Brainwide functional networks associated with anatomically- and functionally-defined hippocampal subfields using ultrahigh-resolution fMRI

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The hippocampus is critical for learning and memory and may be separated into anatomically-defined hippocampal subfields (aHPSFs). Hippocampal functional networks, particularly during resting state, are generally analyzed using aHPSFs as seed regions, with the underlying assumption that the function within a subfield is homogeneous, yet heterogeneous between subfields. However, several prior studies have observed similar resting-state functional connectivity (FC) profiles between aHPSFs. Alternatively, data-driven approaches investigate hippocampal functional organization without a priori assumptions. However, insufficient spatial resolution may result in a number of caveats concerning the reliability of the results. Hence, we developed a functional Magnetic Resonance Imaging (fMRI) sequence on a 7 T MR scanner achieving 0.94 mm isotropic resolution with a TR of 2 s and brain-wide coverage to (1) investigate the functional organization within hippocampus at rest, and (2) compare the brain-wide FC associated with fine-grained aHPSFs and functionally-defined hippocampal subfields (fHPSFs). This study showed that fHPSFs were arranged along the longitudinal axis that were not comparable to the lamellar structures of aHPSFs. For brain-wide FC, the fHPSFs rather than aHPSFs revealed that a number of fHPSFs connected specifically with some of the functional networks. Different functional networks also showed preferential connections with different portions of hippocampal subfields.

The hippocampus plays a major role in learning and memory1–3. Cytoarchitectural, connectomic, and neurophysiological studies have revealed that the hippocampus is a heterogeneous neural structure with complex efferent and afferent connections with multiple brain cortices. Therefore, it is not surprising that extensive efforts have been devoted to defining hippocampal subfields as well as constructing brain-wide hippocampal-subfield specific functional networks. In particular, two different approaches have been widely used to determine hippocampal subfields, namely anatomically and functionally defined subfields. Anatomically, the hippocampal formation consists of multiple, lamellar subdivisions organized along the longitudinal axis of hippocampus, including cornu ammonis 1 (CA1), CA3, CA4, dentate gyrus (DG) and subiculum4–6. Hippocampal subfields have been reported to be differentially vulnerable to a variety of disorders, including autism, Alzheimer’s disease, schizophrenia, and bipolar disorder7–11, as well as prodromal disease states and in individuals clinically normal, yet at-risk for disease12–19.

In contrast, recent investigations, including genetic, as well as functional studies using electrophysiological or fMRI techniques, have revealed that the hippocampal functions are regionally specialized. Using data-driven approaches examining the blood-oxygen-level-dependent (BOLD) signal characteristics of the entire hippocampus, e.g., independent component analysis (ICA), prior studies have reported that functionally independent hippocampal components were spatially arranged along the longitudinal axis16, as well as medial–lateral axis17.
Similar findings were also reported by Zhong et al. who utilized functional correlations between hippocampal voxels and whole-brain cortical regions. Together these studies appear to suggest that the general feature of hippocampal subfields is independent of the approaches (anatomy or functions) used to defined them. That is, the hippocampal subfields are organized along the longitudinal axis as well as potentially mediolateral axis. Despite this consistent general feature, discrepancies exist using the two approaches to define hippocampal subfields. First, there are distinct anatomical demarcations between anatomically and functionally defined subfields. The anatomically-defined hippocampal subfields have a lamellar structure and the functionally-defined hippocampal subfields exhibit a barrel-shaped structure. Second, the brain-wide networks defined using the two approaches differ. Specifically, one of the main assumptions of using anatomically defined subfields as seeds to construct brain-wide functional networks is that within-subfield functional characteristic is homogenous, while across-subfield functional characteristic is heterogeneous. Such an assumption, however, has been challenged by several studies, reporting that resting-state activities between hippocampal subfields are highly correlated and that the functional networks constructed using anatomically defined subfields highly resembles that obtained using the entire hippocampus as a seed region. In contrast, using functionally defined subfields, the brain-wide networks associated with anterior and posterior hippocampus exhibit distinct patterns. Furthermore, Zhong et al. operatively subdivided the hippocampus into head, body and tail portions and showed that the head portion of hippocampus demonstrated a higher connectivity strength with the sensorimotor network than that of body and tail portions.

A caveat to the above findings, however, is that most of the prior hippocampal-subfield studies employed relatively low spatial resolutions (≥ 2 mm isotropic) and/or an insufficient anatomical coverage. Insufficient spatial resolution may not only lead to partial volume effects at subfield boundaries, but also reduce the effective size of unmixed voxels. As a result, the mixture of subfield signals could result in false positive FC as well as affect the functional parcellation of hippocampus. To mitigate the shortcomings attributed by the limited spatial resolution, several approaches capable of acquiring ultra-high spatial resolution fMRI have been proposed. However, these approaches compromise either the ability to acquire whole brain images and/or to achieve a high temporal resolution (< 3 s). A limited spatial coverage makes it impractical to assess brain-wide hippocampal subfield functional networks whereas low temporal resolution renders the images vulnerable to signal contamination from other physiological signals (cardiac and/or respiratory signals).

To this end, we developed a fMRI sequence achieving 0.94 mm isotropic resolution, a repetition time (TR) of 2 s, and whole-cerebrum coverage on 7 T, mitigating the aforementioned limitations (See Supplementary information for the details of the imaging methods). Using the newly developed approach, this study aimed to 1) address whether hippocampal functional subfields are similar to anatomically-defined lamellar structures and 2) define and compare the brain-wide FC associated with fine-grained aHPSFs and fHPSFs using ultra-high-resolution fMRI images. To address the first aim, segmentation was performed using ICA and k-means clustering based on the BOLD time-series of hippocampal voxels. For the second aim, the iFC maps associated with aHPSFs and fHPSFs as well as the spatial distribution of iFC maps in relation to the resting-state networks (RSNs) were demonstrated. A quantitative measure was developed to quantify the functional-network specificity for the aHPSF or fHPSF. To understand how different hippocampal segmentations affect the detection and strength of cortical-hippocampal subfield FC, the hippocampal coverages of significant iFC by every cortical vertex were demonstrated. The results associated with aHPSF and fHPSF were compared.

Materials and methods

Participants. Sixteen healthy participants (aged 27.9 ± 7.2 years, 8 females) were recruited for this study. Participants were screened to ensure they have no history of neurological or psychiatric conditions, previous head trauma, or MRI contraindications. Informed consent was obtained from each participant prior to the study in accordance with the experimental protocol approved by the UNC Institutional Review Board.

MR acquisitions. MR images were acquired using a 7 T Magnetom scanner (Siemens Healthcare, Erlangen, Germany). MPRAGE images were acquired first using the following imaging parameters: TR/TE/T1 = 2200/2.79/1050 ms, flip angle = 7°, partition thickness = 0.94 mm, image matrix = 256 × 208, 192 partitions, and FOV = 24.0 cm × 19.5 cm. Subsequently, two 6-min resting-state scans were acquired using the blipped Partition-encoded Simultaneous Multi-Slab (bPRISM) sequence. The details of the bPRISM sequence can be found in the Supplementary Information. The participants were instructed to remain awake with eyes open during the resting-state scans. The spatial resolution was 0.94 mm isotropic using the following imaging parameters: FOV = 169 × 112.7 × 164.5 (r-l×H-F×A-P) mm³; SMS factor = 7; TE = 23 ms; TR per slab volume = 400 ms; effective TR = 2 s; and number of repetitions 180.

Data preprocessing. Time-series data were motion corrected using FSL. The motion-corrected data were decomposed into a number of independent components by MELODIC. With the 32 resting-state runs acquired from 16 participants, we trained the ICA-based denoising tool FIX to auto-classify ICA components into signal and noise components. The noise components were then removed by FIX using a threshold of 20. The time-series signals were band-pass filtered from 0.01 to 0.1 Hz.

Spatial co-registrations. To minimize crosstalk between hippocampal voxels, the analysis of functional connectivity (FC) was performed on the individual subject’s native EPI space in order to avoid spatial blurring induced by either distortion correction or co-registration. The individual T1 images were coregistered onto the EPI using ANTS (Advanced Normalization Tools 2.1.0) as shown in Fig. 1. To improve coregistration quality, exterior CSF signals of the EPI images were automatically labeled using FSL FAST and removed. The T1 images...
were also masked by the union of gray matter and white matter in order to remove interference from the brain meninges. Nonlinear coregistration with the similarity metric of probability mapping was used.

**Anatomical segmentation of hippocampus.** The labels of aHPSFs were obtained using Freesurfer 6.0\textsuperscript{39–41} on T1 images (Fig. 1). This study employed the version released on August 31st, 2017 which is able to subdivide the hippocampal substructures into head, body and tail where applicable. Specifically, the presubiculum, subiculum, CA1, CA3, CA4/DG are subdivided into head and body portions. Together with parasubiculum and hippocampal tail, we identified 24 aHPSFs (12 per hemisphere) for the following analyses. After coregistering the individual T1 images onto the EPI images, seed-based functional connectivity of each subfield was calculated on EPI space before being warped back to the T1 images.

**Functional segmentation of hippocampus.** Spatially-restricted ICA and k-means clustering were used to perform functional segmentation of the hippocampus (see Figure S6). The EPI time-series signals of each individual were coregistered onto the template brain by resampling the image-series once using the concatenated transformations, which included the transformations from individual EPI, T1 to template spaces. The template surface and brain are of ‘fsaverage5’ from FreeSurfer. The transformation from individual EPI to T1 images was the inverse of the transformation from individual T1 images to EPI as shown in Fig. 1. The transformations from individual T1 to template brains were estimated using the masked hippocampus instead of the whole brain. This step is needed owing to the fact that the cortical folding pattern of each brain varies considerably, leading to inaccurate hippocampal coregistration if the whole brain transformation is used between individual brain images and the template.

Subsequently, the hippocampal temporal signals of all subjects were warped onto the hippocampal template. The group ICA was performed using MELODIC\textsuperscript{35}. The number of independent components for MELODIC operation was set as 24, which is the total number of aHPSFs. The outputs of MELODIC included the probability maps of the independent components. The k-means clustering method utilized the probability maps as the feature vector and generated the segmentation maps. Although the cluster number was initially set as 24, the final cluster number was determined automatically according to the elbow criterion of the cluster validity index, which is defined as the ratio of within-cluster distances to between-cluster distances\textsuperscript{42}. The time courses of the resulting fHPSFs across all the subjects were inspected in order to ensure that none of the fHPSFs exhibited motion-like patterns or high-frequency fluctuation. The FC maps associated with all the fHPSFs were also inspected to ensure none of them exhibit strong correlations on non-brain structures such as cerebral spinal fluid (CSF) or large vessels.

**Figure 1.** The procedures of coregistration. Individual T1-weighted images were coregistered to the EPI by ANTs. The warping coefficients were then applied to the labeled images of the hippocampal subfields. The seed-based FC analysis was performed on the EPI space and the FC map was re-sampled onto individual cortical surface before being warped to the template surface. The re-sampling of FC values on EPI space captured only the signal from the white matter surface to the middle depth of cerebral cortex in order to suppress contamination from large vessels.
**Statistical mapping of functional connectivity on the cortical surface.** The time course associated with each seed was obtained by averaging all time courses within each of the aHPSF or fHPSF. To suppress the signal contamination from pial vessels, only the values from the middle depth of the cortex to the gray-white matter boundary were sampled and averaged. Surface-based spatial smoothing was then applied with full-width half-magnitude (FWHM) of 3 mm. This method has been shown to improve inter-subject consistency on functional mapping in our previous work. After the intrinsic-FC (iFC) values were warped from individual EPI space onto the cortical surface of the template brain, statistical significance of the group-averaged Fisher’s z values was calculated by a t-test using the FreeSurfer command ‘mri_glmfit’. In order to minimize the effect of spatially-varied signal loss and gender effects, the averaged tSNR of the seed region and gender were incorporated into the group analysis as covariates. Correcting for multiple comparisons was performed by the Monte Carlo Simulation in FreeSurfer. The number of permutations was set to 4000. The vertex-wise and cluster-wise p-value thresholds were 0.05. The corrected p-values were converted back to z scores for better visualization.

**Measure of functional-network specificity.** To better understand how the spatial distribution of iFC maps is in relation to commonly defined brain functional networks, seven RSNs using the functional atlas reported by Yeo et al. were employed. In this study, we developed a measure called non-uniformity to evaluate whether a hippocampal subregion (aHPSF or fHPSF) was connected mainly with some specific functional networks or uniformly with many functional networks. Here we defined \( c_i \) as the percentage of the \( i \)th functional network that was significantly connected with a hippocampal subregion and \( c \) is the array of the coverage percentages. To test whether a hippocampal subregion connected to the functional networks uniformly or selectively, we defined the non-uniformity \( \nu \) as

\[
\nu = \frac{\|c'\|_2^2 \sqrt{n} - 1}{\sqrt{n} - 1},
\]

where \( c' = c / \sum c \) and \( n \) denotes the total number of functional networks. If a hippocampal subregion connected to all the functional networks with the same coverage, the \( \|c'\|_2 \) becomes 1/\( \sqrt{n} \) and the non-uniformity \( \nu \) becomes 0. If a hippocampal subregion connected to only one functional network, on the other hand, the \( \|c'\|_2 \) is 1 and the \( \nu \) is 1. Hence, \( \nu \) ranges from 0 to 1 where 0 represents complete uniformity and 1 represents the highest non-uniformity, i.e. highest specificity. High specificity in this study implied that a hippocampal subregion connected with the functional networks selectively.

**Results**

The surface reconstructions of aHPSFs and fHPSFs are shown in Fig. 2a,b respectively. A total of 20 fHPSFs were identified. To demonstrate the spatial relationship between aHPSFs and fHPSFs, the overlapped percentages of each fHPSF with every aHPSF are shown in Fig. 2b. In general, aHPSFs spatially extended along the longitudinal axis of hippocampus, while fHPSFs consisted of barrel-shaped components primarily along the longitudinal axis, consistent with many reports of longitudinal functional specialization. No functionally-defined hippocampal subfields demonstrate lamellar structures similar to aHPSFs. Contrary to the anatomically defined hippocampal subfields which are bilaterally symmetrical, 13 out of 20 fHPSFs were unilateral, suggesting that the iFC of hippocampus is not symmetrical between the two hemispheres. For example, the tail of hippocampus is separated into two functional components, fHPSF #15 and #20, for right and left of hippocampus, respectively. Furthermore, the fHPSF #3 spatially spanned across head, body and tail of hippocampus and overlapped with 56% and 41% of the head and body portion of left CA1, respectively.

The iFC maps associated with aHPSFs and fHPSFs are shown in Fig. 3a,b respectively. The first two maps from the left were the iFC maps overlapped on the left functional map and the right maps were the median view of cortical surface. The maps in the left and right sides of Fig. 3a are associated with the aHPSFs in the left and right hemispheres respectively. The z scores of FC are color-encoded as indicated by the color bar. The merged iFC maps across aHPSFs and fHPSFs covered 67.3% and 64.3% of cortical surface respectively. Additionally, the overlapping percentages of merged iFC maps with each of the resting-state functional networks are listed in Table 1. The overlapping ratio between the two merged iFC maps was 72.5%, indicating that the aHPSFs and fHPSFs have similar topologies of the merged iFC maps. Hence, the corresponding analyses of hippocampal subfields are minimally biased by the detection sensitivity difference of iFC between aHPSF and fHPSF.

To demonstrate the spatial distribution of iFC maps in relation to the commonly defined brain networks, the percentages of the iFC maps with respect to the resting-state functional networks are shown by radar charts in Fig. 4. The radar charts of aHPSFs and fHPSFs were sorted together based on the functional-network specificity (see Materials and Methods) and divided into three groups with the high 1/3, medium 1/3 and lowest 1/3 of specificity. In the high-specificity group, 10 out of 15 charts are of fHPSF even though the total number of fHPSF is lower than that of aHPSF. These results suggest that the fHPSF provides higher sensitivity than aHPSF to detect the iFC specificity to functional networks. Furthermore, for the fHPSF and aHPSF in the group of high specificity, the functional networks that received the highest coverages of iFC are visual, sensorimotor or default-mode networks although the iFC coverages of aHPSF are lower. The highest iFC coverage of visual network is 67% while those of sensorimotor and default-mode networks are 48% and 27% respectively. The FC coverages of the other functional networks are all below 16%, suggesting that the high functional-network specificity is associated more with visual, sensorimotor and default-mode networks than other networks in resting state.

For better visualization, we collated the information from all the charts in Fig. 4 and reorganized in Fig. 5. Only the RSNs with the 4 highest coverage of iFC are shown. The head portions of CA1 connect extensively with visual and sensorimotor networks (maximum at 42% and 20%, respectively) but the body portions show much
less cortical connections (maximum at 12%). On the contrary, the body portions of CA3 and CA4/DG show more cortical connections with visual and sensorimotor networks (maximum at 42% and 43% respectively) than the head portions (maximum at 13%). Among all the aHPSFs, the body portion of pre-subiculum has most extensive FC with DMN (31%).

For the iFC of fHPSFs, as shown in Figs. 3b and 4b, the fHPSF #3 consists of head, body and tail portions and has the most extensive FC with visual network, sensorimotor network and DMN (67%, 48% and 27% respectively). The FC patterns of fHPSF #3 and the head portion of left CA1 are similar (correlation coefficient = 0.49, \( p < 0.001 \)), which reflect the fact that fHPSF #3 overlaps with 56% of the CA1 head and 41% of the CA1 body. From the RSN's perspective, the visual network shows extensive FC with fHPSF #1 (31%), #5 (27%), #6 (28%), #10 (28%), #16 (26%) and #18 (61%). Likewise, the sensorimotor network has extensive FC with fHPSF #5 (29%), #6 (26%), #13 (30%), #15 (45%) and #18 (39%). These results show that the visual and sensorimotor networks connect with distributed anatomical regions along the hippocampal longitudinal axis but some fHPSFs in the posterior hippocampus are more prominent. DMN shows the FC primarily with fHPSF #3 (27%), #8 (25%), #13 (20%), #16 (22%), #17 (24%) and #18 (24%). Almost all of the corresponding fHPSFs are either close to or in the body portion but not the head portion of hippocampus. Different from the visual network, sensorimotor network and DMN, the FPN has much less FC with hippocampus. The fHPSF #15 is the only fHPSF that connects with 22% of FPN.

The brain-wide FC profiles strongly depend on the selection of hippocampal seeds. The selection of seed regions that are functionally heterogeneous will cause reduced detection sensitivity and strength of FC. To

Figure 2. Anatomically and functionally defined hippocampal subfields and their spatial relationships. (a) Surface reconstruction of aHPSF. (b) Surface reconstruction of fHPSF and the percent of spatial overlap between aHPSF and fHPSF. The overlapped percentages with different aHPSFs are color-encoded as indicated in the figure legend. The indices of fHPSFs were arranged from anterior to posterior. The index #1 was assigned to the most anterior fHPSF and #20 to the most posterior fHPSF.
Figure 3. Whole brain functional connectivity maps using (a) aHPSFs and (b) fHPSFs as seed regions, respectively. The statistical significances of iFC maps were assessed using a vertex-wise non-parametric permutation test with 4000 permutations. The statistical maps of iFC were corrected for multiple comparisons and cluster-wise threshold set to $p < 0.05$. The z scores are color-encoded as indicated by the color bar. H head; B body; Sub subiculum; CA cornu ammonis; DG dentate gyrus.
understand how different hippocampal segmentations affect the spatial coverage of significant cortical-hippocampal subfield FC. Fig. 6a,c demonstrate the hippocampal coverage of significant iFC associated with aHPSF and fHPSF respectively. For each cortical vertex, the hippocampal coverage of iFC is defined by the volumetric ratio of the significantly-connected subfields to the whole hippocampus. The cortical patterns associated with aHPSF and fHPSF appear to be similar but the values of the maps are different. To quantify the similarity, we generated the overlapping map (see Fig. 6e) with the coverage threshold of 0. The green color encoded the overlapped areas which connected with at least one of the aHPSFs and fHPSFs. The blue and red colors encoded the areas which connected only with aHPSF and fHPSF respectively. The corresponding dice coefficient is 0.84. Figure 6f showed that the overlapped area covered a majority of cortical surface, including 88% of visual network, 80% of sensorimotor network, 59% of DMN and 50% of frontoparietal network. On the other hand, the non-overlapped area did not show clear preferential FC with any functional network. Despite the similar pattern, nevertheless, the values of hippocampal coverages associated with aHPSF and fHPSF are apparently different. The radar charts

| Table 1. The overlapping percentages of merged iFC maps with different resting-state functional networks. Vis visual network; SM sensorimotor network; dAtt dorsal attention network; vAtt ventral attention network; Limbic limbic network; FP frontoparietal network; DMN default-mode network. |
|-----------------|-----------------|-----------------|-----------------|-----------------|-----------------|
|                | Vis (%) | SM (%) | dAtt (%) | vAtt (%) | Lim (%) | FP (%) | DMN (%) |
| aHPSF          | 90.7    | 87.1%  | 63.8%    | 60.2%     | 52.8%    | 73.4%  | 70%     |
| fHPSF          | 94.9    | 85.5%  | 56.5%    | 53.8%     | 50.1%    | 57.8%  | 71.2%   |

Table 1. The overlapping percentages of merged iFC maps with different resting-state functional networks. Vis visual network; SM sensorimotor network; dAtt dorsal attention network; vAtt ventral attention network; Limbic limbic network; FP frontoparietal network; DMN default-mode network.

Figure 4. The percentages of iFC coverage of each RSN associated with aHPSFs and fHPSFs. The radar plots in the right-most columns of each panel shows the percentages of each iFC map with respect to the seven intrinsic functional networks. The innermost and outermost isoparametric curves in the radar plots represent 0% and 80% respectively. The increment between adjacent isoparametric curves is 20%. The charts of aHPSFs and fHPSFs are sorted together based on the hippocampal specificity and then classified into three groups. L left; R right; other abbreviations are as described in Fig. 3 and Table 1.
Figure 5. The percentages of iFC coverage of each RSN associated with (a) aHPSFs and (b) fHPSFs. The radar charts from left to right panels correspond to visual network, sensorimotor network, DMN and FPN. The names of aHPSF are only illustrated in the left most panel in (a). The head and body portions of aHPSF are highlighted by curly brackets in the other panels. For the radar charts of fHPSF in (b), the indices of fHPSFs were arranged from anterior to posterior. The index #1 was assigned to the most anterior fHPSF and #20 to the most posterior fHPSF. The innermost and outermost isoparametric curves in the radar plots represent 0% and 80% respectively. The increment between adjacent isoparametric curves is 20%. L left; R right; H head; B body; Sub subiculum; CA cornu ammonis; DG dentate gyrus; FPN frontoparietal network; DMN default-mode network.

Figure 6. Hippocampal coverage of iFC associated with aHPSF and fHPSF. (a) The cortical map of hippocampal coverage associated with aHPSF. The contours of functional networks are outlined. (b) The radar chart of averaged hippocampal coverage within each functional network associated with aHPSF. (c) The cortical map of hippocampal coverage associated with fHPSF. (d) The radar chart of averaged hippocampal coverage within each functional network associated with fHPSF. (e) The overlap between the binarized cortical maps in (a) and in (b). The overlapping area is highlighted in green. The area which connects only with aHPSF is highlighted in blue; fHPSF in red. (f) The percentages of the overlapping, 'aHPSF only' and 'fHPSF only' area within each functional network. L left; R right; Lat lateral view; Med medial view; Vis visual network; SM sensorimotor network; dAtt dorsal attention network; vAtt ventral attention network; Limbic limbic network; FP frontoparietal network; DMN default-mode network.
Functional characteristics within hippocampus. The fine-grained functional segmentation shown in Fig. 2b demonstrated a spatial arrangement along hippocampal longitudinal axis, that largely reflected longitudinal functional specialization \(^{19,20}\) without lamellar structures comparable to aHPSFs. As ultrahigh-resolution images minimize the signal mixture between subfields (as shown in Figure S4), our results mitigate the concern of partial volume effects on functional segmentation, as well as support the statement that anatomically-defined subfields do not represent the functional units of hippocampus at rest.

While this study employed ICA for the functional segmentation of hippocampus, alternative approach that employed dynamic FC was also reported \(^{34}\). Although our ICA approach utilized the hippocampal time courses rather than the dynamic FC profiles of hippocampal voxels, both of the approaches took into account the temporal variations of hippocampal signal. Moreover, the dynamic FC-driven parcellation demonstrated longitudinally-arranged functional regions, which is comparable to our segmentation results. Therefore, using the dynamic FC-driven parcellation may produce slightly different segmentation results but would not create systematic difference.

A recent study divided post-mortem human hippocampus into three portions (head, body and tail) and found that adjacent, but not distant, portions of each subfield were anatomically connected \(^{46}\). This finding dovetails with the result that nearly all the functional segments did not extend longitudinally from hippocampal head to tail. Outside of the hippocampus, the afferent input to the entorhinal cortex have been reported to be organized in an anterior-to-posterior fashion \(^{31,47}\), which is largely preserved in its connections to the hippocampus \(^{47}\). That is, anterior entorhinal cortices connect with anterior hippocampus and posterior entorhinal cortices connect with posterior hippocampus. These findings are consistent with our segmentation results as the longitudinally-arranged input pathways to the hippocampus corresponded with the longitudinally-arranged fHPSFs. The spatial arrangement of fHPSFs also clarified why the adjacent portions within or across aHPSFs were correlated, but distant portions were not \(^{46}\), as most of the fHPSFs formed a cluster in proximity.

Although Dalton et al. \(^{30}\) reported that adjacent, but not distinct, segments within each aHPSF were correlated, CA1 was an exception. Interestingly, a similar phenomenon was observed in the current study. FHP SF #3 comprised ~56% of CA1 head and 41% of CA1 body. As shown in Fig. 2b, the functional segment expanded from the anterior to posterior hippocampus. None of the other fHPSFs exhibited such a longitudinal expansion in shape. This finding dovetails with the result that nearly all the functional segments did not extend longitudinally from hippocampal head to tail. Outside of the hippocampus, the afferent input to the entorhinal cortex have been reported to be organized in an anterior-to-posterior fashion \(^{31,47}\), which is largely preserved in its connections to the hippocampus \(^{47}\). That is, anterior entorhinal cortices connect with anterior hippocampus and posterior entorhinal cortices connect with posterior hippocampus. These findings are consistent with our segmentation results as the longitudinally-arranged input pathways to the hippocampus corresponded with the longitudinally-arranged fHPSFs. The spatial arrangement of fHPSFs also clarified why the adjacent portions within or across aHPSFs were correlated, but distant portions were not \(^{46}\), as most of the fHPSFs formed a cluster in proximity.

The longitudinal fHPSFs parcellation observed in the current study is generally consistent with a recent report of functional segmentation \(^{21}\). However, the functional segmentation results in that study could not provide information about functional lateralization because the segmentations of left and right hippocampus were performed separately \(^{17}\). In contrast, we performed functional segmentation using left and right hippocampi jointly. Our results showed 13 out of 20 fHPSFs were unilateral, suggesting the coexistence of the unilateral and bilateral functional subfields. These results of hippocampal functional lateralization are in concord with a number of prior studies. For example, the dynamic FC states demonstrate asymmetric patterns even though the static iFC networks are generally symmetric. Many of the dynamic hippocampal connectivity states even exhibited strong asymmetry \(^{18}\). Additionally, the hippocampus in different hemispheres may be responsible for different functions. Left hippocampus is associated with more verbal memory processes, while right hippocampus has more spatially-dependent processes \(^{46-50}\).

Additionally, the CA1-like subregion in fHPSF was not observed in the earlier reports \(^{16-18}\). Many factors may contribute to this difference, including the spatial resolution of the fMRI data, the used de-noising approaches, and the employed clustering methods. Since one of our goals is to interrogate the respective brain-wide functional networks, the optimization of the functional segmentation was beyond the scope of this study. However, future investigation of the reliability and stability among functional segmentation approaches is an important area of inquiry.

Functional characteristics related to resting-state functional networks. Of the few studies examining FC in different portions of a hippocampal subfield \(29, 30, 47, 51\), they investigated FCs at different portions of a subfield with the surrounding cortical areas, including posterior parahippocampal cortex (PHC) and perirhinal cortex (PRC). The features of brain-wide coverage and ultrahigh resolution in this study shifted the investigation of cortical-hippocampal subfield interactions from a regional level to a network level.
The functional segmentation of hippocampus provided more insights into the cortical-hippocampal subfield connectivity. The brain-wide iFC profiles across aHPSFs had been reported to be similar. With the fine-grained functional segmentation of hippocampus, a number of iHPSFs showed the specificity of iFC with some of the functional networks, including visual, sensorimotor and default-mode networks. The functional-network specificity was not obvious with aHPSFs.

Our results demonstrate that visual and sensorimotor networks preferentially interact with different portions of CA1, CA3 and CA4/DG, as shown in Fig. 5a. The visual and sensorimotor networks connect with the CA1 head more than CA1 body. On the contrary, the two functional networks show strong connections with CA3 body and CA4/DG body rather than the heads of CA3 and CA4/DG. As shown via histological studies, CA1 receives projections from inferior visual system and from V4 through parahippocampal and entorhinal cortices. Moreover, the ventral CA1 (corresponding to human anterior CA1) has direct projection to the olfactory bulb and several other primary olfactory cortical areas in rats and monkeys, lending support to the result that anterior CA1 connects with primary sensory cortices. In contrast, dorsal CA3 and CA4/DG are involved in the dorsal hippocampal network which is known to mediate cognitive process such as learning, memory, and exploration. Histological studies indicate CA3 and CA4/DG receive projections from superior visual system. Our results of aHPSFs suggest that visual networks, as well as sensorimotor networks, connect with anterior CA1, posterior CA3 and CA4/DG. Consistent with our results of fHPSF, both the anterior and posterior subfields connect with visual and sensorimotor networks, but the posterior subfields are more dominant.

For DMN, the posterior pre-subiculum showed the most extensive FC. As documented in several recent subfield studies, the subiculum shows more FC with DMN than CA1, CA3 and CA4/DG. Considering the close proximity between presubiculum and subiculum, the difference of spatial resolution and coregistration method approach may contribute to the discrepancy. For fHPSF, 5 out of the 6 subfields that show extensive FC with DMN are anterior subfields. This is largely consistent with previous reports showing that posterior hippocampus is involved in the posterior medial (PM) system encompassing the dorsolateral/parietal components of DMN.

The FPN demonstrated smaller spatial extent of significant FC with hippocampal subfields compared to visual network, sensorimotor network and DMN in this study. Similar trend was also reported with respect to subiculum, combined subfield CA1-3 and CA4/DG. Since FPN serves as a flexible hub of cognitive control, weaker FC with hippocampal subfields during resting state is biologically plausible. Moreover, our results demonstrated that the lateral prefrontal and posterior parietal cortices in FPN were anticorrelated with the head portion of CA1 (see Fig. 3a). Such an anticorrelation of FPN with anterior hippocampus was also observed in a recent study. Although the anticorrelation between DMN and FPN had been reported to reflect the competitive balance between internally self-referential processing and externally oriented cognitive processing, the anticorrelation between FPN and hippocampal subfields is less explored. More experiments are needed to clarify its biological role. Our group has ongoing research to investigate the functional role of FPN-hippocampal subfield anticorrelation by different task challenges.

Selection of hippocampal segmentation: Anatomical or functional? The iFC profiles of hippocampus with respect to the RSNs are determined by the selection of hippocampal segmentation. While results of the current study suggest that using iHPSFs as the seeds may have the advantage of higher FC strength (see Fig. 6), functional segmentation of the hippocampus requires a certain level of statistical power and is usually more complicated, especially for clinical applications. Other than the difference of FC strength, however, the cortical regions that are connected with any of aHPSF are highly overlapped with those of iHPSF (see Fig. 6e). In other words, using aHPSF for iFC analysis may compromise the strength of FC, but not the presence of FC. An enhanced understanding of functional segmentation of the hippocampus and the respective cortical networks holds a great promise for clinical applications. The hippocampus is the earliest and most severely affected structure in neurodegenerative disorders such as Alzheimer’s disease (AD) and other disorders associated with the hippocampus.

Limitations. Some limitations should be considered in the current study. First, all of the function images were acquired using a 7 T scanner. Therefore, strong susceptibility artifacts may affect the signal quality around the air-tissue interface. In this study, the structures around the anterior and inferior temporal lobes such as the entorhinal cortex, perirhinal cortex, fusiform gyrus, temporal pole experienced severe signal loss (see Figure S5 in the Supplementary Information for details). The inferior part of the hippocampus was also affected. Although we analyzed the temporal signal-to-noise ratio (tSNR) of each hippocampal subfield and regressed out the tSNR variation for the group analysis, the reduction of iFC sensitivity due to signal loss is inevitable. As such, the results related to those regions may not be as reliable.

Secondly, the anatomical segmentation in this study was performed based on the T1-weighted image with 0.94 mm isotropic resolution. Since FSL stores the discrete segmentation volume at 0.33 mm resolution, it is unclear whether the 0.94 mm resolution is sufficient for accurate hippocampal segmentation. To test this issue, we acquired 16 datasets of T2-weight images with 0.6-mm isotropic resolution and T1-weighted images with 1-mm isotropic resolution. The hippocampal segmentation was performed using T1-weighted and T2-weighted images.
separately. The similarity of the hippocampal segmentations was measured by dice coefficient. As shown in Figure S7 (see Supplementary Information), the two segmentation results were highly similar with an overall dice coefficient of 0.99. This finding suggests that the quality of hippocampal segmentation with 0.94-mm resolution is relatively reliable.

**Technical notes.** Of note, it is not our intention to claim that the proposed functional organization of hippocampus is the absolutely accurate model. Rather, this study demonstrates efforts to probe the functional organization of hippocampus by minimizing the biases from insufficient spatial resolution and a priori subfield delineation. Imaging with lower spatial resolution may not only compromise the boundary details, but also reduce the statistical power for multiple comparisons. For studies that focus on subtle changes of subfield boundaries, such as the aging effect on the functional organization of hippocampus, ultrahigh-resolution functional imaging provides unique advantages.

**Conclusions**

This study developed and utilized fMRI data acquisition with submillimeter isotropic resolution, whole-cerebrum coverage, and TR of 2 s to 1) investigate whether hippocampal functional segmentation with ultrahigh-resolution data demonstrate similar anatomical, lamellar structures in the hippocampus, and 2) define and compare the brain-wide FC associated with fine-grained aHPSFs and fHPSFs. No clear lamellar structures were observed in our results of functional segmentation, suggesting that anatomically-defined hippocampal subfields do not represent the functional units of hippocampus at rest. The spatial arrangement of fHPSFs was primarily along the longitudinal axis. While most of the aHPSFs are quite different from the fHPSFs, CA1 was the unique exception. For brain-wide functional networks, we found that the fHPSFs rather than aHPSFs revealed specific FC with some of the functional networks. The visual and sensorimotor networks preferentially interact with different portions of CA1, CA3 and CA4/DG. Visual, sensorimotor, and default-mode networks showed more prominent FC with the body portion of hippocampus than with the head portion. Moreover, lateral prefrontal and posterior parietal cortices in FPN were shown to be anticorrelated with the head portion of CA1. Taken together, our results of fine-grained functional segmentation and the respective functional networks could provide valuable insights into the applications of neurodegenerative diseases.

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W.-T. C. and W.L. wrote the main manuscript text. W.-T. C. and S. L. acquired the data. W.-T. C., S.A., P.-T. Y., Y.T. and H.Z. analyzed the data. K.G. and W.L. designed the experiments. All authors reviewed the manuscript.

Competing interests
The authors declare no competing interests.
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