Supplementary Information

Typical and disrupted brain circuitry for conscious

awareness in full-term and preterm infants

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Supplementary Methods

Participants

Full-term neonates: Data from 342 full-term neonates were available. Only one subject was scanned twice, at 37 weeks postmenstrual age (PMA) and after 37 weeks PMA, but we only used the later one for the full-term group. 61 scans were discarded because of excessive movement (see Data Analyses section). Thus, the full-term neonate group had 282 participants (gestational age (GA) at birth = 40.0 weeks \pm 8.6 days; PMA at scan = 41.2 weeks \pm 12.0 days; 160 males)*.*

Preterm neonates. 121 preterm neonates had fMRI data in the second dHCP public data release. 74 of them were scanned once: 40 participants scanned at TEA and 34 participants scanned before TEA. 47 of the 121 preterm neonates were scanned twice, at and before TEA. Thus, for preterm neonates, we have 87 obtained at TEA and 81 before TEA. 14/87 scans collected at TEA and 8/81 scans collected before TEA were discarded because of excessive movement (see Data Analyses section).

Adults. All participants were right-handed, native English speakers, and had no history of neurological disorders. All participants were tested at Washington University in accordance with the protocol approved by the Washington University institutional review board. Informed consent was obtained for each participant prior to the experiment. Specific details and procedures of subject recruitment can be found in Van Essen et al.¹. The subset used in the current study passed stringent quality control measures relative to the larger HCP.² Detailed information about exclusion criteria can be found in Ito et al.² and the full list of the 176 participants used in this study is available here: [https://github.com/ito-takuya/corrQuench.](https://github.com/ito-takuya/corrQuench)

Data Acquisition

dHCP data acquisition. Full details of dHCP data acquisition can be found at Fitzgibbon et al.³. Prior to scanning, neonates were fed, swaddled, and comfortably positioned in a vacuum jacket to promote natural sleep.

HCP. Full details of HCP data acquisition can be found at Van Essen et al.¹. Rs-fMRI data were collected in four runs of 14.4 minutes each, two runs in one session and two in another session. The two sessions were conducted in two days separately and we only used the data collected in the first session. Within each session, oblique axial acquisitions alternated between phase encoding in a right-to-left (RL) direction in one run and phase encoding in a left-to-right (LR) direction in the other run.

Data pre-processing

dHCP. The dHCP rs-fMRI data were pre-processed by dHCP group using the project's inhouse pipeline optimized for neonatal imaging and specifically developed for this dataset, detailed in Fitzgibbon et al.³. This pipeline includes: 1) Motion and distortion correction: corrects for intra-volume movement artefacts and for artefacts associated with susceptibilityinduced off-resonance field changes (susceptibility-by-movement artefacts), and estimates motion nuisance regressors; 2) Registration: aligns all functional images with the native T2 space and the neonatal template space, which refers to the week-specific 40-week template from the dHCP volumetric atlas⁴; 3) Temporal high-pass filter: 150s high-pass cut-off; 4) Denoising: Estimates artefact nuisance regressors and regresses all nuisance regressors from the functional data obtained from the first step.

HCP. The rs-fMRI data of HCP were pre-processed by HCP group using the following pipeline: 1) Distortion correction: correction of gradient-nonlinearity-induced distortion and

phase-encoding-direction-induced distortion; 2) Motion correction: realigns the timeseries to correct for subject motion by using a 6 DOF FLIRT (Oxford Centre for Functional MRI of the Brain Fs Linear Registration Tool) registration of each frame to the single-band reference image; 3) Aligns the original EPI data (rs-fMRI data) to Montreal Neurological Institute (MNI) template space: EPI to T1w from FLIRT BBR, fine tuning of EPI to T1w with BBR-register, nonlinear T1w to MNI template; 4) Intensity normalization to mean of 10000 and bias field removal; 5) Temporal high-pass filter: 150s high-pass cut-off; 6) Denoising: removes artefactual or "bad" components using ICA-FIX to automatically. Detailed pre-processing procedure can be found in Glasser et al.⁵. Additionally, we performed a temporal low-pass filter (0.08 Hz low-pass cut-off) on the denoised rs-fMRI data and removed the first five volumes. Supplementary Fig. 1 provides a schematic of the processing steps for HCP fMRI data. As the selection of the subset of HCP had controlled head motion (i.e., exclusion of participants that had any fMRI run in which more than 50% of TRs had greater than 0.25mm framewise displacement), and adults generally have smaller maximal head motion than neonates,⁶ we did not apply the same scrubbing method used in the dHCP dataset to adult data. To assess the effect of this, we compared the mean framewise displacement (FD) value in adults and neonates before/after the scrubbing procedure.^{7,8} The FD value indexes the movement of the head from one volume to the next and is defined as the sum of the absolute values of the differential realignment estimates (the six realignment parameters). It has been widely used to index head movement and exclude subjects of high motion.⁹⁻¹¹ Independent-samples *t*-tests showed that adults had significantly lower head motion compared to neonates before (*t* (491.45) = -12.49, *p* < 0.001) and even after scrubbing (*t* (580.56) = -9.92, *p* < 0.001; Supplementary Fig. 4).

Supplementary Figure 1. Data processing steps for neonate and adult data. (A) Processing steps for the neonate rs-fMRI data. The rectangles with rounded corners indicate the steps in the fMRI neonatal pre-processing pipeline of the Developing Human Connectome Project. The frames in light purple indicate additional processing steps in this study, and the dotted black line rectangles represent the steps for correcting motion outliers. The frames in dark purple represent the data that we obtained. (B) Processing steps for the adult fMRI data. The rectangles with rounded corners indicate the steps in the fMRI pre-processing pipeline of the Human Connectome Project. The frames in light blue indicate additional processing steps in this study, and the frame in dark blue represents data that we obtained. Abbreviations: dHCP, developing Human Connectome Project; DVARS, D referring to temporal derivative of time courses and VARS referring to root mean squared variance over voxels; ROI, regions of interest; IQR, Inter Quartile Range; FC, functional connectivity; HCP, Human Connectome Project.

Network definition

We first aligned these ROIs with 40-week dHCP T1w template.⁴ This involved: 1) trimming the dura from the 40-week dHCP T1w template, and the cerebellum from both the 40-week dHCP T1w template and MNI T1w template; 2) aligning the 40-week dHCP T1w template to MNI T1w template using non-linear registration (ANTs SyN) (See Supplementary Fig. 2 for the registration accuracy between them); 3) applying the warp file generated in the last step to the ROIs in MNI space with 40-week dHCP T1w template as a reference. In the next step, we needed to align these ROIs in 40-week dHCP T1w template space with neonate native space. We inverted the func-to-template warp provided by dHCP group and applied this inverted warp to ROIs in the 40-week dHCP T1w template space. Thus, we obtained ROIs in each neonate native functional space (Supplementary Fig. 3). For adults, as the denoised HCP data had been aligned to MNI space, we used these ROIs in MNI space directly.

Supplementary Figure 2. Registration of the MNI T1 template to the 40-week dHCP T1w template. The 40-week dHCP T1w templates shown in red lines were overlaid on the MNI T1 template registered to the 40-week dHCP T1w template spaces.

Supplementary Figure 3. The three networks in the MNI space and neonate native space. The functional images represented here were from one neonate (sub-CC00518XX15), which was randomly chosen. (A) default mode network, (B) dorsal attention network, and (C) executive control network. Abbreviations: IPS: Intraparietal sulcus, MT: Middle temporal area, PFC: Prefrontal cortex; R, right; L, left.

Data analyses

Hierarchical clustering analyses. We captured the structure of the three networks in different groups with hierarchical clustering analysis, $12,13$ which has proven informative in prior studies.^{14,15} This hierarchical clustering algorithm builds up an entire cluster tree in which neighbouring regions are joined if their similarity is maximal among all pairs of neighbouring regions. Here, we used the time-course extracted from the 19 ROIs as input to access the hierarchical relationship among the ROIs. For the neonate data, we first calculated initial pairwise distance between ROIs using one minus the linear correlation between the scrubbed time-courses extracted from the 19 ROIs at the individual level. For adults, the initial pairwise distance between ROIs was calculated using one minus the linear correlation between the timecourses of 1195 time points extracted from the 19 ROIs at the individual level. Then, we averaged the pairwise distances between ROIs within each group to get the group-level pairwise distances, which were submitted to hierarchical clustering analysis to create a hierarchical cluster tree of the 19 ROIs for each group respectively. The cophenetic correlation coefficient was used to create a dendrogram for each group. The length of each C link in the dendrogram represents the distance between regions/clusters.

Multidimensional scaling analysis. Non-metric multidimensional scaling (MDS) was also used to facilitate visualizing the similarity of ROIs functional response for adults and neonate groups. The non-metric MDS performs non-metric multidimensional scaling on the dissimilarity matrix of item−item (i.e., ROI−ROI dissimilarity matrix) to compute a configuration.¹⁶ Then, the Euclidean distances between items (i.e., ROIs) in the configuration were obtained. The difference between the monotonic transformed dissimilarities in the item−item (i.e., ROI−ROI) matrix and the Euclidean distances between items (i.e., ROIs) in this configuration were minimized and items (ROIs) were represented in a low-dimensional space (i.e., a 2-D space). The ROI−ROI dissimilarity matrix (one minus the linear correlation between the time-courses) for each group from the hierarchical clustering analysis was submitted to non-metric MDS analysis implemented in MATLAB.

Supplementary Results

Comparison of head motion in neonates and adults. Independent-sample *t*-tests were applied to compare the head motion in the adults and neonates before/after the scrubbing procedure. We found that adults had significantly lower head motion compared to neonates before (*t* $(491.45) = -12.49$, $p < 0.001$) and after (*t* (580.56) = -9.92, $p < 0.001$) scrubbing procedure (Supplementary Fig. 4). A paired-*t* test was applied to detect the difference in head motion in neonates before and after the scrubbing procedure. Results showed that neonates had significantly lower head motion after scrubbing relative to before scrubbing (*t* (427) = -9.69, *p* < 0.001) (Supplementary Fig. 4). Bonferroni correction for multiple comparisons was applied to statistical results.

Supplementary Figure 4. Head motion in neonates and adults. The red dot indicates the mean framewise displacement value in each group. Independent-samples *t*-tests were applied to detect the difference between the adults and neonates. A paired-*t* test was applied to detect the difference between the neonates before and after scrubbing procedure. Abbreviations: FD, framewise displacement; $** = p$ < 0.005 .

Comparison of head motion in neonate groups after scrubbing and adults. We conducted a one-way ANOVA to compare the difference in head motion between the adults and neonate groups after the scrubbing procedure and found a significant main effect of group (*F* (3, 600) = 16.46, *p* < 0.001). Independent-sample *t*-tests were applied to compare head motion between every two groups. We found that that adults had significantly lower head motion relative to the full-term neonates (t (368.44) = -8.67, p < 0.001), preterm neonates scanned at TEA (t (79.69) $= -4.14$, $p < 0.001$), and preterm neonates scanned before TEA (*t* (76.97) = -4.16, $p < 0.001$) (Supplementary Fig. 5). Bonferroni correction for multiple comparisons was applied to statistical results.

Supplementary Figure 5. Head motion in the neonate groups after censoring and the adults. The red dot indicates the mean framewise displacement value in each group. Independent-sample *t*-tests were applied to detect the difference between every two groups. Abbreviations: FD, framewise displacement; TEA, term-equivalent age; $** = p < 0.005$.

High-order networks functional connectivity in adults. In adults, a 2×3 repeated measure ANOVA [*type of FC* (within-network, between-network) \times *network* (DMN, DAN, ECN)] showed a significant main effect of type of FC ($F (1, 175) = 2323.00, p < 0.001$), which was driven by higher overall connectivity for the within- relative to between-network $(t (175) =$ 48.11, *p* < 0.001). We also found a main effect of network (*F* (1.93, 337.773) = 37.50, *p* < 0.001), that was driven by lower overall connectivity for the DMN relative to the DAN (*t*(175) $= -8.72, p < 0.001$) and ECN (*t* (175) = $-3.71, p < 0.001$) and lower overall connectivity for the ECN relative to DAN $(t (175) = -5.10, p < 0.001)$. Finally, a significant interaction effect of type of FC by network $(F (2, 350) = 96.49, p < 0.001)$ was driven by a smaller within- vs between-network FC difference in ECN relative to DMN (t (175) = -11.20, $p < 0.001$) and DAN $(t$ (175) = -13.03, $p < 0.001$). Paired-t tests showed significantly higher within- to between-network FC for each network (DMN: $t(175) = 30.69$, $p < 0.001$; DAN: $t(175) =$ 42.00, $p < 0.001$; ECN: $t(175) = 27.40$, $p < 0.001$) (Supplementary Fig. 6), confirming that each of the three networks was differentiated as a cohesive unit in adults. Bonferroni correction for multiple comparison was applied to statistical results.

Supplementary Figure 6. Network functional connectivity in adults. The between-network connectivity depicts the average FC of each network with the other two. The red dot indicates the mean within/between-network functional connectivity of each network. Paired-*t* tests were applied to detect the difference between within-network and between-network FC for each network. Abbreviations: DMN, default mode network; DAN, dorsal attention network; ECN, executive control network; FC, functional connectivity; $** = p < 0.005$.

The reciprocal relationship between the DMN and prefrontal networks in adults. A one-way ANOVA with repeated measures for between-network FC (DMN−DAN, DMN−ECN, DAN–ECN) showed a significant main effect $(F (1.93, 337.22) = 118.92, p < 0.001)$, which was driven by significantly lower FC in the DMN−DAN relative to the DMN−ECN (*t* (175) = -6.19, *p* < 0.001) and DAN−ECN pairings (*t* (175) = -14.01, *p* < 0.001), and significantly lower FC in the DMN–ECN relative to DAN–ECN pairing $(t (175) = -9.61, p < 0.001)$ (Supplementary Fig. 7). Bonferroni correction for multiple comparisons was applied to statistical results. The lower FC between the DMN and DAN relative to the other pairings suggested a reciprocal relationship between the two networks in adults.

Supplementary Figure 7. Between-network functional connectivity (FC) in adults. FC values were Fisher-*z* transformed. The red dot indicates the mean functional connectivity in each network-pairing. Paired-*t* tests were applied to detect the difference between every two between-network FCs. Abbreviations: DMN−DAN, FC between the default mode network and dorsal attention network; DMN−ECN, FC between the default mode network and executive control network; DAN−ECN, FC between the dorsal attention network and executive control network; $** = p < 0.005$.

Comparison of DMN-frontoparietal functional connectivity in neonates and adults. To investigate between-network connectivity in neonates relative to adults, we created a GLM that controlled for head motion, and compared each neonate group to the adult group. For full-term neonates, we found significant main effects of group for all of the pairings (DMN−DAN: *F* (1, 455) = 278.77, *p* < 0.001; DMN−ECN: *F* (1, 455) = 195.83, *p* < 0.001; DAN−ECN: *F* (1, 455) $= 15.59$, $p < 0.001$), which was driven by significantly lower FC in DMN-DAN and DMN−ECN, and higher FC in DAN−ECN in the adults related to full-term neonates (Fig. 7A). These results suggested that the DMN was more functionally differentiated from the DAN and ECN, and thus, suggesting a stronger reciprocal relationship in adults relative to full-term neonates. For preterm neonates scanned at TEA, we found a significant main effect of group for DMN−DAN (*F* (1, 246) = 105.83, *p* < 0.001), DMN−ECN (*F* (1, 246) = 91.39, *p* < 0.001), which was driven by significantly lower FC in DMN-DAN and DMN-ECN in the adults related to preterm neonates scanned at TEA (Fig. 7B). Similarly, to full-term neonates, these results demonstrated that the DMN was more functionally differentiated from DAN and ECN, suggesting a stronger reciprocal relationship in adults relative to preterm neonates scanned at TEA. We also compared between-network connectivity in preterm neonates scanned before TEA and adults using a GLM that controlled for head motion, although we did not observe a reciprocal relationship between DMN and frontoparietal network in that neonate group. We found a significant main effect of group for DMN–DAN ($F(1, 246) = 257.78$, $p < 0.001$) and DMN–ECN $(F (1, 246) = 45.54, p < 0.001)$, which was driven by significantly lower FC in DMN−DAN and DMN−ECN in the adults relative to the preterm neonates scanned before TEA (Fig. 7C).

Supplementary Tables

Supplementary Table 1. Previous rs-MRI studies of brain network development in neonates

Abbreviations: GA, gestational age; PMA, postmenstrual age; ICA, Independent Component Analysis; SCA, Seed Correlation Analysis; FD, framewise displacement; TEA, term-equivalent age; SMN, sensorimotor network; VIS, visual network; DMN, default mode network; AUD, auditory network; FPN, frontoparietal network; ECN, executive network; SN, salience Network; ACC, anterior cingulate cortex; PCC, posterior cingulate cortex; MPFC, medial prefrontal cortex; LPFC, lateral prefrontal cortex. The studies in bold are the most relevant to the current research, based on direct focus on the presence of the DMN/DAN/ECN and similar infant age range.

Studies	DMN	DAN	ECN
Fransson et al., ¹⁷	\times	\times	\times
Fransson et al., ¹⁹	\times	\times	\times
Lin et al. 20			
Doria et al., ²¹	$\sqrt{}$	$\sqrt{\text{FPN}}$	$\sqrt{}$
Smyser et al., ²²	$\sqrt{\text{(precursor)}}$	\times	\times
Alcauter et al., ²³	$\sqrt{\text{(incomplete)}}$	$\sqrt{\text{(incomplete FPN)}}$	\times
Gao et al., 24	$\sqrt{\text{(incomplete)}}$		
Gao et al., 25	$\sqrt{\text{(incomplete)}}$	$\sqrt{\text{(incomplete)}}$	\times
Gao et al., 26	$\sqrt{\text{(incomplete)}}$	$\sqrt{\text{(incomplete FPN)}}$	\times
Gao et al., 27	$\sqrt{\text{(incomplete)}}$	$\sqrt{\text{(incomplete FPN)}}$	\times
Wylie et al., 28	V	\times	$\sqrt{}$
Damaraju et al ²⁹	$\sqrt{}$	\times	\times
He et al., 31	\times	$\sqrt{\text{FPN}}$	$\sqrt{}$
He et al., 33	$\sqrt{}$	$\sqrt{\text{FPN}}$	$\sqrt{}$
Ball et al., 34	\times	\times	\times
Weinstein et al., 36			
Cui et al., 37	\times	\times	\times
Link et al., 38	$\sqrt{}$	\times	$\sqrt{}$
Rajasilta et al., 40	$\sqrt{}$	\times	\times
Eyre et al., 41	\times	$\sqrt{\text{(FPN)}}$	\times

Supplementary Table 2 Previous findings on the presence of the three high-order networks in infancy.

Abbreviations: DMN, default mode network; DAN, dorsal attention network; ECN, executive network; FPN, frontoparietal network. " $\sqrt{$ " indicates that network was identified while " \times " indicates not; "-" means that network was not investigated in that study. The studies in bold are the most relevant to the current research, based on direct focus on the presence of the DMN/DAN/ECN and similar infant age range.

Group	Sex	Birth age $(GA, weeks \pm$ days)	Scan age (PMA, weeks \pm days)	Birth weight (Kg)
Full-term neonates	160M/122F	40.0 ± 8.6	41.2 ± 12.0	3.35 ± 0.54
Preterm neonates scanned at TEA	41M/32F	32.0 ± 25.6	40.9 ± 14.5	1.76 ± 0.79
Preterm neonates scanned before TEA	50M/23F	32.5 ± 13.4	34.6 ± 13.4	1.78 ± 0.61

Supplementary Table 3. Information obtained from the developing Human Connectome Project for scans used in the present study.

Abbreviations: GA, gestational age; PMA, postmenstrual age; TEA, term-equivalent age; M, male; F, female.

Abbreviations: IPS, Intraparietal sulcus; MT, Middle temporal area; PFC: Prefrontal cortex.

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