Towards probabilistic atlases of the T1/T2 relaxation times from the CONNECT/ARCHI database

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Purpose The ARCHI Database [1] is a large human brain MRI database (79 healthy subjects) acquired under the framework the European CONNECT project which aims at inferring the connectome atlas of the human brain from functional and anatomical MRI data at 3T. Among the data, T1, T2 and relaxometry dataset have been acquired for all the subjects. To our knowledge, there does not exist any atlas of the relaxation times. This work is therefore focused on the construction of two probabilistic atlases of the quantitative T1 and T2 parameters of the human brain from the CONNECT/ARCHI database. We combined these atlases with the fiber density maps of the major white matter bundles we extracted from the diffusion data of the same subjects to compute the mean T1/T2 for each bundle.

Methods Data were collected on a population of 79 healthy young subjects (CONNECT/ARCHI database), under the framework of the European connectome project (CONNECT), on a Tim Trio 3T MRI system with a 12-channel head coil (Siemens, Erlangen), and the MRI protocol included the acquisition of two datasets: 1) *a relaxometry T1 dataset*: SE-EPI-single-shot / Fov=220mm; TH=1.7mm; 70 slices; TE/TR=30ms/20.6s; 128x128; GRAPPA2; PF=5/8; RBW=1502Hz/Pixel; 10 TI between 300ms-3000ms; 2) *a relaxometry T2 dataset*: GE-EPI-single-shot / Fov=220mm; TH=1.7mm; 70 slices; TR=23.2s; 128x128; GRAPPA2; PF=6/8; RBW=1502Hz/Pixel; 10 TE between 30ms-200ms

The data have been corrected for susceptibility and the T1/T2 maps of each subject have been computed by fitting the following two equations to the T1 and T2 datasets, respectively: $S=\rho^*(1-2^*exp(-TI/T1))$ (T1 relaxometry); $S=\rho^*exp(-TE/T2)$ (T2 relaxometry). A Levenberg-Marquardt algorithm was used to obtain the optimum parameters at the voxel level and we took care to adequately choose the initial parameters to take account of the high sensitivity of the algorithm to its initialization. We initialized T1/T2 close to known values of the literature at 3T: T1=1000ms, T2=90ms. All the subjects were coregistered using the diffeomorphic tensor-based technique provided in DTI-TK [2] using the HARDI data included in ARCHI database. We used the IXI template [3] to define a common frame for all the subjects. The individual T1 and T2 quantitative maps were then transformed to this frame in order to create a novel relaxometry atlas. We finally used this atlas in conjunction with the probabilistic maps of the different bundles described in [1] to compute the statistics of their T1 and T2 relaxation time. The probabilistic fiber density maps naturally define the regions of interest corresponding to the various WM fibre bundles.

Results & Discussion Fig 1 shows the T1/T2 quantitative maps for one axial slice from the atlases. The maps clearly depict a good contrast between the various brain structures. Fig 2 shows the bundle atlas we employed to perform the automatic bundle segmentation and described in [4]. Table 1 summarizes the statistics (mean and standard deviation) of T1 and T2 obtained for a series of well-known WM bundles of interest. In future work, we will take into account a priori anatomical information to correct partial volume effect that may affect the mean T1/T2 values for small bundles. No T1/T2 atlases have been provided to the community for now with such a large population. Combining these relaxometry data with the diffusion data of this database could be of a great interest to microstructure studies. We showed here one of the potential application, namely the possibility to

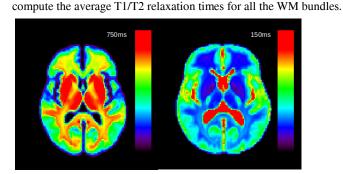


Fig1 a. T1 Relaxometry atlas b. T2 Relaxometry atlas

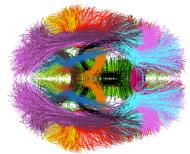


Fig2. Bundle atlas

Bundles T1 (ms) T2 (ms) Arcuate Anterior Left 596.257+-111.79 77.2202+-13.2459 Arcuate Anterior Right 598.477+-104.491 77.0886+-12.7391 Arcuate Right 644.837+-60.6078 72.2389+-7.63583 Arcuate Left 621.403+-81.7891 71.9173+-5.90902 Arcuate Posterior Left 523.559+-127.653 80.114+-10.8715 Arcuate Posterior Right 545.2+-110.9 78.0578+-8.85713 Cingulum Long Right 591.541+-115.351 70.4856+-6.82386 594.832+-109.505 Cingulum Long Left 70.3922+-6.74432 Cingulum Short Left 537.935+-137.194 75.2391+-10.5083 Cingulum Short Right 539.14+-138.412 74.2926+-10.0084 Cingulum Temporal Right 558.603+-168.904 88.8107+-29.2417 Cingulum Temporal Left 580.539+-146.432 86.3448+-24.2756 CorpusCallosum Splenium 612.022+-109.48 95.957+-53.798 CorpusCallosum Genu 534.173+-140.854 95.4439+-56.9274 CorpusCallosum Rostrum 539.599+-111.854 83.6835+-44.3861 CorpusCallosum Body 575.399+-128.337 75.7437+-32.5826 CorticoSpinalTract Left 653.052+-111.093 80.0061+-22.4178 CorticoSpinalTract Right 651.087+-113.588 81.6412+-24.2388 Fornix Left 470.705+-200.348 149.393+-101.835 Fornix Right 467.513+-209.468 148.871+-105.42 InferiorFrontoOccipital Left 652.348+-67.6198 74.698+-9.69363 InferiorFrontoOccipital Right 657.95+-64.387 75.6785+-11.7558 InferiorLongitudinal Left 607.984+-92.9507 71.8923+-18.2364 InferiorLongitudinal Right 622.621+-87.0801 73.307+-17.3845 Uncinate Right 598.165+-151.511 71.5941+-9.56687 Uncinate Left 613.999+-138.806 71.8141+-10.1184

Table 1. Mean T1 and T2 and standard deviation values for well-known bundles

References [1] B. Schmitt et al, ESMRMB 2012; [2] H. Zhang et al, MedIA 2006 10(5):764-85;[3] H. Zhang, WBIR 2010 [4] P. Guevara et al, 2012, NeuroImage 61: 1083–1099