

Functional network interactions during typical development in infancy and early childhood

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Target Audience: MR Image Analysts, Pediatric Imaging Neuroscientists, Neuroimaging Scientists

Purpose Functional networks derived from resting-state fMRI are presumed to reflect the specific cognitive and sensory networks of the brain. Developmental changes in both integration and segregation of these networks have been demonstrated in infants¹ and older children². Using a large cohort of typically developing infants and toddlers in an under-sampled cross-sectional age range (3 months to 5 years) we investigate the effect of maturation on inter-network coupling.

Methods Sample 80 infants, toddlers and young children aged between 84 and 1619 days (mean (SD) age of 688d (477d), 34 female) were scanned during natural sleep³. Imaging Resting state fMRI data were collected using a gradient echo-planar imaging sequence (TR=2.5, TE=34, FA=80°) with resulting resolution of 3mm in-plane over 32 3.6mm slices. A high flip angle T₁-weighted SPGR image collected as part of a multicomponent relaxometry sequence (mcDESPOT⁴) was used for registration purposes (TR/TE=5/10ms, FA=14°, 1.7mm isotropic resolution). Analysis After standard fMRI preprocessing (motion and slice timing correction), functional images were transformed to a common space in 2 stages. Initially, an example functional image was registered to the anatomical T₁ image using boundary-based registration and this T₁ image was then registered non-linearly to an age-appropriate template⁵. Finally spatially normalised data were smoothed to a global 5mm FWHM using 3dBlurToFWHM, ensuring that the effective spatial resolution was the same for all subjects⁶. fMRI data were analysed using independent component analysis (ICA⁷) and the dual regression and FSLNets packages⁸. Motion parameters and motion spikes (>0.2mm interscan motion) were included as confounds. In this *exploratory* analysis, we tested the relationship between functional coupling between ICA-derived networks and both age and cognitive ability in the same children (measured using the Mullen Scale). To account for multiple testing and comparisons, false discovery rate was applied across all contrasts at $q < 0.05$.

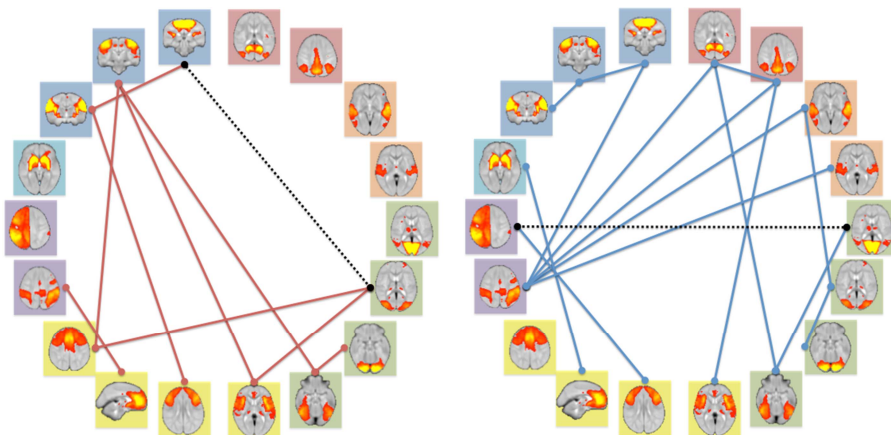


Figure 1: Functional networks from ICA decomposition shown colour coded according to gross functional system (motor blue, striatal aqua-marine, fronto-parietal purple, frontal yellow, visual green, auditory orange, and default red). Red lines (left) and blue lines (right) indicate networks demonstrating increasing and decreasing coupling with age respectively ($p < 0.05$ corrected).

significantly positive relationships). Networks showing decreased coupling tended to be spatially adjacent to each other (especially clear for sensory networks), though the left-dominant fronto-parietal especially and default mode networks showed decreased coupling with spatially disparate regions. We hypothesized that there would be a relationship between coupling in resting state networks and cognitive scores, but only one positive (figure 1, left, dashed line) and negative (figure 1, right, dashed line) relationship was detected.

Conclusions: Our results demonstrate age-varying interactions between functional networks, replicating trends seen in younger¹ and older² cohorts, though with differing patterns. This indicates distinct functional network changes during this important stage of development. Individual differences in network coupling do not show *strong* relationships with cognition in this typically developing cohort.

References: 1 Gao et al, Brain Structure & Function, 2014 | 2 Fair et al, PLoS Comp Biol, 5, e1000381 | 3 Manning et al, Neuroimage, 83, 288-293 2013 | 4 Deoni et al, Neuroimage, 63, 1038-1053, 2012 | 5 Avants et al, Neuroimage, 49, 2457-2466, 2010 | 6 Scheinost et al, Neuroimage, 95, 13-21 | 7 Beckmann & Smith Neuroimage, 25, 294-311, 2005 | 8 <http://fsl.fmrib.ox.ac.uk/fsl/fslwiki/FSLNets>