

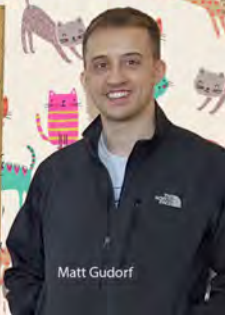
Revolution will not be Twitterized!

Spatiotemporal Cat

A Theory of Turbulence



Predrag Cvitanović



Matt Gudorf

**a spatiotemporal theory of
turbulence
computational challenges**

Predrag Cvitanović and Matt Gudorf

working notes
Georgia Tech

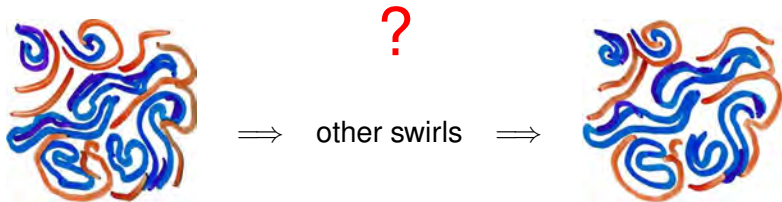
December 6, 2021

overview

- 1 what this talk is about
- 2 turbulence in large domains
- 3 space is time
- 4 bye bye, dynamics

how do clouds solve PDEs?

do clouds **integrate** Navier-Stokes equations?



are clouds Navier-Stokes supercomputers in the sky?

part 1

- 1 turbulence in large domains
- 2 space is time
- 3 spacetime
- 4 bye bye, dynamics

goal : enumerate the building blocks of turbulence

Navier-Stokes equations

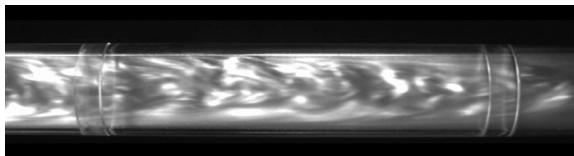
$$\frac{\partial \mathbf{v}}{\partial t} + (\mathbf{v} \cdot \nabla) \mathbf{v} = \frac{1}{R} \nabla^2 \mathbf{v} - \nabla p + \mathbf{f}, \quad \nabla \cdot \mathbf{v} = 0,$$

velocity field $\mathbf{v} \in \mathbb{R}^3$; pressure field p ; driving force \mathbf{f}

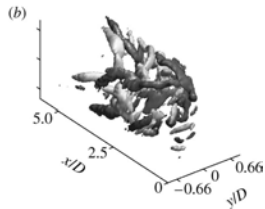
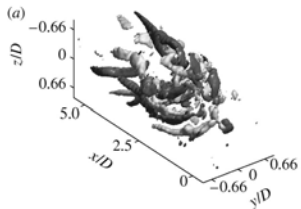
describe turbulence

starting from the equations (no statistical assumptions)

challenge : experiments are amazing

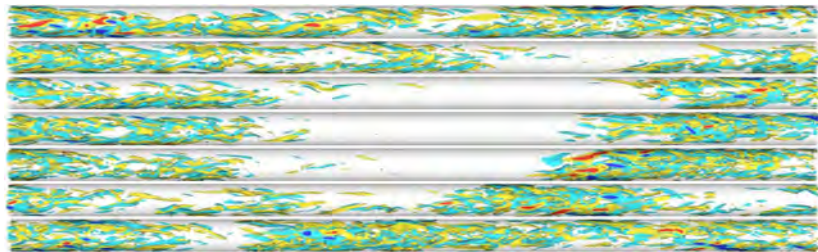


T. Mullin lab



B. Hof lab

can simulate **large** computational domains



pipe flow close to onset of turbulence ¹

but we have **hit a wall** :

exact coherent structures are too unstable to compute

¹M. Avila and B. Hof, Phys. Rev. E **87** (2013)

goal : we can do 3D turbulence, but for this presentation

Navier-Stokes equations

$$\frac{\partial \mathbf{v}}{\partial t} + (\mathbf{v} \cdot \nabla) \mathbf{v} = \frac{1}{R} \nabla^2 \mathbf{v} + (\dots)$$

velocity field $\mathbf{v}(\mathbf{x}; t) \in \mathbb{R}^3$

not helpful for developing intuition

we cannot visualize 3D velocity field at every 3D spatial point

look instead at 1D 'flame fronts'

(1+1) spacetime dimensional “Navier-Stokes”

Navier-Stokes equations

(1822)

$$\frac{\partial \mathbf{v}}{\partial t} + (\mathbf{v} \cdot \nabla) \mathbf{v} = \frac{1}{R} \nabla^2 \mathbf{v} + (\dots)$$



Kuramoto-Sivashinsky (1+1)-dimensional PDE

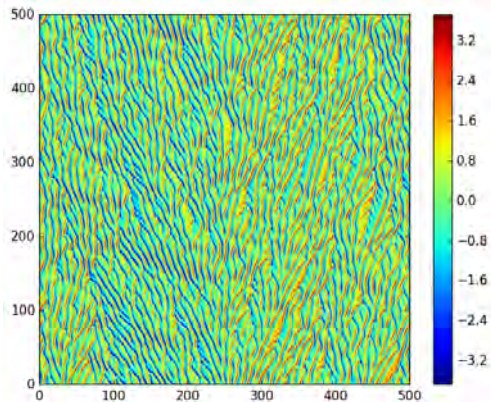
(1975)

$$u_t + u \nabla u = -\nabla^2 u - \nabla^4 u, \quad x \in \mathbb{R},$$

describes spatially extended systems such as

- flame fronts in combustion
- reaction-diffusion systems
- ...

an example : Kuramoto-Sivashinsky on a large domain

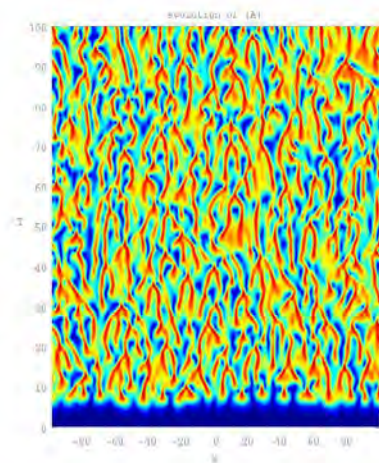


[horizontal] space $x \in [0, L]$ [up] time evolution

- turbulent behavior
- simpler physical, mathematical and computational setting than Navier-Stokes

another example of large spacetime domain simulation

complex Ginzburg-Landau

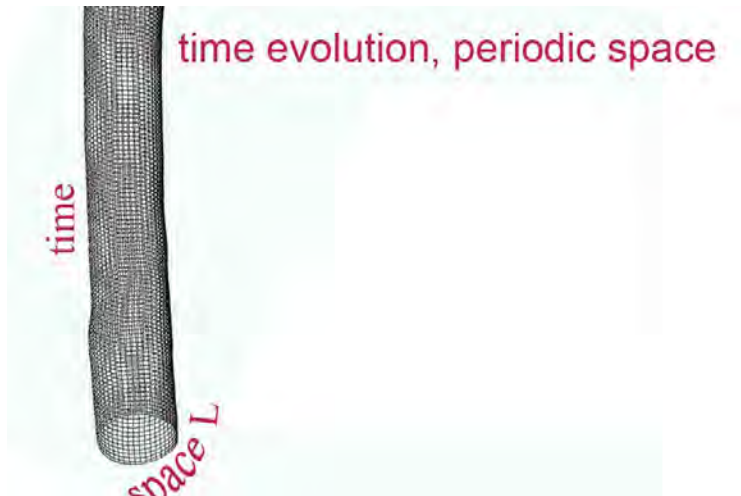


(will return to this)

[horizontal] space $x \in [-L/2, L/2]$

[up] time evolution

compact space, infinite time cylinder



so far : Navier-Stokes on compact spatial domains, all times

compact space, infinite time

Kuramoto-Sivashinsky equation

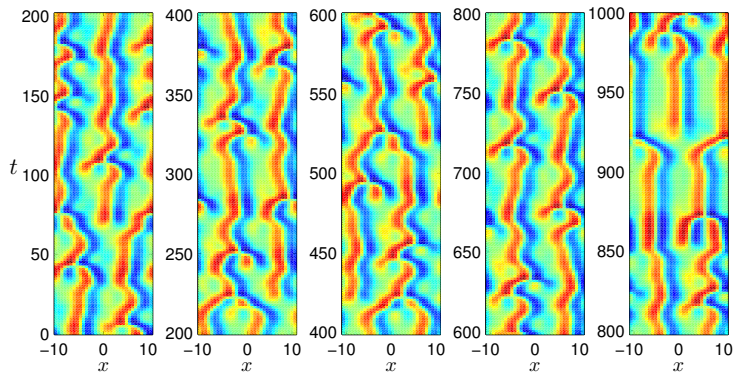
$$u_t = -(\nabla^2 + \nabla^4)u - u\nabla u, \quad x \in [-L/2, L/2],$$

in terms of discrete spatial Fourier modes

N ordinary differential equations (ODEs) in time

$$\dot{\tilde{u}}_k(t) = (q_k^2 - q_k^4) \tilde{u}_k(t) - i \frac{q_k}{2} \sum_{k'=0}^{N-1} \tilde{u}_{k'}(t) \tilde{u}_{k-k'}(t).$$

evolution of Kuramoto-Sivashinsky on small $L = 22$ cell



horizontal: $x \in [-11, 11]$

vertical: time

color: magnitude of $u(x, t)$

part 2

- 1 turbulence in large domains
- 2 **space is time**
- 3 spacetime
- 4 bye bye, dynamics

yes, but

is space time?

compact time, infinite space

rewrite Kuramoto-Sivashinsky

$$u_t = -uu_x - u_{xx} - u_{xxxx}$$

as 4-fields vector

$$\mathbf{u}^\top = (u, u', u'', u''')$$

where $u' \equiv u_x$, $u'' \equiv u_{xx}$, $u''' \equiv u_{xxx}$

equation $\frac{d}{dx}\mathbf{u}(x) = \mathbf{v}(x)$ now 1st order in spatial derivative

Kuramoto-Sivashinsky = four coupled 1st order PDEs

$$\begin{aligned} \frac{du}{dx} &= u', & \frac{du'}{dx} &= u'' \\ \frac{du''}{dx} &= u''', & \frac{du'''}{dx} &= -u_t - u'' - uu' \end{aligned}$$

compact time, infinite space

1st order in **spatial** derivative

evolve four 1st order PDEs for $\mathbf{u}(x)$ in x ,



$$\frac{d}{dx}\mathbf{u}(x) = \mathbf{v}(x)$$

- compact in time, periodic boundary condition

$$u(x, t) = u(x, t + T)$$

- initial data

$$\mathbf{u}_0^\top = (u(x_0, t), u'(x_0, t), u''(x_0, t), u'''(x_0, t))$$

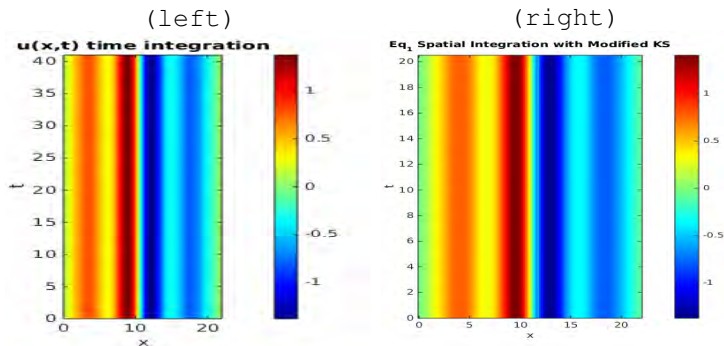
specified for all $t \in [0, T)$, at a fixed space point x_0

can do : compact time, infinite space cylinder

space evolution, periodic time

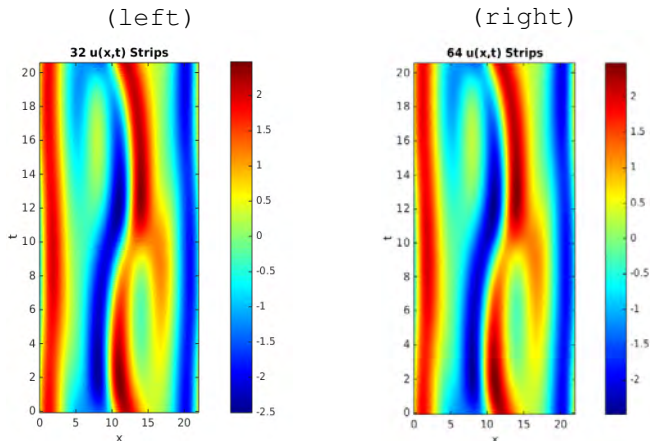


a time-invariant equilibrium, spatial periodic orbit



evolution of EQ_1 : (left) in time, (right) in space
initial condition for the spatial integration is the time strip
 $u(x_0, t)$, $t = [0, T)$, where time period $T = 0$, spatial x period is
 $L = 22$.

a spacetime invariant 2-torus integrated in either time or space



(left) old : time evolution. (right) new : space evolution
 $x = [0, L]$ initial condition : time periodic line $t = [0, T]$

but integrations are uncontrollably unstable

neither time nor space integration works
for large domains

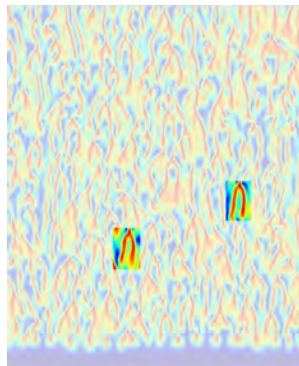
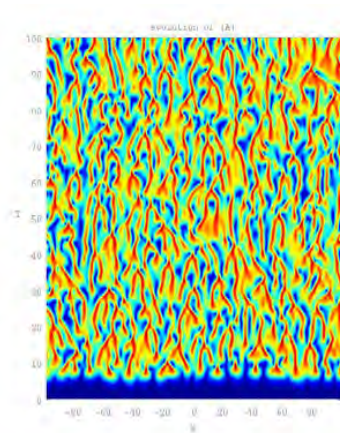
rethink the calculation

part 3

- 1 turbulence in large domains
- 2 space is time
- 3 **spacetime**
- 4 bye bye, dynamics

complex Ginzburg-Landau on a large spacetime domain

goal : enumerate nearly recurrent chronotopes

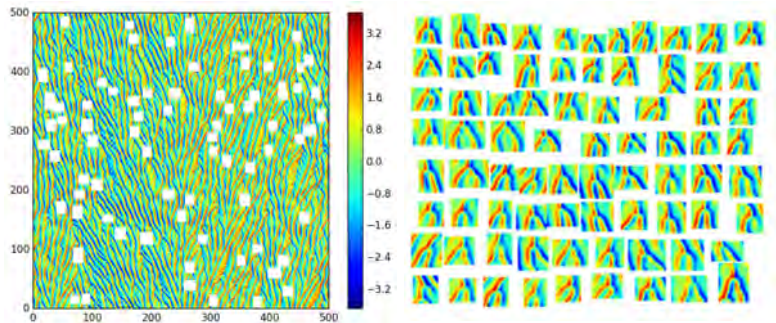


[left-right] space $x \in [-L/2, L/2]$

[up] time $t \in [0, T]$

Kuramoto-Sivashinsky on a large spacetime domain

the same small tile recurs often in a turbulent pattern



goal : define, enumerate nearly recurrent tiles

In literary theory and philosophy of language, the chronotope is how configurations of time and space are represented in language and discourse.

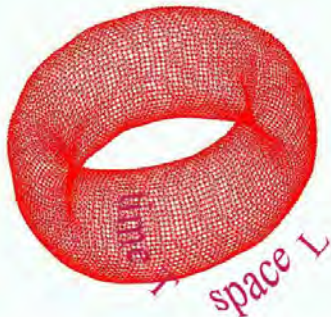
— [Wikipedia : Chronotope](#)

- Mikhail Mikhailovich Bakhtin (1937)

²S. Lepri et al., J. Stat. Phys. **82**, 1429–1452 (1996).

use spatiotemporally compact solutions as chronotopes

periodic spacetime : 2-torus



this 'exact coherent structure'

shadows a small patch of spacetime solution $u(x, t)$

periodic orbits generalize to d -tori

1 time, 0 space dimensions

a state space point is *periodic* if its orbit returns to it after a finite time T ;

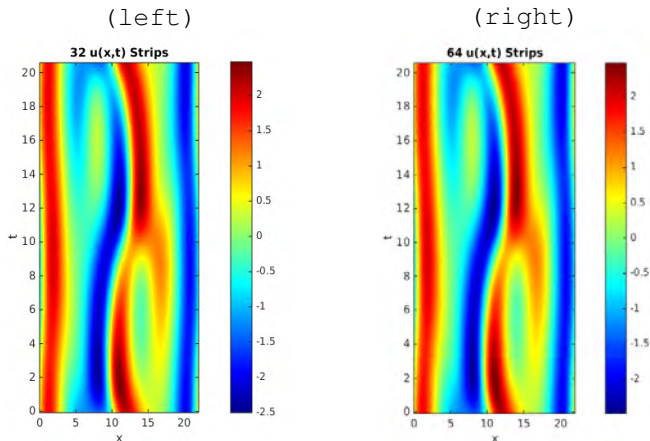
such orbit tiles the time axis by infinitely many repeats

1 time, $d-1$ space dimensions

a state space point is *spatiotemporally periodic* if it belongs to an invariant d -torus \mathcal{R} ;

such torus tiles the spacetime by infinitely many repeats

a spacetime invariant 2-torus integrated in either time or space



(left) old : time evolution $t = [0, T]$

initial condition : space periodic line $x = [0, L]$

(right) new : space evolution $x = [0, L]$

initial condition : time periodic line $t = [0, T]$

every compact solution is a fixed point on a discrete lattice

discretize $u_{nm} = u(x_n, t_m)$ over NM points of spatiotemporal periodic lattice $x_n = nL/N$, $t_m = mT/M$, Fourier transform :

$$\tilde{u}_{k\ell} = \frac{1}{NM} \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} u_{nm} e^{-i(q_k x_n + \omega_\ell t_m)}, \quad q_k = \frac{2\pi k}{L}, \quad \omega_\ell = \frac{2\pi \ell}{T}$$

Kuramoto-Sivashinsky is no more a PDE,
but an algebraic $[N \times M]$ -dimensional problem
of determining global solution \mathbf{u} to

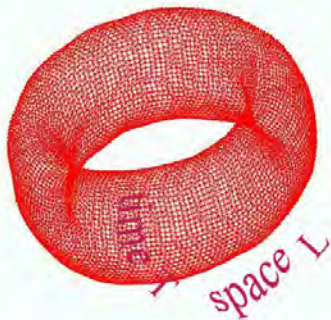
fixed point condition

$$\left(-i\omega_\ell - (q_k^2 - q_k^4)\right) \tilde{u}_{k\ell} + i\frac{q_k}{2} \sum_{k'=0}^{N-1} \sum_{m'=0}^{M-1} \tilde{u}_{k'm'} \tilde{u}_{k-k', m-m'} = 0$$

every calculation is a spatiotemporal lattice calculation

field is discretized as \tilde{u}_{kl} values
over NM points of a periodic lattice

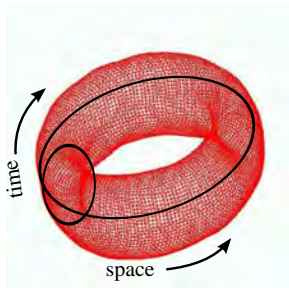
periodic spacetime : 2-torus



professor Zweistein forgets to take his meds

statement : **HA!**

You are imposing by hand the space & time periods L , T !



answer : **NO!**

nature chooses L & T , they are free parameters.

there is no more time evolution

solution to Kuramoto-Sivashinsky is now given as

condition that

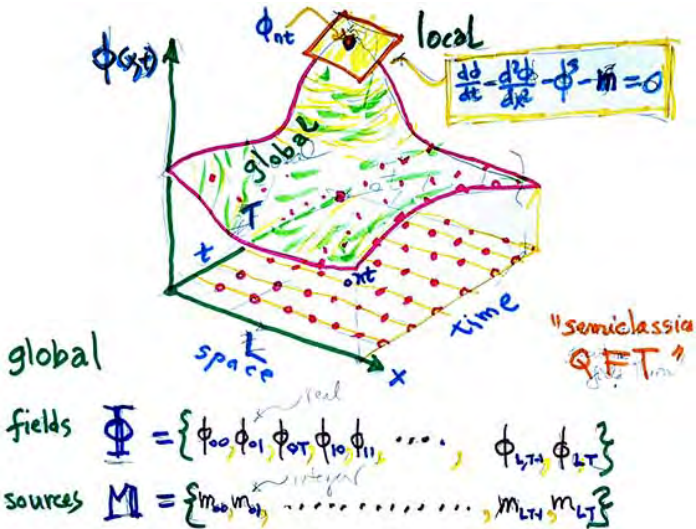
at each lattice point $k\ell$
the tangent field at $\tilde{u}_{k\ell}$

satisfies the equations of motion

$$\left[-i\omega_\ell - (q_k^2 - q_k^4) \right] \tilde{u}_{k\ell} + i\frac{q_k}{2} \sum_{k'=0}^{N-1} \sum_{m'=0}^{M-1} \tilde{u}_{k'm'} \tilde{u}_{k-k', m-m'} = 0$$

this is a **local** tangent field constraint on a **global** solution

think globally, act locally



for each symbol array M , a periodic lattice state X_M

unexpected gift from nature

robust : no exponential instabilities

as there are no finite time / space integrations

no need for $\sim 10^{-11}$ accuracies,

SO

accuracy to a few % suffices,

you only need to get the shape of a solution right

part 4

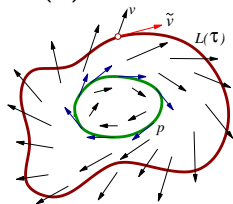
- 1 turbulence in large domains
- 2 space is time
- 3 spacetime
- 4 **spacetime computations**
- 5 bye bye, dynamics

how to find solutions ? an ODE example

the law of motion : $\dot{x} = v(x)$

guess loop tangent $\tilde{v}(\tilde{x}) \neq v(\tilde{x})$

periodic orbit $\tilde{v}(\tilde{x}), v(\tilde{x})$ aligned



cost function

$$F^2[\tilde{x}] = \oint_L ds (\tilde{v} - v)^2; \quad \tilde{v} = \tilde{v}(\tilde{x}(s, \tau)), \quad v = v(\tilde{x}(s, \tau)),$$

penalize³ misorientation of the loop tangent $\tilde{v}(\tilde{x})$ relative to the true dynamical flow tangent field $v(\tilde{x})$

³Y. Lan and P. Cvitanović, Phys. Rev. E 69, 016217 (2004).

how do clouds solve PDEs?

clouds do not **NOT** integrate Navier-Stokes equations



⇒ other swirls ⇒



do clouds satisfy Navier-Stokes equations?

yes!

they satisfy them **locally**, everywhere and at all times

the equations imposed as local constraints

Kuramoto-Sivashinsky equation

$$F(u) = u_t + u_{xx} + u_{xxxx} + uu_x = 0$$

for example, minimize over the entire 2-torus

cost function

$$G \equiv \frac{1}{2} \|F(u)\|_{L^2 \times T}^2$$

need your help !

adjoint descent

cost function

$$G = \frac{1}{2} \mathbf{F}^\top \mathbf{F}.$$

introduce fictitious time (τ) flow by differentiation of cost function.

$$\partial_\tau G = (\mathbf{J}^\top \mathbf{F})^\top (\partial_\tau \mathbf{x})$$

“adjoint descent” method defined by choosing⁴

$$\partial_\tau \mathbf{x} = -(\mathbf{J}^\top \mathbf{F})$$

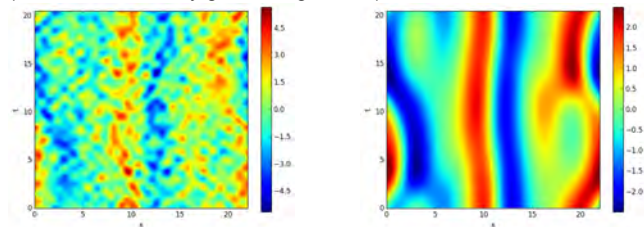
⁴M. Farazmand, J. Fluid M. 795, 278–312 (2016).

does it work at all ?

add strong noise to a *known* solution,
twice the typical amplitude

only the first test

(not how we actually generate guesses)



(left) initial guess: a known invariant 2-torus

$(L_0, T_0) = (22.0, 20.5057459345) + \text{strong random noise}$

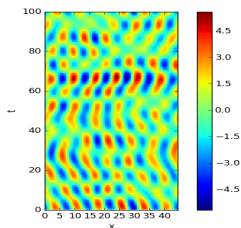
(right) the resulting adjoint descent converged invariant 2-torus

$(L_f, T_f) = (21.95034935834641, 20.47026321555662)$

initial guess generation ?

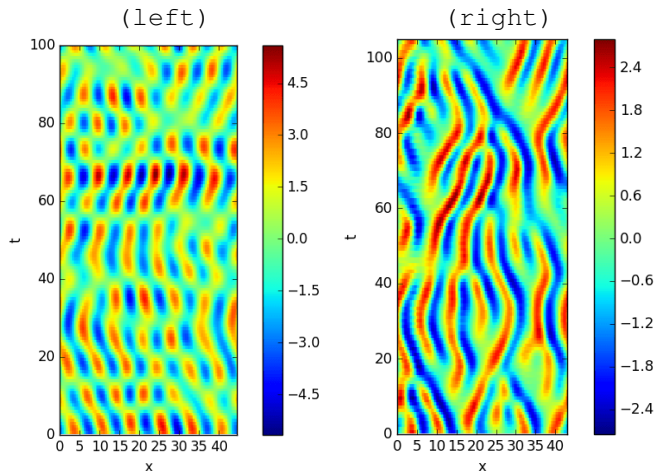
the time scale : the shortest ‘turnover’ scale characterized by the period of the shortest periodic orbit? Or perhaps the Lyapunov time?

the spatial scale : $\bar{L} = 2\pi\sqrt{2}$, the most unstable spatial wavelength of the Kuramoto-Sivashinsky



initial : spatial \bar{L} -modulated random guess

KS invariant 2-torus found variationally



(left) initial : $\bar{L} = 2\pi\sqrt{2}$ spatially modulated “noisy” guess

(right) adjoint descent : converged invariant 2-torus

initial guesses, embedded in ergodic sea?

Historically,

guesses extracted from close recurrences
observed in long turbulent simulations

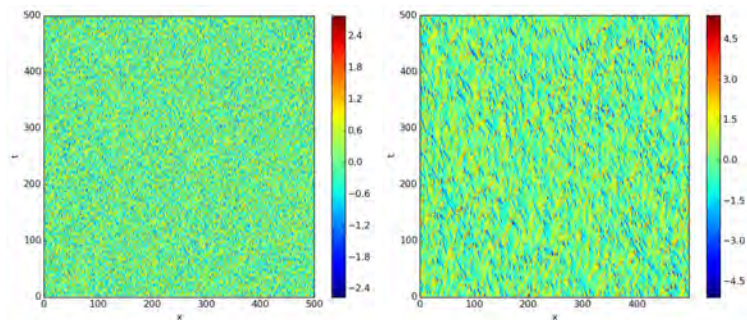
- 1 inefficient, finds only the shortest, least unstable orbits^{5,6}
- 2 can integrate only not far in time

need spatiotemporal guesses

⁵D. Auerbach et al., Phys. Rev. Lett. **58**, 2387–2389 (1987).

⁶J. F. Gibson et al., J. Fluid Mech. **611**, 107–130 (2008).

guesses extracted from large spacetime domains



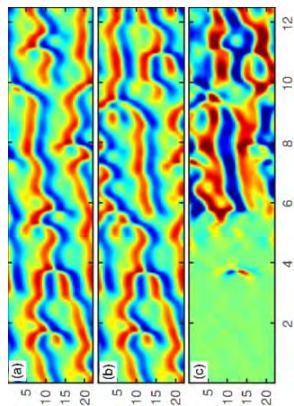
(left) random initial state on $(L, T) = (500, 500)$

(right) adjoint descent \rightarrow typical Kuramoto-Sivashinsky state

finite windows are our starting guesses for invariant 2-tori

another, much twittered : machine learning guesses

“reservoir computing” example⁷



- (a) data:
Kuramoto-Sivashinsky simulation
- (b) reservoir computing prediction
- (c) two subtracted agree to
 ~ 5 Lyapunov times

Q : how would you learn this data?

⁷J. Pathak et al., Phys. Rev. Lett. **120**, 024102 (2018).

embarrassment of riches

what to do?

Matthew N. Gudorf

has 1 000's of such invariant 2-tori

part 5

- 1 turbulence in large domains
- 2 space is time
- 3 spacetime
- 4 **fundamental tiles**
- 5 bye bye, dynamics

building blocks of turbulence

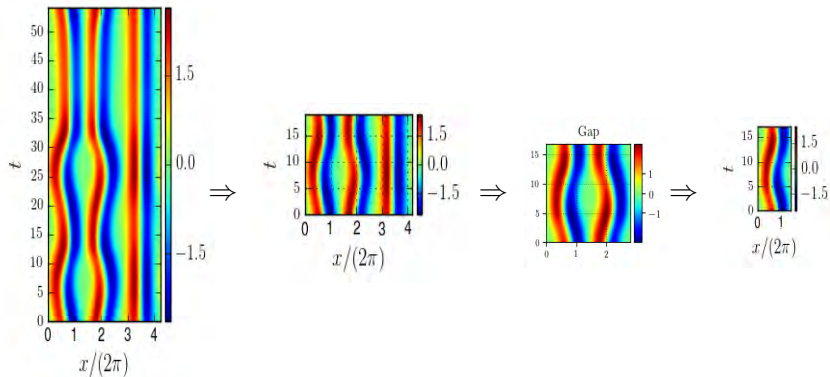
how do we **recognize** a cloud?



by recurrent shapes!

so, construct an **alphabet** of possible shapes

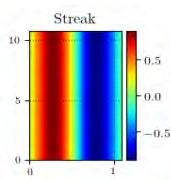
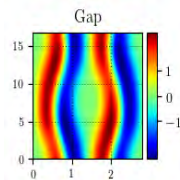
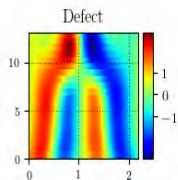
extracting a fundamental tile



- 1) invariant 2-torus
- 2) invariant 2-torus computed from initial guess cut out from 1)
- 3) "gap" invariant 2-torus, initially cut out from 2)
- 4) the "gap" prime invariant 2-torus fundamental domain

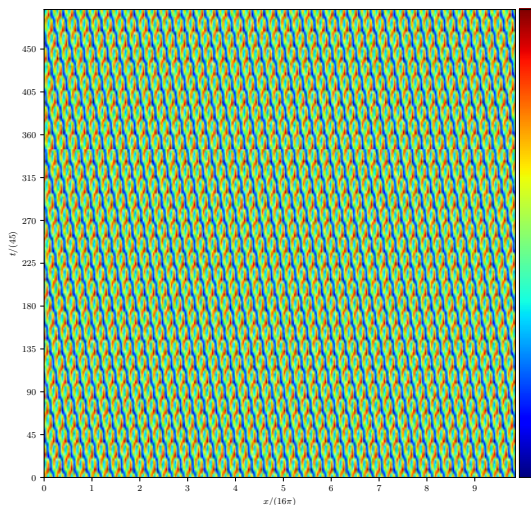
a trial set of prime (rubber) tiles

an alphabet of Kuramoto-Sivashinsky fundamental tiles



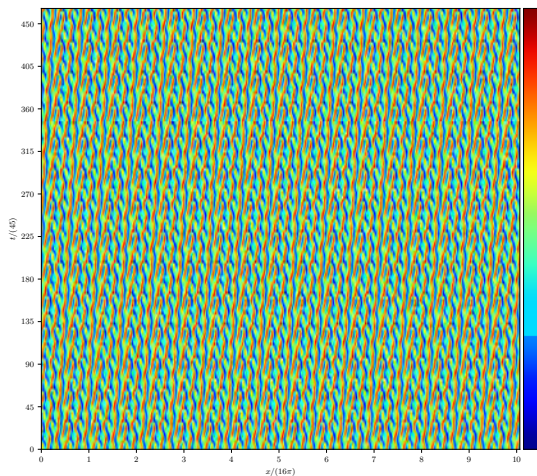
utilize also discrete symmetries :
spatial reflection, spatiotemporal shift-reflect, . . .

Kuramoto-Sivashinsky tiled by a small tile



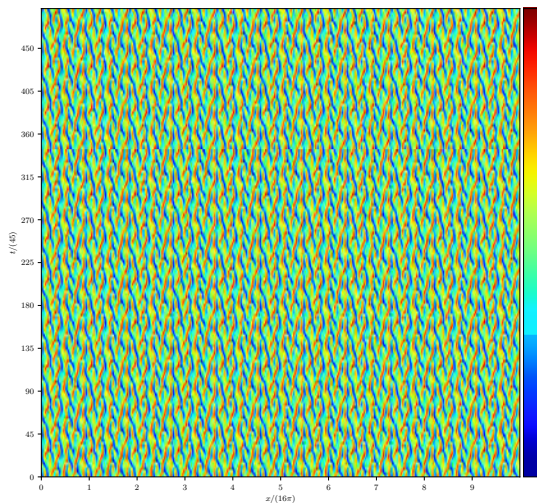
tiling by relative periodic invariant 2-torus
 $(L, T) = (13.02, 15)$

spacetime tiled by a larger tile



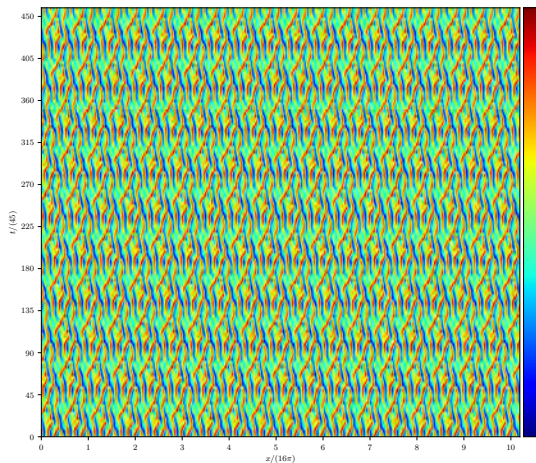
tiling by relative periodic invariant 2-torus
 $(L, T) = (33.73, 35)$

spacetime tiled by a tall tile



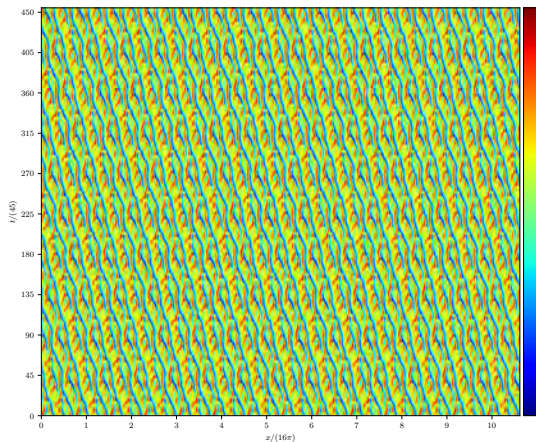
tiling by shift-reflect invariant 2-torus
 $(L, T) = (55.83, 24)$

spacetime tiled by a larger tile



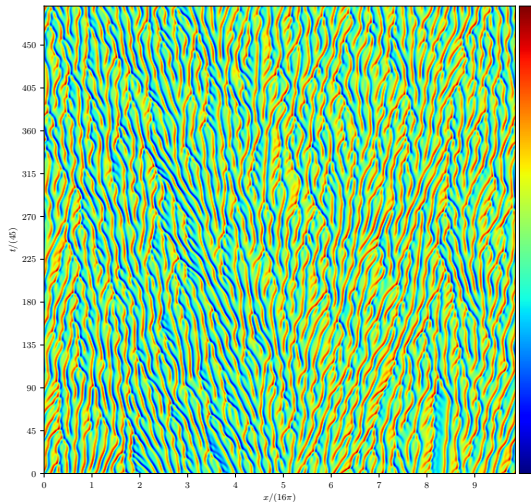
tiling by relative periodic invariant 2-torus
 $(L, T) = (32.02, 51)$

spacetime tiled by a larger tile



tiling by relative periodic invariant 2-torus
 $(L, T) = (44.48, 50)$

any particular tiling looks nothing like turbulent
Kuramoto-Sivashinsky!



[horizontal] space $x \in [-L/2, L/2]$

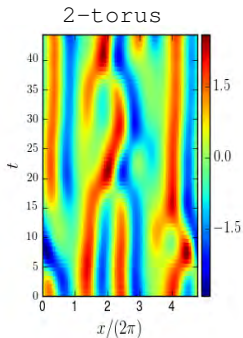
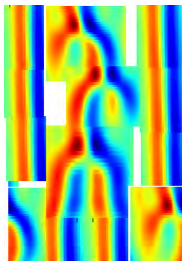
[up] time evolution

part 6

- 1 turbulence in large domains
- 2 space is time
- 3 spacetime
- 4 fundamental tiles
- 5 **gluing tiles**
- 6 bye bye, dynamics

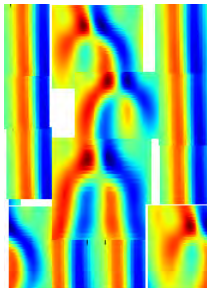
a qualitative tiling guess

a tiling and the resulting solution



turbulence.zip : each solution has a unique symbolic name

symbolic dynamics is 2-dimensional!



S	HalfD		S
	HoD*		
HoD	S	S	HoD

0	2	0	
0	1	0	
0	1*	0	
0		0	
1	0	0	1
	0	0	

- each symbol indicates a corresponding spatiotemporal tile
- these are “rubber” tiles

part 7

- 1 turbulence in large domains
- 2 space is time
- 3 **bye bye, dynamics**

in future there will be no future

goodbye

to long time and/or space integrators

they never worked and could never work

life outside of time

the trouble:

forward time-integration codes too unstable

multishooting inspiration: replace a guess that a point is on the periodic orbit by a guess of the entire orbit.

→

spatio-temporally periodic solutions of classical field theories can be found by variational methods

the equations solved as global optimization problems

impose the equations as local constraints

$$F(u) = u_t + u_{xx} + u_{xxxx} + uu_x = 0$$

minimize globally

perhaps using cost function

$$G \equiv \frac{1}{2} \|F(u)\|_{L \times T}^2$$

can computers

do this ?

the answer is

scalability

compute locally, adjust globally

Navier-Stokes codes

- T. M. Schneider : developing a matrix-free variational Navier-Stokes code, machine learning initial guesses
- D. Lasagna and A. Sharma : developing variational adjoint solvers to find periodic orbits with long periods
- Q. Wang : parallelizing **spatiotemporal** computation is FLOPs intensive, but more robust than integration forward in time

it's rocket science^{8,9,10}

⁸T. M. Schneider, *Variational adjoint methods coupled with machine learning*, private communication, 2019.

⁹D. Lasagna et al., *Periodic shadowing sensitivity analysis of chaotic systems*, 2018.

¹⁰Q. Wang et al., *Phys. Fluids* **25**, 110818 (2013).

towards scalable parallel-in-time turbulent flow simulations

future :

processor speed \rightarrow limit

number of cores $\rightarrow 10^6 \rightarrow \dots$

*Wang et al (2013)*¹¹ :

next-generation : spacetime parallel simulations,
on discretized 4D spacetime computational domains,
with each computing core handling a spacetime lattice cell

compared to time-evolution solvers: significantly higher level of
concurrency, reduction the ratio of inter-core communication to
floating point operations

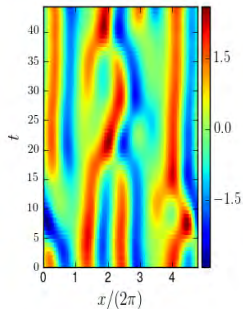
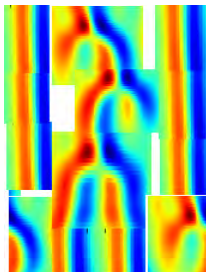
\Rightarrow a path towards exascale DNS of turbulent flows

¹¹Q. Wang et al., Phys. Fluids 25, 110818 (2013).

enumerate hierarchically spatiotemporal patterns

2D symbolic encoding \Rightarrow admissible solutions

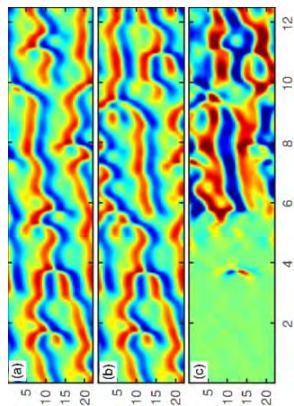
0	2		0
0	1		0
0	1*		0
0			0
1	0	0	1
	0	0	



- each symbol indicates a minimal spatiotemporal tile
- glue them in all admissible ways

machine learning will be needed

“reservoir computing” example¹²



- (a) data:
Kuramoto-Sivashinsky simulation
- (b) reservoir computing prediction
- (c) two subtracted agree to
 ~ 5 Lyapunov times

Q : how would you learn this data?

¹²J. Pathak et al., Phys. Rev. Lett. **120**, 024102 (2018).

take home : clouds do not integrate PDEs

do clouds integrate Navier-Stokes equations?



at any spacetime point Navier-Stokes equations describe the local tangent space

they satisfy them **locally**, everywhere and at all times

summary

- 1 study turbulence in infinite spatiotemporal domains
- 2 theory : classify all spatiotemporal tilings
- 3 numerics : future is spatiotemporal

there is no more time

there is only enumeration of spacetime solutions

spatiotemporally infinite spatiotemporal cat



part 8

- 1 turbulence in large domains
- 2 space is time
- 3 spacetime
- 4 fundamental tiles
- 5 gluing tiles
- 6 bye bye, dynamics
- 7 **theory of turbulence ?**

are d -tori

a theory of turbulence ?

part 9

- 1 (semi-)classical field theories
- 2 state space
- 3 symbolic dynamics

Dreams of Grand Schemes : solve

Navier-Stokes

$$\rho \frac{\partial u_i}{\partial t} + \rho u_j \frac{\partial u_i}{\partial x_j} = \rho X_i - \frac{\partial p}{\partial x_i} + \mu \nabla^2 u_i$$

Einstein

$$R_{ik} - \frac{1}{2} g_{ik} R = \frac{8\pi k}{c^4} T_{ik}$$

$$R^i{}_{klm} = \frac{\partial \Gamma^i{}_{km}}{\partial x^l} - \frac{\partial \Gamma^i{}_{lm}}{\partial x^k} + \Gamma^i{}_{ne} \Gamma^n{}_{km} - \Gamma^i{}_{nm} \Gamma^n{}_{ke}$$

Yang-Mills

$$\mathcal{L} = -\frac{1}{4} F_{\mu\nu}^a F_a{}^{\mu\nu}$$

$$F_{\mu\nu}^a = \partial_\mu A_\nu^a - \partial_\nu A_\mu^a + g C_{abc} A_\mu^b A_\nu^c$$

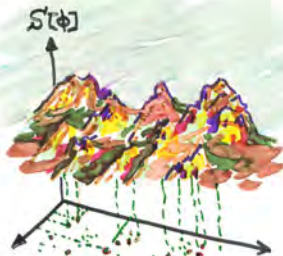
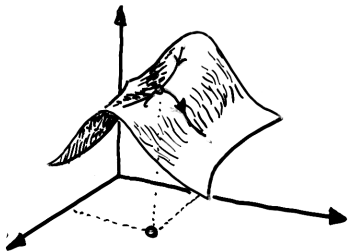
Quantum?

QFT path integrals : semi-classical quantization

a fractal set of saddles

TURBULENT Q.F.T. ?

a local unstable extremum



$$(\text{observable}) = \sum_{\text{set}}^{\text{fractal}} \frac{e^{i S_n[\phi_c]/\hbar}}{\sqrt{\frac{\partial^2 S}{\partial \phi_i \partial \phi_j}}}$$

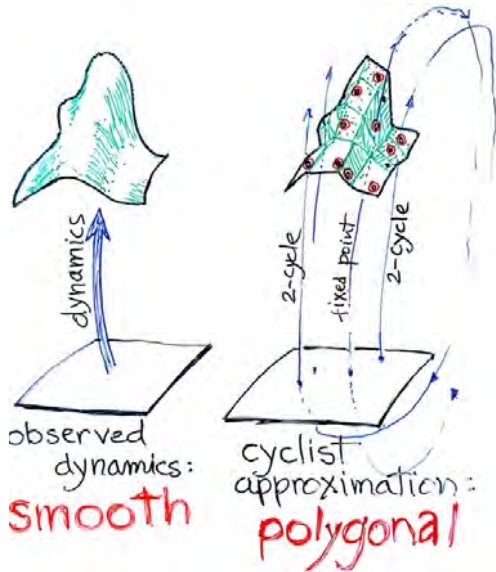
learn to **count** + weigh unstable saddles

the very short answer : POT



if you win : I teach you how

(for details, see ChaosBook.org)



tessellate the state space by recurrent flows

classical trace formula for continuous time flows

$$\sum_{\alpha=0}^{\infty} \frac{1}{s - s_{\alpha}} = \sum_p T_p \sum_{r=1}^{\infty} \frac{e^{r(\beta A_p - s T_p)}}{|\det(\mathbf{1} - M_p^r)|}$$

relates the spectrum of the evolution operator

$$\mathcal{L}(x', x) = \delta(x' - f^t(x)) e^{\beta A(x, t)}$$

to the unstable periodic orbits p of the flow $f^t(x)$.

classical trace formula for averaging over 2-tori

something like

$$\sum_{\alpha=0}^{\infty} \frac{1}{s - s_{\alpha}} = \sum_p V_p \sum_{r=1}^{\infty} \frac{e^{r(\beta A_p - s V_p)}}{|\det \mathcal{J}_{p^r}|}$$

weighs the unstable relative prime (all symmetries quotiented) d -torus p by the inverse of its Hill determinant, the determinant (state space volume) of its orbit Jacobian matrix \mathcal{J}_p

$$\det \mathcal{J}_p$$

and V_p is the volume

$$V_p = T_p L_p$$

of the prime spacetime tile p

extras

speculation : code discrete Lagrangian methods?

the idea : construct a discrete counterpart to the considered system

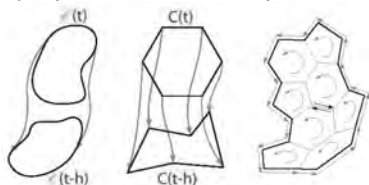
variational integrator : evolution map that corresponds to the discrete Euler–Lagrange equations

Discrete Lagrangian methods

action $S(q) = \int_0^T dt L(q, \dot{q})$ + Hamilton's principle $\delta S(q) = 0$

$$\text{discretize } \int_{t_k}^{t_{k+1}} L(q, \dot{q}) dt \approx \Delta t L(q_k, q_{k+1}).$$

symplectic methods preserve phase-space areas¹³



(left) Kelvin's circulation advected by the flow is constant
(middle) the discrete version, on a Voronoi loop
(right) circulation is constant on any discrete loop.

¹³J. E. Marsden and M. West, *Acta Numerica* **10**, 357–514 (2001).

Discrete Lagrangian codes ?

so far, no codes for
discretized spatiotemporal action / Lagrangian density

$$S = \int dq^d \mathcal{L}(q)$$

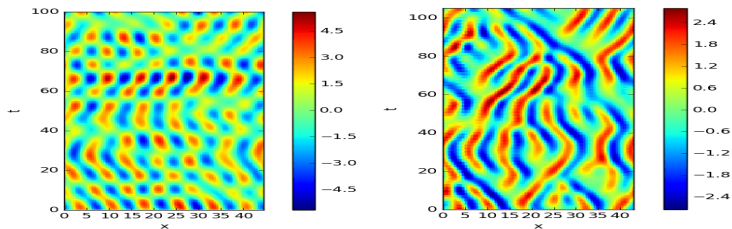
symplectic Euler incompressible fluid dynamics time-evolution
codes exist¹⁴

claim : can apply to non-conservative system

Navier-Stokes?

¹⁴D. Pavlov et al., Physica D **240**, 443–458 (2011).

an intermediate spacetime domain

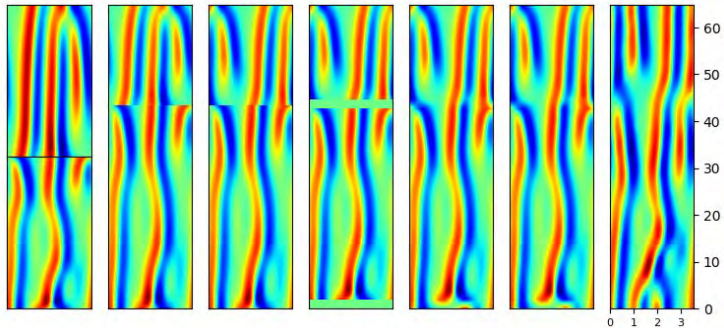


(left) $\bar{L} = 2\pi\sqrt{2}$ modulated initial random guess
 $(L_0, T_0) = (5\bar{L}, 100) = (44.4, 100)$

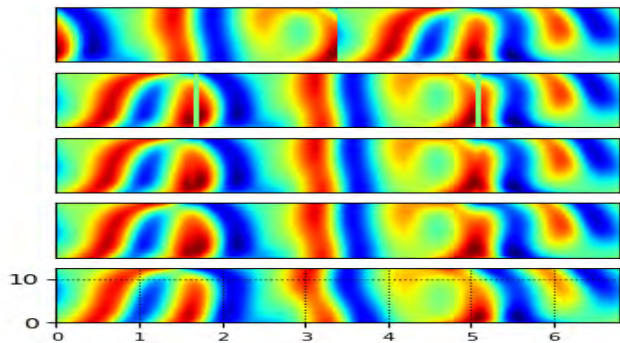
(right) Resulting invariant 2-torus
 $(L_f, T_f) = (43.066, 105.08) = (L_0 - 1.363, T_0 + 5.08)$

Adjoint descent took only 7 laptop CPU seconds

temporally glued Frankenstein

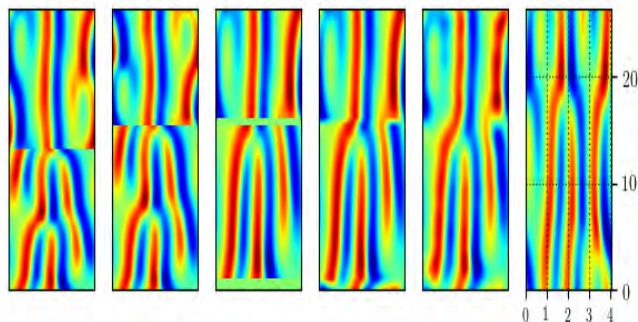


spatial gluing of two invariant 2-tori



- 1) two invariant 2-tori side by side
- 2) initial invariant 2-tori split into smaller tiles
- 3) a guess invariant 2-torus obtained by gluing / smoothing
- 4) converges to a larger invariant 2-torus

temporal gluing of two invariant 2-tori



- 1) an invariant 2-torus atop another invariant 2-torus
- 2) initial invariant 2-tori split into smaller tiles
- 3) a guess invariant 2-torus obtained by gluing / smoothing
- 4) converges to a larger invariant 2-torus

KS invariant 2-tori found by rocket science

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the initial guess

the converged solution $u(x, t)$