Dynamic cerebral reorganization in the pathophysiology of schizophrenia: a MRI-derived cortical thickness study

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Background. A structural neuroanatomical change indicating a reduction in brain tissue is a notable feature of schizophrenia. Several pathophysiological processes such as aberrant cortical maturation, progressive tissue loss and compensatory tissue increase could contribute to the structural changes seen in schizophrenia.

Method. We studied cortical thickness using surface-based morphometry in 98 clinically stable patients with schizophrenia and 83 controls. Using a pattern classification approach, we studied whether the features that discriminate patients from controls vary across the different stages of the illness. Using a covariance analysis, we also investigated if concurrent increases accompany decreases in cortical thickness.

Results. Very high levels of accuracy (96.3%), specificity (98.8%) and sensitivity (88%) were noted when classifying patients with <2 years of illness from controls. Within the patient group, reduced thickness was consistently accompanied by increased thickness in distributed brain regions. A pattern of cortical amelioration or normalization (i.e. reduced deviation from controls) was noted with increasing illness duration. While temporo-limbic and fronto-parietal regions showed reduced thickness, the occipital cortex showed increased thickness, especially in those with a long-standing illness.

Conclusion. A compensatory remodelling process might contribute to the cortical thickness variations in different stages of schizophrenia. Subtle cerebral reorganization reflecting the inherent plasticity of brain may occur concomitantly with processes contributing to tissue reduction in adult patients with schizophrenia.

Introduction

Robust changes in grey matter are observed using structural magnetic resonance imaging (MRI) in patients with schizophrenia both before and after the onset of psychosis, and are linked to the clinical features of the illness (Glahn et al. 2008; Bora et al. 2011; Chan et al. 2011). Rather controversially, these neuroanatomical changes have been argued to be progressive in nature, indicating a deteriorating pathophysiological process (Weinberger & McClure, 2002; DeLisi, 2008; Hulshoff Pol & Kahn, 2008; Andreasen et al. 2011; Olabi et al. 2011; Vita et al. 2012). Longitudinal MRI studies report loss of grey matter at a rate that is often implausible (e.g. ~0.43% of frontal grey-matter volume/year in the first 12–15 years of illness; Andreasen et al. 2011) and if maintained, would result in little brain tissue being left after 2–3 decades of illness in an individual (Weinberger & McClure, 2002). In addition, the putative neuroprogressive changes do not seem to reflect the severity of clinical deterioration (McGlashan, 2006) even in the samples showing such significant tissue loss (Zipparo et al. 2011).
2008), although conflicting data exist in the literature (Van Haren et al. 2008; Andreasen et al. 2011). Some studies with notable sample sizes have failed to observe neuroprogressive changes (Schaufelberger et al. 2011), and instead report a significant increase in the volume of some brain regions (Roiz-Santíañez et al. 2014). In light of this evidence, neuroprogression, if present, is likely to be limited not only in time (as highlighted by Van Haren et al. 2008) but also in spatial distribution (Rosa et al. 2015), and occur alongside compensatory changes in the opposite direction.

As pointed out recently (Zipursky et al. 2013), the idea of ‘neuroprogression’ or ‘structural degeneration’ is somewhat inconsistent with the improvement and/or stabilization in clinical and functional domains in a large number of patients (Arndt et al. 1995; Levine et al. 2011). Diffuse patterns of neuroanatomical changes reflect clinical features of schizophrenia, for example, grey-matter reduction in left-perisylvian regions is associated with positive symptoms, extended frontolimbic reductions with negative symptoms and temporal, insular and medial prefrontal changes with disorganization (Koutsouleris et al. 2008). Either a reversal of these changes, or the appearance of compensatory changes could be expected to occur in those with long-standing illness. McGlashan (2006) argued that the pathological brain changes in schizophrenia are a result of both destructive and ameliorative processes and suggested that chronic illness may lead to concomitant ‘disuse atrophy in some circuits and overuse hypertrophy in other circuits’. This prospect has not been examined in detail to date. Functional MRI studies note that increase in functional activation in certain regions compensates for the reduced activation of the primary task-relevant brain regions (Quintana et al. 2003; Tan et al. 2006), thus producing a rescuing effect in patients with a longer duration of illness (Faget-Agius et al. 2013). In terms of brain structure, if tissue reduction occurs as a primary change, then concomitant (covarying) increase in tissue in other brain regions can be expected to compensate for the deficits. Grey-matter increases are shown to be associated with clinical improvement in some patients with schizophrenia (Ansell et al. 2015), suggesting that such increases are likely to be ‘ameliorative’ or compensatory processes. But structural studies have mostly focused on localizing the regions with most consistent morphological change (Glahn et al. 2008; Bora et al. 2011; Chan et al. 2011), and have not paid close attention to the nature of relationship (covariance) among brain regions that can uncover the subtle changes in opposite direction that occur concomitantly.

Several distinct observations in a cross-sectional sample of patients with varying illness duration could provide indirect evidence for ameliorative or compensatory neuroanatomical changes in schizophrenia. First, volumetric studies in schizophrenia suggest that while some brain regions with reduced grey matter in early stages show more pronounced deficits in chronic schizophrenia, several other regions with early reduction do not appear to have the same degree or more pronounced deficits in later stages (Ellison-Wright et al. 2008; Chan et al. 2011). This latter observation has not been adequately highlighted in the past but is readily evident in an extensive meta-analysis of stage-specific volumetric changes reported by Chan et al. (2011) (see Fig. 3 and Table 4). Thus, in the presence of an ameliorative process, brain regions showing pronounced changes in the early stages of illness will be expected to show less pronounced deviation from healthy controls in those with more chronic illness (hypothesis 1). Second, in the absence of a pathophysiological process affecting distributed brain regions (i.e. in healthy controls), widespread age-related decline occurs in cortical thickness, with few regional exceptions (e.g. primary visual cortex, medial temporal lobe) (Brans et al. 2010; Thambisetty et al. 2010; Storsve et al. 2014). Thus, within the population of healthy controls, the structural covariance between any two brain regions is more likely to show a positive than a negative relationship brought about by regionally non-specific determinants of tissue volume (i.e. both regions show concomitant increase or decrease in thickness across individuals). In contrast, provided active regional remodelling/compensation accompanies tissue reduction, a large number of negative covarying pairs of regions can be expected in patients (i.e. as one region shows thinning, another may show thickening across individuals) (hypothesis 2). Third, meta-analytical anatomical likelihood estimates of morphometric changes consistently demonstrate regions with relative tissue reduction but not tissue increase in different stages of schizophrenia (Chan et al. 2011; Fusar-Poli et al. 2011). This suggests that compensatory increases in grey matter, if present, are likely to be highly variable across individuals. As a result of the higher inter-individual variability of putative compensatory changes, we can expect a reduction in the discriminatory ability when using neuroanatomical patterns to differentiate controls from patients with longer illness duration (hypothesis 3).

We tested these hypotheses in a sample of 98 clinically stable patients receiving treatment for schizophrenia and 83 healthy controls using a pattern classification approach and covariance analysis of cortical thickness. We chose MRI-derived cortical thickness as several studies have previously established that altered cortical thickness is robust feature of schizophrenia (Kuperberg et al. 2003; Narr et al. 2005; Nesvåg et al.
Further thickness is more likely to reflect illness-related factors than surface area (Sprooten et al. 2013; Haukvik et al. 2014), making it a suitable index when studying the malleability of the cortex.

**Method and materials**

**Participants**

Ninety-eight patients at various stages of schizophrenia (80 males, 18 females) were recruited for this study. The mean illness duration assessed using case notes, patient interview and referral information as the time from the point of first contact with psychiatric services with psychotic symptoms to the date of scan acquisition for these patients was 5.5 years (median = 3 years, range 6 months–29 years). The diagnosis of schizophrenia (DSM-IV criteria) was made in accordance with the procedure of Leckman et al. (1982) using data from all available sources [case notes, a standardized clinical interview based on the Signs and Symptoms of Psychotic Illness scale (Liddle et al. 2002) and reports from other informants when required]. Subjects with neurological disorders, current substance dependence, IQ < 70 using the Quick Test (Ammons & Ammons, 1962), and diagnosis of any other Axis I disorder were excluded. Seventy-eight out of 98 patients were receiving psychotropic medications (average dose in chlorpromazine equivalents 490 mg, range 25–3000 mg). Healthy controls were recruited from the local community via advertisements and comprised 83 subjects free of any psychiatric or neurological disorder matched groupwise in age (±3 years) and socio-economic status (measured using National Statistics – Socio Economic Classification; Rose & Pevalin, 2003) to the patient group. Controls had similar exclusion criteria to patients; in addition subjects with history of psychotic illness in first-degree relatives were excluded. Regional Ethics Committees (Nottinghamshire and Derbyshire) approved the study and all participants provided written informed consent. More details are given in Table 1.

**Image acquisition and morphometry**

For details on image acquisition see Appendix 1 in the Supplementary material. Image processing for morphometry was carried out by a single researcher (L.P.) applying previously used criteria to reduce Freesurfer-related processing artefacts (Appendix 1, Supplementary material) (Palaniyappan & Liddle, 2012). The cortical thickness was measured by calculating the Euclidean distance between linked vertices on the inner and outer cortical surfaces (Fischl & Dale, 2000), as shown in Fig. 1a, and in accordance with standard descriptions (Dale et al. 1999). Anatomical parcellations were obtained using the Destrieux sulcogyral atlas (Destrieux et al. 2010), which follows the anatomical conventions of Duvernoy (Duvernoy & Bourgouin, 1999). The details of the 148 resulting parcellations are presented in Supplementary Fig. S1 and Table S1. Regional thickness values were obtained from the mean thickness of all tessellations that were confined within the boundaries of the parcellated sulcogyral regions. Compared to the 148 values obtained per subject for regional thickness measures, vertexwise data provide more than 150,000 values per subject for pattern classification. Nevertheless, regional measures provide anatomically intuitive data that enables a meaningful interpretation of individual data points. Increasing the number of data dimensions in the presence of a limited sample size will affect the trade-off between complexity and error margin, resulting in overfitting (Song et al. 2011). Further, covariance between two anatomical regions is more readily interpretable than the covariance between two vertices in the cortical surface. For these reasons, we used parcellation-based regional thickness for pattern classification and covariance analysis as detailed below.

**Pattern classification and covariance analysis**

For pattern classification analysis, we used Support Vector Machine (SVM), a learning machine for two-class problems. SVM toolkit libsvm written by Lin Chih-Jen from Taiwan University (http://www.csie.ntu.edu.tw/~cjlin/libsvm/) was used with a radial basis function (RBF) as kernel function. In a different dataset published elsewhere (Guo et al. 2014) we tested optimal parameters for employing RBF-based SVM, and chose the values of $\gamma = 2$ and parameter $C = 10$ to achieve a robust trade-off between model complexity and training error for our classification problem. Other SVM parameters were kept as default values, in line with our previous work (Guo et al. 2014). RBF has been shown to perform optimally when region-based neuroimaging data is used in SVM classification (Song et al. 2011). The trade-off parameter $C = 10$ is the best one to get the highest accuracy rate between parameter range (1–15). To measure the test performance and to validate the classifier, a leave-one-subject-out cross-validation approach was employed, where the classifier is trained on all subjects except one, which is used as test data. Balanced accuracy, specificity, sensitivity, and predictive values for each classifier were obtained and statistical significance of these measures was determined by way of permutation testing ($n = 1000$ permutations with random assignment of patient/control labels to the training data). Mean
Table 1. Demographic and clinical characteristics

<table>
<thead>
<tr>
<th></th>
<th>Schizophrenia patients ((n = 98))</th>
<th>Healthy control ((n = 83))</th>
<th>(p) value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age (years)</td>
<td>30.3776 ± 8.6781</td>
<td>29.4819 ± 8.7239</td>
<td>0.4910 ((t) test)</td>
</tr>
<tr>
<td>Sex (M/F)</td>
<td>80/18</td>
<td>70/13</td>
<td>0.6303 ((\chi^2) test)</td>
</tr>
<tr>
<td>Intracranial volume</td>
<td>1371.2 ± 238.02</td>
<td>1399.8 ± 240.12</td>
<td>0.4233 ((t) test)</td>
</tr>
<tr>
<td>Illness duration (years)</td>
<td>5.5357 ± 6.212</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>Chlorpromazine dose equivalents</td>
<td>389.9 ± 501.7</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>SSPI score</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Psychomotor poverty</td>
<td>2.8673 ± 3.1447</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>Reality distortion</td>
<td>2.4388 ± 2.4371</td>
<td>--</td>
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</tr>
<tr>
<td>Disorganization</td>
<td>0.9898 ± 1.2141</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>Total score</td>
<td>10.898 ± 6.8946</td>
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</tbody>
</table>

SSPI, Signs and symptoms in psychotic illness.

Fig. 1. (a) Coronal (left) and horizontal (right) slices of the left hemisphere with gray/white and pial surfaces overlaid. Thickness measures were obtained by calculating the shortest perpendicular distance between the two surfaces across the entire volume for global thickness but limited to boundaries of anatomical parcellations for regional thickness. (b) Comparison of mean hemispheric thickness between controls (black bar) and schizophrenia patients (white bar) of left and right hemisphere. There are significant differences between these two groups on both hemispheres. (c) Group difference between patients and controls in mean global thickness was seen in males but not in females. (d) Thickness has significant negative correlation with age both for controls and patients. (e) There was no significant relationship between residual of thickness after regressing out the influence of age and duration of illness in the patient group for both hemispheres.
discrimination accuracy, sensitivity and specificity was obtained for the entire sample. To study the effect of illness duration, we categorized patients into multiple bins on the basis of their illness duration as shown in Table 2 and Supplementary Table S6. The SVM classification was carried out for each bin independently, estimating the classification accuracy against the same set of controls using regional thickness measures after removing the linear effect of age and gender.

For covariance analysis structural cortical networks were constructed from the regional cortical thickness measurements of all 148 regions after removing the effects of age and gender, in line with previous studies (He et al. 2007; Zhang et al. 2012). More details of the mathematical computation are provided in the Supplementary material.

Demographic and clinical characteristics were compared using either two-sample t test or χ2 test. Correlation analyses among normally distributed variables or residuals were conducted using Pearson’s test.

Results

Global measures of thickness

Mean global thickness measures compared using two-sample t test revealed reduced thickness in patients compared to controls in both hemispheres (left: $p = 9.8 \times 10^{-7}$; right: $p = 6.9 \times 10^{-9}$), as shown in Fig. 1b. Group difference in mean global thickness was seen in males ($p < 0.001$) but not in females ($p = 0.50$), as shown in Fig. 1c. There was a significant negative relationship between age and global thickness in both groups (controls: $r = −0.55$, $p = 7.07 \times 10^{-8}$, schizophrenia: $r = −0.49$, $p = 3.7 \times 10^{-7}$), as shown in Fig. 1d. After regressing out the influence of age and gender, global thickness had no significant association with illness duration, although the correlation was numerically positive for both hemispheres (left: $r = 0.06$, $p = 0.57$; right: $r = 0.08$, $p = 0.41$), as shown in Fig. 1e. There was no significant difference of global thickness between two different scanners for all samples ($p = 0.7720$), for controls ($p = 0.3880$) and for patients ($p = 0.9566$) after controlling for duration, age and gender.

Regional measures of thickness

When all patients were compared to controls, 45 regions showed significantly reduced thickness (Bonferroni corrected). Most significant reduction was seen in parahippocampal gyrus ($p < 10^{-20}$ for left and right), supramarginal gyrus ($p < 10^{-20}$ for left and right), precentral gyrus ($p < 10^{-20}$ for left and right), temporal pole (left: $p < 10^{-20}$; right: $p = 10^{-16}$) and right superior insula ($p = 10^{-15}$). Eleven regions including inferior occipital gyrus (left: $p < 10^{-9}$; right: $p < 10^{-11}$), lateral fissures (left: $p < 10^{-6}$; right: $p < 10^{-8}$), occipital pole (left: $p < 10^{-8}$; right: $p < 10^{-8}$) and calcarine sulcus (left: $p < 10^{-6}$; right: $p < 10^{-9}$) showed significantly greater thickness in patients. The results are shown in Fig. 2.

Within the patient group, several regions showed significant correlation (uncorrected $p < 0.05$) with illness duration after regressing out the influence of age and gender (listed in Supplementary Table S2). Rather strikingly, the regions that showed significant reduction in thickness in the patients v. controls comparison, showed a positive association with illness duration, while those that had increased thickness showed a negative association. In other words, with increasing illness duration, the group differences in thickness had a tendency to ‘normalize’.

To study if this ‘normalization’ across long illness duration occurs in regions that show notable differences in the early stage of illness (<2 years), we plotted the group difference of the mean residual of regional thickness for three regions with maximum group difference in the early stage (left and right parahippocampal, and left supramarginal gyrus), against the illness duration. As shown in Fig. 3a, a pattern of reducing deviation is seen with increasing illness duration across the group.

SVM classification

When the entire sample of patients and controls were compared on the basis of age- and gender-adjusted regional thickness, a discrimination accuracy of 81.77% with specificity 86.75% and sensitivity 77.55% was obtained. When patients were categorized to different illness duration bins, the discrimination accuracy reduced steadily when patients with longer durations were included in the classification matrix. We further studied this effect by studying four distinct subgroups of illness duration (<2, 2–4, 4–10, > 10 years). Very high levels of accuracy (96.3%), specificity (98.8%) and sensitivity (88%) were noted for those with <2 years of illness. This was not affected by scanner or sample size differences (Supplementary Tables S7 and S8). The discrimination accuracy is shown in Table 2 and the best predictors in the different bins are shown in Supplementary Table S3.

Structural covariance

Pearson correlation matrices constructed for patients and controls are displayed in Fig. 3b. A large number of divergent relationships (negative covariance) were noted in patients compared to controls. When the strongest 100 relationships (irrespective of the spatial location) were considered in each group, 36 divergent (negative) relationships were seen in patients while
only one relationship was divergent in controls, indicating that in patients, a number of regions increase in (‘gain’) thickness while some regions decrease (‘loss’). The 36 regional pairs identified in this analysis are listed in Supplementary Table S4. When the within-group, inter-individual variation in the covariance scores were plotted for these 100 regional pairs, as shown in Fig. 3d, patients with schizophrenia had greater variances than controls (mean standard deviation in patients = 2.1814, controls = 1.0801; t test = −13.1107, p < 0.001), suggesting that the reciprocal (putatively compensatory) relationships do not always involve the same pairs of brain regions in patients. The correlation matrices were thresholded into a set of binarized matrices that describe the topological organization of the structural cortical networks for two groups. We examined different correlation thresholds (0–1 in steps of 0.05) to examine the proportion of divergent links (negative covariance) for all possible regional pairs at the different thresholds values. As shown in Fig. 3c, we noted that patients had more negative links for all thresholds, with the gap between the two groups widening notably when stronger pairwise relationships are considered. This pattern was not related to the dose of antipsychotics used by patients (Supplementary Fig. S3). These covariance results indicate that the patient group shows a larger than expected number of reciprocal changes in regional thickness.

A hub-and-spokes plot showing the covariance patterns for two brain regions with maximal group difference in each direction (left parahippocampal and supramarginal regions with reduced thickness and bilateral inferior occipital region with increased thickness in patients) is shown in Fig. 4. In these regions, patients had a number of divergent relationships with a distributed set of regions. More details are presented in Supplementary Table S5.

Discussion

In a large sample of patients recruited at a single centre (Nottingham), we investigated the effect of illness duration on cortical thickness and report four key observations. In accordance with our prediction, a pattern of concomitant group differences in both directions (i.e. relative increase and decrease) resulting in an overall reduction in anatomical deviation from controls was noted in later stages of schizophrenia. The ability to discriminate patients from controls on the basis of cortical thickness is greatest during the early phase of the illness, and reduces with longer illness duration. The features that discriminate a patient from controls vary across the different stages of the illness.

At the outset, it is important to note that our results neither imply an overall increase in thickness nor refute the appearance of more extensively distributed tissue reduction in later stages of schizophrenia. In fact,
the most prominent group difference in the entire sample, irrespective of illness duration, is a reduction in thickness seen in patients. We noted that in each duration bin, the best predictors in each classifier were regions showing reduced thickness in patients (Supplementary Table S3). This is consistent with meta-analyses of volumetric studies that show pronounced grey-matter reduction in the later stages of schizophrenia (Bora et al. 2011; Chan et al. 2011; Hajima et al. 2013). We add two important cross-sectional observations to this literature. (1) Across subjects with schizophrenia, concomitant increases in thickness accompany reductions. (2) Regions showing most pronounced changes (increase or decrease) among patients in early stage of illness are not necessarily ‘worse’ among the patients in later stages of illness. Taken together, these observations are suggestive of a compensatory/remodelling process contributing to the cortical thickness variations in schizophrenia.

Using a structural covariance analysis we noted a higher number of reciprocal changes in thickness.
within the patient group than in controls. Most previous studies of cortical thickness in schizophrenia have sought the brain regions showing the largest magnitude of group differences, and predominantly report reduced thickness in patients (Kuperberg et al. 2003; Narr et al. 2005; Venkatasubramanian et al. 2008; Schultz et al. 2010). To the best of our knowledge no study has investigated the concomitant changes in the brain that relate to localized changes in thickness. We note that regional thickness reduction is frequently accompanied by thickness increase in other brain regions in patients, with a high degree of between-subjects variation in the spatial distribution of the covariance patterns in patients. Further, the brain regions

Fig. 3. (a) Traces of mean residual of thickness of top three regions of interest with significant difference between early stage (~2 years) illness and controls as a function of illness duration. The y axis represents the difference of the mean residual of regional thickness in controls compared to patients. The histogram in each time bin represents mean residual of regional thickness between patients with the corresponding illness v. all controls. (b) Correlation matrices constructed from thickness measure for two groups, where the lower triangular matrix represents the correlation coefficients of patients and the upper triangular matrix represents the correlation coefficients of controls. (c) The percentage of the divergent pairwise relationships for different correlation thresholds for all possible pairwise connections in the two groups. (d) The error bar of normalized covariant coefficients for the top 100 significant links in each group. The solid line represents mean normalized covariant coefficients of each link. Controls are in the upper row, patients in the lower row.
Fig. 4. Hub-and-spoke representation of the covariance between regions with significant group difference and rest of the brain. The central node represents the region with significant increase (occipital) or decrease (parahippocampal and supramarginal) in thickness. The outer nodes represent all brain regions that have a significant covariance with the central node either in the patient group or the control group. Regions with increased thickness (diamonds) and decreased thickness (circles) are separately identified from those that show no significant group differences in thickness (octagons). The spokes represent the relationship between the central and outer nodes. Long dash spokes represent negative (divergence of covariance) relationships while solid spokes represent positive (convergence of covariance) relationships. The numbers shown in the figure correspond to specific brain regions as listed in Supplementary Table S1. A colour version of this figure is provided as Supplementary Fig. S4.

with the most significant thickness reduction that best discriminate patients with <2 years of illness from controls, do not show the same magnitude of group difference in those with longer duration of illness. In line with this obliteration, these regions do not retain their discriminatory ability when patients with longer duration of illness are compared with controls. Due to the cross-sectional nature of our sample, we cannot
infer whether this obliteration is taking place in the same set of individuals over the illness course. But these observations lead us to speculate that a secondary remodelling process that influences brain structure may be present in schizophrenia. In a large sample of patients with differing illness duration (range 6 months–29 years) followed up for 5 years, the rate of change of cortical thickness (reduction) was notably pronounced in patients, irrespective of their age (Van Haren et al. 2011). This higher rate of thinning over the course of 5 years indicates a dynamic morphometric process, but not necessarily ‘degeneration’ or a progressive deviation from normality. Interestingly, in line with our results, in that study several regions that showed higher baseline thickness revealed increased thinning over the course of illness, resulting in ‘normalization’ though this was not explicitly examined (Van Haren et al. 2011). Taken together, these observations contradict a ubiquitous degenerative process, and highlight the dynamic/plastic nature of the morphological changes that index a reorganization process whereby during the course of illness, the difference between patients and controls may become less pronounced.

In our sample, the most significant group differences in thickness involved the bilateral parahippocampal, supramarginal and temporal pole regions, and the pre-central and superior anterior insula regions, while increase in thickness was mostly limited to regions in the occipital cortex. Thinning of parahippocampal region is also a well-documented feature of schizophrenia (Kuperberg et al. 2003; Van Haren et al. 2011). Parahippocampal thinning may be related to the risk of psychosis, with many studies reporting parahippocampal thinning/volume reduction in ultra-high-risk subjects (Jung et al. 2011; Mechelli et al. 2011; Tognin et al. 2014) and in those with the genetic risk (Goghar et al. 2007; Palaniyappan et al. 2012). Both occipital thinning (Szieszko et al. 2012; Yoon et al. 2007; Mané et al. 2009; Van Haren et al. 2011; Xiao et al. 2015) and lack of thinning (Nesvåg et al. 2008) in schizophrenia has been noted in previous studies. In particular (Szieszko et al. 2012) found that patients responding to antipsychotic treatment have a higher occipital thickness than non-responders, suggesting that the occipital thinning is of prognostic importance. In line with this, Mitelman & Buchsbaum (2007) suggested that the loss of grey matter in the posterior aspect of the brain (including occipital cortex) is a feature of poor outcome in schizophrenia. Occipital thickening appears to arise early and is seen in conjunction with thinning in patients with untreated first episode schizophrenia (Xiao et al. 2015). While the plasticity of multimodal, representational cortex may be reduced in schizophrenia (Fisher et al. 2013) the increase in grey matter restricted to the visual cortex suggest that putative compensatory responses in schizophrenia may be more localized to the primary sensory cortices than in other regions.

We achieved a very high degree of accuracy (96.3% for the first 2 years) when discriminating patients from controls, after removing effect of age and gender. This high accuracy deteriorated steadily when all patients irrespective of their stage of illness were included in the discrimination model. We consider this finding as important in the context of increasing relevance of machine learning approaches in diagnosis and outcome prediction in schizophrenia (Lawrie et al. 2011; Borgwardt & Fusar-Poli, 2012; Koutsouleris et al. 2014). A number of studies, using varied neuroimaging features, report an accuracy ranging between 65% and 90% in discriminating patients from controls (Orru et al. 2012; Veronese et al. 2013; Zarogianni et al. 2013). Models that provide >90% accuracy can substantially reduce the proportion of uncertainty inherent to classification (Kasparek et al. 2011; Iwabuchi et al. 2013). For the first time, we have examined the effect of illness duration on the utility of machine learning in schizophrenia, and note that the features that differentiate patients from controls are likely to be specific to the stage of illness. A classifier to separate a patient with schizophrenia from healthy control is redundant for clinical use even if it is >90% accurate, as such an accurate separation can be readily made by clinicians. But it is worth noting that a great deal of investment is being made on prognostic prediction using neuroanatomical features. Our results suggest that acknowledging the process of obliteration of pathophysiological changes and incorporating stage-specific predictors could improve the sensitivity of classification approaches across various imaging modalities, and move us closer to the goal of using neuroanatomical features for prospective patient selection in future clinical trials in schizophrenia.

In clinical samples, patients in early stage of illness are likely to have varied outcome trajectories and thus more diverse pathophysiological signatures compared to patients with more established chronic illness. With chronicity the degree of clinical heterogeneity reduces to some extent as only those who require continued support remain with services. Despite this, we note that the neuroanatomical features in chronic sample are less informative than the pattern seen in earlier stages, suggesting that the degree of putative anatomical reorganization in more representative samples may indeed be greater than that captured by this study.

Our study has several strengths. The sample was recruited from a single site, with identical acquisition parameters used in the two scanners and identical
image processing. Further, we demonstrate (Supplementary Table S7) that the scanner variation does not influence the observed results. We studied the entire cortical surface without restricting our analysis to a priori regions. Studies estimating lobar volumes are likely to miss the reciprocal regional changes; in contrast, we examined anatomically defined sulcogyral subdivisions to capture thickness changes in both directions.

The limitations of the present study include the lack of longitudinal data to estimate illness onset, exposure to recreational substances and antipsychotic medications in a systematic manner. Structural changes are suspected to be linked to exposure to antipsychotics (Ho et al. 2011). We did not have the data for cumulative antipsychotic exposure for this sample. Nevertheless, when the effect due to the prescribed antipsychotic dose in our sample of clinically stable patients (dose unchanged at least for a period of 6 weeks at the time of scan) was removed from the analysis, the results were not altered (Supplementary Fig. S3). We cannot conclude whether the putative ameliorative changes reported here depend on exposure to antipsychotics. Longitudinal follow-up, ideally starting during the prodrome, is the best design to address the question of compensatory brain changes; cross-sectional data such as the one presented here are likely to be affected by a confound between generational or cohort differences and age. We controlled for age in our analyses, but acknowledge that this may not be sufficient to eliminate the cohort effects in a cross-sectional design. To date, most longitudinal structural studies have had a modest range of follow-up periods (median time of 2.4 years), and rarely capture the brain changes right from the onset or beyond 7 years of illness (Vita et al. 2012). In addition, multiple changes in image acquisition procedures that often take place during longitudinal studies greatly affect the ability to observe spatially circumscribed changes (as opposed to total brain volume or lobar volume) in a reliable manner (Takao et al. 2013). The neurohistological basis of changes in MRI-derived cortical thickness is unclear, although this measure corresponds closely to biopsy-based measurement of cortical thickness (Cardinale et al. 2014). Nevertheless, in the wake of the limitations discussed here, our results regarding putative ‘normalization’ at later stages of schizophrenia must be considered preliminary until examined in a follow-up design.

Neuroimaging-based machine-learning approaches are considered as promising tools for a stratified approach in psychiatry (Lawrie et al. 2011; Borgwardt & Fusar-Poli, 2012; Koutsouleris et al. 2014). Considering the neuroanatomy of schizophrenia as a dynamic reorganization rather than static alteration will greatly improve the accuracy of diagnostic/outcome prediction and the clinical utility of MRI scans. Our observations also raise the possibility that if the compensatory brain changes can be further characterized in schizophrenia, harnessing brain’s plasticity optimally for therapeutic purposes may become feasible in the near future.

Supplementary material
For supplementary material accompanying this paper visit http://dx.doi.org/10.1017/S0033291716000994.

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Declaration of Interest
All authors report no biomedical financial interests or potential conflicts of interest.

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