Visualization of Cortical Lamination Patterns with Magnetic Resonance Imaging

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The ability to image the cortex laminar arrangements in vivo is one of the holy grails of neuroscience. Recent studies have visualized the cortical layers ex vivo and in vivo (on a small region of interest) using high-resolution $T_1/T_2$ magnetic resonance imaging (MRI). In this study, we used inversion-recovery (IR) MRI to increase the sensitivity of MRI toward cortical architecture and achieving whole-brain characterization of the layers, in vivo, in 3D on humans and rats. Using the IR measurements, we computed 3D signal intensity plots along the cortex termed corticograms to characterize cortical substructures. We found that cluster analyses of the multi-IR images along the cortex divides it into at least 6 laminar compartments. To validate our observations, we compared the IR-MRI analysis with histology and revealed a correspondence, although these 2 measures do not represent similar quantities. The abilities of the method to segment the cortex into layers were demonstrated on the striate cortex (visualizing the stripe of Gennari) and on the frontal cortex. We conclude that the presented methodology can serve as means to study and characterize individual cortical architecture and organization.

Keywords: BAs, cortical layers, IR-MRI, microstructure, rat and human brain

Introduction

More than 100 years ago, Brodmann, Von-Economo, and others discovered the cellular cytoarchitecture of neuronal tissues (cortical cytoarchitecture) (von Economo and Koskinas 1925; Brodmann and Garey 1999). Six distinct layers of cell structures were identified, having similar composition and order throughout the cortex but differing in their fractional volumes between regions. This histological, cytoarchitectonic feature of the cortex is the basis of its parcellation into neuroanatomical regions (e.g., Brodmann’s areas [BAs]). The cortical layers can be characterized not only by their cytoarchitecture but also by the myeloarchitecture (the density and content of myelin across the cortex) (Vogt 1911). Such staining was also the base of brain parcellation and provided similar neuroanatomical maps to the cytoarchitecture segmentation (Annese et al. 2004).

There is a growing interest to obtain an in vivo imaging modality that enables visualization of macroscopic cell structure arrangements for the whole brain and in 3D. Magnetic resonance imaging (MRI) is the noninvasive, in vivo imaging modality that provides the best spatial anatomical details of the human brain. Visualizing the cortical layers presents various challenges: 1) extremely high resolution is required to resolve the layers, leading to long acquisition times and the need for a high magnetic field to overcome low signal-to-noise levels; 2) MRI detects water and thus is not specific to cell morphology, in general and to neuronal cell architecture in particular; and 3) the MRI signal profile along the cortex at conventional methods and resolution seems to be iso intense and therefore cannot be segmented into cortical substructures.

Despite these limitations, $T_1$ and $T_2$ MRI contrast mechanisms were suggested for discrimination between the layers. Already in 1992, Clark et al. (1992) showed that high-resolution ($\sim 400$ $\mu m^3$) $T_1$-weighted MRI can be used to visualize the border between the striate (BA 17) and extrastriate cortex (BAs 18 and 19). The ability to discriminate between these regions was attributed to the heavily myelinated layer IV of area 17 (striata of Gennari). A discrimination of the striata of Gennari was obtained by $T_2$-weighted images as well (Barbier et al. 2002; Walters et al. 2003), indicating that high-resolution MRI is potentially sensitive to the cortical microstructural arrangement and architecture. Indeed, this observation was recently demonstrated on the mouse brain (Boretius et al. 2009).

Several works followed this approach and acquired high-resolution anatomical MR images of excised cortical tissue and small sections of in vivo human brains (Barbier et al. 2002; Fatterpekar et al. 2002; De Vita et al. 2003; Walters et al. 2003; Bridge et al. 2005; Clare and Bridge 2005; Eickhoff et al. 2005; Bridge and Clare 2006; Duyn et al. 2007). Variations of the MR signal profiles perpendicular to the cortex were suggested to be sensitive to the underlying microstructure and cyto/myeloarchitectonic arrangements (Barbier et al. 2002; Fatterpekar et al. 2002; Walters et al. 2003; Eickhoff et al. 2005; Duyn et al. 2007). Nevertheless, the contrast between the layers in most cortical regions as seen in the $T_1$ or $T_2$-weighted cortical profiles is not strong enough to allow a robust whole-brain segmentation of the cortical layers with MRI, leading to the conclusion that increased image contrast is needed. A few works did attempt to enhance the contrast between the layers by using contrast agents that are less applicable for routine human brain imaging (Angenstein et al. 2007; Silva et al. 2008).

In the present study, we utilized inversion-recovery (IR) MRI to acquire multi $T_1$-weighted images with increased contrast between cortical substructures (for additional background on the IR-MRI method, please refer to the Supplementary Material—additional background). As indicated above, it is well-accepted that myelination causes shortening of the $T_1$, and $T_1$ will therefore decrease from the outer to the inner layers of the cortex (Clark et al. 1992). We exploited the $T_1$ properties...
of gray matter (as imaged with IR-MRI) to discriminate between laminar compartments within the cortex. To that end, we hypothesize that cortical substructures will have distinct $T_1$ properties that become visible and analyzable using the high contrast of IR-MRI. In the present work, we argue that by using a series of IR images, a laminar pattern of cortical substructures can be obtained for the whole brain, in 3D. We propose a framework to visualize (corticogram) and quantify (by cluster analysis) such information. We show that our approach can identify and quantify cortical substructures. We applied this approach to discriminate between the cortical laminar compartments in vivo in human and rat brains and validated the results with histology.

Materials and Methods

Subjects

Humans

Fifteen healthy subjects were recruited for this study, 8 males and 7 females, 25–35 years of age. Subjects were neurologically and radiologically healthy, with no history of neurological diseases, and normal appearance of clinical MRI protocol. The imaging protocol was approved by the institutional review board of the Tel Aviv Sourasky Medical Center, where the MRI investigations were performed. All subjects provided signed informed consent before enrollment in the study.

Rats

Two male, 7-month-old Wistar rats were used in this study. During the imaging protocol, the rats were anesthetized with approximately 2% isoflurane (in oxygen). Following the MRI, the rat brains were euthanized and prepared for histology (see below). The protocol was approved by the Tel Aviv University ethics committee on animal research.

Magnetic Resonance Imaging

Human Experiments

MRI was performed on a 3T General Electric (GE) scanner with an 8-channel radio frequency coil and a 50 mT/m gradient system. The protocol included conventional clinical screening sequences (sagittal spin-echo $T_2$, axial fast spin-echo [FSE] $T_2$, and axial fluid attenuated inversion recovery [FLAIR]) covering the whole brain and lasting 15 min. Then, a series of IR FSE images were acquired. Here, we utilized the inversion time ($T_1$) parameter in IR to produce a series of different image contrasts. More in-depth description of the IR methodology is given in the Supplementary Material. The IR FSE sequence was acquired with the following parameters: time repetition (TR)/echo time (TE) = 11 000/14 ms with TI varied at the following values: 230, 432, 575, 760, 920, 1080, and 1380 ms. The TI of 230 ms was used to produce white matter (WM) nulled image. One whole left hemisphere acquisition was done in the sagittal plane with a matrix of 512 × 512 (reconstructed to 256 × 256) with a final pixel size of 125 × 125 µm² and slice thickness of 400 µm. Total acquisition time for the IR-MRI data set was around 1 h.

Image Analysis

$T_1$ Analysis

$T_1$ analysis was done on a pixel-by-pixel basis by fitting the entire IR data set to an exponential decay function, according to the following formula: $M(T_1)/M_0 = 1 - 2e^{-T_1/T_1}$, Where $M(T_1)$ is the magnetization at the $i$th TI and $M_0$ is the magnetization at $T_1 = 0$ ms. Following successful fitting, each pixel was assigned a single $T_1$ value. Although an optimal calculation of the $T_1$ will require a more lengthy experiment, we used this analysis to get an estimate of the relative $T_1$ variations along the cortex.

Corticogram

Characterization of the IR properties along the cortex was achieved by a 3D plot of IR signals variation along the cortex (corticogram). In that plot, a rectangular region of interest (ROI) perpendicular to the cortex was defined; the $y$-axis of this ROI represents the cortical axis from the outer edge to its inner edge, whereas the $x$-axis represents the region borders. The signal was averaged along the $x$-axis and plotted in 3D versus the TI and the cortical axis ($y$-axis of the ROI). This procedure was replicated for different cortical regions and emphasizes the power of IR-MRI in characterizing the cortical architecture and distinguishing between different cortical regions.

Cluster Analysis

Visualization and quantification of the lamination structure from the multi-IR images was done according to a multispectral image analysis (Yovel and Assaf 2007). Briefly, following the coregistration (to correct for head movements), the cortex was automatically segmented based on the SPGR, FLAIR, and WM-zeroed IR images. Basic contrast enhancement was used on the segmented cortex to exclude outlier pixels, thereby stretching the dynamic range of the image. Cluster analysis based on the multiparametric data was then performed, including the following steps: 1) normalizing the data to create a uniform scale between the different imaging methods, 2) transforming the data into its principle component analysis space to increase the variance, and 3) running a clustering algorithm such as ‘‘$k$-means’’; the number of $k$-cluster is set to 5 or 6 according to the results of the Gaussian mixture analysis (Fig. 1 and see Supplementary Fig. S1).

Comparison with Brodmann’s Map

The results of the MRI methodology were validated by comparing the relative thickness or fractional volume of each of the IR clusters with histology data according to the standard Brodmann (Brodmann and Garey 1999) cytoarchitectonic segmentation as an histological reference. A template of Brodmann’s map (MRicron, Chris Roden) was registered and normalized to the high-resolution SPGR of each subject, so the cortical substructures’ regional clusters could be assigned to different BA’s. We report here this analysis on 10 BA of the frontal lobe, although a similar analysis can be applied to any set of regions. Statistical analysis was performed using a factorial 5 (IR layers) by 10 (BA regions) repeated-measure analysis of variance (ANOVA).

Histology

The rat brains were extracted and fixated with formalin 4%, washed with tap water, and dehydrated with a graded series of ethanol (70%, 96%, and 100%) and chloroform until they sank to the bottom of the test tube. The brains were then embedded in a paraffin solution (Merck, NJ) for a few days for hardening, after which the whole brain was cut serially in the coronal plane (2-µm section thickness) with a sliding microtome (Leica Microsystems SM2500, Nussloch, Germany). The sections were mounted on gelatin slides and postfixed with a solution of sodium phosphoric acid and formalin 37% over 2 days. A deparaffinization phase was executed by Xylol and a downgraded
series of ethanol (100%, 96%, and 70%), and sections were washed twice with distilled water for 15 min.

Pretreatment for cytoarchitecture staining was performed by placing the slides in formic acid 4% for 3 h. The slides were then stained overnight in a mixed solution of formic acid, double distilled water (DDW), and hydrogen peroxide, followed by several washings with DDW. The slides were either stained for cell bodies using the modified Gallyas staining for paraffin sections (Merker 1983) or stained for myelin with the Gallyas silver technique (Gallyas 1979), modified for slide-mounted tissue, similar to Zilles (Zilles 1985). The cytoarchitecture and myelo architecture sections of the whole brain were digitized with a stereotactic microscope (V12 Lumar, Zeiss, Oberkochen, Germany).

Results

The main finding of this study is that multi IR-MRI can provide unique image contrasts that enable visualization and characterization of laminar compartments in human and rat cortices. The first evidence that IR-MRI is able to divide the human cortex into substructures is based on a histogram analysis of the $T_1$ values throughout the cortex. The cortex is characterized by a wide range of $T_1$ values rather than a specific one. Such histogram shows that the $T_1$ values are distributed into at least 6 clusters both in human and rat brains (Fig. 1). It should be indicated that this pattern was repeated in all human subjects and at different image resolutions (see supplementary material, additional results section—Supplementary Fig. S1), ruling out the possibility that the distribution stems from voxel partial volume effects (PVE). We exploited the results presented in Figure 1 and Supplementary Figure S1, to acquire a series of IR images at different TI that yield contrasts with high sensitivity to each of the $T_1$ components.

Figure 2 demonstrates the IR data set and analysis pipeline on rat cortex. Figure 2A shows IR-MRI images of a representative slice at different TIs depicting the large contrast shift in the cortex at different TIs. Note that at each TI, different parts of the tissue are being null ed (e.g., the CSF intensity is zeroed at TI of 2000 ms, bottom-right image at Fig. 2A). Figure 2B demonstrates the $k$-means clusters ($k = 8$) of a subset of the rat cortex. The 8 clusters represent 6 cortical laminar compartments, the white matter, and the CSF. The clusters found within the hippocampus and the subcortex were combined into a separate cluster (no. 9) for visualization purposes. To understand the basis for the cluster arrangement, we plotted the IR signal intensities (as shown in Fig. 2A but only for a subset of the cortex) on a mesh-grid colored according to their cluster assignment. Using such presentation, the $T_1$ characteristics of each cluster are better visualized. For example, layer no. 5 (pale green) is distinguishable from adjacent layers (6—light orange and 4—light blue) mainly by the contrast in TIs 1080 and 860 ms. Figure 2D shows the cluster analysis for the entire cortex of the right hemisphere. On top of the clusters, we marked 4 cytoarchitectonically defined regions: PA (parietal association cortex), S1BF (primary sensory cortex, barrel field area), AuD (secondary auditory cortex, dorsal area), and Au1 (primary auditory cortex). Variations between these regions were demonstrated by the corticograms plots (Fig. 2E) based on the IR signal along the cortical axis, which allow their differentiation. These corticograms indicate that the IR profiles of different cortical regions are different and that within a region, the variation of the IR intensity along the cortical axis can be used to define the border between layers (see white arrows in AuD corticogram).

Figure 2 shows that IR-MRI enables visualization and quantification of cortical substructures; yet, their relationship to the histological cytoarchitectonic layers is not straightforward. To that end, we compared rat brain IR-MRI analysis with histological cyto- and myeloarchitectonic analysis. As shown in Figure 3 on a representative slice, changes in the lamination pattern of the MRI segmentation appeared to follow the borders between cyto- and myeloarchitectonically defined cortical regions. However, as expected, the IR layers do not have a one-to-one correspondence with the histologically defined regions, since the 2 techniques are not measuring exactly the same parameter. Nevertheless, this comparison demonstrates that there is a good relationship between the histological and IR layers and that the clustering of the MRI data into 6 layers is adequate.

Figure 1. $T_1$ distribution of the cortex. (A) A histogram of the longitudinal relaxation time ($T_1$) distribution across the entire cortex averaged for 6 subjects (that were measured with slice thickness of 1.5 mm) scanned at 3-T MRI scanner. Gaussian mixture analysis revealed that the histogram can be characterized with 6 distinct Gaussians centered at 660, 820, 970, 1080, 1220, and 1540 ms. Error bars represent standard deviations across subjects. (B) $T_1$ histogram across the rat cortex scanned at 7-T MRI scanner. Similar multi-Gaussian pattern was revealed with 5 distinguishable peaks. Note that the cortical $T_1$ characteristics were shifted toward $-1500$ ms due to the magnitude of the magnetic field (7T).
Figure 2. The cortical $T_1$ characteristics of the rat brain. (A) An IR data set of the rat brain (at TI of 480–2000 ms) was used for $k$-means clustering analysis (B). Eight clusters were found representing 6 cortical laminar structures, the white matter, and the CSF. The clusters found within the hippocampus and the subcortex were combined into a separate cluster (no. 9) for visualization purposes. (C) IR-MRI signal intensities mesh plots at different TIs of the cortical region shown in (B) colored according to their clustering assignment. This presentation strengthens the clustering results where each cluster is nulled at different TI. For example, CSF (marked in red) is eliminated at TI of 2000 ms, while exists at other TIs. (D) Cluster analysis for the entire cortex of the right hemisphere with 4 cytoarchitectonically defined regions: PtA (parietal association cortex), S1BF (primary sensory cortex, barrel field area), AuD (secondary auditory cortex, dorsal area), and Au1 (primary auditory cortex) indicated in dashed squares. Variations between these regions were demonstrated by corticograms (E) based on the IR-MRI signal perpendicular to the cortical axis, which allow their differentiation.

Figure 3. A comparison of IR-MRI clusters with cyto- and myeloarchitectonic analysis of the rat cortex: (A) cyto- and (B) myeloarchitectonic analysis of the rat cortex; red lines denote the cytoarchitectonic border between regions, and blue lines denote the border between the different layers. (C) Cluster analysis of an IR data set acquired in vivo on the rat brain; the white and yellow markings refer to the cytoarchitectonic analysis performed on histological sections of the rat brain shown in (A) and (B). Note that the IR layers resemble the histological laminar appearance, although there are some mismatches. Data indicate that the MRI laminar patterns are able to define some of the cytoarchitectonically defined borders between cortical regions.
In order to demonstrate the ability of the IR contrast to resolve the cortical architecture in the human brain, we chose to focus on BA 17 (Fig. 4A,B). In that BA, the heavily myelinated layer IV (also termed stria of Gennari) becomes distinguishable in the TI range of 575–760 ms, known to zero most of the cortex signal. Note the bright signal band that indicates a region with similar $T_1$ characteristics to white matter tissue. (D) Clustering analysis focused on this region ascribe a specific cluster representing the stria of Gennari (red cluster). (E) Corticograms were computed for the clusters ROI shown in (D) based on the IR-MRI signal at different TIs along the cortical axis. The corticogram presentations allow detection of the stria of Gennari at several TIs according to the multipeak pattern along the cortical axis, where the stria of Gennari IR signal is similar to white matter tissue. The corticograms demonstrate the ability of IR-MRI to detect and characterize tissue’s underlying architecture.

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In the following analysis, a comparison between the clusters’ composition in different BAs was performed. Ten regions of the frontal lobe were selected for analysis: BAs 4, 6, 8, 9, 10, 11, 44, 45, 46, and 47. The volume fraction of each of the MRI-IR clusters was calculated for these regions and a 5 (cluster volume fraction) x 10 (BAs) repeated-measures ANOVA performed for all subjects (see Supplementary Material, additional results—Fig. S4). The interaction between the region and IR layer (cluster) volume fraction was highly significant.
indicating that the regions can be differentiated using the IR clusters volume fractions. Post hoc pairwise ANOVA comparison between regions revealed that 31 of the 45 possible comparisons turned out to be statistically significant (see Supplementary Table S2). The comparisons between adjacent regions are of particular interest, since if they are statistically significant, the IR layers can be used to outline the border between these regions. Eleven of the 17 comparisons between adjacent BA regions were statistically significant (9 when correcting for multiple comparisons) (Fig. 6). As the figure shows, the frontal lobe can be segmented using the IR layers into 5 regions: region 1 includes BA4, region 2 includes BAs 6 and 8, region 3 includes BAs 9, 10, and 11, region 4 includes BAs 44, 45, and 46, and region 5 includes BA 47. Thus, the IR composition of the layers differentiates between anatomical regions.

Discussion

The main result of this work is that a laminar pattern of cortical substructures can be identified and characterized with IR-MRI, in vivo for the whole brain, in 3D. The visualization and characterization of the cortical lamination pattern are based on the $T_1$ properties of the cortex. Using the high contrast of IR-MRI taken at different TIs, the IR clusters appeared to have a laminar structure, with the cluster that is characterized with low $T_1$ lying adjacent to the white matter and the cluster with the high $T_1$ lying adjacent to the CSF. This observation was obtained at various image resolutions and hence could not be attributed to image PVE. If PVE was the origin of this observation, it would have disappeared or decrease at higher resolution yet the phenomena sustain at a wide range of image resolution values. Moreover, the cortical laminar compartments were reproduced in all subjects and in both the human and rat brains. We demonstrated the ability of this method to define cortical substructures on several brain regions with different layer architecture: striate cortex (Fig. 4), allocortex (Supplementary Fig. S2), and motor cortex (Supplementary Fig. S3).

Although cortical architecture has been demonstrated with various MRI methodologies (Barbier et al. 2002; Fatterpekar et al. 2002; De Vita et al. 2003; Walters et al. 2003; Bridge et al. 2005; Clare and Bridge 2005; Eickhoff et al. 2005; Bridge and
mismatches. This is not surprising since $T_1$ is not a direct measure of cellular architecture but rather describes the ability of the water molecules’ protons to pass their magnetization to the surrounding environment. Thus, it is reasonable to expect that this parameter will be affected by tissue architecture, cellular density, and water/fat ratio (i.e., extent of myelin).

The fact that the $T_1$ values become shorter as one moves from the outer parts of the cortex (layers I, II) to its inner parts (layers V, VI) suggests that the tissue become more dense and rigid (evidenced by the increase of myelinated axons) and viscous, making the relaxation time shorter. However, although there is a correlation between $T_1$ and amount of myelin, one should be cautious about linking the 2: $T_1$, even when measured at high resolution as in this work, averages the contributions of many cellular components including the neurons and their processes and the glial cells and their processes. Thus, even though the degree of myelination might contribute significantly to $T_1$, it is not the sole parameter, and $T_1$ might represent both myelo- and cytoarchitecture, as was indicated in previous studies (Barbier et al. 2002).

The widths of the different lamellas of IR-MRI change significantly between cortical regions (as defined by Brodmann) (Figs 5 and 6), indicating that the IR layers have a morphological meaning and can be potentially used for regional cortical architecture studies. The segmentation of the frontal lobe into zones (e.g., BA’s 6 and 8 in Fig. 6) indicates that those can be distinguished based on their IR lamination pattern. It is expected that the distinguishability between adjacent cortical regions will be different between IR-MRI and histological segmentation. Thus, the biological meaning of cortical segmentation of IR-MRI still needs to be explored. It should be noted that by acquiring additional IR images or with higher resolution (probably at higher magnetic fields $>7$T), a more robust segmentation into additional regions might become feasible.

The morphological characteristics of the cortex represent a fingerprint of brain development and affect cognitive abilities and might be useful for the study of disease. The in vivo measurement of this quantity will have a tremendous impact on neuroscience, including neurobiology, psychology, and neurology. The current knowledge on brain neuroanatomical organization is gleaned from a surprisingly small number of postmortem histological samples from subjects who underwent whole-brain cytoarchitectonic analysis. Yet only in such measurements, the cytoarchitecture could be defined. However, it is argued that these samples are hardly representative of the entire population. Without underpinning the use of histological method to study brain parcellation, in vivo measurement of cortical substructures might enable large population studies in which intersubject variability can be assessed and correlated with subject-specific cognitive abilities or behavior. As brain function and brain architecture are linked (i.e., the functional and anatomical organizations of the brain are similar), measuring variations in the anatomical organization of the brain may help explain functional and behavioral differences. For example, measuring and understanding the cytoarchitecture manifestation of learning and memory, and the development of high-order cognitive skills, language, and other mental abilities, would have a huge impact on our understanding of brain function and plasticity.

Further research is needed on the methodology presented here for it to become a tool for in vivo assessment of cortical architecture.
substructures and lamination patterns. Intersubject variability over a large population data set needs to be determined. Indeed, studying the regional variation in various neurological and psychiatric diseases and disorders such as dyslexia, autism, schizophrenia, and Alzheimer’s disease might shed new light on a neuroanatomical component of these diseases and its relation to disease pathogenesis and etiology. This line of research should also examine the age dependency of the layers’ formation and the development of blurring between the gray matter–white matter border frequently observed at elderly subjects. Still another basic issue is the regional developmental profile of the layers from the postnatal stage to adulthood. With the potential embedded in the IR-MRI segmentation of the cortex, the above-mentioned issues could be systematically explored over large population studies. Those studies will determine the sensitivity and specificity of the proposed methodology toward cortical architecture and its temporal dynamics.

**Supplementary Material**

Supplementary material can be found at: http://www.cercor.oxfordjournals.org/.

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**Notes**

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**References**


