

# Resting state functional connectivity and cognitive task-related activation of the human claustrum

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## ABSTRACT

Structural and functional analyses of the human claustrum, a poorly understood telencephalic gray matter structure, are hampered by its sheet-like anatomical arrangement. Here, we first describe a functional magnetic resonance imaging (fMRI) method to reveal claustrum signal with no linear relationship with adjacent regions in human subjects. We applied this approach to resting state functional connectivity (RSFC) analysis of the claustrum at high resolution (1.5 mm isotropic voxels) using a 7T dataset ( $n = 20$ ) and a separate 3T dataset for replication ( $n = 35$ ). We then assessed claustrum activation during performance of a cognitive task, the multi-source interference task, at 3T ( $n = 33$ ). Extensive functional connectivity was observed between claustrum and cortical regions associated with cognitive control, including anterior cingulate, prefrontal and parietal cortices. Cognitive task performance was associated with widespread activation and deactivation that overlapped with the cortical areas showing functional connectivity to the claustrum. Furthermore, during high cognitive conflict conditions of the task, the claustrum was significantly activated at the onset of the task, but not during the remainder of the difficult condition. Both of these findings suggest that the human claustrum can be functionally isolated with fMRI, and that it may play a role in cognitive control, and specifically task switching, independent of sensorimotor processing.

## 1. Introduction

In its mediolateral dimension, the claustrum is thin (submillimeter at certain points), but its rostrocaudal and dorsoventral dimensions are roughly equivalent to that of the striatum. Decades of tract tracing studies in rodents and non-human primates indicates that the claustrum is bidirectionally connected with many cortical areas (Edelstein and Denaro, 2004; Crick and Koch, 2005; Goll et al., 2015; Riche and Lanot, 1978; Mathur et al., 2009; Mathur, 2014; Reser et al., 2017; Sherk, 1986; Park et al., 2012; White et al., 2017; Wang et al., 2017) and is estimated by volume to be among the most highly connected structures in the brain (Torgerson et al., 2015). These observations have fueled several hypotheses that the claustrum: 1) binds multimodal sensory information for the generation of conscious perception (Crick and Koch, 2005; Koubeissi et al., 2014); 2) coordinates somatosensory and motor cortical

information (Smith et al., 2012) and; 3) acts as a cortico-cortical relay center supporting attention (Mathur, 2014; Goll et al., 2015).

While the exact function of the claustrum is still largely theoretical (Van Horn, 2019), recent work suggests a role in cognition. Comprehensive analyses of mice on how the cortical mantle connects with the claustrum demonstrate that the claustrum weakly innervates primary sensorimotor cortices, while heavily innervating frontal cortices including the anterior cingulate cortex (ACC) and the prefrontal area of the medial prefrontal cortex (PFC) (White et al., 2017). A strong input from the ACC to the claustrum also exists in rats (Smith and Alloway, 2010; White et al., 2017) and mice (Qadir et al., 2018), and this input encodes a top-down preparatory signal that is proportional to task difficulty (White et al., 2018). These findings suggest that the claustrum may subserve frontal cortical function, including top-down executive processes (Atlan et al., 2018; Goll et al., 2015; Mathur, 2014; White and

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Mathur, 2018). While some neuroimaging studies have been performed on the claustrum (Krimmel et al., 2019; Smith et al., 2017), evidence for a role of the human claustrum supporting any of the aforementioned functional hypotheses, including cognitive processing, is particularly lacking.

While the anatomical boundaries of the human claustrum (outside of the ventral subdivision) can be resolved with relative ease using high-resolution structural magnetic resonance imaging (MRI), functionally resolving this structure for analysis of blood-oxygenation level-dependent (BOLD) signal with functional MRI (fMRI) is challenging, as signal extracted from the claustrum is heavily mixed with the signal from the neighboring insula and putamen using standard methods. Analysis of BOLD data using standard methods results in similar patterns of functional connectivity (correlation of signal between regions) when comparing claustrum, insula, and putamen. This contrasts with data from multiple tract tracing studies, which instead show unique patterns of anatomical connectivity across these regions (Nakashima et al., 2000; Mathur et al., 2009; Pan et al., 2010; Sato et al., 2013; Wang et al., 2017). Thus, standard fMRI analyses are not capable of functionally resolving the claustrum and may yield inaccurate functional connectivity and activation results.

In an effort to elucidate claustrum function, the current study has three goals: 1) devise new fMRI methodology called Small Region Confound Correction (SRCC) to functionally distinguish the claustrum from the insula and putamen, creating a corrected claustrum timeseries; 2) perform resting state functional connectivity analyses with the corrected claustrum timeseries to reveal functional coupling of the claustrum in humans; and 3) test the hypothesis that the claustrum, owing to strong connectivity with frontal cortices and recent data in mice suggesting its involvement in top-down cognitive processing (White et al., 2018), is activated during a cognitive conflict task. Our data indicate that extraction of unique claustrum signal using fMRI is possible and that the claustrum displays expansive functional connectivity that overlaps with regions involved in cognitive control. During task performance, we found the claustrum to be active at the onset of – or switch to – cognitive conflict task engagement.

## 2. Methods

### 2.1. Overview

Three datasets were analyzed. The first dataset, *7T-Rest*, is publicly available and consisted of scans from 20 healthy humans scanned with 7 Tesla (T) MRI (Gorgolewski et al., 2015). The second dataset, *3T-Rest*, consisted of scans from 35 healthy humans acquired with 3T MRI. The third dataset, *3T-Task*, used the same subjects as *3T-Rest* and consisted of scans from 33 healthy humans performing a cognitive interference task acquired with 3T MRI. In *7T-Rest*, we performed seed-based whole-brain resting state functional connectivity (RSFC) analyses using the claustrum and surrounding structures (insula/putamen) as seeds. To examine claustrum functional connectivity while controlling for the influence of insula and putamen signal, we regressed out the timeseries of insula/putamen sub-regions from the non-smoothed claustrum, creating a corrected claustrum signal that we then used to make a claustrum specific RSFC map. We used *3T-Rest* to replicate findings from *7T-Rest*. We applied similar methods used in resting state data to *3T-Task* in order to isolate BOLD signal from the claustrum, and analyzed this signal to determine claustrum activation during the cognitive task.

### 2.2. Participants and MRI data

*7T-Rest*: Data from 22 subjects were acquired from a publicly available dataset scanned on a 7T MR scanner (MAGNETOM 7T, Siemens Healthcare, Erlangen, Germany). We used data from only the first session of this dataset. These data included a T1-weighted structural 3D MP2RAGE image that was used in preprocessing (TR = 5000 ms,

TE = 2.45 ms, voxels = 0.7 mm isotropic). *7T-Rest* also included two eyes-open resting state scans using echo planar imaging (EPI) to measure BOLD fMRI while fixating on a plus sign (whole brain coverage, TR = 3000 ms, TE = 17 ms, voxels = 1.5 mm isotropic, slices = 70, duration = 300 TR). Details of the scans can be found in (Gorgolewski et al., 2015). Two subjects were excluded, one from errors induced by preprocessing where despite repeated efforts, quality normalization could not be achieved and a second from different scanning parameters than the other participants, leaving 20 subjects for analysis (average age = 25, s.d. = 2, 10 women).

*3T-Rest*: 36 healthy subjects were recruited as a control sample for an ongoing clinical trial and only a baseline scan was used for the following analyses. MRI data were acquired at the University of Maryland, Baltimore Medical Imaging Facility with a Siemens 3T Tim Trio scanner with 32 channel head coil (n = 22) or a Siemens 3T Prisma scanner with a 64 channel head coil (n = 14) due to a scanner upgrade during data acquisition. We acquired a T1-weighted structural 3D MPAGE scan that was used in preprocessing (whole brain coverage, TR = 2300 ms, TE = 2.98 ms, voxels = 1.00 mm isotropic). We also acquired eyes-open resting state scans using EPI while subjects fixated on a plus sign (whole brain coverage, TR = 2000 ms, TE = 28 ms, voxels =  $3.4 \times 3.4 \times 4.0$  mm, slices = 40, duration = 300 TR). One subject was excluded due to poor coverage leaving 35 subjects (average age = 37, s.d. = 13, 31 women).

*3T-Task*: Participants in the *3T-Rest* also performed the multi-source interference task (MSIT; Bush et al., 2003) as a measure of cognitive conflict. Subjects were first trained on the MSIT outside of the scanner. During the MRI session, they performed the MSIT in two runs of about 5 minutes each while echo planar imaging with whole brain coverage was acquired (TR = 2500 ms, TE = 30 ms, voxels = 3.00 mm isotropic, slices = 44, duration = 121 TR). The task was the same as previously reported (Seminowicz and Davis, 2007; Seminowicz et al., 2011), but is briefly described here. On each trial, the volunteer was presented with an array of three numbers. In each array, two numbers were the same and one number was different. The volunteer was instructed to press a button that corresponded to the number that was different from the two other numbers or characters presented on the screen for that given trial. The control condition was a sequential tapping task in which an asterisk appeared in the same order moving from the left to the right of the screen and the subject pressed a button corresponding to the position of the asterisk. There were two levels of task difficulty (easy, difficult), which were performed in separate 20s blocks (10 stimuli per block). In the easy condition, the different number indicated the position of that number in the array (e.g. “1-2-2”, “3-2-3”, “1-1-3”), and in the difficult condition, the different number did not indicate the position of that number in the array (e.g. “3-2-2”, “3-1-3”, “1-1-2”). Three subjects were excluded because of poor coverage and missing data, leaving a sample size of 33 (29 women, average age = 37 s.d. = 12).

### 2.3. Ethics statement

The collection of *3T-Task* and *3T-Rest* was approved by the University of Maryland, Baltimore Institutional Review Board, and informed written consent was obtained from each participant prior to any study procedures. The study for *7T-Rest* was approved by the Ethics Committee of the University of Leipzig and all subjects gave written informed consent prior to the study procedures (Gorgolewski et al., 2015).

### 2.4. Initial data preprocessing

Data were preprocessed in SPM12 (<http://www.fil.ion.ucl.ac.uk/spm/software/spm12/>), including slice timing correction, realignment (motion correction), coregistration of the T1-weighted structural scan to the mean realigned functional image, segmentation of the structural scan, normalization of the structural and realigned functional images to a standard MNI template using a 4th degree B-Spline interpolation, and smoothing with a 6 mm full width at half maximum (FWHM) Gaussian

kernel for 7T-Rest, 3T-Rest, and 3T-Task. We elected to use the same smoothing kernel for 3T-Rest and 7T-Rest to have as comparable processing pipelines as possible. We did not observe obvious qualitative differences for smoothing 7T-Rest at 6 mm FWHM vs a 3 mm FWHM kernel (Supplemental Fig. S1).

## 2.5. Analysis of resting state claustrum connectivity

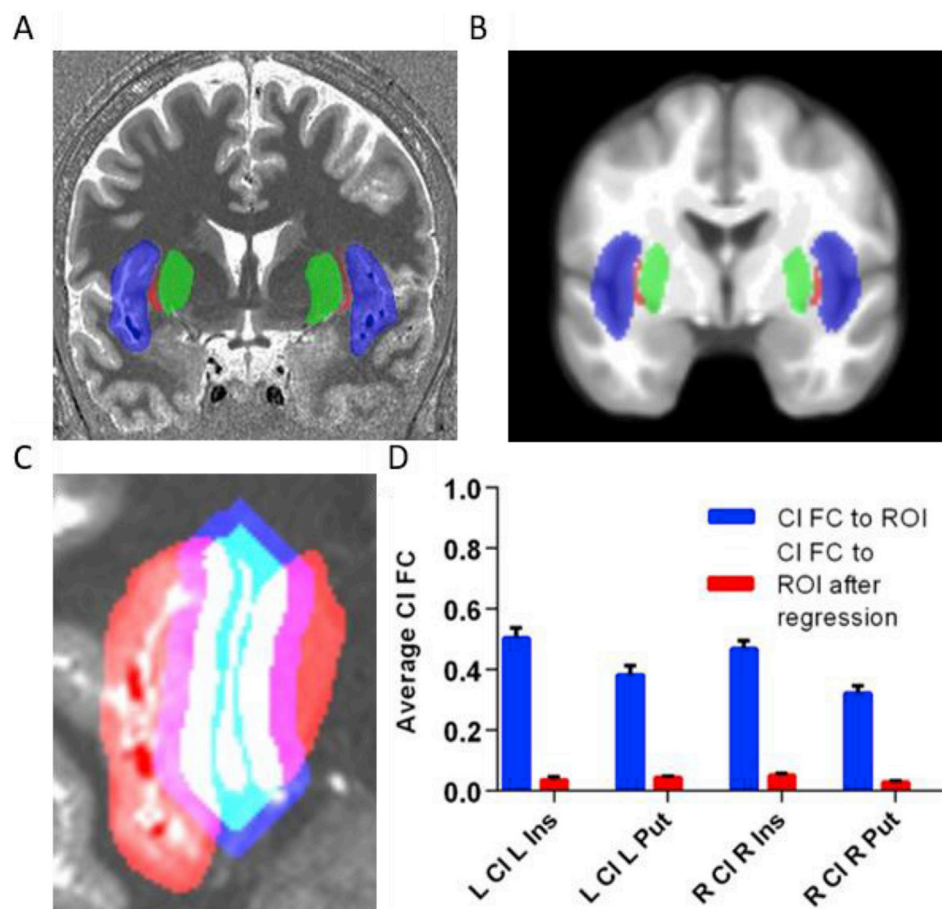
**Resting-state preprocessing.** Whole-brain 7T-Rest and 3T-Rest data underwent the following further preprocessing before functional connectivity analyses were performed. Resting-state preprocessing and seed-based analyses were conducted in the Conn toolbox version 17f (<http://www.nitrc.org/projects/conn>). Given continued controversy, we elected not to control for global signal in our analyses (Murphy and Fox, 2017). To control for noise present in white matter and CSF, we used aCompCor (Behzadi et al., 2007; Muschelli et al., 2014) to determine the first five eigenvectors of white matter and CSF. As we did not want to remove global signal, we used a twice eroded CSF mask and a white matter mask with four erosions (that did not include external or extreme capsules), as this level of erosion has been shown to no longer contain global signal (Power et al., 2017). We removed motion related signals estimated from the realignment parameters along with the first order derivatives of these parameters, in addition to the 10 eigenvectors from white matter and CSF. To avoid the reintroduction of noise while removing low frequency artifact, we used linear detrending and also simultaneously high-pass filtered our voxelwise and regressor data with a cutoff of 0.008 Hz. No low-pass filter was used because signal is present above standard cutoffs (~0.08 Hz, Smith et al., 2013). Despiking was employed after these steps to remove any additional artifact that had not yet been removed (Patel et al., 2014).

## 2.6. Standard analysis

Following artifact removal, ROI analyses were performed by extracting the average timeseries of all voxels within a given set of ROIs using normalized, but not smoothed data. For each subject in 7T-Rest, left and right claustrum ROIs were hand drawn on the subject's normalized structural image, along with ROIs for the insula and putamen. Fig. 1A shows an example tracing for one subject and Fig. 1B shows the average ROIs on a group template. In 3T-Rest, we used mean ROIs obtained from 7T-Rest, after confirming that these mean images fit well onto the normalized structural data for 3T-Rest. Functional connectivity was calculated as the Pearson correlation between the timeseries for each ROI and all voxel timeseries across the brain. In the standard analysis for 7T-Rest, we used the un-corrected claustrum, whole insula, and whole putamen timeseries from individually drawn ROIs to compute RSFC across the whole brain.

## 2.7. Corrected claustrum analysis with Small Region Confound Correction (SRCC)

To remove the possible influence of insula and putamen signals on claustrum timeseries and hence FC, we determined the timeseries of parts of the insula and putamen that 'flanked' the claustrum and treated these as confounding sources for the claustrum. These flanked ROIs were calculated on the individual level for 7T-Rest by dilating the claustrum 4 functional voxels (6 mm), and determining the overlap between the dilated claustrum and the insula/putamen seeds that were at least 2 functional voxels (3 mm) separated from the original claustrum. This resulted in 'flanking' regions that were continuous with the insula and putamen and similar in shape to the claustrum, but still distant enough



**Fig. 1.** The claustrum is a thin sheet of gray matter and its fMRI signal is normally confounded by neighboring structures. A) Structural MRI image with insula (blue), claustrum (red), and putamen (green) for one subject. These regions were hand-drawn for each subject on their anatomical image. B) Average insula (blue), claustrum (red), and putamen (green) averaged over 20 hand-drawn images displayed on an average template. C) To remove signal from neighboring structures, the claustrum was dilated, adding 6 mm around the claustrum, while maintaining its shape (blue). This dilated claustrum was overlapped with the neighboring insula and putamen (red). Voxels that contained insula/putamen and dilated claustrum separated by at least 3 mm from the original claustrum were categorized as flanking regions. Flanking regions, shown in purple, were regressed out of the claustrum signal to create a corrected claustrum timeseries. D) Correlation (FC) of claustrum timeseries with L/R insula/putamen before and after regressing out flanking regions (i.e. following Small Region Confound Correction (SRCC)). The average connectivity between claustrum and its neighbors approaches zero following this approach. Error bars show standard error of the mean. Cl = claustrum, Ins = insula, Put = putamen, ROI = region of interest, FC = functional connectivity.

from the claustrum to ensure that they contained no claustrum data within the seed, as shown in Fig. 1C. We call this approach Small Region Confound Correction (SRCC), as it is designed to eliminate partial volume effects that are particularly problematic in small regions of the brain. In 3T-Rest, this process was repeated, but for the mean ROIs only. We regressed each claustrum timeseries (left and right) on the timeseries from the ipsilateral ‘flanking’ regions, along with previously described artifacts (motion, CSF, etc.), and the residuals from this analysis constituted a corrected claustrum timeseries (e.g. corrected for signal in the flanking insula and putamen sources) that was then used as the seed in whole-brain seed-to-voxel analysis.

To determine statistical significance of seed-to-voxel functional connectivity maps, we performed one sample t-tests in SPM12 for each RSFC map (left and right corrected claustrum). These outputs were used for qualitative comparisons between regions. A cluster-forming threshold of  $p < 0.001$  was used for all analyses and significant clusters based on FWE correction are reported, as these thresholds have been shown to adequately control for false positive rates (Woo et al., 2014).

## 2.8. Modelling task data

We determined the reaction time (RT) for every trial presented in the MSIT. If the participant did not respond to a particular trial, we designated the RT for that trial as the maximum possible trial duration (i.e. 1500 ms). We did not analyze accuracy, as the average was over 90 percent. Typically, the MSIT is modelled in a block design, using three types of blocks (tapping, easy, and difficult). Because of our interest in the effects of cognitive load, we modelled the initial events in a task block as block onset. During this period, reaction times are slower, likely reflecting the cognitive adaptations (or switching) to a new set of rules, and we sought to model this period. We hypothesized that switches would occur relatively early in the block (the first third), and that the switch would be associated with a relatively large RT for the participant. This adaptation period could vary from block to block and between individuals, so, we defined block onset as the first one, two, or three trials of the block, depending on the RT for those trials. We used z scores for each of the first three trials compared to the mean of trials 4 to 10 of the same block type (easy or difficult) to select the duration of the onset adaptively. The block was modelled as the remainder of the block not included in onset. We used this hybrid event-block design for all task analysis of MSIT.

## 2.9. Preprocessing

For whole brain task analyses, we did not treat insula and putamen signals as a source of noise. For analyses of claustrum activation we used similar approaches as described in *Analysis of resting state claustrum connectivity and corrected claustrum analysis with SRCC* (see above) except we did not include despiking nor the first derivative of realignment parameters. This approach created a corrected claustrum timeseries during MSIT that was then used to determine claustrum activation during the task. We then estimated whole brain and corrected claustrum ROI activation patterns.

## 2.10. Analysis

To determine significant task activation for whole brain analyses, we performed one sample t-tests for contrast maps with a cluster-forming threshold of  $p < 0.001$  and a FWE cluster correction. Contrast maps for difficult and easy blocks and onsets were calculated based on a tapping baseline (e.g. [difficult > tapping] and [easy > tapping]). To determine claustrum activation, we performed a one-sample t-test on extracted activation estimates for each contrast extracted using MarsBaR v 0.44 (<http://marsbar.sourceforge.net/>). To examine the overlap between claustrum connectivity and task-related activation during MSIT, we created an overlap map between areas of significant deactivation/

activation and areas with significant claustrum functional connectivity. We then used this overlap map to calculate the percentage of task-responding voxels that functionally connected with the claustrum. Average RT for each task condition (tapping, easy, difficult) was calculated for each subject, and these subject-averaged RTs were compared between each task type using a one-way repeated measures ANOVA with Greenhouse-Geisser correction. Post-hoc t-tests were used to identify specific task type differences with Tukey's correction for multiple comparisons. We also performed a one-way ANOVA with Greenhouse-Geisser correction and used a Dunnett's test to compare the first trial to all other trials. Whole brain contrast of difficult vs easy tasks gave task-positive (difficult > easy, extrinsic mode network: EMN) and task-negative (easy > difficult, default mode network: DMN) networks. For event-related plots in Fig. 4C, we extracted timeseries data for the claustrum and for EMN and DMN. These timeseries were averaged across all voxels within subjects, and then averaged across subjects, yielding a single vector of length equal to the number of TRs for MSIT. This vector was then averaged over each experimental block, yielding a vector for each experimental condition.

## 2.11. Data and code availability

Data and corresponding code are available through request. This complies with requirements of our institution and funding agencies.

## 3. Results

### 3.1. Resting state connectivity of the claustrum at 7T: methodological approach

We used high spatial resolution fMRI data in 7T-Rest and hand drawn ROIs of claustrum, insula, and putamen to analyze whole brain functional connectivity of these ROIs. Despite excellent resolution, we found that functional connectivity between the claustrum and the insula/putamen in this standard analysis was high (average FC of L/R claustrum with insula/putamen ranged from 0.32 to 0.5; Fig. 1D). This contrasts with known unique connectivity of the claustrum relative to insula and putamen (Nakashima et al., 2000; Mathur et al., 2009; Pan et al., 2010; Sato et al., 2013; Wang et al., 2017), and suggests that even with high spatial resolution and well identified ROIs, that insula and putamen signals are sampled in claustrum voxels. This confounds the interpretation of small structure studies in fMRI even when the regions can be identified with relative ease in structural scans.

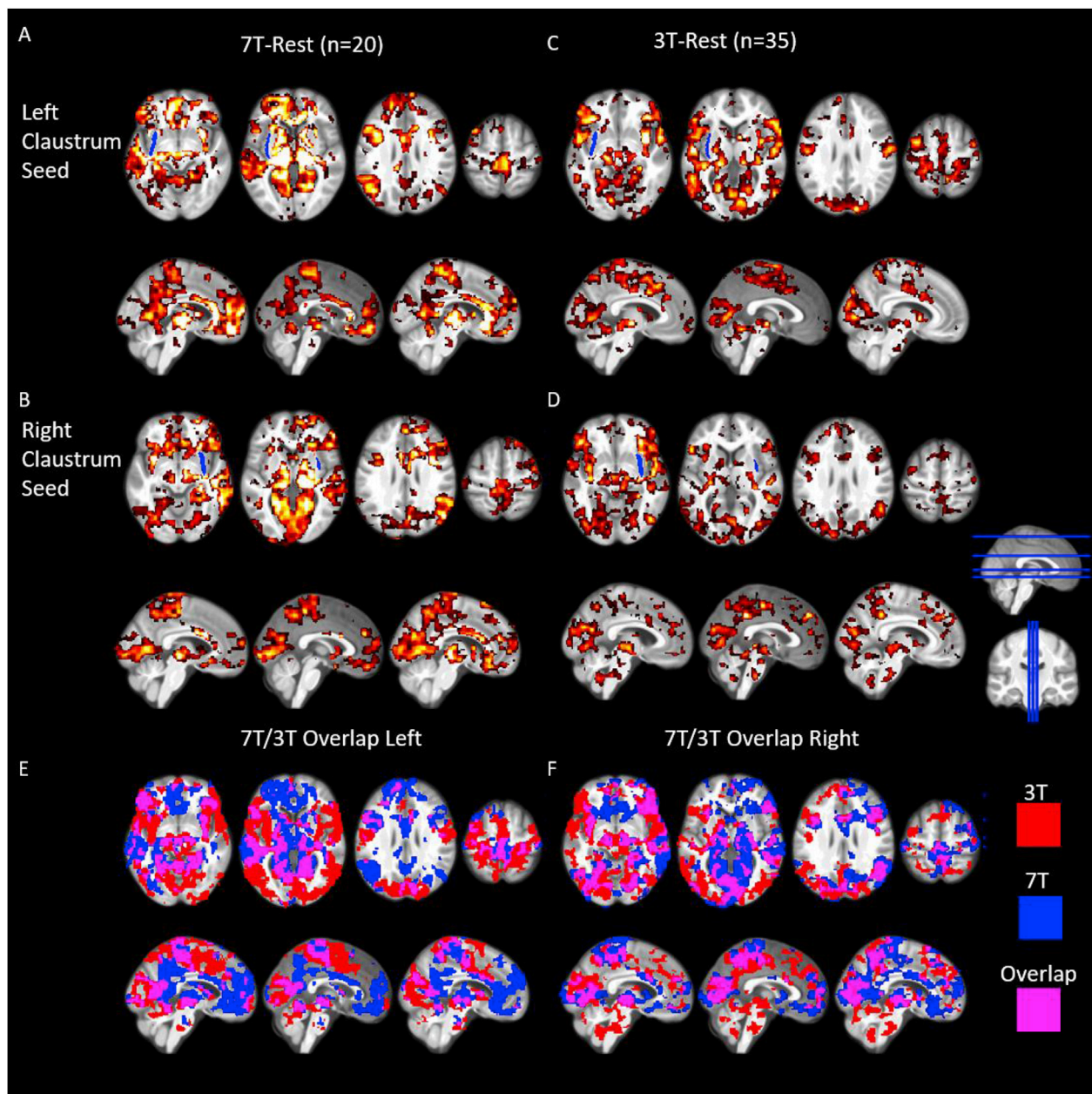
### 3.2. Creating a corrected claustrum signal with SRCC

To mitigate the effect of insula/putamen sampling in claustrum voxels, we generated a corrected claustrum timeseries using SRCC by regressing out signal from insula and putamen regions separated by two voxels from claustrum (3 mm) as sources of noise, identical to how nuisance white matter (WM) and cerebro-spinal fluid (CSF) signals are treated (see Fig. 1C). This approach eliminated the linear relationship between claustrum and the surrounding structures, and the resulting average correlation between the corrected claustrum timeseries and neighboring regions reduced effectively to zero (Fig. 1D).

### 3.3. Resting state connectivity of the corrected claustrum at 7T and 3T

We estimated functional connectivity of the corrected claustrum timeseries across the whole brain in two datasets. RSFC is a measure of inter-region coupling and may reflect the efficiency of information transfer (McIntosh, 2000), and has a complex relationship to structural connectivity (Greicius et al., 2009). RSFC cannot be directly measured, but is typically estimated by correlating BOLD timeseries across the brain. We anticipated human claustrum RSFC to be widespread and to feature functional connectivity with the cingulate cortex, PFC, visual



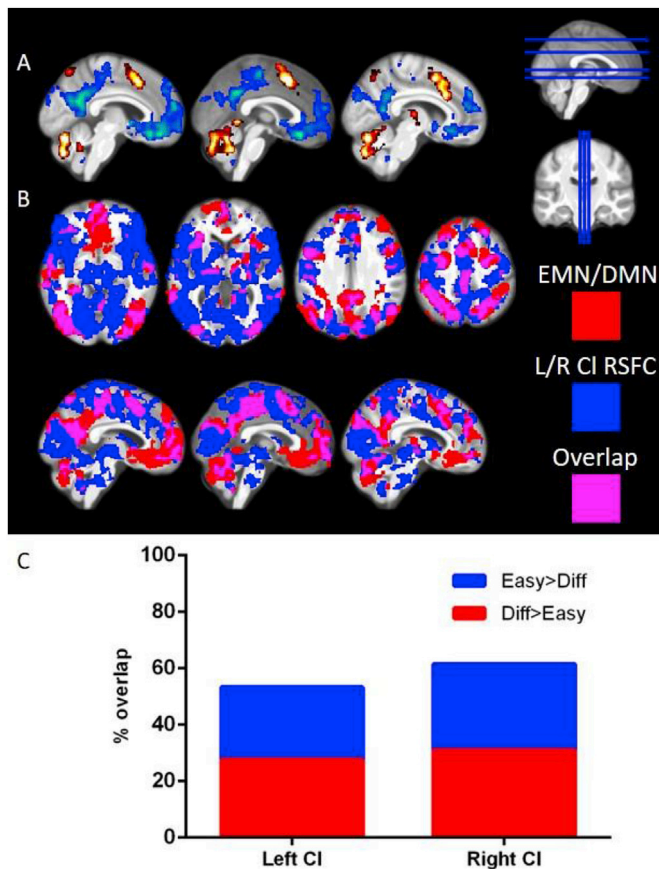


**Fig. 2.** Resting state connectivity of the left and right claustrum at 7T and 3T showing widespread connectivity to cortical and subcortical regions. A) RSFC of the left claustrum in 7T-Rest. B) RSFC of the right claustrum in 7T-Rest. C) RSFC of the left claustrum in 3T-Rest. D) RSFC of the right claustrum in 3T-Rest. E) and F) show the overlap of these thresholded RSFC maps. Data were voxelwise thresholded at  $p < 0.001$  followed by FWE cluster correction. Blue ROI in A–D represents an average claustrum dilated by 1.5 mm (for visualization only). Cl = claustrum.

cortex, and intraparietal sulcus (IPS), given that regions homologous to these in the rodent display claustrum connectivity (White et al., 2017). In 7T-Rest, this analysis identified connectivity with the: thalamus (particularly the pulvinar); nucleus accumbens; parahippocampal gyrus; cingulate cortex (subgenual anterior cingulate cortex (sACC), pregenual anterior cingulate cortex (pACC), anterior mid cingulate cortex (aMCC), posterior mid cingulate cortex (pMCC), and ventral posterior cingulate cortex (vPCC)); superior and inferior temporal gyri; PFC (including the dorsal lateral PFC, medial PFC, and ventral lateral PFC); sensorimotor cortex; precuneus; angular gyrus; IPS; and visual cortex (Fig. 2A and B and Tables S1 and S2).

Overall, we found a more extensive pattern of claustrum functional connectivity in 3T-Rest, however, the regions displaying FC with claustrum were largely similar to 7T-Rest. The more extensive pattern of claustrum functional connectivity observed in 3T-Rest may be a result of

the larger sample size compared to 7T-Rest (35 for 3T vs. 20 for 7T), and may result from different signal to noise ratios between the datasets. In 3T-Rest we identified connectivity with: the thalamus, mainly the pulvinar; nucleus accumbens; parahippocampal gyrus; cingulate cortex (pMCC, and dorsal PCC); PFC; temporal gyri; sensorimotor cortex; precuneus; angular gyrus; IPS; and visual cortex (Fig. 2C and D and Tables S3 and S4). While the patterns of connectivity were largely similar across datasets and 3T-Rest overall had a more extensive RSFC pattern, 7T-Rest displayed greater RSFC in the sACC, pACC, dPCC, superior frontal gyrus, and angular gyrus on the left side than 3T-Rest, and on the right side 7T-Rest displayed more sACC RSFC. The similar pattern of claustrum functional connectivity across 7T-Rest and 3T-Rest, and the bilateral nature of claustrum functional connectivity (i.e. left claustrum functionally connects with right) suggest our confound-corrected claustrum timeseries is not artifactual.



**Fig. 3.** Claustrum functional connectivity substantially overlaps with DMN and EMN. A) Contrasting the difficult block and the easy block of MSIT reveals DMN (easy > difficult, shown in blue) and EMN (difficult > easy, shown in red). B) Areas functionally connected with L/R claustrum in 3T-Rest (blue) and the DMN and EMN shown above (red) show a high degree of overlap (purple) C) Quantification of DMN/EMN voxels from panel A and the percent that also have FC with claustrum. Claustrum FC maps are the same as shown in Fig. 2B. Task contrasts maps used a  $p < 0.001$  threshold with FWE cluster correction. Easy = easy condition of MSIT, Diff = Difficult condition of MSIT, CI = claustrum, DMN = default mode network, EMN = extrinsic mode network, RSFC = resting state functional connectivity, L = left, R = right.

### 3.4. Extrinsic and default mode network overlap with claustrum functional connectivity

Given the high degree of claustrum functional connectivity with regions involved in cognitive control (e.g. ACC) and prior literature showing claustrum involvement in top-down cognitive processing (White et al., 2018), we sought to quantify the overlap between resting-state claustrum connectivity, the cognitive conflict task positive network (i.e. extrinsic mode network; EMN) (Hugdahl et al., 2015) and the task negative network (i.e. default mode network; DMN) (Raichle et al., 2001) evoked by the multi-source interference task (MSIT; Bush et al., 2003). MSIT features a difficult condition where there is conflict between the position and identity of a number that must be selected and an easy condition, where there is no conflict. The neural activation associated with these conditions is contrasted to a motor and visual control condition. Examining the difficult-block greater than easy-block contrast identified the standard EMN, including fronto-parietal network (FPN), dorsal-attention network (DAN), and dorsal ACC (Fox et al., 2005; see Fig. 3A and Table S5). The easy-block greater than difficult-block contrast identified the DMN. We found left and right claustrum functional connectivity overlapped with more than half of the voxels in the combined extrinsic mode and default mode networks (Fig. 3C). Additionally, every major region with significant task activation was also

functionally connected to the claustrum (i.e. at least partially overlapped).

### 3.5. MSIT performance

Consistent with past literature, we found a main effect of reaction time (RT;  $F_{1.76,56.38} = 374.2$ ,  $p < 0.05$ ), with RT for tapping < easy < difficult trials within the MSIT, and all trial types differed from one another in terms of RT (Fig. 4A).

When analyzing RT at the group level, we observed that RT was significantly higher for the first trial of easy and difficult blocks than the remaining 9 trials, a finding observed in many cognitive tasks (Blais et al., 2014) (Fig. 4B) (Difficult:  $F_{6.27,200.7} = 44.17$ ,  $p < 0.05$ ; Easy:  $F_{4.33,138.5} = 25.43$ ,  $p < 0.05$ ). These data suggest that after the first trial participants transition into a stable level of performance and therefore the beginning or onset of difficult and easy conditions should be modelled separately from the block (creating 5 conditions to be modelled: tapping, easy onset, easy block, difficult onset, difficult block).

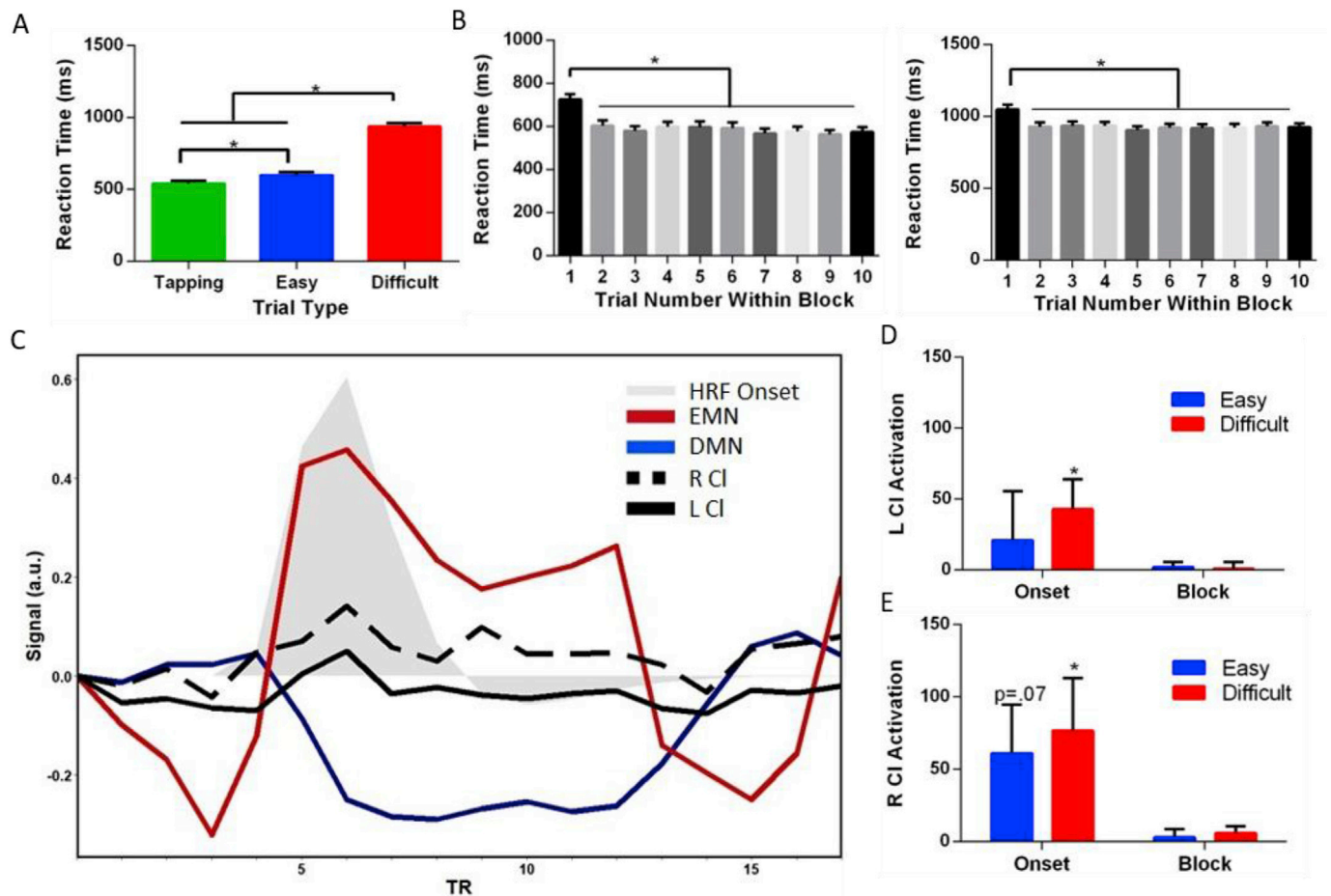
### 3.6. Claustrum responds to the switch to block onset

The onset of tasks involved switching from performing a tapping visuomotor control task to performing a block of either a no-cognitive conflict easy task or a cognitive conflict difficult task. Using a hybrid event-block design for all task analysis of MSIT, we found activations revealing the EMN and deactivations revealing the DMN for the onset and block for both easy and difficult conditions across the brain (Tables S5, S6, S7, and S8). These results show MSIT analyzed under our framework is a robust task to elicit task positive and task negative responses, similar to other analyses of the MSIT. This hybrid model allowed us to examine switches in cognitive operations and we determined the onset of the block in an adaptive manner based on RT where the onset duration was defined as the first 1, 2, or 3 events out of 10 in the block. When analyzing claustrum activation independent of surrounding insula and putamen using this RT adaptive model, we found that claustrum significantly responded to difficult onset (Left:  $t_{32} = 2.1$ ,  $p < 0.05$ ; Right:  $t_{32} = 2.2$ ,  $p < 0.05$ ), though the same effect was not found for easy onset (Left:  $t_{32} = 0.6$ ,  $p > 0.2$ ; Right:  $t_{32} = 1.8$ ,  $p = 0.07$ , Fig. 4C–E). However, we found no evidence that claustrum responded to the difficult block nor to the easy block and found no evidence that claustrum responded to difficult onset more than to easy onset (all  $p > 0.2$ ). This, along with the robust EMN activation during block onset, suggests that claustrum is involved in switching to cognitive control, rather than maintaining performance during high cognitive demand.

## 4. Discussion

In this study, we provide a novel approach we term Small Region Confound Correction to detect the activity and functional connectivity of the human claustrum. In doing so, we find that the claustrum is strongly functionally connected to cingulate and prefrontal cortices at rest and that there is considerable overlap between claustrum connectivity maps and cognitive task-related networks. Supporting a role of the claustrum in cognition and task switching specifically, we show that the claustrum is activated at the onset of a demanding cognitive task, which is also associated with the onset of EMN engagement.

We isolated a distinct claustrum signal by regressing putamen and insula signals from the claustrum. This is a conservative approach, assuming that any linear relationship between claustrum and insula/putamen is a result of a partial volume effect, defined as signal from outside structures being erroneously incorporated into an ROI by virtue of the ROI only occupying a portion of a measured volume (Dukart and Bertolino, 2014; Du et al., 2014). Thus, while our approach mitigates the partial volume effect, a degree of true claustrum signal may be dampened. However, our resting state analyses showing similar RSFC across left/right claustrum, similar RSFC over two datasets, bilateral claustrum



**Fig. 4.** Task-related claustrum activation. A) Reaction time performance in MSIT varies based on the condition, with the difficult blocks having the slowest reaction times. B) The first trial of the easy (left graph) and difficult (right graph) conditions is behaviorally unique from the other trials in the block. As a result, we modelled the onset separately from the remainder of the block for difficult and easy conditions in an adaptive model to capture individual variability. C) Time course of claustrum from group averaged data shows a response during the difficult task, but exclusively during the onset of that task. The contrast map localizing EMN and the contrast map localizing DMN (Fig. 3A) were used as regions of interest. We extracted the signal from these areas along with the left and right claustrum averaged over every difficult condition for all subjects. The task starts at the third TR and ends at the 11th. TRs 1:2 and 12:15 show the control condition. Gray shaded areas is the canonical hemodynamic response function (HRF) convolved with an onset of duration 3.2 s (two trials in duration). D) Left and E) right claustrum average parameter estimate (regression slope) shows activation at the onset – but not block – of the difficult MSIT condition. Error bars show standard error of the mean. \* represents a p value of <0.05. HRF=Hemodynamic Response Function, CI = claustrum, DMN = default mode network, EMN = extrinsic mode network.

RSFC, and task response of the claustrum, argue that the signal that we do derive is sufficiently robust. This approach offers a generalizable method to assay the function of the claustrum, or other small or oddly shaped neural structures, independent of partial volume effects from surrounding structures. SRCC serves as a platform for studying a host of regions across the brain, such as the habenula (Shelton et al., 2012; Hetu et al., 2016) and other thalamic association nuclei.

Our claustrum functional connectivity data reveals co-activation of claustrum with executive cortical regions including ACC and medial PFC. This is in line with decades of neuronal tract tracing studies from rodents to monkeys (for review see Mathur, 2014), and particularly in line with reports in rats indicating dense claustrum connections with ACC and prelimbic PFC (Smith and Alloway, 2010; Smith et al., 2017; White et al., 2017), and with area 24 in the common marmoset (Reser et al., 2017). These data are also consistent with previous rodent functional imaging studies of the claustrum (Krimmel et al., 2019; Smith et al., 2017). We do not believe that these findings contradict recent work showing claustrum stimulation causes feed-forward-inhibition of the ACC (Jackson et al., 2018), because the ACC strongly excites the claustrum (White et al., 2017), likely leading to a net positive RSFC estimate. The present work also indicates that claustrum is functionally connected to posterior cingulate cortex, precuneus, angular gyrus, cuneus, visual cortex, and

sensorimotor cortex, which is in line with connections from claustrum to parietal association cortex and sensorimotor cortices in rats (White et al., 2017), connections from claustrum to visual cortex in cats (LeVay and Sherk, 1981) and projections from claustrum to parietal cortex in monkeys (Gamberini et al., 2017). The additional functional connectivity seen in 3T-Rest with midbrain and nucleus accumbens in particular, is likely to reflect indirect structural connectivity as direct connections between these regions are not currently supported in the tract tracing literature. Alternatively, functional connectivity of claustrum with these structures may reflect connections that are unique to the human brain.

Our data also suggest involvement of the claustrum at the start of a difficult condition when new rules come into play that require a change of cognitive strategy and task set instantiation (Dosenbach et al., 2006). Significant claustrum activation was only observed at the switch from the stimulus-response-based tapping task to the difficult condition of MSIT. As subjects transitioned to this new rule set, average reaction time decreased significantly compared with later trials in the task block. This transition was also met with an emergence of the task positive/extrinsic mode network. A possible interpretation of these data could be that the claustrum is involved in action inhibition as cognitive demand soars. However, neither optogenetic inhibition nor activation of axon terminals of a major excitatory input source to the claustrum, the ACC, affects



motor activity in mice (White et al., 2018). Our results cannot be explained through alterations in sensory binding or motor processes, which are both proposed roles for the claustrum, as we observed claustrum activation when controlling for sensory input and motor responses. While we did not observe a significant difference between easy block onset and difficult block onset in claustrum activation, easy block onset claustrum activation did not reach our statistical threshold, and therefore we cannot conclude that the claustrum serves a general task switching computation. The data also do not suggest a role for claustrum in resolving cognitive conflict, as claustrum did not show sustained activation during the block of the difficult condition, whereas the EMN did (Fig. 3A).

In summary, we showed that even at 1.5 mm spatial resolution attained with 7T fMRI, the claustrum BOLD signal bears unsettling similarity with the insula and putamen, and an additional processing step of SRCC allows for isolation of a claustrum signal that is uncorrelated to the surrounding regions. Using this method we find that in the human – like in other species – the claustrum has wide-ranging cortical connectivity; this includes functional connectivity with default and extrinsic mode networks and to sensory cortices. Additionally, we show that claustrum activity peaks when switching to a cognitive conflict task and that this activity could not be directly explained by changes in consciousness or sensorimotor processing. These data support a role of the claustrum in cognitive control and task switching.

## Conflicts of interest

No conflicts of interest to disclose.

## Author contributions

DAS and BNM conceived of the research. DAS, SK, MGW, and BNM designed research. MP generated claustrum masks. DAS, SK, FSB, MGW and BNM analyzed data. DAS, SK, and BNM wrote the manuscript.

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## Appendix A. Supplementary data

Supplementary data to this article can be found online at <https://doi.org/10.1016/j.neuroimage.2019.03.075>.

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