

Subspace Fluid Re-Simulation Theodore Kim and John Delaney University of California, Santa Barbara

200 x 266 x 200 simulation, 7 hours 2 minutes

Original MacCormack Simulation

Vorticity Confinement set to zero

Outline

- Previous Work
- Subspace Basics
- The Cubature Approach
- Other Features
- Results
- Discussion and Conclusions

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Fluid Simulation



[Foster and Metaxas 1997]



[Stam 1999]

Semi-Lagrangian Advection Implicit Integration

Vortex Methods



[Brochu et al. 2012] [Pfaff et al. 2012]

Viscosity Methods



[Mullen et al. 2009]

Adaptive Methods





[Losasso et al. 2004] [Ando et al. 2013]

Turbulence Methods



[Kim et al. 2008] [Schechter and Bridson 2008] [Narain et al. 2008]

Subspace Methods



[Pentland and Williams 1989]



[Barbic and James 2005]



[Kim and James 2011]



[Harmon and Zorin 2013]

Subspace Methods



[Treuille et al. 2006] [Wicke et al. 2009] [Stanton et al. 2013]



[Treuille et al. 2006]





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Diffusion



Diffusion



 $\mathbf{U} \in \mathbb{R}^{r \times N} \quad r \ll N$



Stable Fluids

- Diffusion
- Projection
- Advection



Diffusion



Semi-Lagrangian Advection



Diffusion



Semi-Lagrangian Advection



 $(\mathbf{u} \cdot \nabla)\mathbf{u}$

Finite Difference Advection [Treuille et al. 2006]





Finite differences, *not* Semi-Lagrangian

Exponential, *not* implicit

Semi-Lagrangian Advection

Semi-Lagrangian Advection



[Stanton et al. 2013]

Enables \div and $\sqrt{}$

Semi-Lagrangian Advection

Finite Difference Advection





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Optimizing Cubature for Efficient Integration of Subspace Deformations [An et al.] 2008







 $\mathbf{f}(\mathbf{x})$







$$\tilde{\mathbf{f}}(\mathbf{x}) \approx \sum_{i=1}^{n} w_i \, \tilde{\mathbf{f}}(\mathbf{x}_i)$$



$\mathbf{f}(\mathbf{x}) = \mathbf{A}(\mathbf{u})$





The Greedy Algorithm










Non-Negative Least Squares





Lawson-Hanson NNLS solve: $O(P^3)$

Greedy search for P cubature points: $O(P^4)$













Fig. 11a. Distribution pattern of jittered samples.



Fig. 11b. Fourier transform of the pattern in Figure 11a.

[Cook 1986] [Pharr and Humphreys 2010] ... many others









$$PDF(\mathbf{x}_p) = R\left(\frac{\left|\mathbf{a}_p \cdot \mathbf{r}\right|}{\mathbf{r} \cdot \mathbf{r}}\right)$$



	Example 1	Example 2	Example 3	Example 4
L_2 error, Iteration 1	0.0428803	0.0592719	0.0409149	0.0330917
L_2 error, Iteration 2	0.0148716	0.0184463	0.0145316	0.0118481
L_2 error, Iteration 3	0.0107379	0.0112989	0.0106133	0.00650847
L_2 error, Iteration 4	0.00866083	0.00865156	0.00871744	converged
Total Time	01h 18m 07s	03h 05m 58s	09h 28m 29s	05h 29m 02s

Greedy search for P cubature points: $O(P^4)$

Importance sampled cubature: $O(P^3)$

[Harmon and Zorin 2013]



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Fast Diffusion-Projection

$$\widetilde{\mathbf{u}}_{2} = \widetilde{\mathbf{D}}\widetilde{\mathbf{u}}_{1}$$
$$\widetilde{\mathbf{d}} = \widetilde{\mathbf{W}}\widetilde{\mathbf{u}}_{2}$$
$$\widetilde{\mathbf{p}} = \widetilde{\mathbf{X}}^{-1}\widetilde{\mathbf{d}}$$
$$\widetilde{\mathbf{u}}_{t+1} = \widetilde{\mathbf{u}}_{2} + \widetilde{\mathbf{Y}}\widetilde{\mathbf{j}}$$

Fast Diffusion-Projection



Fast Diffusion-Projection

 $\tilde{\mathbf{u}}_{2} = \tilde{\mathbf{D}}\tilde{\mathbf{u}}_{1}$ $\tilde{\mathbf{d}} = \tilde{\mathbf{W}}\tilde{\mathbf{u}}_{2}$ $\tilde{\mathbf{p}} = \tilde{\mathbf{X}}^{-1}\tilde{\mathbf{d}}$ $\tilde{\mathbf{u}}_{t+1} = \tilde{\mathbf{u}}_{2} + \tilde{\mathbf{Y}}\tilde{\mathbf{p}}$ $\tilde{\mathbf{u}}_{t+1} = \tilde{\mathbf{u}}_{2} + \tilde{\mathbf{Y}}\tilde{\mathbf{p}}$

Internal Obstacles



Iterated Orthogonal Projection

[Molemaker et al. 2008]



Iterated Orthogonal Projection

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Iterated Orthogonal Projection

[Molemaker et al. 2008]

 $\widetilde{\mathbf{u}}_{t+1} = \widetilde{\mathbf{Z}}\widetilde{\mathbf{u}}_1$ $\widetilde{\mathbf{u}}_{t+1} = \widetilde{\mathbf{Z}}\widetilde{\mathbf{P}}\widetilde{\mathbf{u}}_1$

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Stam Plume, 200×266×200 06h 57m 30s

Original Semi-Lagrangian Simulation

Buoyancy Constant Halved



Stam Plume example

Solver Only:18ms9326x fasterWith Vel. Recon.:4.2s39x faster

Total preprocessing:

09h 50m 23s

12-core, 2.66 Ghz Mac Pro



MacCormack example

Solver Only:96ms1764x fasterWith Vel. Recon.:5.6s30x faster

Total preprocessing:

09h 27m 16s


03h 35m 00s



Dirichlet example

130ms 661x faster Solver Only: With Vel. Recon.: 5.1s 17x faster

Total preprocessing: 19h 00m 58s





Vorticity confinement = 20, originally 1.5

Neumann example

Solver Only:34ms2435x fasterWith Vel. Recon.:5.7s14x faster

Total preprocessing:

18h 53m 55s

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Contributions

- Fast *re-simulation* of an existing simulation
- A *cubature* approach to subspace advection
- Practical cubature training via *importance sampling*
- Internal obstacles via subspace *iterated orthogonal projection*

Limitations

- Memory intensive (Mac Pro had 96 GB)
- Time-consuming pre-process
- How well does it generalize?

Future Work

• Basis enrichment (XFEM?)

• Better basis compression (HSS?)

• Liquid re-simulation?



[Richardson et al. 2011]



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MacCormack, Vorticity Confinement = 6

Thank you



Tbank you