The thermodynamics of thinking: connections between neural activity, energy metabolism and blood flow

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'Significant relationships between non-invasive functional neuroimaging and the underlying neuronal activity'

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Supplementary Information is contained in a separate file

Summary

Functional magnetic resonance imaging (fMRI) and other current functional neuroimaging methods are sensitive to cerebral metabolism and cerebral blood flow (CBF) rather than the underlying neural activity itself. Current studies have shown that the connections between metabolism, flow and neural activity are complex and somewhat counterintuitive: CBF and glycolysis increase more than seems to be needed to provide oxygen and pyruvate for oxidative metabolism in the mitochondria; the oxygen extraction fraction is relatively low in the brain and *decreases* when oxygen metabolism increases; and it appears that inhibitory neural activity is an important driver of CBF, even though such activity is likely to have less of an energy cost in terms of ATP consumption compared with excitatory activity. This work lays a foundation for the idea that this unexpected pattern of physiological changes is consistent with basic thermodynamic considerations related to metabolism. In the context of this thermodynamic framework, the apparent mismatches in metabolic rates and CBF are related to preserving the entropy change of oxidative metabolism, specifically the O_2/CO_2 ratio in the mitochondria. However, the mechanism supporting this CBF response is likely not due to feedback from a hypothetical O_2 sensor in tissue, but rather is consistent with feed-forward control by signals from both excitatory and inhibitory neural activity. Quantitative predictions of the thermodynamic framework, based on models of O_2 and CO_2 transport and possible neural drivers of CBF control, are in good agreement with a wide range of experimental data, including responses to neural activation, hypercapnia, hypoxia and high altitude acclimatization.

1. Introduction: The challenge of interpreting metabolism and blood flow dynamics in terms of the underlying neural activity

Current functional non-invasive neuroimaging methods such as functional magnetic resonance imaging (fMRI), positron emission tomography (PET), and near infrared spectroscopy (NIRS) do not measure neural activity directly, but instead are sensitive to metabolic and blood flow changes that accompany changes in neural activity. Consequently, understanding the links between metabolism, flow and neural activity is an active goal of current neuroimaging research. Intuitively, we expect strong connections because neural activity is energetically costly. Excitatory synaptic activity generates sodium currents from the extracellular to intracellular space [1, 2]. To recover from this activity, neurons must then pump sodium back out of the cell, against a steep thermodynamic gradient. This is accomplished by coupling sodium transport to the conversion of adenosine triphosphate (ATP) to adenosine diphosphate (ADP), a thermodynamically highly favorable reaction, within the sodium/potassium pump. ATP, the ubiquitous energy currency of the cell, must then itself be restored, which is accomplished through glucose metabolism, first by the conversion of glucose to pyruvate (Pyr) in the cytosol, and then oxidative metabolism of Pyr in the mitochondria with the generation of CO_2. Blood flow delivers the O_2 and glucose and carries away the CO_2.

The simplest scenario we might have imagined for this net process would be a serial chain of events with each triggering the next: recovery from neural activity depletes ATP, reduced ATP stimulates glycolysis and oxidative metabolism, and increased metabolism stimulates blood flow. In addition to this simple chain of driving mechanisms, we might have anticipated proportional changes of the cerebral metabolic rate of glucose (CMRGlc), the cerebral metabolic rate of oxygen ($CMRO_2$), and cerebral blood flow (CBF), all matched to the degree of neural activity change. Instead, though, current research suggests a much more complicated picture (**Figure 1, upper panel**), with several counterintuitive features, and it is the complexity of this process that currently presents a barrier to more quantitative interpretations of metabolism and flow dynamics in terms of the underlying neural dynamics.

The most striking departure from the simple picture described above is the apparent mismatch of the rates of different processes. The CMRGlc change with increasing neural activity is much larger than the $CMRO_2$ change [3], and some of the excess pyruvate created is converted to lactate. The production of lactate is surprising given that there appears to be no lack of available oxygen, and this phenomenon has been called 'aerobic glycolysis' [4]. From the perspective of increasing ATP production, the added glycolysis has little impact: for example, even a 50% increase of glycolysis alone would increase ATP production by only about 3%. In short, even with this added glycolysis most of the needed increase in ATP production is due to the increased oxidative metabolism, reflected in $CMRO_2$ [5, 6].

Blood flow also increases much more than $CMRO_2$ [7], so that oxygen delivery to tissue is increased much more than the actual increase in the rate at which it is metabolized. A natural idea is that the large blood flow increase might be necessary to support the large change in CMRGlc, even though the function served by the large change in glycolysis is still unclear. However, studies limiting the increase of CBF nevertheless found normal increases of CMRGlc [8, 9], indicating that the large change in CBF is not necessary to support CMRGlc.

The phenomenon of a much larger increase of CBF than $CMRO_2$ creates the seemingly paradoxical effect that the oxygen extraction fraction (OEF), the fraction of delivered oxygen that is extracted and metabolized, *decreases* with increased neural activity, and this effect is at the heart of the blood oxygenation level dependent (BOLD) changes of the measured fMRI signal [10]. The physical origin of the BOLD effect is that deoxyhemoglobin has paramagnetic properties that create magnetic field distortions and reduce the measured MR signal [11]. The reduction of OEF with increased neural activity then reduces those field distortions, creating a slight increase of the MR signal. Importantly, the magnitude of the BOLD signal depends on the degree of mismatch between CBF and $CMRO_2$, as well as the amount of deoxyhemoglobin present in the baseline state, creating a critical challenge to any quantitative interpretation of the BOLD

signal in neural terms [11, 12]. In short, the mismatch of CBF and $CMRO_2$ makes possible BOLD-fMRI, but what function is served by this mismatch, and under what circumstances might the degree of mismatch change?

The primary idea for understanding this mismatch is that a larger CBF increase than the $CMRO_2$ increase tends to maintain the tissue O_2 level [13-15]. As $CMRO_2$ increases, the gradient of O_2 concentration from blood to tissue must increase to support a higher diffusive flux of O_2 to the mitochondria (**Figure 1, lower panel**). In principle, this gradient could be increased by lowering the partial pressure of oxygen (PO_2) in tissue or by raising blood PO_2, and we made the early suggestion that tissue O_2 concentration is so low to begin with that the only option is to raise blood O_2 [5]. By reducing the OEF, the capillary PO_2 is increased, increasing the PO_2 gradient from blood to mitochondria. However, other studies indicate that the tissue O_2 level is reasonably high, about 25 mmHg [16, 17], and until recently the dominant view was that oxygen metabolism was not affected until the tissue PO_2 is reduced significantly below 1 mmHg [18]. Why does the brain maintain such a high tissue PO_2 when it seems to be unnecessary? More recently, though, the work of Wilson [19] has been a strong challenge to previously held views about how oxygen concentration limits tissue function. The basic finding was that the oxygen metabolic rate can be maintained down to very low O_2 concentrations, but the phosphorylation potential that governs energy metabolism in the cell begins to degrade at a much higher concentration, a partial pressure of about 12 mmHg [20]. This finding is a primary motivation for the thermodynamic framework developed in this paper.

Several other aspects of CBF also are puzzling. Even in the baseline state, OEF in grey matter is relatively low (~40%), so that the high baseline CBF is already delivering more oxygen than is needed, and OEF drops further with neural stimulation (to ~30% for a strong neural stimulus [21]). In contrast, heart muscle has a baseline OEF of about 70-80% that does not change much as O_2 metabolism increases [22], even though the rates of oxidative metabolism are roughly similar for the resting heart and grey matter. Why are the set points and dynamic behavior of blood flow in these two organs so different? Another blood flow effect is that CBF increases strongly with inhaled CO_2, a

phenomenon that has been recognized for more than a century [23], but it is not clear what function this serves. In terms of oxygen delivery, though, the effect is to reduce OEF (to about 30% on inhalation of a 5% CO_2 gas [24]).

Finally, the simple picture of serial mechanisms driving the metabolic and flow changes appears to be wrong, or at least a simple path like this is not the primary mechanism involved. Instead, the current picture is that aspects of neural activity drive CBF changes in a feed-forward way, essentially anticipating the upcoming need for increased metabolism, and this control is applied by a wide variety of mechanisms [25-27]. In this way CBF and $CMRO_2$ are driven in parallel by neural activity, and there is growing evidence that the aspects of neural activity that drive the CBF increase may not be the same aspects that entail the largest ATP cost and so account for most of the increased $CMRO_2$. As noted above, the primary ATP cost is in restoring ion gradients after neural signaling, primarily excitatory synaptic activity. In contrast, inhibitory synaptic activity often involves opening chloride channels, and because the intracellular/extracellular concentrations are near equilibrium with the resting membrane potential there is little ionic current. As a result, the recovery from inhibitory synaptic activity is likely to be less costly in terms of ATP than the recovery from excitatory activity. Although recovery from spiking of both excitatory and inhibitory neural populations also consumes ATP, estimates for the human brain are that excitatory synaptic activity dominates the overall ATP cost of neural signaling [2]. For this reason, we would expect excitatory activity to be a strong driver of CBF, and a number of studies support this [28]. Interestingly, though, some aspects of inhibitory activity also have a surprisingly strong effect on increasing CBF [29]. The release of nitric oxide (NO), a potent vasodilator, has been associated with the activity of inhibitory neurons [30], and adenosine, which often has an inhibitory neural effect, is also a strong vasodilator [31]. In a recent study using optogenetic methods to stimulate only inhibitory neurons the positive CBF response was approximately the same size as with a more natural stimulus [32]. Why has evolution favored a strong role of inhibitory neural activity to increase CBF?

This complicated picture of how different aspects of neural activity drive $CMRO_2$ and CBF suggests the possibility that the balance of changes in CBF and $CMRO_2$ may vary with the mix of underlying neural activity [12]. While this would create even more of a problem for any quantitative interpretation of the BOLD signal alone, it potentially opens the door for quantitative physiological fMRI, measuring the changes in CBF and $CMRO_2$ [33], to provide a deeper and more nuanced interpretation of the underlying neural activity. For this potential to be realized, though, we need a much better understanding of the connections between neural activity, metabolism, and blood flow.

The central paradox of neurovascular and neurometabolic coupling described above is that the mechanisms supplying the raw materials to support neuronal metabolism appear to be over-reactive; the delivery of oxygen and pyruvate to the mitochondria increases beyond the needs of oxidative metabolism. However, the delivery of these metabolic substrates alone may not be the limiting factors for neuronal metabolism. The goal of this paper is to lay a foundation for the theory that the puzzling aspects described above are consistent with thermodynamic limitations on oxygen metabolism [14]. **Section 2** outlines the basic theory, describing the general thermodynamic foundations and how the preservation of the entropy change associated with oxidative metabolism is consistent with the observed physiological behavior. A key result of this development is the importance of maintaining the tissue O_2/CO_2 concentration ratio, primarily by modulating CBF. **Section 3** illustrates the implications of preserving tissue O_2/CO_2 with modeling studies based on a model of gas transport in blood and tissue, and a simple neural dynamics model of feed-forward drivers of CBF and $CMRO_2$, to develop predictions of the theory and compare the predictions with experimental data on CBF responses to neural activity, hypercapnia and hypoxia. Also, in considering high altitude acclimatization another physiological factor comes into play in preserving the tissue O_2/CO_2 ratio: the increased ventilation rate that lowers blood CO_2.

2. Proposed thermodynamic framework

(a) The role of entropy change

The central argument of the thermodynamic framework is that the apparent mismatches of the changes in CMRGlc and CBF compared to $CMRO_2$, as well as the CBF response to inhaled CO_2, can all be viewed as preserving the entropy change of oxidative metabolism in the mitochondria, and through that the oxidative metabolic rate. Entropy is directly related to the number of different molecular states—defined by the positions and velocities of all the particles—which are consistent with given macroscopic constraints. For chemical transformations, such as a chemical reaction or transport of an ion across a cellular membrane, the macroscopic constraint is the average concentrations of different molecules. The basic principles underlying the current theory are that for any transformation of such a system: 1) the net entropy change ΔS of all the processes involved in the transformation cannot be negative (Second Law of Thermodynamics); 2) the net entropy change ΔS has a simple mathematical form that depends on the concentrations of the molecules involved; and 3) the steady-state rate of the chemical transformation depends on both kinetic factors related to the process and on the entropy change ΔS, with that rate going to zero as ΔS goes to zero. All three of these effects follow in a general way from the basic equations of motion of the molecules, specifically that as the molecular states evolve over time they remain as distinct states and do not converge (this is developed in more detail in the *Supplementary Information A: Thermodynamic Basis* based on work in the late twentieth century by E.T. Jaynes [34] and C.H. Bennett [35], and on the implications of the Fluctuation Theorem [36]).

Applying these ideas, for any chemical transformation in a cell the net entropy change cannot be negative. Cellular work involves processes with a negative entropy change, and for these processes to take place they must be coupled to another transformation with a positive entropy change with a larger magnitude. For many cellular processes, the coupled process with a positive entropy change is the breakdown of ATP to ADP and Pi (inorganic phosphate). Another important source of a positive entropy change is the movement of a sodium ion down its electrochemical gradient across the cellular membrane from outside to inside, often used for co-transport of other molecules

(e.g., clearance of glutamate from the synaptic cleft by uptake into astrocytes). The sodium gradient is then restored by the sodium/potassium pump, by coupling sodium transport to ATP consumption, and in the brain the activity of the sodium/potassium pump accounts for most of the ATP consumed [37]. The ATP is then restored, involving a negative entropy change, by coupling the process to oxidative metabolism, providing a stronger positive entropy change.

(b) Key relationships from thermodynamics

The Supplementary Information contains an extended derivation and discussion of the thermodynamic ideas underlying the current theory, leading to the two principal results whose implications are developed below.

Entropy change of a chemical process. The first thermodynamic result is the general form of the entropy change ΔS for a chemical transformation. Suppose that a chemical transformation involves the recombination of several reactants (R_1, R_2, …) to form several products (P_1, P_2,…), and consider one minimal instance of this molecular transformation (e.g., one molecule of R_1 plus one molecule of R_2) so that the overall concentrations are not significantly changed. The entropy change associated with this minimal transformation is:

$$\Delta S = k_B \ln \frac{\Phi}{\Phi_0} \qquad [1]$$

where k_B is Boltzmann's constant and the parameter **Φ** is the ratio of the reactants to the products of the chemical transformation:

$$\Phi = \frac{[R_1][R_2]\ldots}{[P_1][P_2]\ldots} \qquad [2]$$

The parameter Φ_0 is the equilibrium value of Φ such that the chemical transformation involves no change of entropy: $\Delta S=0$. The value of Φ_0 depends on the specific context in which the chemical transformation occurs, including temperature, the energy change of the chemical system related to the different net binding energies of the reactants and products, environmental interactions such as those between the reactants and products with the surrounding water molecules or with pH, and the conditions under which the chemical transformation takes place (e.g., constant volume or constant pressure). In general, it is difficult to precisely quantify the value of Φ_0 in a biological setting, but the important result for the current theory is simply the mathematical form of Eq [1].

Implications: net entropy change for a linked series of chemical transformations. If a net process contains several steps, such as oxidative metabolism, the net entropy change is the sum of the entropy changes for each step, each of the form of Eq [1]. Mathematically, the net entropy change will then depend on the products of the Φ terms for each step. If the linked steps involve an intermediate chemical that is a product of one step that is consumed as a reactant by the second step, the concentration of that intermediate drops out of the expression for the net entropy change. That is, the concentration of an intermediate will appear in the denominator of Φ_1 and the numerator of Φ_2, and the net entropy change will depend on $\Phi_1\Phi_2$. As a result, for an extended process of chemical transformations, the net entropy change depends only on the concentrations of molecules changed by the net process. However, the concentration of the intermediate can play a useful role in balancing the separate entropy changes of the sequential steps (e.g., increasing the concentration of the intermediate will reduce the entropy change of the first step but increase the entropy change of the second step, without altering the net entropy change), and this idea is further discussed in *Supplementary Information A: Thermodynamic Basis*.

Rate of a chemical process. The second important relationship, following from the Fluctuation Theorem (derived in *Supplementary Information A: Thermodynamic Basis*), is that the net rate of a chemical process can be expressed as:

$$R = R_0 \left[1 - \exp\left(-\frac{\Delta S}{k_B}\right)\right]$$

[3]

In this form, the rate of a process has a kinetic term R_0 and a thermodynamic term depending on the entropy change ΔS. We refer to the parameter R_0 as the kinetic rate, and it may depend on the reactant concentrations in a simple way for first order kinetics, or in principle could be independent of the reactant concentrations and controlled by enzyme kinetics (see below). The rate R_0 is the rate a process will have if the entropy change is large, and as ΔS goes to zero, the net rate of the process also goes to zero.

Implications: Kinetic and thermodynamic effects on the rate of a chemical process. Eq [3] suggests two ways in which the rate of a chemical process could be modified: a kinetic mechanism, changing R_0; or a thermodynamic mechanism, changing ΔS. This distinction is important both for considering how a chemical process can be controlled, and also for considering how a reduction of the concentration of a reactant, such as a reduction of the O_2 concentration in the mitochondria, can affect the rate of the process through both a kinetic and a thermodynamic limitation. In general, the kinetic term R_0 for a process will depend on the enzyme kinetics affecting the mechanics of the process. For example, with Michaelis-Menten enzyme kinetics R_0 has the form:

$$R_0 = \frac{v_{max}}{1 + {K_m}/{C}}$$

[4]

where C is the concentration of the reactant, and v_{max} and K_m are parameters describing the enzyme kinetics. If $C << K_m$ the kinetics become first order, and R_0 is proportional to C. However, for $C >> K_m$ the kinetic rate R_0 is $v_{\max}$, independent of C. In this case, the rate could be controlled by modulating the enzyme kinetics (i.e., modulating v_{max}), with no change in C. However, the net rate of the process is fully determined by those enzyme

kinetics only when the entropy change ΔS is large and positive, and as *C* is reduced ΔS also is reduced by Eq [1]. In principle, the reduction of *C* could begin to degrade the rate of the process by a thermodynamic limitation (reduced ΔS) even though the process is not kinetically limited because *C* is still larger than K_m.

(c) Oxidative metabolism in the brain

The oxidative metabolism of Pyr and generation of ATP is a complicated extended process consisting of a series of chemical transformations in which some of the products of one process are the reactants for the next process: 1) in the mitochondria, the TCA cycle metabolizes Pyr, coupled to the conversion of NAD^+ to NADH and the production of CO_2; 2) the NADH contributes electrons to the electron transfer chain, which are eventually transferred to O_2 to produce water, and this electron transfer is coupled to the transport of hydrogen ions across the inner membrane of the mitochondria, creating a proton/potential gradient; and 3) movement of protons down the proton gradient is coupled to conversion of ADP plus Pi to ATP.

For the entropy changes involved, we can consider this extended process as two net processes based on the molecules that are changed (consumed or produced). From Eq [1], the net entropy change associated with the consumption of Pyr and O_2 and the production of CO_2 depends on the ratio:

$$\Phi_{OX} = \frac{[Pyr][O_2]^3}{[CO_2]^3} \qquad [5]$$

This positive entropy change from oxidative metabolism of pyruvate must be larger than the negative entropy change associated with the conversion $ADP+P_i \rightarrow ATP$, determined by the phosphorylation potential:

$$\Phi_{ATP} = \frac{[ATP]}{[ADP][P_i]}$$

[6]

If the O_2 concentration begins to fall (e.g., by decreased O_2 delivery), Φ_{OX} will be reduced and the net entropy change ΔS will be reduced. The net positive entropy change can be maintained as Φ_{OX} falls, and $CMRO_2$ preserved, if Φ_{ATP} falls as well, as in the study of Wilson and colleagues [20]. In this scenario the metabolic rate is maintained, but at the expense of the entropy change available from ATP to drive cellular work. In the context of the proposed thermodynamic framework, this is interpreted as a distinction between kinetic and thermodynamic limits on metabolism. In this interpretation, the kinetic limit is not reached until $[O_2]$ drops to a very low level, but the thermodynamic limit is reached at higher O_2 levels, marked by when Φ_{ATP} begins to degrade to maintain the metabolic rate.

(d) Preserving the entropy change of oxidative metabolism and the importance of the tissue O_2/CO_2 ratio

The core of the proposed thermodynamic framework is that the physiological changes described in **Section 1** serve to preserve Φ_{OX} (Eq [5]) and support the metabolic rate of oxygen. Preserving Φ_{OX} can be done by maintaining the tissue concentrations of O_2 and Pyr, and the physiological challenge of doing this is the need for transport of these molecules to the mitochondria by diffusion (**Figure 1, lower panel**). Unlike enzymatically controlled processes, where in principle the rate of the process can be controlled by modulation of the enzyme activity with no change in the concentration of metabolic substrates, transport by diffusion necessarily involves concentration gradients. In order to increase the flux of O_2 to the mitochondria while preserving the mitochondrial O_2 concentration, the gradient can be increased by raising the O_2 concentration in blood, and that requires reducing the OEF by increasing CBF more than $CMRO_2$. Similarly, to increase the pyruvate flux from the cytosol to the mitochondria, while preserving the mitochondrial pyruvate concentration, the pyruvate gradient can be increased by increasing CMRGlc to increase the cytosolic pyruvate concentration. In this context, the key roles of glycolysis and blood flow are their effects in modulating tissue

concentrations of pyruvate and oxygen, respectively, to increase diffusion gradients while maintaining concentrations in the mitochondria. In short, even though empirically it appears that $CMRO_2$ may be a lesser player compared with CMRGlc and CBF, because the two latter processes are much more responsive to increased neural activity, within the proposed thermodynamic framework $CMRO_2$ is the key function that shapes the complex physiology.

Based on the thermodynamic arguments it is not necessary to preserve the individual concentrations that make up Φ_{OX}, but just to preserve Φ_{OX} itself. The role of glycolysis as the first metabolic step, modulating the Pyr concentration, is discussed in detail in the Supplementary Information. In **Section 3(a-f)** we focus on the implications of maintaining Φ_{OX} by maintaining the tissue O_2/CO_2 ratio under different conditions, including increased $CMRO_2$, hypercapnia, and hypoxia.

(e) Neural control of blood flow

Given this picture, with the central role of CBF being to maintain the tissue O_2/CO_2 ratio, we might expect that CBF would be strongly driven by an oxygen sensor in tissue, but there is currently no known mechanism that could operate in this way. In addition, another factor is critical: O_2 dissolves so poorly in water that there is very little O_2 in tissue to serve as a buffer against a drop in O_2 concentration. The amount of O_2 dissolved in tissue would support grey matter $CMRO_2$ for only about one second [14]. Given the lack of an O_2 sensor and the lack of a buffer of O_2, a feed-forward system with a CBF increase triggered by aspects of neural activity could increase O_2 delivery, and specifically the O_2 gradient driving O_2 flux to the mitochondria, in anticipation of the upcoming metabolic need and would help prevent a drop in $[O_2]$ and thus a drop in Φ_{OX}. To this end, all aspects of neural activity could potentially be beneficial as drivers of CBF, and empirically many mechanisms linking neural activity to CBF have been found. In addition, though, inhibitory activity may be a particularly useful predictor of upcoming metabolic needs, if the general pattern of increasing neural input to a region is a tapering increase of excitatory activity as the inhibitory activity continues to increase. The

inhibitory activity then becomes a warning signal for a larger activity change, and the larger $CMRO_2$ costs that will be incurred in recovery. More quantitatively, in combination with excitatory activity driving a CBF change the added contribution from inhibitory activity would help produce the nonlinear CBF increase needed to maintain the tissue O_2 level in the face of a larger change in $CMRO_2$. In **Section 3(g)**, a simple neural network model is used to show the feasibility of this idea.

3. Modeling Results and Discussion

The modeling studies in this section focus specifically on how blood flow could serve to preserve Φ_{OX} by maintaining the tissue O_2/CO_2 ratio. A quantitative model for O_2 and CO_2 transport in blood and exchange with tissue (described in detail in *Supplementary Information B: Oxygen Transport Model*) was used to develop predictions based on the thermodynamic framework, and compare the predictions to experimental results. Starting from an assumed normal baseline state, the model predicts the change in tissue O_2 and CO_2 concentrations as CBF, $CMRO_2$, and arterial blood gases are modulated.

(a) Neural activation CBF response

When $CMRO_2$ increases, the gradient of O_2 concentration between blood and tissue must increase, and this can be accomplished by increasing CBF more than $CMRO_2$ so that the O_2 concentration in blood rises (**Figure 1, lower panel**). Based on the transport model, the CBF required to maintain the tissue O_2/CO_2 ratio as $CMRO_2$ increases is shown in **Figure 2A**. The CBF/$CMRO_2$ coupling ratio n, defined as the fractional change in CBF divided by the fractional change in $CMRO_2$, varies from about 2 to about 3 as $CMRO_2$ increases, in good agreement with experimental studies (reviewed in [14]). Note that the large increase of CBF required is due to the relatively low baseline OEF (40%), and if instead baseline OEF was larger, less of a CBF change would be needed to preserve tissue O_2/CO_2.

(b) Hypercapnia CBF response

Increased CBF, as a mechanism to raise tissue [O_2], also may explain the function served by the strong sensitivity of CBF to hypercapnia. Breathing a gas with increased CO_2 content will increase arterial and tissue CO_2 levels, and increased CBF can raise tissue [O_2] to preserve the tissue O_2/CO_2 ratio. **Figure 2B** shows the CBF change needed to maintain tissue O_2/CO_2 when the partial pressure of CO_2 in arterial blood ($PaCO_2$) is raised by inhaling a gas mixture with elevated CO_2. Experimental measurements of the CBF response, expressed as % change per mmHg change in $PaCO_2$, vary widely, and the model prediction is consistent with the low end of the experimental results [38, 39].

(c) Limiting O_2/CO_2 values in tissue

To estimate the limiting O_2/CO_2 ratio when the ATP phosphorylation potential Φ_{ATP} begins to degrade, and test for consistency with Wilson's earlier studies [19] finding impairment when tissue PO_2 drops to about 12 mmHg, we used the transport model to analyze the extensive data reported by Nioka et al [40] in a study of different stages of hypoxia in a canine model. Importantly, they measured changes in Φ_{ATP} as well as CBF, $CMRO_2$ and blood gases (**Figure 3A**). When analyzed with the transport model, Φ_{ATP} begins to degrade when the tissue O_2/CO_2 ratio is reduced to about 50% of the baseline value. **Figure 3B** shows the modeled tissue PO_2 when Φ_{ATP} begins to degrade, in good agreement with Wilson's limit of about 12 mmHg.

(d) Sensitivity to reduced oxygen delivery

An important consequence of the thermodynamic framework, emphasizing tissue O_2/CO_2, is the prediction that the sensitivity of the brain to reduced delivery of O_2 will depend on exactly how that delivery is reduced, and not just on the degree of reduction itself. That is, O_2 delivery depends on both blood flow and the O_2 content of arterial blood, and the impairment of the tissue O_2/CO_2 ratio depends on which of these factors is reduced. The model results (**Figure 4A**) are that the reduction of tissue O_2/CO_2 is more

severe in the face of reduced O_2 delivery when the arterial PO_2 is reduced than when CBF is reduced. Importantly, when blood PO_2 is reduced and CBF is increased to restore O_2 delivery to the baseline level, this is not sufficient to restore the tissue O_2/CO_2 ratio (dotted curve in **Figure 4A**).

(e) High altitude acclimatization

Long-term exposure to high altitude involves a number of physiological changes, some of which, such as increased hematocrit, serve to restore O_2 delivery. In addition, though, a key response is an increased ventilation rate that lowers arterial CO_2, and this response helps to maintain tissue O_2/CO_2. For the calculations for the acclimatized state, as arterial PO_2 was lowered, hematocrit was increased so that O_2 delivery remained at the baseline level, and there was no change in CBF or $CMRO_2$. As in **Figure 4A**, though, maintaining O_2 delivery, here by increasing the carrying capacity of blood, was not sufficient to preserve tissue O_2/CO_2. The model was then used to calculate the degree of reduction of arterial PCO_2 needed in addition to restore tissue O_2/CO_2 (**Figure 4B**). Also plotted are the data reported in the classic study of Rahn and Otis [41], reporting alveolar PO_2 and PCO_2 for subjects acutely exposed to simulated altitude in a pressure chamber along with values reported from several studies of acclimatized individuals, showing the effect of increased ventilation rate with acclimatization. The model curve is consistent with the experimental data for acclimatized subjects, with the acute subjects falling above the model curve, corresponding to reduced tissue O_2/CO_2 if there is no increase of CBF. This too is consistent with other experiments finding increased CBF on acute exposure to high altitude that resolves back to baseline CBF over about a week as subjects acclimatize [42].

(f) Consequences of the low baseline OEF in the brain

As noted above, a typical OEF in brain is about 40%, and yet for the heart muscle the OEF is typically much higher. At first glance, one might assume that the lower OEF in brain allows for a greater dynamic range, in some sense a larger buffer of unused

capacity. However, the dynamic range of O_2 metabolism in the heart is several times larger than in the brain, and with increased $CMRO_2$ the OEF decreases, rather than increases. In **Figure 2A**, the requirements for matching blood flow to increased O_2 metabolism are less severe when the baseline OEF is larger, so that a given range of blood flow can support a larger range of O_2 metabolism (as in the heart). Where the lower OEF in brain does confer an advantage, though, is in ischemia, and it is here that the "unused capacity" can come into play to preserve a reduced level of $CMRO_2$ with increased OEF. For example, calculations with the transport model considered the question of how much of the baseline O_2 metabolic rate can be maintained as CBF deceases so that the tissue O_2/CO_2 level is above 70% of baseline, which from **Figure 3A** is a level where the phosphorylation potential is still maintained. At this reduced level, and for a normal baseline OEF of 40%, $CMRO_2$ can be as high as ~65% of normal when CBF is reduced to about ~30% of normal. For a baseline OEF of 60%, though, at this reduced level the maximum $CMRO_2$ is reduced to ~50% of baseline. Note also that increased OEF by itself is not necessarily a sign of critical impairment: for this example with the normal baseline OEF of 40%, that level of maintained $CMRO_2$ is achieved with OEF rising to about 60%. These modeling results suggest that a full evaluation of stroke conditions should involve measurements of both CBF and $CMRO_2$, as a more modest reduction of $CMRO_2$ than the CBF reduction may be sufficient to maintain tissue O_2/CO_2.

(g) Neural control of CBF

Although the central argument here is that the function served by a large CBF change is to maintain tissue PO_2, there is not yet a known mechanism—an O_2 sensor—that could be the basis of a feed-back system [14]. Instead, a growing body of evidence points to neural activity driving the rapid control of CBF in a feed-forward way. As a first step in linking neural activity to CBF and $CMRO_2$, a model was developed to interpret the experimental data we reported in [21] measuring BOLD and CBF responses in humans as the contrast of a visual stimulus was increased (**Figure 5**). The basic experimental result was that as the stimulus contrast was increased the BOLD signal

increased more than CBF, which was consistent with a gradual plateauing of the $CMRO_2$ response as CBF continued to increase with increasing contrast. Interestingly, a parallel electrophysiological pattern was previously observed in animal studies, in which inhibitory neuronal activity continued to increase while excitatory neuronal activity began to plateau [43]. Given the high entropy cost of excitatory activity, due to the associated sodium currents and required sodium transport in recovery, this led us to propose the qualitative idea that $CMRO_2$ is largely driven by excitatory neuronal activity while CBF is driven by both excitatory and inhibitory activity [12].

Here we implemented a simple quantitative model as an initial test of the feasibility of this idea with a version of the Wilson-Cowan model [44] describing the interaction of excitatory (E) and inhibitory (I) populations of neurons. The model is described in detail in *Supplementary Information C: Neural Model*, but the general structure is illustrated in **Figure 5A**, and the behavior of the E and I populations with increasing input is shown in **Figure 5B.** Assuming that $CMRO_2$ is primarily due to E activity and CBF is driven by a combination of E and I activities can create a CBF/$CMRO_2$ response curve that is a reasonably good approximation to the curve calculated to be necessary to maintain the tissue O_2/CO_2 ratio (**Figure 5C**). The curve is not identical, though, with slight differences in the balance of CBF and $CMRO_2$, although these differences have little impact on the general preservation of the entropy change. Interestingly, though, because the BOLD signal is sensitive to the exact balance of CBF and $CMRO_2$ changes, when the BOLD signal is calculated for each curve (**Figure 5D**) the WC model gives a much better fit to the experimental data. We should be cautious about over-interpreting this result—it is one experimental data set and one implementation of the model. Nevertheless it supports the feasibility of modeling the full connections between neural activity and the BOLD signal in a way that is consistent with the proposed thermodynamic framework.

4. Conclusions

The goal of this work was to lay a foundation for a thermodynamic framework for understanding a number of aspects of blood flow and metabolism in the brain that have emerged from experimental studies over the last several decades. Within this framework, blood flow serves to modulate tissue O_2 to maintain the O_2/CO_2 ratio at the mitochondria. This, in turn, serves to maintain the entropy change of oxidative metabolism and consequently maintain the metabolic rate. However, due to the low solubility of O_2 in water, and to the lack of an appropriate O_2 sensor, blood flow is likely driven in a feed-forward way by neural activity rather than a feedback mechanism from O_2. That is, the thermodynamic framework described here addresses the question of *why* the physiology responds as it does in terms of the function served, but does not address the question of *how* this is accomplished—the specific mechanisms that drive CBF. As a first test of modeling a neural feed-forward connection, a simple Wilson-Cowan model is sufficient to explain experimental data for increasing stimulus amplitude. The thermodynamic framework also makes predictions related to hypoxia that are in good agreement with experimental data. More work is needed, though, to understand the mechanisms involved and the full implications for high altitude acclimatization, stroke, and other instances of hypoxia.

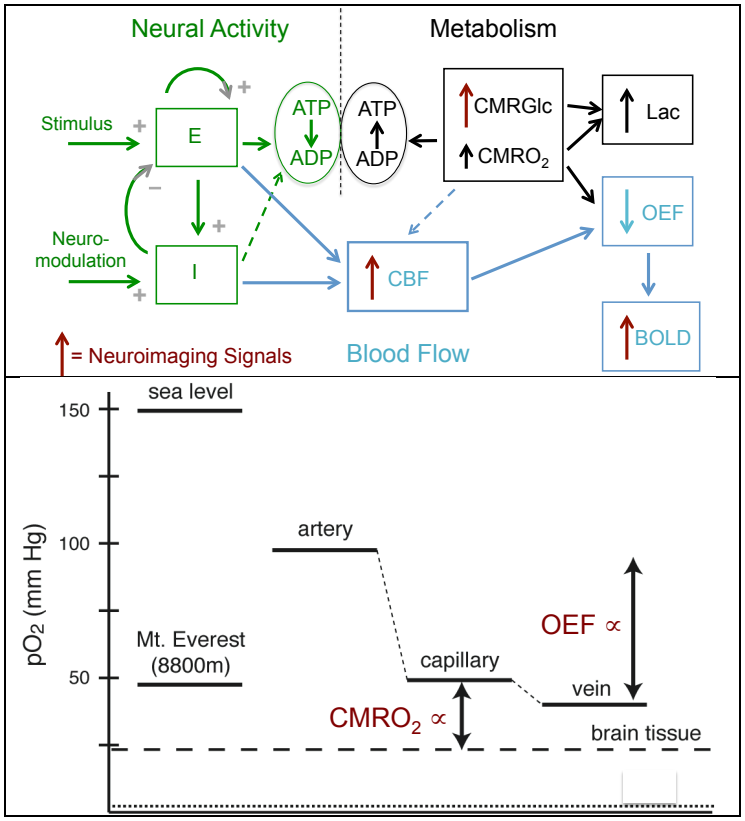

Figure 1. *Schematic of the connections between neural activity, metabolism, blood flow and neuroimaging signals*. **Upper Panel:** Input stimulus and neuromodulatory signals evoke activity of interacting excitatory (E) and inhibitory (I) neural populations (left, green), and ATP is consumed in recovery from that activity, primarily in restoring ion gradients. The ATP is restored through oxidative metabolism of glucose (right, black), with oxygen and glucose delivered by blood flow (middle, blue; CMRGlc = cerebral metabolic rate of glucose (glycolysis), $CMRO_2$ = cerebral metabolic rate of O_2, CBF = cerebral blood flow). Solid/dashed connecting arrows reflect strong/weak drivers, respectively, and the vertical arrows reflect the fractional increase of each rate. The larger increase of glycolysis than $CMRO_2$ leads to increased lactate production (Lac), and the larger increase of CBF than $CMRO_2$ leads to decreased oxygen extraction fraction (OEF), which produces the blood oxygenation level dependent (BOLD) signal measured in fMRI. The sources for current neuroimaging signals (fMRI and PET) are indicated by red arrows. The seemingly paradoxical features addressed in this paper are: the mismatch of CMRGlc and $CMRO_2$, leading to increased Lac production despite the availability of O_2; the mismatch of CBF and $CMRO_2$, leading to the decreased OEF; and the presence of a strong feed-forward drive from the inhibitory neural population (I) to CBF, even though the I activity is likely to be only a small contribution to the ATP costs and thus $CMRO_2$. **Lower Panel:** Diagram of O_2 partial pressures, with atmospheric values on the left for comparison and physiological values on the right. To increase $CMRO_2$, the diffusion gradient of PO_2 between the capillary blood and the mitochondria (tissue) must increase.

A consequence of increasing CBF more than $CMRO_2$ is that the oxygen extraction fraction (OEF) is reduced, raising capillary PO_2. In this way the PO_2 gradient can be increased, supporting increased $CMRO_2$, while maintaining the tissue PO_2 at a constant level. The larger increase of CMRGlc than $CMRO_2$ may serve an analogous function by increasing cytosolic Pyr to increase the diffusion gradient to the mitochondria without reducing mitochondrial Pyr. In the context of the proposed thermodynamic framework, the large increases of CBF and CMRGlc serve the broader role of maintaining the entropy change of oxidative metabolism. For CBF, a key element is that it controls the tissue O_2/CO_2 level, and this has additional implications for the CBF response to hypercapnia and hypoxia.

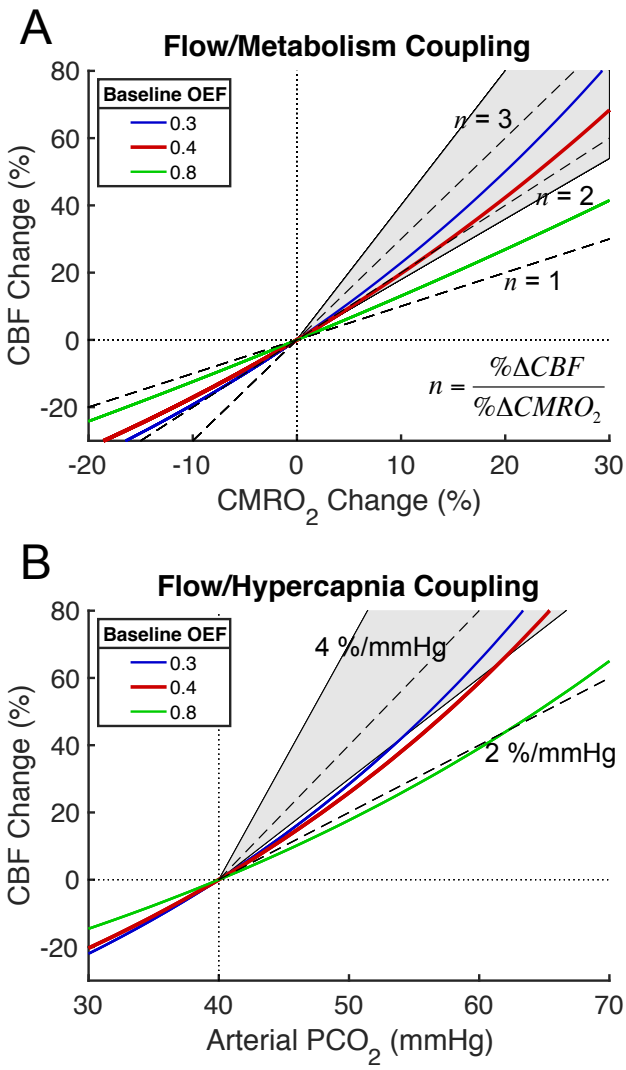

Figure 2. *Modeled CBF responses needed to preserve the $[O_2]/[CO_2]$ ratio in tissue in response to increased $CMRO_2$ and hypercapnia.* **A)** With increasing O_2 metabolism, the required CBF change is shown for three different values of the baseline OEF (the red line is the typical value for brain). Dashed lines show constant coupling ratios n of the fractional CBF change to the fractional $CMRO_2$ change. The shaded area between n=1.8 and n=4 shows the approximate range of experimental results. **B)** The required CBF change for increasing arterial CO_2 shown for the same values of baseline OEF, with dashed lines indicating constant slope values, and the shaded area between slopes of 3 and 7 %/mmHg approximates the range of reported experimental values.

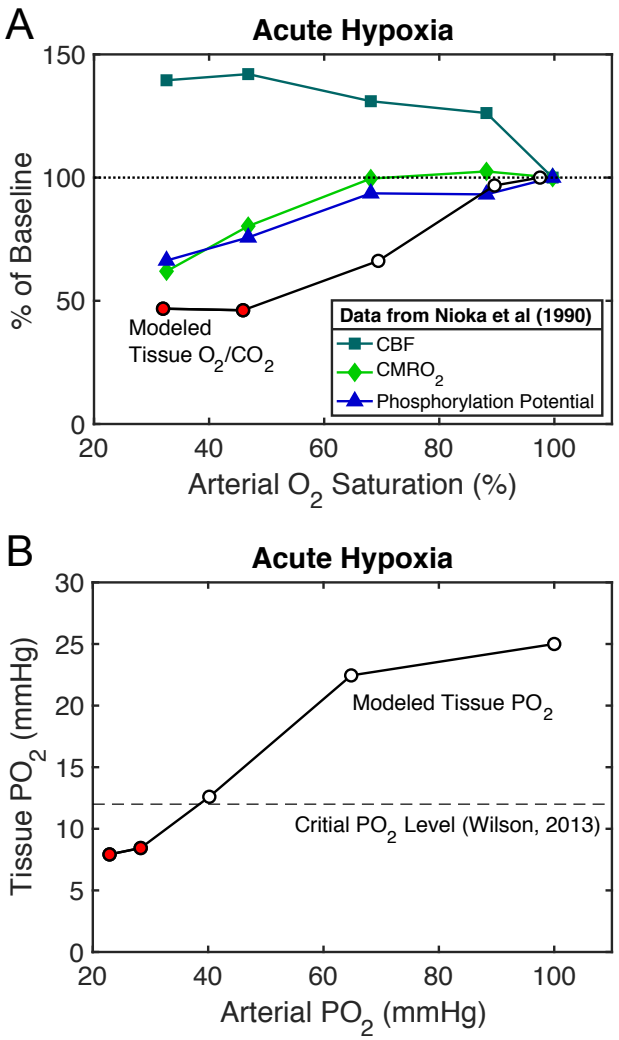

Figure 3. *Modeling experimental data to determine the critical level of tissue O_2/CO_2.* **A**) Data from Nioka et al (1990) for several stages of hypoxic hypoxia in a canine model as arterial O_2-hemoglobin saturation was reduced, including measurement of the ATP phosphorylation potential, are plotted as a percentage of the normoxic value. The transport model was applied to these data to calculate the O_2/CO_2 ratio in tissue (circles). The ATP phosphorylation potential was degraded for the two most extreme hypoxic conditions, and for these states the tissue O_2/CO_2 ratio was reduced to about 50% of the normoxic baseline value (red circles). **B**) For the same model, the calculated tissue PO_2 values are plotted as a function of the arterial PO_2 values. Based on a number of studies of mitochondria, Wilson (2013) concluded that a PO_2 of about 12 mmHg was a critical threshold below which the ATP phosphorylation potential would degrade (dashed line), in good agreement with the data of Nioka et al analyzed with the transport model.

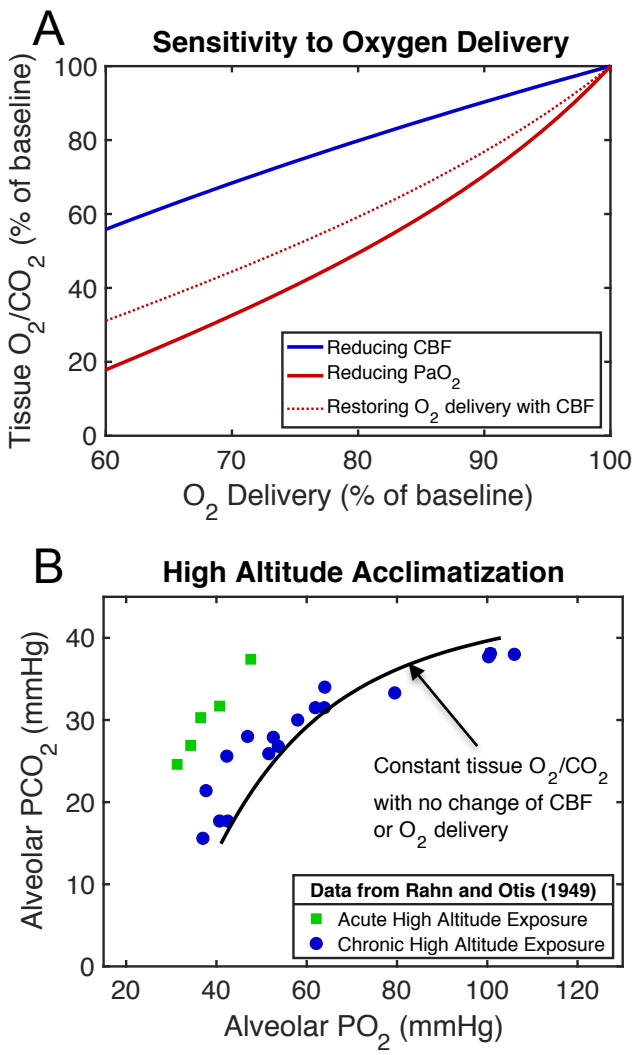

Figure 4. *Implications of the thermodynamic framework for sensitivity to oxygen delivery*. **A**) Curves for the reduction of tissue O_2/CO_2 for two ways of reducing O_2 delivery: by reducing CBF (blue) or by reducing arterial PO_2 (red). The brain is more sensitive to reduced O_2 delivery due to reduced blood PO_2. Importantly, for the case of reduced arterial PO_2, restoring baseline O_2 delivery by increasing CBF does not restore tissue O_2/CO_2 (red dotted curve). **B**) Classic data from Rahn and Otis (1949) showing the effect of increased ventilation rate as a key element of successful acclimatization to the low inspired O_2 concentration at high altitude, reducing arterial CO_2 ($PaCO_2$) and increasing arterial O_2 (PaO_2), creating a shift diagonally to the right. The solid model curve shows the value of $PaCO_2$ needed to compensate for the reduced PaO_2 and maintain the tissue O_2/CO_2 ratio with no change in $CMRO_2$ or CBF from the normoxic baseline, and constant O_2 delivery. (An alveolar/arterial PO_2 difference of 3 mmHg was assumed in making the plot.)

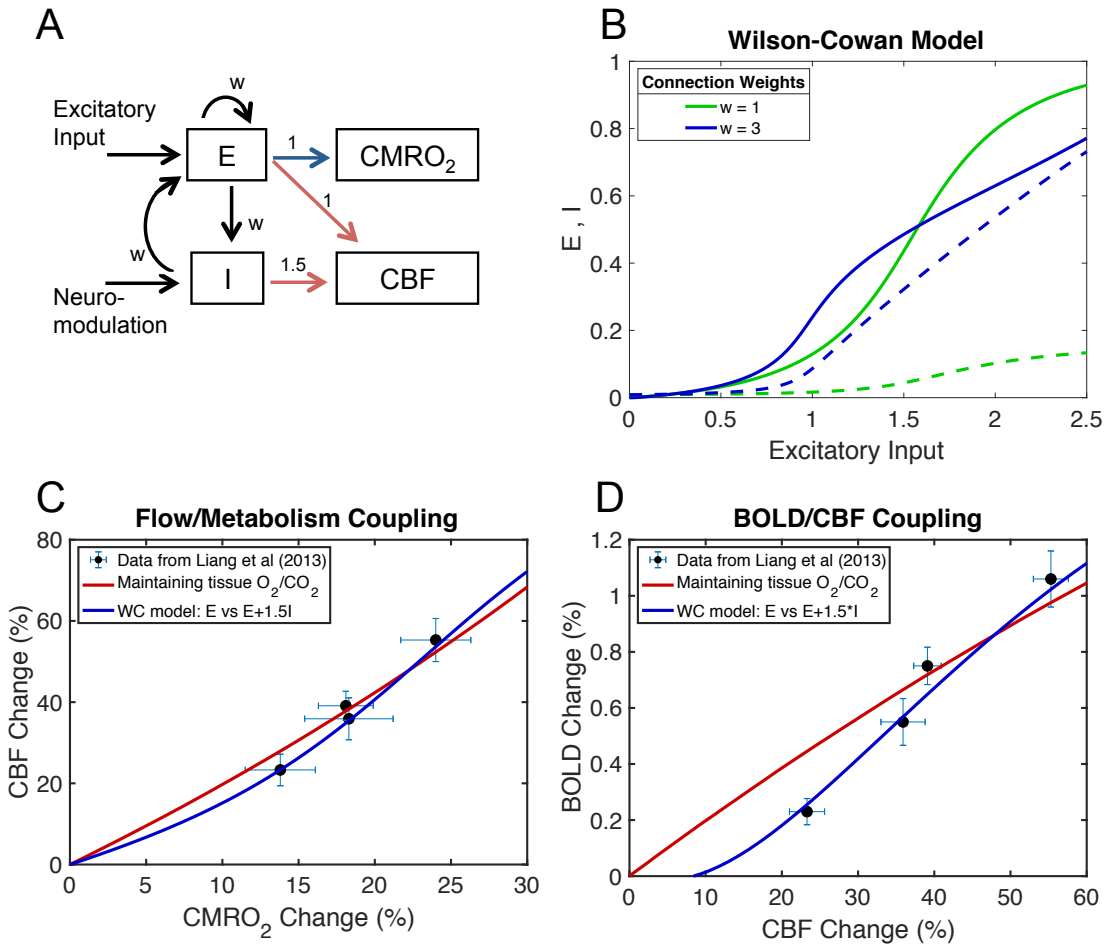

Figure 5. *A speculative mechanism for control of CBF through feed-forward signals from neural activity.* **A**) The Wilson-Cowan (WC) model was used to model in a simple way the dynamics of interactions of excitatory (E) and inhibitory (I) neural populations driven by an increasing external excitatory input to the E population. The weights for the different interactions were taken to be the same value *w*. As a proof of concept, $CMRO_2$ was assumed to be driven by E (blue arrow) and CBF was assumed to be driven by both E and I (red arrows), with a higher weighting for inhibitory activity. **B**) Curves of the activities of the E and I populations as the input increases are shown for *w*=1 and *w*=3. For higher *w* the E population is more sensitive to weaker input values, with activity rising faster due to the increased self-excitation of the E population, and at higher input values E rises more slowly due to a strong steady rise in I activity. **C**) Data from Liang et al (2013) showing the estimated $CMRO_2$ changes and the measured CBF changes (mean +/- SEM) from a calibrated BOLD study of effects of increasing contrast of a visual stimulus in humans. The red curve is the curve calculated to preserve tissue O_2/CO_2 (from **Figure 2A**). The blue curve is calculated from the WC model with *w*=3 with the assumption that $\%\Delta CMRO_2 \sim E$ and $\%\Delta CBF \sim E+1.5I$, with the same proportionality constant for both chosen to make the largest input value for the curves in panel **B** correspond to $\%\Delta CMRO_2 = 30\%$. **D**) Measured BOLD and CBF responses from Liang et al (2013) and the predicted BOLD curves for the constant O_2/CO_2 model (red) and the

WC model (blue). The WC feed-forward model gives a reasonably good approximation of the CBF/$CMRO_2$ coupling needed to preserve tissue O_2/CO_2 (panel **C**), but the BOLD sensitivity to subtle differences in the balance of CBF and $CMRO_2$ creates some divergence of the predictions of the BOLD signal for the two models, with the WC model providing a better fit to the experimental data (panel **D**).

References

1. Attwell, D. and Laughlin, S.B. (2001) An energy budget for signaling in the grey matter of the brain. J Cereb Blood Flow Metab 21 (10), 1133-45.
2. Attwell, D. and Iadecola, C. (2002) The neural basis of functional brain imaging signals. Trends Neurosci 25 (12), 621-5.
3. Fox, P.T. et al. (1988) Nonoxidative glucose consumption during focal physiologic neural activity. Science 241, 462-464.
4. Raichle, M.E. and Mintun, M.A. (2006) Brain work and brain imaging. Annu Rev Neurosci 29, 449-76.
5. Buxton, R.B. and Frank, L.R. (1997) A model for the coupling between cerebral blood flow and oxygen metabolism during neural stimulation. J Cereb Blood Flow Metab 17 (1), 64-72.
6. Lin, A.L. et al. (2010) Nonlinear coupling between cerebral blood flow, oxygen consumption, and ATP production in human visual cortex. Proc Natl Acad Sci U S A 107 (18), 8446-51.
7. Fox, P.T. and Raichle, M.E. (1986) Focal physiological uncoupling of cerebral blood flow and oxidative metabolism during somatosensory stimulation in human subjects. Proc. Natl. Acad. Sci. USA 83, 1140-1144.
8. Powers, W.J. et al. (1996) Hypoglycemia. Am. J. Physiol. 270, H554-H559.
9. Cholet, N. et al. (1997) Local uncoupling of the cerebrovascular and metabolic responses to somatosensory stimulation after neuronal nitric oxide synthase inhibition. J. Cereb. Blood Flow and Metabol. 17, 1191-1201.
10. Kwong, K.K. et al. (1992) Dynamic magnetic resonance imaging of human brain activity during primary sensory stimulation. Proc Natl Acad Sci U S A 89 (12), 5675-9.
11. Buxton, R.B. (2013) The physics of functional magnetic resonance imaging (fMRI). Rep Prog Phys 76 (9), 096601.
12. Buxton, R.B. et al. (2014) Variability of the coupling of blood flow and oxygen metabolism responses in the brain: a problem for interpreting BOLD studies but potentially a new window on the underlying neural activity. Front Neurosci 8, 139.

13. Buxton, R.B. (2009) Introduction to Functional Magnetic Resonance Imaging: Principles and Techniques, Cambridge University Press.
14. Buxton, R.B. (2010) Interpreting oxygenation-based neuroimaging signals: the importance and the challenge of understanding brain oxygen metabolism. Front Neuroenergetics 2, 8.
15. Devor, A. et al. (2011) "Overshoot" of o2 is required to maintain baseline tissue oxygenation at locations distal to blood vessels. J Neurosci 31 (38), 13676-81.
16. Ances, B.M. et al. (2001) Temporal dynamics of the partial pressure of brain tissue oxygen during functional forepaw stimulation in rats. Neurosci Lett 306 (1-2), 106-10.
17. Thompson, J.K. et al. (2003) Single-neuron activity and tissue oxygenation in the cerebral cortex. Science 299 (5609), 1070-2.
18. Gnaiger, E. et al. (1998) Mitochondrial oxygen affinity, respiratory flux control and excess capacity of cytochrome c oxidase. J Exp Biol 201 (Pt 8), 1129-39.
19. Wilson, D.F. (2013) Regulation of cellular metabolism: programming and maintaining metabolic homeostasis. J Appl Physiol (1985) 115 (11), 1583-8.
20. Wilson, D.F. et al. (1979) The oxygen dependence of cellular energy metabolism. Arch Biochem Biophys 195 (2), 485-93.
21. Liang, C.L. et al. (2013) Luminance contrast of a visual stimulus modulates the BOLD response more than the cerebral blood flow response in the human brain. Neuroimage 64, 104-11.
22. Duncker, D.J. and Bache, R.J. (2008) Regulation of coronary blood flow during exercise. Physiological Reviews 88 (3), 1009-1086.
23. Roy, C.S. and Sherrington, C.S. (1890) On the regulation of the blood-supply of the brain. J. Physiol 11, 85-108.
24. Perthen, J.E. et al. (2008) Caffeine-induced uncoupling of cerebral blood flow and oxygen metabolism: A calibrated BOLD fMRI study. Neuroimage 40 (1), 237-47.
25. Villringer, A. and Dirnagl, U. (1995) Coupling of brain activity and cerebral blood flow: basis of functional neuroimaging. Cerebrovascular and Brain Metabolism Reviews 7 (3), 240-276.

26. Hamel, E. (2006) Perivascular nerves and the regulation of cerebrovascular tone. J Appl Physiol 100 (3), 1059-64.
27. Koehler, R.C. et al. (2009) Astrocytes and the regulation of cerebral blood flow. Trends Neurosci 32 (3), 160-9.
28. Bonvento, G. et al. (2002) Does glutamate image your thoughts? Trends Neurosci 25 (7), 359-64.
29. Cauli, B. et al. (2004) Cortical GABA interneurons in neurovascular coupling: relays for subcortical vasoactive pathways. J Neurosci 24 (41), 8940-9.
30. Estrada, C. and DeFelipe, J. (1998) Nitric oxide-producing neurons in the neocortex: morphological and functional relationship with intraparenchymal microvasculature. Cereb Cortex 8 (3), 193-203.
31. Pelligrino, D.A. et al. (2010) Caffeine and the control of cerebral hemodynamics. J Alzheimers Dis 20 Suppl 1, S51-62.
32. Uhlirova, H. et al. (2016) Cell type specificity of neurovascular coupling in cerebral cortex. Elife 5.
33. Liu, E.Y. et al. (2019) The potential for gas-free measurements of absolute oxygen metabolism during both baseline and activation states in the human brain. bioRxiv, 705186.
34. Jaynes, E.T. (1965) Gibbs Vs Boltzmann Entropies. American Journal of Physics 33 (5), 391-+.
35. Bennett, C.H. (1982) The Thermodynamics of Computation - a Review. International Journal of Theoretical Physics 21 (12), 905-940.
36. Sevick, E.M. et al. (2008) Fluctuation theorems. Annual Review of Physical Chemistry 59, 603-633.
37. Ames, A., 3rd (2000) CNS energy metabolism as related to function. Brain Res Brain Res Rev 34 (1-2), 42-68.
38. Grubb, R.L., Jr. et al. (1974) The effects of changes in PaCO2 on cerebral blood volume, blood flow, and vascular mean transit time. Stroke 5 (5), 630-9.
39. Reivich, M. (1964) Arterial Pco2 and Cerebral Hemodynamics. Am J Physiol 206, 25-35.

40. Nioka, S. et al. (1990) Oxidative phosphorylation system during steady-state hypoxia in the dog brain. J Appl Physiol 68 (6), 2527-35.
41. Rahn, O. and Otis, A.B. (1949) Man's respiratory response during and after acclimitization to high altitide. Am J Physiol 157 (3), 445.
42. Wolff, C.B. (2000) Cerebral blood flow and oxygen delivery at high altitude. High Alt Med Biol 1 (1), 33-8.
43. Contreras, D. and Palmer, L. (2003) Response to contrast of electrophysiologically defined cell classes in primary visual cortex. J Neurosci 23 (17), 6936-45.
44. Wilson, H.R. and Cowan, J.D. (1972) Excitatory and inhibitory interactions in localized populations of model neurons. Biophys J 12 (1), 1-24.