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Clinical Applications of Near-Infrared Diffuse Correlation Spectroscopy and Tomography for Tissue Blood Flow Monitoring and Imaging

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TOPICAL REVIEW

Clinical applications of near-infrared diffuse correlation spectroscopy and tomography for tissue blood flow monitoring and imaging

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4 **Clinical Applications of Near-infrared Diffuse Correlation Spectroscopy and**
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7 **Tomography for Tissue Blood Flow Monitoring and Imaging**
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Abstract

Blood flow is one such available observable promoting a wealth of physiological insight both individually and in combination with other metrics. Near-infrared diffuse correlation spectroscopy (DCS) and, to a lesser extent, diffuse correlation tomography (DCT), have increasingly received interest over the past decade as noninvasive methods for tissue blood flow measurements and imaging. DCS/DCT offers several attractive features for tissue blood flow measurements/imaging such as noninvasiveness, portability, high temporal resolution, and relatively large penetration depth (up to several centimeters). This review first introduces the basic principle and instrumentation of DCS/DCT, followed by presenting clinical application examples of DCS/DCT for the diagnosis and therapeutic monitoring of diseases in a variety of organs/tissues including brain, skeletal muscle, and tumor. Clinical study results demonstrate technical versatility of DCS/DCT in providing important information for disease diagnosis and intervention monitoring.

Table of Contents

1		
2		
3		
4		
5		
6		
7	1. Introduction	4
8		
9	2. DCS/DCT Methods	6
10		
11	2.1 Diffuse Correlation Spectroscopy (DCS)	6
12		
13	2.2 Diffuse Correlation Tomography (DCT)	9
14		
15	2.3 DCS/DCT Instrumentation	11
16		
17	3. Clinical Application Examples	12
18		
19	3.1 Brain	13
20		
21	3.1.1 Diagnosis of cardio-cerebral diseases	13
22		
23	3.1.2 Therapeutic monitoring of cardio-cerebral diseases	17
24		
25	3.2 Skeletal muscle	19
26		
27	3.2.1 Diagnosis of muscular diseases	19
28		
29	3.2.2 Therapeutic monitoring of muscular diseases	21
30		
31	3.3 Tumor	22
32		
33	3.3.1 Diagnosis of tumors	22
34		
35	3.3.2 Therapeutic monitoring of tumor treatments	24
36		
37	4. Summary and future perspectives	26
38		
39	Acknowledgments	30
40		
41	References	31
42		
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1. Introduction

Near-infrared (NIR) diffuse optical technologies have gained popularity in recent years as simple, fast, continuous, portable, and relatively inexpensive methods for noninvasive quantification of hemodynamics and metabolism in deep tissues up to several centimeters (Vardi and Nini, 2008; Schachner *et al.*, 2008; Bouye *et al.*, 2005; Boas *et al.*, 2001; Ferrari *et al.*, 2011; Jobsis, 1977; Murkin and Arango, 2009; Shuler *et al.*, 2009). One type of NIR technology, namely near-infrared spectroscopy (NIRS) or diffuse optical spectroscopy (DOS), takes advantage of the low absorption spectrum of biological tissues in the NIR range (650-950 nm) and penetrates deep tissues to detect light absorption by oxy-hemoglobin (HbO₂) and deoxy-hemoglobin (Hb) in red blood cells (RBCs) (Kim *et al.*, 2005; Strangman *et al.*, 2003). As a result, oxy-, deoxy-, total- hemoglobin concentrations (i.e., [HbO₂], [Hb], and THC) and tissue blood oxygen saturation (StO₂) in local tissue microvasculature can be quantified (Liu *et al.*, 1995; Boas *et al.*, 2001; Wolf *et al.*, 2003; Fantini *et al.*, 1995). Over past decades, a variety of NIRS/DOS technologies have been extensively explored to study various diseases affecting blood oxygenation levels in local and regional tissues (Liu *et al.*, 1995; Wolf *et al.*, 2003; Quaresima *et al.*, 2004; Fantini *et al.*, 1995; Patterson *et al.*, 1989; Al-Rawi and Kirkpatrick, 2006). Correspondingly, diffuse optical tomography (DOT) based on NIRS/DOS has also been developed for 3-dimensional (3-D) imaging of oxygenation distributions in organs or regional tissues (Intes *et al.*, 2010; Eggebrecht *et al.*, 2014; Dehghani *et al.*, 2009; Zhang *et al.*, 2013; Arridge and Hebden, 1997; Jermyn *et al.*, 2013).

Another emerging NIR technology, diffuse correlation spectroscopy (DCS), has also been developed to directly measure blood flow variations in deep tissue microvasculature (Boas *et*

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4 *al.*, 1995; Maret and Wolf, 1989; Pine *et al.*, 1988; Yu *et al.*, 2005a; Jaillon *et al.*, 2007; Boas,
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7 1996; Durduran, 2004; Choe, 2005; Zhou, 2007; Irwin, 2011; Gurley, 2012; Cheng, 2013; He,
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9
10 2015; Dong, 2015). DCS uses *coherent* NIR light to penetrate deep tissues and monitors
11
12 temporal light intensity fluctuation caused by moving scatterers (primarily RBCs in the
13
14 microvasculature) to extract a blood flow index (BFI) (Irwin *et al.*, 2011; Cheung *et al.*, 2001).
15
16
17 The relative change of blood flow (rBF) is calculated by normalizing BFI to its baseline value
18
19 before the physiological change occurs. DCS measurements of tissue blood flow changes
20
21 have been extensively validated against other standards, including power spectral Doppler
22
23 ultrasound (Yu *et al.*, 2005b), Doppler ultrasound (Buckley *et al.*, 2009; Roche-Labarbe *et al.*,
24
25 2010), laser Doppler (Durduran, 2004; Shang *et al.*, 2011a), Xenon computed tomography
26
27 (Xenon-CT) (Kim *et al.*, 2010), fluorescent microsphere flow measurement (Zhou *et al.*,
28
29 2009), and arterial spin labeling magnetic resonance imaging (ASL-MRI) (Yu *et al.*, 2007). In
30
31 some studies, DCS has been combined with NIRS/DOS in hybrid instruments to
32
33 simultaneously measure tissue blood flow and oxygenation (Durduran *et al.*, 2004; Yu *et al.*,
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35 2005a; Cheng *et al.*, 2012; Munk *et al.*, 2012; Shang *et al.*, 2012; Gurley *et al.*, 2012), which
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37 allows for the derivation of metabolic rate of tissue oxygen consumption.
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47 Despite advances in DCS applicability, there have been limited tomographic imaging
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49 realizations. An early contact-measurement based 3-D diffuse correlation tomography (DCT)
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51 approach using an optical fiber array was applied to tissue phantoms with a semi-infinite
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53 geometry (Boas and Yodh, 1997), but is disadvantaged *in vivo* due to the irregular geometries
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55 of biological tissues. A few noncontact-measurement based DCT examinations have been
56
57 recently conducted to avoid tissue hemodynamic variations induced by the probe compression
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4 or the disturbance of sensitive tissue areas (Culver *et al.*, 2003; Huang *et al.*, 2015a; Zhou *et*
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6
7 *al.*, 2006; Lin *et al.*, 2014; He *et al.*, 2015; Huang *et al.*, 2015b). In these noncontact
8
9 measurement systems, lenses were positioned between a sample and an optical fiber array
10
11 connected to the light sources and detection elements. The noncontact DCS/DCT has been
12
13 successfully applied in examining blood flow distributions in animal models (Zhou *et al.*,
14
15 2006; Culver *et al.*, 2003) and human subjects (Li *et al.*, 2013; He *et al.*, 2015; Lin *et al.*,
16
17 2012; Huang *et al.*, 2015c; Huang *et al.*, 2015b).

22
23 Since NIRS/DOS/DOT technologies for tissue blood oxygenation measurements have
24
25 been broadly reviewed (Durduran *et al.*, 2010a; Ferrari *et al.*, 2011; Ghosh *et al.*, 2012; Wolf
26
27 *et al.*, 2007), our review focuses on the DCS/DCT technologies for tissue blood flow
28
29 measurements in the clinic. We first introduce the principle and instrumentation of DCS/DCT.
30
31 We then provide typical clinical examples of DCS/DCT for the diagnosis of diseases and
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33 therapeutic monitoring of interventions in a variety of organs/tissues such as brain, skeletal
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35 muscle, and tumor. Finally, we highlight the limitations of DCS/DCT and point out future
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37 perspectives in technology development and clinical applications.
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43 44 45 **2. DCS/DCT Methods**

46 47 48 49 **2.1 Diffuse Correlation Spectroscopy (DCS)**

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51 DCS originates from the concept of “dynamic light scattering (DLS)”, a technology that
52
53 can be used to quantify the motion of moving scatterers in a thin solution through
54
55 investigating the correlation of light electric field after photons are scattered once by the
56
57 moving scatterers (i.e., single light scattering) (Brown, 1993; Fletcher, 1976). The extension
58
59 from single light scattering to multiple light scattering has been explored since 1980s (Maret
60

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4 and Wolf, 1987; Pine *et al.*, 1988; Boas *et al.*, 1995). This extension, called diffusing-wave
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6 spectroscopy (DWS) (Pine *et al.*, 1988; Li *et al.*, 2005; Jaillon *et al.*, 2006) or diffuse
7
8 correlation spectroscopy (DCS) (Cheung *et al.*, 2001; Yu *et al.*, 2005a; Gagnon *et al.*, 2008;
9
10 Boas and Yodh, 1997; Boas *et al.*, 1995), makes it possible to quantify the motion of moving
11
12 scatterers in a thick sample, such as red blood cells (RBCs) flowing through biological
13
14 tissues.
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18
19

20 The principle and instrumentation of DCS have been described elsewhere (Boas *et al.*,
21
22 1995; Cheung *et al.*, 2001; Durduran and Yodh, 2014; Irwin *et al.*, 2011; Yu, 2012b, a; Yu,
23
24 2012c). Briefly, a long-coherence laser delivers NIR light via an optical fiber to the tissue
25
26 (**Fig. 1a**). Photons entering into the tissue are either absorbed or more dominantly, scattered
27
28 by static scatterers (e.g., organelles and mitochondria) and dynamic scatterers (mainly moving
29
30 RBCs). Due to the scattering effect photons may change directions many times while
31
32 traveling throughout the tissue, resulting in light diffusion. Only a few photons can eventually
33
34 reach tissue surface and be collected by a single-mode detector fiber placed millimeters or
35
36 centimeters away from the source fiber. Photons collected by the detector fiber are detected
37
38 by a single photon-counting avalanche photodiode (APD) to yield light intensity signals (**Fig.**
39
40 **1b**).
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49 The detected light intensity fluctuates with time (**Fig. 1b**), which is caused by the motion
50
51 of moving scatterers (mainly RBCs) inside the tissue volume measured. To quantify the
52
53 motion of RBCs (i.e., blood flow) the normalized light intensity temporal autocorrelation
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55 function ($g_2(\tau)$) is calculated by a hardware correlator board or a software correlator, which
56
57 can then be converted to the normalized electric field temporal autocorrelation function ($g_1(\tau)$),
58
59
60

Fig. 1c) through the Siegert relation (Rice, 1954):

$$g_2(\vec{r}, \tau) = 1 + \beta |g_1(\vec{r}, \tau)|^2 \quad (1)$$

where τ is the correlation delay time, \vec{r} is the position vector, and β is a parameter which is dependent of laser stability, coherence length, and the number of speckles detected.

The $g_1(\tau)$ can be expressed in an integral form (Boas and Yodh, 1997; Li *et al.*, 2013; Pine *et al.*, 1988; Maret and Wolf, 1987)

$$g_1(\tau) = \frac{\langle E(0)E^*(\tau) \rangle}{\langle |E(0)|^2 \rangle} = \int_0^\infty P(s) \exp\left(-\frac{1}{3}k_0^2 \langle \Delta r^2(\tau) \rangle \frac{s}{l^*}\right) ds \quad (2)$$

Here $E(0)$ and $E^*(\tau)$ are the scattered light electric field at time 0 and its conjugation at time τ , respectively. $P(s)$ is the normalized distribution of detected photon path length s , k_0 is the wave vector magnitude of the light in the medium, and l^* is the photon random-walk step length, which is equal to $1/\mu_s'$ (μ_s' is the medium reduced scattering coefficient).

The unknown $\langle \Delta r^2(\tau) \rangle$ represents the mean-square-displacement of moving scatterers, which is conventionally difficult to be solved directly using **Eq. 2**. Alternately, the integral form of autocorrelation function can be converted to a partial differential equation form (Pine *et al.*, 1990). As such, the unnormalized electric field autocorrelation function $G_1(\tau) = \langle E(0)E^*(\tau) \rangle$ satisfies a correlation diffusion equation (Cheung *et al.*, 2001; Boas and Yodh, 1997)

$$\left(D\nabla^2 - \nu\mu_a - \frac{1}{3}\nu\mu_s'k_0^2 \langle \Delta r^2(\tau) \rangle \right) G_1(\vec{r}, \tau) = -\nu S(\vec{r}) \quad (3)$$

Here, ν is the light speed in the medium; μ_a is the medium absorption coefficient;

$D \approx \nu/3\mu_s'$ is the medium photon diffusion coefficient, and $S(\vec{r})$ is the continuous-wave

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4 isotropic source. The form of $\langle \Delta r^2(\tau) \rangle$ depends on a flow model specifically adopted. For
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7 example, a diffuse model, i.e., $\langle \Delta r^2(\tau) \rangle = 6D_B\tau$, has been found to fit experimental data well
8
9
10 over a wide range of different tissues (Cheung *et al.*, 2001; Irwin *et al.*, 2011). Here, D_B is the
11
12 effective diffusive coefficient. To account for the fact that not all scatterers are “moving” in
13
14 the tissue, a factor α , representing the ratio of “moving” scatterers to the total number of
15
16 scatterers, is added to $\langle \Delta r^2(\tau) \rangle$ (i.e., $\langle \Delta r^2(\tau) \rangle = 6\alpha D_B\tau$). The combined term, αD_B , is referred
17
18 to as blood flow index (BFI) in biological tissues. The αD_B can be extracted by fitting the
19
20 measured autocorrelation function curve to an analytical solution of the correlation diffusion
21
22 equation (Eq. 3) under certain geometries (e.g., semi-infinite boundary) (Cheung *et al.*, 2001;
23
24 Dong *et al.*, 2012a; Roche-Labarbe *et al.*, 2010; Irwin *et al.*, 2011).

2.2 Diffuse Correlation Tomography (DCT)

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33 The extension of DCS to DCT is comparable to that of DOS extension to DOT, which
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35 images the optical property distribution based on photon diffusion equation and its inverse
36
37 solutions (i.e., image reconstruction). A DOT instrument typically illuminates tissue and
38
39 measures the diffusive light leaving the tissue with multiple sources and detectors on the
40
41 tissue boundary (Intes *et al.*, 2010; Eggebrecht *et al.*, 2014; Dehghani *et al.*, 2009; Zhang *et al.*,
42
43 2013). The diffusive propagation of light in tissue can be characterized with photon
44
45 diffuse equation and parameterized in terms of the unknown $\mu_a(\vec{r})$ and $\mu'_s(\vec{r})$ at different
46
47 locations (\vec{r}) of the measured tissue volume. One can then “invert” the propagation model to
48
49 recover these unknown parameters. Imaging with DOT has been described in many papers
50
51 covering computer simulations, phantom tests, and *in vivo* applications (Franceschini *et al.*,
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53 2006; Durduran *et al.*, 2010a; Eggebrecht *et al.*, 2014; Arridge and Lionheart, 1998; Arridge
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and Hebden, 1997; Jermyn *et al.*, 2013; Dehghani *et al.*, 2009).

Similarly, DCT requires a large number of sources and detectors to measure autocorrelation functions on the tissue boundary. Early DCT approaches rely on analytical solutions of **Eq. 3** with the assumption of simple semi-infinite geometries and/or strict heterogeneities (e.g., spherical anomaly) of tissues (Zhou *et al.*, 2006; Culver *et al.*, 2003), which precludes the transition to complex boundaries and imperfect heterogeneities in realistic tissues.

The finite-element-method (FEM) framework has been applied to model light transport in highly diffuse media to resolve the geometry and heterogeneity limitations for DOS/DOT (Arridge and Lionheart, 1998; Arridge and Hebden, 1997; Eggebrecht *et al.*, 2014; Dehghani *et al.*, 2009; Jermyn *et al.*, 2013). For example, NIRFAST (www.nirfast.org) (Dehghani *et al.*, 2009; Jermyn *et al.*, 2013) and TOAST++ (<http://web4.cs.ucl.ac.uk/research/vis/toast/>) (Schweiger and Arridge, 2014) are publically available image reconstruction toolboxes based on the diffusive model and FEM for DOT

Exploiting the high mathematical similarity of the forward and inverse problems (e.g., boundary conditions and mathematical assumptions) between DOT and DCT, our group first reported the application of FEM in DCT (Lin *et al.*, 2014; Huang *et al.*, 2015a; He *et al.*, 2015; Huang *et al.*, 2015b). We introduced this concept into the FEM-based light transport and image reconstruction modules in NIRFAST (Dehghani *et al.*, 2009), as a shortcut for FEM implementation of DCT. Specifically, we used the “effective $\mu_a(\vec{r})$ ” (i.e., $\mu_a(\vec{r}) + 2\mu_s'(\vec{r})k_0^2\Delta\alpha D_b(\vec{r})\tau$) and measured $G_1(\vec{r}, \tau)$ to replace the $\mu_a(\vec{r})$ and photon fluence rate $\Phi(\vec{r}, t)$ respectively in the NIRFAST for the accomplishment of DCT in tissues

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4 with complex boundaries (Lin *et al.*, 2014; Huang *et al.*, 2015a; He *et al.*, 2015; Huang *et al.*,
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6
7 2015b). Computer simulations, phantom tests, and *in-vivo* studies have verified the accuracy
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9
10 of our modifications of NIRFAST adapted for DCT. As a result, the realm of DCT becomes
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12 available for comprehensive imaging under complex geometries and heterogeneous optical
13
14
15 properties.

16 17 **2.3 DCS/DCT Instrumentation**

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19
20 Typical DCS/DCT systems are illustrated in **Fig. 2**. The DCS/DCT systems consist of
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22 long-coherence NIR lasers (e.g., 785 nm, Crystalaser, USA), single photon-counting APDs
23
24 (e.g., SPCM-AQR-12, Perkin Elmer Inc., Canada), and digital correlator boards (e.g.,
25
26 FLEX03LQ, www.correlator.com, USA). A control panel (laptop or desktop) is used to
27
28 control the DCS/DCT system for data collection and calculation of temporal autocorrelation
29
30
31 function via the correlator board. Blood flow index (αD_B) can be extracted by fitting the
32
33 measured autocorrelation function curve to an analytical solution of **Eq. 3** under certain
34
35 geometries. In contrast to the portable DCS device (**b**) with a limited number of sources and
36
37 detectors, the DCT instrument (**c**) has a large number of sources and detectors for 3-D flow
38
39
40
41
42 imaging.

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45
46
47 A variety of fiber-optics probes have been designed for DCS (**Fig. 3**) and DCT (**Fig. 4**)
48
49 measurements in different applications (Huang *et al.*, 2015c; Durduran *et al.*, 2005; Yu *et al.*,
50
51 2011; Shang *et al.*, 2011b; Huang *et al.*, 2015b) (see **Section 3**). These probes are connected
52
53 to the DCS/DCT instruments (**Fig. 2**) through optical fibers for light delivery and detection.
54
55
56 Examples of DCS probes for the studies of brains, skeletal muscles, and tumors are illustrated
57
58 in **Fig. 3a-e**, respectively. For instance, a hand-held probe was designed to scan over a breast
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4 tumor in both horizontal and vertical directions for diagnostic purpose (Durduran *et al.*, 2005)
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6
7 (**Fig. 3d**). For the tissues whose shapes promote hemodynamic variations induced by
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9
10 compression or disturb sensitive areas, DCS probes were designed in a noncontact manner
11
12 (Huang *et al.*, 2015c) (**Fig. 3f**). In the noncontact measurement design, a lens system with
13
14 separated source and detector paths was utilized to focus the light from the source and
15
16 detector fibers on the measured tissue surface. The noncontact DCS (ncDCS) probe was also
17
18 extended to a noncontact imaging probe head for DCT measurements, which consisted of a
19
20 linear array of 15 photodetectors and two laser sources connected to a mobile lens-focusing
21
22 system (**Fig. 4**) (Lin *et al.*, 2014; He *et al.*, 2015; Huang *et al.*, 2015b). A motorized stage was
23
24 utilized for automatic scanning (**Fig. 4a**), enabling large ROI coverage and flexible S-D
25
26 arrangements without greatly increasing hardware requirements and costs. This new design of
27
28 noncontact DCT (ncDCT) probe, combined with a novel FEM framework for DCT image
29
30 reconstruction, has been validated in a tissue-like phantom with anomaly flow contrast design
31
32 (**Fig. 4d**) (Lin *et al.*, 2014) and has been applied in detection of breast tumors (He *et al.*, 2015;
33
34 Huang *et al.*, 2015b). In practice, any innovations adopted in DOS/DOT probes can be
35
36 adapted for the design of DCS/DCT probes, and a combined probe integrating DOS/DOT and
37
38 DCS/DCT measurements can be constructed by adding extra optical fibers.
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50 **3. Clinical Application Examples**

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53 DCS/DCT technologies have been extensively explored to be used in animal models and
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55 human subjects for the diagnosis and therapeutic monitoring of diseases in various tissues and
56
57 organs including brains (Cheng *et al.*, 2014; Hou *et al.*, 2014; Busch *et al.*, 2016a; Kim *et al.*,
58
59 2014; Kim *et al.*, 2010; Favilla *et al.*, 2014; Lin *et al.*, 2016; Dehaes *et al.*, 2014; Durduran *et*
60

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4 *al.*, 2010b; Shang *et al.*, 2011b; Buckley *et al.*, 2013; Zirak *et al.*, 2014), skeletal muscles
5
6
7 (Henry *et al.*, 2015; Mesquita *et al.*, 2013; Yu *et al.*, 2005a; Shang *et al.*, 2012; Gurley *et al.*,
8
9 2012), and tumors (Chung *et al.*, 2015; Choe *et al.*, 2014; Zhou *et al.*, 2007; Durduran *et al.*,
10
11 2005; He *et al.*, 2015; Sunar *et al.*, 2006; Dong *et al.*, 2012b; Dong *et al.*, 2016). Since this
12
13 review focuses on the clinical applications, only typical clinical examples are presented in the
14
15 following subsections based on different types of tissues/organs. For more details, readers are
16
17 encouraged to read primary papers that are cited but not discussed here and to learn from
18
19 recent reviews (Yu, 2012a; Yu, 2012c; Durduran and Yodh, 2014; Mesquita *et al.*, 2011).
20
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25 **3.1 Brain**

26 **3.1.1 Diagnosis of cardio-cerebral diseases**

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30
31 DCS has been explored for the diagnosis and evaluation of cardio-cerebral diseases
32
33 affecting cerebral blood flow (CBF) in adults, children, and neonates. For example, DCS was
34
35 utilized to investigate CBF responses in adult patients with cerebral diseases including
36
37 vasovagal syncope (Cheng *et al.*, 2014), obstructive sleep apnea-hypopnea (Hou *et al.*, 2014;
38
39 Busch *et al.*, 2016a), traumatic brain injury (Kim *et al.*, 2010; Kim *et al.*, 2014), and ischemic
40
41 stroke (Durduran *et al.*, 2009; Favilla *et al.*, 2014).
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48 Vasovagal syncope (VVS) is the sudden loss of consciousness, which can be fatal if
49
50 occurring in public such as driving and flying. In a study of CBF variations to predict VVS
51
52 (Cheng *et al.*, 2014), a 70-degree head-up-titling (HUT) protocol was applied to 14 healthy
53
54 adults, and relative changes of CBF (rCBF) to the baseline were continuously monitored
55
56 using a fiber-optic probe illustrated in **Fig. 3a**. In addition, relative changes of main artery
57
58 blood pressure (rMAP) were also continuously monitored using a noninvasive finger sensor
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60

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3
4 (Portapres, FMS Inc., Netherlands). In the subjects having presyncope symptoms during HUT,
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6
7 physiological responses to the tilting were observed to have two stages; while Stage I showed
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9
10 a small changes, Stage II demonstrated a rapid and dramatic decreases in both rCBF and
11
12 rMAP that were coincided with presyncope symptoms (**Fig. 5**). On average, rCBF reached the
13
14 Stage II earlier and decreased larger ($76 \pm 8\%$, from a baseline of 100%) than rMAP ($39 \pm$
15
16 19% , assigning a baseline value of 100%) during presyncope. Moreover, a threshold of $\sim 50\%$
17
18 rCBF decline was determined to completely separate the subjects with or without presyncope.
19
20
21 This study suggests that continuous monitoring of CBF variations by the portable DCS device
22
23
24 may provide predictive information to prevent VVS (e.g., using an automatic feedback to
25
26
27 maintain MAP and CBF at normal levels).
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31
32 Another example of DCS applications is the study of obstructive sleep apnea-hypopnea
33
34 (OSAH) (Hou *et al.*, 2014), a disease characterized by repetitive pausing of breath resulted
35
36 from upper airway obstruction during sleep. The impeded airflow during OSAH may cause
37
38 cerebral ischemia and disturb CBF. Although OSAH can be diagnosed through monitoring of
39
40 overnight sleep with a polysomnography, evaluation of cerebral ischemia and hypoxia
41
42 induced by OSAH is not routinely conducted, due to lack of appropriately technologies.
43
44
45 Using a dual-wavelength DCS flow-oximeter (Shang *et al.*, 2009), our group conducted
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47
48 continuous and simultaneous measurements of relative changes of CBF, oxy- and deoxy- and
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50
51 total hemoglobin concentrations (rCBF, $\Delta[\text{HbO}_2]$, $\Delta[\text{Hb}]$, ΔTHC) in adult subjects with
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53
54 OSAH for ~ 8 hours overnight (Hou *et al.*, 2014). Two fiber-optic probes connected to the
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56
57 DCS flow-oximeter were taped on both sides of the subject's forehead for cerebral monitoring
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60 during sleep (illustrated in **Fig. 3b**). To minimize the disturbance to patients, we remotely

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4 operated the optical measurements in a control room and designed special fiber-optic
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7 connectors which could easily disconnect the probes when the patient went to bathroom. The
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9
10 results showed that apneic events caused significant variations in rCBF and Δ THC. Moreover,
11
12 the degrees of these hemodynamic variations were significantly correlated with the severity of
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14
15 OSAH.
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17
18 Similarly, DCS was also adopted to measure CBF responses to hypercapnia during
19
20 wakefulness rather than during sleep in children with obstructive sleep syndrome (OSA)
21
22 and/or snores (Busch *et al.*, 2016a). Both OSA and snore subjects were found to have
23
24
25 significant less changes in CBF during hypercapnia than health controls, indicating the
26
27
28 diseased-induced blunted brain responses.
29

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31 DCS was also utilized to monitor CBF variations in adults with traumatic brain injury
32
33 (Kim *et al.*, 2014; Kim *et al.*, 2010) or ischemic stroke (Favilla *et al.*, 2014; Durduran *et al.*,
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35
36 2009). In those studies, CBF changes were found to be associated with large variations in
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39 cerebral blood pressure induced by physiological manipulations (Kim *et al.*, 2010; Favilla *et*
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41
42 *al.*, 2014; Durduran *et al.*, 2009), indicating the impairments of cerebral autoregulation to
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44
45 maintain a constant CBF.
46

47
48 In addition to the applications in adults and children, DCS was employed to assess CBF
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51 in neonates with malfunctions/defects and during surgical treatments, including those with
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54 congenital heart disease (CHD) (Durduran *et al.*, 2010b) , hypoxic ischemic encephalopathy
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56
57 (Dehaes *et al.*, 2014), single-ventricle CHD undergoing surgery (Dehaes *et al.*, 2015), as well
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60 as the neonates during open heart surgery (Busch *et al.*, 2016b). In those studies, DCS

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4 measurements benefited from higher signal-to-noise ratio (SNR) due to the thinner skulls of
5
6 neonates.
7

8
9 When subjected to a protocol of CO₂ inhalation for inducing hypercapnia (Durduran *et*
10
11 *al.*, 2010b), CBF values in neonates with CHD were elevated significantly ($158 \pm 6\%$,
12
13 assigning a baseline of 100%), which were detectable by both DCS and ASL-MRI
14
15 measurements. Since complex CHD may lead to impaired cerebral autoregulation and
16
17 low-baseline CBF, this study explored the potential of DCS for longitudinally probing
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19 cerebral defects in critically ill neonates.
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24
25 Apart from physiological challenging mentioned above, CBF values before and after
26
27 heart surgeries were measured and compared between the neonates with single-ventricle CHD
28
29 and healthy controls (Dehaes *et al.*, 2015). The results showed impaired cerebral development
30
31 in diseased neonates, as characterized by decreased CBF. Furthermore, CBF was
32
33 continuously monitored during neonatal open heart surgery, an intervention aiming to treat
34
35 heart defects (Busch *et al.*, 2016b). Significant variations in CBF during deep hypothermia,
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37 circulatory arrest, and rewarming were observed, indicating the capability of DCS for cerebral
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39 hemodynamic monitoring in a highly challenging clinical environment.
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48 In study of cerebral hemodynamics in premature neonates (Roche-Labarbe *et al.*, 2010),
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50 a steady increase (28%, from a baseline of 100%) in CBF was found over the first six weeks
51
52 of life, indicating that the neonates gained improvements in brain microcirculation since birth.
53
54 In another study of extremely preterm neonates, the subjects with hemorrhage were found to
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56 have lower CBF when compared with those without hemorrhage (Lin *et al.*, 2016). Similarly,
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4 lower CBF levels in neonates with hypoxic ischemic encephalopathy were reported,
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6 compared to age-matched healthy neonates (Dehaes *et al.*, 2014).
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10 In summary, many cardio-cerebral diseases affect nervous system, cerebral
11
12 microvasculature, and cerebral oxygen kinetics, leading to abnormality in CBF (Durduran and
13
14 Yodh, 2014; Mesquita *et al.*, 2011). Therefore, CBF is considered as an important biomarker
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16 of brain health, which links the oxygen demand, supply, and consumption. Continuous
17
18 measurements of CBF in the clinical setting, however, are not easily achievable due to
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20 technology unavailability. DCS offers a continuous, fast, portable, and low-cost tool to
21
22 noninvasively monitor CBF variations at the bedside of the clinic. Studies presented in this
23
24 subsection demonstrate that CBF abnormalities are associated with a variety of cerebral
25
26 diseases. While the explicit relationships between CBF abnormalities and clinical outcomes
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28 need to be further investigated, these studies support DCS as a useful tool for the diagnosis of
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30 cerebral diseases in adults, children, and neonates.
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39 **3.1.2 Therapeutic monitoring of cardio-cerebral diseases**

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41 DCS has been used to evaluate the effects of surgical interventions on CBF in adults and
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43 infants. Those surgeries included carotid endarterectomy in adults (Shang *et al.*, 2011b),
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45 thrombolysis in adults (Zirak *et al.*, 2014), and cardiac surgery in infants (Buckley *et al.*,
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47 2013).
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52 Carotid endarterectomy (CEA) is a surgical intervention to restore the blood supply to
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54 brain through removal of the blockage in carotid artery (Shang *et al.*, 2011b). During CEA,
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56 some of major vessels are temporarily clamped, causing a shutdown of blood supply to local
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58 cerebral tissues. Thus, intraoperative monitoring of CBF variation is particularly important as
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4 it permits acute assessment of cerebral hemodynamic level during surgery and post-surgical
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6 hemodynamic improvement in the brain. Electroencephalography (EEG) is an often-used
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8 technology for cerebral monitoring during CEA, which offers indirect assessment of cerebral
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10 ischemia through analysis of brain waves. In a comparison study of EEG and DCS
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12 measurements for evaluating cerebral hemodynamic variations during CEA, EEG electrodes
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14 were placed all over the scalps of eleven patients undergoing CEA, and two optical probes
15
16 were taped on both sides of forehead for simultaneous measurements of CBF and cerebral
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18 oxygenation. The internal carotid artery (ICA) clamps during CEA resulted in significant
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20 CBF decreases ($-24.7 \pm 7.3\%$, assigning a baseline of 100%) at the surgical sides (**Fig. 6a**).
21
22 On the other hand, post-CEA CBF values were significantly higher ($+43.2 \pm 16.9\%$, assigning
23
24 a baseline of 100%) than pre-CEA CBF values. Flow compensation via circle of Willis was
25
26 also observed at the nonsurgical side (**Fig. 6b**). CBF responses to ICA clamping were found
27
28 to be significantly faster, larger, and more sensitive than EEG responses (**Fig.7**).
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30 Simultaneous monitoring of CBF and EEG provided a comprehensive evaluation of cerebral
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32 physiological status and showed potential for the adoption of acute interventions (e.g.,
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34 shunting, medications) during CEA to reduce the risks of severe cerebral ischemia and
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36 cerebral hyperperfusion syndrome.
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49 DCS was also applied to a patient with acute ischemic stroke, who received
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51 thrombolytic therapy to restore CBF (Zirak *et al.*, 2014). CBF was monitored by DCS
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53 throughout the 60-minute thrombolytic therapy. DCS data showed an acute improvement in
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55 CBF after thrombolytic therapy, which agreed with the improvement in stroke scale score.
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60 This study suggested the potential of continuous DCS monitoring for therapeutic evaluation

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4 of acute stroke.
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7 DCS measurements of CBF during interventions were also reported in infants. For
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9 example, CBF, cerebral oxygenation, and cerebral oxygen metabolism were continuously
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11 monitored during the period of postoperative neonatal cardiac surgery using a hybrid optical
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13 instrument (Buckley *et al.*, 2013). Due to the complicated communications between the
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15 cerebral and cardiac vessels, the surgeries on hearts were found to significantly affect cerebral
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17 oxygen extraction fraction, but not CBF and cerebral metabolic rate of oxygen consumption.
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23 In summary, many intervention strategies for cardio-cerebral diseases are to restore
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25 tissue blood flow. Studies presented in this subsection show high sensitivity of DCS for
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27 continuous monitoring of CBF variations during interventions, which holds the potential for
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29 dynamic assessment and optimization of cardio-cerebral interventions to maintain CBF in a
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31 normal level.
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35 36 37 **3.2 Skeletal muscle**

38 39 40 **3.2.1 Diagnosis of muscular diseases**

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43 DCS has been adopted to evaluate the vascular diseases affecting skeletal muscle
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45 perfusion such as fibromyalgia and peripheral arterial disease (Shang *et al.*, 2012; Mesquita *et*
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47 *al.*, 2013). Evaluation of muscular diseases was often performed in a dynamic manner
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49 because those diseases restrict patients from performing leg or arm exercises. Blood flow
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51 responses to exercise reflect how skeletal muscles properly perform physiological functions.
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57 In a study of fibromyalgia (FM) (Shang *et al.*, 2012), women with FM were instructed to
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59 perform knee extension exercises with steadily increasing intensity. Calf muscle
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4 hemodynamics was continuously monitored by a hybrid instrument during exercise using a
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6 probe illustrated in **Fig. 3c**. Results showed that subjects with FM had significantly lower
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8 oxygen extraction rate during exercise than healthy controls, and the time of oxygenation
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10 recovery were significantly longer. These results suggested an alteration of muscle oxygen
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12 utilization in the FM population.
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18 In another study investigating the influence of peripheral arterial disease (PAD) on
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20 skeletal muscle hemodynamics (Mesquita *et al.*, 2013), calf blood flow values were found to
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22 occur at more depressed levels in PAD population after treadmill and pedal exercises,
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24 compared to those in age-matched healthy people. These results indicated adverse impact of
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26 PAD on blood flow responses to exercise.
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32 Limitations existed when using DCS techniques to continuously monitor muscle
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34 hemodynamic changes *during* exercise, such as the nature of relative measurement (rBF) and
35
36 the motion artifact induced by the exercise. To overcome these limitations, our group recently
37
38 created a calibration protocol for absolute flow measurements and a gating algorithm to
39
40 minimize motion artifacts during exercise (Gurley *et al.*, 2012; Henry *et al.*, 2015). Briefly, a
41
42 hybrid DCS/NIRS instrument along with a pre-exercise arterial occlusion protocol was
43
44 utilized to quantify the absolute baseline blood flow value before exercise. The absolute
45
46 baseline flow value was then used to calibrate the measured rBF during exercise in order to
47
48 obtain absolute flow measurements over the entire exercising period. Motion artifacts were
49
50 minimized using a novel dynamometer-based gating algorithm embedded in the DCS control
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52 software. Essentially, this software determined muscle contraction status based on signals
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54 from the dynamometer and recorded data only when muscle fiber motion was minimal.
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4 The innovative calibration and gating algorithms have been recently applied on patients
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7 with FM and compared with age-matched healthy controls. Blood flow responses in forearm
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10 muscles to handgrip exercise were continuously monitored and patients with FM exhibited
11
12 less capability to perform handgrip exercise than healthy controls, evidenced by the weaker
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14
15 blood flow responses during exercise (**Fig. 8**).

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17 Overall, deficiency of skeletal muscle function may result from mitochondrial
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20 dysfunction (Cordero *et al.*, 2010), lower capillary density (Lindh *et al.*, 1995; Morf *et al.*,
21
22
23 2005), reduced capillary permeability (Grassi *et al.*, 1994) and impaired vasodilatory capacity
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25
26 (Kasikcioglu *et al.*, 2006; McIver *et al.*, 2006), all of which are difficult to measure
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28
29 noninvasively. Hybrid DCS/NIRS devices enable noninvasive dynamic quantification of
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32 blood flow, blood oxygenation, and oxidative metabolism during exercise, providing unique
33
34
35 and comprehensive diagnostic information for skeletal muscle diseases.

36 **3.2.2 Therapeutic monitoring of muscular diseases**

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39 A few studies have been reported using DCS to investigate the impacts of surgical
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41
42 interventions on skeletal muscle diseases (Yu *et al.*, 2011; Huang *et al.*, 2015c). For muscular
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44
45 revascularization, ultrasound Doppler is routinely used in surgical rooms to test post-surgical
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48 reperfusion in large vessels. However, ultrasound Doppler cannot directly assess the
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51 restoration of tissue microcirculation, which is generally the major goal of revascularization.

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53 Using a portable DCS flow-oximeter and the probe illustrated in **Fig. 3c** (Shang *et al.*,
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55
56 2009), we continuously monitored calf muscle blood flow variations on twelve human limbs
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59 undergoing artery revascularization (Yu *et al.*, 2011). The high sensitivity of DCS in
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detecting blood flow changes was confirmed through clamping or ballooning on/off the

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4 femoral or aorta arteries during surgeries. Furthermore, immediate hyperemia following the
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6
7 release of occlusions and post-surgery flow improvements in calf muscles were observed,
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9
10 indicating the potential of the DCS for objective assessment of revascularization effects.

11
12 Another representative application of DCS is the assessment of blood flow in free
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14 transfer muscle flaps following salvage surgeries in patients with head and neck cancer.
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16
17 Conventional approaches to evaluate the success of tissue flaps rely on visual examination
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20 along with blood flow measurements in large vessels by ultrasound Doppler. DCS offers a
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22
23 tool to directly assess tissue blood flow in peripheral muscle flaps, which is crucial to
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25
26 evaluate the success of surgeries. For intraoperative monitoring of reconstructive flaps, we
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28
29 designed a lens system which focused the sources and detectors on the tissue surface,
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31
32 permitting noncontact measurements of flap blood flow (see **Fig. 3f**) (Huang *et al.*, 2015c).
33
34 The noncontact DCS probe was applied to eight free muscle flaps at multiple time points of
35
36
37 blood flow measurements during and post the surgical operations. **Fig. 9** shows comparison
38
39
40 results obtained from seven successful flaps and one unsuccessful flap. Postoperative blood
41
42
43 flow values in the successful flaps were significantly higher than the intraoperative baseline
44
45
46 values, indicating a gradual recovery of flap vascularity after the tissue transfer. By contrast,
47
48
49 postoperative blood flow recovered much less in the unsuccessful flap. Measurement of blood
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51
52 flow recovery after flap anastomosis holds the potential to act early to salvage ischemic flaps.

53 **3.3 Tumor**

54 **3.3.1 Diagnosis of tumors**

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57 The first translational case of DCS for clinical cancer diagnosis was reported on human
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59
60 breast tumors (Durduran *et al.*, 2005). Through scanning over the breast tumors (n = 5) using

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4 a contact optical probe illustrated in **Fig. 3d**, higher blood flow contrasts were found in
5
6 malignant (230% by averaging over three subjects) and benign (153% by averaging over two
7
8 subjects) tumors, respectively, as compared to the surrounding normal tissues (assigning
9
10 100%). Similar measurement configuration was thereafter applied to a larger population with
11
12 breast tumors (n = 32) (Choe *et al.*, 2014), confirming the high blood flow contrasts in breast
13
14 tumors (225~227% in mean value and 190~270% in 95% confidence intervals).
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20
21 Moreover, a recent pilot study investigated the correlations between tumor hemodynamic
22
23 parameters (i.e., blood flow, [HbO₂], [Hb], StO₂, THC) and tumor histopathological
24
25 biomarkers. Increased blood supply to breast tumors was observed, which agreed with the
26
27 high expression level of Ki67 nuclei in the confirmed breast tumors (Chung *et al.*, 2015). This
28
29 study suggested that the macroscopic measurements of tissue hemodynamics could reveal
30
31 pathological properties of breast cancer in microscopic level.
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36
37 A significant problem with contact DCS measurements described above is the deformation
38
39 of soft breast tissue, which may distort blood flow distribution in the soft breast. To solve this
40
41 problem, noncontact DCS/DCT (ncDCS/ncDCT) systems with unique fiber-optic probes were
42
43 recently developed enabling fully noncontact measurement/imaging of blood distributions in
44
45 deep tissue volumes with complex boundaries (Lin *et al.*, 2014; Lin *et al.*, 2012; Li *et al.*,
46
47 2013; Huang *et al.*, 2015c; He *et al.*, 2015; Huang *et al.*, 2015b). The noncontact probe was
48
49 attached to a motorized stage that scanned linearly or rotationally over a ROI (**Fig. 4**). **Fig. 10**
50
51 shows *in vivo* imaging results from two breast carcinomas. Higher blood flow contrasts (5.9-
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53 and 10.9-fold) in the tumor regions compared to the surrounding tissues were observed. The
54
55 reconstructed locations of the two tumors matched ultrasound imaging results when the tumor
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4 was within the sensitivity region of diffuse light. The ncDCT system showed the promise to
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6
7 image blood flow distributions in soft and vulnerable tissues without distorting tissue
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10 hemodynamics.

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12 The autonomic growth and spread of tumors are dependent on increased angiogenesis
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14 arising from the increased metabolic demand. Since functional alternations in tumors often
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17 appear earlier than detectable morphological changes, functional imaging of tumor blood flow
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20 by DCT is a new strategy for early cancer diagnosis and localization.
21

22 **3.3.2 Therapeutic monitoring of tumor treatments**

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24
25 It has been found that the patients with hypoxic tumors show incomplete clinical
26
27
28 responses when receiving chemo-radiation therapy that requires tissue oxygen for treatment
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30
31 efficacy (Busch *et al.*, 2000; Carlson, 2006). Many tumors are hypoxic because of abnormal
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34 vasculature, malignancy-related anemia, and/or high oxygen consumption by tumor cells.
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36
37 Studies have shown that pretreatment tumor hypoxia is associated with significantly poor
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40 responses to the therapy when compared to oxygenated tumors. However, some
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42
43 well-oxygenated tumors failed to respond while some hypoxic tumors responded well,
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45
46 possibly due to the dynamic changes during treatment in tumor oxygen status induced by
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48
49 radiation. Therefore, repeated monitoring of individual tumor hemodynamic status during
50
51
52 therapy may provide predictive information for treatment outcomes.

53
54 To date, DCS has been utilized to monitor the tumor hemodynamic responses to
55
56
57 chemo-radiation therapy in breast tumors (Zhou *et al.*, 2007) and head/neck tumors (Sunar *et*
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59
60 *al.*, 2006; Dong *et al.*, 2012b; Dong *et al.*, 2016). In a case study of breast cancer using a
hybrid DCS/NIRS device, significant changes in tumor blood flow and blood oxygenation

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4 were observed in the first week of chemo-radiation therapy (Zhou *et al.*, 2007).
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7 In another pilot study of head/neck tumors using a hybrid DCS/NIRS instrument, tumor
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9 hemodynamic responses in a small group of patients (n = 8) were continually measured once
10
11 a week over the period of chemo-radiation therapy (Sunar *et al.*, 2006). Tumors exhibited
12
13 significant dynamic flow and oxygenation changes during the first four weeks of the
14
15 treatment. This study, however, was limited by the small number of patients examined and
16
17 only one patient out of 8 showed a partial response to the treatment.
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23 Very recently, our group employed a hybrid DCS/NIRS instrument to continually
24
25 monitor tumor hemodynamic responses to chemo-radiation therapy for early prediction of
26
27 treatment outcomes in a relative large patient population with head/neck cancers (Dong *et al.*,
28
29 2016). Forty-seven patients were measured once per week over 7 weeks of treatment period to
30
31 evaluate the hemodynamic status of clinically involved cervical lymph nodes. Patients were
32
33 classified into two groups: complete response (CR) (n = 29) and incomplete response (IR) (n
34
35 = 18). Interestingly, tumor hemodynamic responses were found to be associated with clinical
36
37 outcomes (CR/IR), wherein the associations differed depending on human papillomavirus
38
39 (HPV-16) status. In HPV-16 positive patients, significantly lower levels in tumor [HbO₂] at
40
41 weeks 1 to 3, THC at week 3, and StO₂ at week 3 were found in the IR group. In HPV-16
42
43 negative patients, significantly higher levels in tumor BFI and μ_s' at week 3 were observed in
44
45 the IR group. These hemodynamic parameters exhibited significantly high accuracies for
46
47 early prediction of clinical outcomes, within the first three weeks of therapy, with the areas
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49 under the receiver operating characteristic curves (AUCs) ranging from 0.83 to 0.96 (**Fig. 11**).
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4 Overall, tumor hemodynamic parameters exhibited significantly high accuracies for early
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6 prediction of clinical outcomes. Such predications, based on frequent optical measurements,
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8 may ultimately be used to optimize individual therapeutic outcomes at an early time of
9
10 therapy. For example, treatment outcomes may be improved by dynamically promoting
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12 oxygenation levels (e.g., hyperbaric oxygen therapy) in HPV-16 positive tumors or inhibiting
13
14 angiogenesis (e.g., anti-VEGF antibody) in HPV-16 negative tumors.
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20 **4. Summary and future perspectives**

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22
23 In contrast to large imaging modalities such as CT, MRI, and PET, optical instruments
24
25 such as NIRS/DOS/DOT and DCS/DCT are portable, fast, inexpensive, and suitable for
26
27 continuous measurements at the bedside of clinical settings. NIRS/DOS/DOT techniques have
28
29 been extensively used to measure tissue oxygenation in the clinic. Compared to
30
31 NIRS/DOS/DOT, DCS/DCT is a relatively new technique that enables direct measurement of
32
33 tissue blood flow. DCS/DCT or hybrid DCS/NIRS is being increasingly employed worldwide
34
35 in a large variety of clinical contexts for the diagnosis and therapeutic monitoring of various
36
37 diseases affecting tissue hemodynamics and metabolism. Overall, these pilot clinical studies
38
39 in relatively small populations have shown the technique's versatility and demonstrated that
40
41 the method provides new and complementary information about patient pathophysiology in a
42
43 noninvasive fashion. It should be noticed that DCS/DCT has only been applied to a small
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45 population of patients with short measurement time frames (minutes to several weeks).
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4 There has been a concern on the physical modeling of blood flow (i.e., the motion of red
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7 blood cells) in the microvasculature. It was found that diffusive motion (see Section 2.1) fits
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9
10 the experimental autocorrelation curves rather well over a broad range of tissue types
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12 (Durduran, 2004; Zhou, 2007). Intuitively, however, random ballistic flow would be
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14 considered as a better model to fit the DCS/DCT data. Modified mixture models were
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16 proposed to include both ballistic flow and diffusive motion (Carp *et al.*, 2011; Boas *et al.*,
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18 2016). Experimental data supported the proposed models for capturing the transition from
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20 early ballistic to subsequent diffusive motion. Computer simulations on the tissue with varied
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22 vessel diameters and spacing showed that the diffusive motion dominates the correlation
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24 decay in typical DCS measurements, and the blood flow index is modulated proportionally by
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26 the concentration of hemoglobin and the average diameter of blood vessels. Nevertheless,
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28 more clinical investigations are needed for accurate extracting of blood flow information in
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30 deep tissues.
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38 Some technical issues exist when applying NIR diffuse optical technologies to deep
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40 tissues with heterogeneous properties. Since NIRS and DCS techniques rely on the transport
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42 of NIR light through top layer tissues (e.g., skin, skull), both methods must account for partial
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44 volume contributions from the top layer-structure to the deep tissue signals (e.g., brain,
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46 muscle, and tumor). In addition, measurement signal-to-noise ratio (SNR) and penetration
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48 depth are two interrelated and important parameters that affect the utility of both DCS and
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50 NIRS. Furthermore, most research with DCS reports only relative changes of blood flow
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52 index with respect to some baseline condition.
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4 The “partial volume effect” substantially affects the accuracy of DCS measurements, as
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7 discussed in literature (Durduran and Yodh, 2014; Yu, 2012a; Strangman *et al.*, 2003).
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9 Slab-layered models were proposed to reduce the partial volume effect (Jaillon *et al.*, 2006; Li
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11 *et al.*, 2005; Verdecchia *et al.*, 2016). A recently developed method, “Modified Beer-Lambert
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13 law for blood flow”, has been proved to be effective in pressure modulation experiments to
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15 reduce the skin-effect on cerebral blood flow measurements (Baker *et al.*, 2014; Baker *et al.*,
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17 2015). However, those methods ignored the influence of irregular tissue geometries. Recently,
18
19 we created a new algorithm integrating a linear model of autocorrelation function with the
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21 Monte Carlo simulation of photon migrations in heterogeneous tissues with arbitrary
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23 geometries for simultaneous extraction of blood flow indices in multiple layered tissues
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25 (Shang and Yu, 2014).
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33 Compared to NIRS/DOS/DOT, DCS/DCT measurements have relatively lower SNRs
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35 when probing deep tissue blood flow with large S-D separations (e.g., >2.5 mm). This is due
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37 to the utilization of single-mode detector fibers with a small core diameter of 5 μm in
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39 DCS/DCT measurements to ensure the detection of autocorrelation functions of light intensity.
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41 Efforts have been made to improve SNR through the spatial average of multiple
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43 autocorrelation functions detected by a detector fiber bundle (Dietsche *et al.*, 2007) or the
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45 temporal average of DCS signals obtained by a fast software correlator (Wang *et al.*, 2016).
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52 A potential remedy for the absolute blood flow measurement problem is to calibrate
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54 DCS/DCT against a gold standard. As mentioned early, comparison studies in small
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56 populations have been done in skeletal muscles, tumors, and brains against other established
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58 methods including power spectral Doppler ultrasound (Yu *et al.*, 2005b), Doppler ultrasound
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4 (Buckley *et al.*, 2009; Roche-Labarbe *et al.*, 2010), laser Doppler (Durduran, 2004; Shang *et*
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7 *al.*, 2011a), Xenon-CT (Kim *et al.*, 2010), fluorescent microsphere flow measurement (Zhou
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10 *et al.*, 2009), and ASL-MRI (Yu *et al.*, 2007). Results from these studies show the promise to
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12 obtain absolute flow measurements although further validation works need to be done with
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15 different types of tissues in large populations.
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18 DCT/ncDCT enables 3-D imaging of blood flow distributions in deep tissues. While
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20 effective, ncDCS/ncDCT employs a limited number of expensive APDs for blood flow
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23 detection, leading to low spatiotemporal resolution and high instrument cost (Lin *et al.*, 2014;
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26 He *et al.*, 2015; Huang *et al.*, 2015b). With current ncDCT, the ROI must be mechanically
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28 scanned for a complete tissue coverage, which can take up to 40 minutes and may result in
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31 motion artifacts. To overcome these limitations, we recently developed and tested a
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34 non-scanning, noncontact, fast, portable, cost-effective device, namely speckle contrast
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37 diffuse correlation tomography (scDCT), for 3-D imaging flow distributions (Huang *et al.*,
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40 2015a; Huang *et al.*, 2016). The scDCT used a charge-coupled-device (CCD) as a 2-D
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43 detector array to cover a ROI, thus eliminating the mechanical scanning and consequent
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46 problems. Thousands of detectors provided by the CCD significantly improved
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49 spatiotemporal resolution and reduced instrument cost/size. The scDCT prototype has been
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52 tested using computer simulations and tissue phantoms with anomaly flow contrast design.
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55 We are currently translating this novel optical technique to the clinic.

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58 It is well known that many diseases are associated with tissue hypoxia, which is
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61 influenced by oxygen supply (i.e., blood flow) and tissue oxygen consumption. Simultaneous
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64 measurements of tissue blood flow and blood oxygenation using combined DCS and NIRS

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4 instruments enable the evaluation of tissue metabolic rate of oxygen consumption (Durduran
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7 *et al.*, 2004; Roche-Labarbe *et al.*, 2010; Lin *et al.*, 2016; Dehaes *et al.*, 2014; Durduran *et al.*,
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10 2010b; Buckley *et al.*, 2013; Henry *et al.*, 2015; Shang *et al.*, 2012; Gurley *et al.*, 2012;
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12 Chung *et al.*, 2015). This metabolic parameter is potentially a more direct indicator of tissue
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15 metabolic activities, which integrates many factors and provides further insight about tissue
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18 pathophysiology. It is expected that with further technology development and more clinical
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21 applications, DCS/DCT and hybrid DCS/NIRS technologies will be eventually utilized as
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24 routine diagnostic and intervention monitoring tools for clinical investigations of various
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27 diseases.

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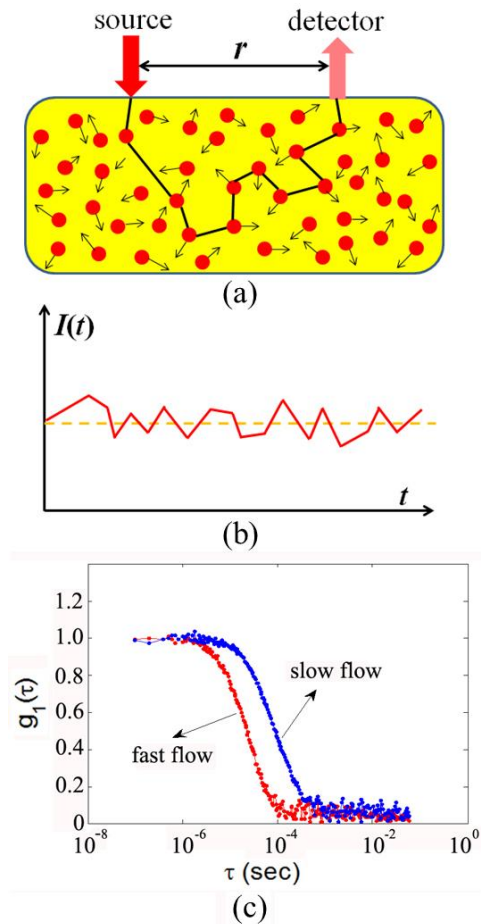
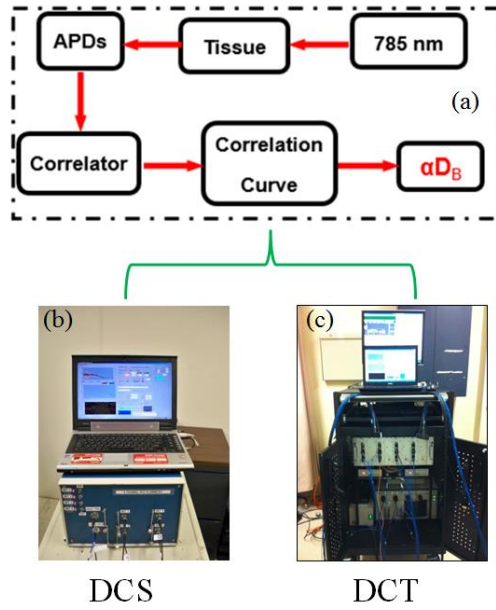


Fig. 1: The principle of diffuse correlation spectroscopy (DCS) for blood flow measurements in deep/thick tissues. (a) Source and detector fibers are placed on the tissue surface within a distance of r for light delivery and collection; (b) Light intensity $I(t)$, detected by the APD, fluctuates with time due to the motions of RBCs; (c) Blood flow change can be quantified from the shift of temporal electric field autocorrelation function curves.



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Fig. 2: The schematic diagram (a) and instrument photos of typical DCS (b) and DCT (c) systems.

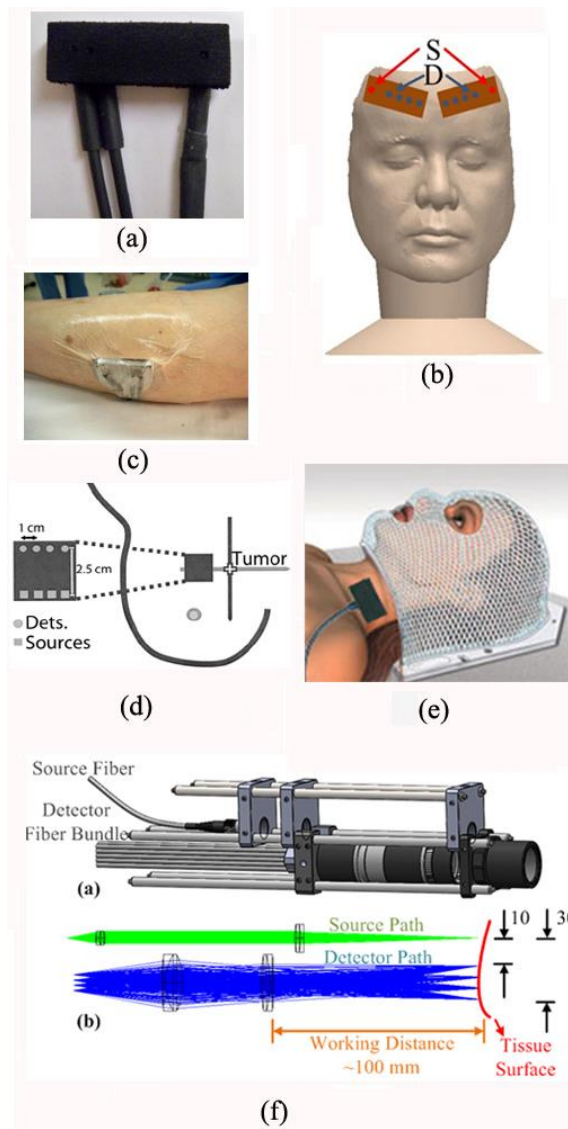


Fig. 3: A variety of optical probes for DCS/DCT measurements. (a) A typical DCS probe consisting of source and detector fibers; (b) Two DCS probes taped on both sides of a frontal head for cerebral blood flow measurements; (c) A DCS probe taped on top of the calf muscle; (d) A hand-held DCS probe scanning over a breast tumor; (e) A DCS probe placed on top of the head/neck tumor; (f) A noncontact DCS probe with lens system focusing the source and detectors on the surface of tissue. Some of the figures are reproduced from the subfigures in the references (Huang *et al.*, 2015c; Durduran *et al.*, 2005)

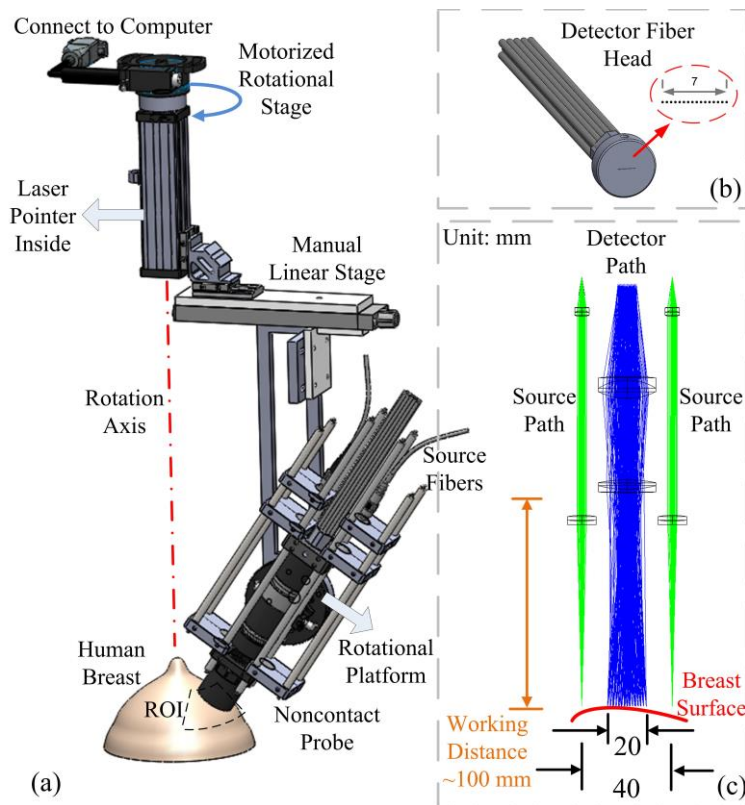


Fig. 4: A noncontact DCT probe with rotational scanning system. (a) Noncontact DCT probe head was scanned over a region of interest (ROI) on the breast by a motorized rotational stage, (b) A linear array of single-mode detector fibers, (c) Source and detector fibers were projected on the breast surface using achromatic lenses; two source paths were attached to the sides of the detector path. This figure is reproduced from Fig. 1 in the reference (Huang *et al.*, 2015b).

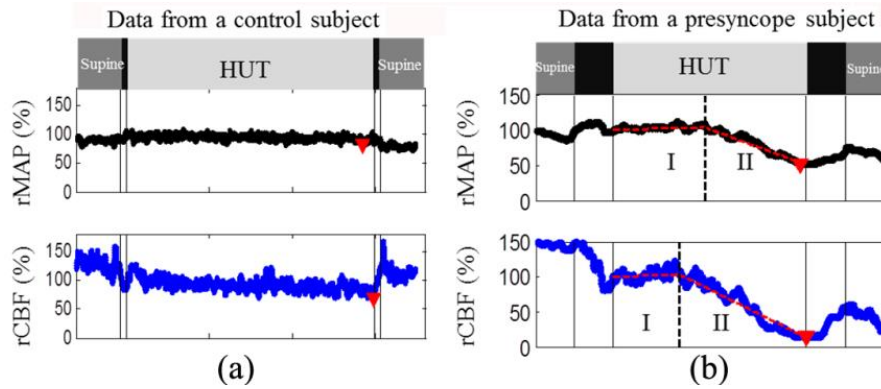


Fig. 5. Relative changes in mean artery blood pressure (rMAP) and cerebral blood flow (rCBF) from a control subject **(a)** and a presyncope subject **(b)** during Head-Up-Tilting (HUT). The solid vertical lines indicate the beginning or ending of tilting up and tilting down. The triangles indicate the minima of rMAP and rCBF during HUT. The two connected dashed lines on top of the raw data **(b)** demonstrate the two-line fitting results. The dashed vertical lines indicate the break-points separating Stage I and Stage II. This figure is reproduced from the subfigures of Fig. 1 in the reference (Cheng *et al.*, 2014).

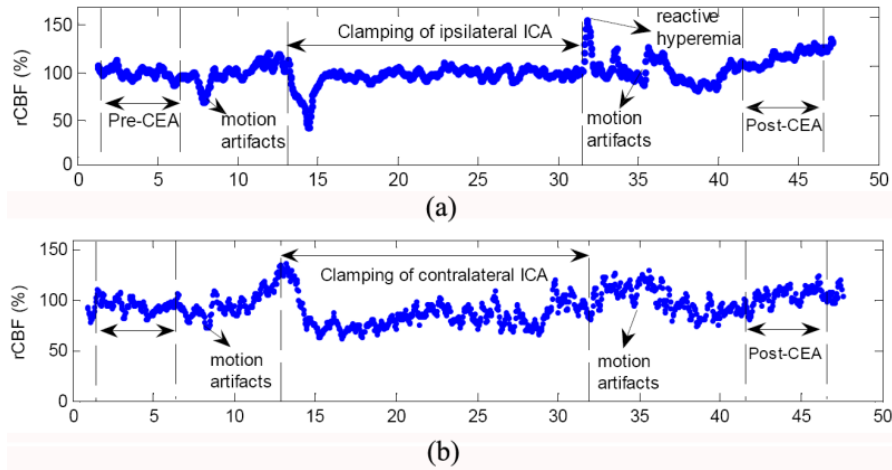


Fig. 6. Typical relative changes in cerebral blood flow (rCBF) **(a)** at the surgical side (ipsilateral) and **(b)** at the nonsurgical side (contralateral). This figure is reproduced from subfigures of Fig. 2 in the reference (Shang *et al.*, 2011b).

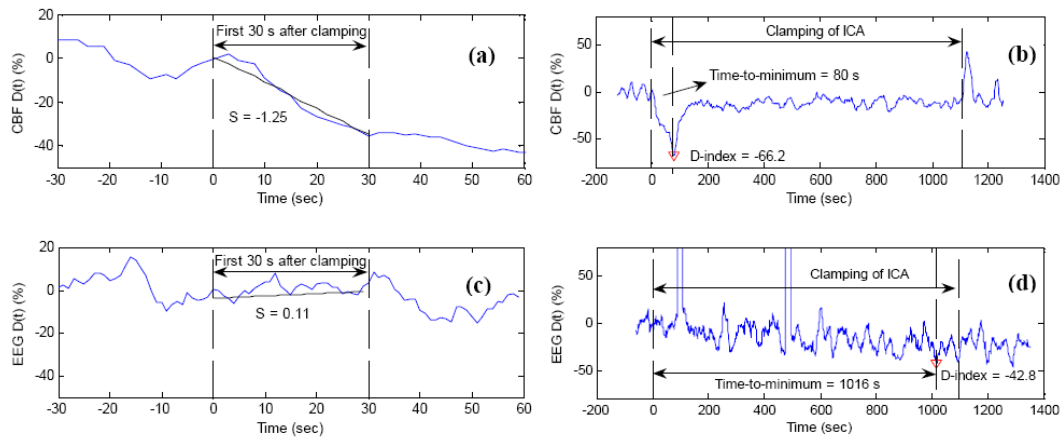


Fig. 7: Comparison of DCS and EEG measurement results in a typical patient during carotid endarterectomy (CEA) when the internal carotid artery (ICA) was temporarily clamped. The resulted ischemic status by the ICA clamping was characterized by the slope (S), D -index, and time-to-minimum for both CBF (top panel) and EEG (bottom panel) at the surgical side. Here the slope (S) is the decreasing rate of variable $D(t)$ during the first 30-second clamping period, and $D(t)$ represents the percentage change of variable (CBF or EEG spectral power) at time t , when compared to pre-clamping baseline. The minimum of $D(t)$ during the entire period of clamping is defined as D -index, representing the most severe cerebral hypoperfusion during ICA clamping. The period from the beginning of clamping to the time of $D(t)$ attaining its minimum (D -index) is defined as time-to-minimum. The large negative CBF slope ($S = -1.25$) during the first 30-second clamping period indicates the rapid CBF decrease due to ICA clamping (a). The time duration of CBF decrease and maximal CBF decrease during the entire clamping period were characterized by time-to-minimum (80 s) and D -index (-66.2%), respectively (b). By contrast, the EEG power changes were small and slow ($S = 0.11$) (c), and reached its minimum (D -index = -42.8%) in a long period of time (time-to-minimum = 1016 s) (d). This figure is reproduced from Fig. 4 in the reference (Shang *et al.*, 2011b).

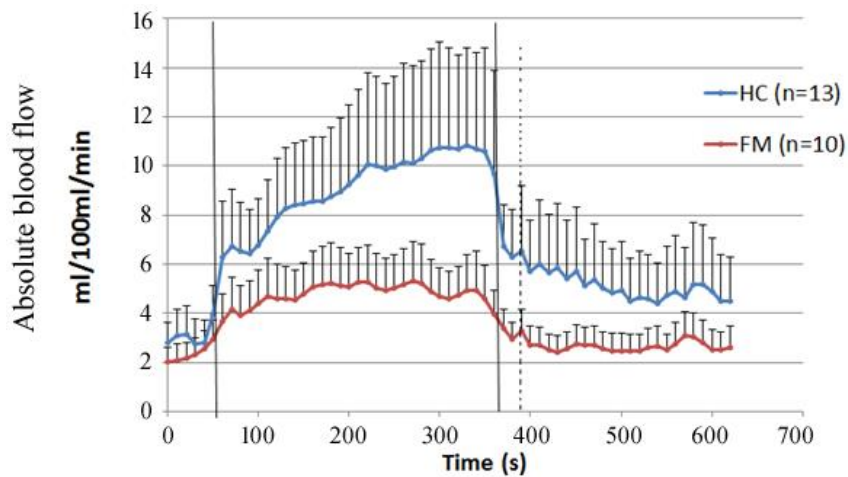


Fig. 8: Average absolute blood flow responses to handgrip exercise in 10 subjects with fibromyalgia (FM) and 13 age-matched healthy controls (HC). The vertical lines indicate the beginning and end of exercise. Average pre-exercise blood flow value, determined as the average of data points 30 s before the onset of exercise, was 3.19 ± 1.03 ml/100 ml/min for the healthy subjects and 2.63 ± 0.71 ml/100 ml/min for the FM subjects ($p = 0.444$). Plateau blood flow value, determined as the average of points 30 s before the end of exercise was 10.71 ± 3.91 ml/100 ml/min for healthy subjects and 4.83 ± 1.42 ml/100 ml/min for FM subjects ($p = 0.018$).

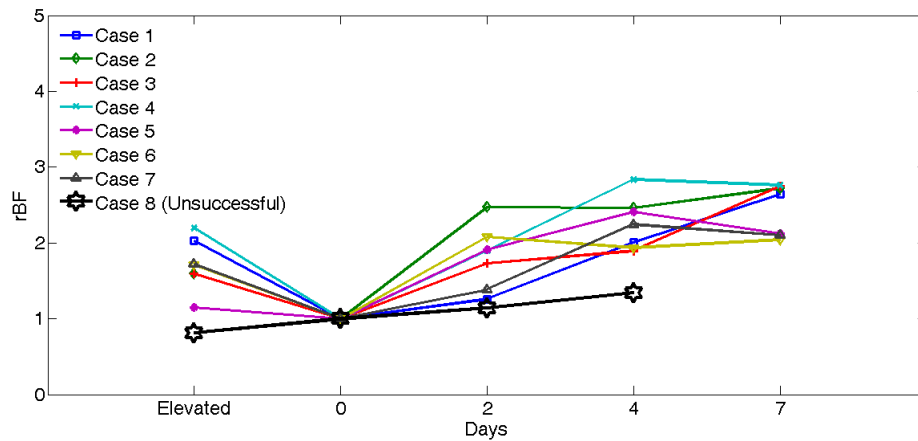


Fig. 9: Comparison of rBF changes between seven successful and one unsuccessful flaps. The rBF values over the seven successful cases (Cases 1 to 8) during operation and on postoperative days 2, 4, 7 are presented individually. rBF from the unsuccessful flap case (Case 8) was measured at the time of flap elevation and on postoperative days 2 and 4. The data from the unsuccessful case (Case 8) at Day 7 are not available due to the subsequent re-flap before that day. This figure is reproduced from Fig. 6 in the reference (Huang *et al.*, 2015c).

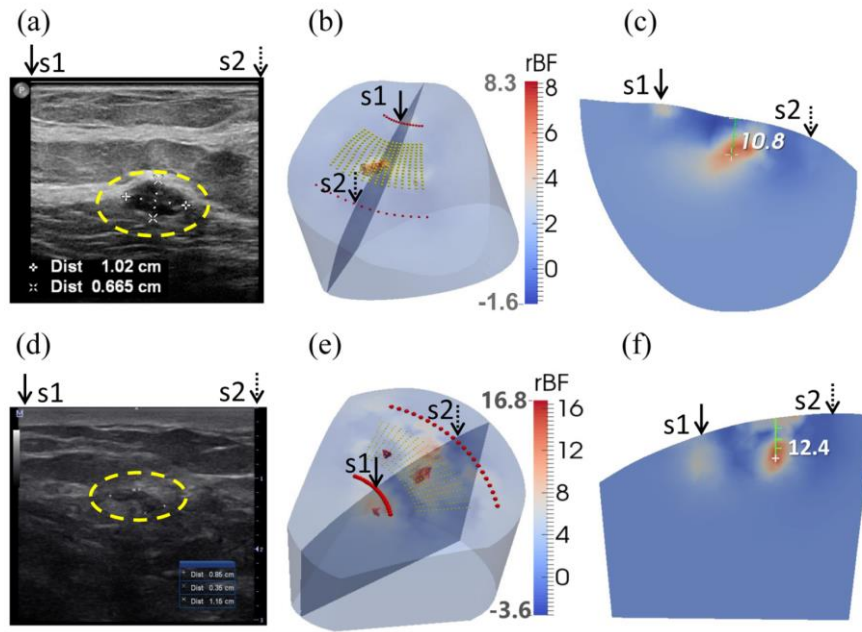


Fig. 10: Clinical examples of two low-grade carcinomas in situ. (a) Ultrasound image of breast with a tumor (inside the yellow dashed circle) in the first patient. The tumor mass center is located at 19.2 mm beneath the skin surface. (d) Ultrasound image of breast with a tumor (inside the yellow dashed circle) in the second patient. The tumor mass center is located at 13.3 mm beneath the skin surface. (b) and (e) show the reconstructed 3-D tumor blood flow contrasts imaged by the ncDCT in the first and second patients, respectively. For the comparison of ultrasound and ncDCT results, an ultrasound imaging plane along the transducer line and across the overlapped two specific light sources (S1 and S2) of ncDCT (see Fig. 4) is presented in the 3-D reconstructed image. The backgrounds are presented with 30% transparency of the original color clarity. (c) and (f) show the cross-section views through the reconstructed tumor centers, which can be directly compared to the 2-D ultrasound tumor images [(a) and (d)], respectively. This figure is reproduced from Fig. 8 in the reference (He *et al.*, 2015).

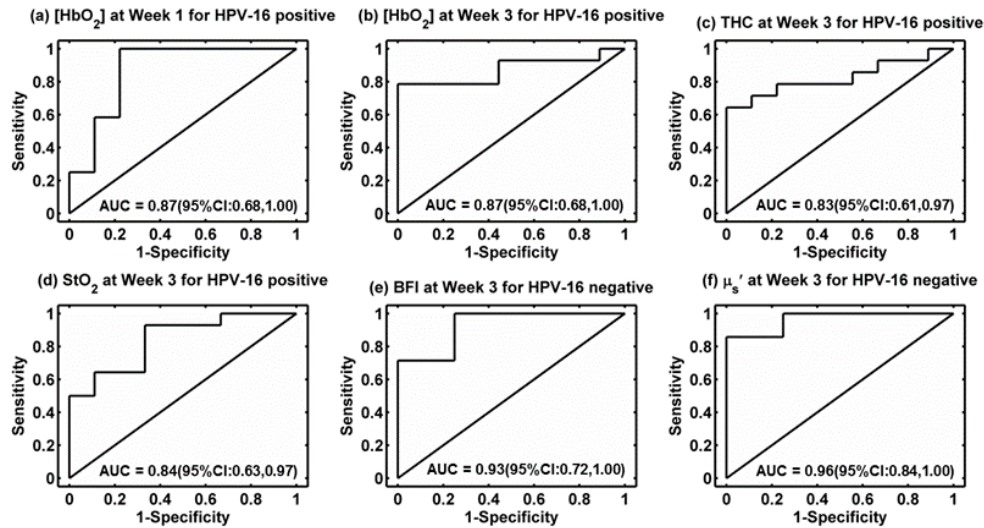


Fig. 11: Receiver operating characteristic curves (ROC) and the associated areas under the curve (AUC) for discriminating tumors with incomplete response (IR) or complete response (CR). (a, b) [HbO₂] at Week 1 and Week 3 in HPV-16 positive patients, (c) THC at Week 3 in HPV-16 positive patients, (d) StO₂ at Week 3 in HPV-16 positive patients, (e) BFI in HPV-16 negative patients at Week 3, and (f) μ_s' in HPV-16 negative patients at Week 3. This figure is reproduced from Fig. 5 in the reference (Dong *et al.*, 2016).