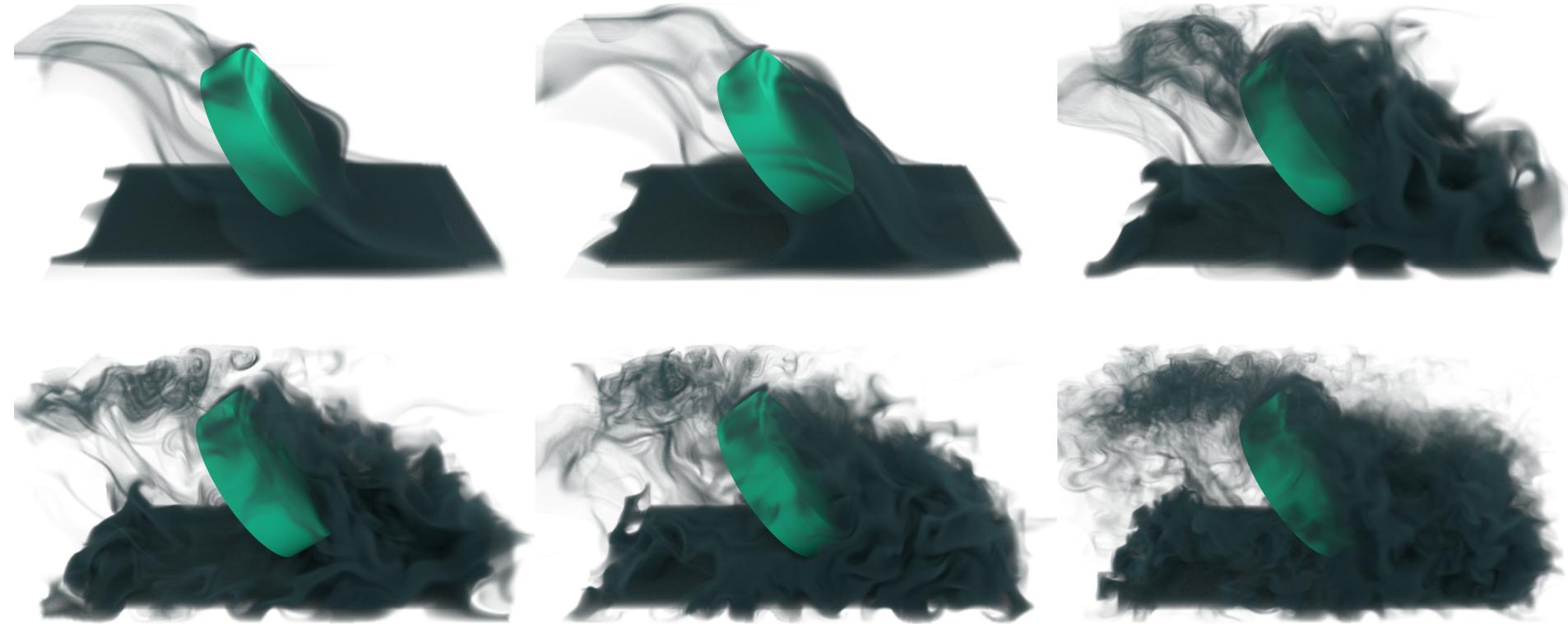


# Scalable Laplacian Eigenfluids



Qiaodong Cui

University of California, Santa Barbara

Pradeep Sen

University of California, Santa Barbara

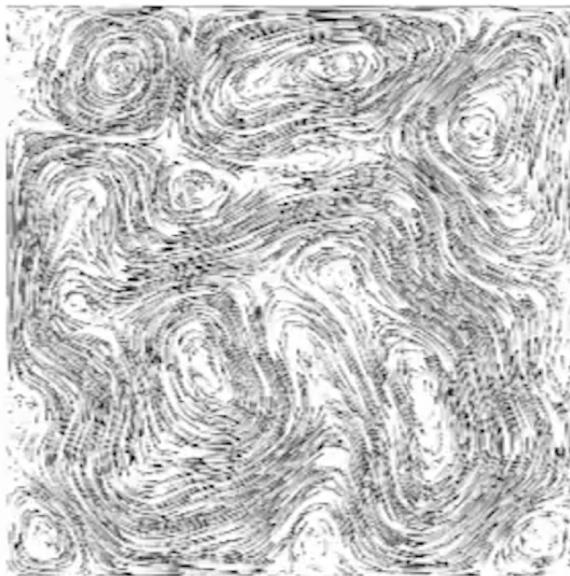
Theodore Kim

Pixar

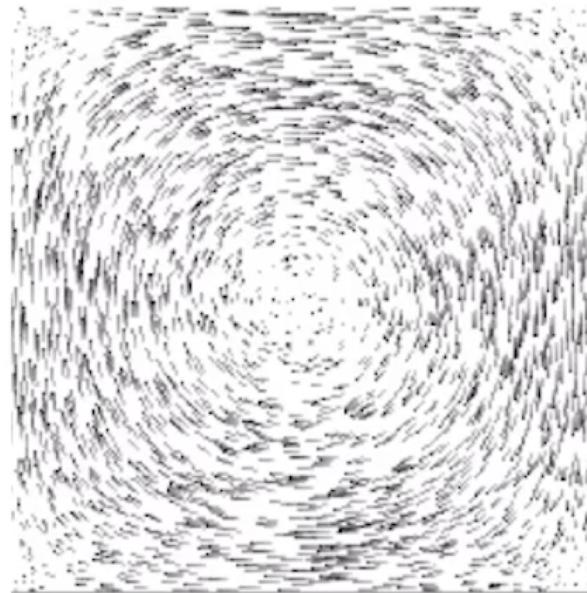
# *Fluid Simulation using Laplacean Eigenfunctions*

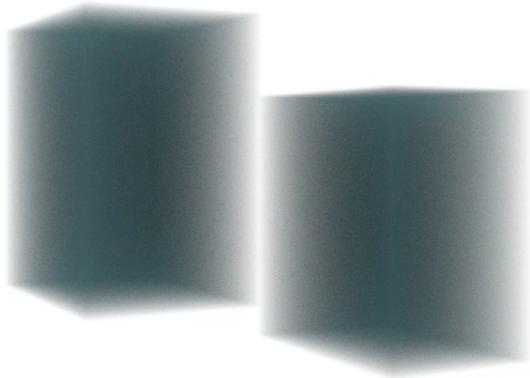
[DeWitt et al. 2012]

Inviscid



Pressure free





200 eigenfunctions [DeWitt et al. 2012]

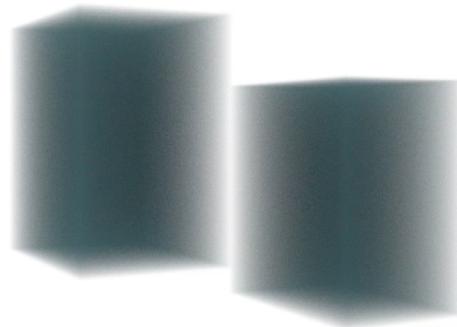
memory: 52 GB

time: 17.2 secs/frame

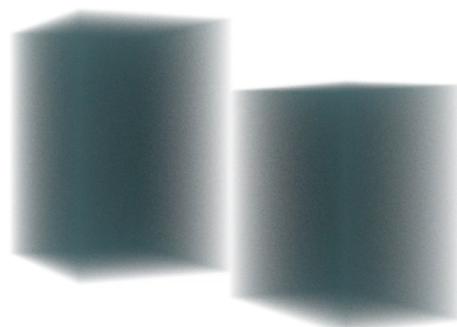
## 24K eigenfunctions:

[DeWitt et al. 2012] memory: 6.1 TB time: 1.84 hrs/frame

Ours memory: **26 GB** time: **13 secs**/frame



200 eigenfunctions



24K eigenfunctions

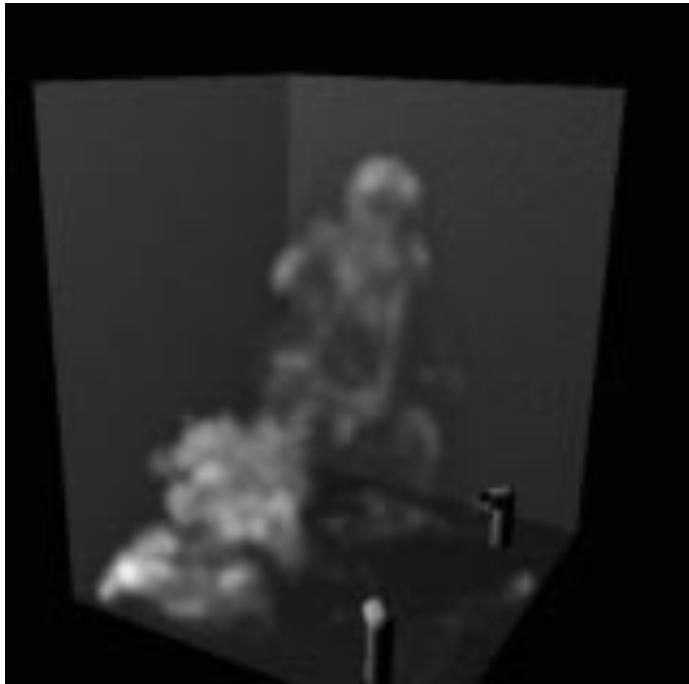
# Outline

- Previous work
- Laplacian Eigenfluids
- Our methods
- Results
- Conclusions and future work

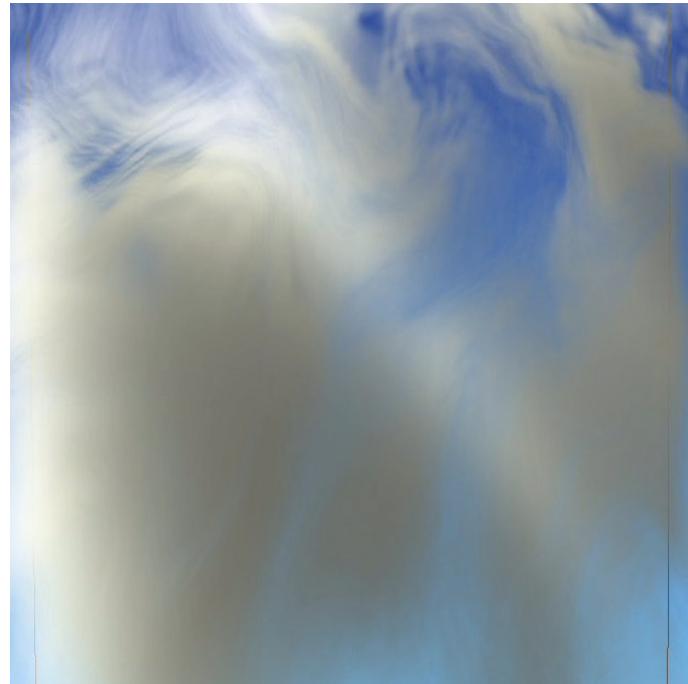
# Outline

- Previous work
  - Laplacian Eigenfluids
  - Our methods
  - Results
  - Conclusions and future work

# Fluid Simulation



[Foster and Metaxas 1997]

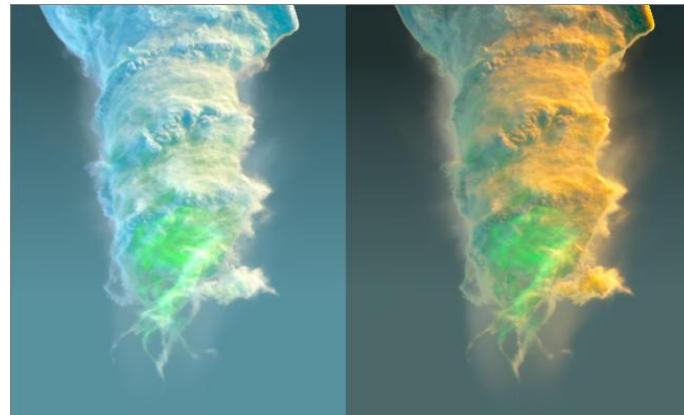


[Stam 1999]

# Spectral Solvers



[Long and Reinhard 2009]



[Henderson 2012]

# Inviscid Methods

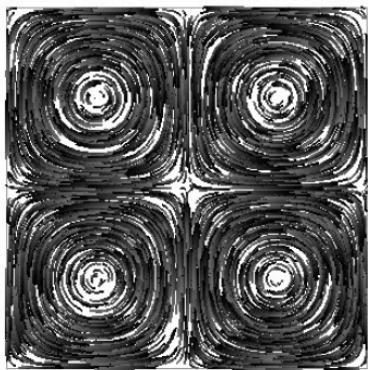


[Mullen et al. 2009]

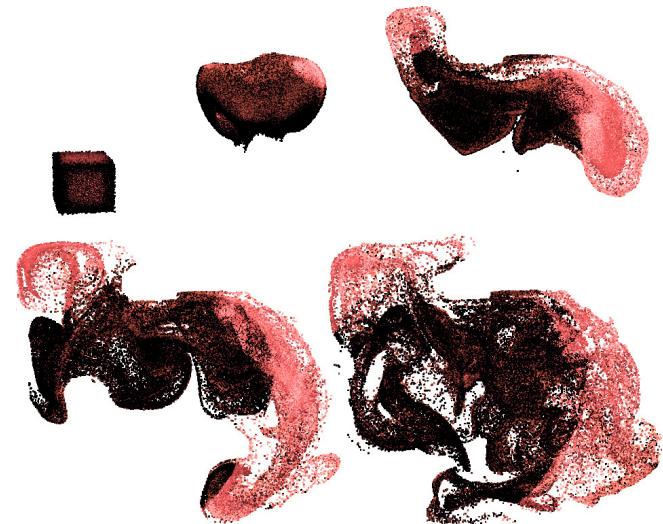
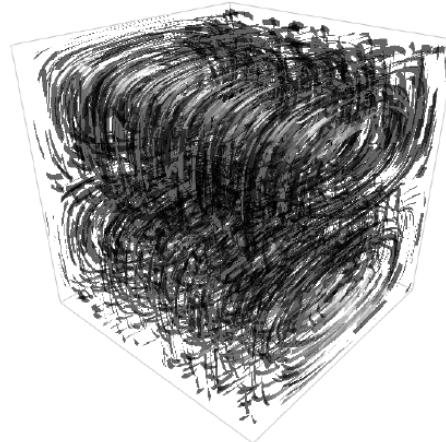


[Chern et al. 2016]

# Laplacian Eigenfluids



[De Witt et al. 2012]



[Liu et al. 2015]

# Outline

- Previous work
- Laplacian Eigenfluids
- Our methods
- Results
- Conclusions and future work

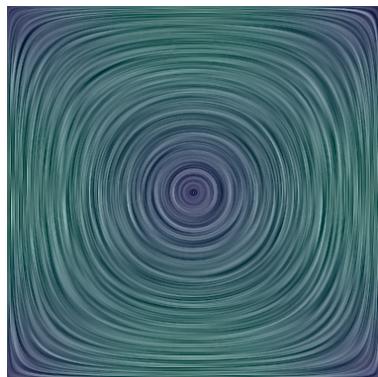
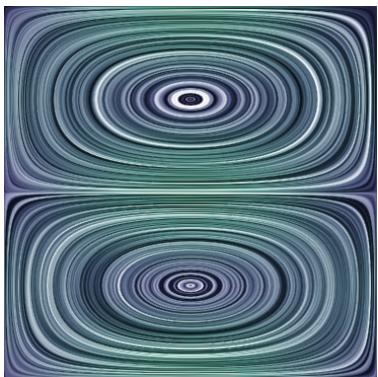
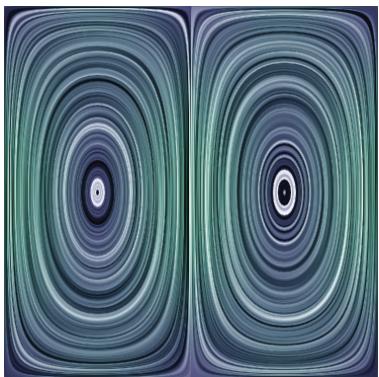
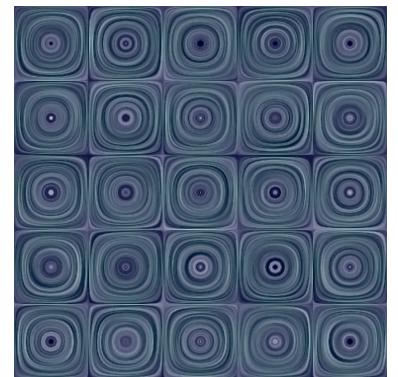
# Laplacian Eigenfluids

$$\dot{\mathbf{u}} = -\mathbf{u} \cdot \nabla \mathbf{u} + \nu \nabla^2 \mathbf{u} - \nabla p + \mathbf{f}$$

$$\nabla \cdot \mathbf{u} = 0$$

$$\mathbf{u} = \sum_{i=1}^r w_i \Psi_i = w_1 \Psi_1 + w_2 \Psi_2 + w_3 \Psi_3 + \dots + w_r \Psi_r$$

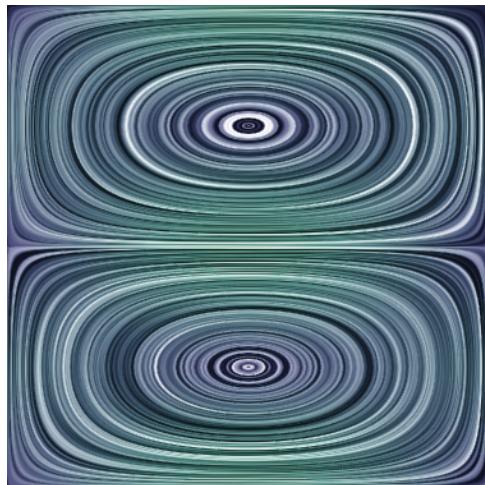
$$\nabla \cdot \Psi_i = 0$$

 $\Psi_1$  $\Psi_2$  $\Psi_3$  $\dots$  $\Psi_r$

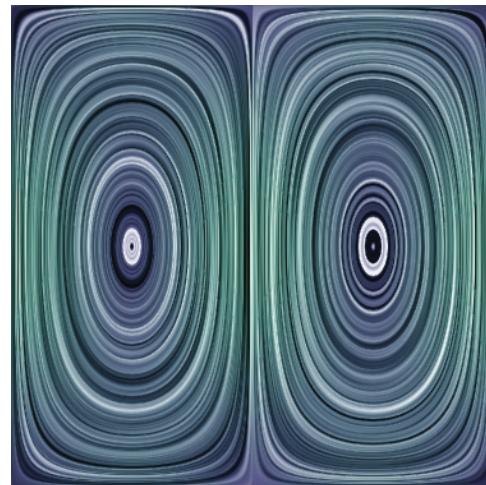
# Laplacian Eigenfunctions

$$\Psi_x(\mathbf{x}, \mathbf{k}) = -\frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \cos(\mathbf{k}_y y)$$

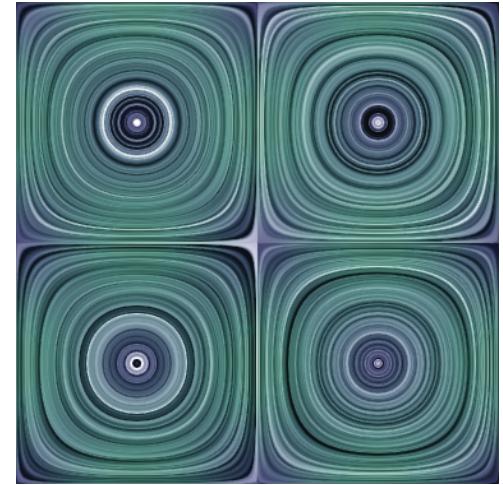
$$\Psi_y(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \cos(\mathbf{k}_x x) \sin(\mathbf{k}_y y)$$



$\mathbf{k}_x = 1, \mathbf{k}_y = 2$

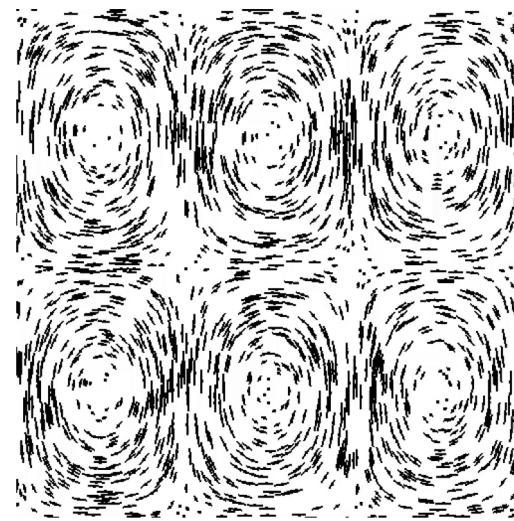
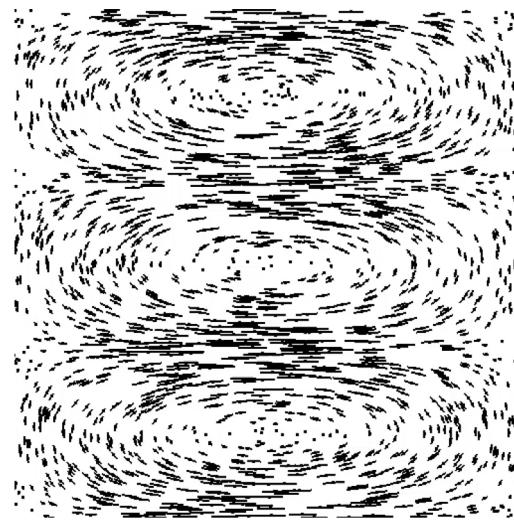
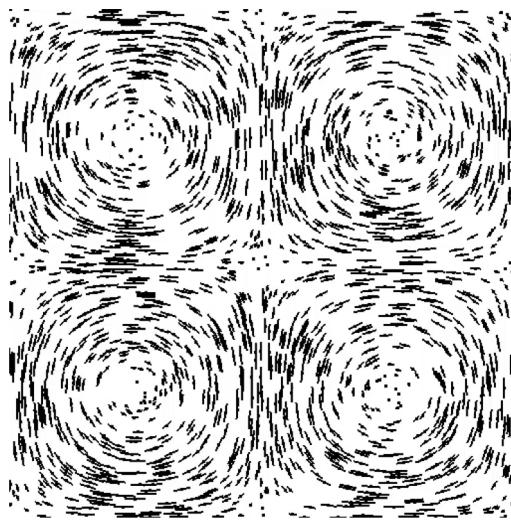
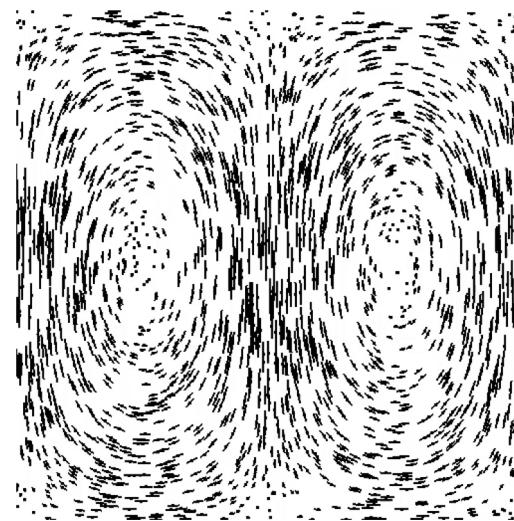
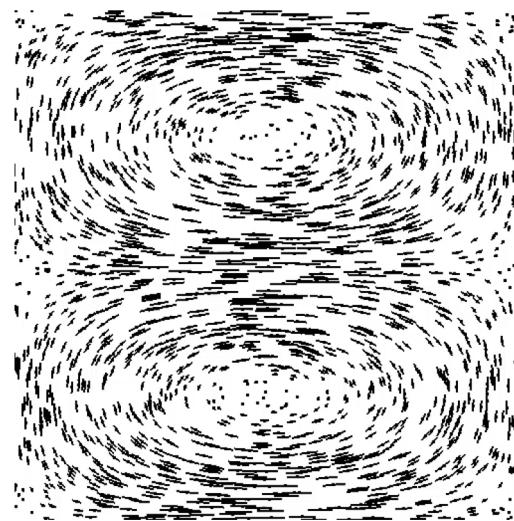
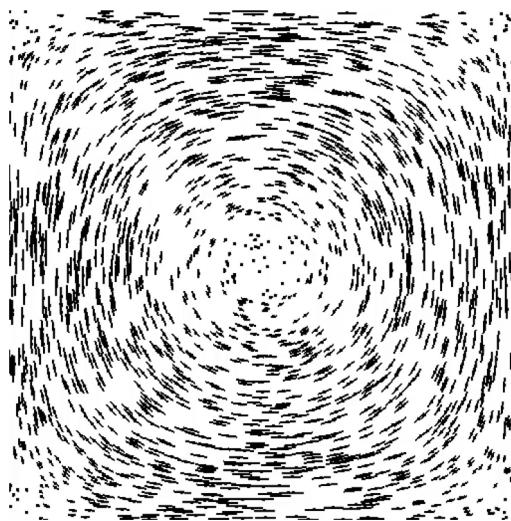


$\mathbf{k}_x = 2, \mathbf{k}_y = 1$

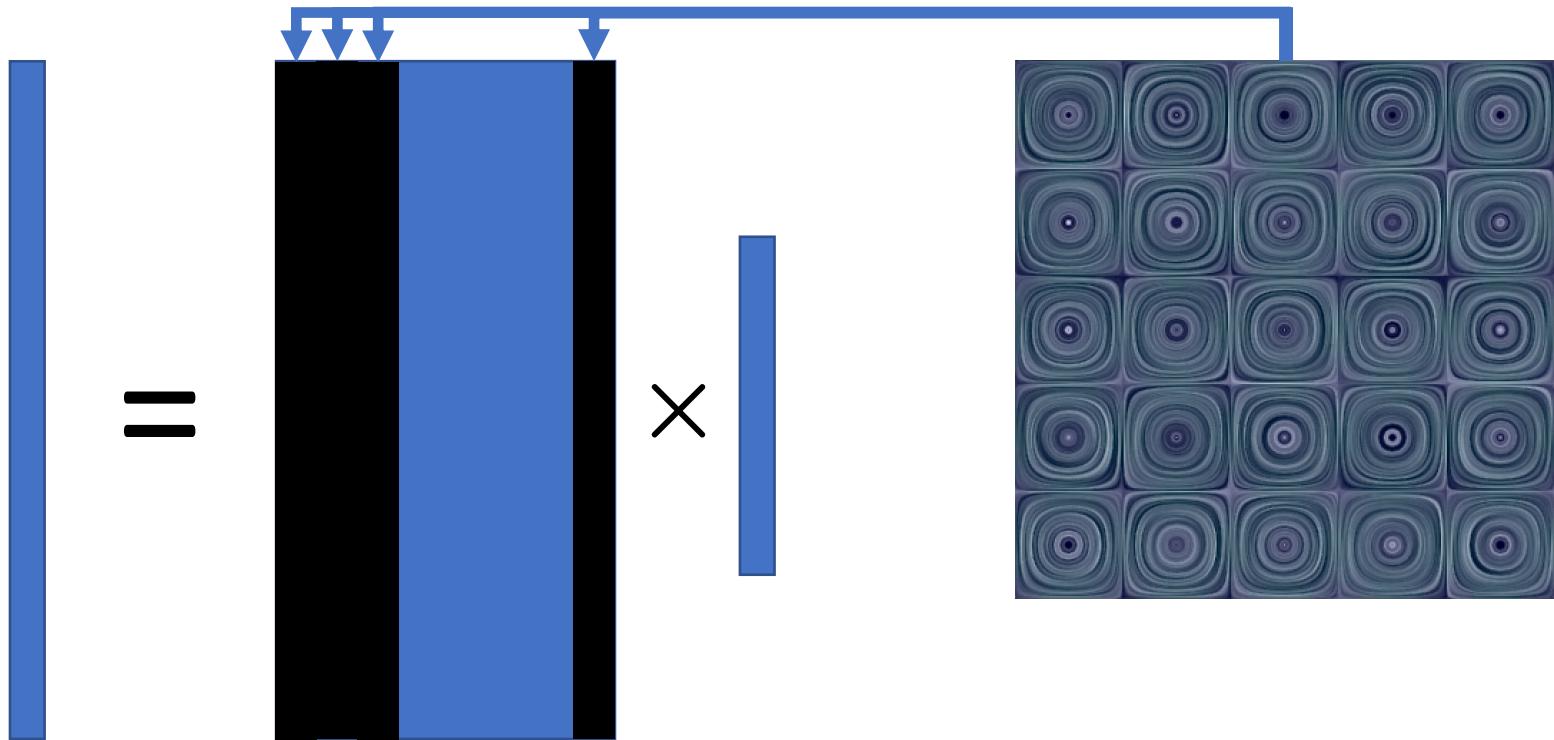


$\mathbf{k}_x = 2, \mathbf{k}_y = 2$

# Laplacian Eigenfunctions



# Reconstruction Bottleneck



$$\mathbf{u} \in \mathbb{R}^{N^3} \quad \mathbf{U} \in \mathbb{R}^{N^3 \times r} \quad \mathbf{W} \in \mathbb{R}^r \quad r \approx 1000 \quad 50.5 \text{ GB}$$
$$\mathbf{U} = \{\Psi_1, \Psi_2, \dots, \Psi_r\} \quad 84\%$$

# Analytical Basis

- Basis take analytical forms under rectangular domain

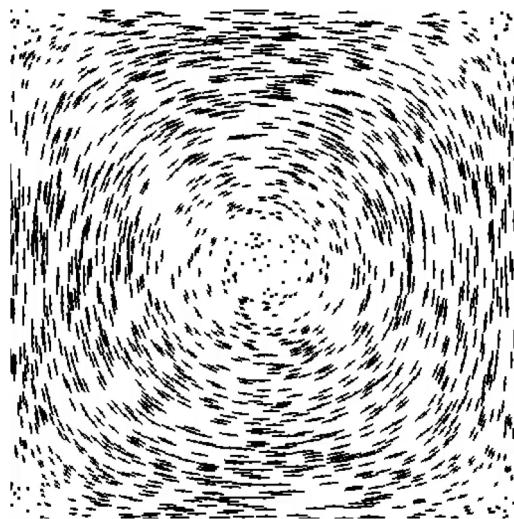
$$\Psi_x(\mathbf{x}, \mathbf{k}) = -\frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \cos(\mathbf{k}_y y)$$

$$\Psi_y(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \cos(\mathbf{k}_x x) \sin(\mathbf{k}_y y)$$

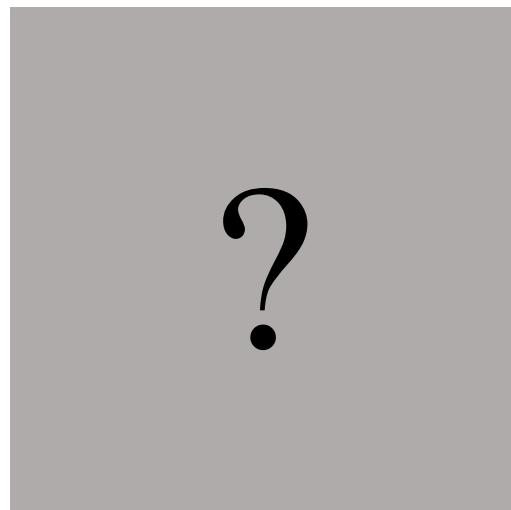
- Removes basis storage
- 5-7 times slower
- $r \approx 1000$    44.1 seconds vs 9.5 seconds

# Boundary Conditions

Dirichlet:



Neumann:



# Outline

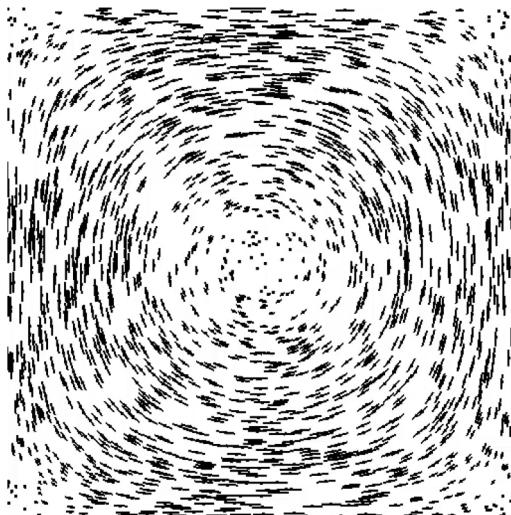
- Previous work
- Laplacian Eigenfluids
- Our methods
  - Analytical basis functions with DCT
  - Dynamics
  - Other features
- Results
- Conclusions and future work

# Analytical Basis with DCT

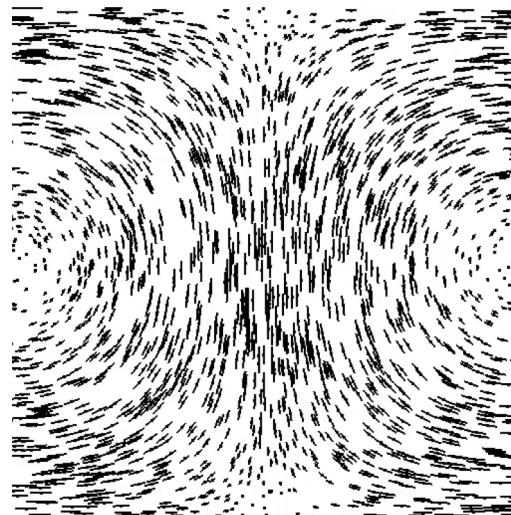
- Memory complexity:  $O(rN^3) \xrightarrow{\hspace{2cm}} O(r)$
- Time complexity:  $O(rN^3) \xrightarrow{\hspace{2cm}} O(N^3 \log(N))$

# Analytical Basis with DCT

Dirichlet:



Neumann:



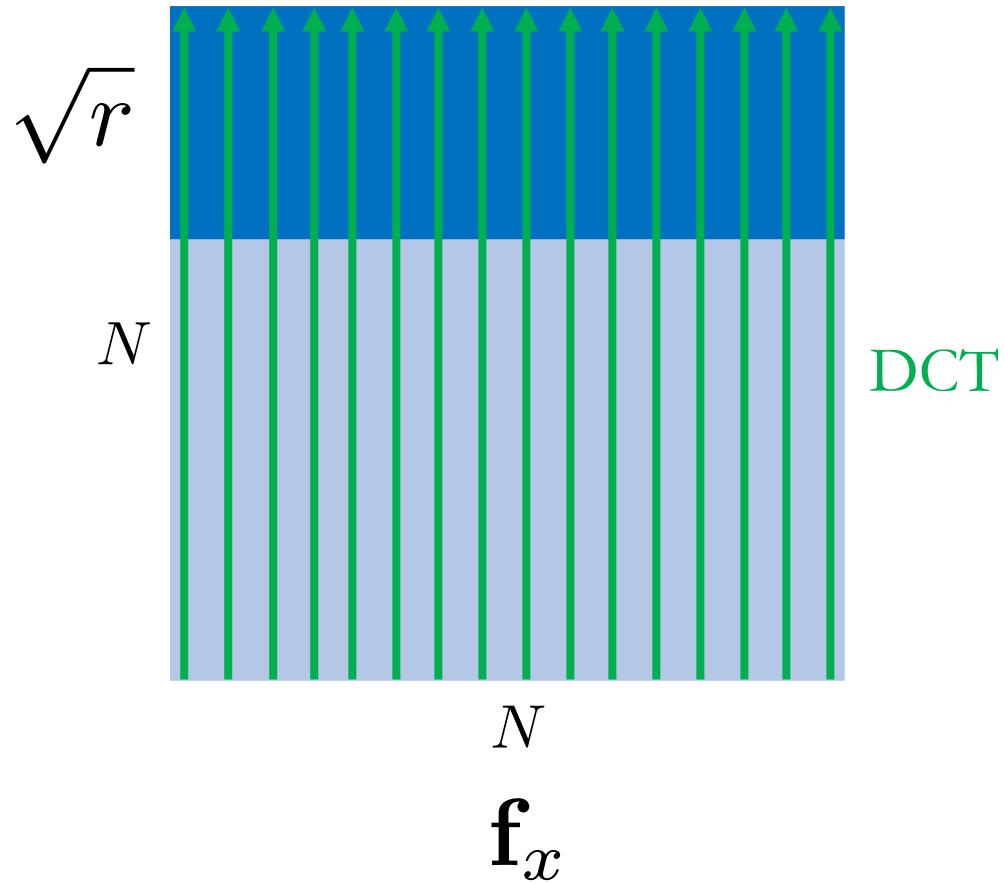
# Basis Transformations with DCT

$$\Psi_x(\mathbf{x}, \mathbf{k}) = -\frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \cos(\mathbf{k}_y y)$$

$$\Psi_y(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \cos(\mathbf{k}_x x) \sin(\mathbf{k}_y y)$$

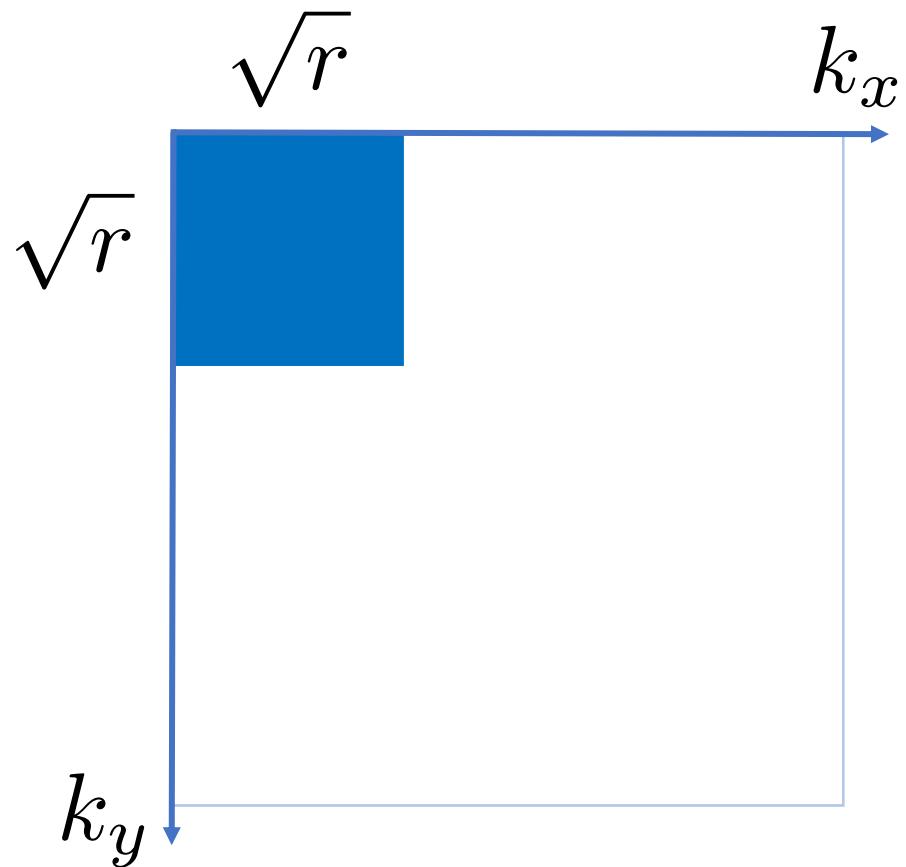
$$\hat{\mathbf{f}}_k = \mathbf{f} \cdot \Psi_k = \langle \mathbf{f}_x, \Psi_x(\mathbf{k}) \rangle + \langle \mathbf{f}_y, \Psi_y(\mathbf{k}) \rangle$$

$$\langle \mathbf{f}_x, \Psi_x(\mathbf{k}) \rangle = -\frac{1}{|\mathbf{k}|} k_y \int_{\Omega} \mathbf{f}_x \sin(k_x x) \cos(k_y y) dx dy$$



The diagram illustrates a DST (Dense SuperTree) structure. It features a central vertical column of blue rectangles representing nodes, with horizontal arrows pointing towards it from both sides. The top node is labeled  $\sqrt{r}$ . To the left of the central column, there is another label  $\sqrt{r}$ . The background is divided into three horizontal regions: a top region in light blue, a middle region in medium blue, and a bottom region in light blue.

$f_x$

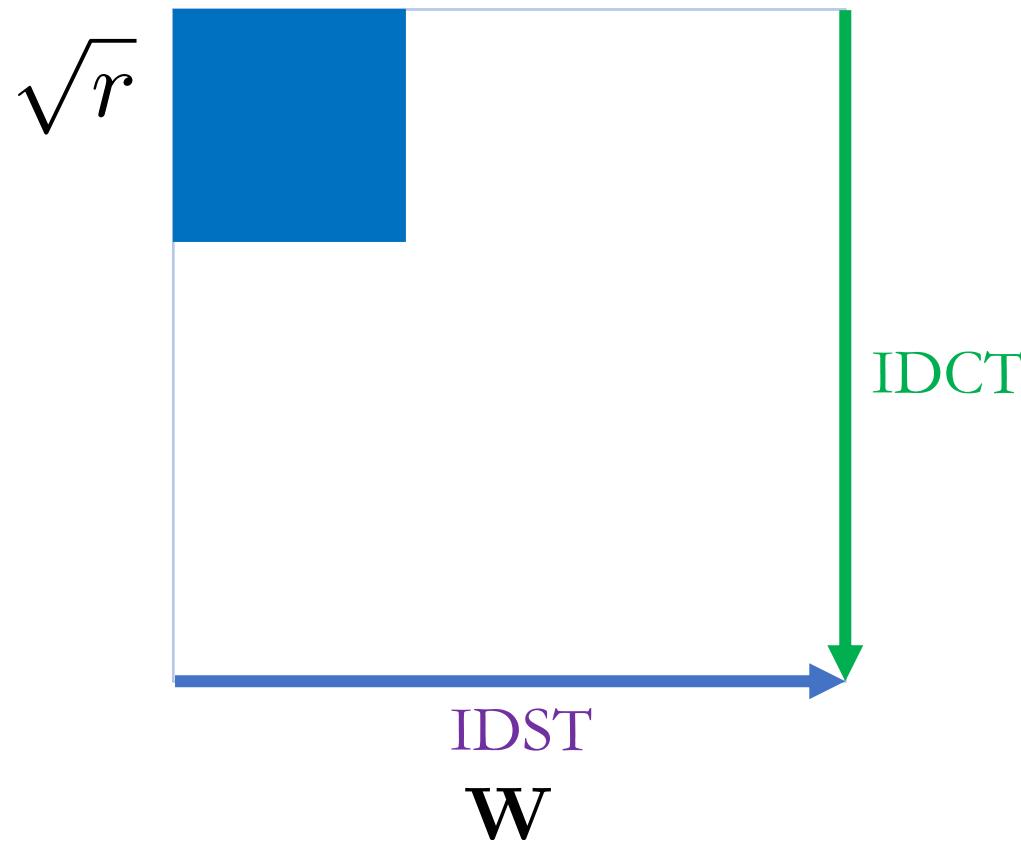


$$\langle \mathbf{f}_x, \boldsymbol{\Psi}_x(\mathbf{k}) \rangle$$

# Velocity reconstruction

$$\Psi_x(\mathbf{x}, \mathbf{k}) = -\frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \cos(\mathbf{k}_y y)$$

$$\sqrt{r}$$



# Analytical Basis with DCT

- Memory complexity:  $O(rN^3) \rightarrow O(r)$  226×
- Time complexity:  $O(rN^3) \rightarrow O(N^3 \log(N))$  95×

$128^3$  Grid, 1000 basis functions

# Dirichlet

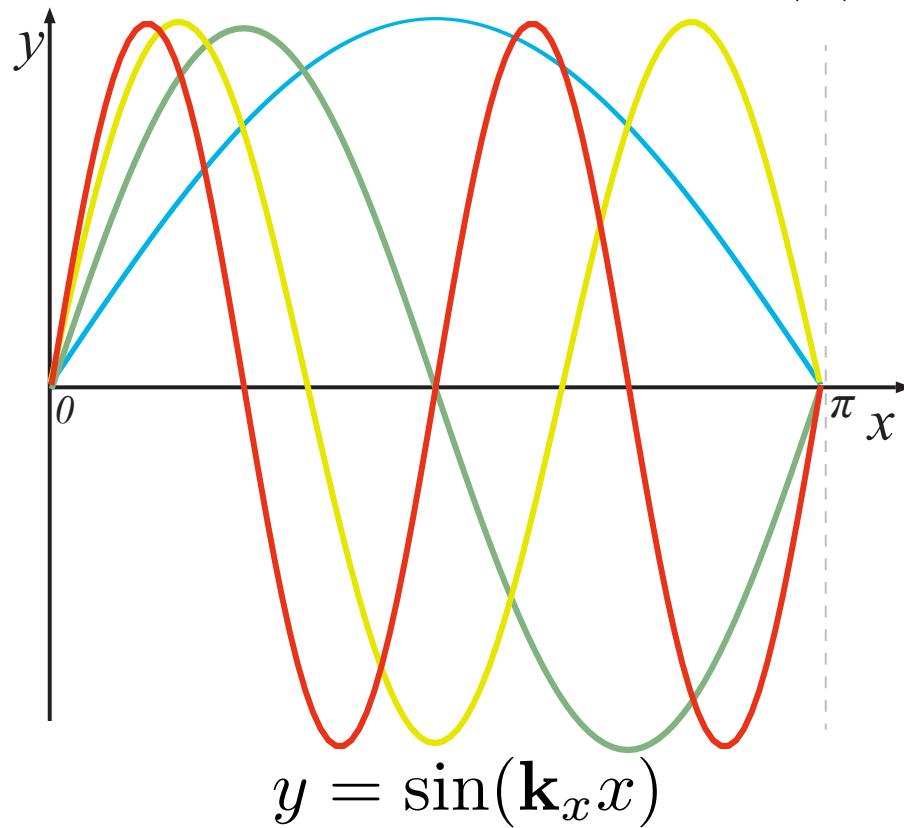
[De Witt et al. 2012]

$$\Psi_x|_{x=0,\pi} = 0$$

$$\Psi_x(\mathbf{x}, \mathbf{k}) = -\frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \cos(\mathbf{k}_y y)$$

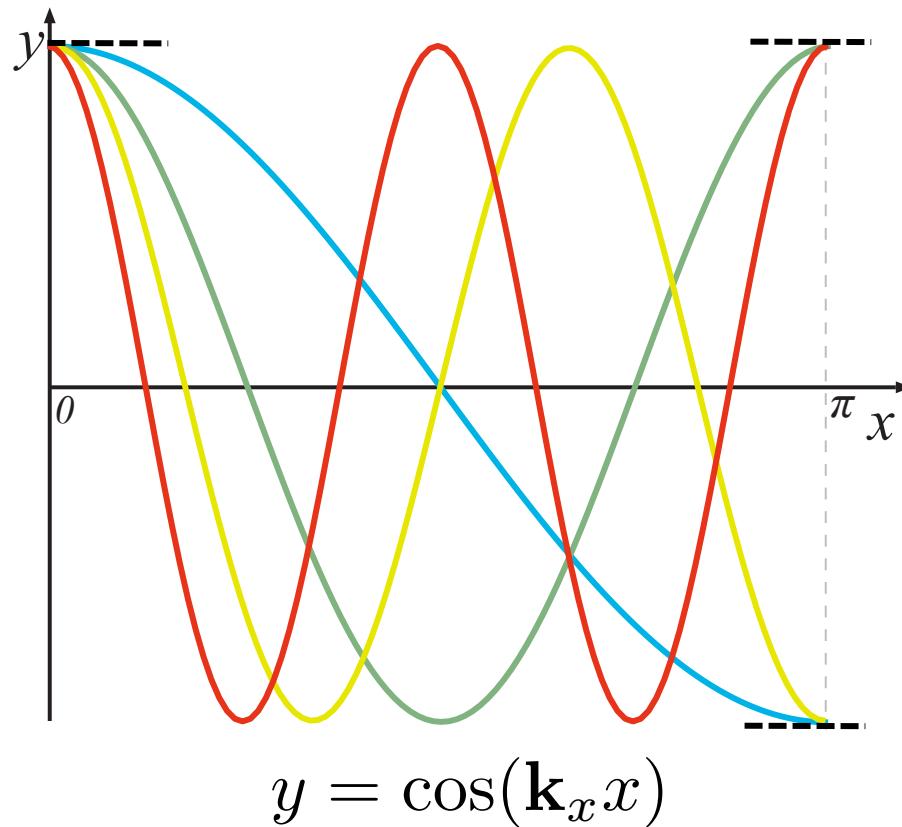
$$\Psi_y|_{y=0,\pi} = 0$$

$$\Psi_y(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \cos(\mathbf{k}_x x) \sin(\mathbf{k}_y y)$$



# Extension to Neumann

$$\frac{\partial \Psi_x}{\partial x} |_{x=0,\pi} = 0$$



# Extension to Neumann

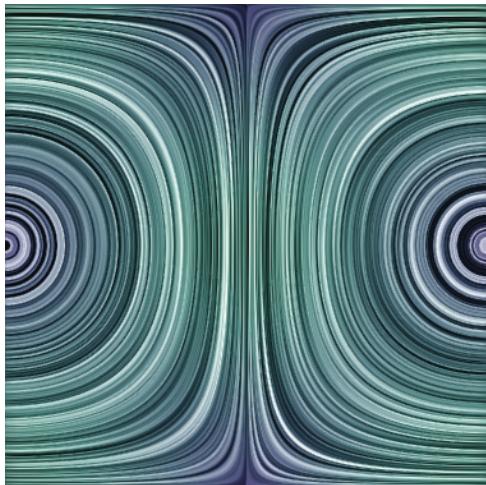
$$\frac{\partial \Psi_x}{\partial x}|_{x=0,\pi} = 0 \quad \Psi_y|_{y=0,\pi} = 0$$

$$\begin{cases} \nabla^2 \Psi(\mathbf{x}, \mathbf{k}) = -|\mathbf{k}|^2 \Psi(\mathbf{x}, \mathbf{k}) \\ \nabla \cdot \Psi(\mathbf{x}, \mathbf{k}) = 0 \end{cases}$$

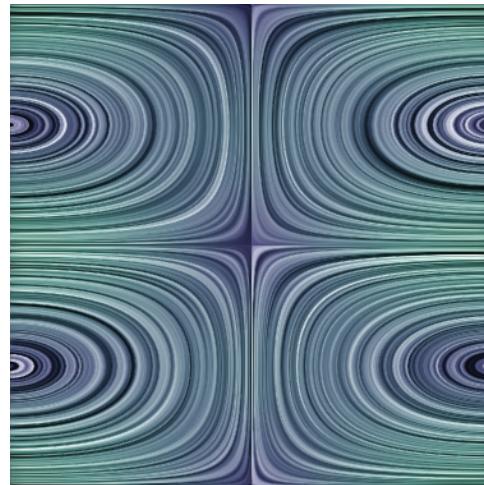
$$\begin{aligned} \Psi_x(\mathbf{x}, \mathbf{k}) &= \frac{1}{|\mathbf{k}|} \cos(\mathbf{k}_x x) \cos(\mathbf{k}_y y) \\ \Psi_y(\mathbf{x}, \mathbf{k}) &= \frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \sin(\mathbf{k}_y y) \end{aligned}$$

# Extension to Neumann

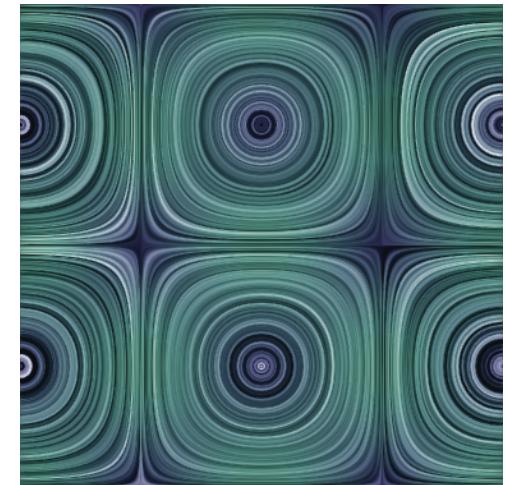
$$\Psi_x(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \cos(\mathbf{k}_x x) \cos(\mathbf{k}_y y)$$
$$\Psi_y(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \sin(\mathbf{k}_y y)$$



$$\mathbf{k}_x = 1, \quad \mathbf{k}_y = 1$$

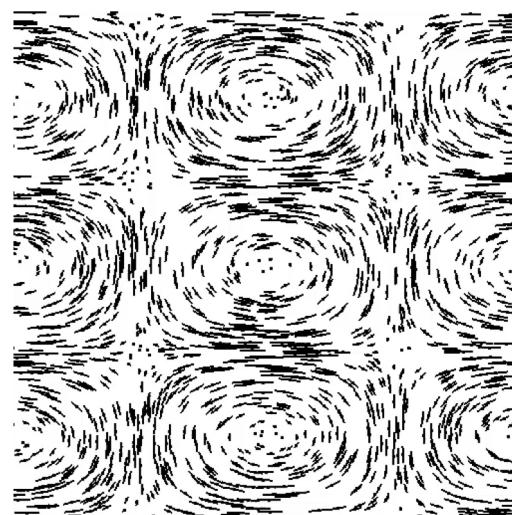
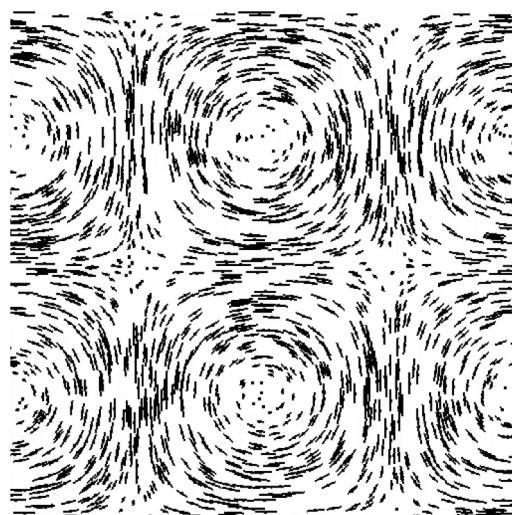
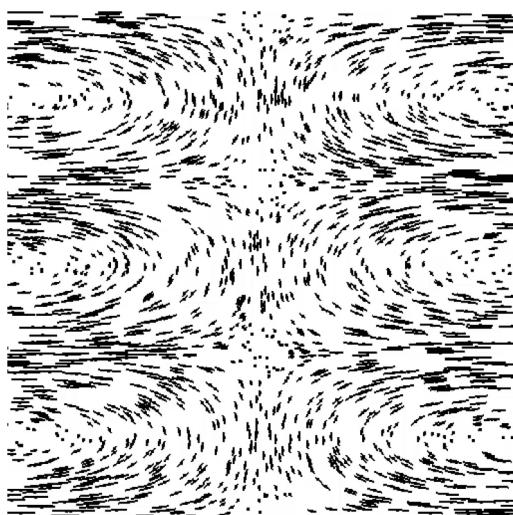
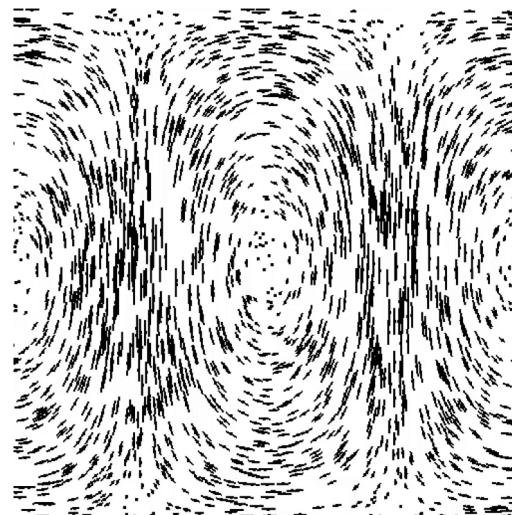
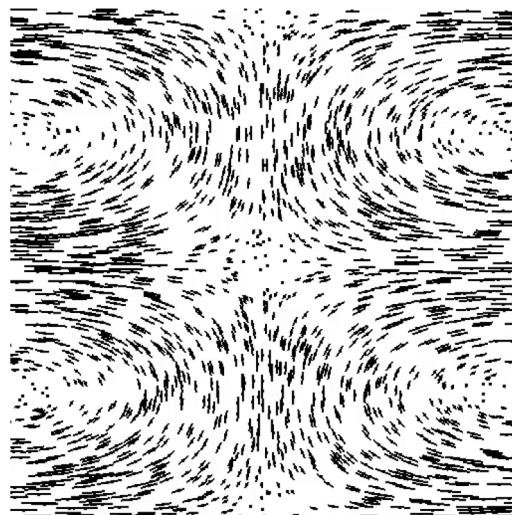
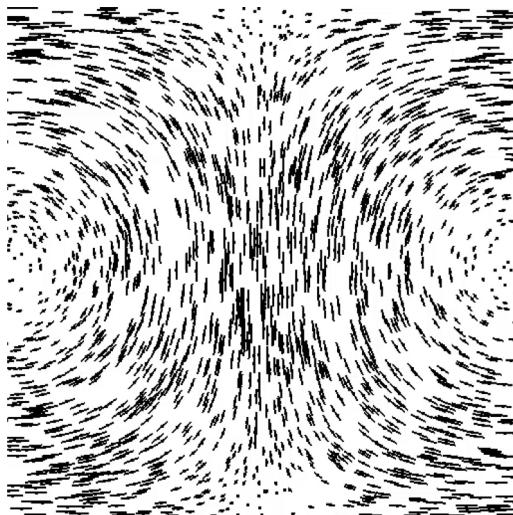


$$\mathbf{k}_x = 1, \quad \mathbf{k}_y = 2$$



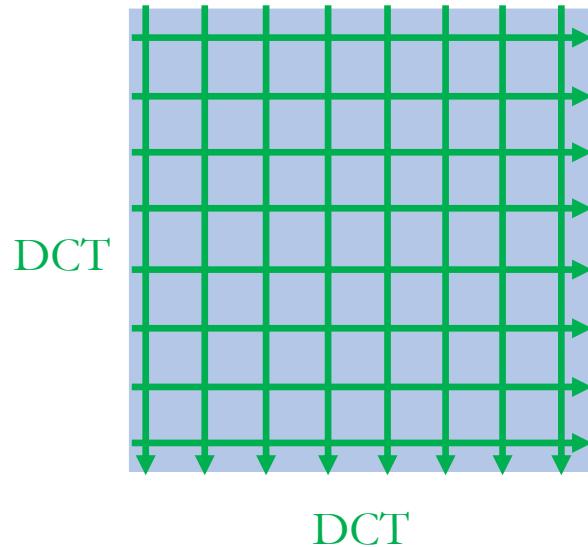
$$\mathbf{k}_x = 2, \quad \mathbf{k}_y = 2$$

# Extension to Neumann

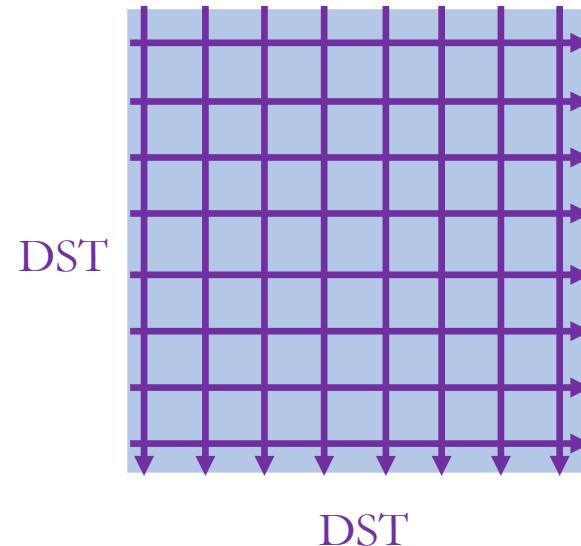


# Extension to Neumann

$$\Psi_x(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \cos(\mathbf{k}_x x) \cos(\mathbf{k}_y y)$$
$$\Psi_y(\mathbf{x}, \mathbf{k}) = \frac{1}{|\mathbf{k}|} \sin(\mathbf{k}_x x) \sin(\mathbf{k}_y y)$$



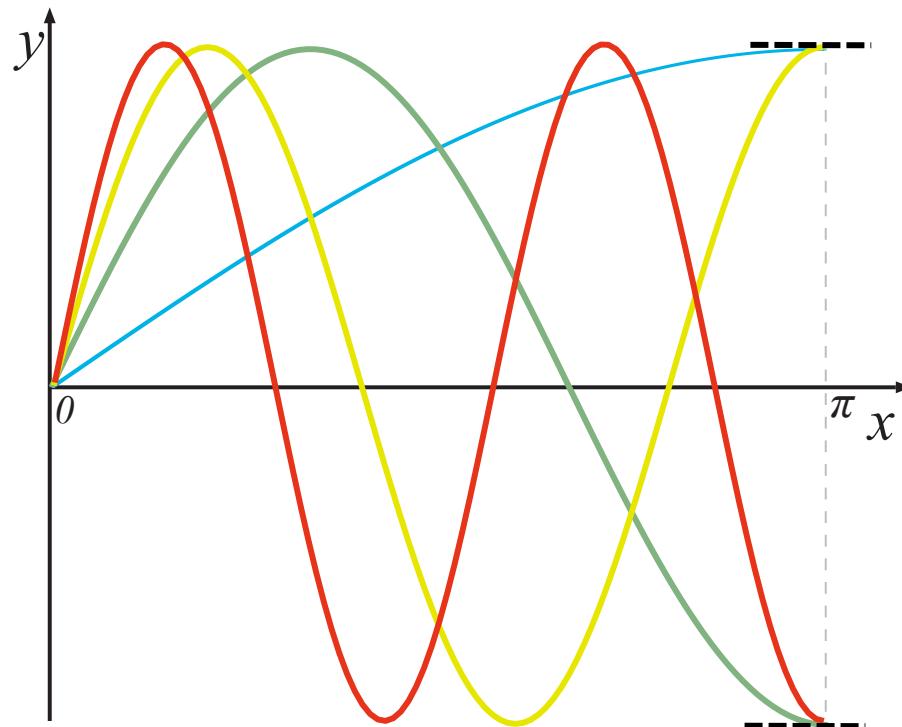
$\Psi_x$



$\Psi_y$

# Mixed Boundary Conditions

$$\Psi_x|_{x=0} = 0 \quad \frac{\partial \Psi_x}{\partial x}|_{x=\pi} = 0$$



$$y = \sin((k_x - 0.5)x), \quad k_x \in \mathbb{Z}^+$$

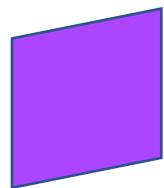
# Mixed Boundary Conditions

$$\Psi_x|_{x=0} = 0 \quad \frac{\partial \Psi_x}{\partial x}|_{x=\pi} = 0$$

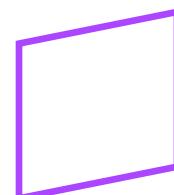
$$\Psi_x(\mathbf{x}, \mathbf{k}) = -\frac{1}{\mathbf{k}} \sin((\mathbf{k}_x - 0.5)x) \cos(\mathbf{k}_y y)$$

$$\Psi_y(\mathbf{x}, \mathbf{k}) = \frac{1}{\mathbf{k}} \cos((\mathbf{k}_x - 0.5)x) \sin(\mathbf{k}_y y)$$

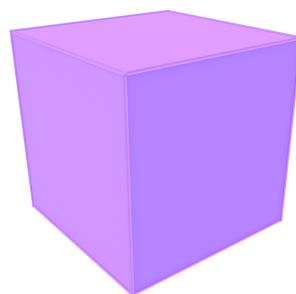
# 3D Basis Functions



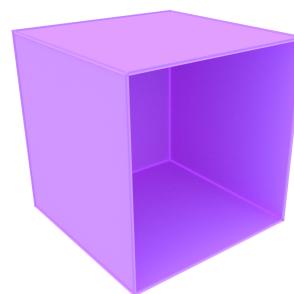
Dirichlet



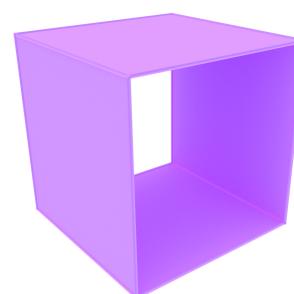
Neumann



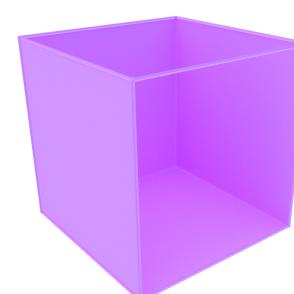
1



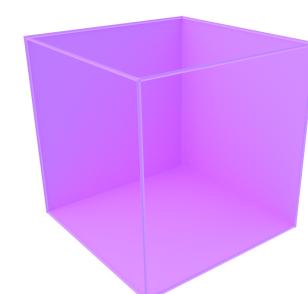
2



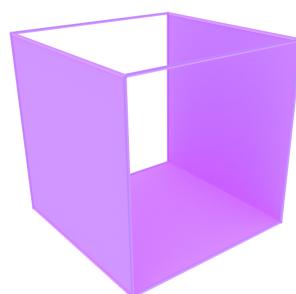
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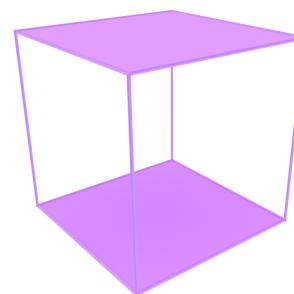
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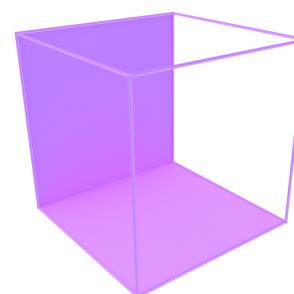
5



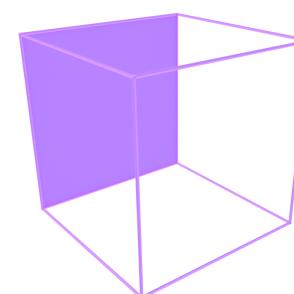
6



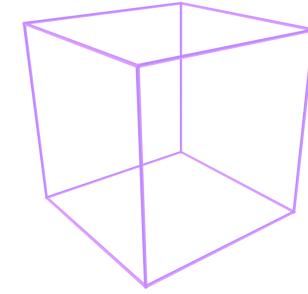
7



8



9



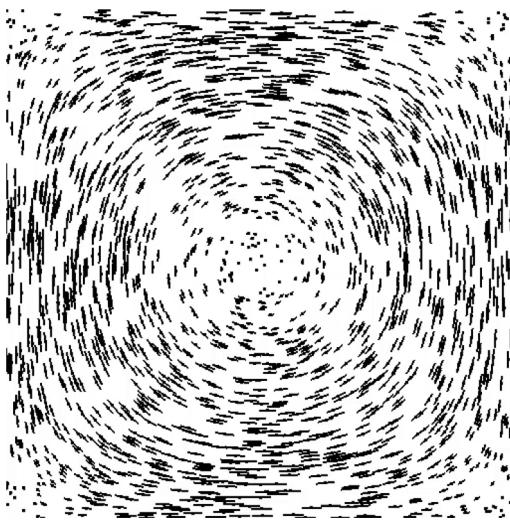
10

# Outline

- Previous work
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  - Analytical basis functions with DCT
  - Dynamics
  - Other features
- Results
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# Dynamics

Dirichlet:



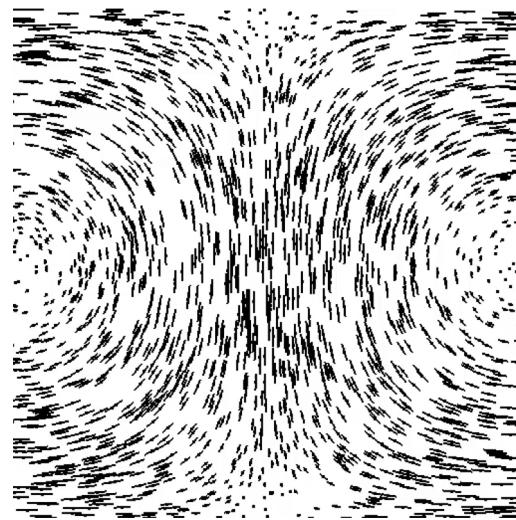
Original  
Eigenfluids



Variational



Neumann:



Blows up



# Dynamics

$$\dot{\mathbf{u}} = -\mathbf{u} \cdot \nabla \mathbf{u} + \nu \nabla^2 \mathbf{u} - \cancel{\nabla p} + \mathbf{f}$$
$$\cancel{\nabla \cdot \mathbf{u} = 0}$$

# Dynamics

$$\mathbf{u} = \sum_{i=1}^r w_i \Psi_i \quad \nabla^2 \Psi_i = \lambda_i \Psi_i$$

Diffusion:  $\dot{\mathbf{u}} = \nu \nabla^2 \mathbf{u}$

$$w_k^{t+1} = w_k^t e^{\nu \lambda_i \Delta t}$$

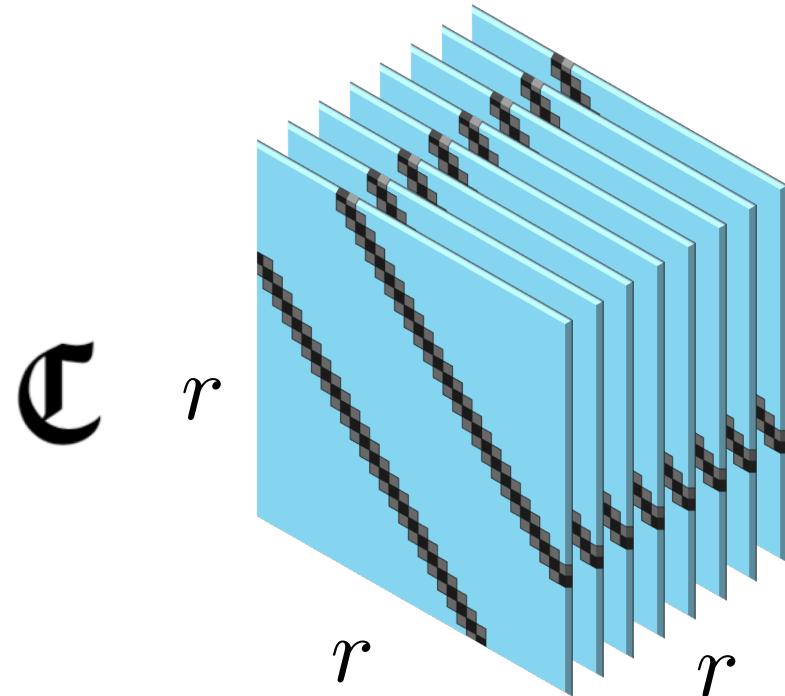
External force:  $\dot{\mathbf{u}} = \mathbf{f}$

$$\hat{\mathbf{f}}_k = \mathbf{f} \cdot \Psi_k \quad w_k^{t+1} = \Delta t \hat{\mathbf{f}}_k$$

# Advection

$$\dot{\mathbf{u}} = -\mathbf{u} \cdot \nabla \mathbf{u} \longrightarrow \dot{\boldsymbol{\omega}} = \nabla \times (\boldsymbol{\omega} \times \mathbf{u})$$

$$\dot{w}_g = \sum_{h=1}^r \sum_{i=1}^r w_h w_i \mathfrak{C}(g, h, i)$$



# Dynamics

$$\dot{w}_g = \sum_{h=1}^r \sum_{i=1}^r w_h w_i \mathbf{C}(g, h, i)$$

$$\phi_h = \nabla \times \Psi_h$$

$$\mathbf{C}(g, h, i) = [\nabla \times (\phi_h \times \Psi_i)] \cdot \phi_g$$

[De Witt et al. 2012]

Dirichlet:      OK

Neumann:      Blows up

# Dynamics

$$\mathbf{C}(g, h, i) = -\mathbf{C}(h, g, i)$$

$$\mathbf{C}(g, h, i) = \int_{\Omega} (\nabla \times \Psi_i) \cdot (\Psi_g \times \Psi_h) d\Omega$$

[Liu et al. 2015]

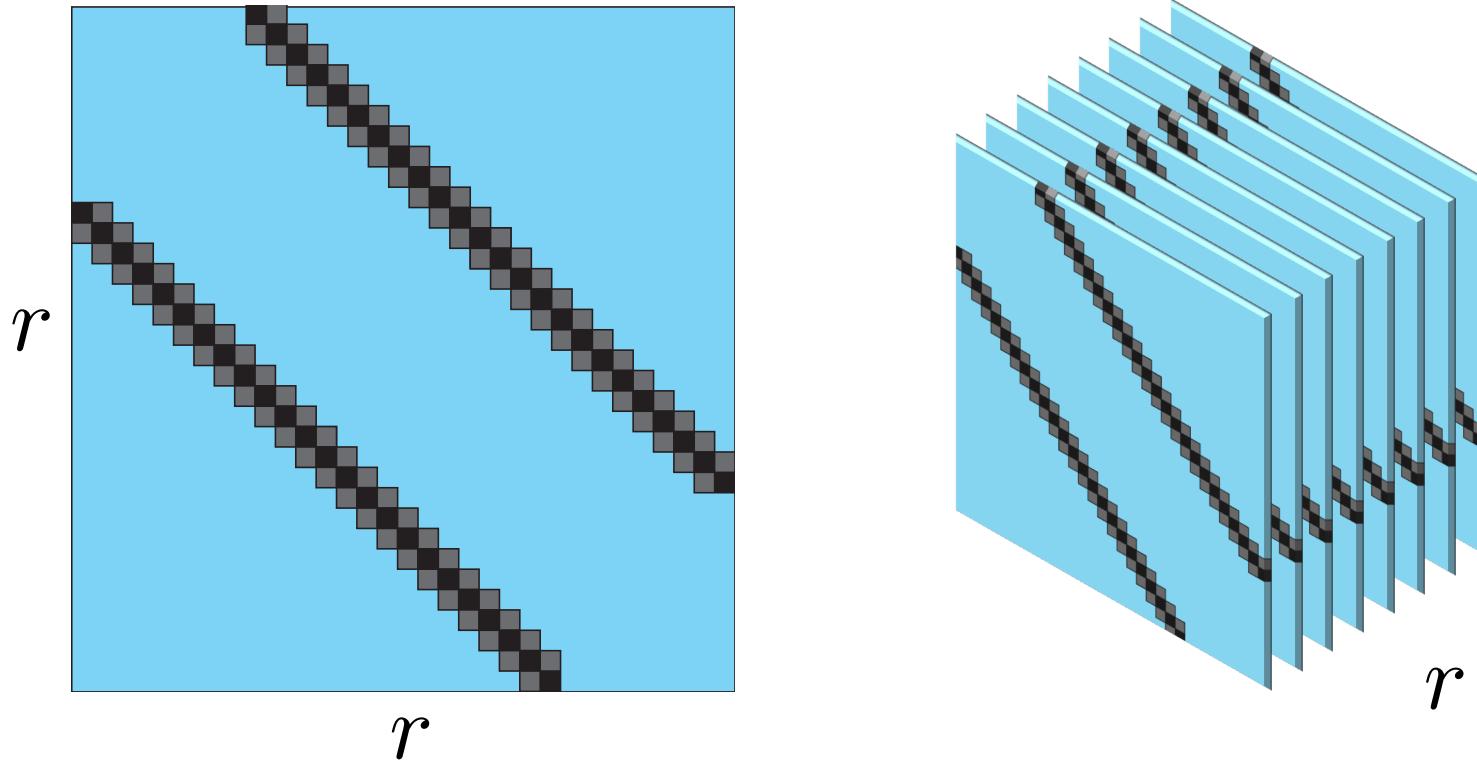
Dirichlet: OK

Neumann: OK

# Outline

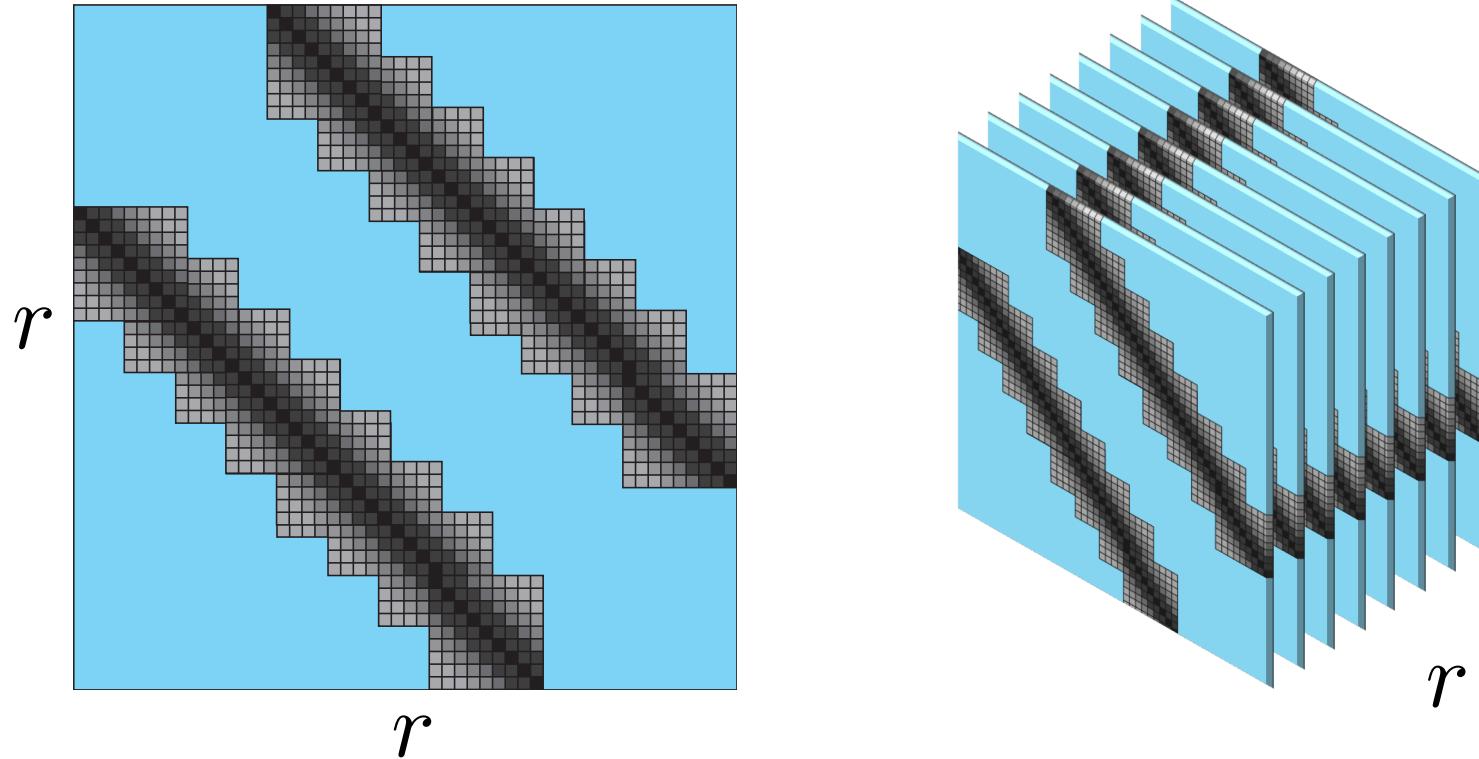
- Previous work
- Laplacian Eigenfluids
- Our methods
  - Analytical basis functions with DCT
  - Dynamics
  - Other features
- Results
- Conclusions and future work

# Tensor Compression



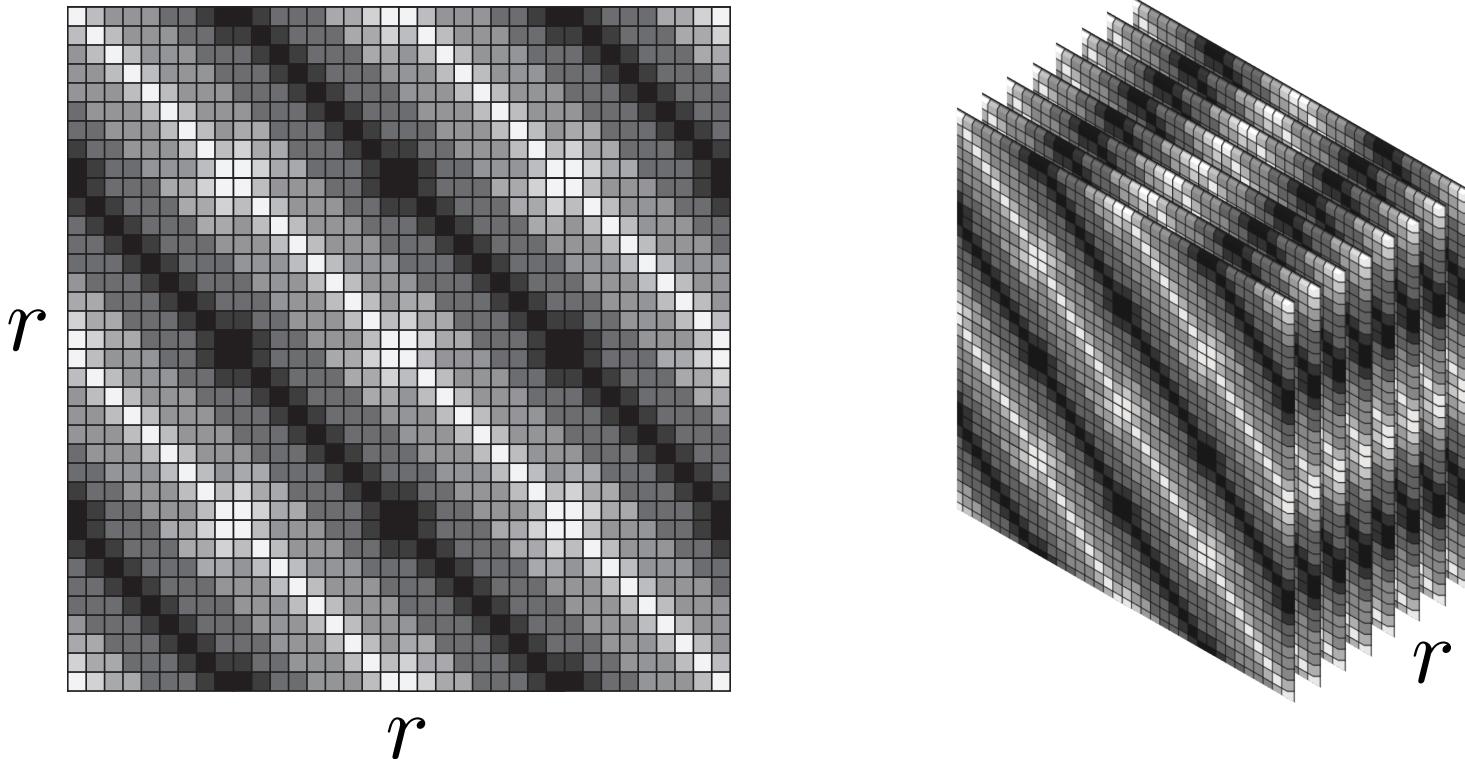
Dirichlet:  $O(r^2)$

# Tensor Compression



One Neumann direction:  $O(r^{2.5})$

# Tensor Compression



Two Neumann directions:  $O(r^3)$

# Tensor Sparsity

Dirichlet

$$O(r^2)$$

One Neumann direction

$$O(r^{2+\frac{1}{3}})$$

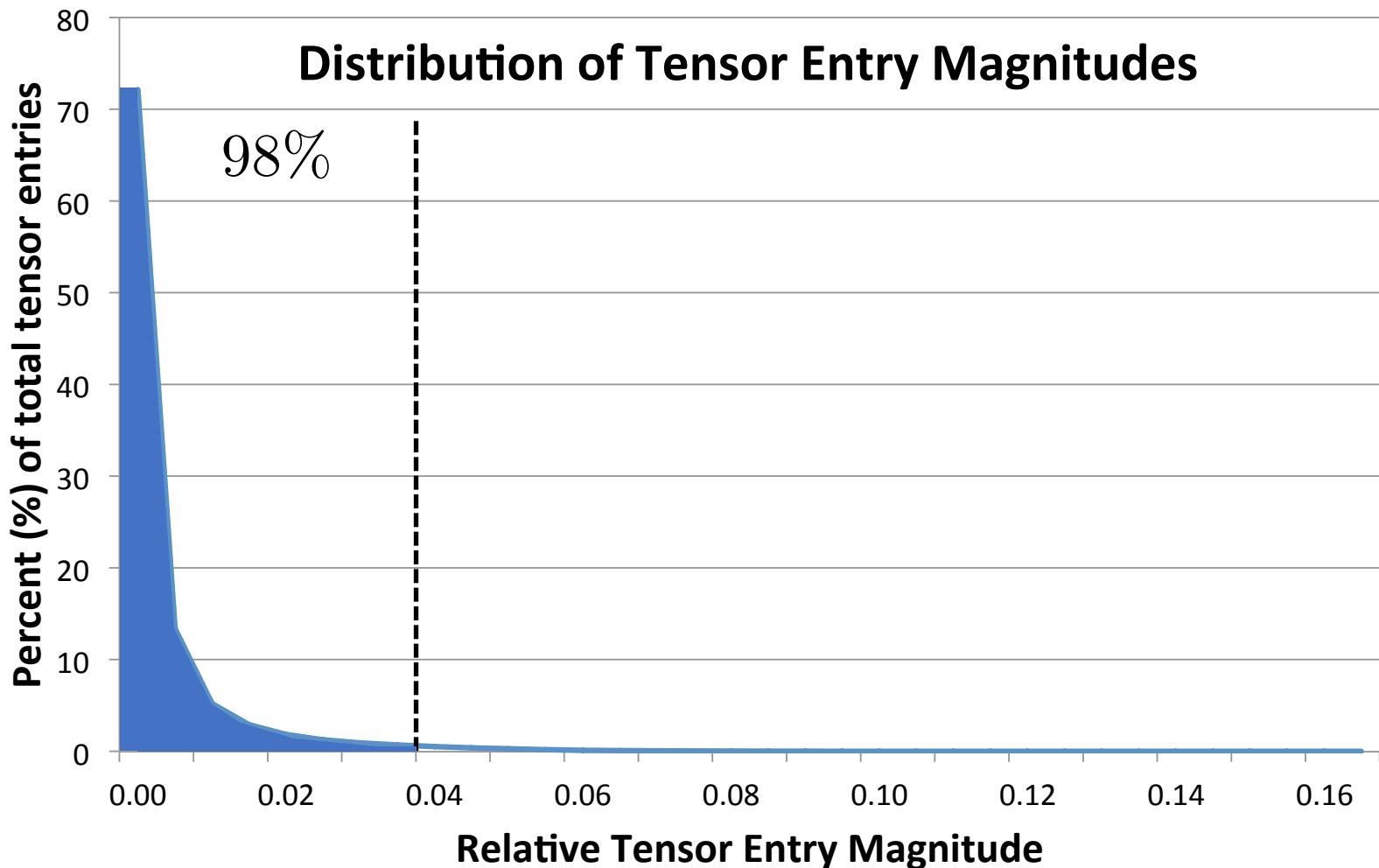
Two Neumann directions

$$O(r^{2+\frac{2}{3}})$$

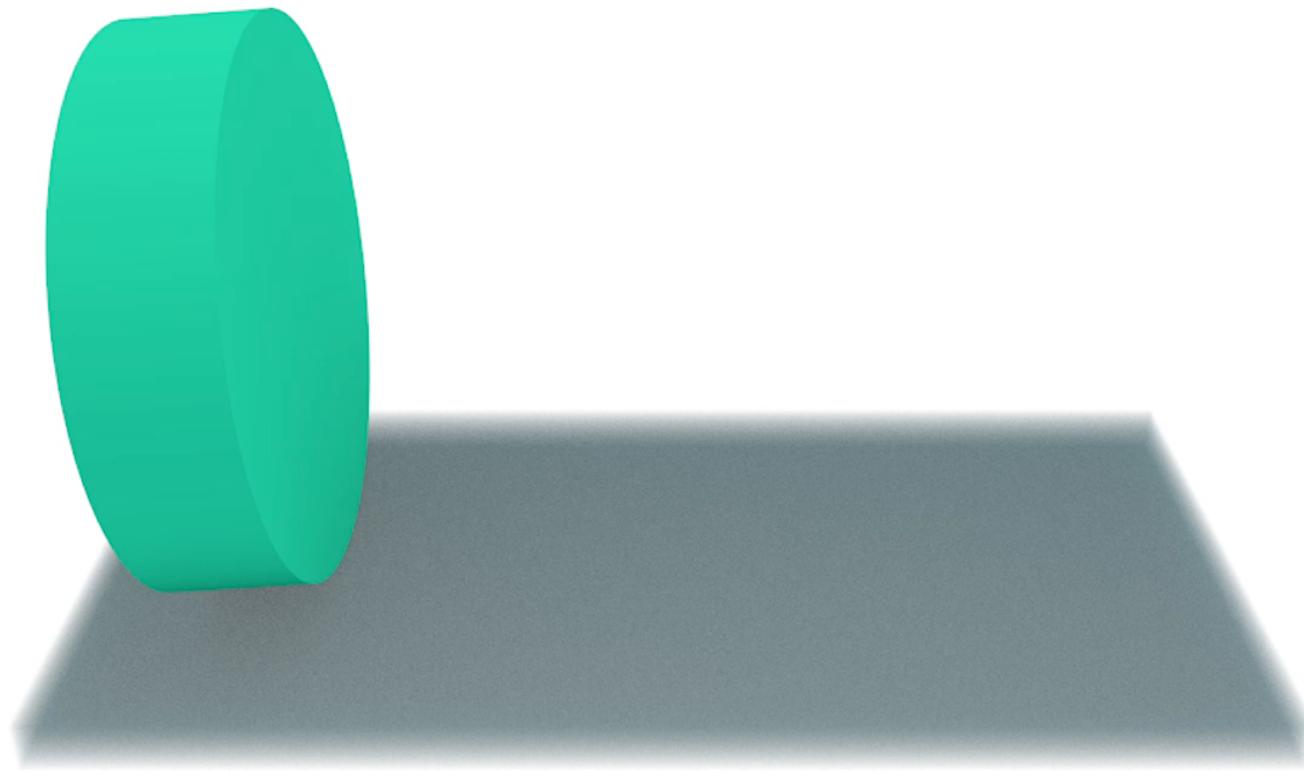
Three Neumann directions

$$O(r^3)$$

# Tensor Compression

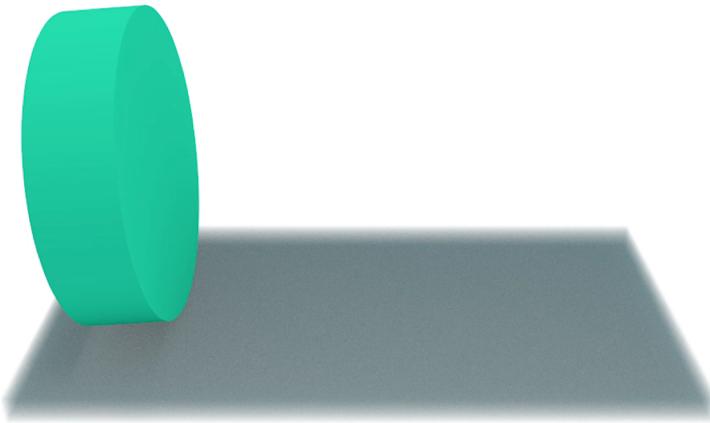


# Tensor Compression

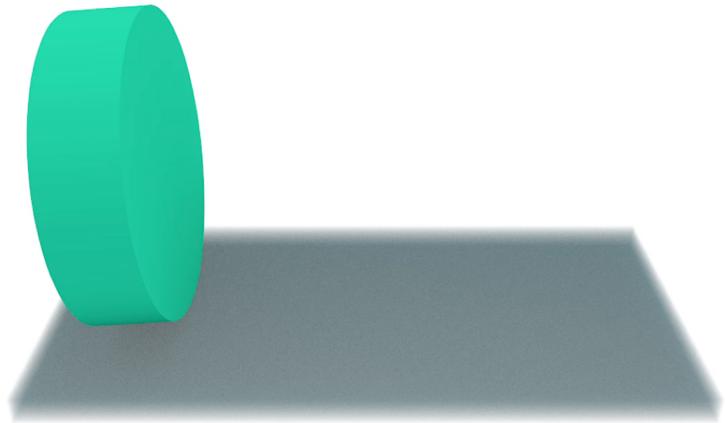


0% discarded, Tensor memory: 26 GB

# Tensor Compression

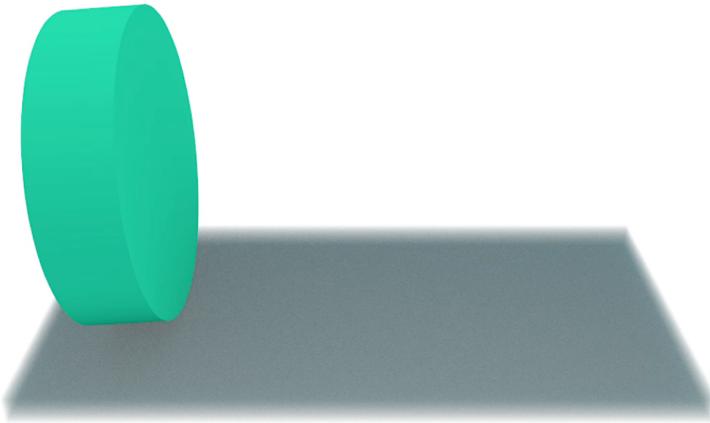


100% discarded, Tensor memory: 0 MB

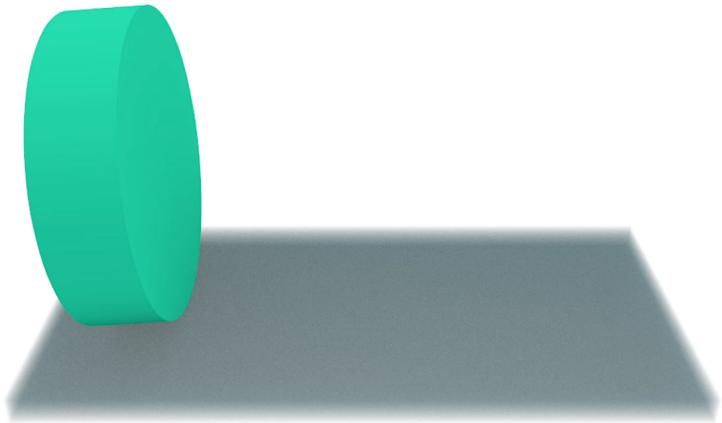


Reference, Tensor memory: 26 GB

# Tensor Compression

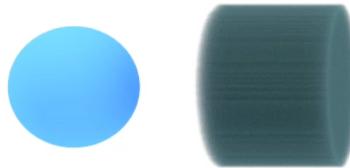


99.9% discarded, Tensor memory: 26 MB

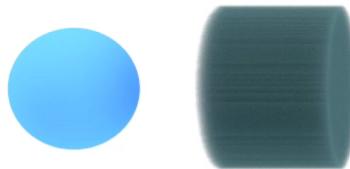


Reference, Tensor memory: 26 GB

# Directable Dynamics

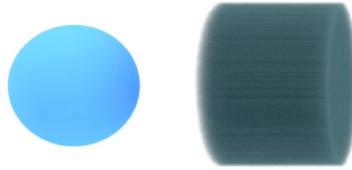


Default tensor

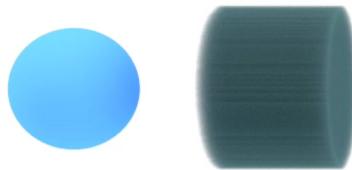


$c = 0.0005$

# Directable Dynamics



Negative weight,  $c = 0.0005$

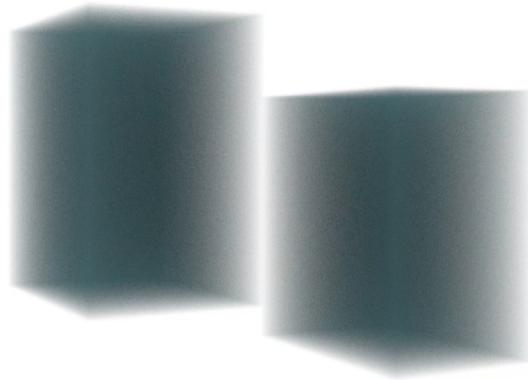


Mixed positive and negative weights

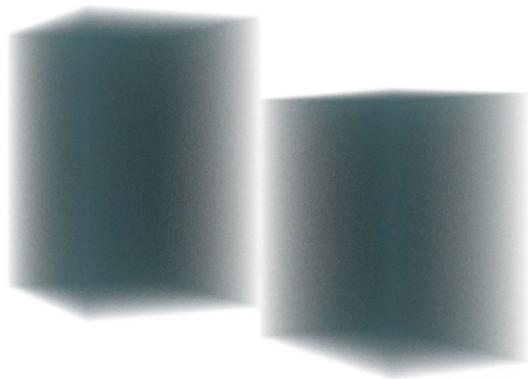
# Outline

- Previous work
- Laplacian Eigenfluids
- Our methods
- Results
- Conclusions and future work

# Scaling effects



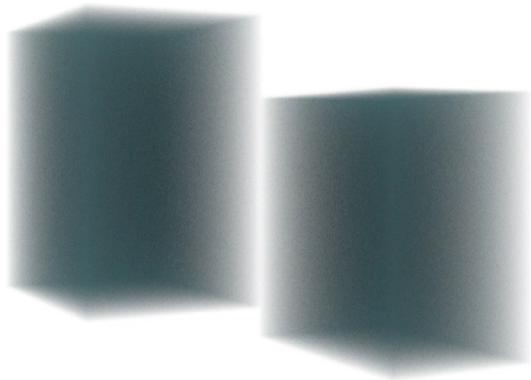
200 basis functions  
[DeWitt et al.] memory: 52 GB  
Our Memory: 0.9 GB



3K basis functions

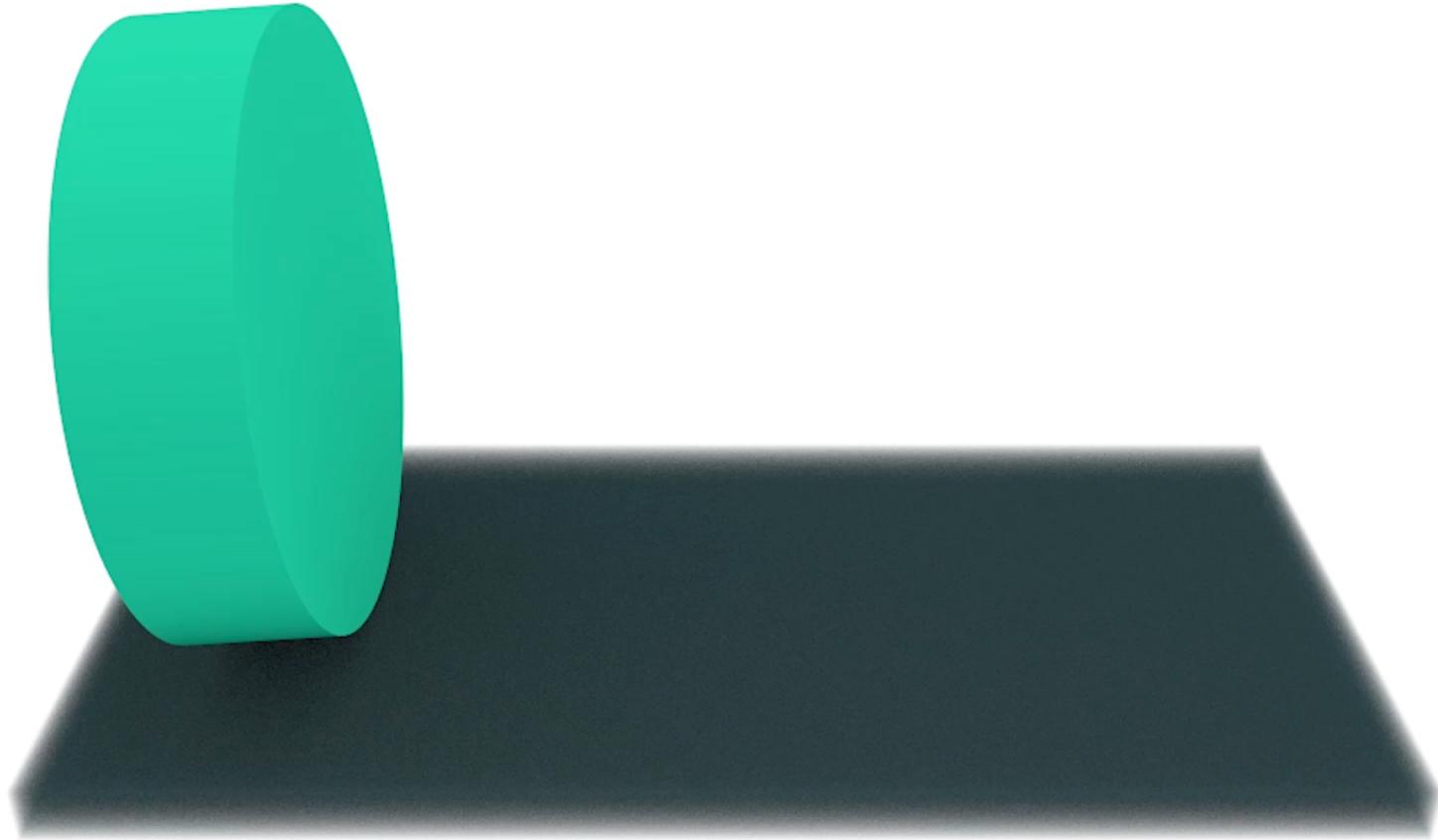
[DeWitt et al.] memory: 0.76 TB

Our Memory: 1.3 GB



24K basis functions  
[DeWitt et al.] memory: 6.1 TB  
Our Memory: 26 GB

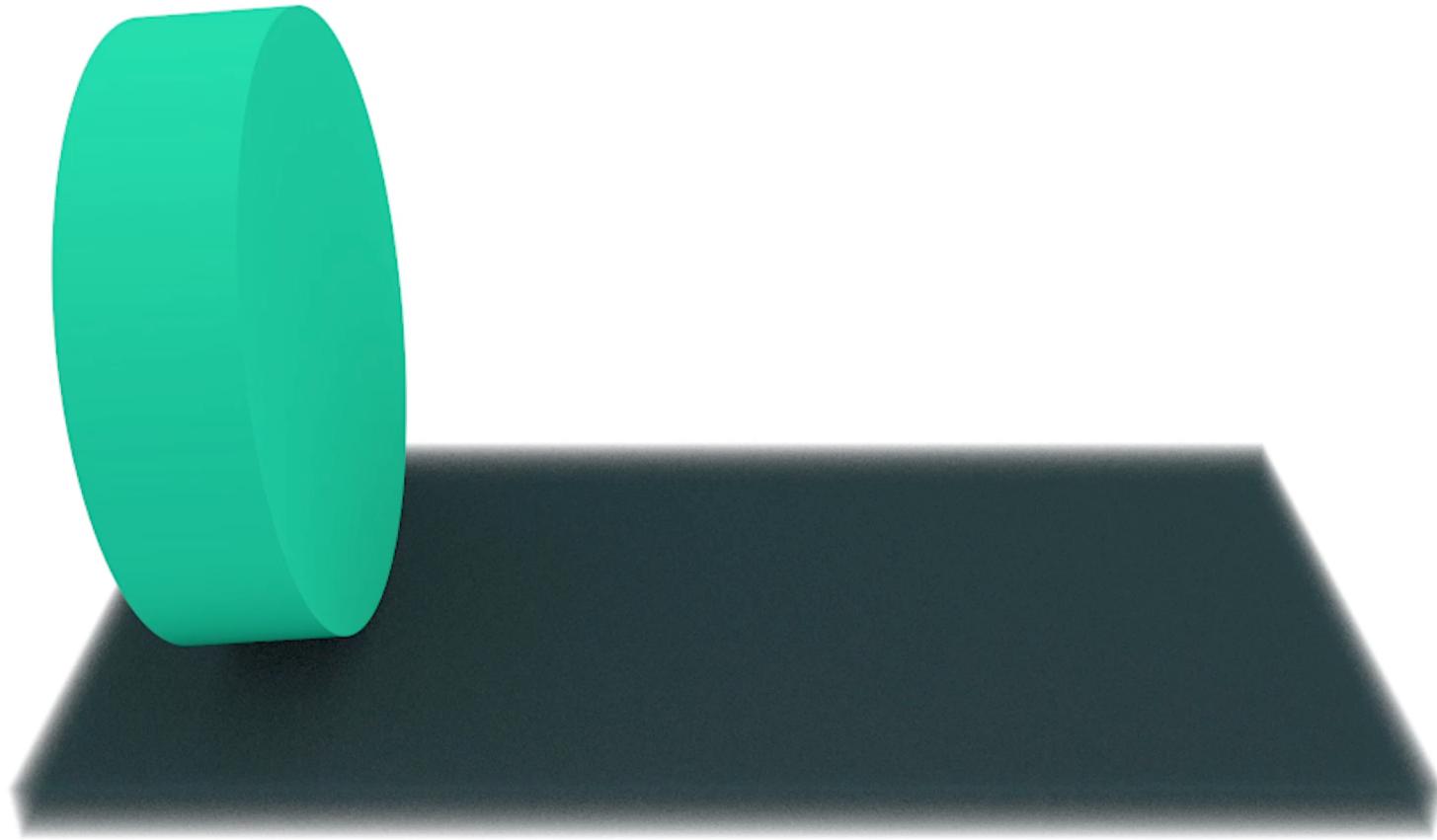
# Paddle Wheel



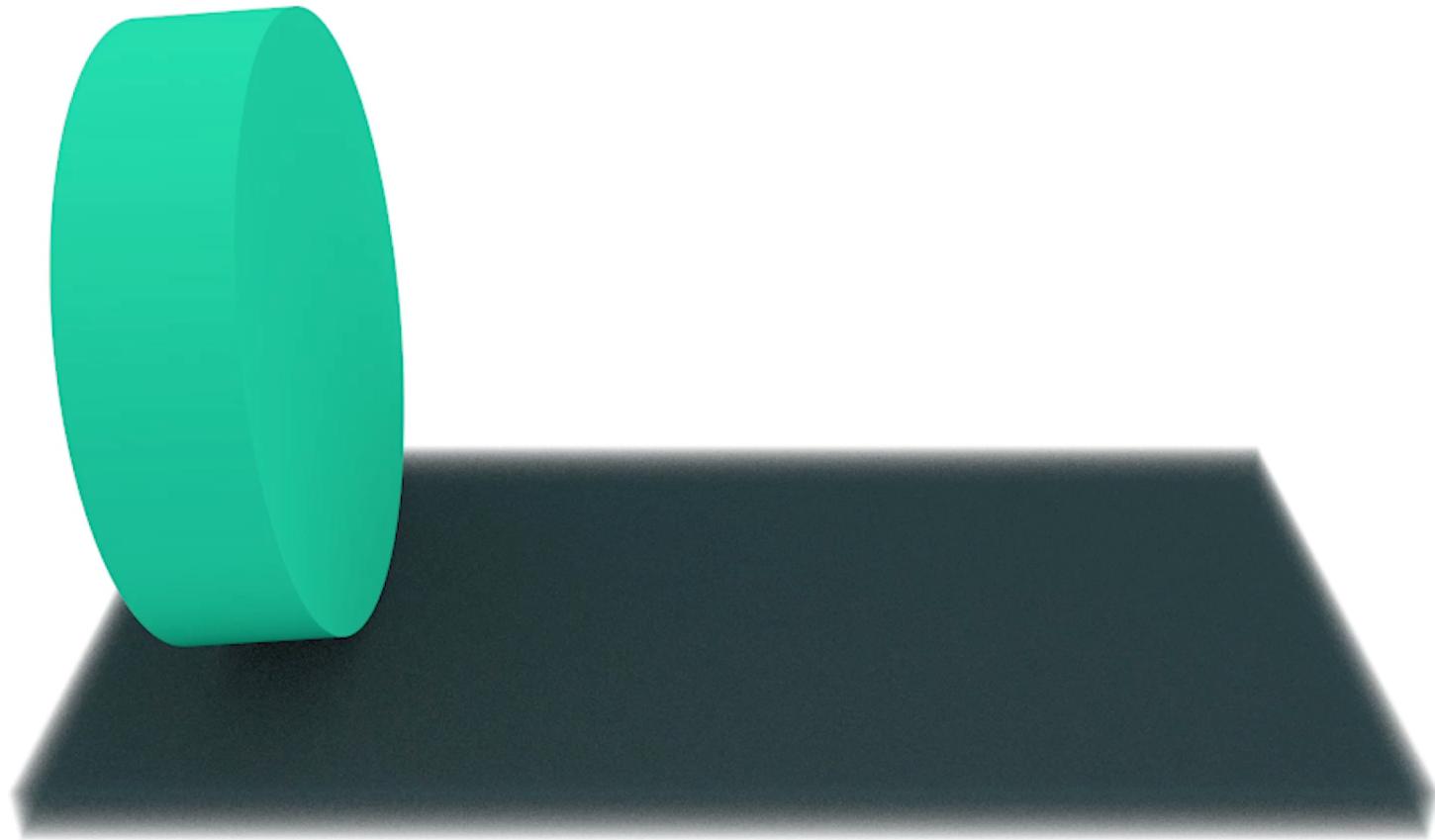
200 basis functions

[DeWitt et al.] memory: 38 GB

Our Memory: 0.9 GB



4 K basis functions  
[DeWitt et al.] memory: 0.8 TB  
Our Memory: 5.1 GB



12 K basis functions  
[DeWitt et al.] memory: 2.3 TB  
Our Memory: 47.1 GB

# Speed Comparison

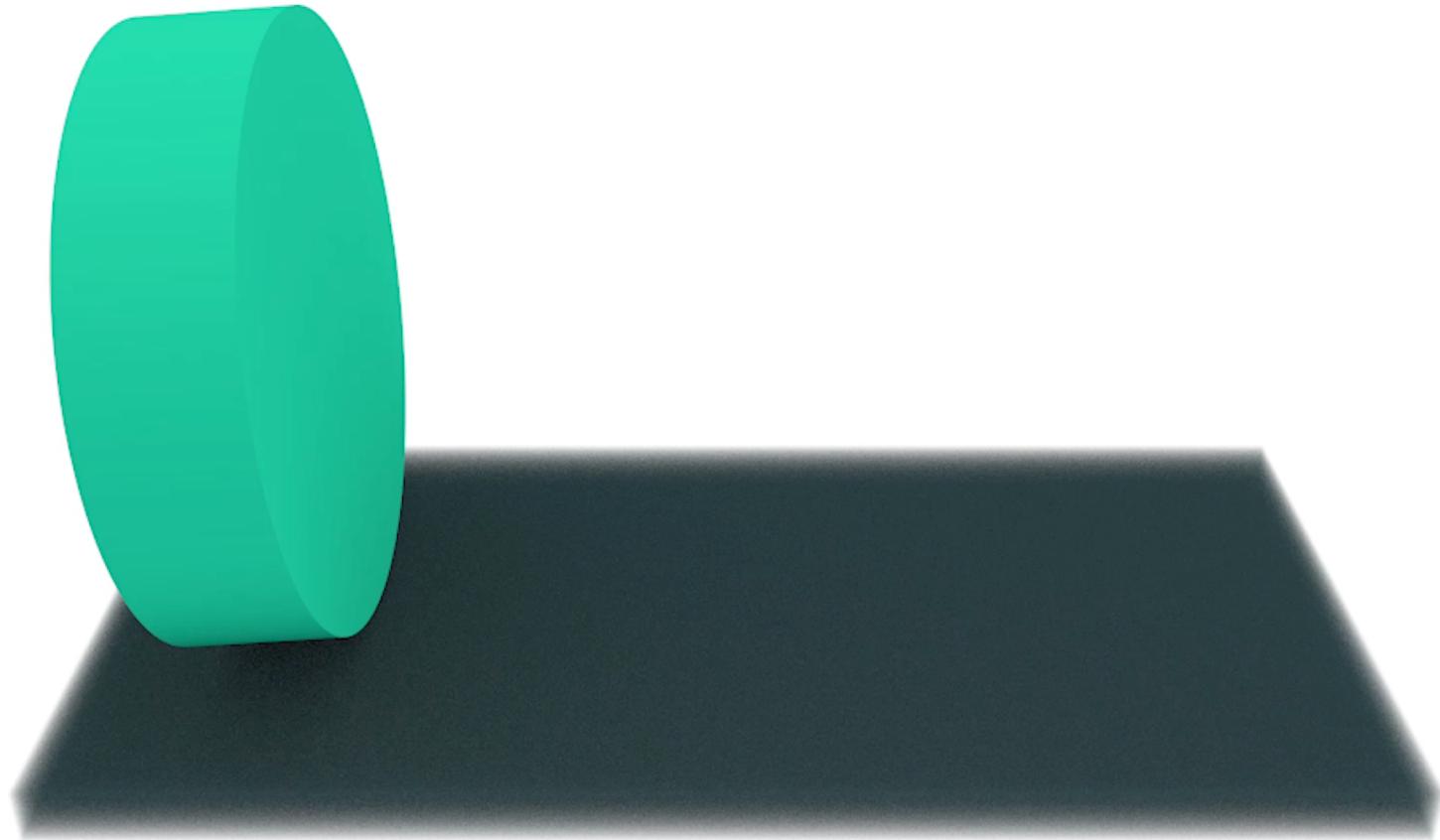
$128^3$ , 1000 basis functions

On the fly basis: 44.10 secs **440×**

Cached basis: 9.54 secs **95×**

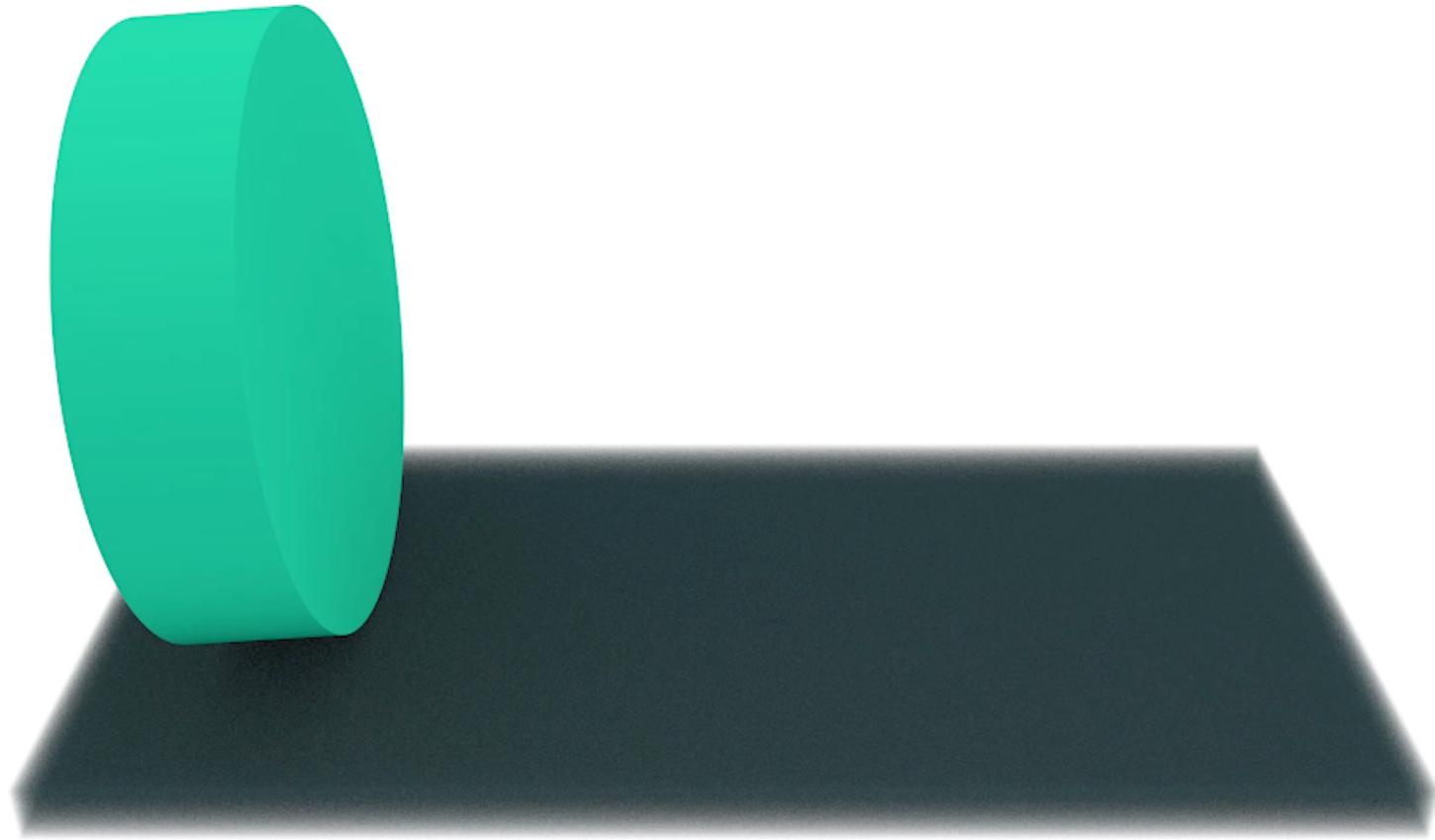
Ours: 0.10 secs

# Varying Viscosity



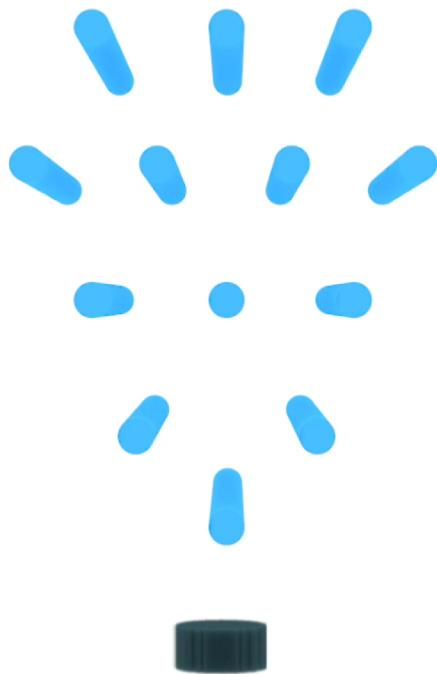
Viscosity = 0.00035

# Varying Viscosity

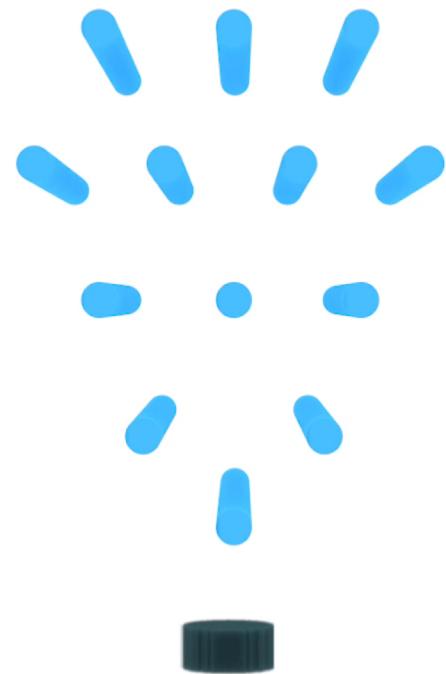


Viscosity = 0.00

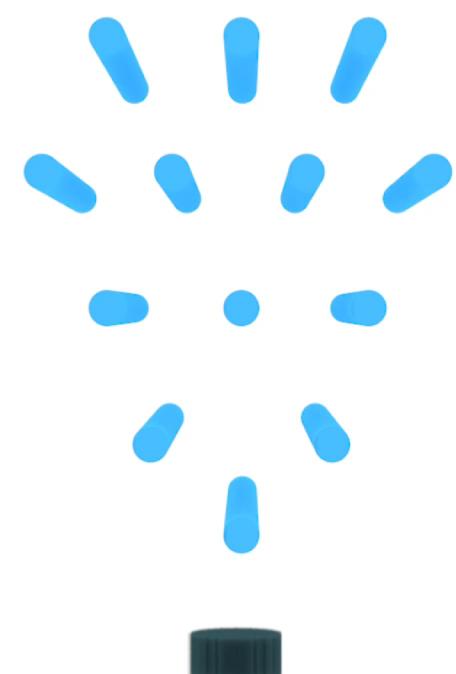
# Interaction with Obstacles



$r = 200$

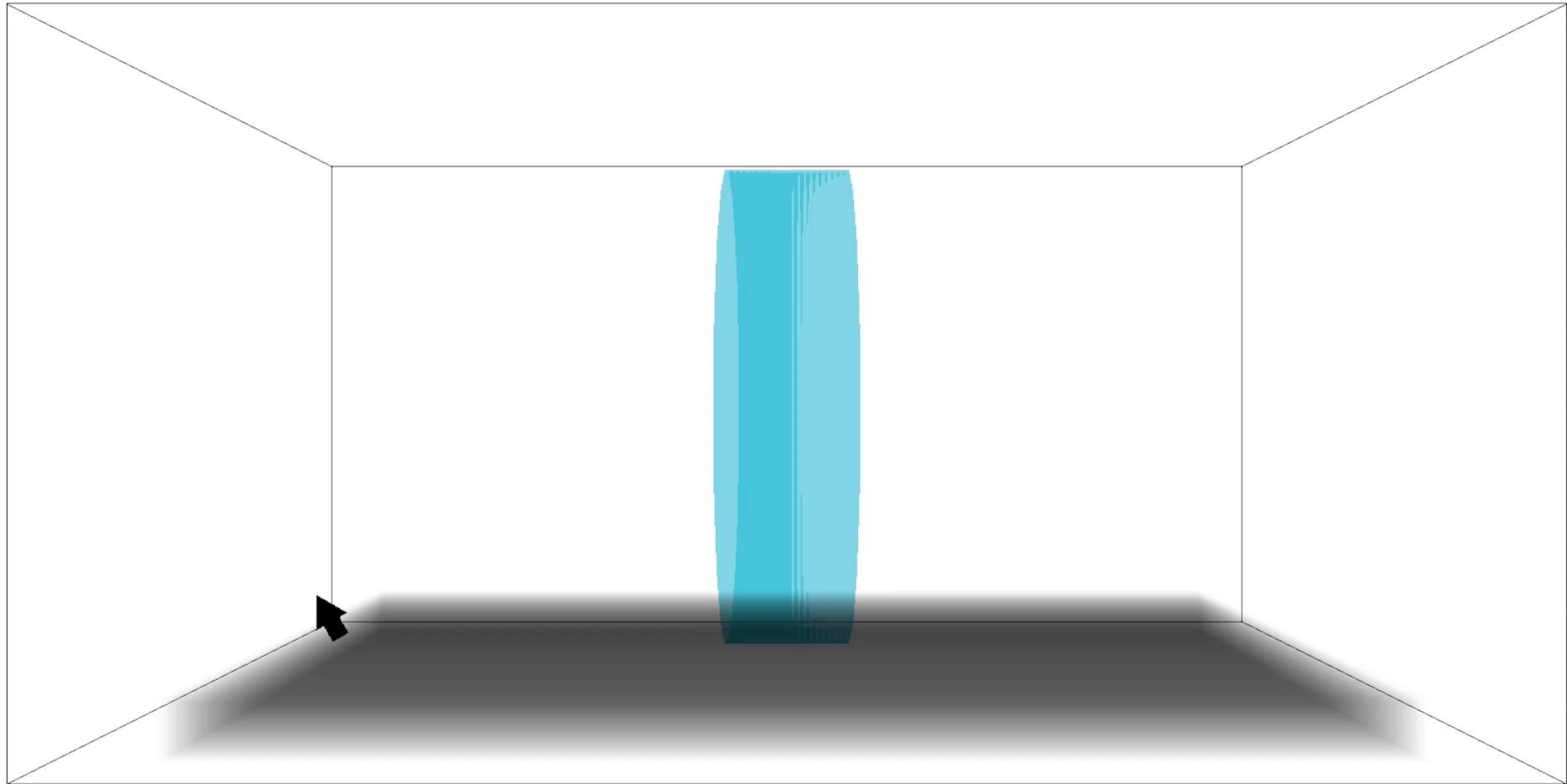


$r = 1000$



$r = 7000$

# Real-time Interaction



13 frames per second

# Outline

- Previous work
- Laplacian Eigenfluids
- Our methods
- Results
- Conclusions and future work

# Conclusions

- Asymptotically superior in memory and time
  - Memory:  $O(rN^3) \longrightarrow O(r)$
  - Time:  $O(rN^3) \longrightarrow O(N^3 \log(N))$
- Support Neumann velocity boundaries
- Directable dynamics
- Advection tensor lossy compression
  - Up to 99.9 %

# Limitations

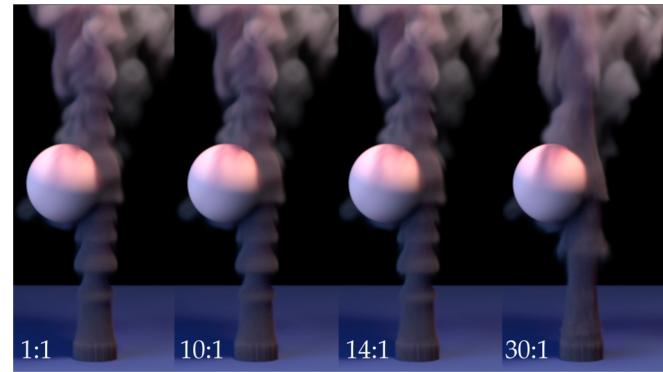
- Rectangular domain
- Uniform boundary condition
- Energy cascade is capped at the highest frequency
- Penalty methods for obstacles

# Future Work

- Advection tensor compression
- Tiled domains
- Wavelets



[Wicke et al. 2009]



[Jones et al. 2016]

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- National Science Foundation (NSF)
  - CAREER award IIS-1253948
  - IIS-1321168
  - IIS-1619376
- UCSB Center for Scientific Computing
  - NSF MRSEC (DMR-1720256)

# Thanks!

Source code is available:

[http://cvc.ucsb.edu/graphics/Papers/SIGGRAPH2018\\_EigenFluid/](http://cvc.ucsb.edu/graphics/Papers/SIGGRAPH2018_EigenFluid/)

