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## **Cerebrovascular reactivity mapping without gas challenges**

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### **Abstract**

Cerebrovascular reactivity (CVR), the ability of cerebral vessels to dilate or constrict, has been shown to provide valuable information in the diagnosis and treatment evaluation of patients with various cerebrovascular conditions. CVR mapping is typically performed using hypercapnic gas inhalation as a vasoactive challenge while collecting BOLD images, but the inherent need of gas inhalation and the associated apparatus setup present a practical obstacle in applying it in routine clinical use. Therefore, we aimed to develop a new method to map CVR using resting-state BOLD data without the need of gas inhalation. This approach exploits the natural variation in respiration and measures its influence on BOLD MRI signal. In this work, we first identified a surrogate of the arterial  $CO<sub>2</sub>$  fluctuation during spontaneous breathing from the global BOLD signal. Second, we tested the feasibility and reproducibility of the proposed approach to use the above-mentioned surrogate as a regressor to estimate voxel-wise CVR. Third, we validated the "resting-state CVR map" with a conventional CVR map obtained with hypercapnic gas inhalation in healthy volunteers. Finally, we tested the utility of this new approach in detecting abnormal CVR in a group of patients with Moyamoya disease, and again validated the results using the conventional gas inhalation method. Our results showed that global BOLD signal fluctuation in the frequency range of 0.02–0.04 Hz contains the most prominent contribution from natural variation in arterial  $CO<sub>2</sub>$ . The CVR map calculated using this signal as a regressor is reproducible across runs  $(ICC=0.91±0.06)$ , and manifests a strong spatial correlation with results measured with a conventional hypercapnia-based method in healthy subjects ( $r=0.88$ ,  $p<0.001$ ). We also found that resting-state CVR was able to identify vasodilatory deficit in patients with steno-occlusive disease,

### **Conflicts of interest**

The authors have no conflicts of interest or financial disclosures to report.

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the spatial pattern of which matches that obtained using the conventional gas method  $(r=0.71\pm0.18)$ . These results suggest that CVR obtained with resting-state BOLD may be a useful alternative in detecting vascular deficits in clinical applications when gas challenge is not feasible.

### **Keywords**

Cerebrovascular reactivity; resting-state BOLD fMRI; global signal; spontaneous  $CO<sub>2</sub>$  fluctuation; hypercapnia (CO<sub>2</sub>) inhalation; Moyamoya disease

### **1. Introduction**

Cerebrovascular reactivity (CVR) is a measure of the dilatory function of cerebral blood vessels. Compared to other vascular measures such as cerebral blood flow (CBF) and cerebral blood volume (CBV), CVR is thought to be a more specific indicator of vascular health. CVR mapping has been shown to provide valuable information in the evaluation of various cerebrovascular conditions, including arterial stenosis (Donahue et al., 2013; Gupta et al., 2012; Mandell et al., 2008; Mikulis et al., 2005), stroke (Geranmayeh et al., 2015), small vessel disease (Greenberg, 2006), brain tumors (Zaca et al., 2014), traumatic brain injury (Chan et al., 2015; Kenney et al., 2016), substance abuse (Han et al., 2008), and normal aging (Gauthier et al., 2013; Lu et al., 2011). CVR also has important utility in normalizing blood-oxygenation level dependent (BOLD) fMRI signal to differentiate neuronal from vascular alternations in brain function (Liu et al., 2013a; Liu et al., 2013b).

At present, CVR mapping is typically performed using hypercapnic gas inhalation as a vasoactive challenge while collecting perfusion sensitive MRI images (Lu et al., 2014; Spano et al., 2013; Wise et al., 2007; Yezhuvath et al., 2009). Hypercapnic gas inhalation increases the blood concentration of  $CO<sub>2</sub>$ , which, as a potent vasodilator, dilates blood vessels and increases cerebral perfusion (Brian, 1998). CVR can then be quantified by the BOLD signal changes associated with  $CO<sub>2</sub>$  inhalation. Although the BOLD signal is not a direct measure of CBF change in the brain and thus may be confounded by changes in cerebral oxygen metabolism (Xu et al., 2011), it is often preferred over perfusion signal (e.g. using arterial-spin-labeling MRI) due to its higher signal-to-noise ratio (SNR). This CVR mapping method has been successfully applied in a number of research studies in healthy and chronic disease patients (Donahue et al., 2013; Donahue et al., 2014; Han et al., 2008; Liu et al., 2013b; Lu et al., 2011; Mandell et al., 2008; Marshall et al., 2014; Mikulis et al., 2005; Thomas et al., 2013). However, the inherent need of gas inhalation and the associated apparatus setup requires additional time and expertise for handling and monitoring, which may limit the applications of this technique. This is especially the case when examining acute patients (e.g., acute stoke, acute traumatic brain injury). Breath-holding is another approach to manipulate blood  $CO<sub>2</sub>$  concentration (Kastrup et al., 1998; Murphy et al., 2011; Tancredi and Hoge, 2013; Zaca et al., 2014). But the requirement of subject's cooperation in performing breath-holding tasks also makes it difficult for patients with acute or severe conditions.

Therefore, in the present work, we aim to show the proof-of-principle of a new CVR mapping approach that does not require gas inhalation. This approach utilizes the natural

variation in respiration over time as an intrinsic vasoactive stimulus. A surrogate of fluctuations in arterial  $CO<sub>2</sub>$  can be extracted from global BOLD signal. Global, non-regionspecific fluctuation in BOLD MRI signal is known to be attributed to several mechanisms, including cardiac cycle, breathing cycle, scanner thermal noise, slow physiological variations, and potentially whole-brain neural fluctuation (Birn et al., 2006; Chang et al., 2009; Glover et al., 2000; Wise et al., 2004; Wong et al., 2013). We hypothesize that the signal component association with respiratory variation can be extracted by advanced acquisition and analysis schemes of the global BOLD signal, which can then serve as a regressor for voxel-wise estimation of CVR, all from the resting-state BOLD data. We conducted four studies to test this hypothesis. Firstly, we identified the component in the global BOLD signal that has the best correspondence to end-tidal (Et)  $CO<sub>2</sub>$  time course at rest. Secondly, we tested the feasibility and reproducibility of performing voxel-wise calculation of CVR using the global BOLD as a regressor. Thirdly, we validated the "resting CVR map" with the conventional CVR map obtained by hypercapnic gas inhalation in healthy volunteers. Finally, we tested the utility of this new approach in detecting abnormal CVR in a group of patients with Moyamoya disease, a cerebrovascular disease characterized by intracranial arterial stenosis, and again validated the results using the conventional gas inhalation methods.

### **2. Materials and methods**

### **2.1 General**

All MR imaging experiments were conducted on 3 Tesla MR system (Philips Medical System, Best, The Netherlands). Foam padding was placed around the head to minimize motion. The study protocols were approved by the Institutional Review Boards of the Johns Hopkins University School of Medicine, the University of Texas Southwestern Medical Center and the University of Texas at Dallas. Written informed consent was obtained from all participants before the MRI scans.

Four studies were performed in separate participants. Study 1 and 4 were performed solely for the purpose of this report. Study 2 and 3 used resting-state BOLD data collected as part of our previous studies (Liu et al., 2013a; Tung et al., 2013). Data from a total of 48 participants are included.

### **2.2 Study 1: Identification of CO2 related signal component from global BOLD**

It is known that the origins of the global BOLD signal include many components. Therefore, the purpose of this study was to experimentally determine which frequency range of the global BOLD signal at rest contains the best correspondence to blood  $CO<sub>2</sub>$  concentration. Five healthy volunteers (age  $35 \pm 13$  years, 3 males and 2 females) were scanned using a 32channel receive-only head coil. The body coil was used for RF transmission. During the resting-state BOLD scan, the subject wore a nasal cannula (Salter Labs, Arvin, CA) through which the  $EtCO<sub>2</sub>$  was recorded using a capnograph device (Capnogard, Model 1265, Novametrix Medical Systems, CT). Imaging parameters of the BOLD scan were: single-shot gradient-echo EPI, Field-of-view (FOV) =  $220 \times 220 \times 30$  mm<sup>3</sup>, voxel size=3.44 $\times$ 3.44 $\times$ 5mm<sup>3</sup>, TR/TE=270/20ms, SENSE factor = 2, 6 axial slices with 1mm gap, 900 dynamics. The

imaging slices were positioned parallel to the anterior commissure -posterior commissure (AC-PC) line, with the last slice centered on the AC-PC line.

Data analysis was conducted using the software Statistical Parametric Mapping (SPM) (University College London, UK) and in-house MATLAB (MathWorks, Natick, MA) scripts. Pre-processing of the BOLD image series included motion correction, smoothing using a Gaussian filter with a full-width-half-maximum (FWHM) of 8 mm, and linear detrending. The  $ECO<sub>2</sub>$  time course was shifted to account for the time it takes for the blood to travel from the lung to the heart and then to the brain, using a step-wise searching procedure described previously (Yezhuvath et al., 2009). The shifted  $E<sub>1</sub>ECO<sub>2</sub>$  time course was in sync with the BOLD images, and was used for the following analysis.

To identify the best frequency component for the extraction of  $CO<sub>2</sub>$ -related signal from the BOLD signal, the BOLD images were filtered into five different frequency bands: 0–0.01 Hz, 0.01–0.02 Hz, 0.02–0.04 Hz, 0.04–0.08 Hz, 0.08–0.2 Hz, and the global BOLD time course for each frequency band was obtained. The frequency width was set to be larger at higher frequencies because signal power decays at higher frequencies. Signal frequencies above 0.2 Hz were not tested because those signal components are expected to be primarily due to cardiac and breathing cycles and there is minimal  $ECO<sub>2</sub>$  signal in that frequency range (Birn et al., 2006; Chang et al., 2009; Wise et al., 2004). The BOLD signal time course was then spatially averaged to minimize scanner thermal noise, generating a reference BOLD time course for each frequency range. The cross-correlation coefficient (r) between the reference BOLD time course and  $ECO<sub>2</sub>$  time course was calculated for each frequency band. Paired t-tests were performed to compare the r values across the frequency bands to identify the band that provides the highest R, which was used as the optimal frequency range for CVR calculations in the following studies.

An exploratory analysis was also performed to compare the statistical significance (indicated by Z score) of voxel-wise regression using the filtered reference BOLD signal as regressor to that using  $ECO<sub>2</sub>$  time course as regressor following a previous study (Golestani et al., 2016). Z scores from all brain voxels were considered. Paired t-tests were performed to evaluate the significance of the comparison between the two methods. The p value <0.05 is considered significant.

### **2.3 Study 2: Feasibility and reproducibility**

Study 2 examined the feasibility of using resting-state BOLD fMRI to map CVR, and evaluated the reproducibility of this approach. This study was done by using resting state data collected in our previous study (Tung et al., 2013). Ten healthy subjects (age  $28\pm7$ years, 6 males and 4 females) underwent a 35-minute session consisting of 7 resting state BOLD scans of 5 minutes each. An 8-channel receive-only head coil was used. The BOLD imaging parameters were: single-shot gradient-echo EPI, TR/TE=1000/25ms, FOV=  $220 \times 220 \times 104$  mm<sup>3</sup>, voxel size  $3.4 \times 3.4 \times 4$  mm<sup>3</sup>, 21 axial slices with 1 mm gap covering the whole cerebrum, 300 dynamics. Compared to Study 1, the data in this study provided a substantially larger spatial coverage, although the temporal resolution was lower.

The data analysis followed the chart shown in Figure 1. The pre-processing was identical to that used in Study 1. The temporal filtering of the BOLD time course used the optimal frequency band determined from Study 1. A whole-brain mask obtained from a  $T_1$ -weighted MPRAGE image was used as a reference mask and the voxels inside the mask were averaged to yield the reference time course. A linear regression analysis was then performed in which the reference time course was the independent variable and individual voxel's time course was the dependent variable, yielding a CVR index for each voxel. By normalizing the CVR index to a mean value calculated from the reference mask, a resting-state CVR map was obtained. We note that the resting-state CVR map is in relative units. To evaluate the reproducibility of the resting state CVR maps, intraclass correlation (ICC) analysis was performed across multiple runs of the data (Shrout and Fleiss, 1979). For this study, twoway model, average-measures ICC values were calculated.

### **2.4 Study 3: Validation of resting-state CVR with gas-inhalation based method**

Twenty-six healthy subjects (age  $28 \pm 5$  years, 12 males and 14 females) underwent a 5 min resting-state BOLD scan and a 7 min hypercapnia BOLD scan (Liu et al., 2013a). An 8 channel receive-only head coil was used. The imaging parameters were identical between the two scans: TR/TE=2000/25ms, FOV=220×220×150.5mm<sup>3</sup>, voxel size  $3.4\times3.4\times3.5$ mm<sup>3</sup>, 43 axial slices without gap. During the hypercapnia scan, subjects breathed room-air and 5%  $CO<sub>2</sub>$  (mixed with 21% O2 and 74% N2) in an interleaved fashion (switching every 1 min) while BOLD EPI images were acquired continuously.  $ECO<sub>2</sub>$  was recorded throughout the breathing task.

A regression analysis between EtCO<sub>2</sub> and the MRI time course yielded the gas-inhalationbased CVR map (Lu et al., 2014; Yezhuvath et al., 2009). The resting-state CVR map (in relative units) was obtained following the procedures developed in Study 2 (Figure 1). Spatial correlation between the resting-state and gas-inhalation CVR maps was calculated as the Pearson product-moment coefficient. For easy comparison, the gas-inhalation CVR map was also normalized to its whole-brain value.

### **2.5 Study 4: Evaluation of Moyamoya patients**

Seven patients with Moyamoya Disease (age  $39 \pm 10$  years, 1 male and 6 females) were recruited. Moyamoya Disease is a steno-occlusive cerebrovascular disease characterized by severe narrowing/blockage of the anterior and middle cerebral vessels. The specific diagnosis of each patient in this study is listed in Table 1. Four patients had unilateral stenosis and three had bilateral stenosis. Each patient underwent a 9 min resting-state BOLD scan and a gas inhalation scan of 9 min. The imaging parameters for both the breathing task and the resting-state scan were: BOLD sequence,  $FOV = 205 \times 205 \times 151$  mm<sup>3</sup>, TR/ TE=1510/21ms, 3.2mm isotropic voxels, whole brain coverage using 36 slices with 1mm gap. The gas inhalation scan utilized the concomitant  $CO<sub>2</sub>/O<sub>2</sub>$  modulation paradigm so that CVR and venous CBV maps were obtained simultaneously. Details of the concomitant  $CO<sub>2</sub>/O<sub>2</sub>$  modulation and data analysis were described previously (Liu et al., 2016). An ASL scan was performed to obtain CBF. The imaging parameters of the ASL scan were: 2D multi-slice acquisition, labeling duration=1650ms, post-labeling delay = 1525ms, TR/ TE=4260/14 ms, FOV=240 $\times$ 240 $\times$ 145.5mm<sup>3</sup>, voxel size 3 $\times$ 3 $\times$ 5mm<sup>3</sup>, whole-brain coverage

using 29 axial slices without gap. Anatomical images of FLAIR,  $T_1$ -weighted MPRAGE, and time-of-flight (TOF) angiogram were also obtained. All patients were scanned with a 32-channel receive-only head coil. The body coil was used for RF transmission.

Resting-state and gas-inhalation CVR maps were obtained using procedures described above. Both maps are in relative units. In these analyses, we used cerebellum gray matter as the reference region because posterior circulation is usually preserved (as verified by the TOF angiogram) in Moyamoya patients. Deficit regions as identified by resting-state and gas-inhalation CVR were compared. Quantitative comparison between the two methods was performed using scatter plot and spatial correlation. CBF maps were obtained following the ASL white paper (Alsop et al., 2015).

### **3. Results**

### **3.1 Study 1**

Figure 2 shows cross-correlation coefficients between  $ECO<sub>2</sub>$  and BOLD signal time courses for different frequency ranges. It can be seen that the correlation is highest when the time courses are filtered at the 0.02–0.04 Hz frequency range. Paired t-tests showed that the correlation at 0.02–0.04 Hz is significantly higher than all other frequency bands except for 0.01–0.02 Hz (Figure 2). Indeed, the peak correlation appeared at the 0.02–0.04 Hz for every participant in the study. These observations suggested that BOLD signal at the frequency range of 0.02–0.04 Hz provides the best estimation of the spontaneous fluctuation of blood CO2 concentration, relative to other frequency bands. Therefore, in the following studies, the BOLD time series were all filtered to 0.02–0.04 Hz for the calculation of CVR map using resting-state data.

Comparing the methods using global BOLD signal (filtered to  $0.02-0.04$  Hz) and EtCO<sub>2</sub> time course, the statistical significance of voxel-wise regression is considerably greater when using the global BOLD signal ( $p=0.001$ ). The resulting voxel-wise Z scores were  $17.9\pm3.2$ and  $4.0\pm2.7$  for global BOLD and EtCO<sub>2</sub> methods, respectively. Similarly, we found that 90 $\pm$ 5% of voxels in the brain has a statistically significance Z score (Z>3) using the global BOLD signal as a regressor, as compared to  $50\pm20\%$  using EtCO<sub>2</sub> as the regressor  $(p=0.014)$ .

### **3.2 Study 2**

Figure 3 demonstrates the resting-state CVR maps obtained from seven consecutive scans from a representative subject. The maps showed clear gray matter – white matter contrast, which is consistent across different scans. The resting-state CVR maps of all scans from all subjects are shown in Supplementary Figure S1. The ICC values were found to be  $0.91 \pm 0.06$ (mean±standard deviation) across 10 subjects (range from 0.79 to 0.98), indicating a good reproducibility of the resting-state CVR maps.

### **3.3 Study 3**

Figure 4a displays group-averaged CVR maps obtained from the resting-state and hypercapnia scans. Visual inspection suggests that the two CVR maps, although obtained by

different techniques, have virtually identical image contrast. The gray matter has larger CVR than white matter, and voxels containing veins shows the highest CVR values due to the blood volume effect. The voxels containing veins were identified by visual inspection based on knowledge of the vessel anatomy of large draining veins. Consistency of the two CVR maps was confirmed by quantitative analysis of their scatter plot, shown in Figure 4b. There was a strong spatial correlation  $(r=0.88)$  between the two maps. The individual resting-state and hypercapnia-derived CVR maps of all subjects are shown in Supplementary Figure S2. To compare the sensitivity of the resting-state CVR method with the standard gas method, Figure 4c shows the histogram of Z-statistics of the resting-state and hypercapnia-based techniques. It can be seen that the resting-state method  $(Z=7.47\pm1.75, N=26)$  is less sensitive compared to the hypercapnia CVR mapping technique ( $Z=11.94\pm2.54$ , N=26), as expected.

### **3.4 Study 4**

The demographic information of the Moyamoya patients is shown in Table 1. Figure 5 shows CVR maps from resting-state and hypercapnia scans of all seven patients. For reference, the patients' MRI angiograms are also shown, illustrating the specific stenosis site of each patient (red arrows).  $T_1$ -MPRAGE, FLAIR, CBF and venous CBV images are also displayed. Reduced CVR was observed in the disease territories in all patients. Resting-state CVR maps were found to reveal a similar deficit pattern as the hypercapnia CVR maps, suggesting that CVR measured with resting-state BOLD fMRI can provide similar information of cerebrovascular dysfunction as that obtained using a conventional hypercapnia-based method. Scatter plots between the two CVR maps are shown in Figure 5 (bottom row). Their spatial correlation coefficient was  $0.71 \pm 0.18$ . We have also investigated resting-state CVR mapping using the frequency bands other than 0.02–0.04 Hz, and found that the resulted resting-state CVR maps were less consistent with the hypercapnia CVR map (Supplementary Figure S3).

Comparing to CBF and venous CBV maps, the two CVR maps showed their unique deficit pattern, suggesting that the resting-state CVR we obtained measures the specific vascular function as the hypercapnia CVR.

### **4. Discussion**

In this report, we propose a new method to map cerebrovascular reactivity without the need of gas inhalation. We identified that global BOLD signal fluctuation in the frequency range of 0.02–0.04 Hz contains the most prominent contribution from natural variation in arterial  $CO<sub>2</sub>$ . We then utilized this signal as a regressor for CVR estimation from the resting-state BOLD fMRI data. Our results showed that the CVR map calculated with this method is reproducible across runs, and manifests a strong spatial correlation with results measured with a conventional hypercapnia-based method. Resting-state CVR was able to identify vasodilatory deficits in patients with steno-occlusive disease, the spatial pattern of which is similar to that using conventional gas method.

CVR indicates the ability of blood vessels to dilate upon demand, and is thought to be a specific marker of vascular health compared to cerebral perfusion, which may be affected by non-vascular factors such as neural activity, metabolism, and consumption of caffeine (Chen

and Parrish, 2009; Fujishima et al., 1971). CVR has shown an important value in research studies involving chronic ischemic conditions such as atherosclerotic arterial steno-occlusive disease and Moyamoya disease (Donahue et al., 2013; Gupta et al., 2012; Mandell et al., 2008; Mikulis et al., 2005; Pandya et al., 2015). However, despite its potential, CVR mapping has not been used in large-scale clinical trials or routine clinical practice. One of the most important limitations of this method is that CVR mapping using gas-inhalation is typically viewed as cumbersome (although recent development in breathing apparatus (Lu et al., 2014; Tancredi et al., 2014) has begun to address these issues) and thus cannot be applied routinely. There are additional concerns of using gas challenges in acute or severely ill patients, as they may not tolerate or report their intolerance of hypercapnia. It has also been suggested that patients with significant pulmonary or cardiac disease should not be given gas challenge (Moreton et al., 2016). Therefore, there is significant clinical value to develop a CVR mapping method that does not require gas inhalation. The present work is one of such efforts to measure CVR using resting-state BOLD data. We have shown that resting-state CVR maps have a strong correlation with conventional, gas-inhalation based CVR maps, although its overall sensitivity, as measured by Z-statistic, is lower. We also note that the resting-state CVR mapping is based on the spontaneous fluctuation of blood  $CO<sub>2</sub>$ content, which is in the range of 2–4 mmHg. This natural variation is much smaller than the change induced by 5%  $CO<sub>2</sub>$  breathing (usually 8–10 mmHg), but is more relevant to the degree of daily neural activation induced blood flow changes. Given the non-linear relationship between blood CO2 content and CBF change (Fox et al., 1992; Ito et al., 2003), it is expected that the BOLD signal may also change non-linearly with blood CO2.

Resting-state fMRI data are known to contain information about the brain's vasculature, in addition to its common use of mapping functional connectivity (Birn et al., 2006; Wise et al., 2004). Kannurpatti and Biswal showed that the resting-state fluctuation amplitude (RSFA), computed as the temporal standard deviation of the BOLD time series, is correlated with CVR measured with breath-holding (Kannurpatti and Biswal, 2008). Jahanian et al. further showed that coefficient of variance (CV, standard deviation divided by mean) of the BOLD signal was significantly higher in hypertensive elderly subjects with chronic kidney disease than in young healthy volunteers (Jahanian et al., 2014). Other studies have explored a parameter that quantifying the degree of low-frequency variation of the BOLD signal, referred to as Amplitude of low-frequency fluctuations (ALFF), and have demonstrated a correlation between ALFF and hypercapnia-derived CVR (Di et al., 2013; Kazan et al., 2016). These methods have not attempted to differentiate various components of the BOLD signal, thus may contain a variety of physiological origins. In our study, we considered the multi-factorial mechanism of the global BOLD signal and developed specific acquisition and analysis schemes to maximally isolate signal component associated with natural fluctuation in blood CO2. The effects of cardiac and respiratory cycles on the BOLD time course were alleviated by acquiring data at a relatively high temporal frequency, thus such effects can be removed by low-pass filtering. Scanner thermal noise was minimized by spatially averaging a large number of voxels, e.g. the whole-brain, and by using the averaged signal as a regressor in CVR calculation. Further effort to obtain the signal component that is most relevant to  $CO<sub>2</sub>$  fluctuation was undertaken by identifying the frequency band (0.02–0.04 Hz) that shows the higher correlation with end-tidal  $CO<sub>2</sub>$ . A few studies have also been

reported to use resting-state fMRI data to estimate blood transit time in the brain (Amemiya et al., 2014; Christen et al., 2015; Tong and Frederick, 2014). We have tried to obtain the voxel-wise delay following a previous study (Christen et al., 2015) and apply it in restingstate CVR mapping. However, the resulted CVR maps were noisy, probably due to the low SNR in resting-state delay mapping. Therefore, we did not account for voxel-wise blood transit delay in our CVR analyses. As a result, the deficit regions in the CVR maps of the Moyamoya patients (Figure 5) reflect both diminished CVR and lengthened blood transit delay in those regions.

Recently, two studies reported the use of  $ECO<sub>2</sub>$  time course recorded during the restingstate BOLD scan as a regressor to map CVR in healthy volunteers (Golestani et al., 2016; Lipp et al., 2015). Compared to these previous studies, the present study is the first to apply these methodologies in patients with cerebrovascular diseases. Moreover, the present work has primarily focused on the use of filtered global BOLD signal as a regressor, rather than  $E<sub>1</sub>EC<sub>2</sub>$  as the regressor, due to the following reasons. First, when comparing the  $E<sub>1</sub>EC<sub>2</sub>$  and global BOLD methods, we found that the global BOLD method provided significantly better CVR maps (in terms of Z-scores). This is because  $E<sub>1</sub>CO<sub>2</sub>$  has a lower sampling frequency, e.g. one measure every 4 seconds, whereas the resulting BOLD signal change has a higher sampling frequency, e.g. one measure every 270 ms. Thus, a regression analysis between these two time series requires substantial temporal interpolation. One could downsample the BOLD images to match the sampling rate of EtCO2 and conduct the regression analysis. However, that means a large fraction of the BOLD data is discarded. Additionally,  $CO<sub>2</sub>$ concentration time course in the lung will be slightly different from that in the brain due to dispersion. This could further reduce the degree of association between  $ECO<sub>2</sub>$  and BOLD signal time courses. Moreover, the placement of  $E<sub>1</sub>CO<sub>2</sub>$  recording apparatus may take some additional setup time and wearing the apparatus for an extended period of time (i.e. having the tips of the nasal cannula at the nostril for the entire session, not just during the resting CVR scan) may cause some discomfort. For most cerebrovascular diseases which are usually regional, a relative map measured by resting-state CVR is sufficient to delineate deficit brain areas. If an absolute CVR (e.g. in  $\%$ /mmHg) is desired, EtCO<sub>2</sub> recording using nasal cannula (as done in our Study 1) can be added to the procedure. However, to obtain the highest statistical significance, we still recommend the use of filtered global BOLD as the regressor. The EtCO<sub>2</sub> recording is thus mainly used for the conversion from relative units to mmHg units.

This study has a few limitations. First, the resting-state CVR maps we obtained are in relative units, instead of in absolute unit of  $\%$ /mmHg CO<sub>2</sub>. As mentioned earlier, recording of  $ECO<sub>2</sub>$  can be added if absolute values are desired. Second, we have empirically found that the global BOLD signal within the frequency range of 0.02–0.04 Hz has the strongest correlation with  $ECO<sub>2</sub>$ . However, it is plausible that the signal in this frequency range still contains some contributions from non- $CO<sub>2</sub>$  related sources, for example, global fluctuations in neural activity due to vigilance changes (Liu, 2013; Wong et al., 2013), subject motion and heart rate changes. We are conducting further studies with simultaneous fMRI/EEG to investigate the extent of this contribution. Third, for Study 1, we did not have hypercapnia CVR map to calculate the spatial correlation between resting-state and hypercapnia CVR maps.

### **5. Conclusions**

In this work, we developed a new method to map cerebrovascular reactivity using restingstate BOLD data without the need of gas inhalation or breath-holding. Our results have shown that resting-state CVR maps obtained by this method are reproducible, and are highly consistent with conventional,  $CO_2$ -inhalation-based CVR maps. This method may be a useful alternative in detecting vascular deficits in clinical applications when gas challenge is not feasible.

### **Supplementary Material**

Refer to Web version on PubMed Central for supplementary material.

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### **Highlights**

- **•** A new method was proposed to map CVR without gas inhalation or breathholding
- **•** This method exploits natural variations in respiration and their effect on BOLD signal
- The resulted CVR maps were reproducible and consistent with CO<sub>2</sub>inhalation CVR maps
- **•** This method can identify CVR deficit in patients with steno-occlusive disease
- **•** This method may be a useful alternative to map CVR when gas challenges is not feasible





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### **Figure 2.**

The correlation between global BOLD signal and  $ECO<sub>2</sub>$  time course at different frequency ranges (N=5). Error bars indicate standard error. The p-values are uncorrected.

### BOLD Scan 1 Scan 2 Scan 3 Scan 4 Scan 5 Scan 6 Scan 7



### **Figure 3.**

CVR maps of seven consecutive resting-state BOLD fMRI scans from a representative healthy subject.

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### **Figure 4.**

Comparison between the group-averaged CVR obtained from resting-state scan (RS) and hypercapnia scan (HC) from 26 young healthy subjects. (a) The CVR maps obtained from the two scans. (b) Voxel-by-voxel scatter plot between the two maps (down-sampled by 5×5×5 voxels). (c) Average histograms of voxel-wise Z scores associated with CVR mapping using RS scan and HC scan, respectively.



### **Figure 5.**

Imaging results from the seven patients with Moyamoya diseases. Row 1: Time-of-flight (TOF) MR angiograms (MRA). Red arrows indicate the sties of arterial stenosis. Row 2: MPRAGE images. Row 3: T2-FLAIR images. Row 4: CBF maps. Row 5: venous CBV maps. Row 6: CVR maps using resting-state scan of these patients. Row 7: CVR maps from hypercapnia scan of these patients. Row 8: Scatter plots between the resting-state CVR map and hypercapnia CVR map for each patients.

# **Table 1**



