

# Not So Short Introduction to VAN

## Data

There are six folders in /data/ folder, each with vasculature graphs for a single mouse.

Each mouse folder contains graph.seed, brOrder.mat, and Parameters.mat. These are the inputs for the model. The computed results from each module are saved in the /mouse#/results/ folder.

The graph.seed file contains the graph of the vasculature. The information is contained in the im2 structure. Here's a list of the most important fields of the im2 structure. To load the file into Matlab, go to the folder of the file and simply type `load('graph.seed', '-mat')`. *If you are only interested in the vasculature, this 'im2' structure is what you are looking for.*

im2.nodePos: the position of nodes. E.g. im2.nodePos(1,:) gives the x, y, z coordinates of the first node in  $\mu\text{m}$ .

im2.nodeEdges: the edges connecting the nodes. E.g. for the first mouse im2.nodeEdges(1,:)=15 41, which means that the 1st edge connects the 15th node and 41st node.

im2.nodeDiam: the diameter of the node in  $\mu\text{m}$ .

im2.nodeType: 1: arterioles; 2: capillaries; 3: venules.

## Code

### Mesh

The VAN\_setup\_mesh.m in the /code/ folder generates mesh, using iso2mesh toolbox. The input file is graph.seed and the output file is mesh.mat in /mouse#/results/ folder.

### Vessel Dilation

Input: mesh.mat, brOrder.mat.

Output: dilate\_vessel\_NCES-Sigmoid#.mat.

This module calculates flow dynamics with arterial dilation. The code used is VAN\_calFlow\_sigmoid.m.

### Advection-diffusion solver

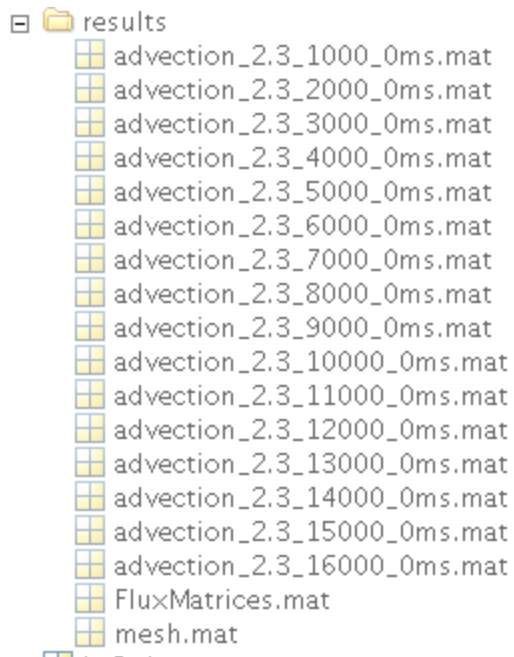
Input: mesh.mat, baseline CMRO<sub>2</sub>.

Output: advection\_CMRO<sub>2</sub>\_time\_dilation ## ms.mat

The code to use is VAN\_advection\_setup.m.

The advection-diffusion module is the most time-consuming part. Intermediate results are saved in /mouse#/results/.

For example, if you have set the 'duration\_simulation' time to be 18, but due to computer crash or some other reasons, you only see 16 files as below



You can use the function `VAN_advection_setup_breaktoparts` to complete the computation instead of rerun the simulation from the very beginning.

```
%% advection-diffusion solver

duration_simulation = 18; % Duration of simulation (sec)

%VAN_advection_setup(cmro2, duration_simulation,savefolder,VesselDiPercent);

VAN_advection_setup_breaktoparts(cmro2,duration_simulation,savefolder,VesselDiPercent,16)
```

### *Monte-Carlo simulations*

Input includes all the computed files from previous steps and the acquisition parameters of magnetic field strength and its orientation  $B_0$ , gradient field  $G_x$ , echo time  $TE$ .

The final result is `BOLD_GE`, which is the BOLD signal.

## **Review of VAN modeling**

### *Review of vascular anatomical network (VAN) modeling*

The VAN model is a bottom-up model that computes oxygen distribution, blood flow and BOLD signal within a micro-vascular network.

Raw data obtained from experiments are converted to a graph which is a mathematical object of nodes and segments representing the vascular network. We have used a suite of custom-designed tools in MATLAB, with manual corrections applied until all segments in the field of view become interconnected. Vessel size

was estimated at each graph node by thresholding the image at a low value of  $\sim 2\%$ . Vessel types and branching orders are manually obtained by following them from the pial surface. Pial arterioles and venules are identified based on  $PO_2$  measurements and their morphology.

Mesh generation is the first step of VAN modeling, providing meshes for the finite element method used to compute oxygen distribution. The toolbox we use is *iso2mesh* (Fang and Boas, 2009).

Oxygen distribution is obtained by solving the advection-diffusion equation

$$\frac{\partial C_T}{\partial t} = \vec{v} \cdot \nabla C_F - \vec{v} \cdot \nabla C_B + \nabla \cdot (D_{O_2} \nabla C_F) - OC,$$

where  $C_F$  and  $C_B$  are the free and bounded oxygen concentrations respectively;  $C_T = C_B + C_F$  denotes the total oxygen concentration,  $\vec{v}$  is the velocity,  $D_{O_2}$  is the oxygen diffusion coefficient and  $OC$  is the tissue oxygen consumption rate. Another equation that governs the equilibrium between  $C_F$  and  $C_B$  is

$$C_B = 4C_{Hb}HctSO_2(C_F),$$

where  $Hct$  is hematocrit;  $C_{Hb}$  is the hemoglobin concentration within a red blood cell,  $SO_2(C_F)$  is the hemoglobin oxygen saturation which is in equilibrium with the concentration of free oxygen. This equation is solved by our hybrid model including a graph-based 1D oxygen advection within the vessel, a 1D oxygen flux conservation model across the vessel wall, and an finite-element-based 3D oxygen diffusion model within the tissue (Fang et al., 2008).

The VAN model computes blood flow within the vasculature from the flow-pressure relationship

$$P_1 - P_2 = F_{12}R_{12},$$

where  $P_1 - P_2$  is the pressure drop of the segment,  $F_{12}$  is the blood flow and  $R_{12}$  is the resistance which is inversely proportional to the square of the volume  $1/V_{12}^2$  (Boas et al., 2008). When the volume  $V_{12}$  changes, pressure and blood flow distributions are recomputed to modulate neural activation with arterial dilation.

The Monte-Carlo method is used to compute the BOLD signal. The volume susceptibility shift is

$$\Delta\chi = \delta\chi_0 Hct(1 - SO_2),$$

where  $\delta\chi_0$  is the susceptibility difference between fully oxygenated and fully deoxygenated haemoglobin. The value of the haematocrit  $Hct$  is 0.3 in capillaries and 0.4 in arteries and veins. The magnetic field inhomogeneity  $\Delta B(\vec{x})$  is computed by convolving  $\Delta\chi$  with the geometrical factor for the magnetic field induced by a unit cube

$$\Delta B_{cube} = \frac{(\frac{2}{\pi})a^3}{r^3} (3\cos^2\theta - 1)B_0.$$

Besides the magnetic field inhomogeneity, intrinsic  $T_2^*$  variations are also included in the modeling. It is very hard to model intravascular signals from microscopic details. We estimate  $T_2^*$  values inside vessels from experimental measurements (Uludağ et al., 2009)

$$T_{2,vessel}^* = (A + C(1 - SO_2)^2)^{-1},$$

where  $A$  and  $C$  are constants that depend on the external magnetic field  $B_0$  as in Table 1.

**Table S1. Constants for  $T_2^*$  within vessels**

Magnetic Field strength	A	C
$B_0 \leq 1.5T$	6.5	25
$1.5T < B_0 \leq 3T$	13.8	181
$3T < B_0 \leq 4T$	30.4	262
$4T < B_0 \leq 4.7T$	41	319
$4.7T < B_0$	100	500

In the tissue,  $T_2$  and  $T_2^*$  are computed as

$$T_{2,tissue} = (1.74 \cdot B_0 + 7.77)^{-1},$$

$$T_{2,tissue}^* = (3.74 \cdot B_0 + 9.77)^{-1}.$$

The diffusion of protons is modeled with a Monte-Carlo method. The initial positions of  $10^7$  protons are randomly distributed within the 3D volume. The diffusion coefficient of the protons is  $D = 1 \times 10^{-5} \text{ cm}^2 \text{ s}^{-1}$ , and the time step  $dt = 0.2 \text{ ms}$ . The position of each proton  $(x_1, x_2, x_3)$  after time  $dt$  is

$$x'_i = x_i + N(0, 2Ddt), i = 1, 2, 3,$$

where  $N(0, 2Ddt)$  is the normal distribution. Protons cannot move across the vessel membrane in our simulations. The phase increment for a proton at each  $dt$  is

$$\delta\phi_{intra} = -T_{2,vessel}^* dt/i,$$

$$\delta\phi_{intra} = -T_{2,tissue}^* dt/i + \gamma \Delta B(\vec{x}) dt.$$

Here  $\gamma = 2.675 \times 10^5 \text{ rad/T/ms}$  is the hydrogen proton precession frequency. The MR signal is computed as

$$S(t) = \text{Re} \left\{ \frac{1}{N} \sum_1^N e^{i\phi_n(t)} \right\}.$$

For simulation of the gradient echo MR signal, a gradient magnetic field  $\Delta G_x x$  is turned on at  $TE/3$  and flipped at  $2TE/3$ . For the spin echo signal, the imaginary part of the phase is inverted at  $TE/2$ ,  $\phi_n(TE/2) = \text{conj}(\phi_n(TE/2))$ . Finally, the BOLD signal is obtained as

$$BOLD = S(TE).$$

We can compute the BOLD signal for various oxygen consumption rates and blood flow distributions to model activation.

## Notes

1. VANmodelrun.m contains information to call different modules above.
2. Matlab versions beyond 2017a currently cannot calculate a large VAN. This is related to the algorithm of '\ ' operator. We will work on this problem in the future.
3. The real  $CMRO_2$  increase can be obtained from the increase of *OCoutput* variable in the advection\_##\_##\_##ms.mat file from the output of the advection-diffusion solver. This value can be different from the input  $CMRO_2$  change, which is currently a discrepancy of our model. This difference is large for mouse2. This mouse was the first animal we dealt with and graphing may not be perfect. So if you are doing activation, you can avoid using this animal for now.

## References

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- Fang, Q., Boas, D.A., 2009. Tetrahedral mesh generation from volumetric binary and grayscale images, in: 2009 IEEE International Symposium on Biomedical Imaging: From Nano to Macro. Presented at the 2009 IEEE International Symposium on Biomedical Imaging: From Nano to Macro, pp. 1142–1145. doi:10.1109/ISBI.2009.5193259
- Fang, Q., Sakadžić, S., Ruvinskaya, L., Devor, A., Dale, A.M., Boas, D.A., 2008. Oxygen advection and diffusion in a three dimensional vascular anatomical network. *Opt. Express* 16, 17530.
- Uludağ, K., Müller-Bierl, B., Uğurbil, K., 2009. An integrative model for neuronal activity-induced signal changes for gradient and spin echo functional imaging. *NeuroImage* 48, 150–165. doi:10.1016/j.neuroimage.2009.05.051