

Comparison of Multiple Large Fluid-Structure Interaction Simulations in Virtual Reality

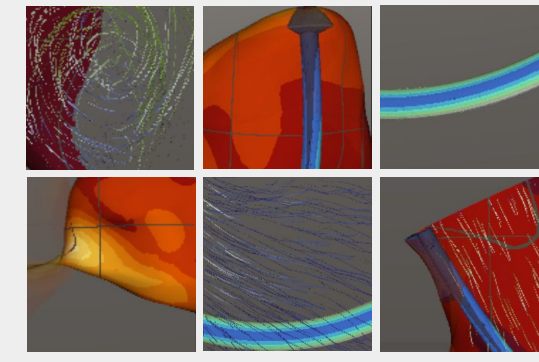
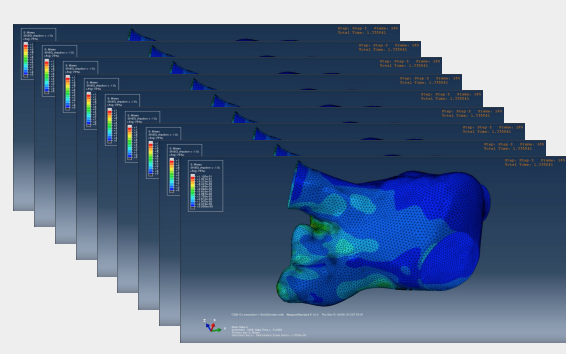
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Introduction and Motivation

How does the stiffness and length of a cardiac lead affect stress and blood flow in the right atrium of the heart [5]?

1) Run Simulations:

- Vary lead length and stiffness parameters
- 10 Instances > 39 GB
- Fluid-Structure Interaction (FSI)



2) Explore 4D Relationships:

- Compare multiple spatial locations simultaneously.
- Full variable access across space & time for all instances.

- How can we analyze the **parameter space** [6,7]?
- Visual relationships in **multimodal scenarios** like FSI are rarely explored [2].

- The data is 10 times larger than the GPU memory (**4 GB < 39 GB**), which requires sophisticated **sampling & rendering** strategies [1].

- **Opportunity for VR** - more than a stereoscopic view [3], and must have **high frame rates**.
- There are only a few tools that **compare multiple large instances in VR** [3,8].

Case Study: FSI comparison in VR for data > GPU memory

Goal: VR Comparative Analysis

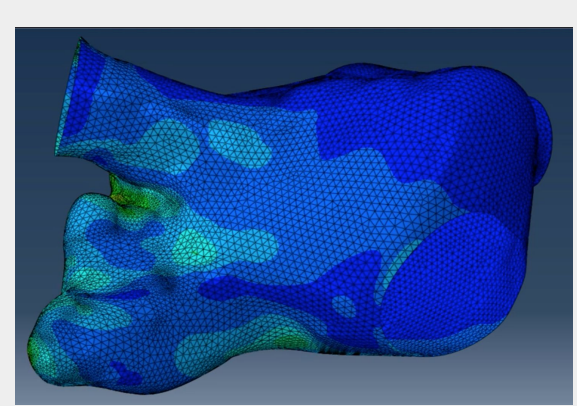


Bento Box: A VR data visualization technique and bimanual 3D interface for exploratory analysis of 4D data ensembles. Comparison is conducted via a grid, with a column per data instance, and a row per view. Views are selected and reframed via 3D gestures, and colormaps can be picked for each view.

Methods

Solid Domain

- 1) Store one triangular mesh with solid-specific variables in GPU buffers.

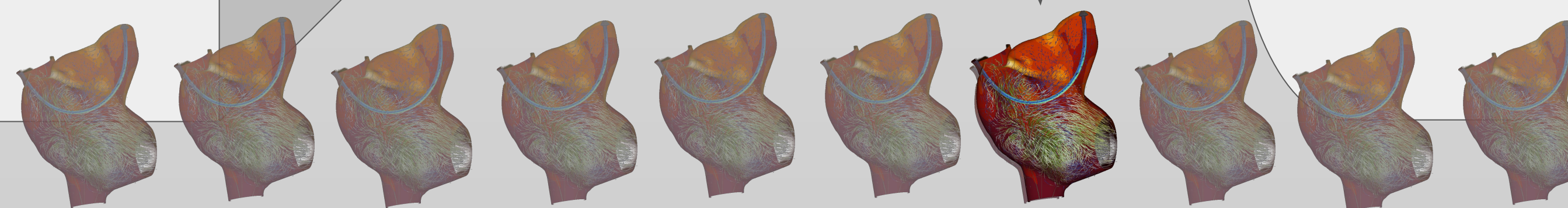
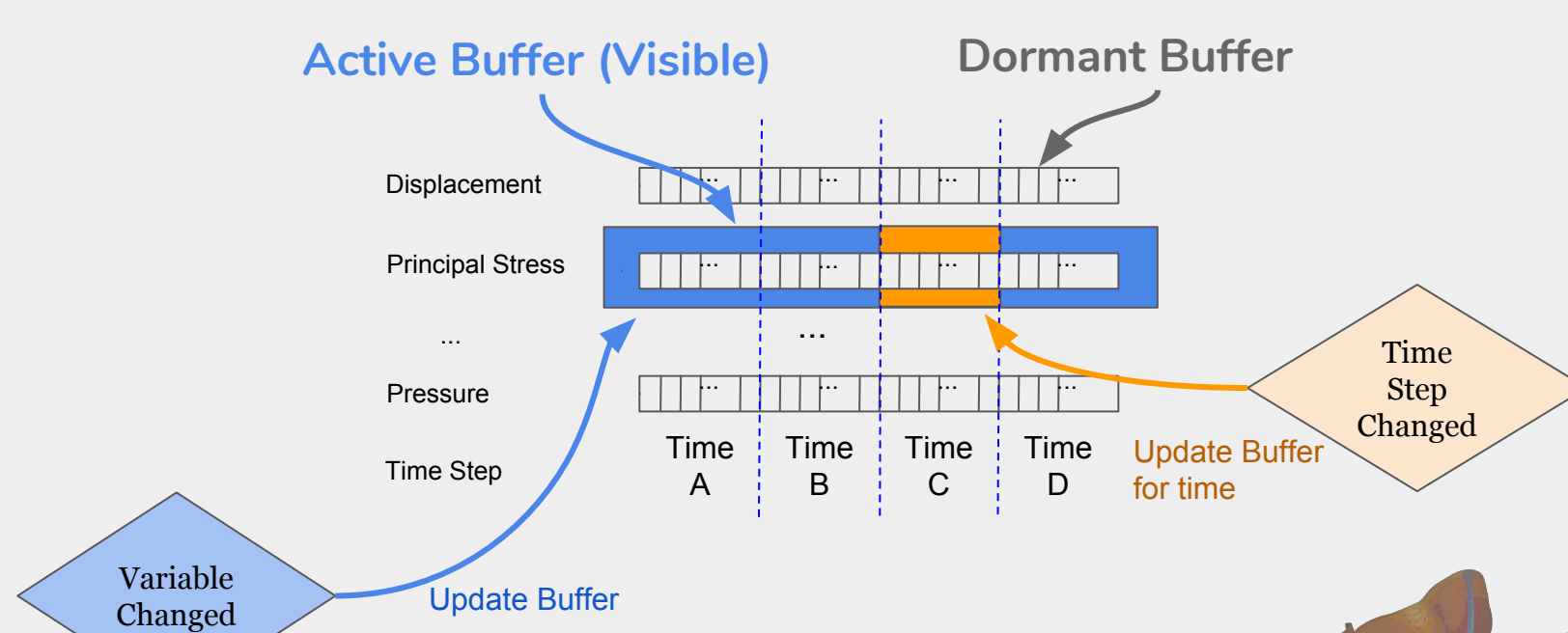


- Displacement
- Principal Stress
- Pressure
- von Mises
- ...
- Stress Tensor

One buffer per variable, where the size is the number of comparable timesteps times the number of vertices. Optimized for static comparison[4].

Render Solid

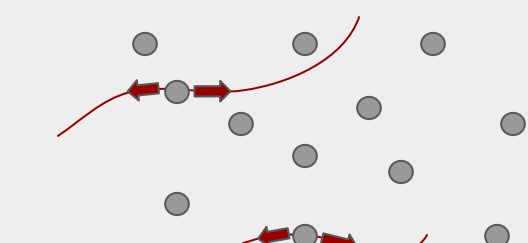
- 2) Update buffers based on interaction.



Render Fluid

Fluid Domain

- 1) Create sampled pathlines from seeds

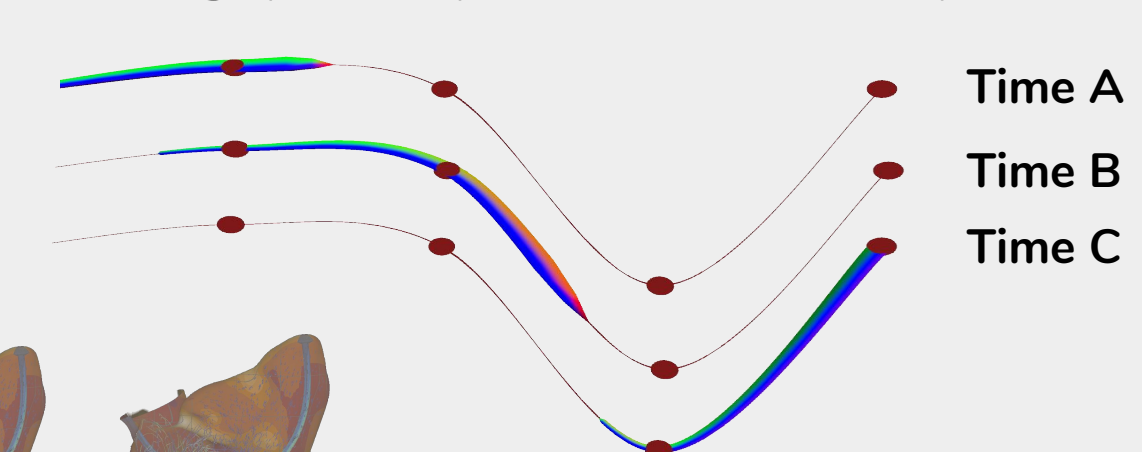


- 2) Store all pathlines and one particle mesh on the GPU

Path	Positions	Velocities
1	(1,1,1)...(0,0,0)	(1,1,1)...
2	(3,2,1)...(10,1,0)	(2,2,2)...
...
n	(3,2,1)...(10,1,0)	(5,5,5)...

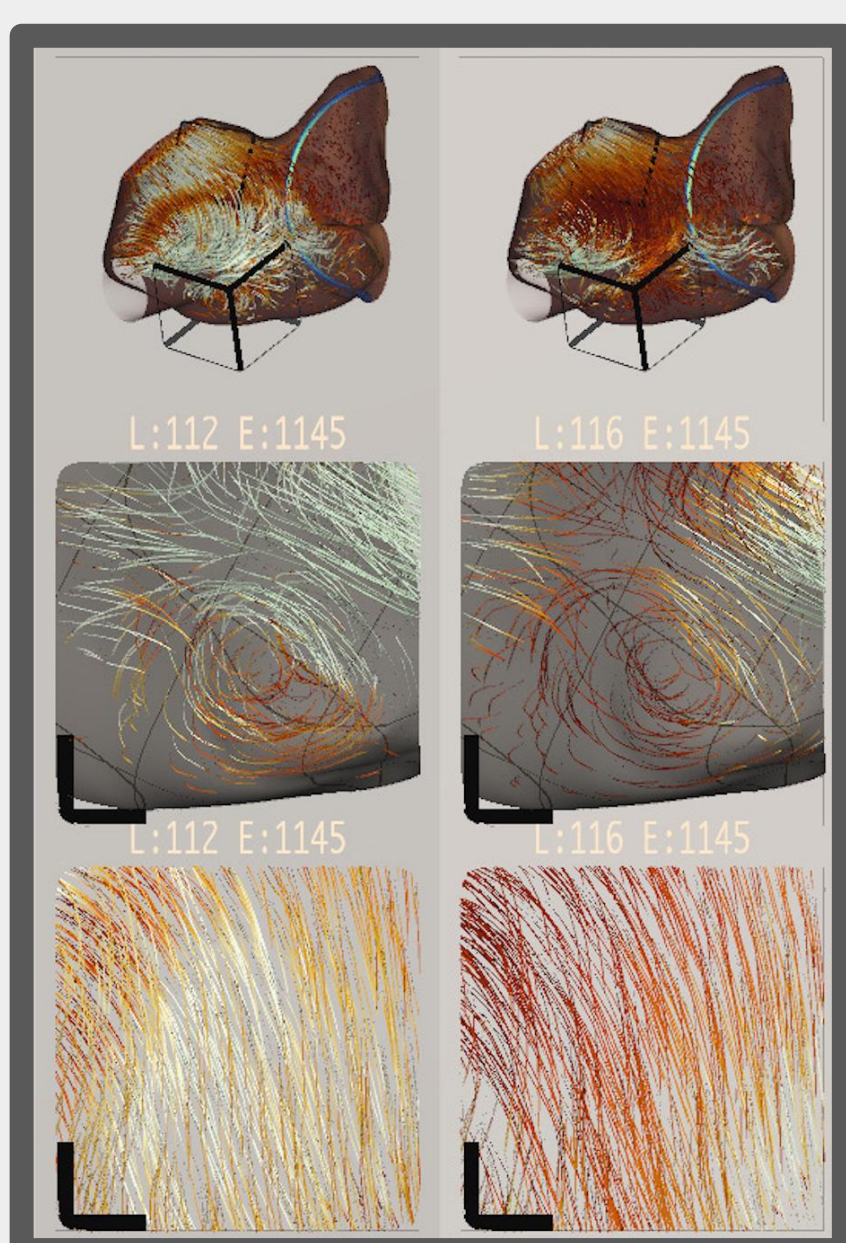


- 3) Instance render the mesh and morph along path points via a spline

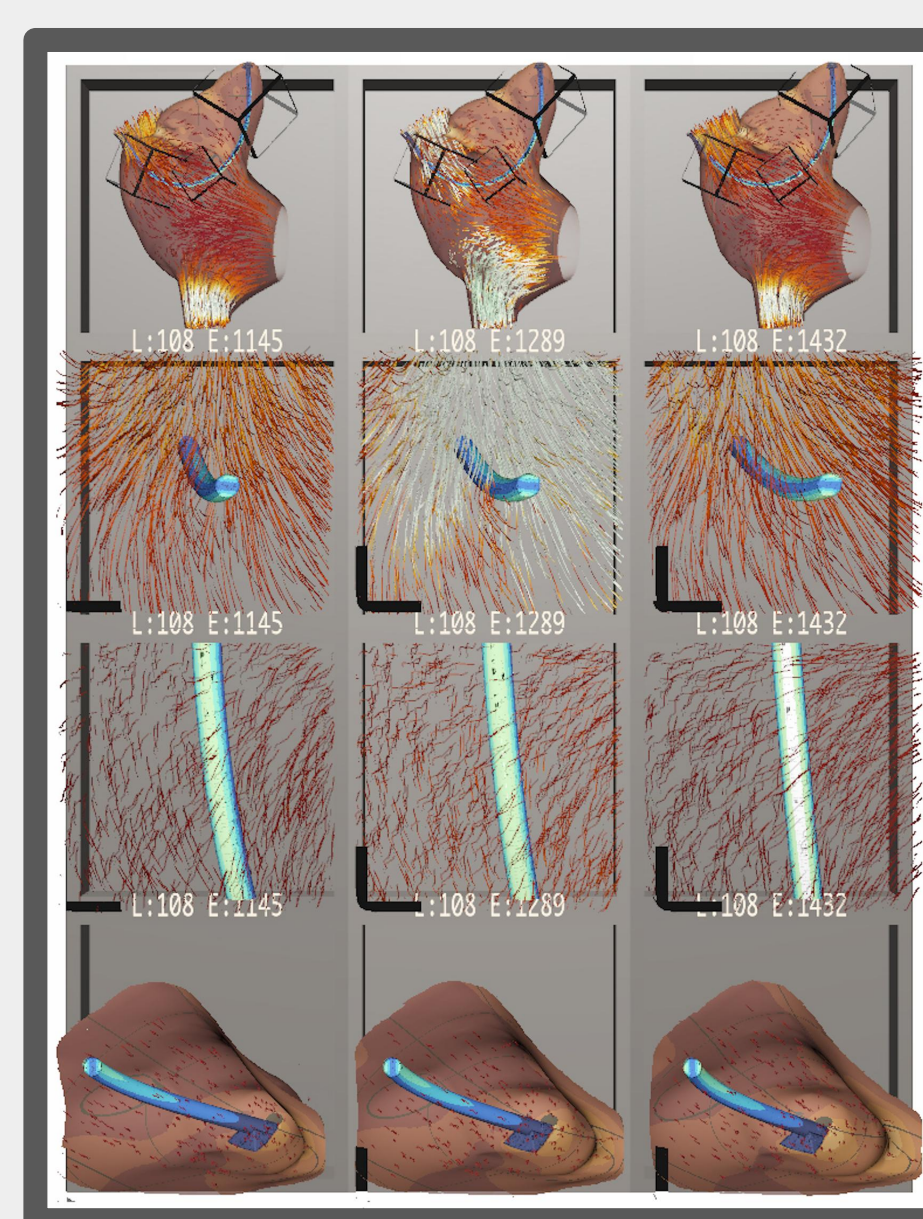


Results

FSI Instance Comparison Using Immersive Virtual Reality Techniques



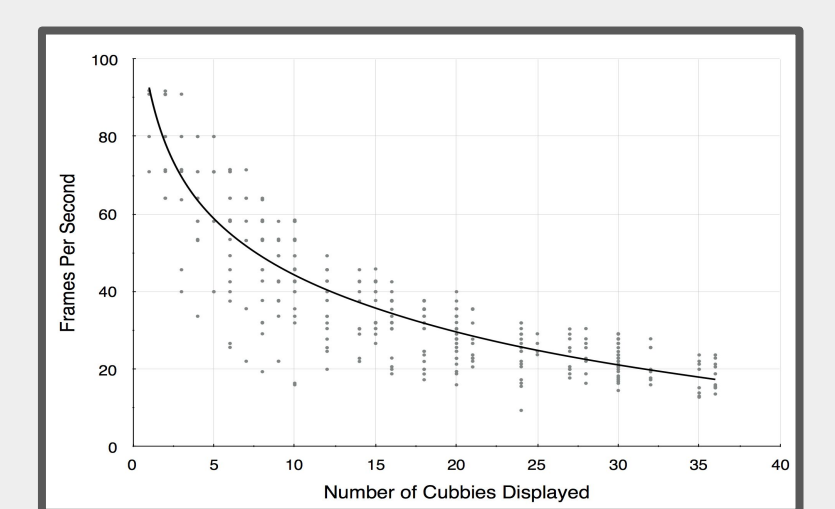
A comparison of the effect of two different lead lengths on the **flow rate** in two regions of the heart. The longer lead length (right) correlates to slower flow on both sides of the heart, indicated by the darker paths.



A comparison of leads of three different **stiffness** at three points along each lead. (Bento Box shows these views as black boxes in the top row.) The second row shows a region where the flow rate is greatest near the lead of medium stiffness.

At Interactive Rates

We performed a systematic sampling of Bento Box grid configurations for this 10-instance data ensemble. The trend is above **30 frames-per-second** for Bento Box Configuration of 20 cells or less, and is in the **40-50 frames-per-second** range for smaller arrangements.



Timings were recorded on a 4 core processor Intel(R) CORE(TM) i7-7700HQ CPU @2.80GHz machine with 16 GB of RAM and a NVIDIA GeForce GTX 1070 graphics card. An HTC Vive was used with a 2160 x 1200 resolution.

For Large Data

After processing the **39 GB of raw data**, the amount of memory needed to accurately visualize the solid and fluid domain is over 8 GB, exceeding a 4 GB GPU hardware limit. Our sampling, rendering, and streaming strategies enable a low memory footprint of **1.2 GB on the GPU** at any time.

ID	Raw (MB)	Solid (MB) Processed	Fluid (MB) Proc. / GPU	Total (MB) Processed	Total (MB) GPU
108-1145	5,252.6	892.6	7.0	115.2	1,007.8
108-1289	4,724.4	682.2	7.0	115.2	797.4
108-1432	4,911.5	682.2	7.0	115.2	797.4
110-1145	2,913.7	660.9	6.7	115.2	776.2
110-1289	2,913.7	660.9	6.7	115.2	776.2
110-1432	2,913.7	660.9	6.7	115.2	776.2
112-1145	4,934.6	687.5	7.0	115.2	802.7
112-1289	4,934.6	687.5	7.0	115.2	802.7
112-1432	4,934.6	687.5	7.0	115.2	802.7
116-1145	1,588.9	832.4	6.9	115.2	947.6
Average	40,073.6	7,134.7	69.1	1,152.3	8,287.9
Limit					1,221.4

There is a 4 GB GPU hardware limit for our 4-wall cave environment, a 2 processor Intel(R) Xeon(R) CPU E5-2640 @2.50GHz machine with two NVIDIA Quadro K5000 cards and 192 GB of RAM

Conclusions

- Our **Bento Box** extension enables comparison of **multiple large FSI simulations** at **interactive rates**.
- The GPU memory required for our approach implies that the number of instances can **scale beyond 10 instances**.
- We hope that this case study can inspire more work in combining **large data visualization techniques** with **interactive exploration**.

References

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Acknowledgments

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