

Massively parallel split-step Fourier techniques for simulating quantum systems on graphics processing units

James Schloss

Advisor: **Thomas Busch**
Quantum Systems Unit

December 9, 2019

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Physics: understanding superfluid vortices

Bose–Einstein Condensate



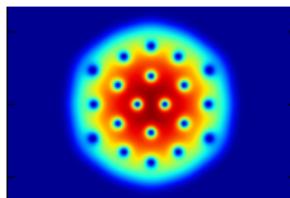
Vortex generation



2D vortex chaos



3D vortex structures



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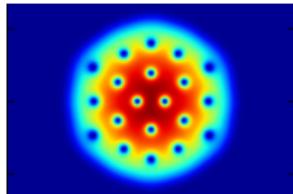
Vortex generation



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Computer Science: SSFM on GPUs

Split-step Fourier method



GPU hardware



GPU codebase



Optimizations

$$\mathrm{GP}^{\binom{\hat{U}}{E}}$$

Physics: understanding superfluid vortices

Computer Science: SSFM on GPUs

Quantum state engineering

(*NJP* 18 (3):035012, 2016)

J Schloss, A Benseny, J Gillet, J Swain,

T Busch

Implementation details

(Tests, formats, etc.)

The Fourier Transform

Fourier Transform:

$$F(\xi) = \int_{-\infty}^{\infty} f(t) e^{-2\pi i t \xi} dt$$

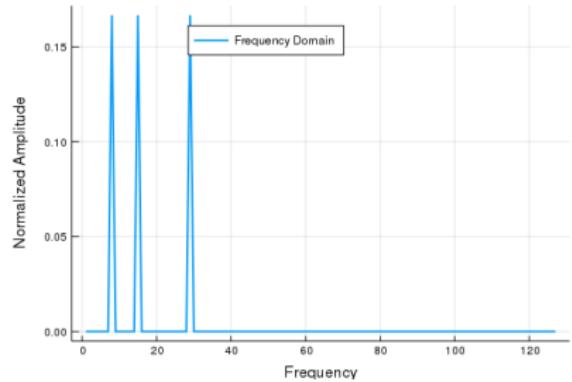
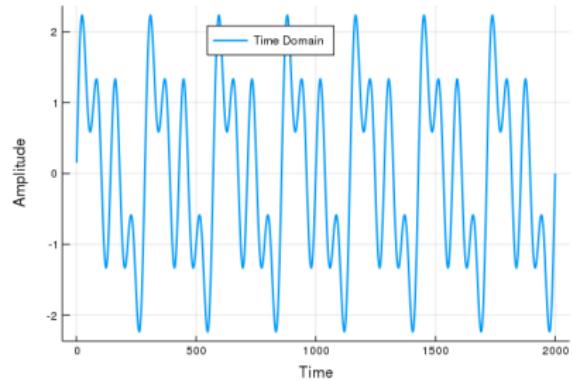
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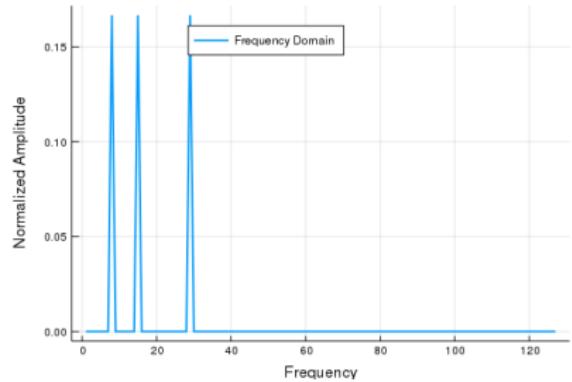
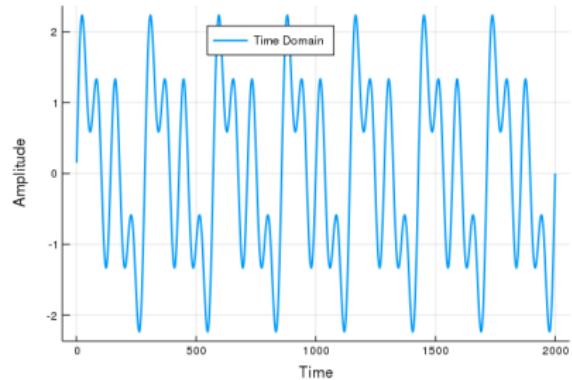
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Discrete Fourier Transform:

$$X_k = \sum_{n=0}^{N-1} x_n e^{2\pi i k n / N}$$

$$x_n = \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{-2\pi i k n / N}$$



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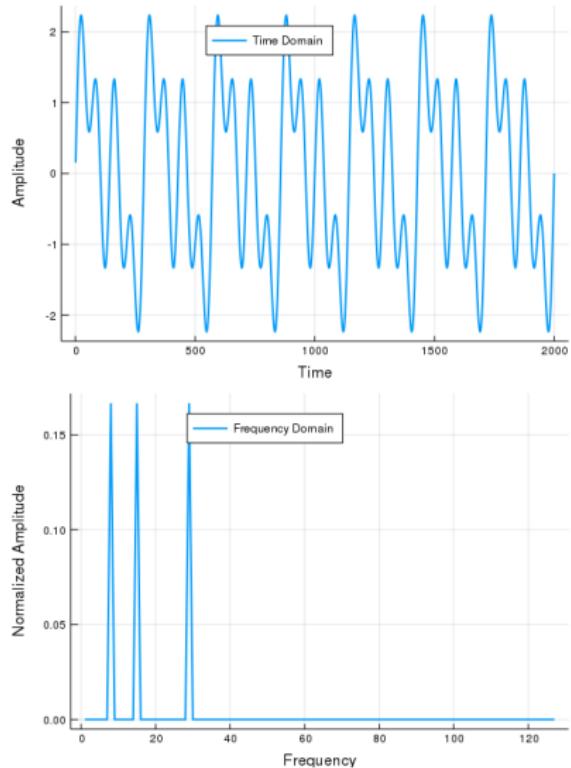
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The FFT is a *global* operation requiring iteration or recursion



Hamiltonian

1D Schrödinger equation:

$$i\hbar \frac{\partial \Psi(x, t)}{\partial t} = \left(\frac{p^2}{2m} + \frac{1}{2} m\omega^2 x^2 \right) \Psi(x, t)$$

Hamiltonian

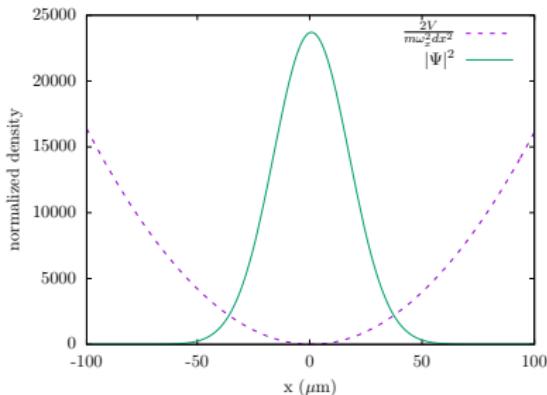
1D Schrödinger equation:

$$i\hbar \frac{\partial \Psi(x, t)}{\partial t} = \left(\frac{p^2}{2m} + \frac{1}{2} m\omega^2 x^2 \right) \Psi(x, t) = \mathcal{H}\Psi(x, t)$$

Splits into:

$$\mathcal{H}_v = \frac{1}{2} m\omega^2 x^2$$

$$\mathcal{H}_p = \frac{p^2}{2m} = \frac{1}{2m} \left(i\hbar \frac{\partial}{\partial x} \right)^2$$



Split-Step Fourier Method (SSFM)



Differential Equations:

$$\Psi(x, t + dt) = \left[e^{-\frac{i\mathcal{H}dt}{\hbar}} \right] \Psi(x, t) = \left[e^{-\frac{i(\mathcal{H}_v + \mathcal{H}_p)dt}{\hbar}} \right] \Psi(x, t)$$

Differential Equations:

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Baker-Campbell-Hausdorff:

$$\Psi(x, t + dt) = \left[e^{-\frac{i\mathcal{H}_v dt}{\hbar}} e^{-\frac{i\mathcal{H}_p dt}{\hbar}} e^{-\frac{[i\mathcal{H}_v, i\mathcal{H}_p]dt^2}{2}} \right] \Psi(x, t)$$

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Matrices:

$$U_v = e^{-\frac{i\mathcal{H}_v dt}{\hbar}}$$

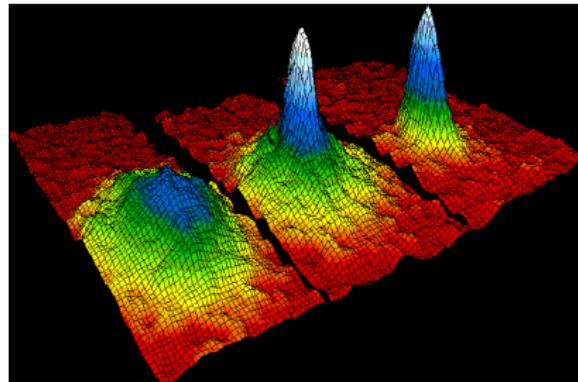
$$U_p = e^{-\frac{i\mathcal{H}_p dt}{\hbar}}$$

$$\Psi(x, t + dt) = \mathcal{F}^{-1} [U_p(dt) \mathcal{F} [U_v(dt) \Psi(x, t)]] + \mathcal{O}(dt^2)$$

Bose-Einstein condensation



Bosons condense into a superfluid BEC

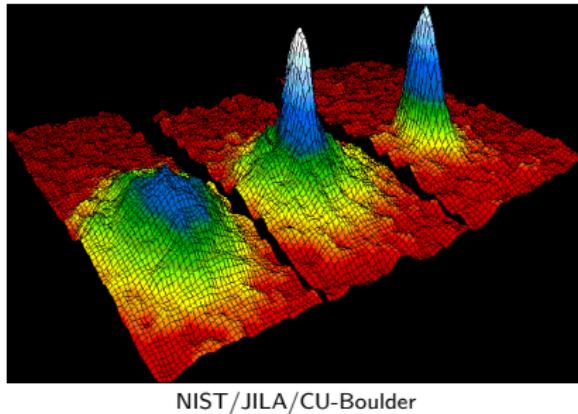


NIST/JILA/CU-Boulder

Bose-Einstein condensation



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Described by the mean field Gross–Pitaevskii equation:

$$i\hbar \frac{\partial}{\partial t} \Psi(x, t) = \left(-\frac{\hbar^2}{2m} \nabla^2 + \frac{1}{2} m \omega^2 x^2 + g |\Psi(x, t)|^2 \right) \Psi(x, t)$$

Superfluid rotation

Rotation leads to many vortices along the axis of rotation

Fluid



Credit: howstuffworks.com

Superfluid rotation

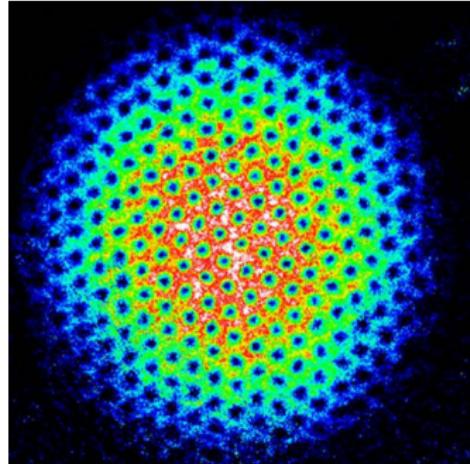
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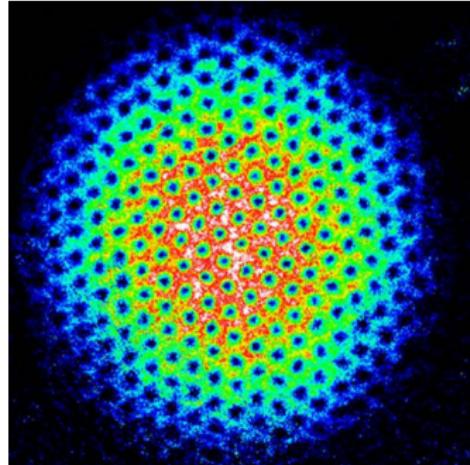
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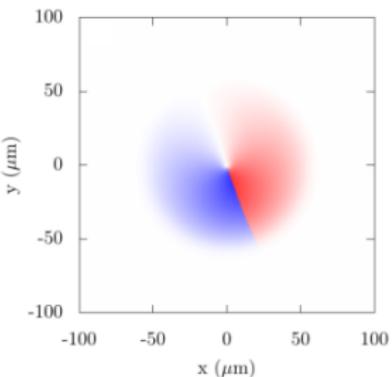
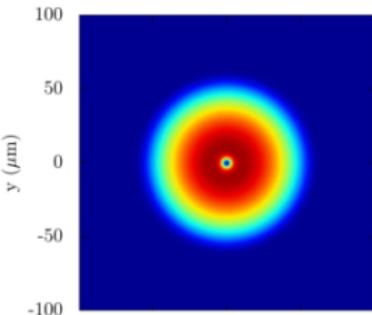
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$$i\hbar \frac{\partial}{\partial t} \Psi(x, t) = \left(-\frac{\hbar^2}{2m} \nabla^2 + V_0(x) + g|\Psi(x, t)|^2 - \Omega L_z \right) \Psi(x, t)$$

$$L_z = (xp_y - yp_x) = -i\hbar \left(x \frac{\partial}{\partial y} - y \frac{\partial}{\partial x} \right)$$

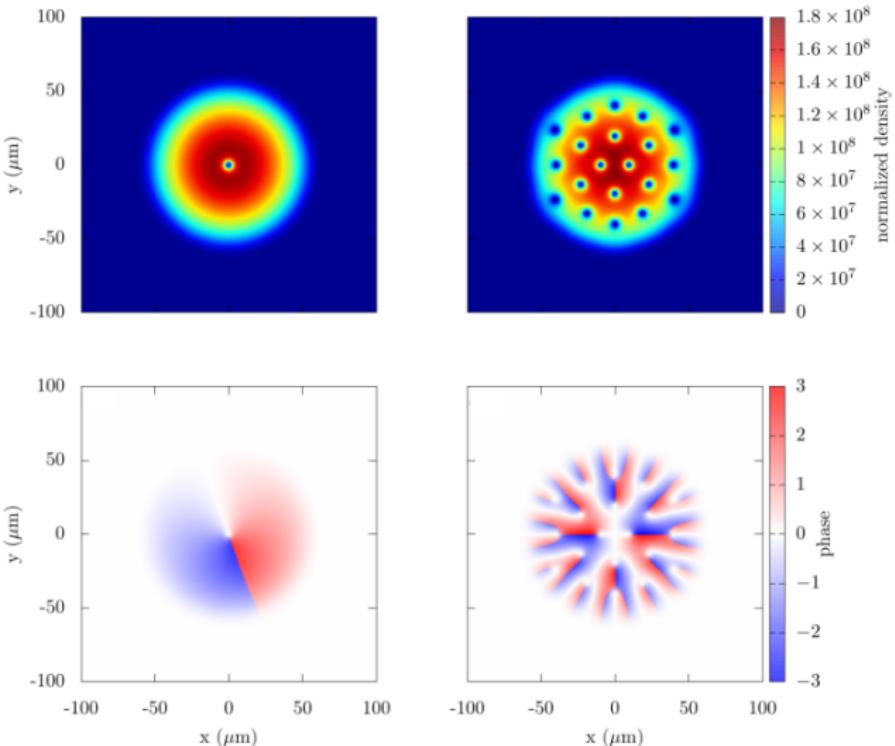
In the co-rotating frame

Superfluid vortex phase



Each vortex has a 2π complex phase winding, $v \sim \nabla\phi$

Superfluid vortex phase

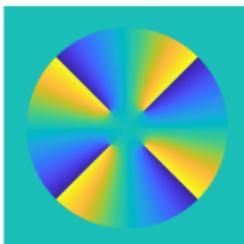


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What else can we do?

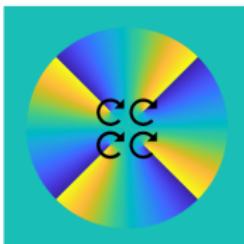
Phase manipulation

Initial

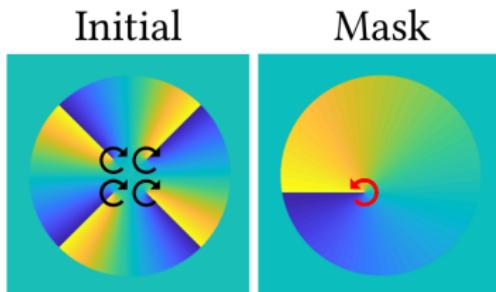


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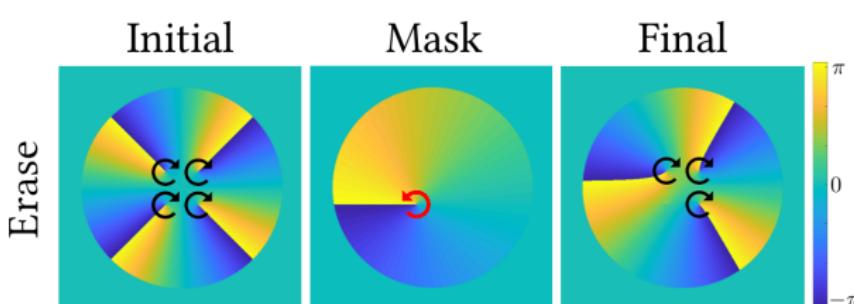


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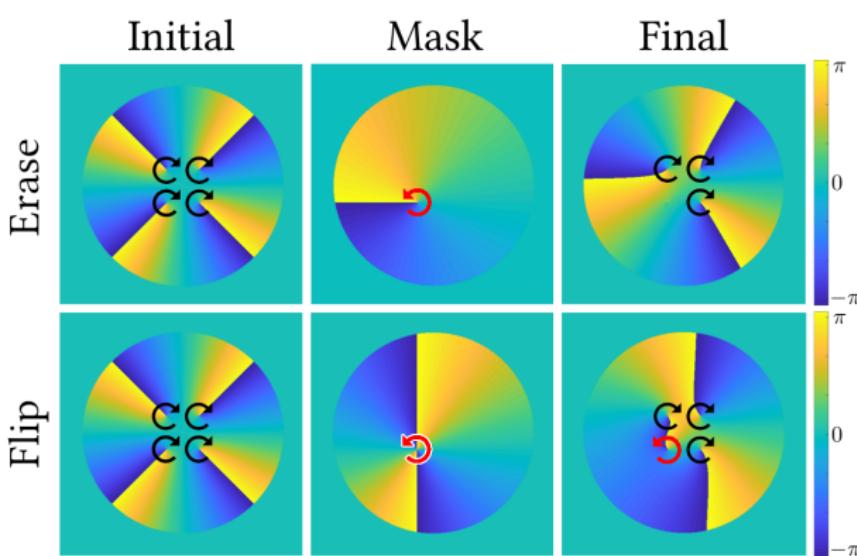
Phase masks can change the vortex distribution

Phase manipulation



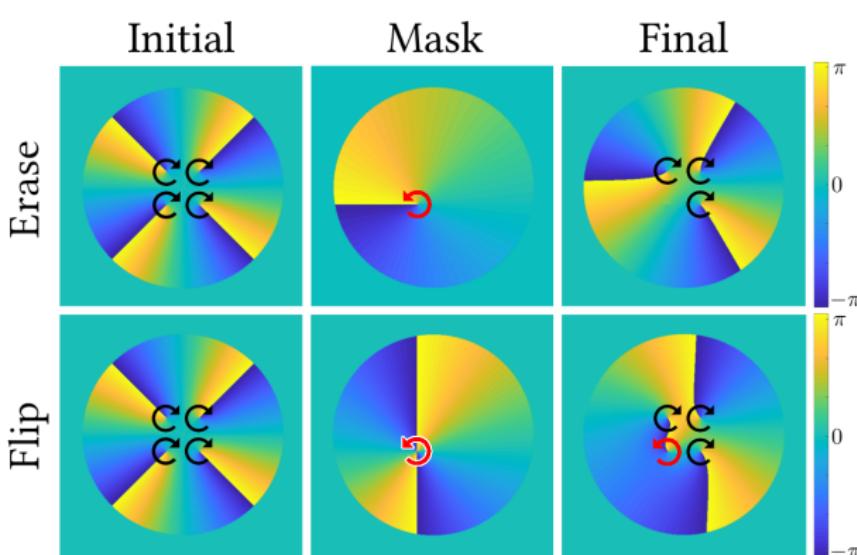
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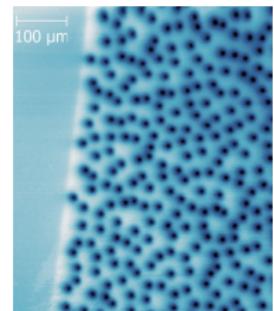


Phase masks can change the vortex distribution
 ... but will create energetic condensates

Artificial magnetic fields



Magnetic fields cause rotation in *charged* particles
(Example: Type II superconductors)



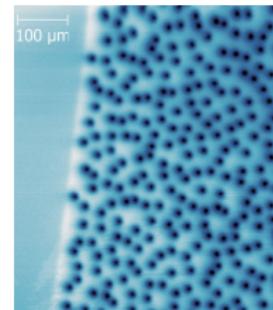
FS Wells *et al.*, *Scientific Reports*
5: 8677, 2015

Artificial magnetic fields

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- ▶ The Hamiltonian of a particle with the Lorentz force law is:

$$\mathcal{H} = \frac{(p - q\mathbf{A}(x))^2}{2m}, \quad \mathbf{B} = \nabla \times \mathbf{A}$$



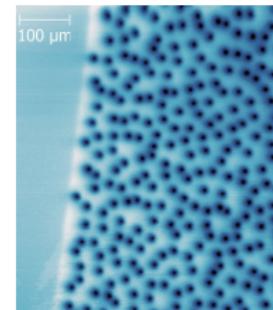
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- ▶ Vortices follow the magnetic field lines

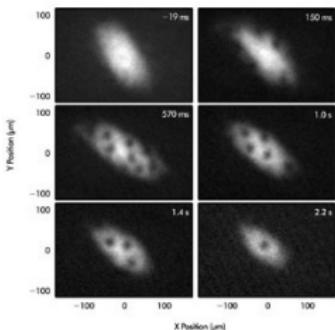
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Ian Spielman, NIST

GPE with gauge fields



With gauge fields, the GPE becomes

$$\mathcal{H} = \frac{(p - m\mathbf{A}(x))^2}{2m} + V_0 + g|\Psi(x, t)|^2$$

GPE with gauge fields

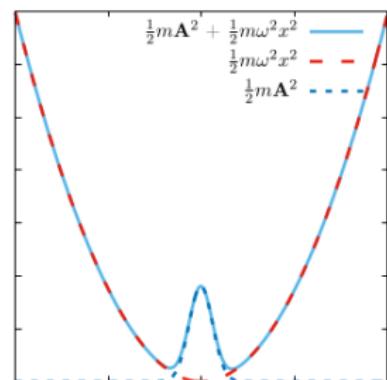
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Which creates

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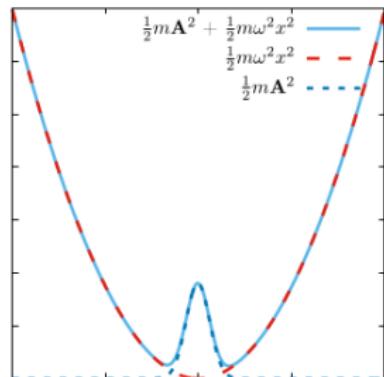
Which creates

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$$\frac{m\mathbf{A}(x)^2}{2}$$

- ▶ Components in position and momentum-space, that **require 1D FFTs**

$$-\left(\frac{p\mathbf{A}(x) + \mathbf{A}(x)p}{2}\right)$$



Physics

- ▶ Quantized vortices can be formed in BECs with rotation, phase imprinting, and artificial magnetic fields

Physics

- ▶ Quantized vortices can be formed in BECs with rotation, phase imprinting, and artificial magnetic fields

Computer Science

- ▶ The SSFM requires a large number of FFTs
- ▶ This system is a great for testing spectral-like methods

GPU computing and the GPUE codebase

J Schloss, LJ O'Riordan

Journal of Open Source Software 3 (32):1037, 2018

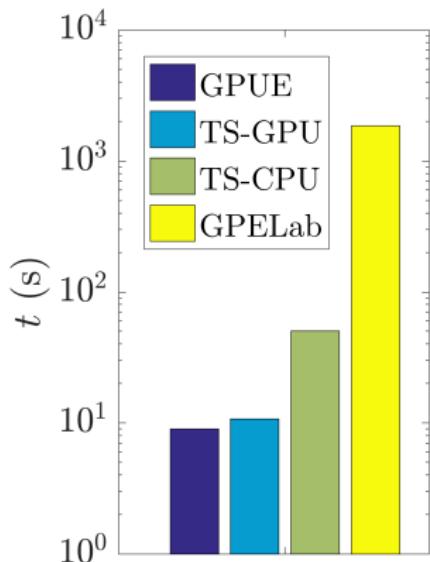
What is a GPU?

- ▶ Graphics processing units (GPUs) are massively parallel computing devices



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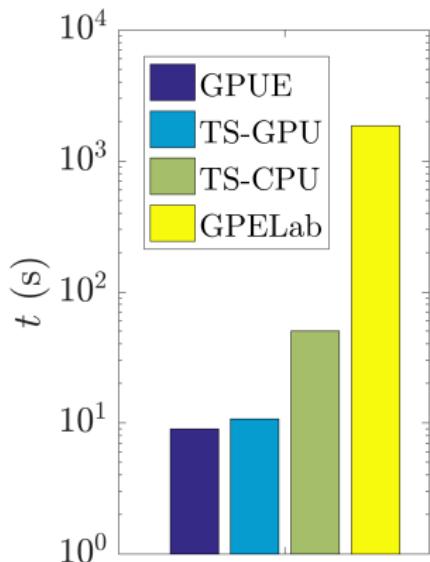
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Total runtime, smaller has higher performance!

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- ▶ Graphics processing units (GPUs) are massively parallel computing devices
- ▶ GPUs are fast for parallel tasks
- ▶ Summit uses GPUs



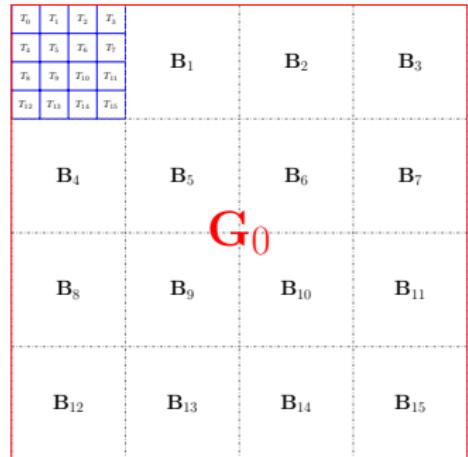
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GPU memory hierarchy



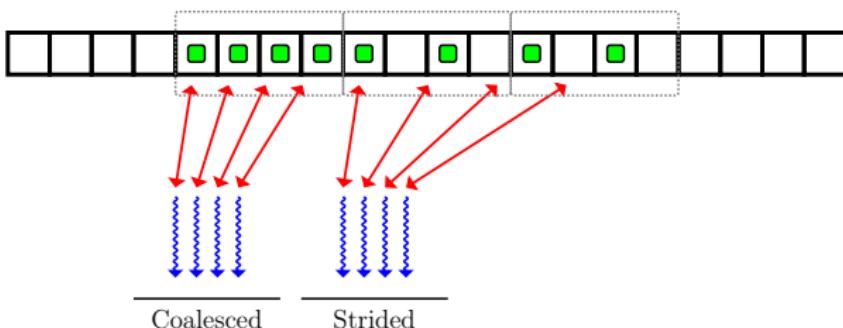
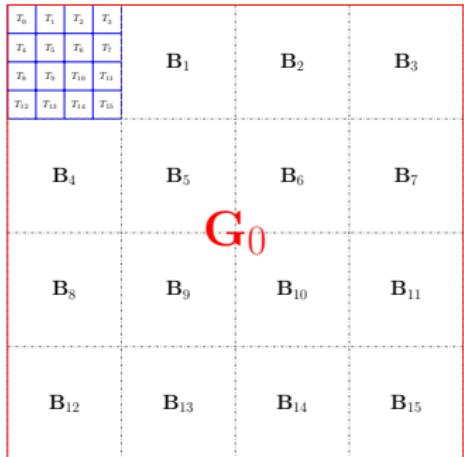
GPU memory hierarchy

- ▶ Computing threads in blocks, blocks in grids



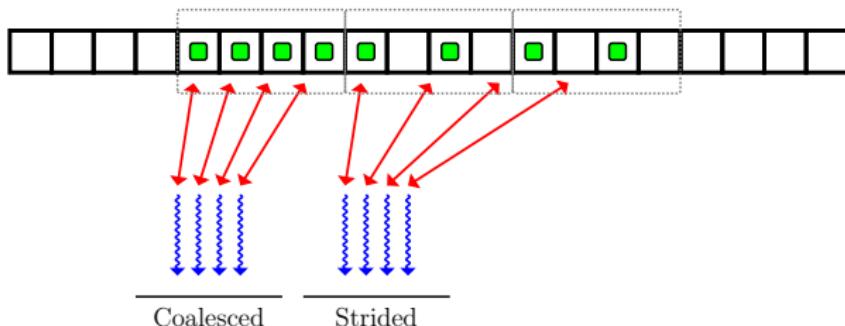
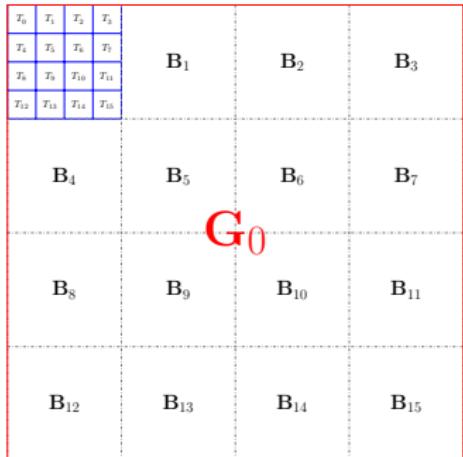
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- ▶ Computing threads in blocks, blocks in grids
- ▶ Memory coalescence



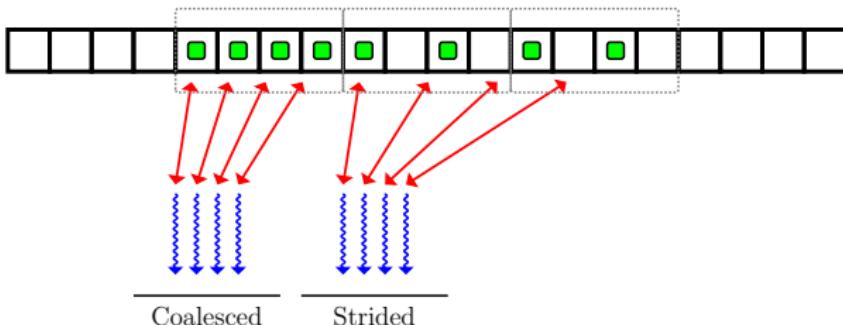
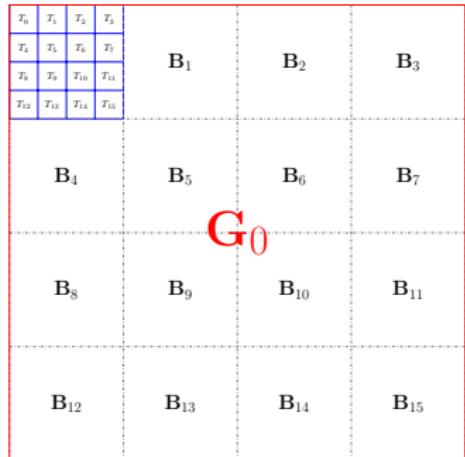
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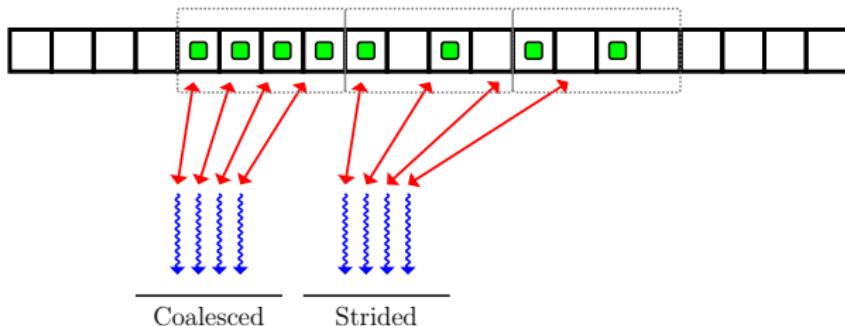
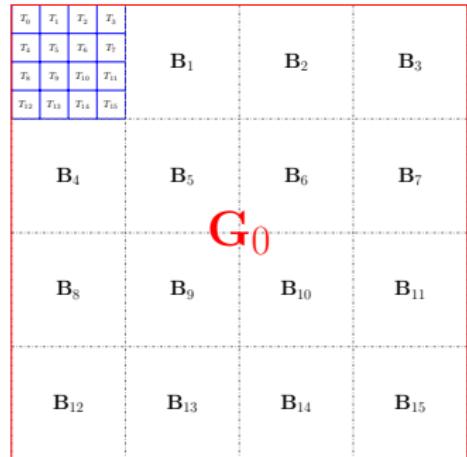
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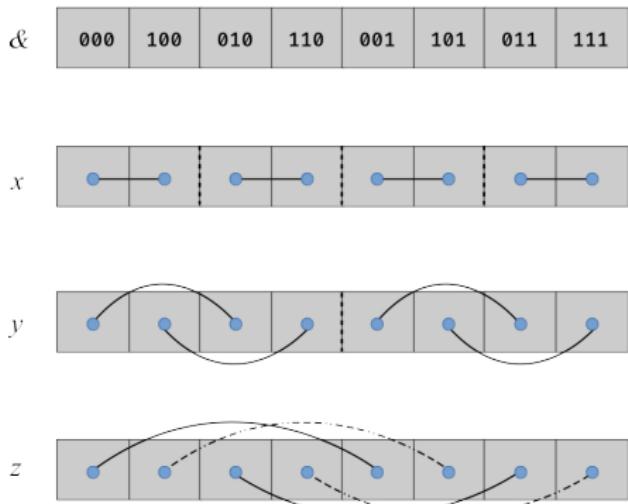
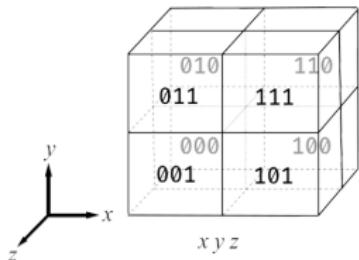


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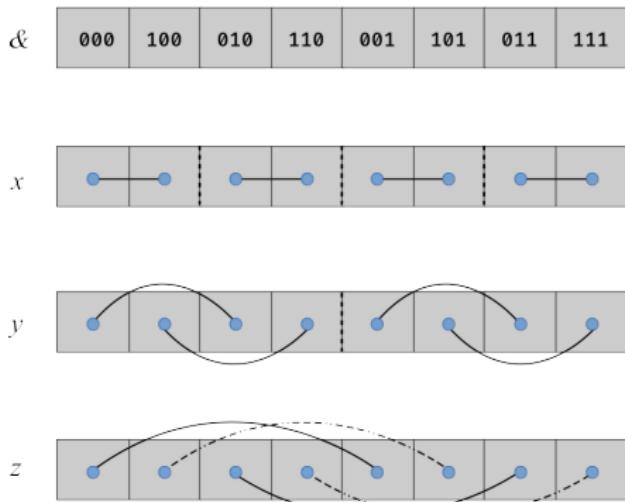
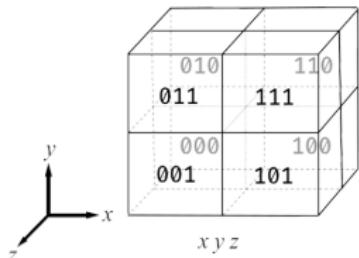
- ▶ Computing threads in blocks, blocks in grids
- ▶ Memory coalescence
- ▶ Data transfer is slow
- ▶ Recursion and iteration is slow
- ▶ Limited memory



- ▶ Global operations
- ▶ 1D FFTs are not supported by CuFFT



- ▶ Global operations
- ▶ 1D FFTs are not supported by CuFFT
- ▶ Transposes are necessary



Two important features

$$\text{GP}^{\hat{U}}_E$$

- ▶ Novel features useful for understanding examples
- ▶ No discussion of implementation details

How do we scale?



Notes:

- ▶ GPUs have limited memory, so we should minimize the memory footprint with dynamic kernels

How do we scale?



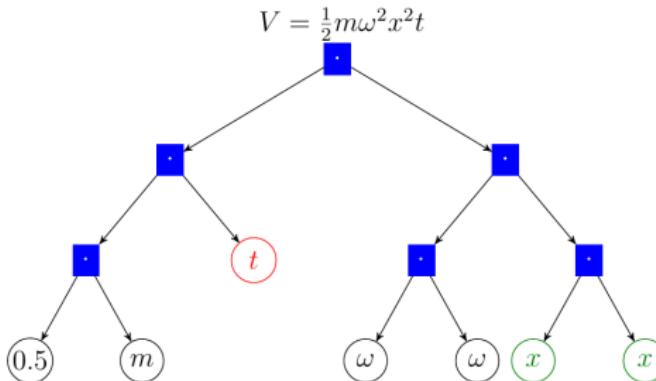
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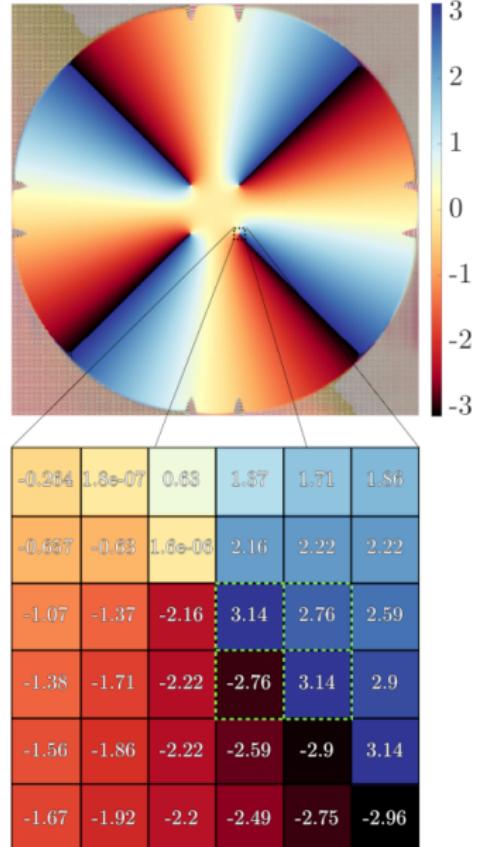
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Expression trees

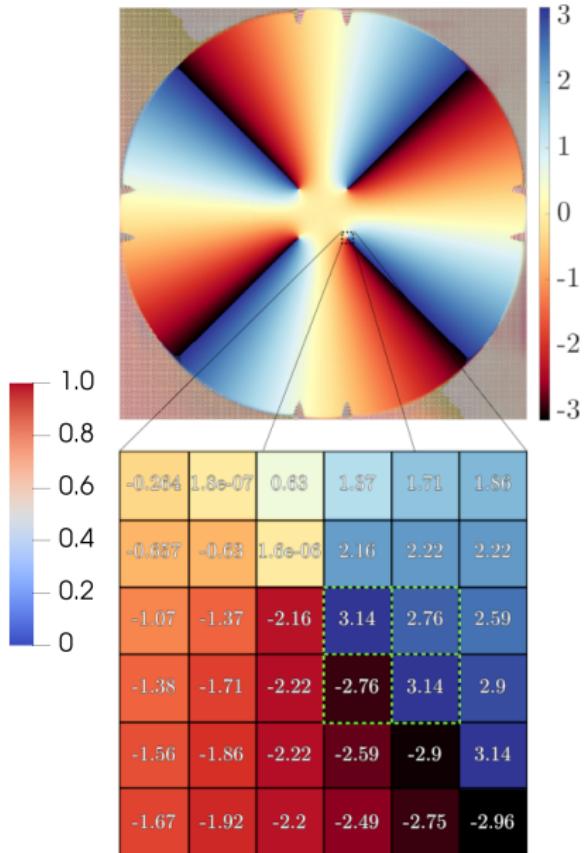
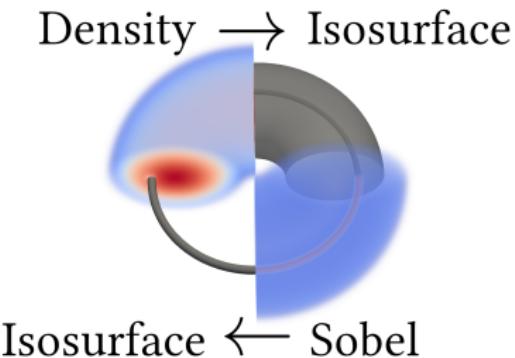
Vortex tracking and highlighting

- ▶ Phase plaquettes in 2D



Vortex tracking and highlighting

- ▶ Phase plaquettes in 2D
- ▶ Vortex highlighting in 3D



Physics

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- ▶ Multi-component simulations and HDF5 have also been implemented

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Computer Science

- ▶ Expression trees save a lot of GPU memory
- ▶ Distributed transpose methods are in development

Chaotic vortex dynamics in 2D BEC

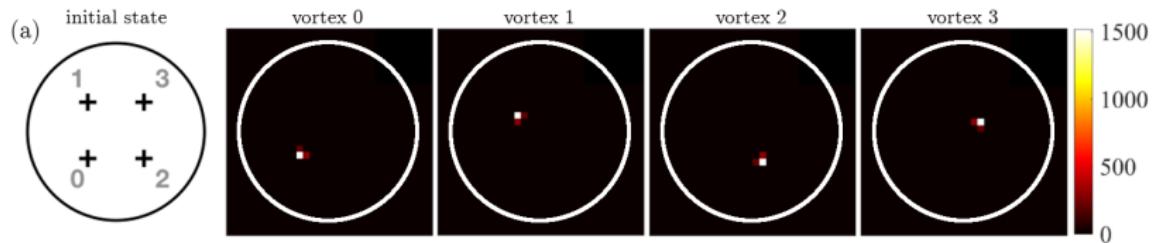
T Zhang, J Schloss, A Thomasen, LJ O'Riordan, T Busch, A White

Physical Review Fluids 4 (5):054701, 2019

Few vortex dynamics



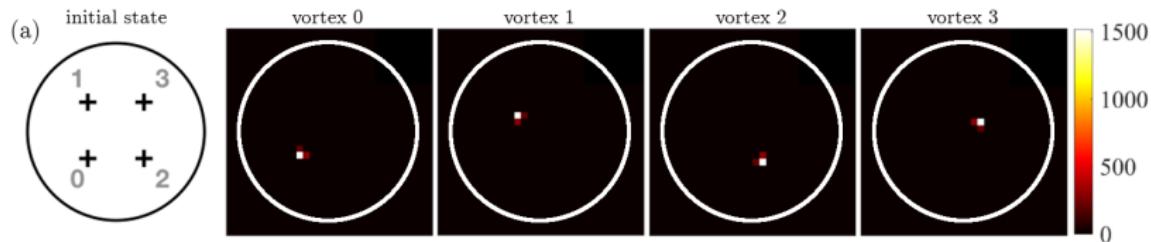
4 vortices:



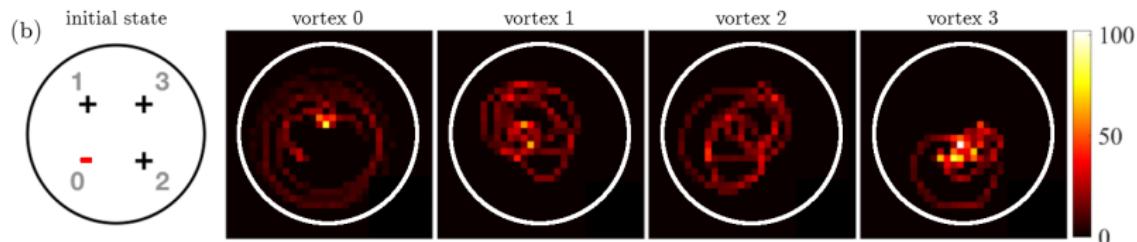
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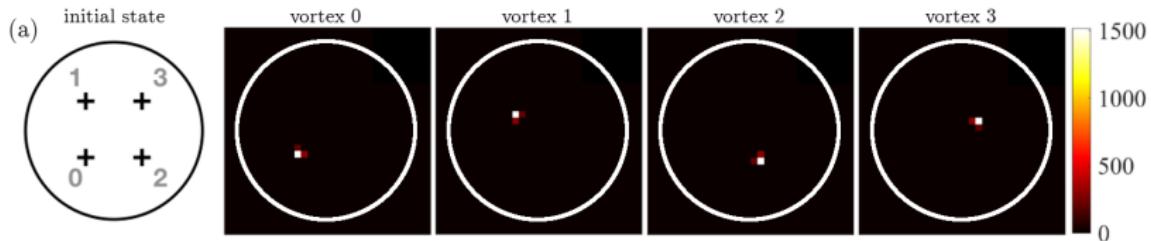
4 Vortex, 1 anti-vortex:



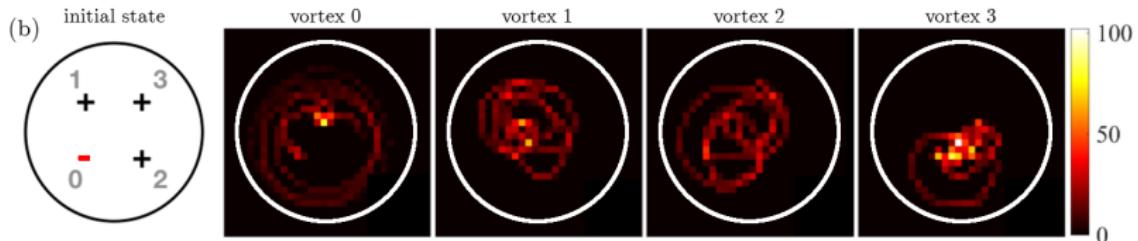
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Is this chaotic?

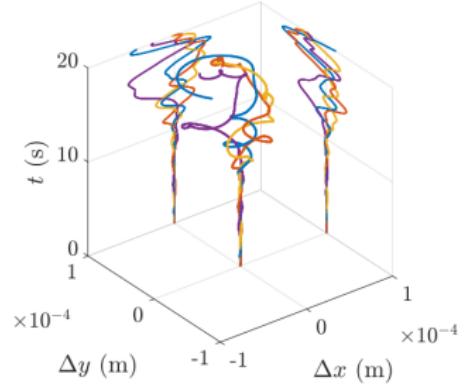
Divergence in trajectories



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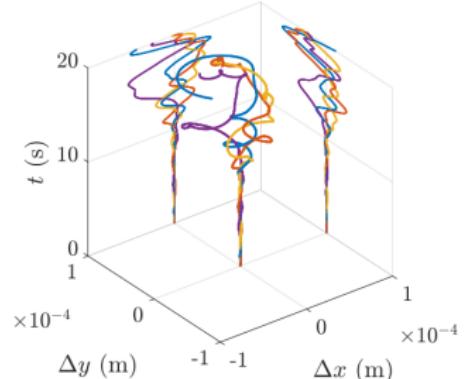
Divergence in trajectories at $\sim 10s$



Divergence in trajectories



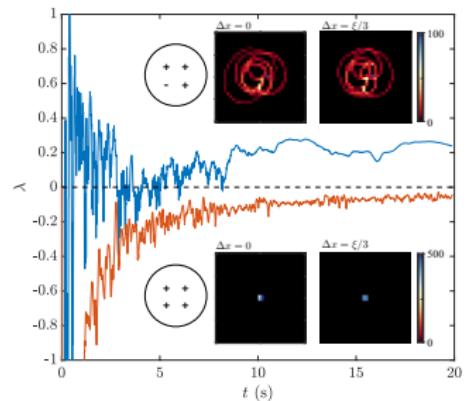
Divergence in trajectories at $\sim 10\text{s}$



Divergence in trajectory when Lyapunov exponent (λ) becomes positive

$$\delta \mathbf{P}(t) = (\delta x(t), \delta y(t), \delta v_x(t), \delta v_y(t))$$

$$|\delta \mathbf{P}(t)| \approx e^{\lambda t} |\delta \mathbf{P}_0|$$



Divergence in trajectories



Physics

- ▶ This is a chaotic system with a controlled initial state
- ▶ Chaotic dynamics of few-vortex systems is accelerated by collisional events

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- ▶ Chaotic dynamics of few-vortex systems is accelerated by collisional events

Computer Science

- ▶ 50 TB of data
- ▶ Reliant on post-processing metrics (Lyapunov exponent, Vortex tracking)
- ▶ 1 hour per run (on K80's), infeasible with other software

3D vortex ring generation in toroidal BEC systems

J Schloss, P Barnett, R Sachdeva, T Busch
arXiv:1910.02364

Submitted to *Physical Review Fluids*

Artificial magnetic fields



Magnetic fields cause rotation in *charged* particles

Artificial magnetic fields

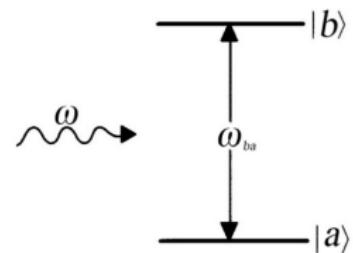
Magnetic fields cause rotation in *charged* particles

- ▶ If a two-level atom moves slowly in a tuned light field, *Berry's connection* is

$$\mathbf{A} = i\hbar \langle \psi_I | \nabla \psi_I \rangle$$

- ▶ The magnetic field is

$$\mathbf{B} = \nabla \times \mathbf{A}$$



Artificial magnetic fields

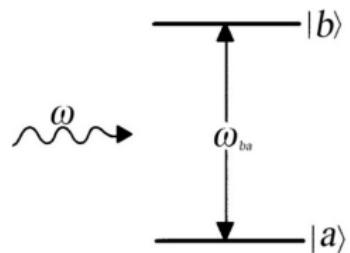
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- ▶ Artificial magnetic fields are created with a strongly varying electric field

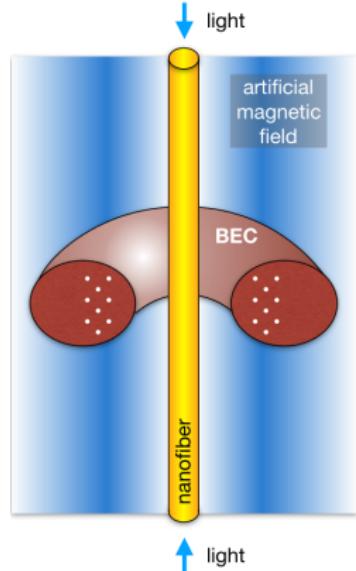
Optical nanofibers and BECs

- ▶ Nanofiber has exponentially decaying evanescent field
- ▶ Nanofiber generates **A**

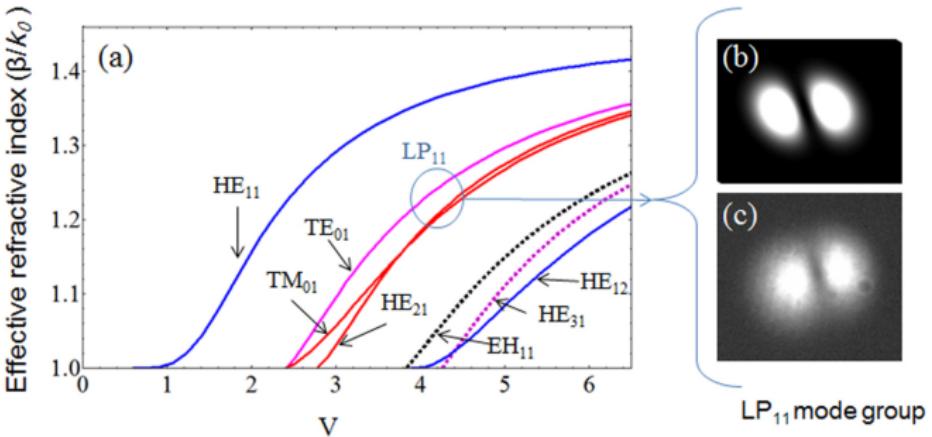


Optical nanofibers and BECs

- ▶ Nanofiber has exponentially decaying evanescent field
- ▶ Nanofiber generates **A**
- ▶ BEC toroidally trapped around nanofiber
- ▶ Vortices follow $\mathbf{B} = \nabla \times \mathbf{A}$



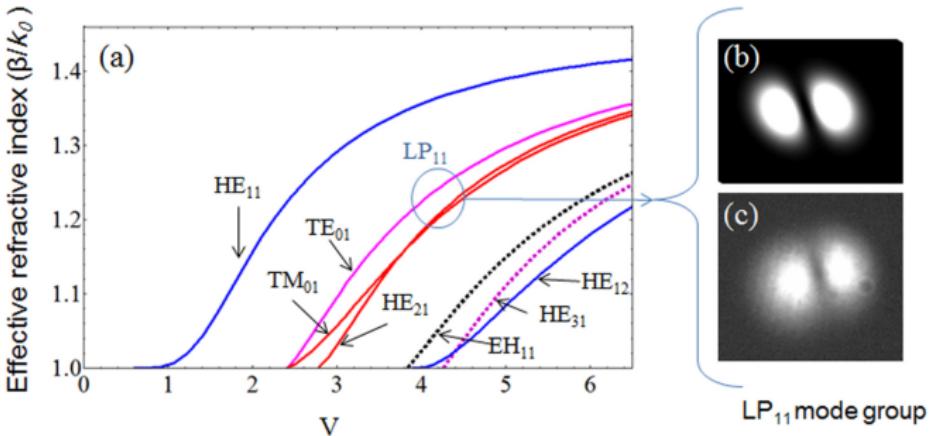
Optical nanofiber modes



Kumar et al. *NJP* 17(1):013026, 2015

$$V = k_0 a \sqrt{n_1^2 - n_2^2}$$

Optical nanofiber modes



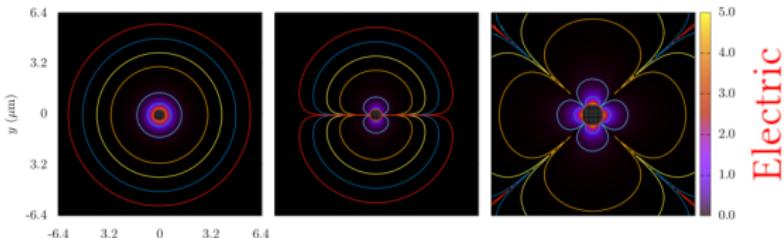
Kumar et al. *NJP* 17(1):013026, 2015

$$V = k_0 a \sqrt{n_1^2 - n_2^2}$$

- ▶ Different optical modes propagate at different fiber widths
- ▶ We will focus on HE11 and HE21 with linear and circular polarization

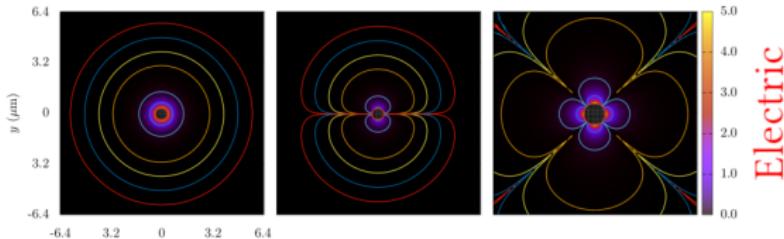
Optical nanofiber fields

HE11 Circ HE11 Lin HE21 Lin



Optical nanofiber fields

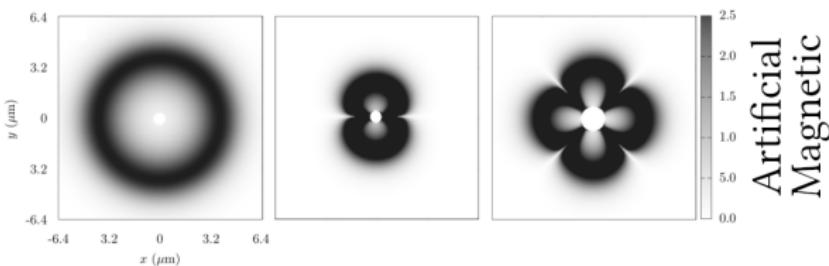
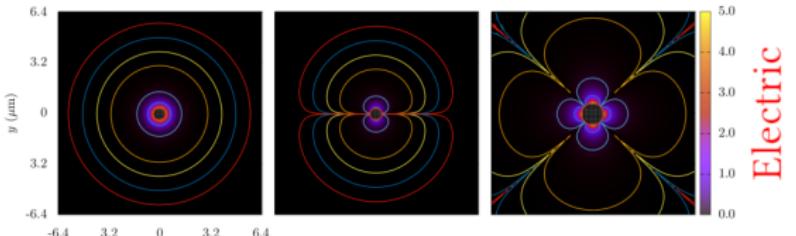
HE11 Circ HE11 Lin HE21 Lin



$$\mathbf{B} = \frac{\hbar\kappa_0 s^2 (n_1 + 1)}{(1 + \tilde{s}^2 |d_r E_r + d_\phi E_\phi + d_z E_z|^2)^2} \times \left[\hat{\phi} \frac{\partial}{\partial r} |d_r E_r + d_\phi E_\phi + d_z E_z|^2 - \hat{r} \frac{1}{r} \frac{\partial}{\partial \phi} |d_r E_r + d_\phi E_\phi + d_z E_z|^2 \right]$$

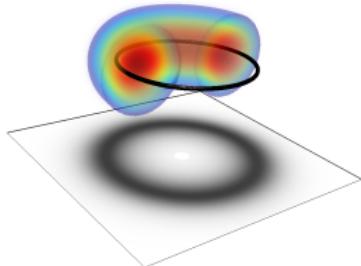
Optical nanofiber fields

HE11 Circ HE11 Lin HE21 Lin

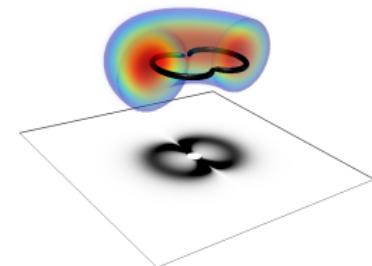


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HE11 modes:

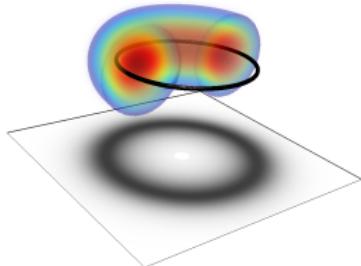


Circular

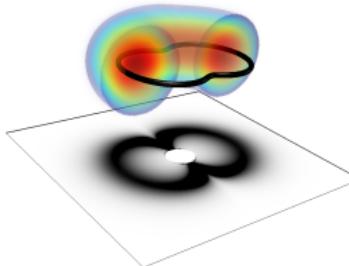


Linear

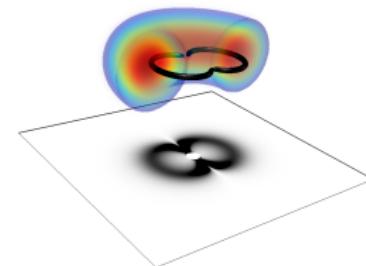
HE11 modes:



Circular



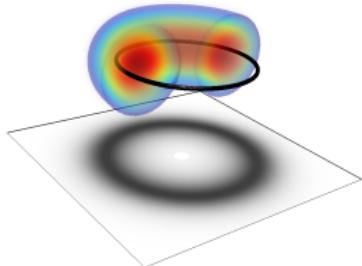
Elliptic



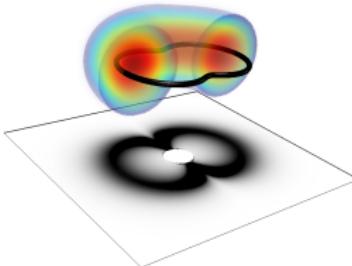
Linear

- ▶ Transition with linear polarization

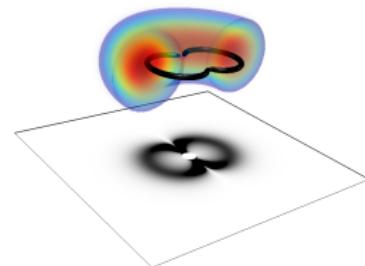
HE11 modes:



Circular

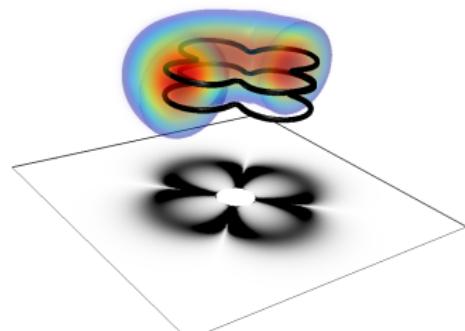


Elliptic



Linear

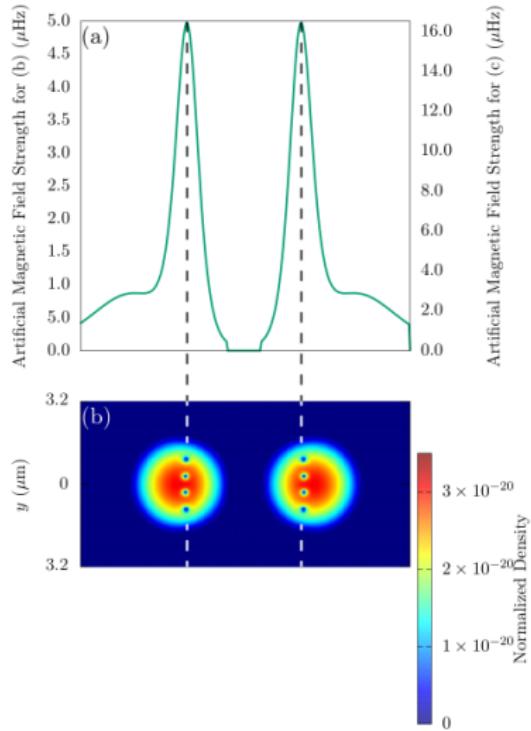
- ▶ Transition with linear polarization
- ▶ Highly controllable vortex structures that align to **B**



HE21 Linear

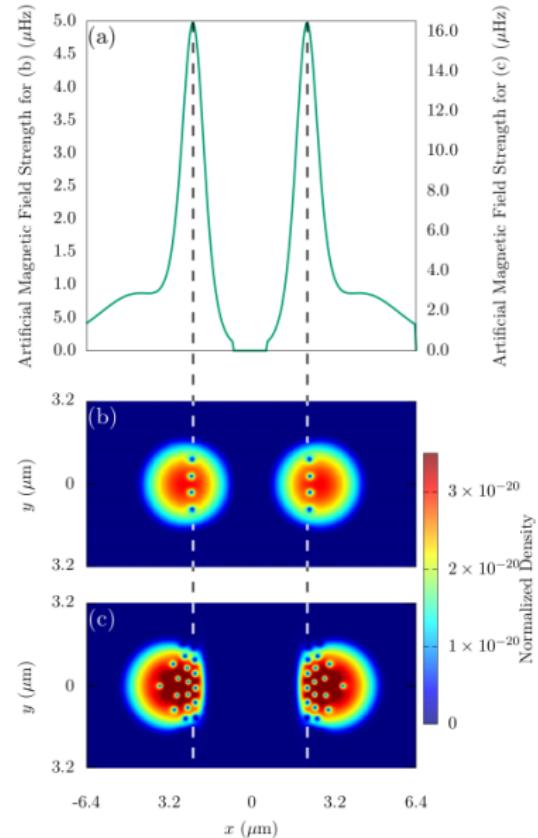
Vortex ring lattices

- ▶ With low magnetic field,
vortices follow magnetic field



Vortex ring lattices

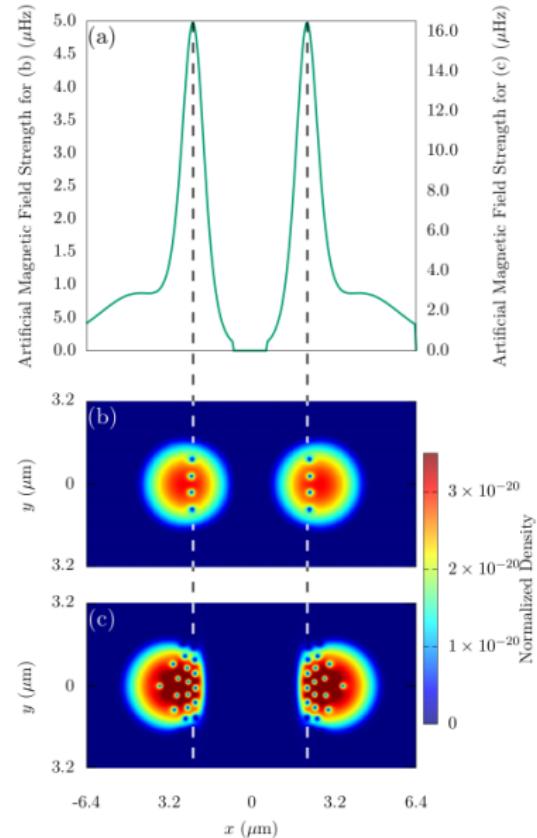
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Vortex ring lattices

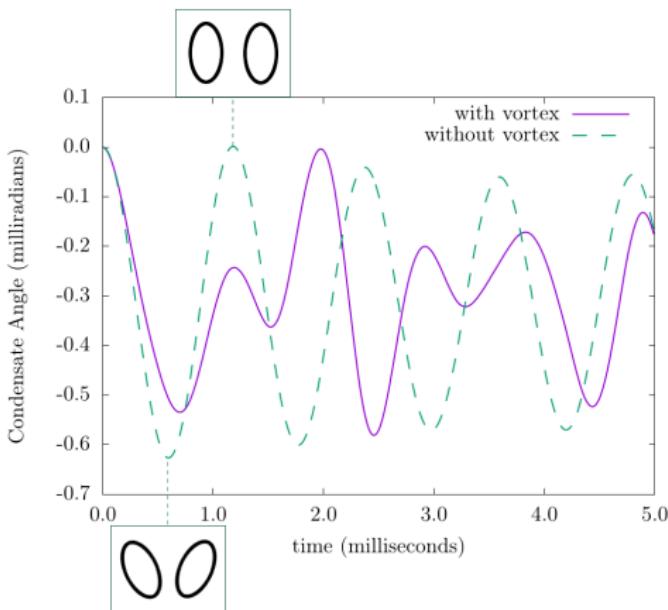
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Vortex detection?



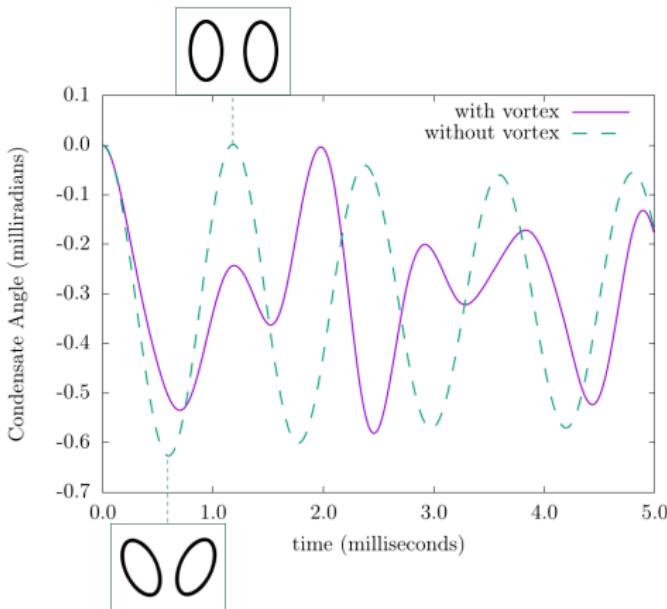
Scissors modes

- ▶ Vortex detection in 3D is difficult
- ▶ Scissors modes provide different oscillations with vortices



Scissors modes

- ▶ Vortex detection in 3D is difficult
- ▶ Scissors modes provide different oscillations with vortices
- ▶ Relies on elliptic-toroidal geometry



Physics

- ▶ This system can generate, control, and detect vortex structures in a BEC around a nanofiber
- ▶ Dynamic simulations are underway!

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Computer Science

- ▶ Dynamic simulations require a large amount of fileIO
- ▶ Scaling to a larger grid requires multiple GPUs

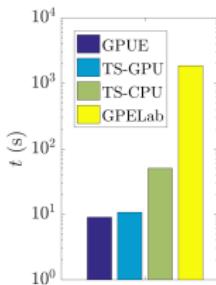
Overall conclusions

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I developed GPUE and GPU methods for

- ▶ Expression trees
- ▶ Vortex tracking

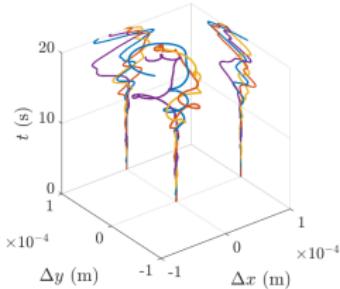
$$\text{GP} \left(\hat{U} \right)_E$$



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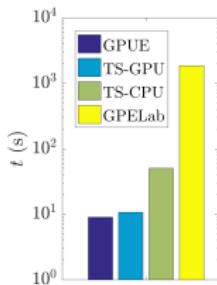


$$GP\left(\hat{U}\right)_E$$

I have also shown...

Chaos can be found in few-vortex systems

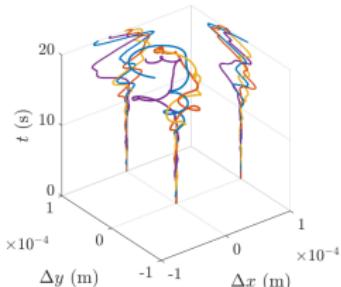
- ▶ Chaos is accelerated by collisions
- ▶ Highly controllable initial conditions



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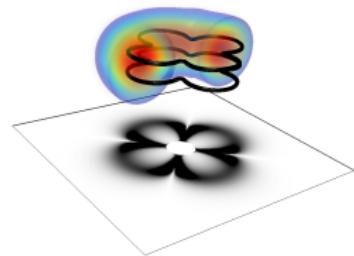
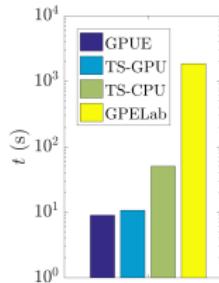
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3D vortex structures can be controlled, generated, and detected

- ▶ Vortex ring lattice
- ▶ Scissors mode



Physics: understanding superfluid vortices

3D vortex dynamics

Vortex knots

Vortex turbulence

Multicomponent simulations

Computer Science: SSFM on GPUs

Distributed Transpose

GPU-E.jl

Expression tree extensions

Compression methods

People



Organizations:



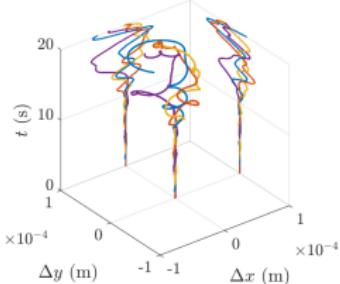
algorithm-archive.org

Overall conclusions

GPU computing is fast

- ▶ Distributed transpose
- ▶ Expression trees

$$GP\left(\hat{U}\right)_E$$

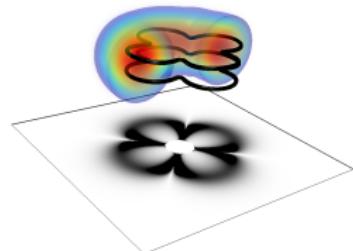
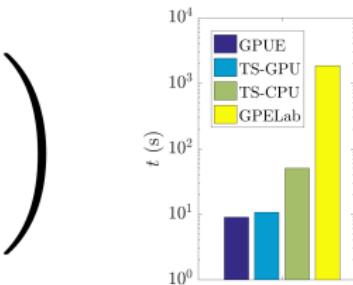


We can create chaotic systems

- ▶ Chaos is accelerated by collisions
- ▶ Highly controllable initial conditions

We can control vortex structures

- ▶ New vortex structures
- ▶ Vortex detection



Quantum state engineering

Quantum optimal control



Overall goal: optimize cost function by twiddling control parameters

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- ▶ For quantum systems, cost function is often the fidelity:

$$\mathcal{F} = |\langle \psi | \phi \rangle|^2$$

which required re-simulation every time a control parameter is changed

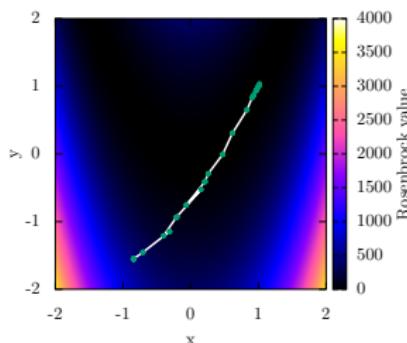
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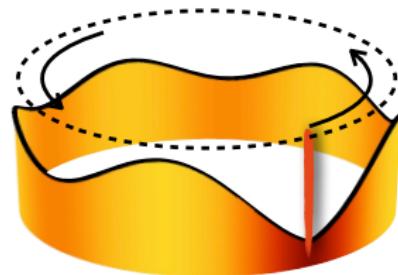
Nelder–Mead



Example Tonks–Girardeau gas system

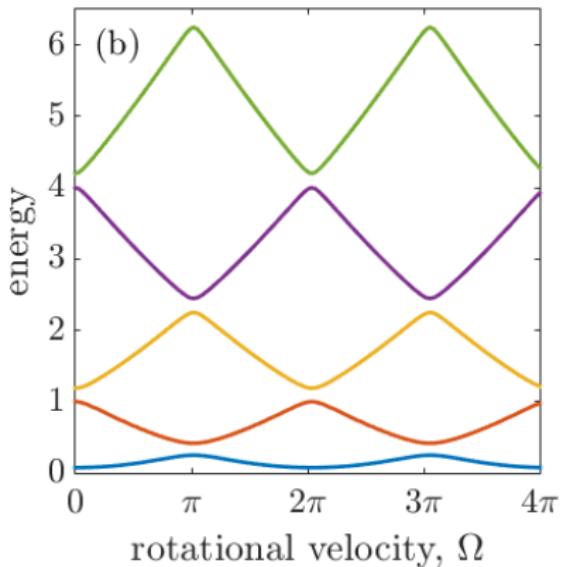
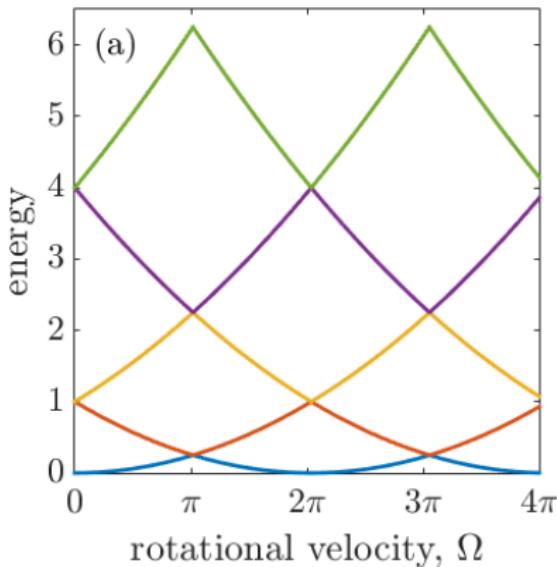
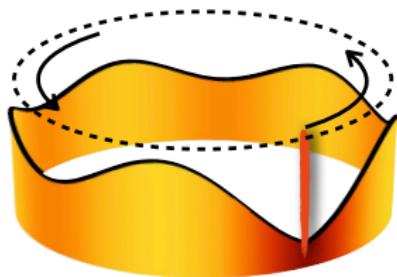


- ▶ NOON state: $|N, 0\rangle + |0, N\rangle$
- ▶ Tonks–Girardeau Gas:
 $g \rightarrow \infty$



Example Tonks–Girardeau gas system

- ▶ NOON state: $|N, 0\rangle + |0, N\rangle$
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An example protocol is the Chopped RAndom Basis (CRAB) optimal control method where...

- ▶ A control parameter is modified with

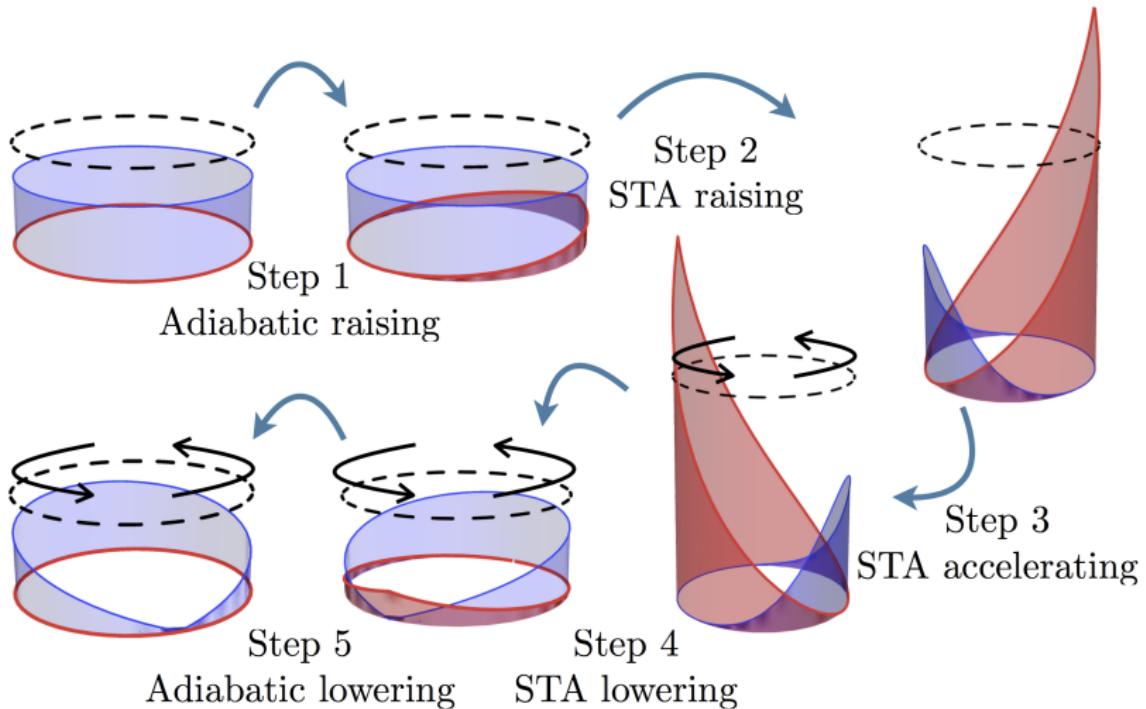
$$\Gamma^{\text{CRAB}}(t) = \Gamma^0(t)\gamma(t)$$

where

$$\gamma(t) = 1 + \frac{1}{\lambda(t)} \sum_{j=1}^J (A_j \sin(\nu_j t) + B_j \cos(\nu_j t))$$

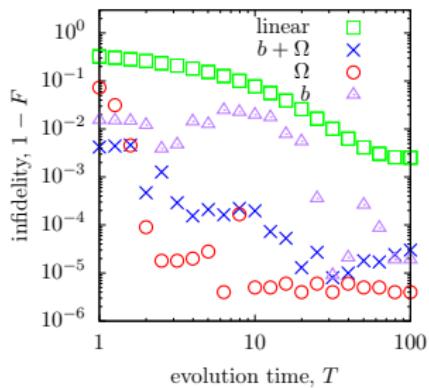
- ▶ Works if $\lim_{t \rightarrow 0} \lambda(t) = \lim_{t \rightarrow T} \lambda(t) = \infty$
- ▶ Creates a $3J$ -dimensional space to optimize (A, B, ν)

STA protocol

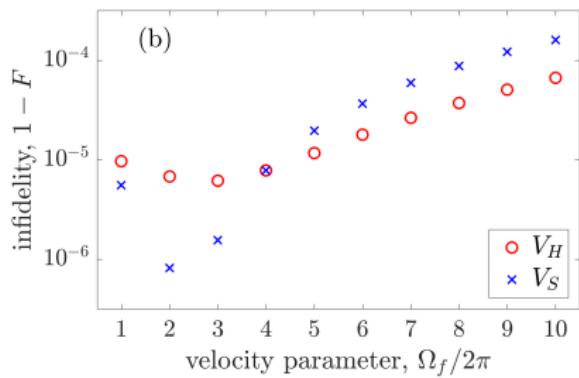


NOON optimization

Optimal control



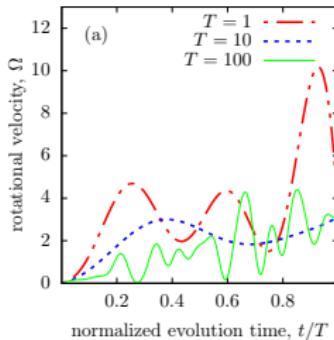
STA



Please ask questions at the end!

Fidelities with optimal control

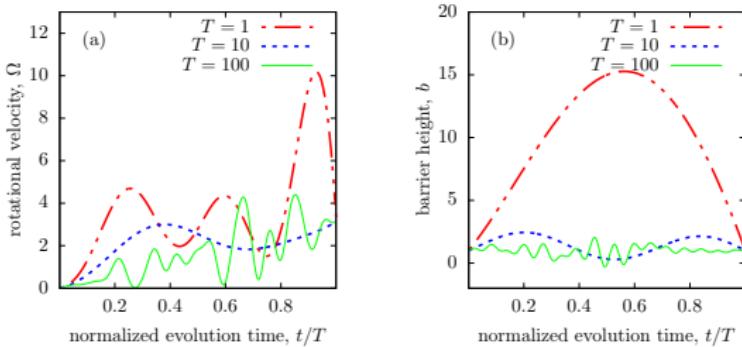
► Rotation $(\Omega(t))$



Fidelities with optimal control



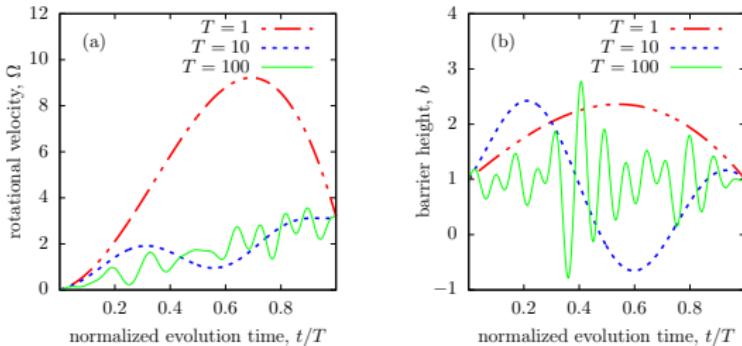
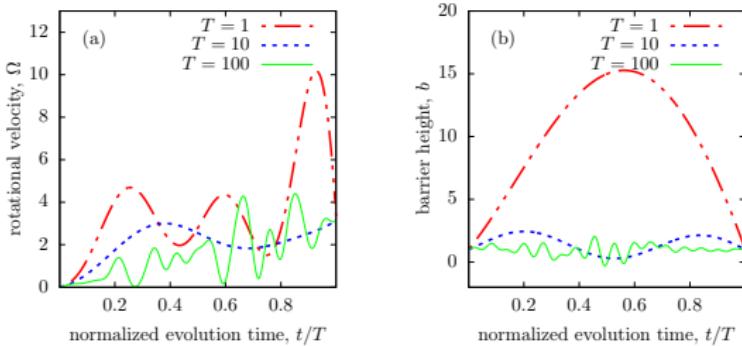
- ▶ Rotation ($\Omega(t)$)
- ▶ Barrier height ($b(t)$)



Fidelities with optimal control



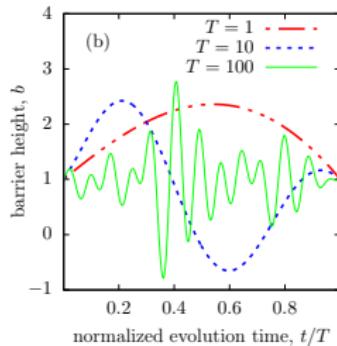
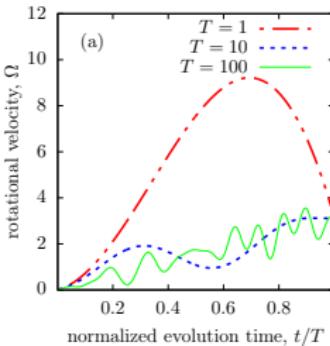
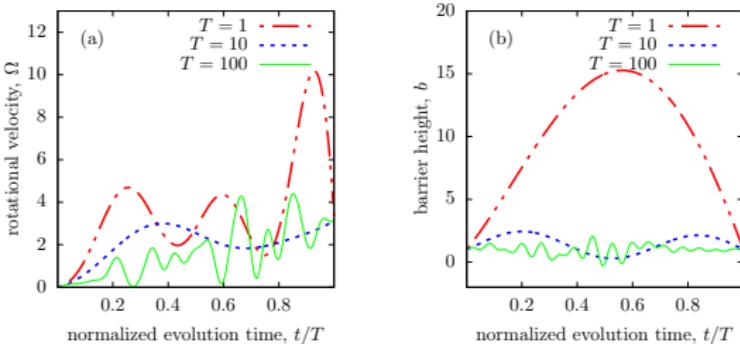
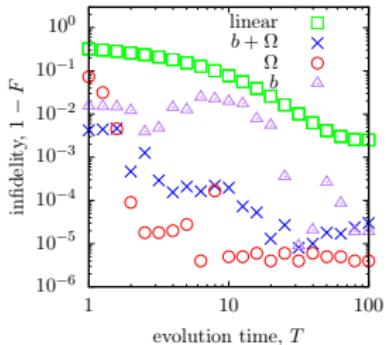
- ▶ Rotation ($\Omega(t)$)
- ▶ Barrier height ($b(t)$)
- ▶ Both



Fidelities with optimal control



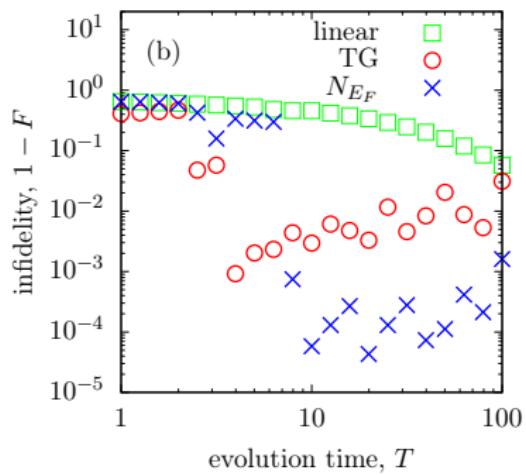
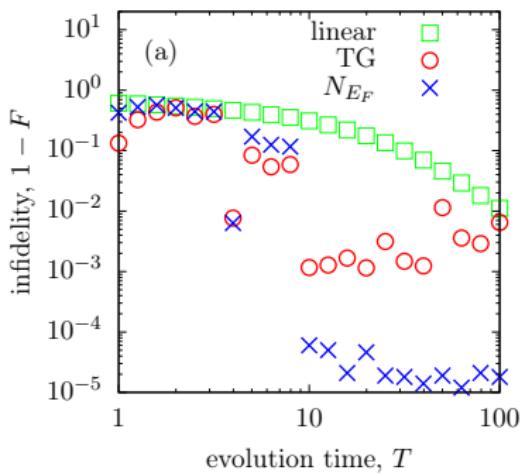
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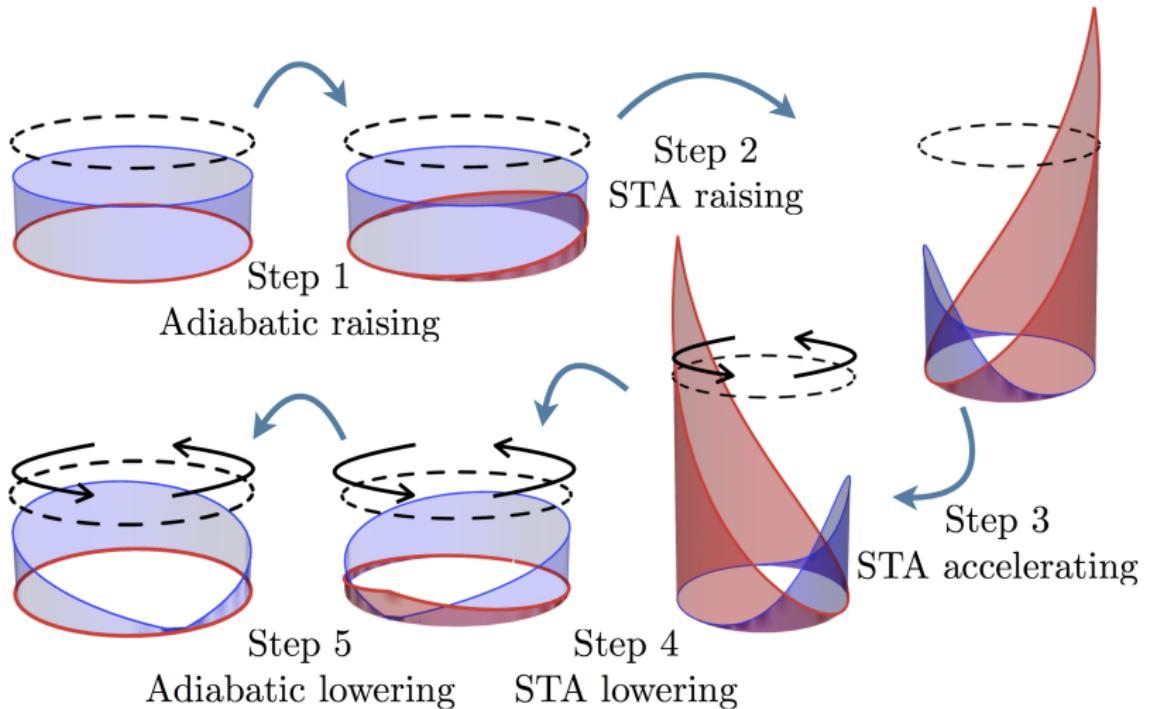
NOON Optimization



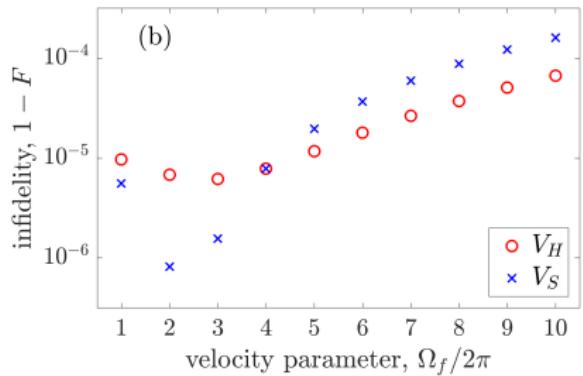
Optimizations of NOON state generation with 3 and 5 particles



STA protocol



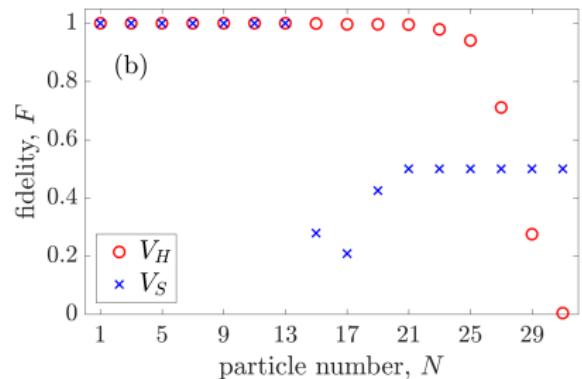
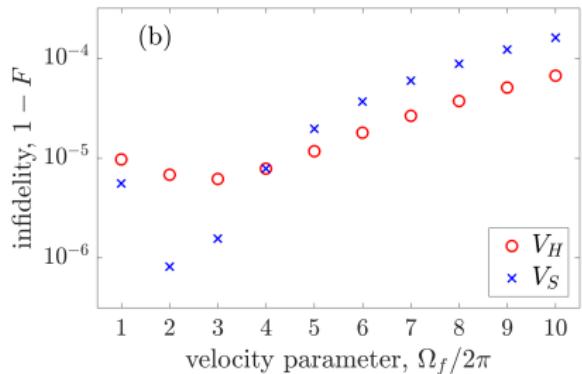
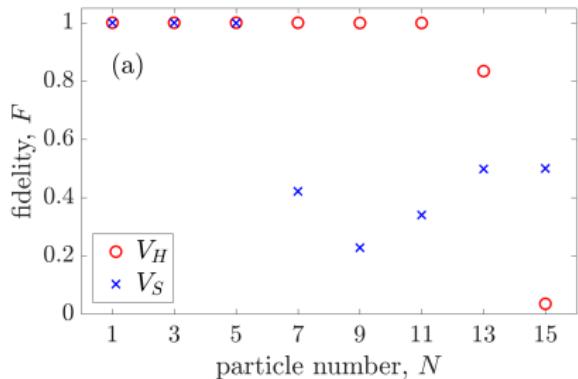
► Fidelities with rotation



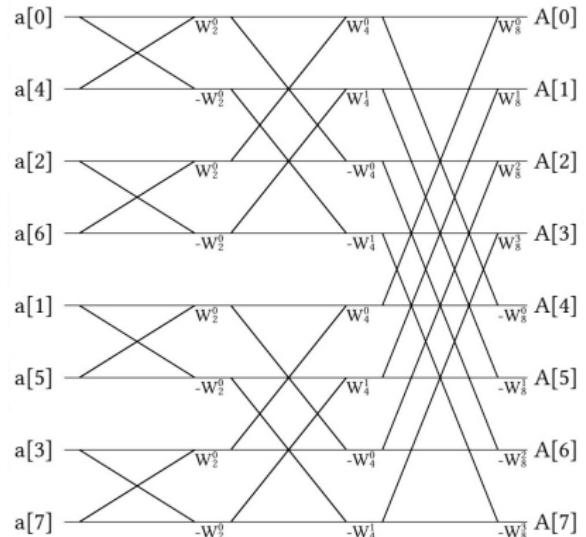
STA fidelities



- ▶ Fidelities with rotation
- ▶ Fidelities with rotation of 100, 200 and higher particle number



- ▶ Recursively subdivides DFT into simple sums with twiddle factors
- ▶ Many known libraries, like FFTW, and CuFFT
- ▶ Hard to parallelize (note for later)



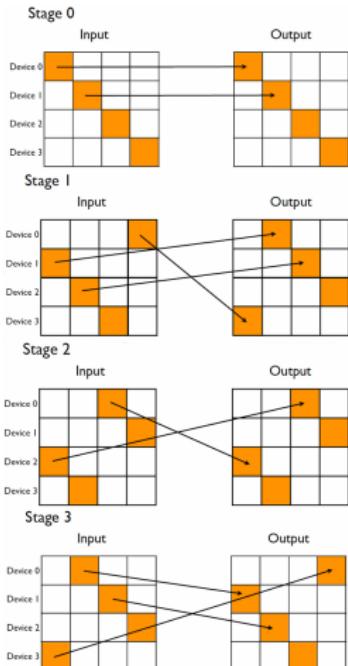
GPU transpose

- ▶ 2D out-of-place transpose \approx copy
(shared memory, coalescence, bank conflicts)
- ▶ In-place transposes are inefficient
- ▶ 3D permutations of arrays have 70% efficiency

Distributed transpose

GPU transpose

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(shared memory, coalescence, bank conflicts)
- ▶ In-place transposes are inefficient
- ▶ 3D permutations of arrays have 70% efficiency
- ▶ 2D distributed example



Ruetsch, G. and Fatica, M. 2013

Strang splitting



$$\Psi(\mathbf{r}, t + dt) = \left[e^{-\frac{i\hat{\mathcal{H}}_V dt}{2\hbar}} e^{-\frac{i\hat{\mathcal{H}}_P dt}{\hbar}} e^{-\frac{i\hat{\mathcal{H}}_V dt}{2\hbar}} \right] \Psi(\mathbf{r}, t) + \mathcal{O}(dt^3)$$

Taylor expansion of exponential:

$$e^{h(\mathbf{A}+\mathbf{B})} \approx \mathbf{I} + h(\mathbf{A} + \mathbf{B}) + \frac{1}{2}h^2(\mathbf{A} + \mathbf{B})^2$$

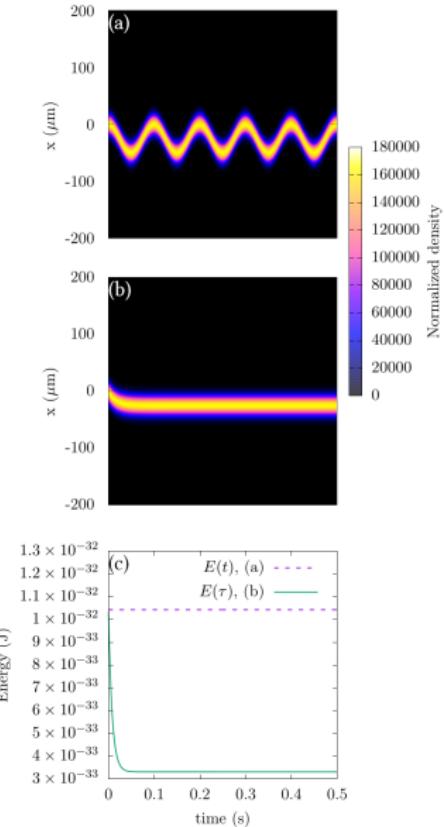
Last term:

$$\frac{1}{2}h^2(\mathbf{A} + \mathbf{B})^2 = \frac{1}{2} (\mathbf{A}^2 + \mathbf{AB} + \mathbf{BA} + \mathbf{B}^2)$$

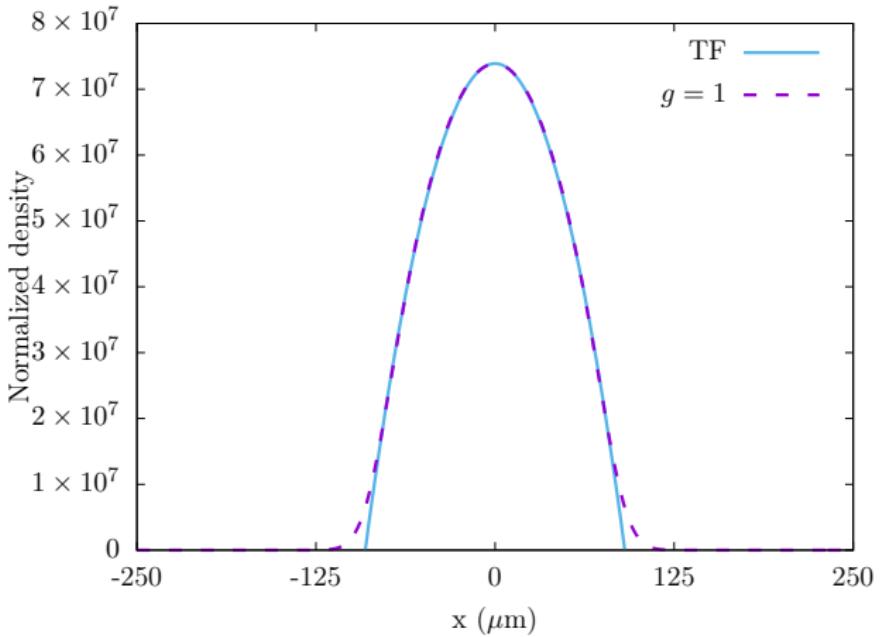
Symmetric splitting:

$$e^{h(\mathbf{A}+\mathbf{B})} \approx e^{h\mathbf{A}/2} e^{h\mathbf{B}} e^{h\mathbf{A}/2}$$

Evolution of SHO



Thomas–Fermi



$$R_{\text{TF}} = \sqrt{\frac{2\mu}{m\omega_i^2}}$$