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A PREEMINENT
RESEARCH
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A fast approach to estimating Windkessel model parameters for patient-specific multi-scale CFD simulations of aortic flow

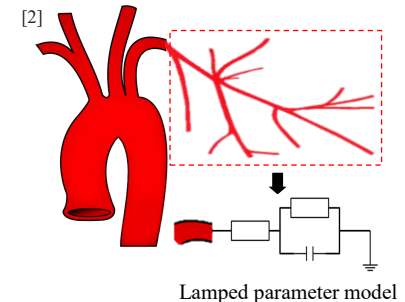
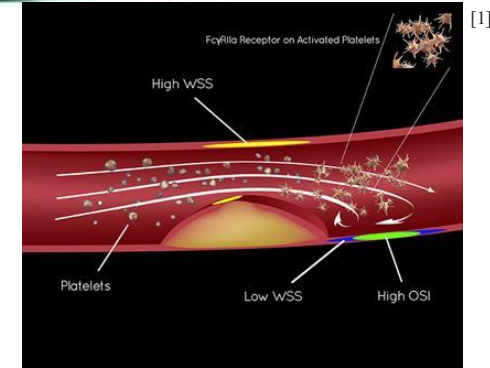
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Introduction:

- ❑ Knowing hemodynamics in vessel is important.
 - Wall shear stress (WSS) is related to thrombosis.
- ❑ Current clinical flow visualization techniques have limitations.
 - Difficult to obtain velocity and pressure fields in vasculature.
- ❑ Computational fluid dynamics (CFD) with Windkessel model.
 - Current way to find parameters for Windkessel model needs iterations of periodic flow simulation.
- ❑ Objective: find a fast way to obtain accurate Windkessel model parameters.



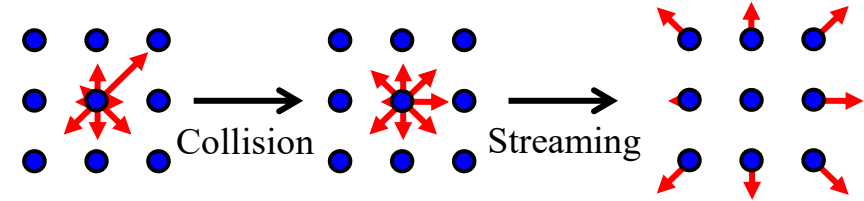
[1] Liebeskind D S, Hinman J D, Kaneko N, et al. Endothelial Shear Stress and Platelet FcγRIIa Expression in Intracranial Atherosclerotic Disease[J]. Frontiers in Neurology, 2021, 12: 244.

[2] https://en.wikipedia.org/wiki/Ascending_aorta

Lattice Boltzmann method (LBM)

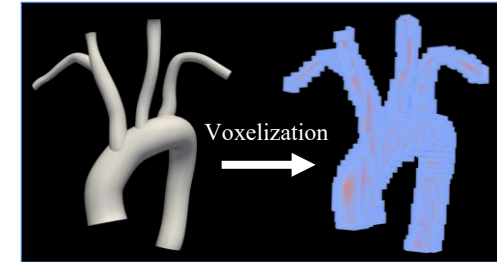
□ Simple two step algorithm:

1. Fluid node collides by the collision operator.
2. Stream to neighbor fluid nodes.

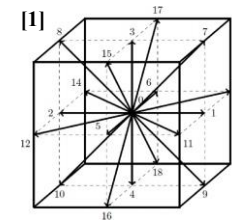


□ Advantages:

- Collision step is local; streaming step takes no computation; natural to parallelize.
- Easy to handle complex geometry by voxelization.



□ BGK collision operator and D3Q19 lattice structure are used in this study.



D3Q19 descriptor

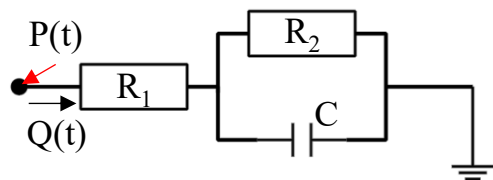
[2]

BGK operator

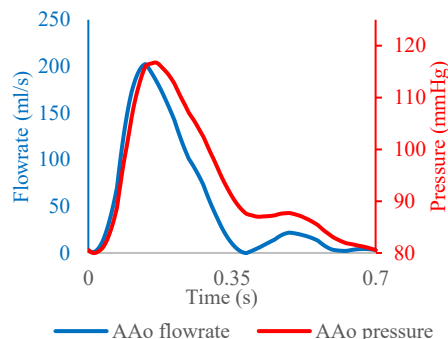
[1] Lenan, Zhang & Jebakumar, Anand Samuel & Abraham, John. (2016). Lattice Boltzmann method simulations of Stokes number effects on particle motion in a channel flow. Physics of Fluids. 28. 063306. 10.1063/1.4953800.

[2] https://en.wikipedia.org/wiki/Bhatnagar%E2%80%93Gross%E2%80%93Krook_operator

Three element Windkessel model

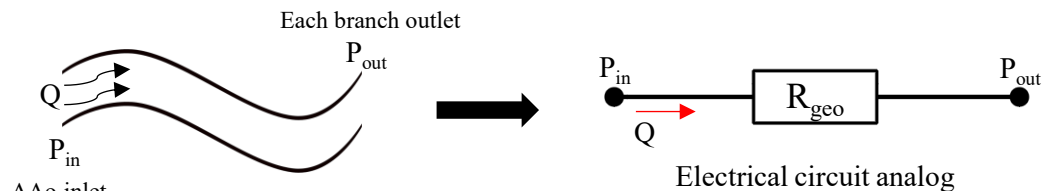


$$\left(1 + \frac{R_1}{R_2}\right) Q(t) + CR_1 \frac{\partial Q(t)}{\partial t} = \frac{P(t)}{R_2} + C \frac{\partial P(t)}{\partial t}$$

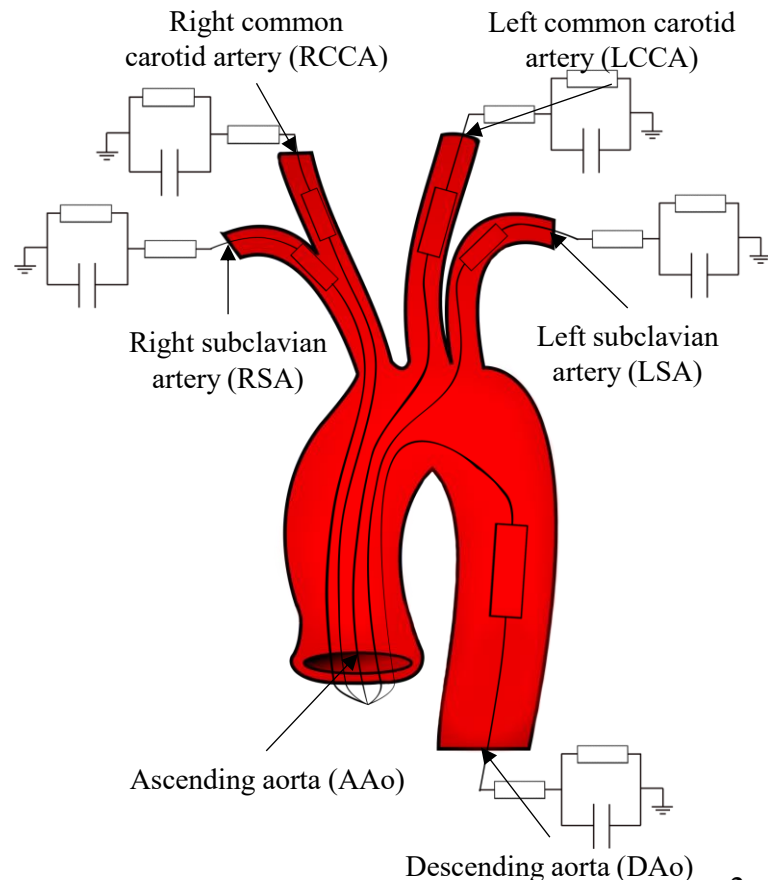


R_1 : characteristic resistance; R_2 : peripheral resistance; C : compliance;

Geometric resistance



$$R_{geo} = \frac{\int_0^T [P_{in}(t) - P_{out_i}(t)] dt}{\int_0^T Q_{out_i}(t) dt}$$



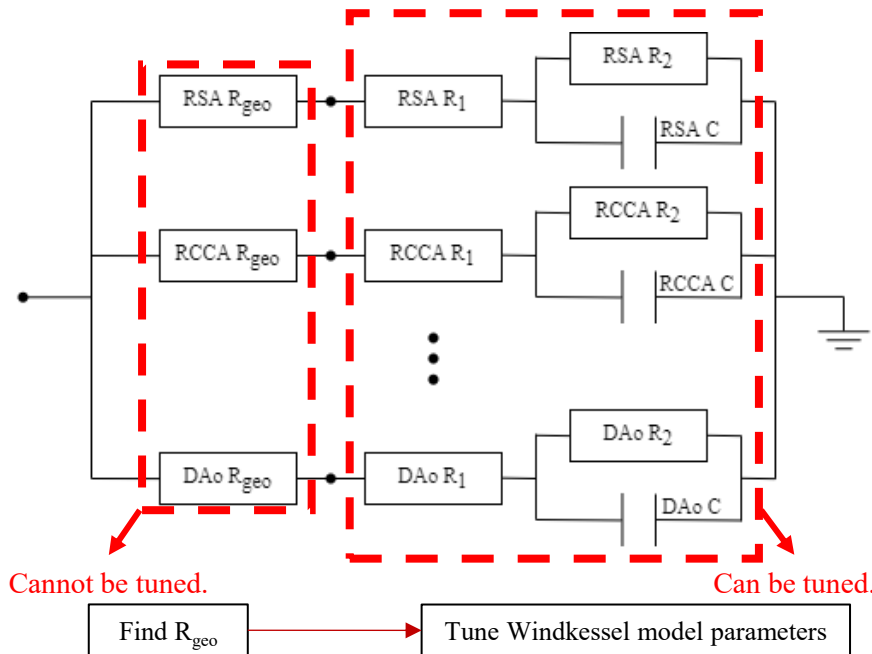
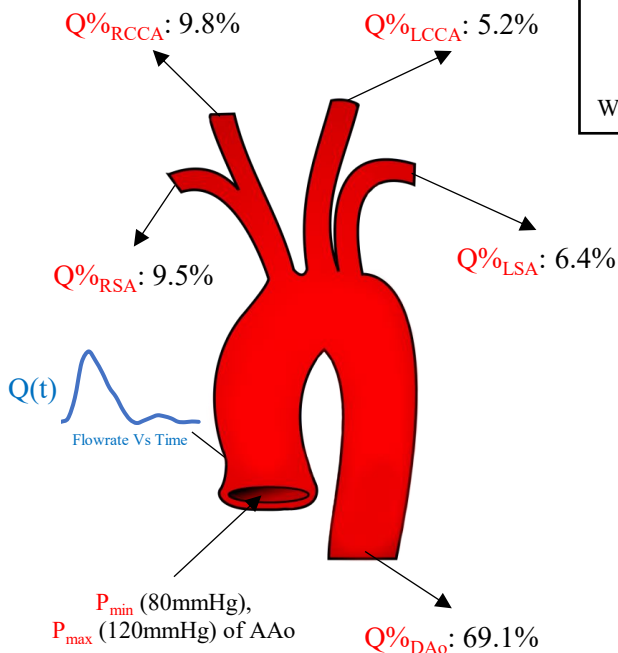
Goal of our algorithm

Input parameter:
Flowrate waveform: $Q(t)$

Controllable parameters:
Max, min pressure at AAo: P_{\max}, P_{\min} ;
Flow distribution of each branch: $Q\%_{RSA},$
 $Q\%_{RCCA}, Q\%_{DAo}, Q\%_{LCCA}, Q\%_{LSA}$

$$Q\%_i = \frac{\int_0^T Q_i(t)dt}{\int_0^T Q_{AAo}(t)dt}$$

Where i suggests each branch.

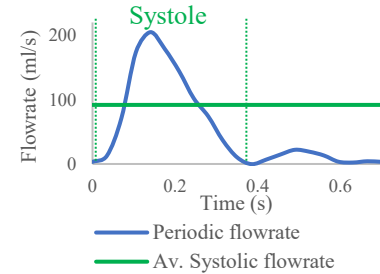


Flow distribution can be controlled by:

$$Q\%_{RSA} : Q\%_{RCCA} : \dots : Q\%_{DAo} = \frac{1}{R_{totalRSA}} : \frac{1}{R_{totalRCCA}} : \dots : \frac{1}{R_{totalDAo}}$$

For each branch: $R_{total} = R_{geo} + R_1 + R_2$

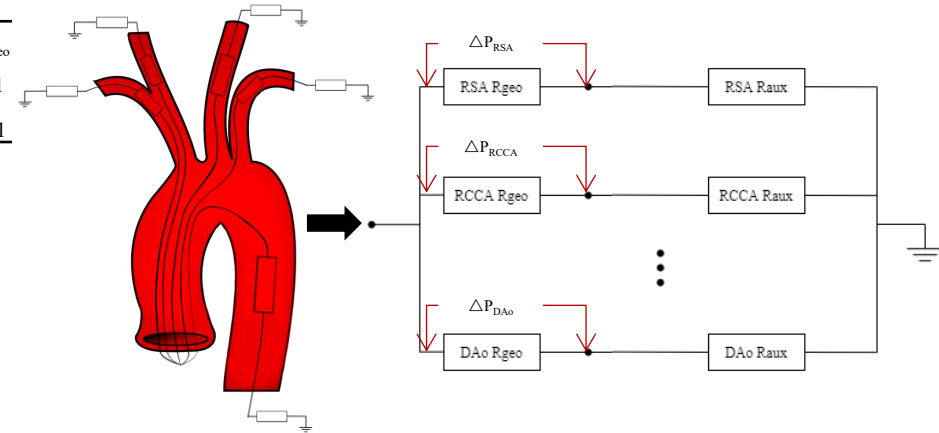
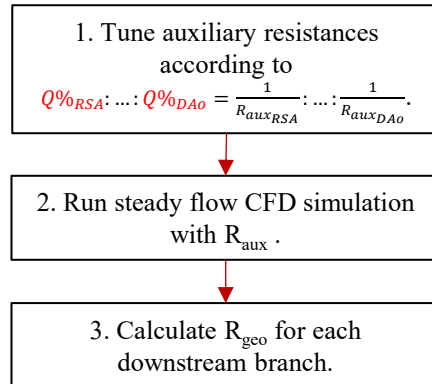
Find geometric resistance



Case	RSA R_{geo}	RCCA R_{geo}	DAo R_{geo}	LCCA R_{geo}	LSA R_{geo}
Steady	7.66E-3	3.45E-1	7.73E-2	6.96E-2	1.76E-1
Periodic	4.51E-4	3.31E-1	7.36E-2	6.67E-2	1.69E-1

Resistance unit in mmHg*s/ml

R_{geo} in both cases are similar, to simplify the algorithm, steady simulation is used.



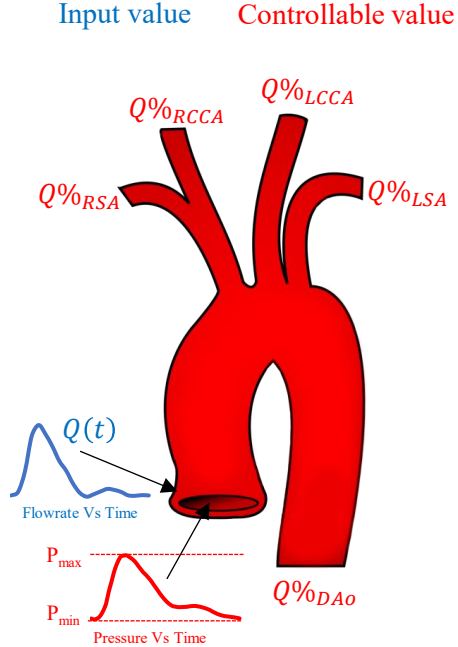
R_{aux} is to control flow distribution of each branch.

$$Q\%_{RSA} : Q\%_{RCCA} : \dots : Q\%_{DAo} = \frac{1}{R_{total_{RSA}}} : \frac{1}{R_{total_{RCCA}}} : \dots : \frac{1}{R_{total_{DAo}}}$$

$$R_{total} = R_{geo} + \boxed{R_{aux}} \uparrow$$

Find Windkessel model parameters

i denotes each branch.



1. Get total resistance for branch i :

$$R_{total_i} = \frac{P_{mean}}{Q_{mean_i}}$$

Where, $P_{mean} = \frac{1}{3}(P_{max} - P_{min}) + P_{min}$;

$$Q_{mean_i} = \frac{\int_0^T Q(t) dt}{T} * Q\%_i$$

2. Get Windkessel model resistance and max min pressure for branch i :

$$R_{WKM_i} = R_{total_i} - R_{geo_i}$$

$$P_{max_i} = P_{max} - Q_{mean_i} R_{geo_i}$$

$$P_{min_i} = P_{min} - Q_{mean_i} R_{geo_i}$$

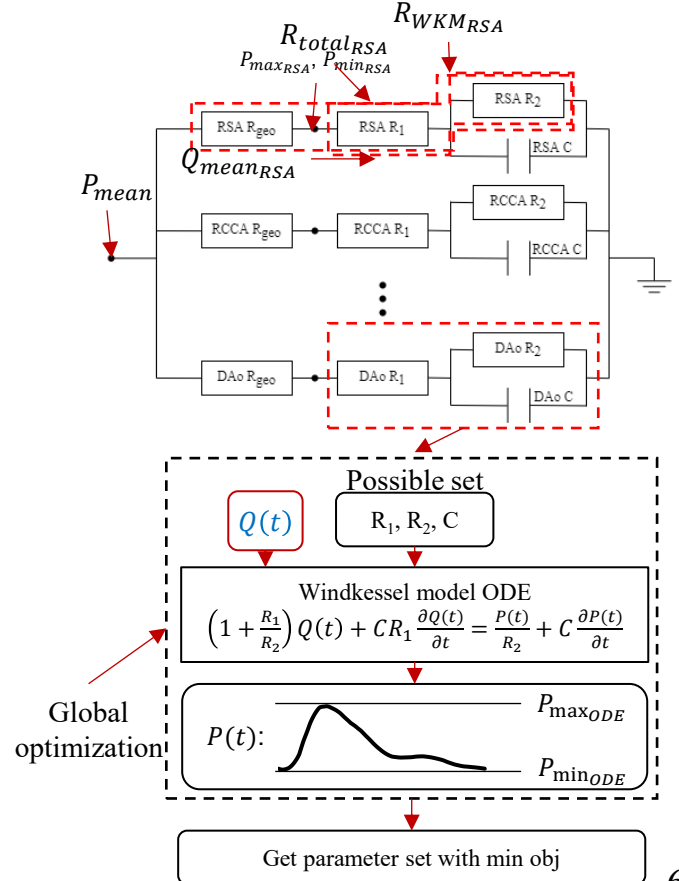
3. Find R_1 , R_2 , and C for branch i by the pattern search.

$$R_{1_i} = 0 \sim R_{WKM_i}; R_{2_i} = R_{WKM_i} - R_{1_i}; C_i = 0 \sim 3 \text{ml/mmHg}$$

Objective function:

$$obj = \sqrt{\frac{1}{2}[(P_{max_i} - P_{max_{ODE}})^2 + (P_{min_i} - P_{min_{ODE}})^2]}$$

Get a set of R_1 , R_2 , and C for all branches.



Simulation setup

Fluid settings:

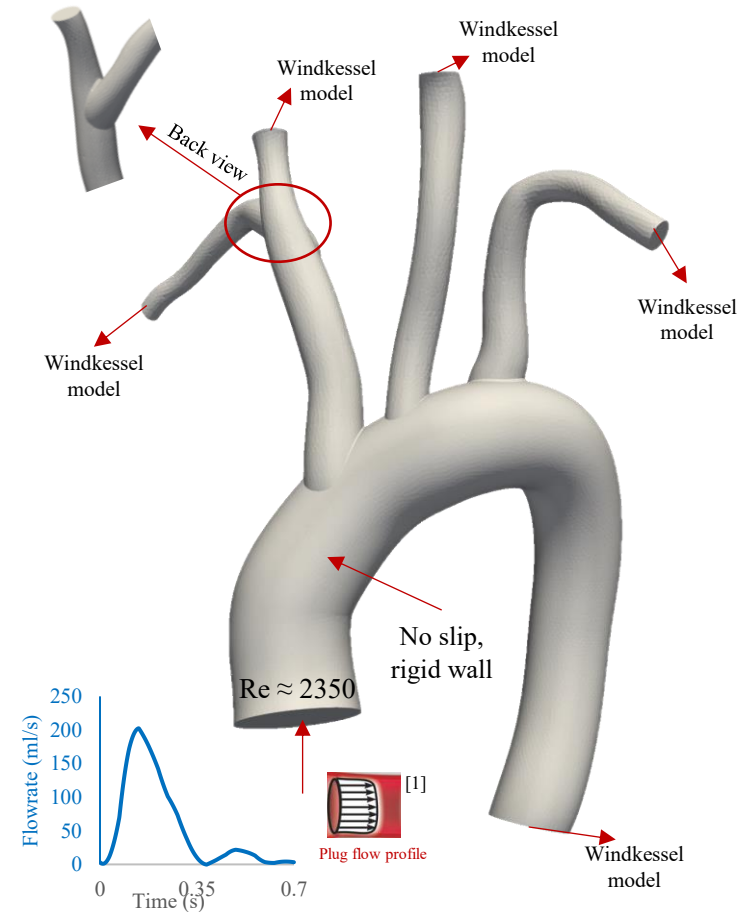
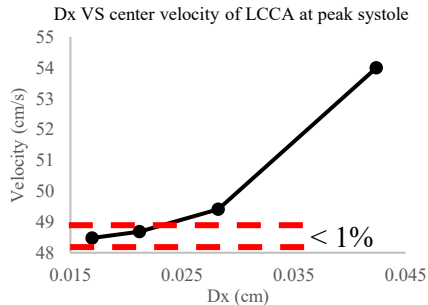
Newtonian fluid;
Density: 1060 kg/m^3 ;
Kinematic viscosity $3.3\text{e-}6 \text{ m}^2/\text{s}$.

Boundary condition:

No slip and rigid wall;
Plug profile periodic velocity inlet;
Flowrate variant pressure outlet with three element Windkessel model.

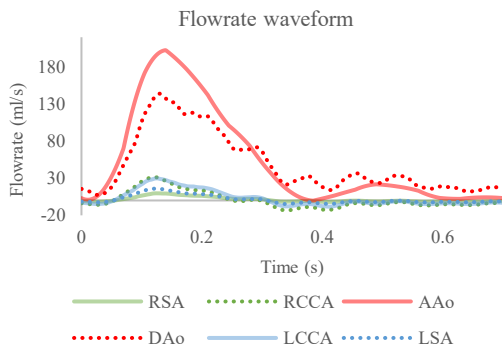
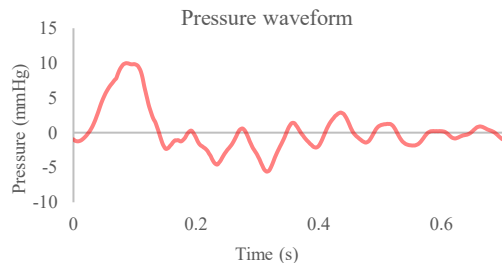
Mesh Independence study:

When $dx \leq 0.02 \text{ cm}$, the difference is smaller than 1%. In the simulation, there are about 21 million fluid cells.



Results

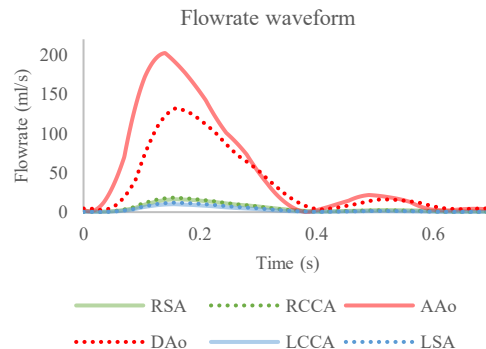
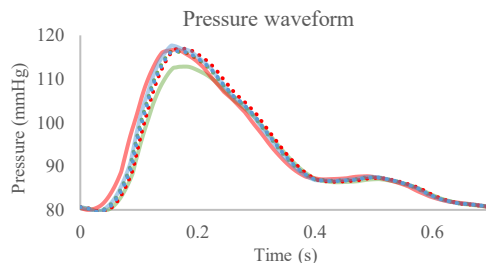
Constant zero pressure outlets



Flow distribution without Windkessel model

Branch	DAo	RSA	RCCA	LCCA	LSA
Distribution	90.06%	2.08%	0.84%	4.82%	2.48%

Our algorithm with Windkessel model



Flow distribution

Branch	Pre-set	Simulation
DAo	69.10%	68.81%
RSA	9.50%	9.36%
RCCA	9.80%	9.69%
LCCA	5.20%	5.17%
LSA	6.40%	6.33%
L2 norm error		0.16%

Pressure range

Unit: mmHg	Pre-set	Simulation
Max pressure	120	119.56
Min Pressure	80	79.60
L2 norm error		0.42

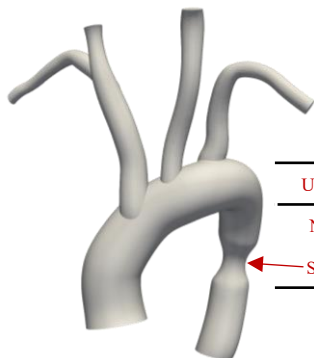
Windkessel model parameters

Branch	R ₁	R ₂	C
DAo	0.26	0.35	1.43
RSA	1.91	2.36	0.19
RCCA	1.85	2.45	0.20
LCCA	3.50	4.66	0.11
LSA	2.84	3.72	0.13
Resistance unit: mmHg*s/mL		Compliance unit: mL/mmHg	

$$L2 \text{ norm error} = \sqrt{\frac{1}{n} \sum_i (val_{sim} - val_{set})^2}$$

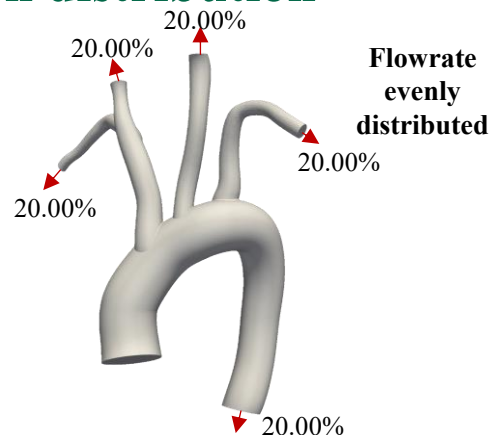
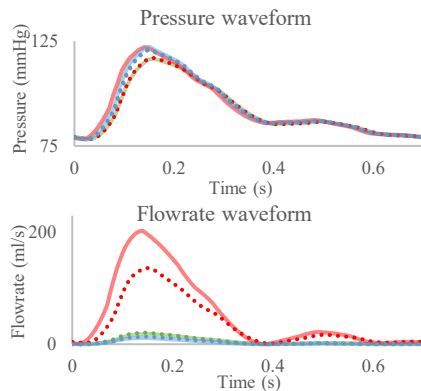
Where val_{sim} is the value in simulation;
 val_{set} is the pre-set value; n is the amount of branch.

Results for simulation with stenosis or even distribution



**Geometry
with stenosis**

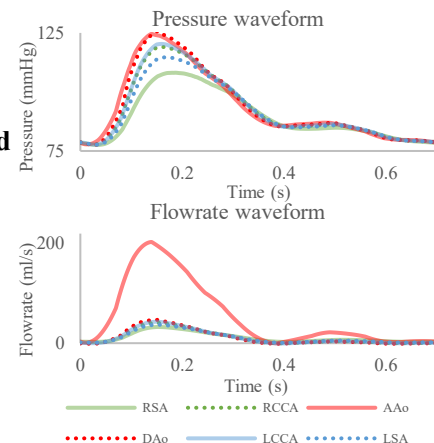
Unit: cm ²	Area
Normal	2.90
Stenosis	0.98



**Flowrate
evenly
distributed**

Flow distribution

Branch	Pre-set	Simulation
DAo	20.00%	20.38%
RSA	20.00%	19.98%
RCCA	20.00%	19.97%
LCCA	20.00%	20.13%
LSA	20.00%	19.79%
L2 norm error	0.20%	



Windkessel model parameters

Branch	R ₁	R ₂	C
DAo	0.91	1.21	0.42
RSA	0.89	0.89	0.38
RCCA	0.91	1.15	0.41
LCCA	0.91	1.15	0.41
LSA	0.90	1.05	0.40
Resistance unit:	mmHg*s/mL		
Compliance unit:	mL/mmHg		

Flow distribution

Branch	Pre-set	Simulation
DAo	69.10%	68.61%
RSA	9.50%	9.42%
RCCA	9.80%	9.75%
LCCA	5.20%	5.21%
LSA	6.40%	6.37%
L2 norm error	0.22%	

Windkessel model parameters

Branch	R ₁	R ₂	C
DAo	0.26	0.33	1.41
RSA	1.91	2.36	0.19
RCCA	1.85	2.45	0.20
LCCA	3.50	4.66	0.11
LSA	2.84	3.72	0.13
Resistance unit:	mmHg*s/mL		
Compliance unit:	mL/mmHg		

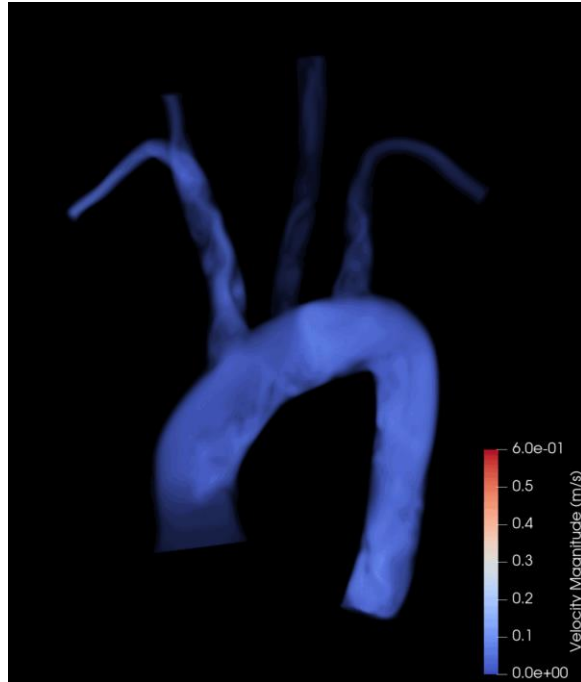
Pressure range

Unit: mmHg	Pre-set	Simulation
Max pressure	120	122.75
Min Pressure	80	78.76
L2 norm error	2.14	

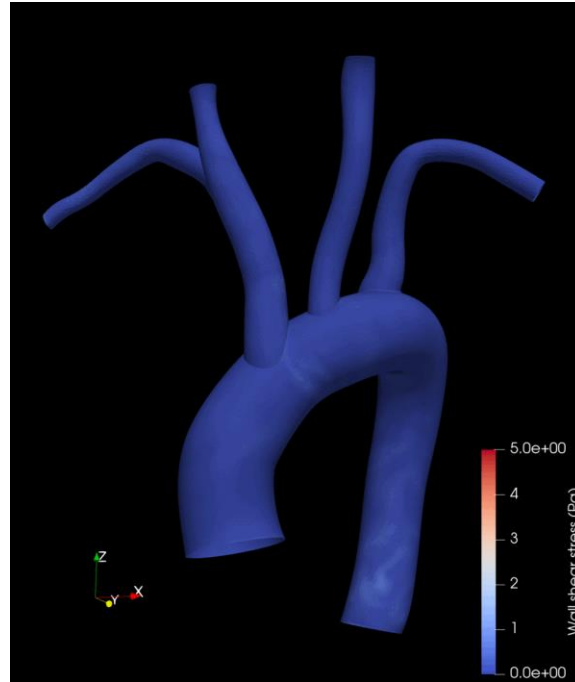
Pressure range

Unit: mmHg	Pre-set	Simulation
Max pressure	120	125.64
Min Pressure	80	78.49
L2 norm error	4.13	

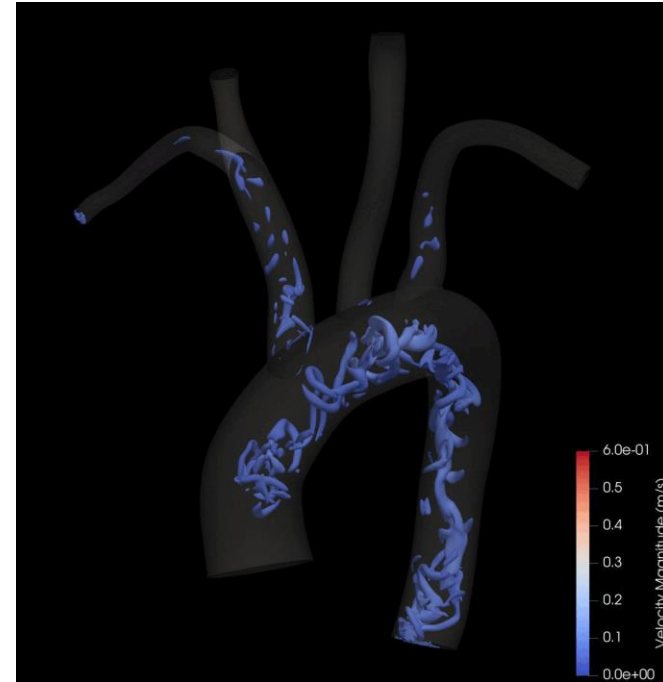
Results: animation from normal geometry with physiological flow distribution



Velocity magnitude field



Wall shear stress



Lambda 2



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Conclusion:

We developed a fast algorithm to approach optimal Windkessel parameters for patient-specific aortic flow simulations:

1. Find geometric resistances by a steady CFD simulation.
2. Get the optimized Windkessel model parameters by the global optimization with the consideration of R_{geo} .
3. Run periodic flow simulation with tuned Windkessel model.

By this algorithm, max, min pressure of ascending aorta and flow distribution of each downstream branch are controllable.



Thank you!

Appendix

Windkessel model parameters			
Branch	R_1	R_2	C
DAo	0.27	2.20	1.53
RSA	1.96	15.98	0.21
RCCA	1.90	15.50	0.22
LCCA	3.59	29.20	0.11
LSA	2.92	23.73	0.14
Resistance unit: mmHg*s/mL		Compliance unit: mL/mmHg	

For 80~120mmHg with normal
geometry and normal flowrate
distribution

Windkessel model parameters			
Branch	R_1	R_2	C
DAo	0.14	2.21	0.64
RSA	0.78	12.23	0.12
RCCA	0.89	13.80	0.10
LCCA	0.89	13.80	0.10
LSA	0.73	11.40	0.12
Resistance unit: mmHg*s/mL		Compliance unit: mL/mmHg	

Madhavan S, Kemmerling E M C. The
effect of inlet and outlet boundary
conditions in image-based CFD
modeling of aortic flow[J]. Biomedical
engineering online, 2018, 17(1): 1-20.