# A Model of the Blood Flow in the Retina of the Eye 

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## Abstract:

Understanding the blood flow in the retina of the eye may provide insights into several eye pathologies and ultimately lead to better treatments. We model the flow as a hierarchical Darcy flow on a curved surface and solve the model numerically using discrete exterior calculus and finite element exterior calculus. Results support the hypothesis that changes in the shape of the retina cause significant changes in the ocular blood flow, which may play a role in the dynamics of open angle glaucoma.

## Outline

Modeling the Eye: Goals and Motivation

```
A Multiscale Model of Blood Flow in the Retina
    Hierarchical Darcy Flow
    Vascular Tree Architecture
Computational Method
    Hierarchical Discretization
    Spatial Discretization
    Curved Surfaces
    Discrete Exterior Calculus (DEC) and Finite Element Exterior
    Calculus (FEEC)
Results
    Ocular Curvature May Play a Role in Glaucoma
Summary and Future Plans
```


## The Eye

- The eye is the organ of sight: the sensory part is the retina.



## Goals

- Advance the theory: most accurate models become theory
- Help reason: simplified models help us think more clearly
- Simulate experiments: experiments may be hard, or even impossible
- Competing goals: tension between simplicity and accuracy
- Solution: a collection of models at various levels of refinement


## Motivation

- Preserve the vision
- Window into the body
- Part of the brain
- Opportunity to study flow on curved surfaces


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A Multiscale Model of Blood Flow in the Retina

## Fundus Images

- A healthy right eye (left) and a healthy left eye (right)



## Regions of Particular Clinical Interest in the Retina



- optic nerve head (ONH)
- inferior quadrant (I)
- superior quadrant (S)
- nasal quadrant (N)
- temporal quadrant (T)
- fovea
- At the level of capillaries and small arterioles and venules, we don't see the details of the vascular tree architecture
- The tissue appears to be a porous medium
- Think of the vascular tree as a hierarchical structure
- Parameterize these hierarchical levels by a continuous parameter $\vartheta$


## Hierarchical Darcy Flow

- Consider averaged (smeared) quantities at position $\mathbf{x}$ in some spatial domain $\Omega$ for each hierarchy level $\vartheta$
- Velocity $\mathbf{v}(\mathbf{x}, \vartheta)$
- Hierarchical velocity $\omega(\mathbf{x}, \vartheta)$
- Pressure $p(\mathbf{x}, \vartheta)$
- Model each $\vartheta$ level by Darcy flow equations and couple "vertically" with hierarchical flow

$$
\begin{array}{rlrl}
\nabla \cdot\left(n_{b} \mathbf{v}\right)+\frac{\partial}{\partial \vartheta}\left(n_{b} \omega\right)=0 & & \mathbf{x} \in \Omega, \vartheta \in(0,1) \\
n_{b} \mathbf{v} & =-\mathbf{K} \nabla p & & \mathbf{x} \in \Omega, \vartheta \in(0,1) \\
n_{b} \omega & =-\alpha \frac{\partial p}{\partial \vartheta} & & \mathbf{x} \in \Omega, \vartheta \in(0,1)
\end{array}
$$

## Parameters and Boundary Conditions

- Parameters
- Tissue porosity $n_{b}(\mathbf{x}, \vartheta)$
- Tissue perfusion $n_{b} \omega$
- Tissue permeability tensor $\mathbf{K}(\mathbf{x}, \vartheta)$
- Hierarchical permeability $\alpha(\mathbf{x}, \vartheta)$
- Hydraulic conductivities $G_{v}, G_{a}$
- Venous and arterial pressures $p_{v}(\mathbf{x})$ and $p_{a}(\mathbf{x})$
- Boundary conditions

$$
\begin{array}{rr}
n_{b} \omega(\mathbf{x}, 0)=-G_{v}\left(p(\mathbf{x}, 0)-p_{v}(\mathbf{x})\right) & \mathbf{x} \in \Omega \\
n_{b} \omega(\mathbf{x}, 1)=-G_{a}\left(p_{a}(\mathbf{x})-p(\mathbf{x}, 1)\right) & \mathbf{x} \in \Omega \\
n_{b} \mathbf{v}(\mathbf{x}, \vartheta) \cdot \mathbf{n}=0 & \mathbf{x} \in \partial \Omega, \vartheta \in[0,1]
\end{array}
$$

- Couple to feeding arteries and draining veins

$$
G_{v}=\alpha_{v} \delta_{v}(\mathbf{x}) \quad \text { and } \quad G_{a}=\alpha_{a} \delta_{a}(\mathbf{x})
$$

- $\alpha_{v}$ and $\alpha_{a}$ are venous and arterial conductances


## Vascular Tree Architecture

- $\delta_{v}(\mathbf{x})$ and $\delta_{a}(\mathbf{x})$ are delta functions which differ from zero only where the feeding arteries and draining veins are located, respectively
- Need to extract feeding arteries and draining veins from fundus images
- Alternatively, model the vascular tree
- Young (1808): Compare parent and daughters vessels
- Assume two identical daughters
- Radius ratio: $\frac{r_{p}}{r_{d}}=2^{\frac{1}{3}}=1.26$
- Area ratio: $\frac{A_{p}}{A_{d}}=2^{-\frac{1}{3}}$
- Murray (1926): Generalize to asymmetric trees
- Similar to Pythagora, except cubes: $r_{p}^{3}=\sum r_{d}^{3}$
- Derives from optimality (variational principle)
- Forgotten for many years, then "rediscovered"


## Optimal Transport $\Longrightarrow$ Murray's Law

- $P_{f}$ power needed to maintain the blood flow (viscous losses)
- $P_{m}$ power needed to metabolically maintain the blood and the vessel
- $P_{t}=P_{f}+P_{m}$ total power
- Minimize $P_{t}$ !
- Assume cylindrical vessels: $f=c p$
- $f$ is volumetric flow rate
- $p$ is pressure difference
$-c$ is conductance coefficient
- Assume Poiseuille flow: $c=\frac{\pi r^{4}}{8 \mu l}$
- $\mu$ is blood viscosity
- $l$ is vessel length
- $\operatorname{Min} P_{f}=p f=a f^{2} r^{-4} \Longrightarrow r \rightarrow \infty$ ?
$-a=\frac{8 \mu l}{\pi}$
- Min $P_{m}=m V=m \pi r^{2} l=b r^{2} \Longrightarrow r \rightarrow 0$ ?
- $m$ is a metabolic coefficient
$-b=m \pi l$
- $\operatorname{Min} P_{t} \Longrightarrow \frac{d P_{t}}{d r}=0 \Longrightarrow-4 a f^{2} r^{-5}+2 b r=0 \Longrightarrow f=k r^{3}$ !
$-k=\left(\frac{b}{2 a}\right)^{1 / 2}$


## Recent Experimental Data and Models

- Takahashi (2009, 2013): $f=k r^{m}$
- Theory (Murray): $m=3$
- Data (Takahashi 2009): $m \approx 2.85$
- Data (Takahashi 2013): significant deviations in $m$ between various organs
- In addition to radius, interested in length and angle
- Theory: open problem!
- Data: rather scarce - open problem!
- Model (Takahashi 2009): $L=7.4 r^{1.15}$
- Open question: what optimality?


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## Hierarchical Discretization

- Standard FE procedure
- Rewrite in weak form
- Introduce piecewise linear basis functions $\left\{\varphi_{i}\right\}$ with $i=0, . ., n$ on $[0,1]$ corresponding to nodes $\left\{\theta_{i}\right\}$ and we let $\varphi=\varphi_{i}$.
- We assume permeabilities and pressure are piecewise linear interpolations of the hierarchical variable

$$
\begin{gathered}
\mathbf{K}(\mathbf{x}, \theta)=\sum_{k=0}^{n} \mathbf{K}\left(\mathbf{x}, \theta_{k}\right) \varphi_{k}(\theta), \quad \alpha(\mathbf{x}, \theta)=\sum_{k=0}^{n} \alpha\left(\mathbf{x}, \theta_{k}\right) \varphi_{k}(\theta) \\
p(\mathbf{x}, \theta)=\sum_{k=0}^{n} p\left(\mathbf{x}, \theta_{k}\right) \varphi_{k}(\theta) \\
-\sum_{j=0}^{n} \int_{\Omega}\left(\mathbf{K}_{i j}(\mathbf{x}) \nabla p_{j}(\mathbf{x}), \nabla q(\mathbf{x})\right) \mathrm{d} \mathbf{x}-\sum_{j=0}^{n} \int_{\Omega} \alpha_{i j}(\mathbf{x}) p_{j}(\mathbf{x}) q(\mathbf{x}) \mathrm{d} \mathbf{x} \\
\text { Computational Method } \quad=\int_{\Omega} f_{i}(\mathbf{x}) q(\mathbf{x}) \mathrm{d} \mathbf{x}
\end{gathered}
$$

## Hierarchical Discretization Cont.

- where

$$
\begin{align*}
\mathbf{K}_{i j}(\mathbf{x}) & =\sum_{k=0}^{n} \mathbf{K}\left(\mathbf{x}, \theta_{k}\right) \int_{[0,1]} \varphi_{k}(\theta) \varphi_{i}(\theta) \varphi_{j}(\theta) \mathrm{d} \theta \\
\alpha_{i j}(\mathbf{x}) & =\sum_{k=0}^{n} \alpha\left(\mathbf{x}, \theta_{k}\right) \int_{[0,1]} \varphi_{k}(\theta) \frac{\partial \varphi_{i}(\theta)}{\partial \theta} \frac{\partial \varphi_{j}(\theta)}{\partial \theta} \mathrm{d} \theta  \tag{1}\\
f_{i}(\mathbf{x}) & =\alpha_{v} p_{v} \delta_{i 0}+\alpha_{a} p_{a} \delta_{i n}
\end{align*}
$$

- The hierarchical discretization leads to a tridiagonal system for the pressures $p_{i}\left(\mathbf{x}, \theta_{i}\right)$.


## Example: 3 layer model

- For three levels $i=0,1,2$ and with isotropic and space independent permeability $\mathbf{K}_{i j}=K_{i j} \mathbf{I}$ per level we have $\nabla \cdot \mathbf{K}_{i j} \nabla=K_{i j} \Delta$
- This yields

$$
-\left[\begin{array}{lll}
K_{00} \Delta+\alpha_{00} & K_{01} \Delta+\alpha_{01} &  \tag{2}\\
K_{10} \Delta+\alpha_{10} & K_{11} \Delta+\alpha_{11} & K_{12} \Delta+\alpha_{12} \\
& K_{21} \Delta+\alpha_{21} & K_{22} \Delta+\alpha_{22}
\end{array}\right]\left[\begin{array}{l}
p_{0} \\
p_{1} \\
p_{2}
\end{array}\right]=\left[\begin{array}{l}
f_{0} \\
f_{1} \\
f_{2}
\end{array}\right] .
$$

- Permeability matrix $K_{i j}$, conductivity matrix $\alpha_{i j}$, and right hand side vector $f_{i}$ as given in equation (1).


## The $K$ and $\alpha$ matrices in the 3 layer model

- The scalar $K_{i j}$ is the $i j$ entry of the matrix

$$
\frac{1}{24}\left[\begin{array}{ccc}
3 K_{0}+K_{1} & K_{0}+K_{1} & 0 \\
K_{0}+K_{1} & K_{0}+6 K_{1}+K_{2} & K_{1}+K_{2} \\
0 & K_{1}+K_{2} & K_{1}+3 K_{2}
\end{array}\right],
$$

- The scalar $\alpha_{i j}$ is the $i j$ entry of the matrix

$$
\left[\begin{array}{ccc}
\alpha_{0}+\alpha_{1}+G_{v} & -\left(\alpha_{0}+\alpha_{1}\right) & 0 \\
-\left(\alpha_{0}+\alpha_{1}\right) & \alpha_{0}+2 \alpha_{1}+\alpha_{2} & -\left(\alpha_{1}+\alpha_{2}\right) \\
0 & -\left(\alpha_{1}+\alpha_{2}\right) & \alpha_{1}+\alpha_{2}+G_{a}
\end{array}\right]
$$

- The scalar $f_{i}$ is the $i$ component of the vector

$$
\left[G_{v} p_{v}, 0, G_{a} p_{a}\right]^{T}
$$

Here $\alpha_{0}, \alpha_{1}, \alpha_{2}$ and $K_{0}, K_{1}, K_{2}$ are the conductivities and permeabilities at the three hierarchical levels and are given in equation (1).

## Exterior Algebra

- Motivation: generalize the cross product $\times$ from $\mathbb{R}^{3}$ to $\mathbb{R}^{n}$
- $V$ an $n$ dimensional real vector space with elements $u, v, \ldots$
- $u \wedge v$ exterior (wedge) product of $u$ and $v$
- anti-symmetric: $u \wedge v=-v \wedge u$
- linear in each factor: $(a u+b v) \wedge w=a u \wedge w+b v \wedge w$
- $\bigwedge^{0} V=\mathbb{R}$
- $\Lambda^{1} V=V$
- $\bigwedge^{2} V=V \bigwedge V, 2$-vectors
- continue recursively, demanding associativity
- $\bigwedge^{p} V$ p-vectors, $\operatorname{dim}\left(\bigwedge^{p} V\right)=\binom{n}{p}$
- $\operatorname{dim}\left(\bigwedge^{n} V\right)=1$
- $V$ inner product space:
- natural isomorphism between $V$ and its dual $V^{*}$
- flat: b:V$\rightarrow V^{*}, u^{b}(v)=(u, v)$
- sharp: $\sharp: V^{*} \rightarrow V, \sharp=b^{-1}$
- inner product on $V$ induces inner product on $\Lambda^{p} V$ (the notion of length induces notions of area, volume, $p$-volume)


## Hodge *

- Motivation: $u \wedge v$ is a 2-vector, so not quite $u \times v$ yet
- Solution: so map $u \wedge v$ to $u \times v$ - call this map Hodge $*$
- Then: $u \times v=*(u \wedge v)$
- $V$ oriented inner product space: * : $\bigwedge^{p} V \rightarrow \bigwedge^{(n-p)} V$
- natural isomorphism
- Let $\left\{e_{1}, e_{2}, \ldots, e_{n}\right\}$ be a basis for $V$
- Let $\sigma=e_{1} \wedge e_{2} \wedge \cdots \wedge e_{n}$ be a chosen orientation
- Hodge $*: u \wedge v=(* u, v) \sigma, \quad \forall v \in \bigwedge^{(n-p)} V$
- Intuition: complementary $n-p$ vector to the given $p$ vector
- orthogonal
- consistent with the orientation
- Side result: $u \cdot v=(u, v)=*(u \wedge * v)=*(v \wedge * u)$


## Hodge * Example

- $V=\mathbb{R}^{2}$ :

$$
\begin{array}{ccc}
* 1 & = & e_{1} \wedge e_{2} \\
* e_{1} & = & e_{2} \\
* e_{2} & = & -e_{1} \\
*\left(e_{1} \wedge e_{2}\right) & = & 1
\end{array}
$$

- $V=\mathbb{R}^{3}$ :

$$
\begin{array}{ccc}
* 1 & = & e_{1} \wedge e_{2} \wedge e_{3} \\
* e_{1} & = & e_{2} \wedge e_{3} \\
\ldots & = & \cdots \\
*\left(e_{1} \wedge e_{2}\right) & = & e_{3} \\
\ldots & = & \cdots \\
*\left(e_{1} \wedge e_{2} \wedge e_{3}\right) & = & 1
\end{array}
$$

## Exterior Calculus

- Differential forms - generalizations of differentials
- Let $M$ be a manifold
- 0-forms are functions
- 1-forms are differentials
- $k$-forms at a point $P$ are $k$ vectors (elements in $\bigwedge^{k} T_{P}^{*} M$ )
- Exterior Derivative $d$
$-d f=\frac{\partial f}{\partial x^{i}} d x^{i}$
$-d\left(f d x^{1} \wedge \cdots \wedge d x^{k}\right)=d f \wedge d x^{1} \wedge \cdots \wedge d x^{k}$
- Adjoint derivative: $\delta=* d *$, with $(\delta \alpha, \beta)=(\alpha, d \beta)$
- Laplacian: $\Delta=(d+\delta)^{2}$
- Interior Product: $\left(i_{X} \alpha\right)\left(v_{2}, \ldots, v_{k}\right)=\alpha\left(X, v_{2}, \ldots, v_{k}\right)$
- Lie Derivative (using Cartan's Magic Formula): $L_{X} \alpha=d i_{X} \alpha+i_{X} d \alpha$


## Curved Surfaces

- Numerically, a major issue is: retina is a curved surface
- Flat: $\left(\frac{\partial}{\partial x^{i}}\right)^{b}=g_{i j} d x^{j},\left(v^{b}\right)_{i}=g_{i j} v^{j}$
- Sharp: $\left(d x^{i}\right)^{\sharp}=g^{i j} \frac{\partial}{\partial x^{j}}$
- Hodge $*:(* \alpha)_{i_{1}, i_{2}, \ldots, i_{n-k}}=\frac{1}{k!} \alpha^{j_{1}, \ldots, j_{k}} \sqrt{\operatorname{det} g} \epsilon_{j_{1}, \ldots, j_{k}, i_{1}, \ldots, i_{n-k}}$


## Algebraic Topology Approach to Space Discretization

- Simplices
- Simplicial Complexes
- Chains
- Cochains
- Boundary operator $\partial$
- Discrete exterior derivative $D=\partial^{T}$
- Discrete version of Stokes theorem


## $M^{D E C}$ : DEC Hodge *

- Marsden, Hirani, Desbrun, et al.: $\frac{1}{|* \sigma|} \int_{* \sigma} * \alpha=\frac{1}{|\sigma|} \int_{\sigma} \alpha$
- $\left[M^{D E C}\right]_{i j}=\frac{\left|* \sigma_{i}^{k}\right|}{\left|\sigma_{i}^{k}\right|} \delta_{i j}$
- Advantage: $M^{D E C}$ diagonal
- Disadvantage: $M^{D E C}$ not positive definite, because circumcenter may be outside the simplex
- Example: standard 2-simplex; vertices: $(0,0),(1,0),(0,1)$

$$
M_{1}^{D E C}=\left[\begin{array}{ccc}
\frac{1}{2} & 0 & 0 \\
0 & \frac{1}{2} & 0 \\
0 & 0 & 0
\end{array}\right]
$$

## $M^{\text {Whit }}$ : FEEC Hodge *

- Arnold, Falk, Winther.
- Whitney 0 -forms are barycentric coordinates:
- Whitney 1-forms: $\eta_{k}=\lambda_{i} d \lambda_{j}-\lambda_{j} d \lambda_{i}$
- $\left[M^{W h i t}\right]_{i j}=\int \eta_{i} \eta_{j} d A$
- Advantage: $M^{W h i t}$ is positive definite, because barycenter is always inside the simplex
- Disadvantage: $M^{W h i t}$ not diagonal, $\left(M^{W h i t}\right)^{-1}$ not sparse
- Example: standard 2-simplex; vertices: $(0,0),(1,0),(0,1)$

$$
M_{1}^{W h i t}=\left[\begin{array}{ccc}
\frac{1}{3} & \frac{1}{6} & 0 \\
\frac{1}{6} & \frac{1}{3} & 0 \\
0 & 0 & \frac{1}{6}
\end{array}\right]
$$

## PyDEC - A Python Implementation of DEC

- Authors: Hirani and Bell
- Efficient operator implementation in terms of sparse matrices
- Data structures
- Simplicial Complex
- Regular Cube Complex
- Operators
- Discrete Exterior Derivative
- Diagonal Sparse Matrix Discrete Hodge *
- First Order Whitney Hodge *
- Provides a nice playground for numerical experiments


## Data

Figure: Retina Fundus Image

webvision.med.utah.edu/book/part-i-foundations/simple-anatomy-of-the-retina/

## Spatial Discretization

Figure: Triangularization by gmsh.


## Example: 3 layer model geometry

Figure: Three layer discretization


The $\theta=1$ layer represents arterioles, the $\theta=\frac{1}{2}$ layer represents capillaries and $\theta=0$ represents venules.

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# Capillary Velocity Field 

Figure: Capillary Blood Velocity


## Application: Myopia

Figure: Capillary Blood Velocity in Myopia
0.102
0.076
0.051
0.026
0.000

The myopic eye is extended along the optical axis.

## Eye Deformations

Figure: Eye Shapes


## Velocity Field Deformations

Figure: Velocity Difference


Caution: to compare the velocity field difference in a meaningful way, all velocity fields are first pulled back to the sphere.

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## Summary

- The hierarchical Darcy equation models blood flow in the retina as a hierarchical porous medium.
- Exterior Calculus generalizes vector calculus to manifold and is convenient for formulating and solving models on surfaces.
- The discrete exterior derivative $d$ is unique and determined by the discrete Stokes theorem.
- The discrete Hodge * operator is an open question and a topic of current research.
- DEC and FEEC are related; they differ in the Hodge * discretization.
- DEC and FEEC can be mixed with traditional FEs.
- Curvature may play an important role in the development of glaucoma.


## Future Plans

- Generalize the model to space dependent parameters.
- Couple the model with elasticity.
- Use the model to make predictions for other eye pathologies.
- The eye is the window into the body. What can we learn about conditions in the rest of the body?
- Personalized medicine: couple the model with image analysis front end.


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## Thank You!

