



VIIP: Central Nervous System (CNS) Modeling

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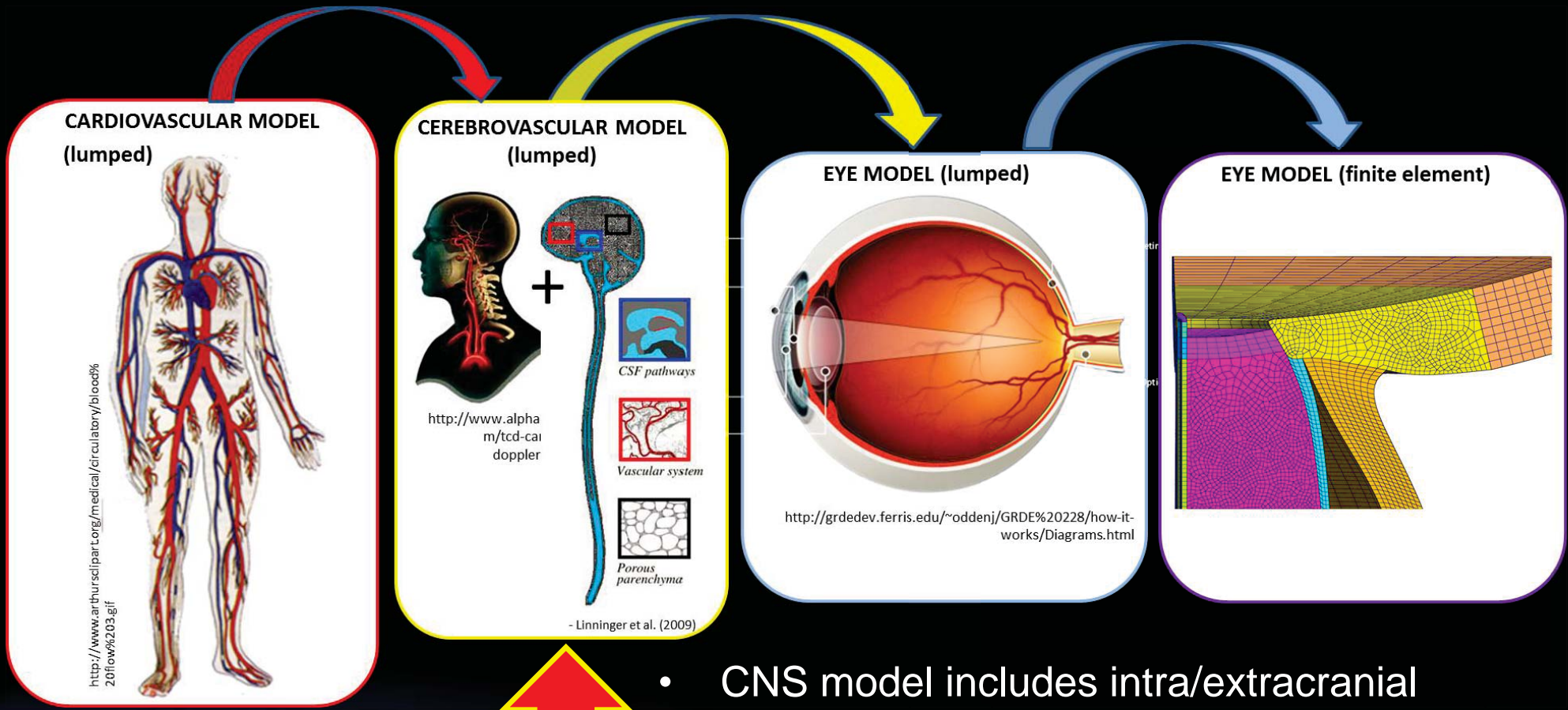
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Multiscale model for VIP research

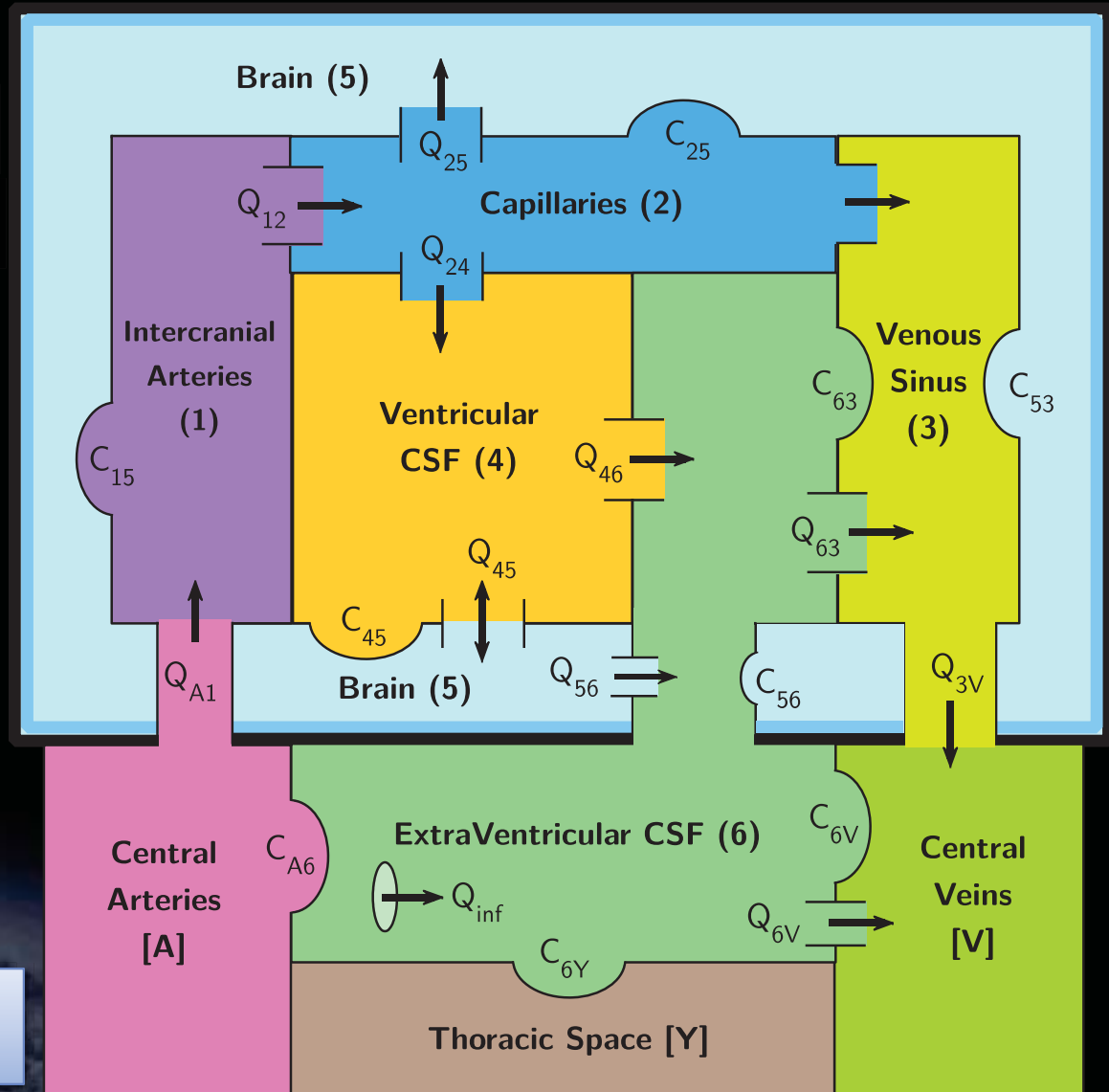


- CNS model includes intra/extracranial cerebrospinal fluid (CSF) and cranial blood compartments
- For details on other modules, see companion works for IWS2015 by Ethier et al., Feola et al., Nelson et al., and Price et al.



CNS Blood flow and pressure model

- Several lumped CNS models exist. Our starting point was a model that had been applied to microgravity (μg) (Stevens et al., 2005; Lakin et al, 2007):
- Time-dependent model composed of 6 fluid compartments (nodes)
 - 3 vascular:
 - Intracranial Arteries (1)
 - Capillaries (2)
 - Venous Sinus (3)
 - 2 cerebrospinal fluid
 - Ventricular CSF (4)
 - Extraventricular CSF (6)
 - 1 Brain node (5)
- Boundary conditions at cranium and whole-body interaction provided by extracranial nodes
 - Central Arteries [A]
 - Central Veins [V]
 - Thoracic Space [Y]



- Stevens et al. (2005)

Q = Flowrates between compartments (ml/min)
 C = Compartment compliance



Governing Equations

- Defining the pressures in the 6 compartments as dependent variables, the system is modeled in matrix form as a system of ordinary differential equations:

$$C \left[\frac{dP}{dt} \right] + ZP = S$$

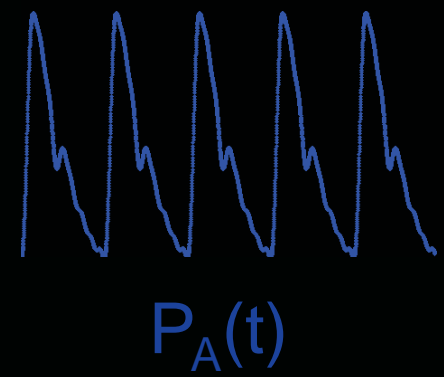
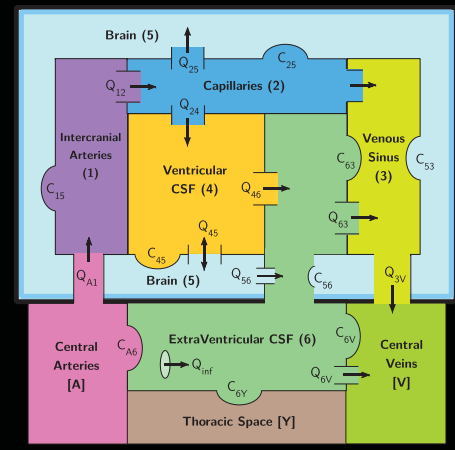
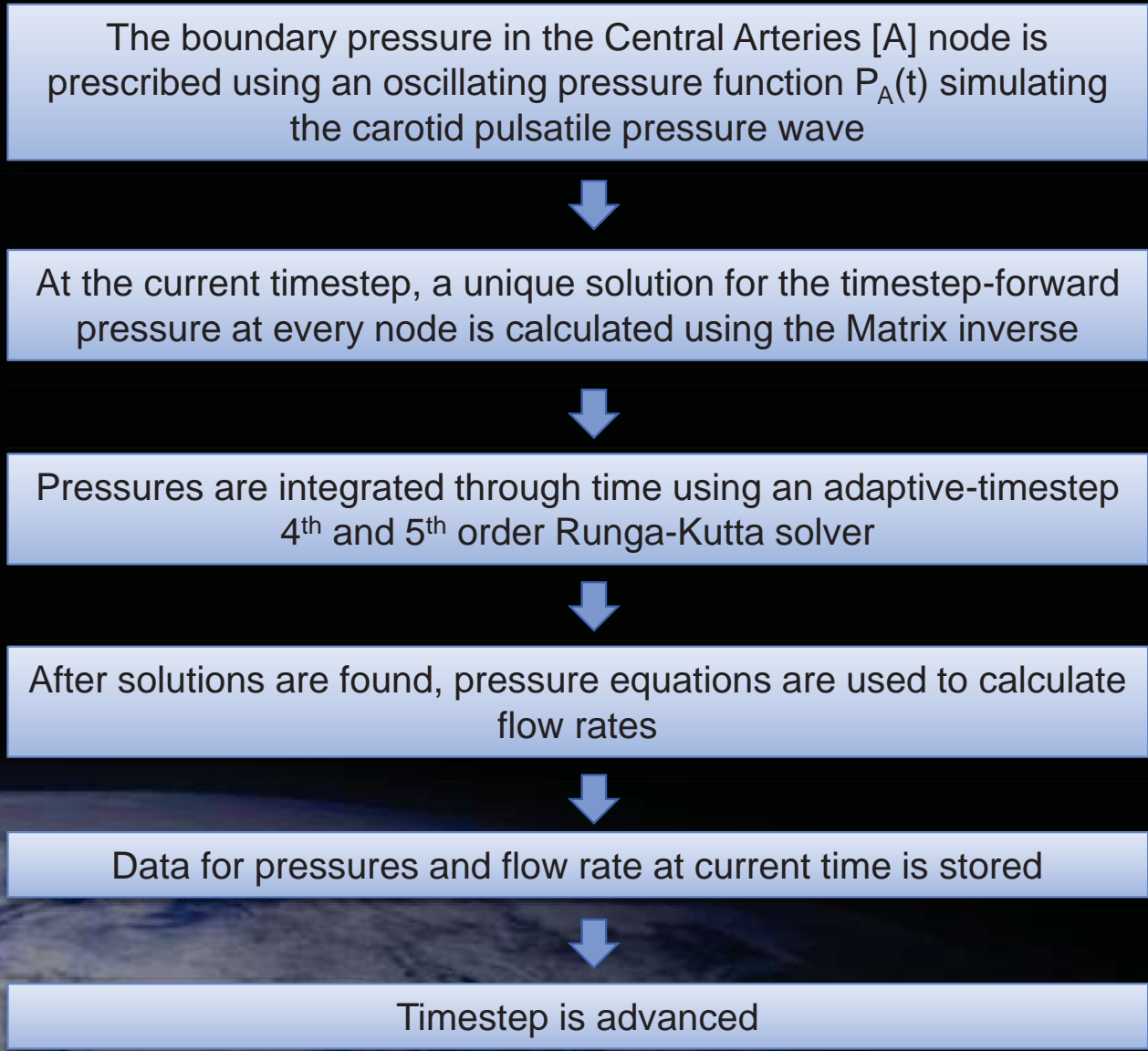
C_{15}				$-C_{15}$		dP_1/dt	Z_{A1}					P_1	$-Z_{A1}(P_A - G_{A1}\sin\theta) - Q_{12}$
	C_{25}			$-C_{25}$		dP_2/dt		$Z_{23} + K_{25}$	$-Z_{23}$		$-K_{25}$	P_2	$Q_{12} - Q_{24} - K_{25}\sigma_{25}(\pi_2 - \pi_5)$
		$C_{53} + C_{36}$		$-C_{53}$	$-C_{36}$	dP_3/dt		$-Z_{23}$	$Z_{23} + Z_{63} + Z_{3V}$		$-Z_{63}$	P_3	$Z_{3V}(P_V - G_{3V}\sin\theta)$
			C_{45}	$-C_{45}$		dP_4/dt				$Z_{45} + Z_{46}$	$-Z_{45}$	P_4	Q_{24}
$-C_{15}$	$-C_{25}$	$-C_{53}$	$-C_{45}$	$C_{25} + C_{45} + C_{53} + C_{36} + C_{15}$	$-C_{56}$	dP_5/dt		$-K_{25}$		$-Z_{45}$	$K_{25} + Z_{45} + Z_{56}$	P_5	$-K_{25}\sigma_{25}(\pi_2 - \pi_5)$
		$-C_{36}$		$-C_{56}$	$C_{56} + C_{36} + C_{6V} + C_{6Y} + C_{A6}$	dP_6/dt			$-Z_{63}$	$-Z_{46}$	$-Z_{56}$	P_6	$-C_{A6}(d_{PA}/dt) + C_{6Y}(d_{PY}/dt) + C_{6V}(d_{PV}/dt) + Z_{6V}(P_V)$

- | | |
|--------------------------|---------------------------------|
| C compliance | Z fluidity ~ 1/resistance |
| G gravity | θ tilt angle |
| K filtration coefficient | π osmotic pressure |
| P pressure | σ reflection coefficient |
| Q flow rate | |
| S source/forcing terms | |

Note that G is explicitly included in the forcing terms in S



MATLAB Implementation





Verification Tests

- 20 independent verification tests that included variation in hydrostatic pressure
- 3 independent users of the code

TEST

- ✓ Short-term head down tilt (HDT)
- ✓ Long-term HDT
- ✓ Microgravity
- ✓ Blood-brain barrier influence

Verification tests also had a validation component

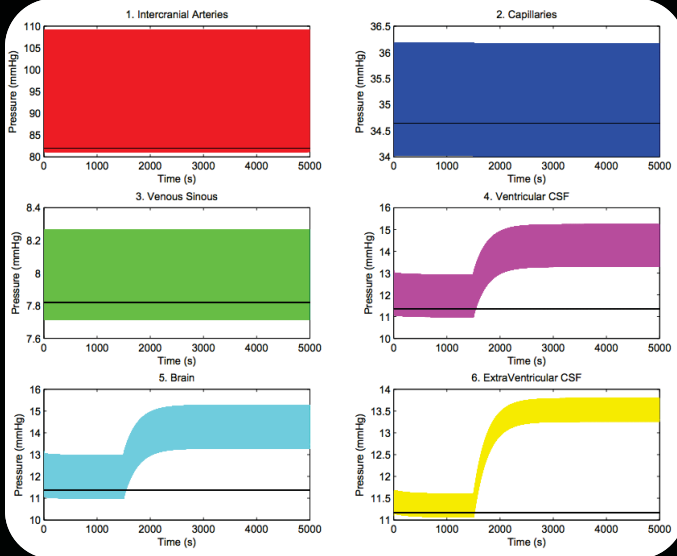
- Used Lakin and Stevens equation structure and parameters, but
- Developed independent implementation, arterial pressure that drives unsteady response and solution methodology



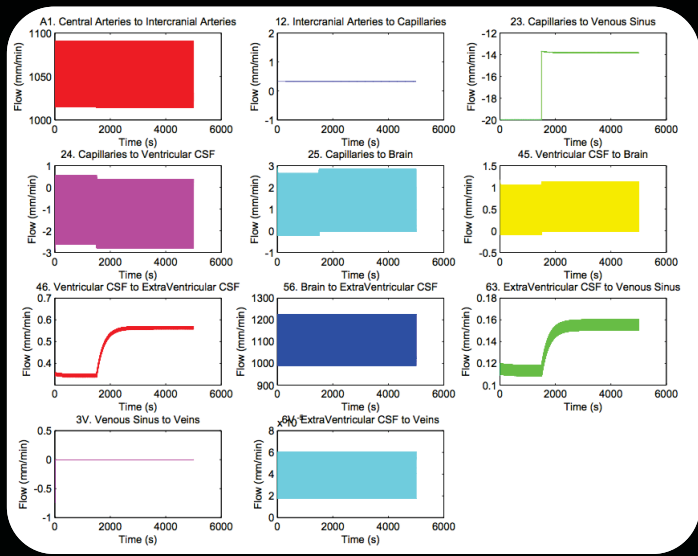
Short-term head down tilt

- Tests called for monitoring of changes in pressure differences pre- to post-tilt:

PRESSURES



FLOWRATES



Tilt angle (°)	$\Delta(P_s - P_v)$ (mmHg)			Δ ICP (mmHg)		
	Expt [1]	Model [2]	Our Model	Expt [3]	Model [2]	Our Model
-6				3.3	2.10 to 3.70	2.20
-10	3.1	3.86	3.66			
-15				6.1	5.18 to 7.78	5.46

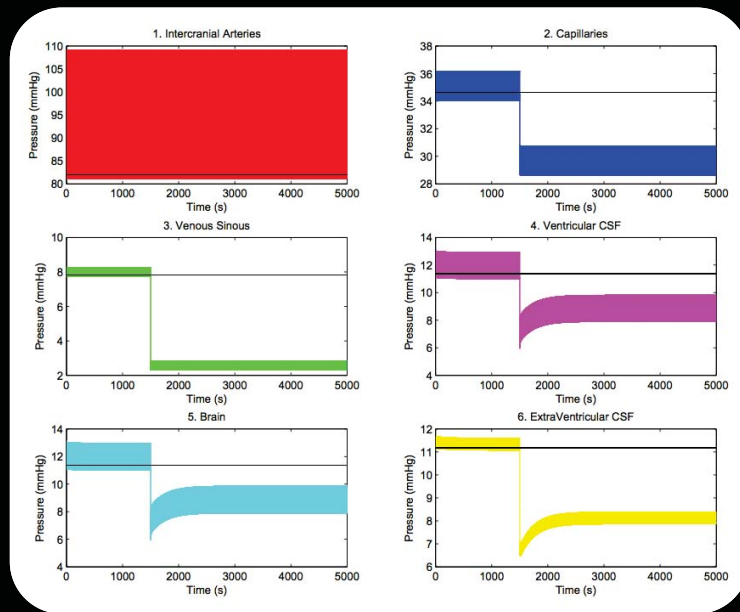
Conclusion: The current model agrees with prior experimental and numerical work.

[1] Katkov and Chestukhin (1980)
 [2] Stevens et al. (2005)
 [3] Murthy et al. (1992)

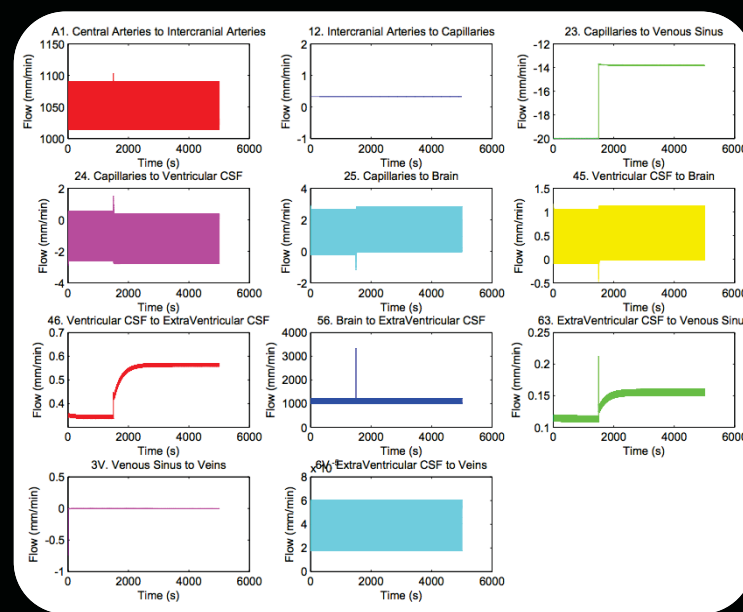


Long-term HDT and Microgravity

PRESSURES



FLOWRATES



Condition	Δ ICP (mmHg)	
	Model [1]	Our Model
Long-term HDT	4.9	4.9
μ g	<0	-3.1

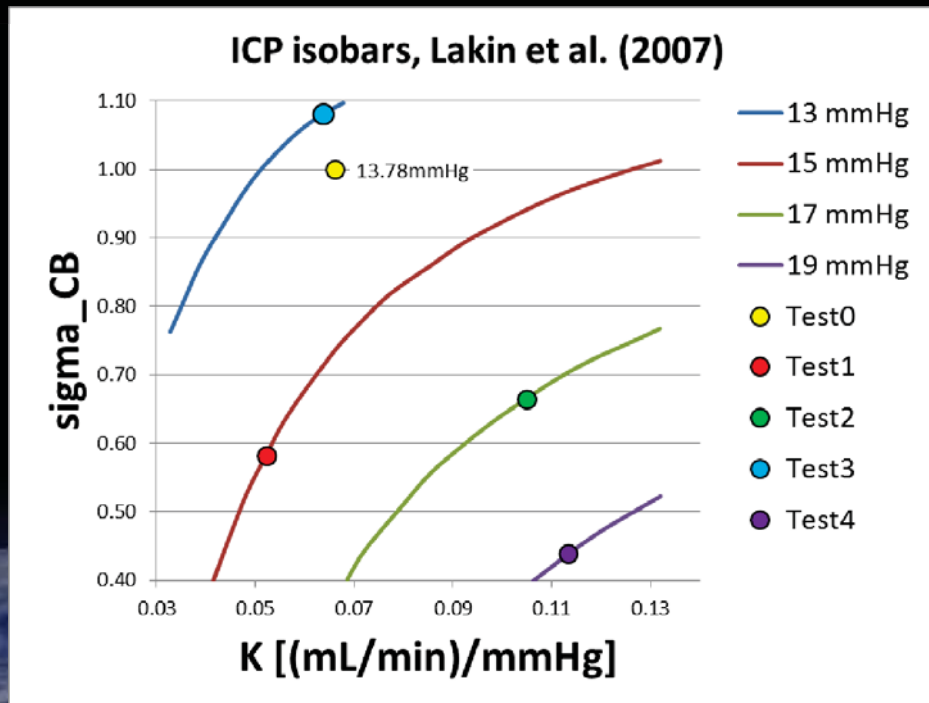
Conclusion: *Using their parameters, our predicted Δ ICP is consistent with the prior model in μ g and long-duration HDT, but are their parameters correct?*

[1] Stevens et al. (2005)



Blood-brain barrier influence

- Later work by the Stevens/Lakin team hypothesized that the blood/brain barrier might weaken in μg
- In Lakin et al. (2007), they performed a sensitivity study for a hypothetical change in the reflection and filtration coefficients
- This ***changed their findings*** on ICP in μg



	sigma	K	Simulated ICP (mmHg)	Target ICP (mmHg)	Error (%)
Test0	1.000	0.066	13.78		
Test1	0.583	0.052	15.15	15	1
Test2	0.665	0.105	17.18	17	1.06
Test3	1.081	0.064	13.23	13	1.77
Test4	0.438	0.113	19.14	19	0.74

Conclusion: Our model agrees with literature results to within 2% or better.

Revising prior findings, authors concluded that ICP could increase in μg .
But how do we assess the credibility of this claim?



Preparing for μg simulations

- Before weighing in on the potential change in ICP in μg , we need to:
 - Re-assess parameters used by Lakin/Stevens based on the most current VIIP research
 - Quantify uncertainty in model parameters
 - Define a physiological envelope for parameters that will be relevant for the astronaut corps on orbit
 - Perform sensitivity studies over a much larger parameter space
 - Examine model predictions against independent studies in HDT, μg , and postural change, particularly for chronic conditions. We need our model to do a good job in predicting:
 - Volumes of intra/extracranial CSF compartments
 - Volumes of intracranial blood compartments
- Only after these steps are taken can we make intelligent predictions about μg response



Sensitivity analysis

- We are analyzing this system by testing model sensitivity
 - Parameters include: compliances, resistances and filtration coefficients
 - Each described by statistical parameters
 - Mean and range of variation (variance)
 - Distribution of variation (density function)
- Methodology
 - Partial Rank Correlation Coefficient (PRCC) Analysis
 - Provides the linear relationships between two variables
 - one input parameter and one output parameter
 - All linear effects of other variables are removed after rank transformation
 - Rank Transformation: transforms nonlinear monotonic relations to linear
- Latin Hypercube sampling
 - Efficient method to randomly characterize the sets of combined parameters
 - Many independent runs with randomly chosen parameter sets provide statistics on the system response



Conclusions

- A CNS lumped parameter model has been produced based on the model developed by Lakin and Stevens
 - Our solution methodology and computational platform is unique
- Our model has been tested and verified
 - ICP predictions agree with Lakin/Stevens in 20 cases of acute and chronic μg and HDT
- CNS model infrastructure is complete, but additional work is needed
 - Re-assess parameters used by Lakin/Stevens
 - Define flight and flight analog derived parameter ranges
 - Perform parameter sensitivity studies
 - Validate against the latest VIIP research
- In the future this model will be
 - integrated with lumped CVS and eye models
 - Used to establish spaceflight responses with fidelity sufficient to supply boundary conditions for more complex VIIP eye simulations.