speech disturbances in psychosis. Furthermore, the observation that several speech/language features predict functional outcome and measures of brain structure/function after allowing for variance in diagnosis supports the hypothesis that some pathological features are distributed in a continuum across psychotic disorders.

The trend of positive correlation between speech graph connectedness and gyriﬁcation indicates that subjects with a neurodevelopmental disturbance indexed by reduced gyriﬁcation produce speech with lower degree of referential ties approaching randomness (low LSCr). We have previously reported that the most signiﬁcant reduction in gyriﬁcation affects the fronto-insular cortex and the superior temporal cortex in psychosis (Palaniyappan and Liddle, 2014; Palaniyappan and Liddle, 2012; Palaniyappan et al., 2013b), in line with several previous studies suggesting a relationship between the dysfunction of these brain regions and formal thought disorder. In the context of the current findings, the relationship between specific components of thought disorder and the gyriﬁcation patterns of these regions require further study. We have previously shown a relationship between speech connectedness and the typical development of intelligence quotient and theory of mind abilities, which allows a prediction of reading acquisition in typical developing children (Mota et al., 2016b) – but note that this study did not investigate gyriﬁcation. Also a slow asymptotic development of speech connectedness, reaching asymptotic stable values around 13 years of schooling, was observed in typical developing individuals, but not in a psychotic population, despite schooling (Mota et al., 2016a).

While the correlations with speech connectedness across diagnostic groups were signiﬁcant, most of the correlations within each diagnostic group were not. No correlations at all were signiﬁcant within the bipolar group, and only correlations with brain function and global functioning were signiﬁcant within the schizophrenia group. This could

Table 3
Spearman Correlations: VCC: Variance of the degree centrality of the core hubs; LGI: Gyrification index; LCC: Largest connected component; LSC: Largest strongly connected component; LCCr: ratio LCC/mean LCC in 100 random graphs; LSCr: ratio LSC/mean LSC in 100 random graphs; GAF: Global Assessment of Function; SOFAS: Social and Occupational Functioning Assessment Scale; DIS: Disorganized thought/language; IMP: Impoverished thought/language; DSST: Digit Symbol Substitution Score; N-Back: working memory test, boldface indicates $p < 0.0062$.

<table>
<thead>
<tr>
<th>Connectedness</th>
<th>VCC</th>
<th>LGI</th>
<th>DIS</th>
<th>IMP</th>
<th>GAF</th>
<th>SOFAS</th>
<th>DSST</th>
<th>N-Back</th>
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<tbody>
<tr>
<td>LCC</td>
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<td>0.25</td>
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<td>−0.24</td>
<td>0.0949</td>
<td>0.32</td>
<td>0.0162</td>
<td>−0.14</td>
<td>0.3132</td>
<td>−0.38</td>
<td>0.0041</td>
</tr>
</tbody>
</table>
be interpreted to support the assumption that the relationships with speech connectedness are restricted to the latter group. But since the R values found on Bipolar group analysis were similar to the R values found when considering the entire sample, and given that sample re-

A major limitation of the current study is that we lacked equivalent speech sample data for healthy controls. As a result, we are unable to found when considering the entire sample, and given that sample re-

et al., 2010; Bearden et al., 2011). Most of our patients were medicated; this could have reduced the severity and the variance of thought disturbances. Yet, as shown in our previous study using this sample (Palaniyappan and Liddle, 2014), we did not find any linear relationship between prescribed medications (Supplement Table 1) and the degree of abnormalities in gyriﬁcation or centrality. Nevertheless, we cannot completely rule out the confounding effects of medications. Another limitation is the use of GAF, which is confounded by the mixture within its rating system of psychopathological symptoms and functional measures.

Neuroimaging of symptoms in psychosis is widely believed to be a powerful means to further understand psychiatric disorders at a systems level (Redpath et al., 2013). The speech graph approach employed here provides an automated, unbiased means of quantifying the subtle speech disorders in psychosis with a short duration of assessment. In combination with clinical ratings, this approach not only predicts real-world outcome measures but also reveals the neurobiological underpinnings of thought disorders, thus providing a robust means to study the pathophysiology of psychosis.

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Conflicts of interest

L. Palaniyappan received speaker fees from Otsuka and educational grant from Janssen in 2017. P F Liddle has received honoraria for academic presentations from Janssen-Cilag and Bristol Myers Squibb; and has taken part in advisory panels for Bristol Myers Squibb. NBM, SR and MC receive support from Boehringer-Ingelheim since 2017. All other authors declare no conﬂict of interest.

Ethical standards

The authors assert that all procedures contributing to this work comply with the ethical standards of the relevant national and institutional committees on human experimentation and with the Helsinki Declaration of 1975, as revised in 2008. The authors assert that all procedures contributing to this work comply with the ethical standards of the relevant national and institutional guides on the care and use of laboratory animals.

Appendix A. Supplementary data

Supplementary data to this article can be found online at https:// doi.org/10.1016/j.pnpbp.2018.07.007.

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