BOLD-based Techniques for Quantifying Brain Hemodynamic and Metabolic Properties – Theoretical Models and Experimental Approaches

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Abstract
Quantitative evaluation of brain hemodynamics and metabolism, particularly the relationship between brain function and oxygen utilization, is important for understanding normal human brain operation as well as pathophysiology of neurological disorders. It can also be of great importance for evaluation of hypoxia within tumors of the brain and other organs. A fundamental discovery by Ogawa and co-workers of the BOLD (Blood Oxygenation Level Dependent) contrast opened a possibility to use this effect to study brain hemodynamic and metabolic properties by means of MRI measurements. Such measurements require developing theoretical models connecting MRI signal to brain structure and functioning and designing experimental techniques allowing MR measurements of salient features of theoretical models. In our review we discuss several such theoretical models and experimental methods for quantification brain hemodynamic and metabolic properties. Our review aims mostly at methods for measuring oxygen extraction fraction, OEF, based on measuring blood oxygenation level. Combining measurement of OEF with measurement of CBF allows evaluation of oxygen consumption, CMRO₂. We first consider in detail magnetic properties of blood – magnetic susceptibility, MR relaxation and theoretical models of intravascular contribution to MR signal under different experimental conditions. Then, we describe a “through-space” effect – the influence of inhomogeneous magnetic fields, created in the extravascular space by intravascular deoxygenated blood, on the MR signal formation. Further we describe several experimental techniques taking advantage of these theoretical models. Some of these techniques - MR susceptometry, and T₂-based quantification of oxygen OEF – utilize intravascular MR signal. Another technique – qBOLD – evaluates OEF by making use of through-space effects. In this review we targeted both scientists just entering the MR field and more experienced MR researchers interested in applying advanced BOLD-based techniques to study brain in health and disease.

Keywords
MRI; brain; BOLD; qBOLD; OEF; CMRO₂; blood; susceptibility

1. Introduction
Quantitative evaluation of brain hemodynamics and metabolism, particularly the relationship between brain function and oxygen utilization, is important for understanding
normal human brain operation (1) as well as understanding the pathophysiology of disorders such as stroke (2), Alzheimer’s disease (3–5), Huntington’s disease (6), Parkinson’s disease (7,8) and other neurological disorders (9–12). It can also be of great importance for evaluation of hypoxia within tumors of the brain and other organs (13,14). Currently, clinically accepted methods for measuring oxygen metabolism rely on PET techniques (e.g., (1)). However, the presence of ionizing radiation, low spatial resolution, and rather restricted availability inhibit broad application of PET-based methods in human research and clinical practice.

A fundamental discovery by Ogawa and co-workers (15) of the BOLD (Blood Oxygenation Level Dependent) contrast opened a possibility to use this effect to study brain hemodynamic properties by means of MRI measurements. The most widely used application of BOLD contrast is based on studying dynamic properties of brain during functional stimulation, i.e. fMRI (functional MRI) (16–19). However, BOLD effect can also be used to studying brain hemodynamic in a resting (baseline) state because, despite heterogeneities in blood flow and metabolism amongst brain regions, a long-standing observation is that in healthy human subjects oxygen extraction fraction (OEF), on average, is remarkably constant across the brain in the resting state (20) and can be used as the primary index of brain function (20,21). This baseline state is usually achieved with the subject resting quietly but awake with eyes closed. Such a baseline state can be described as a default mode of the brain activity. Therefore, it is important to understand the quantitative characteristics describing this default mode. In this regard, it should also be noted that it is the baseline functional activity of the brain that is responsible for the majority of the brain’s enormous energy budget (1).

Regional OEF measurements can assess compensatory responses made by the brain to progressive reductions in cerebral perfusion pressure (CPP) (e.g., (22,23)). When CPP is normal, CBF is closely matched to the resting metabolic rate of the tissue. As a consequence of this resting balance between flow and metabolism, the OEF shows little regional variation. Moderate reductions in CPP have little effect on CBF, as vasodilation of arterioles reduces cerebrovascular resistance, thus maintaining a constant CBF. With more severe reductions in CPP, the capacity for compensatory vasodilation is exceeded and CBF begins to decline. Cerebral oxygen metabolism and brain function at this stage are maintained by progressive increase in OEF (2,24). Importantly, the OEF has been shown to be an accurate predictor of subsequent stroke occurrence in patients with cerebrovascular disease (25,26) supporting the idea that OEF is a critical hemodynamic parameter that provides an important and predictive index of brain functional status.

Combining OEF measurements with measurements of cerebral blood flow (CBF) enables the evaluation of cerebral metabolic rate of oxygen consumption (CMRO2) by means of Fick’s principle. Different methods of CBF quantification will be discussed in other papers of this special issue of NMR in Biomedicine. In this article we focus our attention on MRI-based methods for measuring OEF. Our review is targeted to both scientists just entering the MR field and more experienced MR researchers interested in applying advanced BOLD-based techniques to study brain in health and disease.

2. Fick’s principle

According to Fick’s principle (27), CMRO2 can be calculated by using the following relationship:

\[
CMRO_2 = CBF \cdot C_{\text{blood}} \cdot (Y_a - Y_v) \quad [1]
\]
where $Y_a$ and $Y_v$ are oxygenation levels of arterial and venous blood (i.e., the fraction of hemoglobin in the form of oxyHb; $Y = 1$ corresponds to fully oxygenated blood and $Y = 0$ corresponds to fully deoxygenated blood); and $C_{blood}$ is blood oxygen carrying capacity. Usually, $\text{CMRO}_2$ is measured in $\mu$mol O$_2$/g tissue/min, $\text{CBF}$ in ml blood/g tissue/min, and $C_{blood}$ in $\mu$mol O$_2$/ml blood.

Strictly speaking, Eq. [1] corresponds to oxygen combined with hemoglobin, ignoring oxygen dissolved in blood plasma. The amount of dissolved oxygen is determined by partial pressure of oxygen PO$_2$ in blood (28):

$$[O_2]_{dissolve} = 0.023 \cdot \frac{\text{PO}_2 \text{[mmHg]}}{760} \text{[cc/ml]}, \quad [2]$$

and usually does not exceed 1.5% of total oxygen in blood though it can be higher in abnormal conditions, e.g. hyperoxia (29).

The hemoglobin-based blood oxygen carrying capacity is proportional to blood hematocrit level $Hct$:

$$C_{\text{blood}} = C_{RBC} \cdot Hct \quad [3]$$

where the proportionality coefficient $C_{RBC}$ is oxygen carrying capacity of red blood cells (RBC). Hence, Eq. [1] can be written as follows:

$$\text{CMRO}_2 = C_{RBC} \cdot \text{CBF} \cdot Hct \cdot Y_a \cdot \text{OEF} \quad [4]$$

where oxygen extraction fraction is defined as

$$\text{OEF} = \frac{Y_a - Y_v}{Y_a}, \quad [5]$$

Oxygen carrying capacity $C_{RBC}$ can be calculated as

$$C_{RBC} = n_{Hb} \cdot M_{Hb} \cdot C_{Hb}, \quad [6]$$

where $n_{Hb} = 5.5 \cdot 10^{-6}$ mol/ml is the concentration of hemoglobin in RBC (30), $M_{Hb} = 64.450$ g/mol is the hemoglobin molecular weight (31), and $C_{Hb} = 1.39$ ml/g is amount of oxygen that can be combined with 1g of pure hemoglobin (28). Thus, $C_{RBC} = 0.493$ (ml O$_2$/ml RBC).

According to Eqs. [1], [4], if $\text{CBF}$ and $\text{OEF}$ are known, oxygen consumption $\text{CMRO}_2$ can be calculated. Several reviews in the current issue of this journal are devoted to methods of measuring $\text{CBF}$. Our review aims mostly at methods for measuring $\text{OEF}$. Most of the current MR methods of measuring $\text{OEF}$ are based, in fact, on measuring blood oxygenation level. Some of these methods and theoretical backgrounds behind these methods are discussed in the present review.

### 3. Blood Oxygenation and MR Signal

One of the important parameters characterizing magnetic properties of all tissues is their magnetic susceptibility $\chi$ – a proportionality coefficient between tissue magnetization, $M$, induced by an external magnetic field $B_0$, and the magnetic field strength (note that we are using Gaussian-CGS units throughout):

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Most components of biological tissues, such as water, proteins, lipids, are diamagnetic (their magnetic susceptibility $\chi$ is negative). The diamagnetism is a common property of all atoms and molecules; it is due to the effect of changing microscopic atomic currents of orbiting electrons, sometimes called Ampèrian currents, in the presence of magnetic field $B_0$. If atoms or molecules contain uncompensated electronic spin moments (that is always accompanied by magnetic moment), they also exhibit additional magnetic susceptibility which is positive and is called paramagnetic susceptibility. The paramagnetic effect is due to “orientational nature” of magnetic field that tends to align electron spin magnetic moments against “de-orientational nature” of thermal motion. Paramagnetic susceptibility is described by the Curie law:

$$\chi = N \frac{\mu^2}{3k_B T} \quad [8]$$

where $N$ is a molar content of paramagnetic molecules per unit volume, $\mu$ - their molecular spin-based magnetic moment, $k_B$ – Boltzmann constant and $T$ – absolute temperature.

Biologically relevant examples of paramagnetic molecules include non-heme iron and heme iron in deoxyhemoglobin (see detail discussion in (32–34)), and dissolved oxygen molecule $O_2$. Importantly, heme iron is paramagnetic because of its Fe$^{2+}$ state. When heme iron combines with oxygen, it changes its electronic configuration and the total spin magnetic moment of heme complex becomes zero (35,36). When heme iron releases oxygen, it returns to a paramagnetic state. Hence, magnetic state of heme iron can be used as a biomarker of blood oxygenation level. When blood passes through the capillary bed and releases oxygen, the state of heme iron changes from zero-spin at the arterial side to a high spin at the venous side.

Due to these reversible changes of heme complexes in deoxygenated red blood cells (RBC) (35,36), the blood vessel network in biological tissues modifies MR signal. Importantly, this modification depends on blood oxygenation level. This phenomenon forms the basis of the BOLD (blood oxygen level dependent) contrast in MRI. Two effects should be separated – intravascular and extravascular. The intravascular effect is due to the inhomogeneous magnetic fields created by red blood cells in blood (37). The extravascular (through-space) effect is mainly due to inhomogeneous magnetic fields created by blood vessels in the surrounding tissue (15,38). Because these magnetic field inhomogeneities are tissue specific, they can provide important information on tissue hemodynamic properties. To quantify these changes, MRI techniques based on Free Induction Decay (FID) and different versions of spin echo (SE), including multi-SE sequences, such as CPMG and its variations (e.g. (39)), are usually used. In these experiments, the MR signal $S(t)$ can be described in terms of a signal attenuation function $\Gamma(t)$:

$$S(t)=S(0)\exp(-\Gamma(t)) \quad [9]$$

where $t$ is either time after initial excitation RF pulse in FID and CPMG experiments or spin echo time in SE experiment. It is usually assumed that

$$\Gamma_{SE}(t)=R_2 t, \quad \Gamma_{FID}(t)=R_2^e t - i\phi(t), \quad R_2^e = R_2 + R_2^f \quad [10]$$
where $R_2$ and $R_2'$ are irreversible and reversible (with respect to 180° refocusing RF pulse) MR signal relaxation rate constants and $\phi(t)$ is the MR signal phase contributed by tissue specific sources and also by unwanted macroscopic field inhomogeneities. Note that in multi-SE experiments, $R_2$ depends on inter-echo spacing, $R_2 = R_2(\tau_{CPMG})$, e.g. (40,41).

In fact, Eqs. [10] are rarely accurate in biological tissues because the signal attenuation function $\Gamma(t)$ is a non-linear function of its argument and depends on the type of MR experiment. An obvious source of non-linearity of the signal attenuation function relates to a multi-compartment tissue structure. Another source of the deviation from simple exponential decay (even for a single compartment case) relates to the presence of inhomogeneous magnetic fields (42–58). This type of non-linearity will also be discussed in detail in this review paper.

4. Intravascular MR Signal

4.1 Blood magnetic susceptibility

Studies of the magnetic properties of human blood have been undertaken in numerous publications (35–37,57,59–78). A detailed theoretical consideration of blood magnetic susceptibility and the detailed experimental studies employing in vitro samples that were well representative of human blood in situ were provided in (71). Here we briefly outline these results. The magnetic susceptibility of whole blood is determined by a weighted sum of magnetic susceptibilities of RBC and plasma:

$$\chi_{\text{blood}} = Hct \cdot \chi_{\text{RBC}} + (1-Hct) \cdot \chi_{\text{plasma}} \quad [11]$$

where $Hct$ is the blood hematocrit level.

The magnetic susceptibility of RBC can be expressed as (71):

$$\chi_{\text{RBC}} = (1-n_{\text{Hb}} \cdot \nu_{\text{M,Hb}}) \cdot \chi_{\text{H2O}} + n_{\text{Hb}} \cdot [M_{\text{Hb}} \cdot \chi_{\text{g,prot}} + (1-Y) \cdot \chi_{\text{M,dHb}}] \quad [12]$$

where $Y$ is blood oxygenation level, $\chi_{\text{g,prot}} = -0.587 \cdot 10^{-6}$ mL/g is diamagnetic contribution of the Hb protein (79), $\nu_{\text{M,Hb}} = 48,277$ mL/mol is the molar volume of Hb in solution (79), and the rest of the parameters are defined in Section 2. The paramagnetic contribution to the molar susceptibility of deoxyHb, $\chi_{\text{M,dHb}}$, can be calculated via Eq. [8]. Taking into account that there are 4 iron atoms per heme complex, and that the average magnetic moment of Fe$^{2+}$ in hemoglobin, $\mu = 5.46 \mu_B$/Heme ($\mu_B$ is the Bohr magneton) (36), it can be calculated that at 37°C, $\chi_{\text{M,dHb}} = 48.082 \cdot 10^{-6}$ mL/mol. Thus, the magnetic susceptibility of RBCs can be described as (71):

$$\chi_{\text{RBC}} = -0.736 + 0.27 \cdot (1-Y) \quad \text{ppm} \quad [13]$$

Susceptibility of plasma, $\chi_{\text{plasma}}$, can be estimated based on the contributions of water and the other major plasma constituents (proteins), which accounts for approximately 93% and 7% of the plasma weight, respectively (80). As a result, $\chi_{\text{plasma}}$ is slightly different from susceptibility of water, $\chi_{\text{H2O}} = -0.719$ ppm (71),

$$\chi_{\text{plasma}} = -0.722 \quad \text{ppm} \quad [14]$$

Thus, the total susceptibility of whole blood is

$\chi_{\text{blood}}$.
Important prediction of this equation is that the susceptibility difference between completely deoxygenated \((Y = 0)\) and completely oxygenated \((Y = 1)\) RBC is equal to

\[
\Delta \chi_0 = 0.27 \text{ ppm} \quad [16]
\]

Equation [16] was confirmed by two independent MR and SQUID magnetometer measurements recently by detailed magnetic susceptometry measurements [77]. A rather close value was also reported in [81]. Note that the value \(\Delta \chi_0 = 0.27 \text{ ppm}\) in Eq. [16] is substantially (50\%) larger than that frequently used \((\Delta \chi_0 = 0.18 \text{ ppm})\) - for historical reasons - in the MRI literature.

Magnetic susceptibility of oxygen dissolved in blood plasma, \(\chi_{[O_2]}\), can be readily calculated based on the Curie law, Eq. [8]:

\[
\chi_{[O_2]} = \frac{N_a \cdot [O_2]_{\text{dissolve}}}{V_0} \cdot \frac{\mu_{\text{eff}}^2}{3 k_B T} \quad [17]
\]

where \(N_a\) is Avogadro’s number, \(V_0\) is gas molar volume at normal conditions, and \(\mu_{\text{eff}} = 2.83 \cdot \mu_B\) per 1 molecule of oxygen. Using Eq. [2] for dissolved oxygen concentration in blood, we get

\[
\chi_{[O_2]} = 0.003 \cdot \frac{\text{PO}_2[\text{mmHg}]}{760} \quad [\text{ppm}] \quad [18]
\]

This quantity is rather small as compared to the paramagnetic component of the magnetic susceptibility of deoxyHb. For example, at \(\text{PO}_2 = 100 \text{ mmHg}\), the contribution of oxygen to blood magnetic susceptibility is \(\chi_{[O_2]} = 0.00039 \text{ ppm}\), while at \(Hct = 0.4\) and blood oxygenation level \(Y=0.6\), the contribution of the paramagnetic component of the magnetic susceptibility of deoxyHb to the blood magnetic susceptibility is \(\Delta \chi_{\text{deoxyHb}} = 0.0432 \text{ ppm}\) - two orders of magnitude higher. In what follows, the contribution of dissolved oxygen to blood magnetic susceptibility will be ignored.

Equation [15] determines the susceptibility of the whole blood. For the purpose of study of \(OEF\) and \(\text{CMRO}_2\) in the brain, one should refer \(\chi_{\text{blood}}\) to the magnetic susceptibility of surrounding tissue:

\[
\Delta \chi = \chi_{\text{blood}} - \chi_{\text{tissue}} \quad [19]
\]

The tissue susceptibility \(\chi_{\text{tissue}}\) is determined by a variety of susceptibility inclusions: first of all, proteins, lipids, and non-heme iron. For example, based on cellular content and related physical properties of the essential inclusions in an “average normal” frontal lobe cortical GM, WM and CSF/ISF, the volume magnetic susceptibilities of the inclusions (relative to water) were estimated in [82] and provided here in Table 1. Thus, the magnetic susceptibility difference between blood and surrounding tissue can be calculated as follows:

\[
\Delta \chi = Hct \cdot 0.27 \cdot (1-Y) - 0.736 + (1-Hct)(-0.722) \quad [\text{ppm}] \quad [20]
\]

where \(\Delta \chi_{\text{tissue}} = \chi_{\text{tissue}} - \chi_{H_2O}\) in ppm (see Table 1). This means that the susceptibility difference between blood and surrounding tissue is not zero even for fully oxygenated
blood. Hence, the blood oxygenation level can be determined from the experimentally measured \( \Delta \chi \) as follows:

\[
Y = 0.948 - \frac{1}{Hct} (\Delta \chi + 0.003 + \Delta \chi_{\text{tissue}}) \tag{21}
\]

For arterial and venous blood, \( Y = Y_a \) and \( Y = Y_v \), respectively. OEF can be calculated from these parameters using Eq. [5]. Note that a usually used relationship

\[
\Delta \chi = \Delta \chi_0 \cdot Hct \cdot (1 - Y) \tag{22}
\]

where \( \Delta \chi_0 \) is given in Eq. [16], is an approximation of Eq. [20] that neglects differences between magnetic susceptibilities of tissue, water and plasma.

4.2 MR signal from blood vessel network

The susceptibility difference between blood \( (\chi_{\text{blood}}) \) and tissue \( (\chi_{\text{tissue}}) \) results in difference of the Larmor frequency of protons in water molecules within blood vessels and those in surrounding tissue, \( \Delta \omega \). Approximating a blood vessel by a long circular cylinder, \( \Delta \omega \) can be written as

\[
\Delta \omega = 2\pi \gamma B_0 \Delta \chi \left( \cos^2 \theta - \frac{1}{3} \right) \tag{23}
\]

where \( \theta \) is the angle between the vessel direction and the direction of magnetic field \( B_0 \). The term \( 1/3 \) in Eq. [23] reflects a contribution of the so called Lorentzian sphere (e.g., (84)). One should however note that the Lorentzian sphere approximation might not be adequate for describing influence of tissue microstructure on MR frequency. To accurately predict the Larmor resonance frequency difference between intravascular and extravascular water protons in tissues with long cells (i.e. axons), the generalized Lorentzian theory (82) should be used. However, since susceptibilities of GM and WM tissues (see Table 1) are much smaller than that of venous blood, the influence of this effect on Eq. [23] might be small.

The normalized MR free induction decay (FID) signal from a single vessel with the frequency shift \( \Delta \omega \) is described as a simple exponential oscillation, \( s_i(t) = \exp(-i \cdot \Delta \omega \cdot t) \), (transverse relaxation of blood signal will be introduced later). However, each imaging voxel contains blood vessels with different orientations, and the total MR signal form a voxel is a sum of the signals from individual vessels:

\[
s_i(t) = \int_0^{\pi} d\theta P(\theta) \cdot \exp(-i \Delta \omega \theta t) \tag{24}
\]

where \( P(\theta) \) is the distribution function of vessel orientations. This distribution is usually not isotropic (85), however for low resolution experiments when an imaging voxel contains a multitude of blood vessels, a uniform distribution, \( P(\theta) = \sin \theta / 2 \), can be a reasonable approximation. In this case the signal \( s_i(\theta) \) can be readily calculated (52) (see also (86)):

\[
s_i(t) = \exp(i \cdot \frac{\Delta \omega_0}{3} t) \cdot \left[ C \eta - i S \eta \right] \eta, \quad \eta = \left( \frac{2 \Delta \omega_0 |t|}{\pi} \right)^{1/2} \tag{25}
\]
where $\Delta \omega_0 = 2\pi \gamma B_0 \Delta \chi$, and $C(\cdot)$ and $S(\cdot)$ are the Fresnel functions (note that we use here the definition of the Fresnel functions as in (87) which is different by the factor $\pi/2$ in the argument from that used in (88) and adopted in (86)).

Equation [25] describes contribution to MR signal of blood with a given oxygenation level. However, blood oxygenation is different between arteries, capillaries and veins. To address this issue, Eq. [25] can be modified by introducing a distribution function of characteristic frequencies, $\Delta \omega_0$. This is not difficult to incorporate into theoretical model but accounting for this effect in experiments might be a challenge.

### 4.3 Blood transverse relaxation – $T_2$ and $T_2^*$

The signal decay described by Eq. [25] does not take into account intrinsic $T_2$ (or $T_2^*$) MR relaxation processes in blood, usually described phenomenologically by $R_2$ (or $R_2^*$) relaxation rate constants for SE and FID signals correspondingly. The presence of inhomogeneous magnetic fields created in the blood by deoxyHb, increases $R_2$ and $R_2^*$ relaxation rate constants (37). Generally, these transverse relaxation rate constants of blood can be presented as

$$R_2 = R_{20} + \Delta R_2, \quad R_2^* = R_{20}^* + \Delta R_2^* \quad [26]$$

where the first terms correspond to the case when magnetic susceptibilities of RBCs and plasma are identical. According to Eqs. [13]–[14], it takes place at $1 - Y = 0.05$ (71) (see Eq. [27] below). In this case there is no frequency shift between RBC and plasma, and one would expect that $R_2^*$ should be the same as $R_{20}$. This was indeed observed at 1.5T measurements in (71,74) but not at 4.7T (74). An explanation of this effect is still pending.

Experimental values for $R_{20}$ and $R_{20}^*$ found for human and bovine blood at 1.5T (e.g., (71,74,81,89)), vary in the interval $4$–$7$ s$^{-1}$. The second terms in Eqs. [26] describe the contribution of water diffusion in the inhomogeneous magnetic field induced by the magnetic susceptibility difference between RBC and plasma,

$$\Delta \chi = \chi_{\text{RBC}} - \chi_{\text{plasma}} = \Delta \chi_0 \cdot (1 - Y - 0.05) \quad [27]$$

Description of MR signal in terms of $R_2$ and/or $R_2^*$ relaxation rate constants assumes that the signal decay can be described by a simple exponential. In fact, experimental data in (71) demonstrated a substantial deviation from such a behavior, hence MR signal should be described in general terms of a non-linear signal attenuation function $\Gamma(t)$ in Eq. [9].

Specifically, for a realistic model of blood this function was calculated in (72) using previously developed Gaussian-phase-approximation-based approach (53).

It was demonstrated that for an arbitrary shape of susceptibility inclusions (in our case, RBC), at short time, $t \ll t_D$ ($t_D$ is a characteristic diffusion time, see below), $\Gamma_{\text{FID}}(t) \sim t^2$ for FID signal and $\Gamma_{\text{SE}}(t) \sim 1/t^{5/2}$ for single spin-echo (SE) pulse sequence. The asymptotic behavior of $\Gamma_{\text{FID}}(t)$ and $\Gamma_{\text{SE}}(t)$ at long time, $t \gg t_D$ is as follows

$$\Gamma_{\text{FID}}(t) = \Delta R_2^* \cdot t \cdot 1 - a_{\text{FID}} \cdot (t/t_D)^{-1/2}$$
$$\Gamma_{\text{SE}}(t) = \Delta R_2 \cdot t \cdot 1 - a_{\text{SE}} \cdot (t/t_D)^{-1/2} \quad [28]$$

where
\[ D_{\text{eff}} = D_{\text{RBC}} \cdot \text{Hct} + D_{\text{plasma}} \cdot (1 - \text{Hct}) \]

where \( D_{\text{RBC}} \) and \( D_{\text{plasma}} \) are water diffusion coefficient in RBC’s cytoplasm and blood plasma, respectively; \( t_p = \frac{4 \pi^3}{3} \cdot \gamma \cdot \Delta \chi_p \cdot B_0^{-1} \) is characteristic dephasing time; the double integration in [29] is over the volume of a single RBC \( v \). The coefficients \( a_{\text{FID}} \) and \( a_{\text{SE}} \) in Eq. [28] are RBC shape-specific numerical coefficients. For a spherical shape they are 0.89 and 1.6 for FID and SE experiments, respectively.

Figure 1 illustrates a typical time dependence of the signal attenuation functions \( \Gamma_{\text{FID}}(t) \) and \( \Gamma_{\text{SE}}(t) \) normalized to the characteristic value \( \Gamma_0 \).

\[ \Gamma_0 = \frac{4 \zeta_0 \omega_0^2}{5} \]  

In (72), the shape of RBC was modeled as a biconcave disk which can be obtained by rotating the curve given in Eq. [31] about the ordinate \( z \):

\[ z(x) = l_{\text{bnc}} \cdot (1 - u^2)^{1/2} \cdot (c_0 + c_1 u^2 + c_2 u^4), \quad u = x/l_{\text{bnc}} \]

The curve \( z(x) \) is depicted in Figure 2. For the normal human RBC, \( c_0 = 0.188, c_1 = 1.085, c_2 = -0.896, l_{\text{RBC}} = 3.825 \mu m, \) and the RBC volume \( v = 97.91 \mu m^3 \) (90). For these specific RBC parameters, a numerical evaluation of the integral divided by \( v \) in Eq. [29] gives \( 2.51 \cdot l_{\text{bnc}}^2 \). Using Eq. [27], the quantity \( \Delta R_2 \) can be written in the form:

\[ \Delta R_2 = \kappa \cdot \text{Hct} \cdot (1 - \text{Hct}) \cdot (1 - Y) - 0.05^2, \quad \kappa = 0.96 \cdot t_p / \tau_0^2 \]

where \( t_p^2 = l_{\text{bnc}}^2 / 6D_c \) and \( \tau_0 = 4 \pi^3 \cdot \gamma \cdot \Delta \chi_0 \cdot B_0^{-1} \) are the characteristic diffusion and dephasing times (the numerical coefficient in Eq. [32] differs from that in (72) due to different definition of the characteristic times \( \tau_0 \) and \( t_D \)).

It should be noted that the Gaussian phase approximation is valid under assumption that the characteristic diffusion time is less than the characteristic dephasing time, \( \lambda = t_D / \tau_0 \ll 1 \). For typical value of the average water diffusion coefficient \( D_{\text{eff}} = 2 \mu m^2/\text{ms} \), we get for \( B_0 = 1.5T, 3T \) and \( 4.7T \): \( \lambda = 0.22, 0.44 \) and 0.69, respectively. Consequently, one can expect that our approach should be accurate for \( B_0 = 1.5T \) and \( 3T \) but might not be accurate for \( B_0 = 4.7T \). Indeed, for \( B_0 = 1.5T \), Eq. [32] gives \( \kappa = 51.8 \text{sec}^{-1} \) which is in a good agreement with value \( \kappa = 55 \text{sec}^{-1} \) obtained experimentally in (71). Detail results of measurements at different field strength can be found in (71,74,81,91) for different types of MR experiments.

A semi-phenomenological theory of FID signal relaxation in blood was proposed in (57). In this approach blood is considered as a heterogeneous media characterized by a correlation length \( l_c \). In case of a small variation of the Larmor frequency (characterized by the dispersion \( \delta \Omega \)), the FID signal is described as follows (57):

\[ s(t) = \frac{\alpha^2}{1 - 2\alpha^2} \cdot \exp \left( - \left( \frac{t}{\tau_0} \right) \cdot \frac{1}{1 - \alpha^2} \right) + \frac{1 - \alpha^2}{1 - 2\alpha^2} \cdot \exp \left( - \left( \frac{t}{\tau_0} \right) \cdot \frac{-\alpha^2}{1 - \alpha^2} \right) \]
where $\alpha = \delta \Omega \cdot \tau_D = (1/2 \pi)^2 / D$ is the characteristic diffusion time, $D$ is the diffusion coefficient (assumed to be the same for RBC and plasma). The expression [33] differs from a standard bi-exponential function broadly used in MR literature by a negative sign in the pre-exponential factor in the first term. As demonstrated in (57), the fit of Eq. [33] to experimental data provides a reasonable agreement between the value $\delta \Omega_{\text{fit}} = 0.86 \text{ms}^{-1}$ and the experimental $\delta \Omega = 0.69 \text{ms}^{-1}$.

The previous discussion addressed blood relaxation properties for FID and SE experiments. For a CPMG sequence, when RF excitation pulse follows by a train of 180° refocusing pulses, blood properties were discussed in (37,81,89,92,93) in the frameworks of different models. In the two-compartment Luz-Meiboom exchange model (41), the quantity $\Delta R_2$ in Eq. [26] for CPMG pulse sequence is equal to (81,89),

$$
\Delta R_2 = Hct \cdot 1 - Hct \cdot \Delta \omega^2 \cdot \tau_{\text{exch}} \cdot \left(1 - \frac{2 \tau_{\text{exch}}}{\tau_{\text{CPMG}}} \tanh \left(\frac{\tau_{\text{CPMG}}}{2 \tau_{\text{exch}}} \right)\right)
$$

where $\tau_{\text{exch}}$ is the phenomenological parameter describing the exchange lifetime, $\tau_{\text{CPMG}}$ is the inter-echo spacing. Equation [34] assumes that the two compartments (RBC and plasma) have a characteristic susceptibility-induced Larmor frequency shift $\Delta \omega$.

The expression for $\Delta R_2$ [34] is valid in the limit of fast exchange, $\Delta \omega \cdot \tau_{\text{exch}} \ll 1$, and at short inter-echo spacing, $\Delta \omega \cdot \tau_{\text{CPMG}} < 2$ (see discussion in (93)). A more complex exchange model valid in the slow-exchange regime ($\Delta \omega \cdot \tau_{\text{exch}} > 1$) and for long echo spacing ($\Delta \omega \cdot \tau_{\text{CPMG}} > 2$) was developed in (94,95). In this model, the quantity $\Delta R_2$ is

$$
\Delta R_2 = 1 + \frac{2 \tau_{\text{exch}}}{\tau_{\text{CPMG}}} \cdot \sinh^{-1} Q
$$

$$
Q = \left[ D_s \sinh^2(\tau_{\text{CPMG}} \cdot \operatorname{Re}(U)/2) + D_\omega \sinh^2(\tau_{\text{CPMG}} \cdot \operatorname{Im}(U)/2) \right]^{1/2}
$$

$$
U = \tau_{\text{exch}}^2 + \Delta \omega^2 + 2i \cdot Hct - (1 - Hct)(\Delta \omega / \tau_{\text{exch}})
$$

Equations [34] and [35] are based on the models that treat diffusion of water molecules in blood as a two-cite exchange process. Another approach to account for diffusion in CPMG experiment was proposed by Jensen and Chandra (50) in the framework of the so called weak field approximation. With an additional assumption that the Larmor frequency correlation function monoexponentially decreases with time, they found the same functional form for $\Delta R_2$ as in Eq. [34]. However, as demonstrated in the same study (50) for the model of magnetized spheres the frequency correlation function decreases with time in an algebraic way (see also (53,54)). In this case, the quantity $\Delta R_2$ for a CPMG pulse sequence was found to be (50):

$$
\Delta R_2 = \frac{2 \zeta \delta \omega^2 t_p}{5} \cdot V \left( \frac{4 \tau_{\text{CPMG}}}{t_p} \right)
$$

$$
V(x) = \int_0^\infty dy \exp(-y^2) \left(1 - \frac{\tanh y}{y} \right)
$$

As we discussed above, a general time dependence of MR signal attenuation function $\Gamma(t)$ in the FID and SE experiments is nonlinear. The same might take place for the CPMG experiment, though the CPMG sequence have a tendency to “restore” the linear time dependence of $\Gamma(t)$ as a function of spin echo number ($t = t_p$) in the CPMG echo train. Such a “restoration” was exemplified in the original paper by Carr & Purcell (40), by considering.
diffusion in a constant magnetic field gradient (see also discussion in of this issue in (96)).
Though we should note that the dependence of $\Delta R_2$ on the echo spacing, $\tau_{CPMG}$, still remains non-linear. A detail discussion of this behavior is beyond the scope of this paper.

5. Theoretical models of Extravascular MR Signal

In the previous section we discussed the intravascular contribution to MR signal attenuation from blood vessel network. The total MR signal includes also the signal from surrounding tissue where inhomogeneous magnetic field is induced due to the susceptibility difference $\Delta \chi$ between blood containing paramagnetic deoxyHb and tissue, Eq. [20]. Thus, spins of water protons in the extravascular space sustain different phase shift $\phi$, leading to additional MR signal decay. The signal at time $t$ after excitation is given by the average of the contributions from all the spins across the volume

$$s(t) = \langle \exp(-i\phi(t)) \rangle, \quad [37]$$

where $\phi(t)$ is the phase accumulated by a single spin by time $t$, and $\langle \ldots \rangle$ means averaging over all possible initial positions and trajectories of a spin. In an inhomogeneous magnetic field, $h = h(r)$, the phase $\phi(t)$ of the spin moving along a given trajectory $r = r(t)$ can be written as

$$\phi(t) = \int_0^t dt' \Delta \omega r(t'), \quad \Delta \omega r(t') = \gamma \cdot h(r(t')), \quad [38]$$

Inhomogeneous magnetic field $h(r)$ is created by a multitude of blood vessels with different positions and orientations. As magnetic field obeys linear superposition, a spin’s phase can be found by integrating the sum of frequency shifts $\Delta \omega_n(t)$ caused by each of $N$ vessels:

$$\phi(t) = \int_0^t dt' \sum_{n=1}^N \Delta \omega_n r(t'), \quad [39]$$

For a single blood vessel of cylindrical shape, the local Larmor frequency shift $\Delta \omega(r)$ in tissue is given by a well known expression (e.g., (83)):

$$\Delta \omega(r) = 2\pi \gamma B_0 \Delta \chi \cdot \frac{R^2}{r^2} \cdot \cos 2\varphi \cdot \sin^2 \theta, \quad r \geq R, \quad [40]$$

Several theoretical approaches have been proposed to calculate MR signal in Eqs. [37]–[40] (e.g., (42,44–50,52–54,97)). A Monte-Carlo-based analysis was also used for this purposes (e.g., (55,67,83,98–100)). Herein we will discuss in detail the static dephasing model (42), the models based on a Gaussian phase approximation (44,53,54) and a phenomenological model (100). Two other approaches – the so called weak and strong field approximations – were proposed in (49,50). The so called “strong collision approximation” was introduced in (47,48). The model proposed in (45) exploits the “linear local field approximation”, in which spins are considered to diffuse in a local (constant for each spin) field gradient. A validity analysis of some of these models was conducted in (100). Some of these results will also be presented here.
5.1 Static dephasing regime

An exact analytical expression for the FID signal was first derived in (42) in the important case of a static dephasing regime when the characteristic dephasing time of MR signal decay in the presence of inhomogeneous magnetic fields is smaller than the characteristic diffusion time (see details below). As shown in (42), in the case of blood vessels with a uniform distribution of their orientations, the extravascular FID signal in the presence of magnetic field inhomogeneities induced by blood vessel network can be presented in the form of Eq. [9] with

\[ \Gamma(t) = \zeta \cdot f(t/t_c) + R_2 t, \]  

where zero time \( t \) corresponds to the position of RF excitation pulse, \( R_2 = 1/T_2 \) describes a \( T_2 \)-relaxation in tissue, \( \zeta \) is the deoxyhemoglobin-containing part of cerebral blood volume (dCBV) - veins and pre-venous part of capillaries. Strictly speaking, the presence of blood vessel network in the tissue also creates frequency shift in Eq. [41] that depends on details of tissue structure at both cellular and global levels (82). The characteristic dephasing time is determined as:

\[ t_c = \left( \frac{4\pi \gamma B_0 \Delta \chi}{3} \right)^{-1} \]  

The function \( f(t/t_c) \) in Eq. [41] is given by (42)

\[ f(t/t_c) = \int_0^1 du \frac{(2+u) \sqrt{1-u}}{3u^2} \left[ 1 - J_0 \left( \frac{3u}{2} \cdot \frac{t}{t_c} \right) \right] \]  

where \( J_0 \) is a Bessel function. The function \( f(t/t_c) \) can also be represented using a generalized hypergeometric function \( _1F_2 \) (52):

\[ f(t/t_c) = _1F_2 \left( \left[ -\frac{1}{2} \right], \left[ \begin{array}{c} 3 \\ -4 \end{array} \right], -\frac{9}{16} \left( \frac{t}{t_c} \right)^2 \right) - 1 \]

Comments about the parameter \( \zeta = dCBV \) are in order as it is sometimes confused with CBV - a parameter defined by other techniques, such as PET, which measure the total blood volume. Only blood that has magnetic susceptibility different from the magnetic susceptibility of brain parenchyma contributes to field inhomogeneities that lead to tissue water MR signal decay described in Eqs. [41]–[44]. Recall, that the major contribution to this magnetic susceptibility difference is due to the presence of deoxyhemoglobin in blood, Eq. [20]. This deoxyhemoglobin containing blood constitutes only part of the blood vessel network, namely, the venous blood vessel network and part of the capillary network adjacent to venous side (capillary blood becomes deoxygenated as it moves from arterial to venous side). According to Eq. [20], there is a non-zero magnetic susceptibility difference even between the completely oxygenated blood in arteries ( \( Y = 1 \)) and surrounding brain parenchyma. However, this difference is rather small and usually can be neglected. In order to avoid confusion, we will call the parameter \( \zeta \) the deoxygenated Cerebral Blood Volume (dCBV).

The function \( f(t/t_c) \) in Eq. [44] is illustrated in Figure 3. In the short-time and long-time regimes, the function \( f(t/t_c) \) depends on its argument quadratically and linearly, respectively:
The linear behavior of the function $f(t/t_c)$ at large argument leads to a linear behavior of the signal attenuation function:

$$f(t/t_c) = \begin{cases} 0.3 \cdot (t/t_c)^2, & t/t_c \ll 1 \\ t/t_c - 1, & t/t_c \gg 1 \end{cases} \quad [45]$$

Importantly, the volume fraction $\zeta$ and the susceptibility difference $\Delta \chi$ appear in Eq. [46] only as a product, hence to resolve these two parameters, measurement interval should include short time regime $t < t_c$ (43) where parameters $\zeta$ and $\Delta \chi$ decouple because the function $f(t/t_c)$ exhibits a quadratic time dependence, Eq. [45].

Since the short time interval is important for decoupling of parameters $\zeta$ and $\Delta \chi$, it was proposed in (43) to use GESSE (gradient echo sampling of spin echo) approach (see Figure 4b) when MR signal is sampled around spin echo (SE). This strategy practically doubles the short time interval as compared to the FID experiment. The expression for MR signal around SE (GESSE signal) is given by Eq. [9] with

$$\Gamma_{GESSE}(t) = -\zeta \cdot f(t/t_c) + R_2(t + TE), \quad [47]$$

where time $t$ is counted from spin echo time $TE$ (and can be negative, though $t > -TE/2$).

It is important to note that the results in Eqs. [41] and [47] depend only on the total volume of blood vessels with the partially deoxygenated blood and does not depend on specific distribution of blood vessels. This result is in a very good agreement with the results obtained by others by means of Monte-Carlo computer simulations (83) and (98). This very important theoretical prediction allows application of this theory (42) to in vivo studies where a rather broad microscopic distribution of blood vessels can be expected.

Equations [41] and [47] were derived in the static dephasing regime framework. Here the presence of susceptibility-induced static magnetic field inhomogeneities cause the MR FID signal to decay much faster than a competing process, the averaging of the water $^1$H nuclear spin phases due to molecular diffusion. Validity criteria of the static dephasing regime analysis have been discussed in detail in (42,43). According to (42), the static dephasing regime for the FID signal holds if the characteristic dephasing time $t_c$, Eq. [42], is smaller than the characteristic diffusion time, $t_D = R^2/D$, where $D$ is the water diffusion coefficient (in the brain, $D \sim 1 \mu m^2/\text{ms}$) and $R$ is the characteristic blood vessel radius. For the blood vessel network modeled by randomly oriented cylinders discussed above, this condition is

$$(4\pi/3 \cdot \gamma \cdot \Delta \chi \cdot B_0)^{-1} < R^2/4D \quad [48]$$

The inequality [48] puts a restriction on the blood vessel radii, $R$, for which the static dephasing regime is valid:

$$R > 2(D t_c)^{1/2} \quad [49]$$
For $B_0 = 3T$, $\gamma = 2.675 \cdot 10^8 \text{ sec}^{-1} \text{T}^{-1}$, $\Delta \chi = 0.04 \text{ppm}$ (for venous blood with the Hct=0.4 and $Y=0.6$), the characteristic dephasing time $\tau_c = 7.4 \text{ ms}$, and the inequality [49] results in $R > 6 \mu \text{m}$. Thus, for the FID signal, the static dephasing regime might not quantitatively describe MR signal behavior in the presence of capillaries but is a very good approximation for the post-capillary blood vessel network (venules, veins, etc.).

The effect of water diffusion and validity criteria of the static dephasing regime were discussed in a number of publications, e.g. (42,43,46,53,54,100–102). Some of these approaches will be discussed in detail below.

### 5.2 Gaussian Phase Approximation (role of water diffusion)

To describe the MR signal attenuation for small capillaries, the Gaussian phase approximation can be used (53,54). In the framework of this approach, the expression for the signal, Eq. [37], is presented in the form first proposed in (103):

$$s(t) = \langle \exp(-i\phi(t)) \rangle = \int_{-\infty}^{\infty} d\phi P(\phi) \exp(-i\phi)$$  \[50\]

where $P(\phi)$ is a phase distribution function. As shown in (103), in the case of unrestricted diffusion in the constant field gradient, this function is of the Gaussian type:

$$P(\phi, t) = \frac{1}{2\pi(\phi^2(t))^{1/2}} \exp \left[ -\frac{\phi^2}{2\langle \phi^2(t) \rangle} \right] \quad \[51\]$$

and the MR signal corresponding to $P(\phi, t)$ in Eq. [50] can be described by Eq. [9] with signal attenuation function

$$\Gamma(t) = \frac{1}{2} \langle \phi^2(t) \rangle \quad \[52\]$$

(without loss of generality, we consider $\langle \phi \rangle = 0$). Importantly, averaging $\phi^2(t)$ in Eq. [52] rather than the exponent $\exp(i\phi)$ in Eq. [37] is a substantially less challenging problem because the expression for $\langle \phi^2(t) \rangle$ can be written in a closed form:

$$\langle \phi^2(t) \rangle = 2 \cdot \int_0^{t_1} \int_0^{t_2} G(t_1, t_2)$$

$$G(t_1, t_2) = \frac{1}{V} \cdot \int_{r_1}^{r_2} \int_{r_2}^{r_1} \int_{r_2}^{r_1} \int_{r_1}^{r_2} \Delta \omega r_1(t_1) \Delta \omega r_2(t_2) P(r_1, r_2, t_1 - t_2).$$ \[53\]

where $\Delta \omega r(t)$ is the susceptibility-induced local Larmor frequency shift, $G(t_1, t_2)$ is the frequency correlation function, $P(r_1, r_2, t)$ is a diffusion propagator describing the probability for a molecule to move from point $r_2$ to point $r_1$ during time $t$; integration is over all the initial and final positions of molecules in the system’s volume $V$.

If diffusion is restricted by some barriers or if the field gradients are non-uniform (as in our case of susceptibility-induced field inhomogeneities), the phase distribution function $P(\phi, t)$, is, in general, not Gaussian. However, in some cases it can be well approximated by a Gaussian function in Eq. [51] and the MR signal can be found from Eqs. [52]–[53] – the so-called Gaussian phase approximation (not to be confused with Gaussian diffusion). The adequateness of the Gaussian phase approximation has been discussed by many authors (104–107). A detailed quantitative comparison of the Gaussian phase approximation with exact results for some models of restricted diffusion in the presence of a constant field.
gradient was given in (107,108) for a broad range of system parameters. First, this approximation is valid at short diffusion times, when phase accumulated by diffusing spins is small, $\phi \ll 1$. Assuming that $\langle \phi \rangle = 0$, one can get $\langle \exp(i\phi) \rangle \approx 1 - \langle \phi^2 \rangle/2 \approx \exp(-\langle \phi^2 \rangle/2)$.

Second, it can be valid at long diffusion times (this condition is necessary but not sufficient!), when all diffusing spins have encountered boundaries many times, their trajectories become statistically identical, and the central limit theorem can be applied.

Thus, the quantity $\langle \phi(t) \rangle$ and, consequently, the signal in Eq. [52] can be calculated in systems for which the diffusion propagator $P(r_1, r_2, t)$ is available. For susceptibility induced magnetic field, the MR signal in the Gaussian phase approximation was calculated in (53,54) for different geometries of magnetized objects, including permeable and impermeable infinite cylinders. The FID signal attenuation function due to the magnetic field inhomogeneities induced by a blood vessel network comprised of cylindrical blood vessels (impermeable for water molecules) with uniform distribution of their orientations was found in (54):

$$\Gamma_{\text{FID}}(t) = \Gamma_0 \cdot \int_0^\infty \frac{g_{\text{FID}}(u^2 t/t_D)}{u^6 [J_2^2(u) + N_2^2(u)]} + R_2 t \quad [54]$$

where $J'(u)$ and $N'(u)$ are first derivatives of Bessel functions of the first and second kinds respectively, $\Gamma_0$ is given in Eq. [30] with the characteristic diffusion time $t_D = R^2 / D$, and

$$g_{\text{FID}}(x) = \exp(-x) + x - 1 \quad [55]$$

For GESSE pulse sequence acquiring data around spin echo time $TE$, the signal attenuation function is

$$\Gamma_{\text{GESSE}}(t) = \Gamma_0 \cdot \int_0^\infty \frac{g_{\text{GESSE}}(u^2 t/t_D)}{u^6 [J_2^2(u) + N_2^2(u)]} + R_2 \cdot (TE + t) \quad [56]$$

where

$$g_{\text{GESSE}}(x, y) = 2\exp(-y/2) + 2\exp(-(x+y)/2) - \exp(-(x+y)) + (x+y) - 3 \quad [57]$$

and variables $x = u^2 \cdot t_D$ and $y = u^2 \cdot TEB(t_D)$ ($x \gg y/2$). For a standard SE pulse sequence, when the signal is measured at SE time $TE (t = 0)$, Eq. [57] is simplified to (54):

$$g_{\text{SE}}(y) = g_{\text{GESSE}}(0, y) = 4\exp(-y/2) - \exp(-y) + y - 3 \quad [58]$$

In the short time regime, $t \ll t_D$, expansion of Eq. [54] for FID signal in a time series predicts that a leading term is quadratic with the same coefficient as in the static dephasing approximation, Eq. [45], however higher order expansion terms are different:

$$\Gamma_{\text{FID}} = \Gamma_0 \cdot \left[ \frac{1}{8} t^2 \frac{1}{12} t^3 + \frac{32}{35\pi^{1/2}} t^{7/2} + O(t^4) \right], \quad \tau = t/t_D \ll 1 \quad [59]$$

For a SE experiment, the expansion starts from a cubic term:
In the long-time regime, \( t \gg t_D \), the asymptotic behavior of the signal attenuation functions is more complicated:

\[
\Gamma_{SE} = \Gamma_0 \cdot \left[ \frac{1}{4} \tau^3 - \frac{8(4-\sqrt{2})}{35\pi^{1/2}} \tau^{3/2} + O\left(\tau^1\right) \right], \quad \tau = TE/t_D < 1 \quad [60]
\]

where \( C_1 - C_4 \) are numerical constants. Importantly, the main term in the asymptotic behavior of the attenuation functions at long time is \( t \ln t \), which was first predicted in (109). Hence, even in the long-time regime the MR signal cannot be described in terms of \( R_2 \) or \( R_2^* \) relaxation rate constants.

The model proposed in (44) is based on the same Gaussian phase approximation as that of (54) but also makes an additional assumption that the frequency correlation function \( G(t_1, t_2) \), Eq. [53], monoexponentially decreases with time,

\[
G(t_1, t_2) = \overline{\Delta \omega^2} \cdot \exp \left( \frac{[t_1-t_2]}{\tau_c} \right) \quad [62]
\]

where \( \tau_c \) is the correlation time of the frequency fluctuations and \( \overline{\Delta \omega^2} \) is the mean square Larmor frequency shift, depending on the volume fraction \( \zeta \) and geometry of susceptibility inclusions. In the framework of this approach, the signal attenuation functions for the FID and for a GESSE-type pulse sequence (asymmetric spin echo) turn out to be proportional to \( g_{FID} \), Eq. [55], and to \( g_{GESSE} \), Eq. [57], respectively,

\[
\Gamma_{FID} = \overline{\Delta \omega^2} \cdot \tau_c \cdot \frac{g_{FID}}{\tau_c} \cdot \frac{t}{\tau_c} \quad [63]
\]

\[
\Gamma_{GESSE} = \overline{\Delta \omega^2} \cdot \tau_c \cdot \frac{g_{GESSE}}{\tau_c} \cdot \frac{TE}{\tau_c}
\]

For the case of randomly oriented blood vessels, when \( \Delta \omega \) is given by Eq. [40] (long cylinders), \( \overline{\Delta \omega^2} = 3/5 \cdot \zeta \cdot \tau_c^2 \) (44), where the characteristic dephasing time \( \tau_c \) is defined in Eq. [42].

### 5.3 Monte-Carlo Simulations

Using Monte Carlo methods it is possible to directly simulate the BOLD signal for any tissue type and MRI pulse sequence parameters. Studies taking this approach (55, 67, 83, 98–100, 110, 111) have increased understanding of the link between BOLD signal and the underlying hemodynamics, which in turn has influenced fMRI experimentation design. In (100), Monte-Carlo simulation method was applied to show the validity criteria of a number of models from the literature. Simulation results were then used to construct a simple phenomenological model that is accurate for the physiological range of blood oxygenation, blood volume and vessel radii. First, the simulations were performed for blood vessel networks comprising of vessels with the same radius \( R \) and a uniform distribution of their orientations. Typical time dependence of the simulated FID and GESSE signals for different vessel radii \( R \) is demonstrated in Figure 4. The local Larmor frequency shift \( \Delta \omega(r) \) induced by each blood vessel is calculated by Eq. [40] with the susceptibility difference \( \Delta \chi = \Delta \chi_0 \cdot Hct \cdot OEF \) and the input parameters \( OEF = 0.4, \ \zeta = 0.03, \ \Delta \chi_0 = 0.27 \text{ ppm}, \ Hct = 0.34 \).
approximate Eq. [22] and $Y_c = 1$ are used here). It should be noted that the behavior of MR signal depicted in Figure 4 is similar to that reported by Fujita et al (55).

The signal corresponding to the static regime with $D = 0$ is formally equivalent to the case $R \rightarrow \infty$, when diffusion plays no role. In this case, the FID signal decay is the most rapid and restores its initial value at spin echo time ($t = 0$ in (B)). The FID signals for $R > 9 \mu m$ practically coincide with this case and can therefore be well described by the corresponding analytical expressions, Eqs. [41]–[44]. The signals clearly reveal a quadratic regime at short times and practically a mono-exponential behavior at long times, as predicted in Eqs. [45] and [54]–[55]. All the GESSE signals (except of the case of SDR) reveal a shift of their maxima from the spin echo time $TE ((t = 0$ in (B)) and do not restore their initial values at these maxima even in the absence of the transverse relaxation. This effect however can be described by introducing an additional exponential factor similar to a simple $T_2$ decay. Importantly, this factor does not affect ability of static dephasing model to quantitatively describe experimental data which was demonstrated in (100) (see Figure 5).

Ability of some theoretical models to reproduce the input values of simulation parameters ($OEF$, dCBV) for a broad range of $R$ was analyzed in (100) by fitting the models to the simulated signal. Some of the results are shown in Figure 5 (adapted from (100)). As expected, Eqs. [44], [47] corresponding to the static dephasing regime (red lines), reproduce the input parameters rather well for sufficiently large radii $R$, whereas the Gaussian phase approximation (green lines), Eqs. [54]–[57], is valid for small $R$. The “linear local field approximations” (blue lines) (45) provides reasonable results only in the region where it practically coincides with the static dephasing regime. Importantly, none of the models are accurate for all the radii or for the intermediate range $R \sim 10–20 \mu m$. To address this issue, a phenomenological model was proposed in (100).

5.4 Phenomenological model for extravascular signal

A realistic blood vessel network is comprised of vessels of different radii. According to (85), the distribution of vessels’ radii in the brain can be described by a truncated normal distribution:

$$P \left( \frac{1}{\sqrt{2R}} \right) = \left\{ \begin{array}{ll} N(0.38, 4.9 \times 10^{-3}) & 0 < R < 25 \mu m \\ 0 & \text{otherwise} \end{array} \right. \quad [64]$$

where $N(\mu, \sigma^2)$ is a normal distribution with mean $\mu$ and variance $\sigma^2$. This distribution was used in (100) for performing Monte-Carlo simulations of extravascular BOLD signal and developing a phenomenological model describing this signal. Analyzing simulated data, it was found that they can be extremely well described by the following phenomenological signal attenuation function (100):

$$\Gamma(t) = \zeta \cdot f(t), \quad f(t) = A_1 \exp(-A_2 t) - 1 + A_3 t + A_4 \quad [65]$$

where time $t$ is counted from the initial RF pulse excitation. The parameters $A_n (n = 1, 2, 3, 4)$ are found by fitting Eq. [65] to the simulated data for different values of $OEF$. It turned out that all parameters $A_n$ depend on $OEF$ according to the equation

$$A_n = OEF \cdot B_1 - B_2 \exp(-B_3 \cdot OEF) \quad [66]$$

The values of the parameters $B_{1-3}$ (different for different $A_n$) for the FID and GESSE sequence are shown in Table 2.
6. qBOLD - Quantitative Mapping of Brain Hemodynamics and Metabolism

One of the experimental methods for quantifying brain hemodynamic properties is quantitative BOLD (qBOLD). This technique was proposed in (86) and verified on animal model in (112). It is based on a theory of MR signal formation in the presence of blood vessel network (42), experimental method GESSE proposed and verified on phantoms in (43) and a realistic consideration of multi-compartment tissue structure. We have already discussed in detail theoretical background of the BOLD effect in previous sections of this review. Experimental consideration is provided below.

6.1 GESSE Approach

To extend the range of quadratic signal behavior, Eq. [45], that is essential for separation of contributions from blood oxygenation level ($Y$) and blood volume fraction (dCBV), it was proposed in (43) to sample the MR signal around the spin echo time by using the GESSE sequence (113), which is a modified version of the Gradient Echo Sampling of FID and Echo (GESFIDE) sequence proposed and used earlier (114). The GESSE sequence consists of embedding a set of gradient echoes around the echo of a single spin echo sequence, Figure 6.

6.2 Microscopic, Mesoscopic and Macroscopic Field Inhomogeneities

The magnetic field inside practically any system placed into an MRI scanner is always inhomogeneous due to a number of reasons. The relative scale of field inhomogeneities compared to an imaging voxel can roughly be divided into three categories: microscopic, mesoscopic and macroscopic (43). Microscopic scale refers to changes in magnetic field over distances that are comparable to atomic and molecular lengths. Fluctuating microscopic field inhomogeneities lead to the irreversible signal dephasing characterized by the $T_2$ relaxation time constant and also to the longitudinal magnetization changes characterized by the $T_1$ relaxation time constant.

In the previous sections we mainly discussed the inhomogeneous magnetic fields induced by susceptibility difference between blood and surrounding tissue. This type of inhomogeneities can be referred to as mesoscopic because their characteristic scale (several microns) is on the order of RBC and blood vessels sizes, hence much smaller than typical imaging voxel dimensions (about mm) but much bigger than the atomic and molecular scale.

Macroscopic scale refers to magnetic field changes over distances that are larger than dimensions of the imaging voxel. Macroscopic field inhomogeneities arise from magnet imperfections, boundaries separating different tissues, body-air interfaces, large (compared to voxel size) sinuses inside of the body, etc. These field inhomogeneities are undesirable in most MR imaging experiments because they generally provide little information of physiologic or anatomic interest. Rather, they lead to such effects as signal loss in gradient-echo imaging and image spatial distortions in both gradient-echo and spin-echo imaging. Due to the differences in spatial scale of microscopic, macroscopic and mesoscopic field inhomogeneities, the total qBOLD signal can be represented as a product of the signal that would exist in the absence of macroscopic field inhomogeneities, Eq. [9], multiplied by a function $F(t)$, representing contributions caused by macroscopic field inhomogeneities (43)

$$S(t) = S(0) \cdot \exp \left(-\Gamma(t) \cdot F(t)\right)$$  [67]

Although mesoscopic type of magnetic field inhomogeneities is of main interest for our purpose, the correct interpretation of experimental data requires accounting for the
macroscopic inhomogeneities as well. The consideration of this issue has been provided in multiple papers (e.g. (43,115)) and is beyond the scope of this review paper.

6.3 Phantom Studies

Two phantoms have been used in (43) to test the theoretical concept of MR signal formation in the presence of mesoscopic field inhomogeneities and to validate GESSE approach. The first phantom mimics the theoretical model of randomly distributed cylinders with random axial directions whereas the second phantom mimics the theoretical model of randomly distributed parallel cylinders (42). Here we briefly reproduce results obtained from the second phantom. Data was acquired using GESSE sequence, Figure 6 described in detail in (43).

The right panel in Figure 7 illustrates a specific quadratic behavior of GESSE MR signal around SE predicted by the theory, Eqs.[45] and [47] and also consistent with Eq. [59]. This experimentally measured GESSE signal behavior is similar to that found in Monte-Carlo simulations (55,100) and illustrated in Figure 4 (right panel). The volume fraction of filaments \( \zeta \) and their magnetic susceptibility \( \Delta \chi \) were found by fitting theoretical Eq. [67] with mesoscopic signal attenuation function for parallel cylinders (42) similar to Eq. [43], to experimental data on a voxel-by-voxel basis. Results are presented in Figure 8. Maps in the first row were obtained when macroscopic field inhomogeneities were accounted for by using a Gaussian correction term (43) in \( R(t) \), Eq. [67]. The value of volume fraction \( \zeta \) occupied by the filaments from the map is equal to \( \zeta = 0.063 \pm 0.016 \) (mean \( \pm \) SD) in a good agreement with the directly estimated value of 0.06. The value of the susceptibility difference between filaments and solution was \( \Delta \chi = 0.055 \pm 0.005 \) ppm. Note that the SD for the volume fraction \( \zeta \) is about 25% of its mean value. In contrast, the SD for the susceptibility difference \( \Delta \chi \) is only 9%. This is because, in the case of \( \zeta \), the SD reflects primarily not only measurement error, but also inter-voxel structure variability, which comes about from a non-homogeneous filament distribution and orientations. However, the susceptibility difference \( \Delta \chi \) does not directly depend on the structure variability and, as a result, has a much smaller SD.

Maps in the second row in Figure 8 were obtained without correction for macroscopic field inhomogeneities and, as a result, are heavily contaminated with artifacts. This contamination is especially prominent near the sides of the container where the magnetic field apparently was distorted. The comparison of images in Figure 8 (upper and lower rows) clearly demonstrates the importance of taking into consideration the macroscopic field inhomogeneities.

6.4 qBOLD - Multi-Component Tissue Model

Several attempts (116–119) to directly implement the above described method (43) in vivo provided important insights into the problem. However, the simplistic model used therein, describing brain tissue as a one component structure similar to water in a phantom, is not sufficient to describe real brain tissue. It was long recognized that an adequate description of BOLD effect requires accounting for multi-compartment brain tissue structure (e.g., (74,75,92,99,120–122)). Also, direct measurements of the \( R_2 \) relaxation rate constant in the brain tissue by Fujita et al (55) demonstrated that the \( R_2 \) measurement depends on the spin echo time in the GESSE sequence. In (55), this effect was correctly attributed to the multi-component structure of brain tissue, whereas each component has a different \( T_2 \) relaxation time constant and fractionally weights the measurement differently for different spin echo times. Hence, the correct model of brain tissue should also incorporate the multi-component tissue structure.
A manifestation of multi-compartment structure of brain MR signal in transverse relaxivity measurements was reported previously by Whittal et al. (123). For WM these authors identified three components in brain tissue - fast, intermediate and slow with $T_2 = 15, 77$ and 250 msec, and fractional weights of 11, 84 and 5%, respectively. For GM, these authors found fast and intermediate components with $T_2 = 15$ and 87 msec and fractional weights of 5% and 95%, respectively. For a MR experiment with the spin echo time ranging from 25 ms to 80 ms, the fast component ($T_2$ approximately 15 ms) is “invisible” due to $T_2$ decay and its low concentration. Extracellular fluid in the central nervous system is composed of cerebrospinal fluid (CSF) and interstitial fluid (ISF) in gray and WM. The extracellular space of the brain freely communicates with the CSF compartment. Hence, the composition of the two fluid compartments is relatively similar. Bearing in mind partial volume effects of CSF, we adopt a model of brain tissue that treats both GM and WM as consisting of two components with intermediate (around 70 to 90 msec) and long (more than 250 msec) relaxation time constants. $T_2$-based measurements (124) also support such a model. Note that the $T_2$ values of CSF directly measured within cortical lobes were reported to be more than 1000 ms (125). The estimated $T_2$ values around 250 ms (123) may be due to the ISF compartment.

To reflect this multi-compartment structure of extravascular tissue signal in GESSE experiment, He and Yablonskiy (86) presented the qBOLD model signal as a sum of three contributions:

$$S_{\text{total}}(t) = S(0) \cdot \left[ A_e^{(1)} \cdot s_e^{(1)}(t) + A_e^{(2)} \cdot s_e^{(2)}(t) + \zeta \cdot s_b(t) \right] \cdot F(t), \quad [68]$$

where indices (1) and (2) and $b$ correspond to two extravascular compartments (main tissue and ISF/CSF), and intravascular compartment; the function $R(t)$ describes the macroscopic field inhomogeneities (43), and

$$s_e^{(1)}(t) = \exp \left[ -\zeta \cdot f(t/t_e) - R_2^{(1)} \cdot (t+TE) \right], \quad [69]$$

$$s_e^{(2)}(t) = \exp \left[ -\zeta \cdot f(t/t_e) - R_2^{(2)} \cdot (t+TE) - i \cdot 2\pi \cdot \Delta f \cdot t - i\phi_0 \right], \quad [70]$$

$$s_b(t) = \exp(i \cdot \frac{t}{2t_e} \cdot \frac{C \eta - i \text{sgn}(t) S \eta}{\eta}) \cdot \exp \left[ -R_2^b \cdot (t+TE) \right], \quad [71]$$

Equation [71] is slightly modified as compared to Eq. [25] because in GESSE sequence time $t$ is counted from SE time $TE$ and can be both negative and positive. Note that the second extravascular compartment has also frequency shift and phase shift with respect to the first one (86). As already mentioned in Section 4.2, the expression for blood signal can be generalized by accounting for differences in oxygenation levels of veins, capillaries and arteries.

Separation of extravascular tissue in two compartments is, of course, a simplification. Both the intracellular and interstitial compartments are complex structures and each should be described by means of distribution functions of their parameters (e.g. (108,126)). Also, exchange effects between compartments make such a separation rather conditional; though relatively long exchange times (hundreds-to-thousand ms, e.g., (127–131)) make this separation plausible.
6.5 qBOLD - Human Studies

Mapping of brain tissue OEF and dCBV using described above qBOLD approach that is based on GESSE method (43) and multi-compartment brain tissue structure was first reported in (86). Representative signal intensity evolution profiles from two selected voxels in GM and WM obtained with the GESSE sequence are illustrated in Figure 9.

The residual profile (Figure 9d) indicates that the fitting error of the qBOLD model is comparable to the background noise level. A peak in the residual at the spin echo time reflects the presence of a small concentration of lipids and metabolites in the tissue which are not included in the qBOLD model. The signal contributions from CSF/ISF and intravascular venous blood are illustrated in Figure 9e and Figure 9f, and indicate a significantly higher CSF/ISF signal fraction in GM regions than in WM. Figure 9c shows the extravascular brain tissue signal after removing the effect of $T_2$ decay and the contributions from CSF/ISF and intravascular blood.

As already mentioned above, the qBOLD signal from the brain, Eq. [68], is comprised of contributions from three compartments: brain tissue, CSF/ISF, and blood vessel network. At the same time, numerous publications treat BOLD signal in the framework of a single compartment model which can introduce bias if quantitative results are needed. Although the signal amplitude from CSF/ISF (Figure 9e) and blood vessel network (Figure 9f) are indeed much smaller than the signal from the brain tissue, their presence in the qBOLD model is very important. To demonstrate this fact, the same experimental qBOLD signal from the same representative pixel as in Figure 9 was fitted by means of the model without CSF/ISF signal (the second term in Eq. [68]). The residual of fitting procedure is shown in Figure 10. This residual is substantially bigger than that obtained when using the multi-compartment model (see Figure 9f for comparison) and drastically increases with time. Thus, we can conclude that the model without CSF/ISF signal fails to fit experimental data at long time.

Figure 11 shows maps of the estimated brain parameters. In the maps of dCBV, CSF/ISF volume fraction and brain tissue deoxyhemoglobin concentration, the contrast is sufficient to resolve GM and WM. Meanwhile, the maps of OEF do not show contrast between GM and WM in agreement with the literature data for healthy human subjects (21). The concentration of deoxyHb, $C_{\text{deoxy}}$, is calculated according to the following equation (86):

$$C_{\text{deoxy}} = \zeta \cdot (1-Y) \cdot Hct \cdot n_{Hb} \quad \text{(72)}$$

where $n_{Hb} = 5.5 \cdot 10^{-6}$ mol/ml is the concentration of hemoglobin in RBC (30).

Table 3 summarizes results obtained from nine studies conducted on normal healthy volunteers (86). Within the entire group, an average OEF value is 38.3±5.3%. This variation may reflect different subject’s physiological condition. Both mean OEF values and inter-subject variability are consistent with previous results of PET studies obtained for normal volunteers. Indeed, (20,132–134) reported OEF values of 35 ± 7%, 42.6± 5.1%, 41± 6%, and 40± 9%, correspondingly.

The proposed qBOLD method, Eqs. [68]–[71], requires multiple fitting parameters to describe the contributions from different compartments. This makes the fitting procedure rather challenging. As demonstrated in (43,86,135,136), the robustness of OEF and dCBV estimation depends on a high SNR to resolve small deviation from a linear behavior in Eq. [47], at short time scale ($\tau < 1.5\tau_c$). Recent phantom and in vivo healthy human studies (136) found a strong interdependency between dCBV and OEF estimation. This is expected since $R2^\prime$ of extravascular tissue (directly related to concentration of deoxyHb, and
proportional to the product of dCBV and OEF) can be reliably estimated from the $R2^*$ decay time course before and after the spin echo. Meanwhile, a reliable separation between dCBV and $OEF$ is heavily dependent on the accurate quantification of MR signal behavior around spin echo time ($|t| < 1.5\cdot t_c$). A detail analysis of the uncertainties of the qBOLD parameter estimates in the framework of the Bayesian probability theory (e.g., (137,138)) is provided in (139) where it was demonstrated that a rather high SNR (more than 500) is required for accurate estimate of dCBV and $OEF$.

A robustness of qBOLD quantification can be improved by independent measurements of some model parameters. In the framework of a single compartment model this idea was implemented by Christen, et al, (140–142). In their approach, the change of tissue $R2^*$ decay rate constant after administration of intravascular contrast agent (ultrasmall superparamagnetic iron oxide, USPIO) provides an estimation of total blood volume fraction (CBV). Tissue $R2$ is pre-determined from a multi spin echo experiment. By ignoring vascular contribution and adopting a single extravascular compartment model, a reasonable correlation between the MRI derived local blood oxygenation and the blood-gas measurement of sagittal sinus was established. Note, however, that this method provides an average blood oxygenation within the entire vasculature, which is different than the venous blood oxygenation, dCBV, derived from qBOLD approach. It should also be noted that an accurate application of such an approach requires accounting for a multi-compartment tissue structure.

Further improvements in qBOLD technique can also be achieved by accounting for water diffusion effects in the model (see Section 5 above) and (111).

### 6.6 Validation of the qBOLD technique – animal study

Validation of qBOLD technique was performed in (112), where $OEF$ measurements provided by the qBOLD technique were compared with direct $OEF$ measurements in a rat model. Cerebral venous blood oxygen level of rat was manipulated by utilizing different anesthesia methods and different level of oxygen in inhaled air. The venous blood samples were drawn directly from the superior sagittal sinus. Compared with the jugular venous oxygenation monitoring approach, this method eliminates contamination from extracranial sources (143,144). Blood oxygen saturation level was independently directly measured using an i-STAT Portable Clinical Analyzer. A 3D GESSE sequence was employed and the MR signal was then analyzed using the qBOLD model, Eq [68].

A representative $T1$-weighted image and maps of estimated venous oxygen saturation level $Y_v$ (color bar) from one slice in the same rat under isoflurane and alpha-chloralose anesthesia are demonstrated in Figure 13. Isoflurane is known to slightly elevate cerebral blood flow (CBF) and significantly reduce brain metabolism (145). Alpha-chloralose is known to suppress both CBF and metabolism (146). Hence, one can expect that the venous blood oxygen level under isoflurane anesthesia should be much higher than in the alpha-chloralose case. This is consistent with the Figure 13, which shows a fairly homogenous venous blood oxygen saturation level across the brain, with an estimated mean venous $O2$ level of 77% under isoflurane anesthesia and 62% under alpha-chloralose anesthesia.

Figure 14 illustrates the results obtained from 13 experiments where the venous oxygen level was varied from 44% to 87% by manipulating the concentration of inhaled oxygen and the type of anesthesia. Data show a very good agreement ($R^2=0.92$) between the qBOLD-derived and direct blood gas measured oxygen saturation levels in rat brain providing a direct validation of the qBOLD technique.
6.7 ASL-qBOLD technique for quantitative mapping of CMRO$_2$

Combining qBOLD measurements of OEF with ASL measurements of CBF, allows quantitative mapping of tissue oxygen consumption CMRO$_2$ – Eq. [1]. The method described in (147) is based on a GESSE sequence with arterial spin labeling (ASL) preparation pulses and is similar to previously used for studying water exchange in brain tissue (131). FAIR-based perfusion sensitive preparation pulses (148) are used to generate perfusion weighted GESSE qBOLD images. QUIPSS II technique (149) with saturation pulse train starting at 900 ms before GESSE acquisition is incorporated to shape the labeling bolus for a robust CBF quantification. Figure 15 shows a typical result from CMRO$_2$ quantification from a healthy subject. The estimated CBF map, Figure 15b, (in ml/100g/min), delineates the contrast between GM and WM. The measured mean CBF was 52±10 ml/100g/min in GM. Figure 15c illustrates the estimated OEF map with mean values of 38±9 %. Figure 15d shows the CMRO$_2$ map (in μmol/g/min). Note that the CMRO$_2$ was much higher in the cortical GM than the WM. The mean CMRO$_2$ in GM area was 1.77±0.56 μmol/g/min, which is consistent with that measured by PET imaging (150). The data demonstrate that the ASL-qBOLD method has a potential for mapping brain CBF, OEF and CMRO$_2$ from a single MRI scan.

7. MR Susceptometry-based CMRO$_2$ quantification

Simultaneous estimation of oxygen saturation and cerebral blood flow in the major vessels draining and feeding the brain can be used for rapid non-invasive quantification of whole-brain CMRO$_2$. The vessel of interest often includes internal jugular vein and/or superior sagittal sinus (SSS). The principle of the MR susceptometry of the whole brain is based on the measurement of the susceptibility difference between blood in the draining vein (such as jugular vein or SSS), and its surrounding tissue by measuring the phase difference with a GRE sequence (field mapping) (91,115,151–153). The SSS is often preferred over the internal jugular vein where severe susceptibility artifacts, caused by the proximity of air spaces such as the oral cavity and trachea, may complicate measurements. An additional benefit of the SSS is the elimination of contamination by the blood from extra-cranial sources (144).

In (152), the draining vein was considered as a long cylinder with a fixed orientation relative to the static magnetic field $B_0$. For such a model, the proton Larmor frequency is homogeneous within the cylinder (blood vessel), Eq. [23], and inhomogeneous outside the cylinder (surrounding tissue), Eq. [40]; however, the average frequency in the surrounding tissue is equal to 0 (due to the factor $\cos^2 \phi$ in Eq. [40]). These frequencies are usually measured from a set of phase images acquired at different gradient echo times. A difference between gradient echo MR signal phase inside the vein and an average phase of MR signal in the tissue is equal to (see Eq. [23]):

$$\Delta \phi_1 = 2\pi \gamma B_0 \Delta \chi \left( \cos^2 \theta - \frac{1}{3} \right) \cdot \Delta TE \quad [73]$$

where $\theta$ is the relative angle between blood vessel and the static magnetic field $B_0$ and $\Delta TE$ is the interval between gradient echo times. The fraction of oxygenated hemoglobin $Y$ can then be determined as (152):

$$Y = 1 - \frac{\Delta \phi_1}{2\pi \cdot \gamma \cdot \Delta \chi_0 \cdot Hct \cdot B_0 \cdot \cos^2 \theta - 1/3 \cdot \Delta TE}, \quad [74]$$
This result corresponds to a simplified relationship between blood magnetic susceptibility and blood oxygenation level, Eq. [22]. Note that in (152), the SI units are used whereas in this paper we use Gaussian-CGS units. To reduce the phase artifacts from moving blood, a flow-compensated 2D GRE sequence is often adopted. Blood vessel curvature and tapering may produce potential deviations from the long cylinder model. To address this issue, the accuracy of the “long cylinder” approximation has been evaluated via numerical computation of the induced magnetic field from 3D segmented renditions of three veins of interest (superior sagittal sinus, femoral and jugular vein) (154). The authors concluded that at a typical venous oxygen saturation of 65%, the absolute error via a close-form cylinder approximation was only 2.6% when measured over three locations in three veins studied and did not exceed 5% for vessel tilt angles <300 at any location. The influence of vessel eccentricity was examined theoretically and experimentally in (155). The problem of macroscopic magnetic field inhomogeneities resulting in low spatial-frequency variations in the reference tissue was corrected in (156) by means of a polynomial approximation for a field map.

An example of this technique is illustrated in Figure 16 where GRE signal phase was quantified for a baseline state and during hypercapnia (91). Notice the expected decrease in phase contrast for the SSS, indicating increased blood oxygenation level during hypercapnia. To evaluate CMRO2, in addition to Eq.[74], the blood flow in the vein should also be quantified. In (152) CBF was quantified using a phase-contrast technique (63,157). With a separate CBF quantification by phase-contrast MRI, MR oximetry has been successfully applied to quantify global brain oxygen metabolism at baseline (152). The sensitivity of this technique has also been demonstrated during breath-holding and hypoventilation (115).

In pediatric traumatic brain injury (TBI) patients, depressed global brain OEF has also been observed, while global OEF measured within two weeks of injury was found to be predictive of patient outcome at 3 mo after injury (158). Susceptibility-based MR oximetry has also been adapted to quantify oxygen kinetics of skeletal muscle in peripheral arterial diseases (159,160). With the improvement of MR technique, the acquisition time has been reduced from 4 mins to ~30 seconds, making the technique suitable for studying the temporal variations in CMRO2 in response to physiological challenges (91). In a recent paper (153), the susceptibility-based blood oxygen saturation technique was extended to quantify oxygen levels in small vein segments within gray matter, thus opening an opportunity for measuring regional oxygen metabolism of brain tissue.

8. $T_2$-based CMRO2 quantification

Another approach for quantifying biological tissue hemodynamic properties is based on measuring blood $T_2$ relaxation that is related to blood oxygenation level (see Section 4.3 above). In (89), CPMG pulse sequence was used to measure the blood transverse relaxation rate constant, see Eq. [34]. The reduction in acquisition time when comparing a CPMG sequence with single SE sequences may provide advantages such as the reduction in the motion sensitivity of the $T_2$ determination, and possible inter-scan differences in the physiological reactions to the stimulus (125).

One potential problem of the aforementioned approach is the possible signal contamination of intravascular blood from tissue and CSF/ISF signals due to partial volume effects (81), similarly as reported in (161). The isolation of pure venous blood signal is not a trivial task to carry out in vivo, even in large blood vessels such as the sagittal sinuses. The susceptibility to contamination by CSF/ISF space known as the arachnoid granulations often leads to non mono-exponential MR signal decay (162). In addition, the rapid movement of blood, especially in large vessels, presents a second technical challenge for studies seeking...
an accurate quantification of absolute blood oxygenation. To address these issues, Lu et al (162) proposed a spin-labeling technique, TRUST (T2-Relaxation-Under-Spin-Tagging), which can isolate pure venous blood signal (see also (163)). TRUST technique minimizes the partial volume effect and avoids the need for judicious selection of voxels containing blood. The T2 relaxation time of the TRUST signal can then be determined and converted to venous oxygenation Yv using a calibration plot (74,75,164). To minimize the outflow effect in blood T2 quantification, the T2-weighting of isolated blood signal in major blood vessels is often achieved by using a series of non-slice-selective T2-preparation pulses rather than the conventional spin-echo sequence. Figure 17 illustrated the TRUST sequence diagram (165). In (125) and (166), a non-selective saturation pulse was applied right after the signal acquisition to reset magnetization in the whole brain, thus improving the speed and reliability of TRUST MRI.

The theoretical framework of TRUST MRI is similar to that of ASL, but the scenario is simpler for TRUST MRI because the blood and static tissue compartments are strictly segregated (venous vessel walls are not permeable to water molecules) and there is no exchange effects during the inversion time. This technique has been validated in humans (164). It has shown promise for normalization of fMRI signals (162,167) and for studying cognitive aging (168) and multiple sclerosis (169). Methodological studies have also confirmed the sensitivity of this technique to physiologic procedures such as hypercapnia/hyperoxia (166) and caffeine uptake (162).

Spatial information is largely lost in TRUST or other T2-based approaches. Because of spatial resolution limitation, usable voxels are typically found only in large venous vessels. The QUIXOTIC (quantitative imaging of extraction of oxygen and tissue consumption) technique introduced by Bolar et al (170) separated the post-capillary venous blood signal from decelerating arterial blood and static tissue using velocity selective spin labeling (VSSL) technique. By measuring the T2 of this post-capillary blood, it is possible to generate the brain OEF thus CMRO2 maps, as shown in Figure 18 (170). Reliability of this approach was further improved in (171) by nulling the MR signal from inflow arterial blood and combining with phase sensitive signal detection. The improved method successfully detected the change of oxygenation during a hypoxia challenge.

9. Concluding Remarks

Studying brain hemodynamics and metabolism in health and disease attracts attention of scientists and clinicians, and numerous papers have been published on this subject. In this paper we review several BOLD-based experimental MR methods for studying brain function and discuss in detail theoretical background behind these techniques. These techniques allow evaluation of OEF through measuring blood oxygenation levels and (in combination with measuring CBF) estimation of CMRO2. Of course, in a single review it is impossible to even mention all publications on this issue and our list of references is far from being exhaustive. A direct measurement of CMRO2 is also possible by using tracers such as 17O2. This novel technique was described in detail in a recent review (172) and therefore is not discussed in our paper. We also have not included numerous publications aimed at measuring changes in brain hemodynamics and metabolism during functional activation as they are discussed in other articles of this special issue of NMR in Biomedicine. The present paper gives an updated picture of recently developed MR-based approaches for quantifying brain OEF, dCBV and CMRO2 properties1.

1When this paper was in revision, two publications devoted to development of yet another method for quantification of brain metabolism appeared in press (173, 174). They are based on an extension of the so-called calibrated fMRI approach (175, 176). A detail discussion of this method can be found in (177), see also related publications (148, 178–180).
Acknowledgments

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Figure 1.
Time dependence of the signal attenuation functions $\Gamma_{\text{FID}}(t)$ and $\Gamma_{\text{SE}}(t)$ (normalized to the characteristic value $\Gamma_0$, Eq. [30]) calculated in the Gaussian phase approximation (adapted from (53)).
Figure 2.
The RBC shape described by Eq. [31].
Figure 3.
The function $f(t/t_c)$ in Eq. [44] (adapted from (42)).
Figure 4.
Time dependence of the simulated FID (A) and GESSE (B) signals at different vessel radii $R$. Input parameters are: $OEF = 0.4$, $\zeta = 0.03$. Light blue lines correspond to the case $D = 0$ (static dephasing regime – SDR). All other lines correspond to $D = 1 \mu m^2/ms$. Spin echo time $TE = 60ms$ corresponds to $t = 0$ in (B) (shown by dashed vertical line). The transverse $T_2$ relaxation is ignored. (Modified from (100)).
Figure 5. (adapted from (100)). Values of $OEF(A, B)$ and $dCBV(C, D)$ found when fitting the models to simulated datasets over a range of blood vessel radii and using a variety of BOLD models. The input parameters for all simulations are: $OEF = 0.4$, $dCBV = 0.03$ (shown by dashed gray lines). The lines are only shown to display trends. Subplots A and C represent results for the FID sequence, subplots B and D represent results for the GESSE sequence. Red lines – static dephasing model, Eqs. [44], [47]; blue lines – “linear local field approximation” (45); green lines – Gaussian phase approximation, Eqs. [54]–[58] (for the latter, $dCBV$ is assumed to be known from independent measurement).
Figure 6.
Schematic structure of the GESSE sequence designed to give pixel-wise sampling of the MRI signal around the spin echo; RF – radio frequency pulse, GS – slice selection gradient, GP – phase encoding gradient, and GR – read-out gradient. Both multi-slice 2D (shown here) and 3D versions can be implemented. The signal is sampled only in the presence of same sign read-out gradients. This is especially important in the presence of the macroscopic static magnetic field inhomogeneities, which differently distort images collected in the presence of sign-alternating read-out gradients.
Figure 7. (adapted from (43)). A high resolution image illustrating the structure of the second phantom (left panel) and typical signal behavior (on a logarithmic scale) taken from a voxel in the middle of the phantom (areas with filaments) (right panel).
Figure 8.
(adapted from (43)). Maps of the volume fraction $\zeta$ (first column), the $R'_2$ relaxation rate constant (second column), and the susceptibility difference $\Delta \chi$ (third column).
Figure 9. (adapted from (86)). Representative data and fitting curves obtained with the GESSE sequence (64×64 matrix). (a) signals and the fitted profiles for voxels in GM area (blue line) with dCBV=1.56% and OEF=32.9% and WM area (red line) with dCBV=0.62% and OEF=33.1%; (b) high resolution anatomic $T_1$ weighted image showing selected voxels; (c) extravascular signal contributions after removing signals from CSF/ISF, intravascular blood and adjusting for the $T_2$ decay (multiplying by the factor $\exp(+R_2\cdot TE)$), the black solid lines correspond to the extrapolated signal profile from the asymptotic behavior at long echo times; (d) fitting residuals; (e) magnitudes of the CSF/ISF signals; (f) real parts of intravascular blood signals. In all plots x axis corresponds to a gradient echo time elapsed from the SE time (TE=36.4 msec), y axis represents signal in relative units. The echo spacing is 1.2 ms.
Figure 10.
The residual of the fitting of the model in Eq. [68] without contribution from ISF/CSF to the signal from a representative pixel (same as in Figure 9).
Figure 11. (adapted from (86)). Representative maps of estimated brain parameters obtained with a high resolution (128 × 128) GESSE sequence. Top leftmost image is a high resolution anatomic image. The rest of the maps are: dCBV fraction (%); OEF(%); \( R_2 \) of brain tissue (s\(^{-1}\)); CSF/ISF volume fraction; CSF/ISF frequency shift (Hz); \( R_2 \) of brain tissue (s\(^{-1}\)); and brain deoxyhemoglobin concentration \( C_{deoxy} \) (μM).
Figure 12. (adapted from (86)). Histograms of the estimated brain parameters for the study shown in Fig. 11. The histograms represent dCBV (%), OEF (%), CSF/ISF volume fraction, CSF/ISF frequency shift (Hz); $R_2$ of the brain tissue (s$^{-1}$), and the concentration of deoxyhemoglobin ($\mu$M).
Figure 13.
(adapted from (112)). A representative $T_1$ weighted image and maps of the estimated venous oxygen saturation level $Y_v$ (color bar) from one slice in the same rat under isoflurane and alpha-chloralose anesthesia.
Figure 14. (adapted from (112)). Comparison of $Y_v$ obtained by means of the qBOLD and by direct measurements of the venous oxygen level.
**Figure 15.**
(adapted from (147)). Quantification of CMRO₂ in healthy subject using ASL-qBOLD technique during the resting state. (a) $T₁$-weighted anatomical image; (b) CBF map (in ml/100g/min) delineates the contrast between GM and WM; (c) OEF map is mostly uniform across the whole brain (two spots at the back of brain with extreme high OEF values correspond to large uncompensated $B₀$ field inhomogeneities); (d) the map of CMRO₂ (in $µmol/g/min$) demonstrates much higher value of CMRO₂ in GM than in WM.
Figure 16.
(adapted from (91)). Images from which venous oxygen saturation during baseline, hypercapnia, and recovery were derived. (A) Sagittal localizer angiogram indicating the locations of SSS; (B) axial magnitude image; (C–E) GRE phase-difference images. Note the change in contrast in vessels during hypercapnia.
Figure 17.
(adapted from (162)) TRUST MRI technique Sequence diagram. (a) Pulse sequence diagram for TRUST MRI. The sequence consists of interleaved acquisitions of label and control scans, and each image type is acquired with four different effective TE\(s\) ranging from 0 to 160 ms. For each scan, the sequence starts with a presaturation RF pulse to suppress the static tissue signal, followed by a labeling (or control) RF pulse to magnetically label the incoming blood. A brief waiting period (1.2 seconds) is allowed for blood to flow into the imaging slice. Before data acquisition, a non-selective \(T_2\)-preparation pulse train is applied to achieve the \(T_2\)-weighting. The \(T_2\)-preparation scheme, instead of conventional \(T_2\)-weighted sequence, is used in order to minimize the blood outflow effect. (b) Positions of the imaging slice (yellow) and the labeling slab (green).
Figure 18.
(adapted from (170)). An example of quantitative $Y_i$ (left), $OEF$ (middle), and $CMRO_2$ (right) maps obtained by QUIXOTIC technique.
Table 1
The susceptibility of “average normal” tissue (relative to water, excluding intravascular blood) in frontal lobe (in $10^{-3}$ ppm) (82)

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<th>GM</th>
<th>WM</th>
<th>CSF</th>
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<tr>
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<td>-4.56</td>
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<tr>
<td>Lipids</td>
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<td>5.52</td>
<td>0</td>
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<tr>
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<td>4.40</td>
<td>0.33</td>
</tr>
<tr>
<td>Total</td>
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<td>5.36</td>
<td>0.33</td>
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Table 2

Values of parameters when fitting Eqs. [65]–[66] to simulated FID (first number) and GESSE (second number after//) time courses using a physiological distribution of radii, Eq. [64]; $OEF = 0$–100%; for GESSE time course, $t > TE/2$ and $TE = 60$ ms.

<table>
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<th>$A_3$</th>
<th>$A_4$</th>
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<td>35.3/0.03</td>
<td>0.31/0.26</td>
<td>0/58.1</td>
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<tr>
<td>$B_2$</td>
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<td>0.30/0.18</td>
<td>0/62.9</td>
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<tr>
<td>$B_3$</td>
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<td>−0.003/2.14</td>
<td>3.12/6.15</td>
<td>0/1.58</td>
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</table>
Table 3
(modified from (86)). The estimated brain parameters – average values obtained from nine studies.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>GM (%)</th>
<th>WM (%)</th>
<th>OEF (%)</th>
<th>( R_2^* ) GM (s(^{-1}))</th>
<th>( R_2^* ) WM (s(^{-1}))</th>
<th>dHb GM (μM)</th>
<th>dHb WM (μM)</th>
</tr>
</thead>
<tbody>
<tr>
<td>dCBV GM</td>
<td>1.75±0.13</td>
<td>0.58±0.09</td>
<td>38.3±5.3</td>
<td>2.9±0.4</td>
<td>0.68±0.10</td>
<td>12.4±1.4</td>
<td>4.4±0.8</td>
</tr>
</tbody>
</table>