

a spatiotemporal theory of turbulence computational challenges

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working notes
Georgia Tech

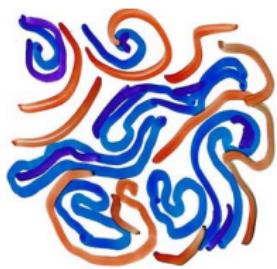
October 9, 2020

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?



?

⇒ other swirls ⇒



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

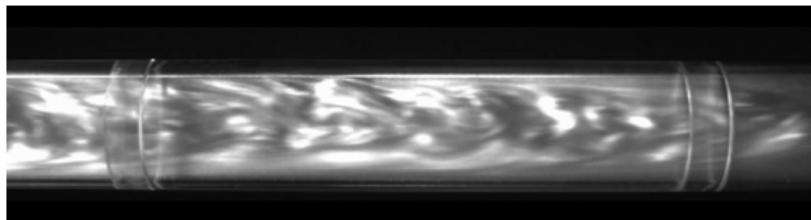
$$\frac{\partial \mathbf{v}}{\partial t} + (\mathbf{v} \cdot \nabla) \mathbf{v} = \frac{1}{R} \nabla^2 \mathbf{v} - \nabla p + \mathbf{f}, \quad \nabla \cdot \mathbf{v} = 0,$$

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

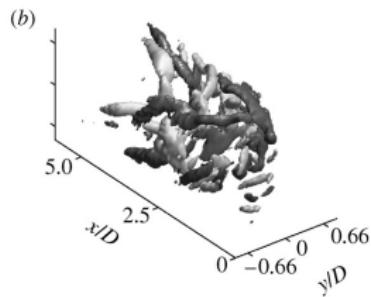
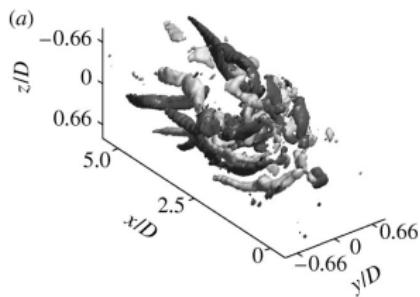
describe turbulence

starting from the equations (no statistical assumptions)

challenge : experiments are amazing

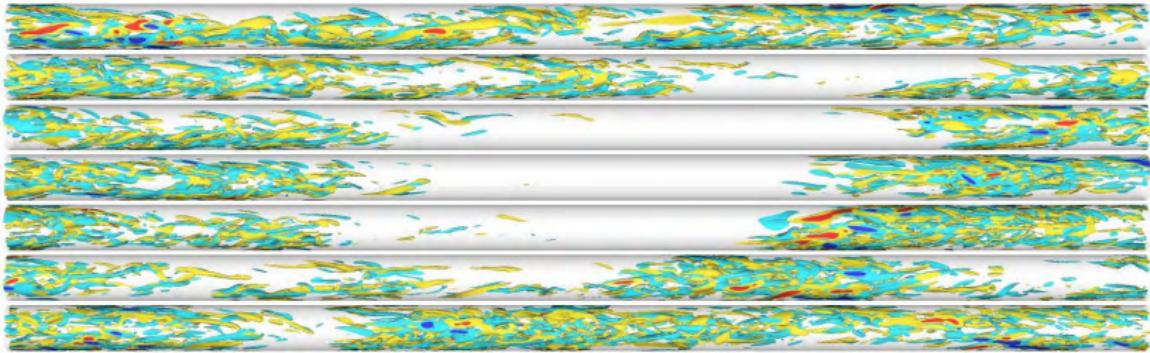


T. Mullin lab



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

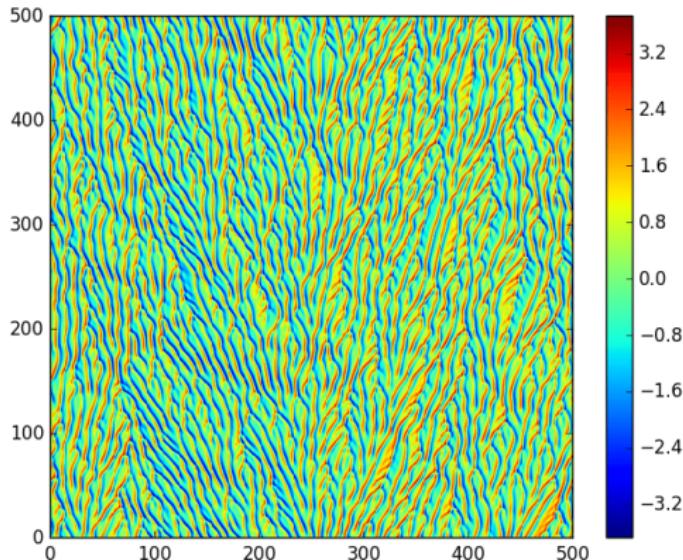
(1975)

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

describes spatially extended systems such as

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

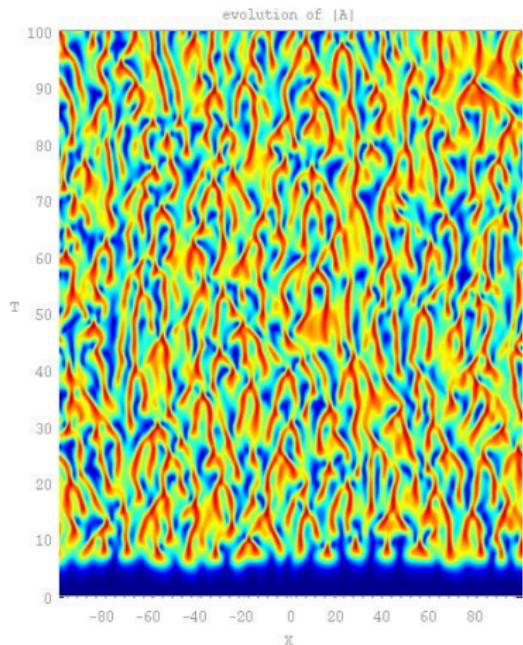
an example : Kuramoto-Sivashinsky on a large domain



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

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

another example of large spacetime domain simulation complex Ginzburg-Landau

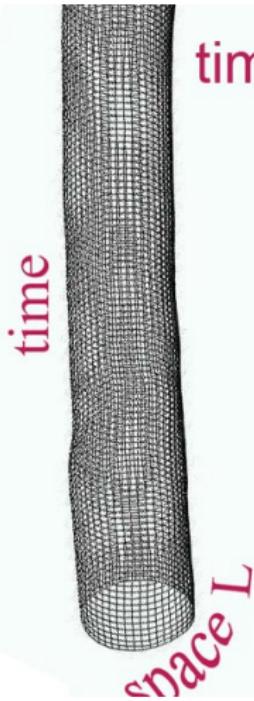


(will return to this)

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

compact space, infinite time cylinder

time evolution, periodic space



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

compact space, infinite time

Kuramoto-Sivashinsky equation

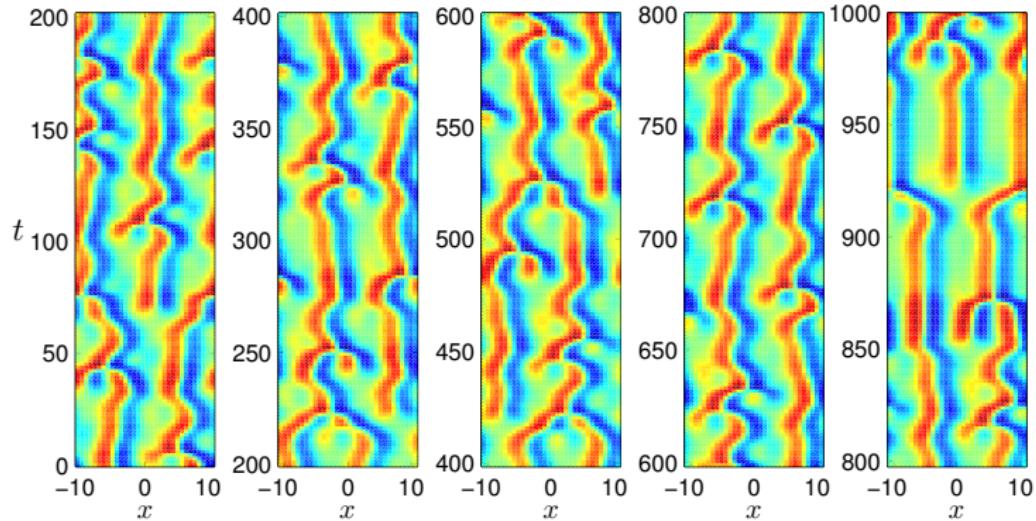
$$u_t = -(+\nabla^2 + \nabla^4)u - u\nabla u, \quad x \in [-L/2, L/2],$$

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

$$\begin{aligned}\mathbf{u}^\top &= (u, u', u'', u''') \\ \text{where } u' &\equiv u_x, \quad u'' \equiv u_{xx}, \quad u''' \equiv u_{xxx}\end{aligned}$$

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

Kuramoto-Sivashinsky = four coupled 1st order PDEs

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

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

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

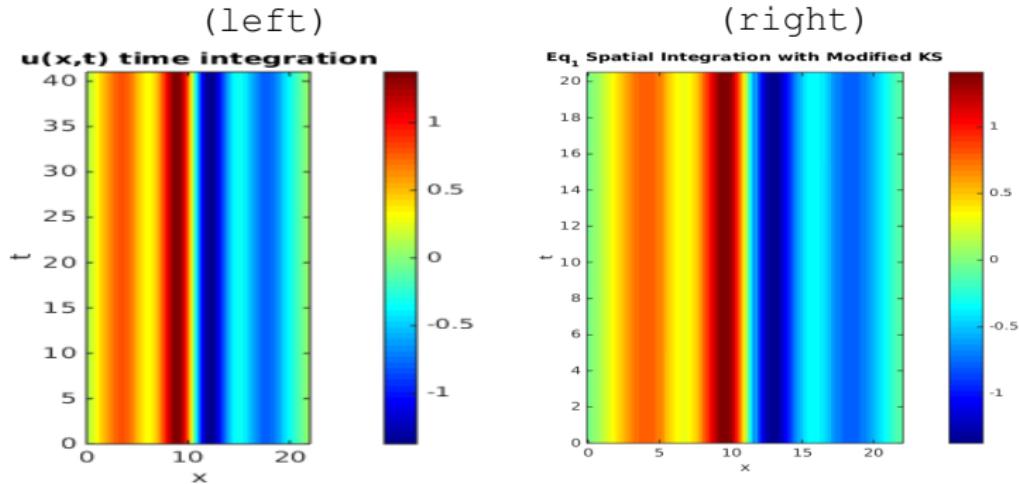
specified for all $t \in [0, T]$, at a fixed space point x_0

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

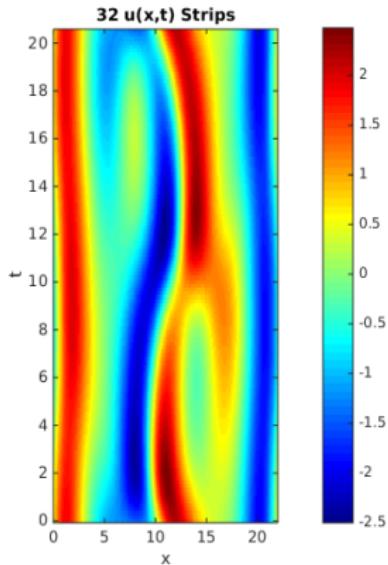
initial condition for the spatial integration is the time strip

$u(x_0, t)$, $t = [0, T]$, where time period $T = 0$, spatial x period is $L = 22$.

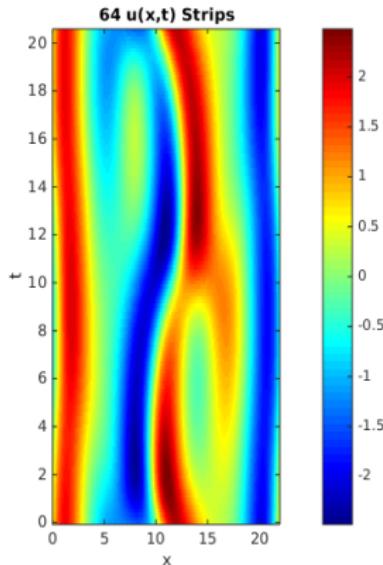
Michelson 1986

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

(left)



(right)



(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**

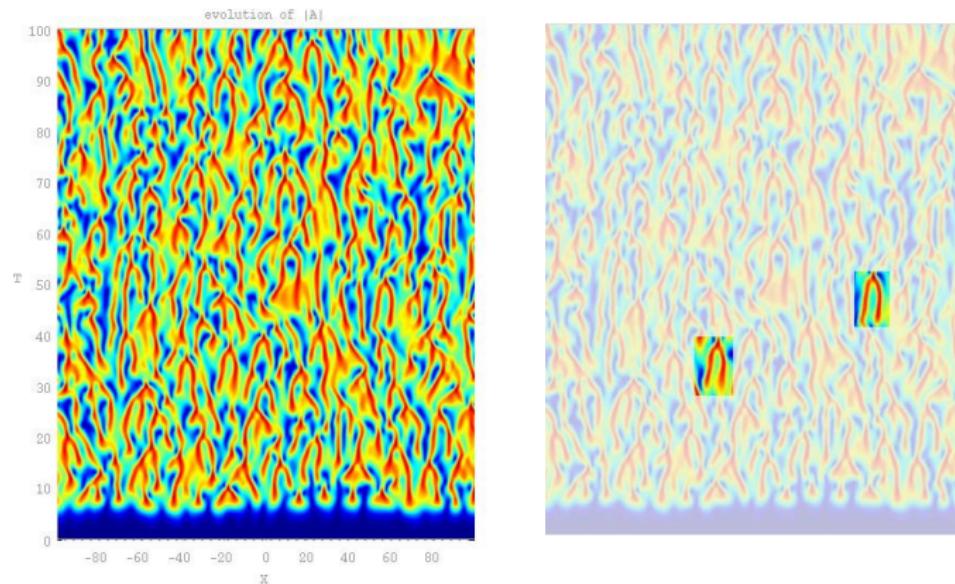
rethink the calculation

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

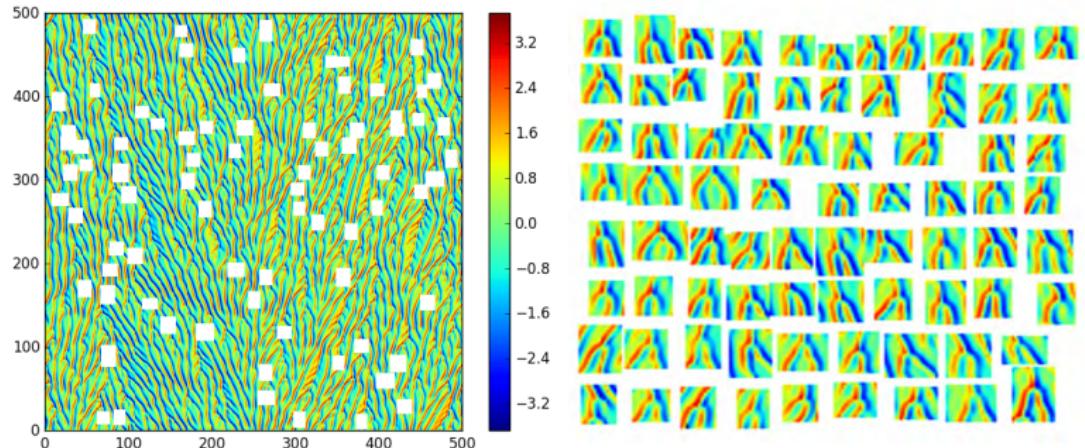


[left-right] space $x \in [-L/2, L/2]$

[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

chronotope²

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

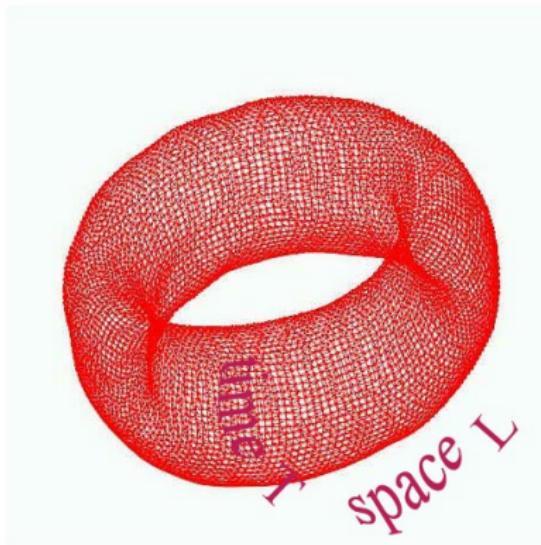
— Wikipedia : Chronotope

- 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

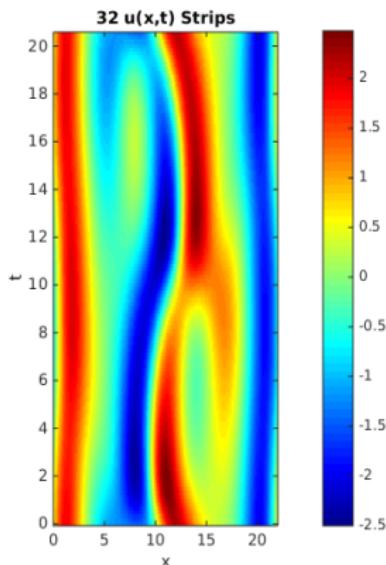
a state space point is *periodic* if its orbit returns to it after a finite time T ;
such orbit tiles the time axis by infinitely many repeats

1 time, $d-1$ space dimensions

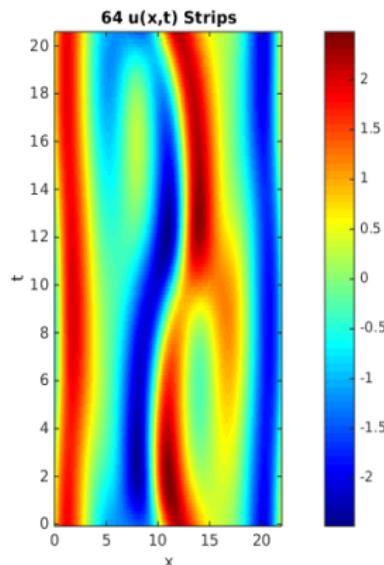
a state space point is *spatiotemporally periodic* if it belongs to an invariant d -torus \mathcal{R} ;
such torus tiles the spacetime by infinitely many repeats

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

(left)



(right)



(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

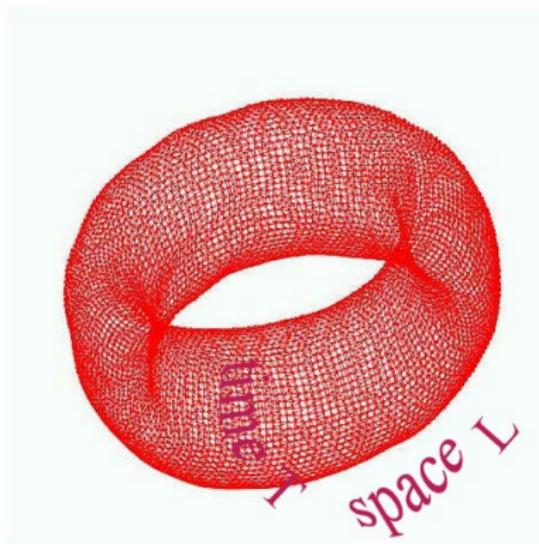
fixed point condition

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

every calculation is a spatiotemporal lattice calculation

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

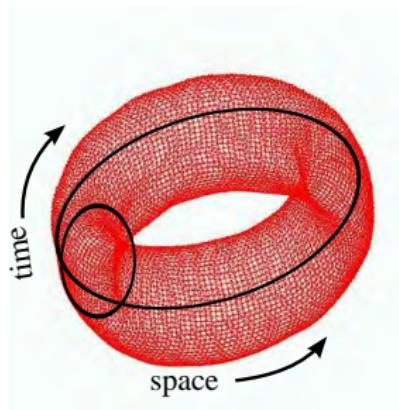
periodic spacetime : 2-torus



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

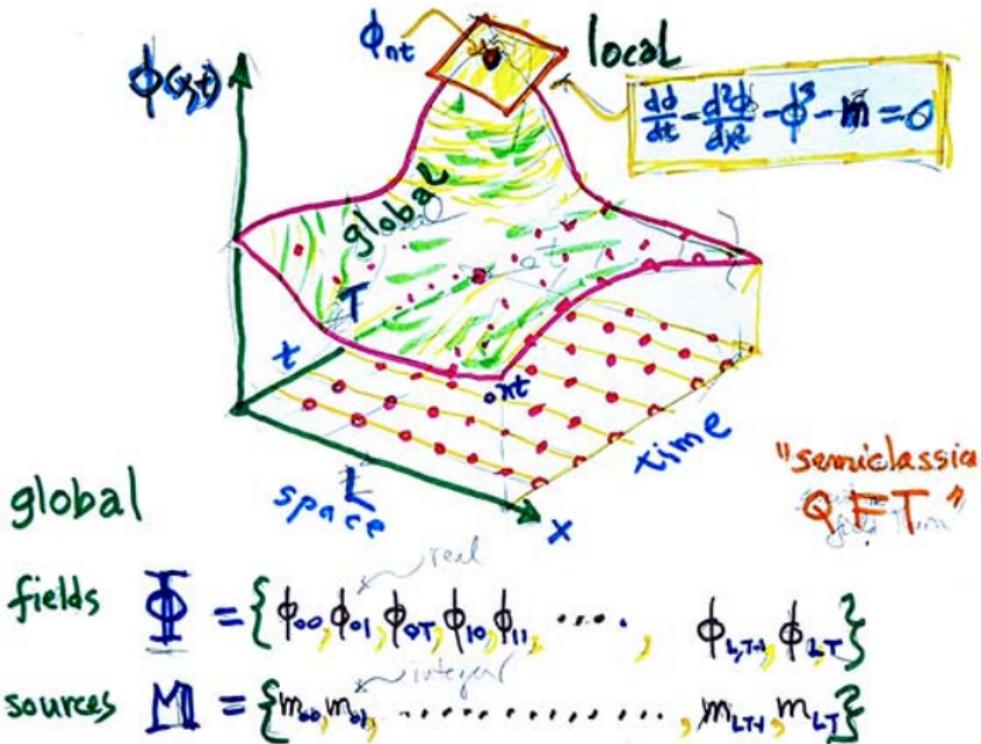
at each lattice point $k\ell$
the tangent field at $\tilde{u}_{k\ell}$

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

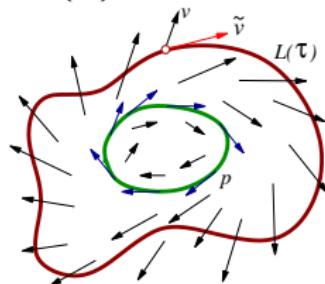
- ① turbulence in large domains
- ② space is time
- ③ spacetime
- ④ spacetime computations**
- ⑤ bye bye, dynamics

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

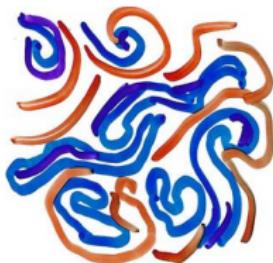
$$F^2[\tilde{x}] = \oint_L ds (\tilde{v} - v)^2; \quad \tilde{v} = \tilde{v}(\tilde{x}(s, \tau)), \quad v = v(\tilde{x}(s, \tau)),$$

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

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

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.

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

“adjoint descent” method defined by choosing⁴

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

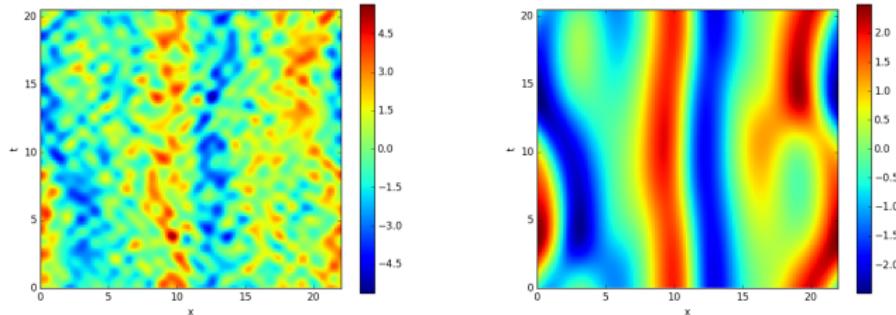
⁴M. Farazmand, J. Fluid M. 795, 278–312 (2016).

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

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

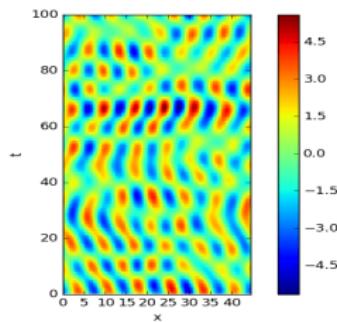
(right) the resulting adjoint descent converged invariant 2-torus

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

initial guess generation ?

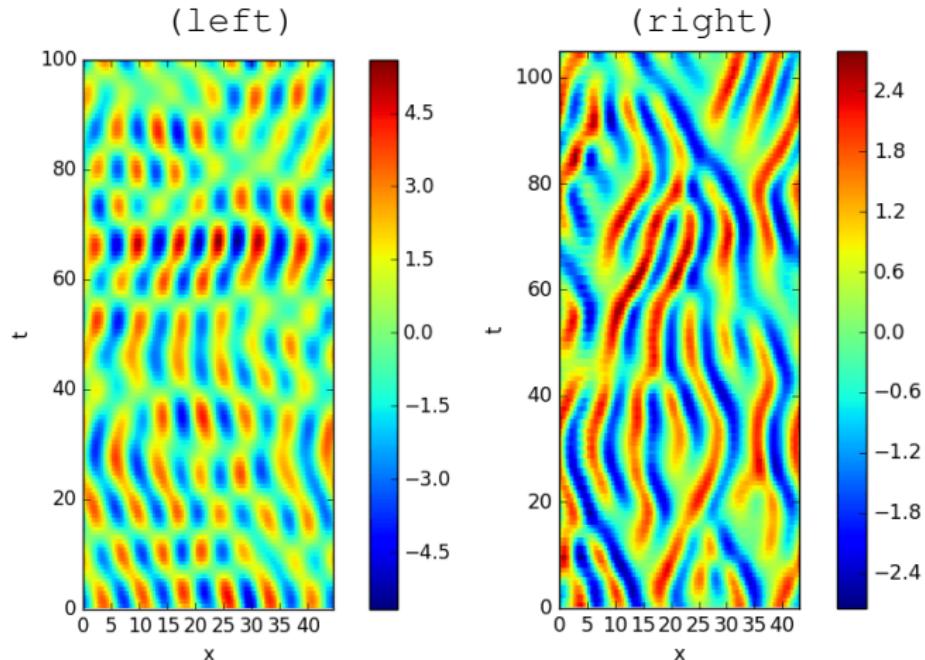
the time scale : the shortest ‘turnover’ scale characterized by the period of the shortest periodic orbit? Or perhaps the Lyapunov time?

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

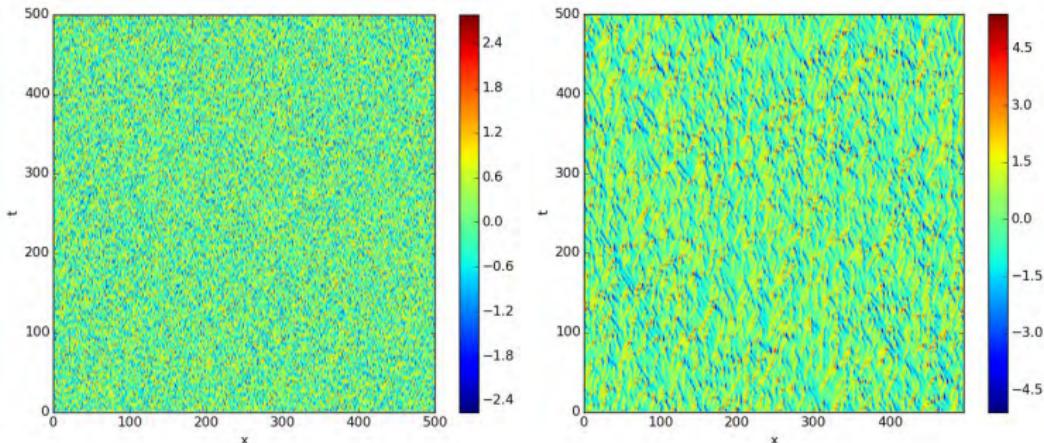
- ➊ inefficient, finds only the shortest, least unstable orbits^{5,6}
- ➋ can integrate only not far in time

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

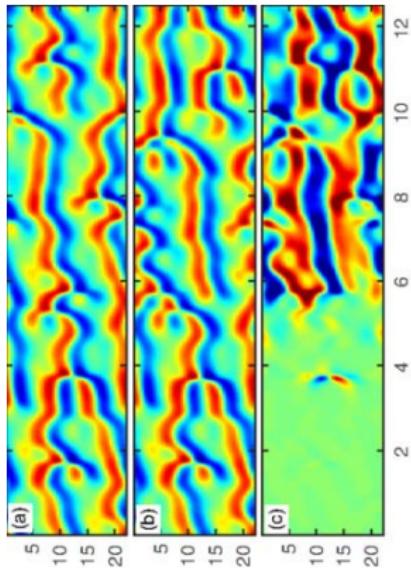


(left) random initial state on $(L, T) = (500, 500)$
(right) adjoint descent \rightarrow typical Kuramoto-Sivashinsky state

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?

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

embarrassment of riches

what to do?

Matthew N. Gudorf

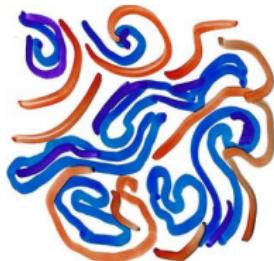
has 1 000's of such invariant 2-tori

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?



WATCH

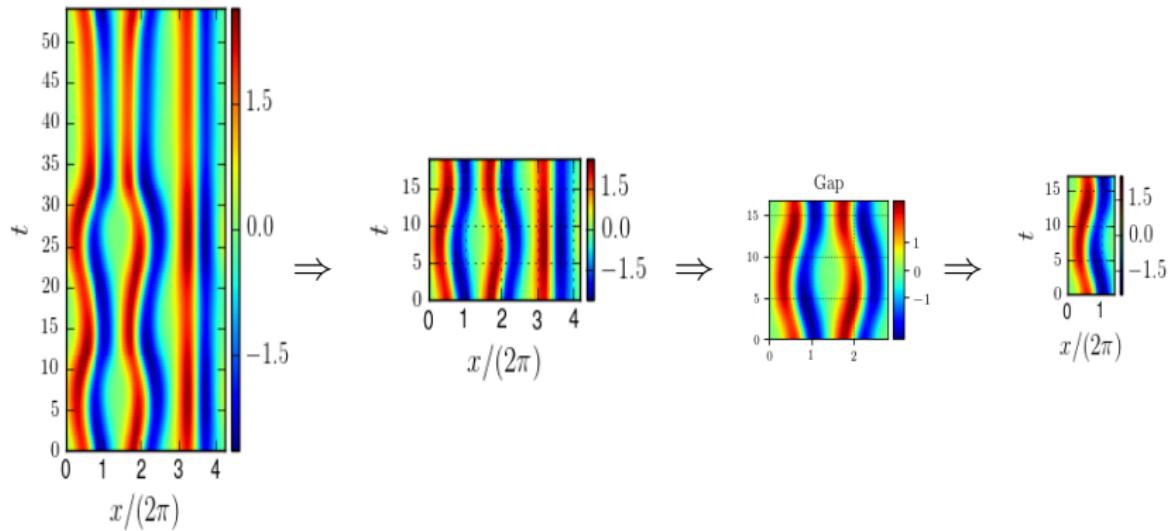
⇒ other swirls ⇒



by recurrent shapes!

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

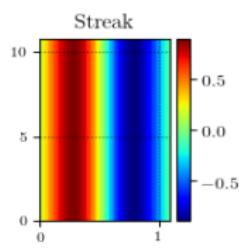
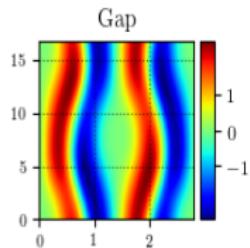
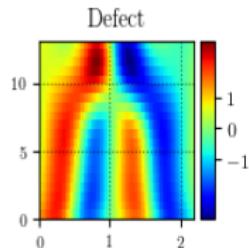
extracting a fundamental tile



- 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

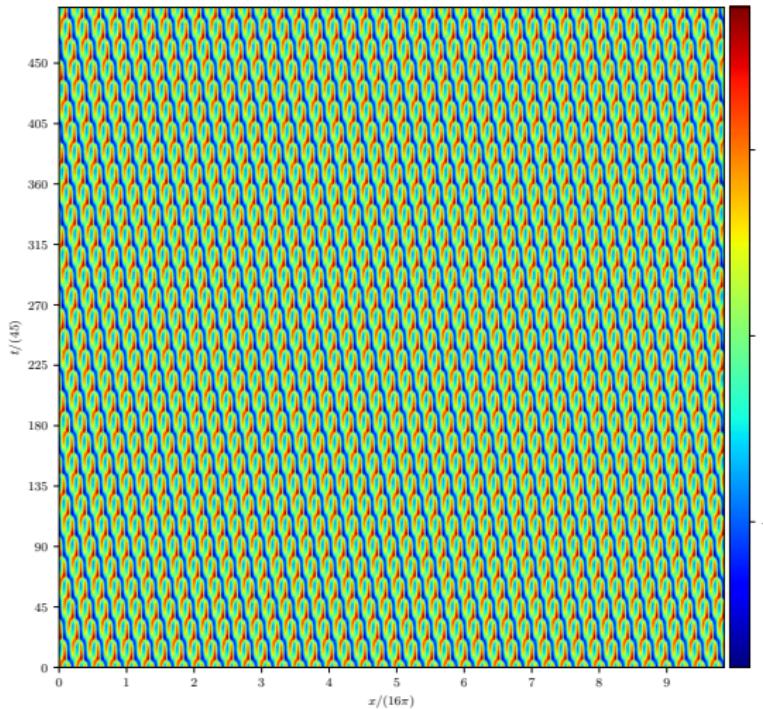
a trial set of prime (rubber) tiles

an alphabet of Kuramoto-Sivashinsky fundamental tiles



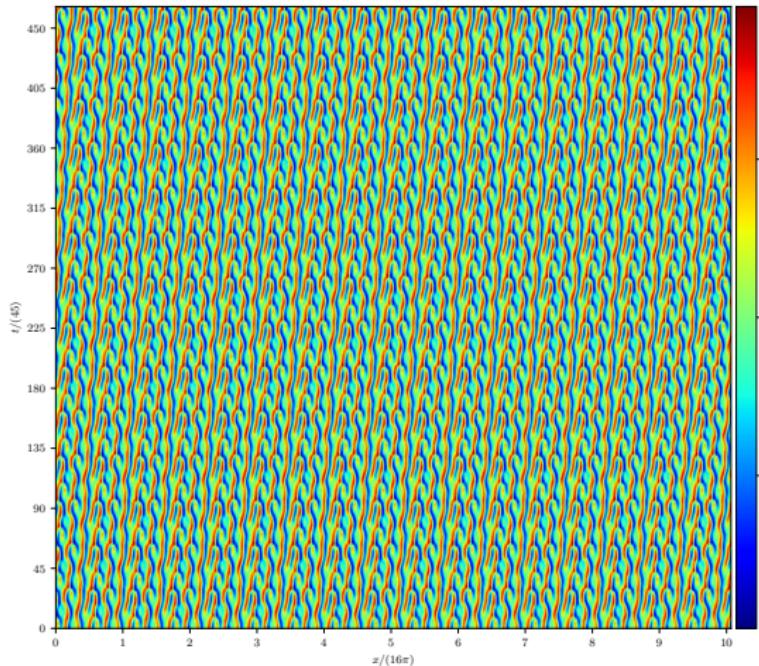
utilize also discrete symmetries :
spatial reflection, spatiotemporal shift-reflect, . . .

Kuramoto-Sivashinsky tiled by a small tile



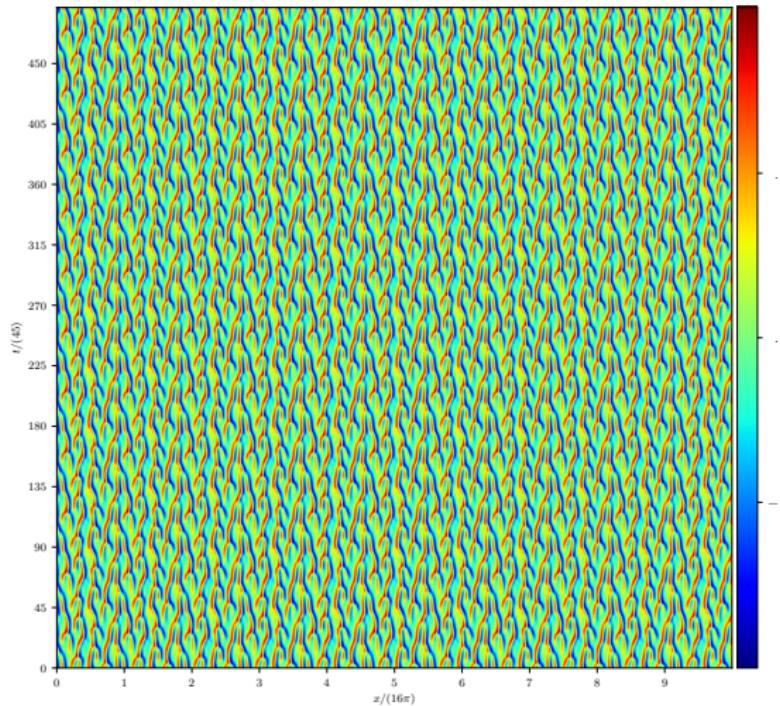
tiling by relative periodic invariant 2-torus
 $(L, T) = (13.02, 15)$

spacetime tiled by a larger tile



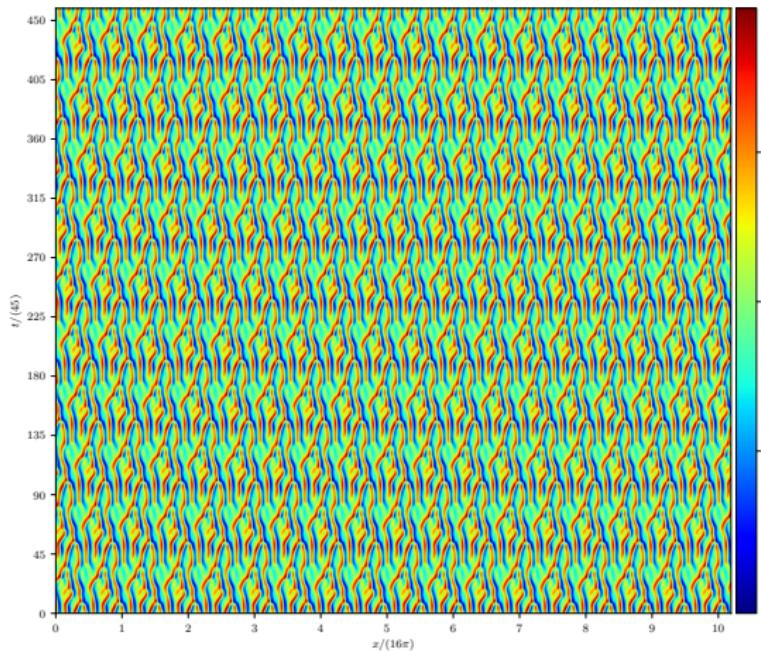
tiling by relative periodic invariant 2-torus
 $(L, T) = (33.73, 35)$

spacetime tiled by a tall tile



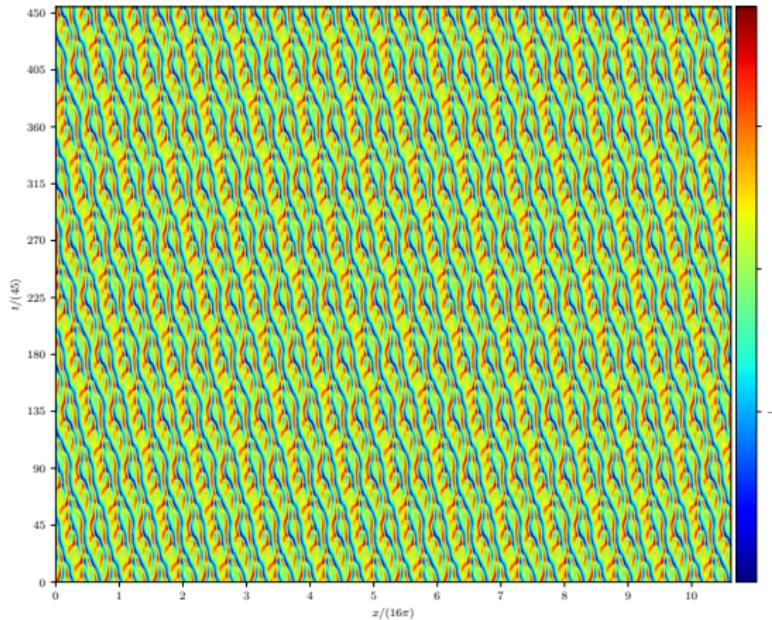
tiling by shift-reflect invariant 2-torus
 $(L, T) = (55.83, 24)$

spacetime tiled by a larger tile



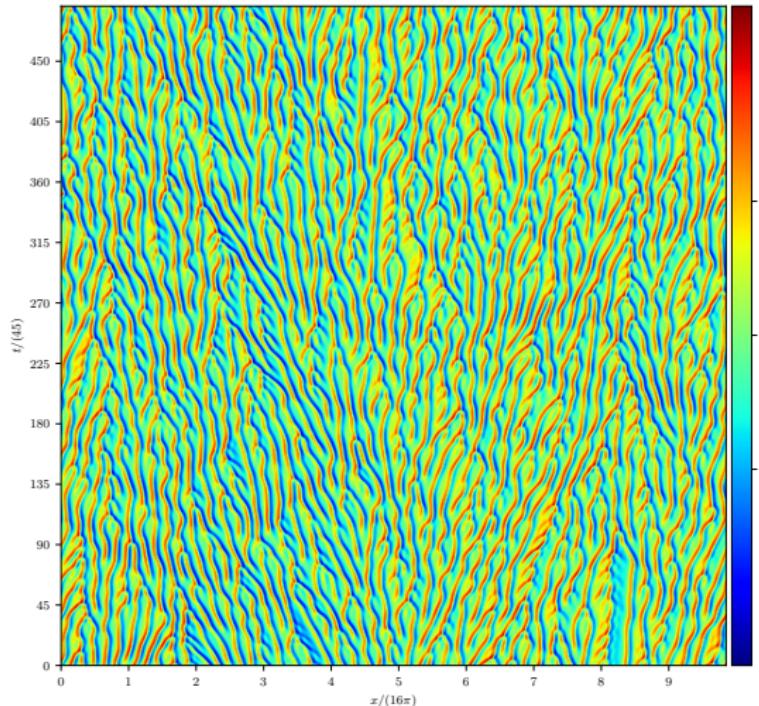
tiling by relative periodic invariant 2-torus
 $(L, T) = (32.02, 51)$

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!



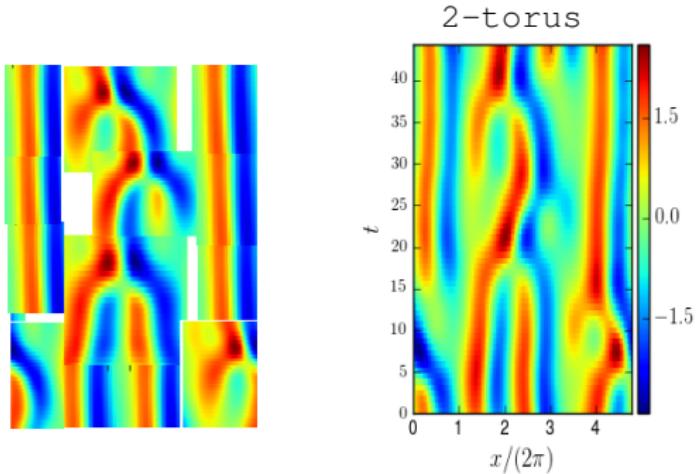
[horizontal] space $x \in [-L/2, L/2]$ [up] time evolution

part 6

- ➊ turbulence in large domains
- ➋ space is time
- ➌ spacetime
- ➍ fundamental tiles
- ➎ gluing tiles**
- ➏ bye bye, dynamics

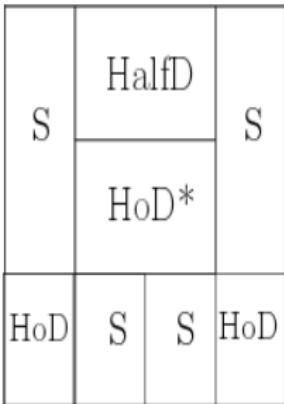
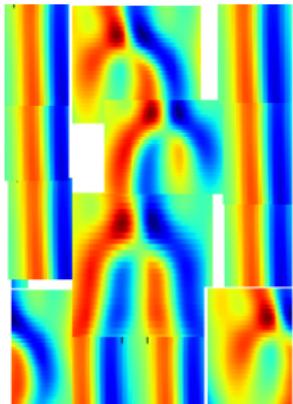
a qualitative tiling guess

a tiling and the resulting solution



turbulence.zip : each solution has a unique symbolic name

symbolic dynamics is 2-dimensional!



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

- 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

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

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 → limit

number of cores → 10^6 → ...

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

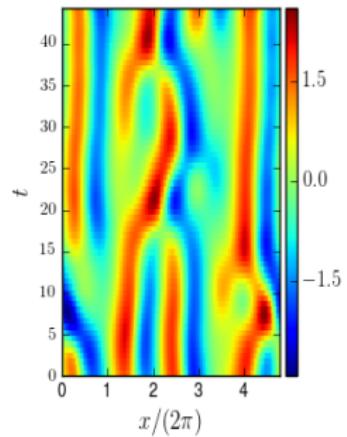
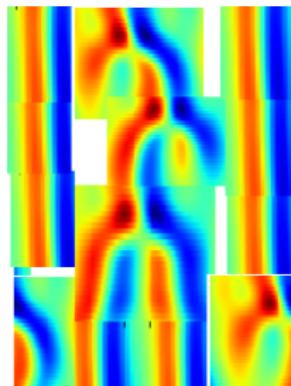
⇒ a path towards exascale DNS of turbulent flows

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

enumerate hierarchically spatiotemporal patterns

2D symbolic encoding \Rightarrow admissible solutions

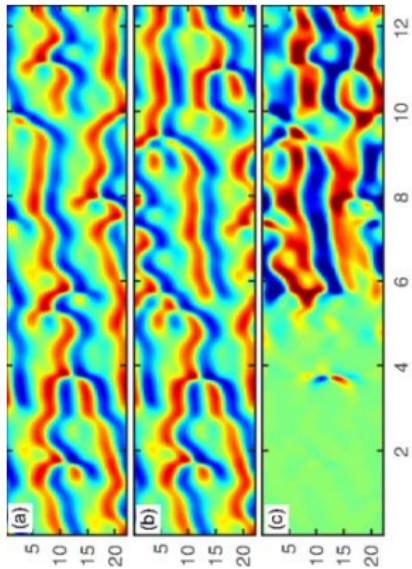
0	2	0
0	1	0
0	1*	
0		
1	0	0
	0	0



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

machine learning will be needed

“reservoir computing” example¹²



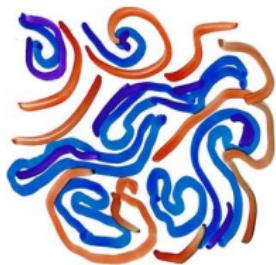
- (a) data:
Kuramoto-Sivashinsky simulation
- (b) reservoir computing prediction
- (c) two subtracted agree to
 ~ 5 Lyapunov times

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?



NO!

⇒ other swirls ⇒



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

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

summary

- ➊ study turbulence in infinite spatiotemporal domains
- ➋ theory : classify all spatiotemporal tilings
- ➌ 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

- ➊ (semi-)classical field theories
- ➋ state space
- ➌ 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

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

$$R^i_{klm} = \frac{\partial \Gamma^i_{km}}{\partial x^l} - \frac{\partial \Gamma^i_{ki}}{\partial x^m} + \Gamma^x_{he} \Gamma^h_{km} - \Gamma^e_{nm} \Gamma^h_{ke}$$

Yang-Mills

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

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

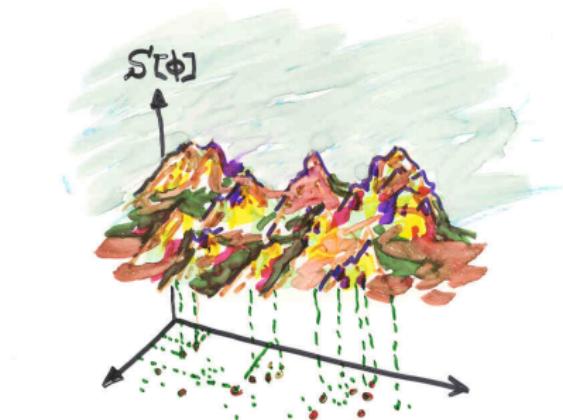
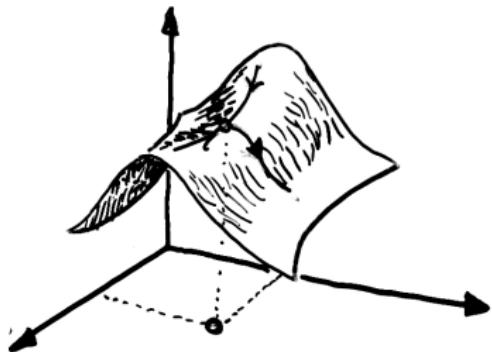
Quantum 3

QFT path integrals : semi-classical quantization

a fractal set of saddles

TURBULENT Q.F.T. 2

a local unstable
extremum



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

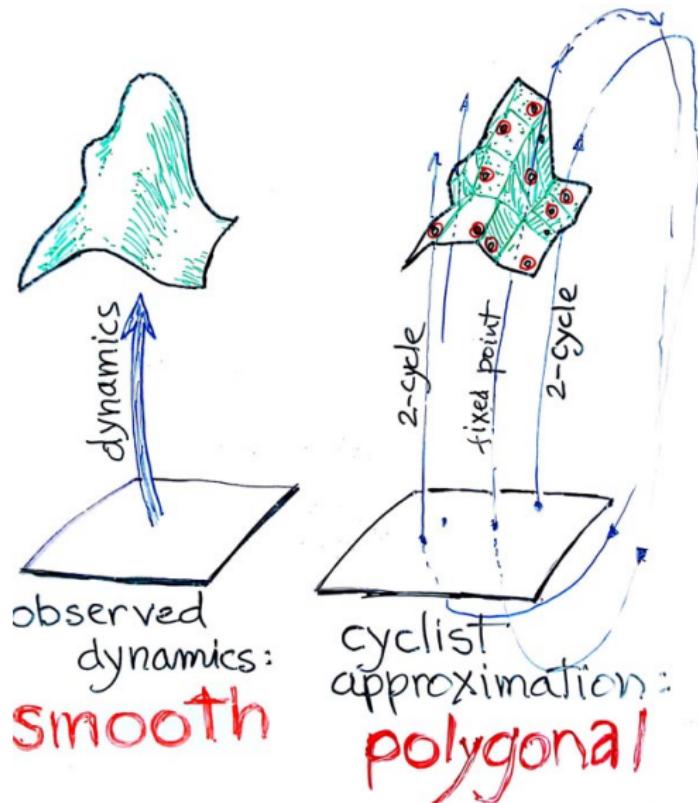
learn to
count + weigh unstable
saddles

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

we conjecture

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

weights the unstable relative prime (all symmetries quotiented)
 d -torus p by its variational Hessian H via Hill's formula

$$\det H_p = \det(1 - J_p)$$

speculation : code discrete Lagrangian methods?

the idea : construct a discrete counterpart to the considered system

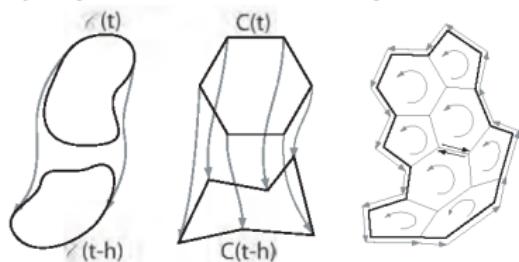
variational integrator : evolution map that corresponds to the discrete Euler–Lagrange equations

Discrete Lagrangian methods

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

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?

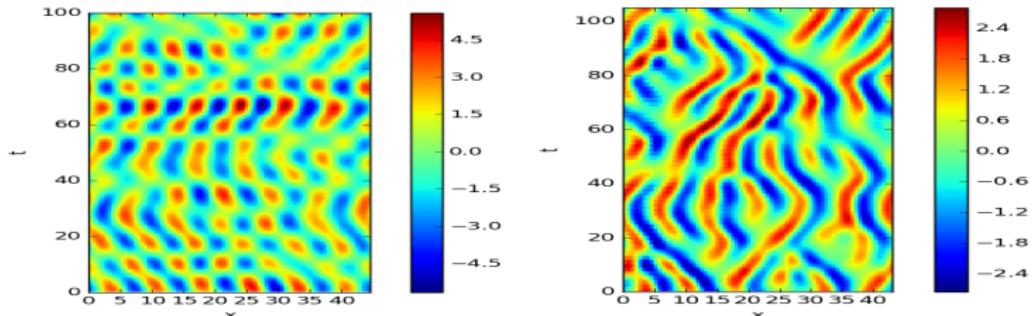
¹⁴D. Pavlov et al., Physica D **240**, 443–458 (2011).

XXX

XXX

XXX

an intermediate spacetime domain



(left) $\bar{L} = 2\pi\sqrt{2}$ modulated initial random guess

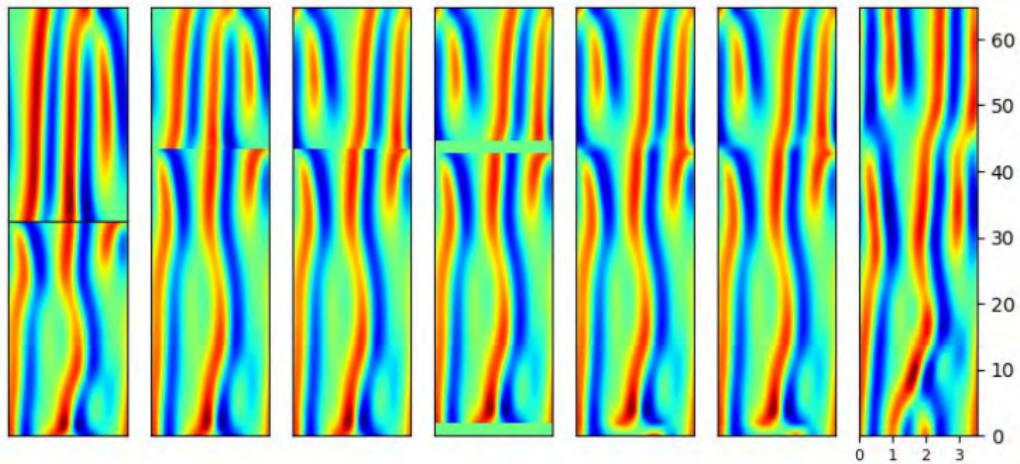
$$(L_0, T_0) = (5\bar{L}, 100) = (44.4, 100)$$

(right) Resulting invariant 2-torus

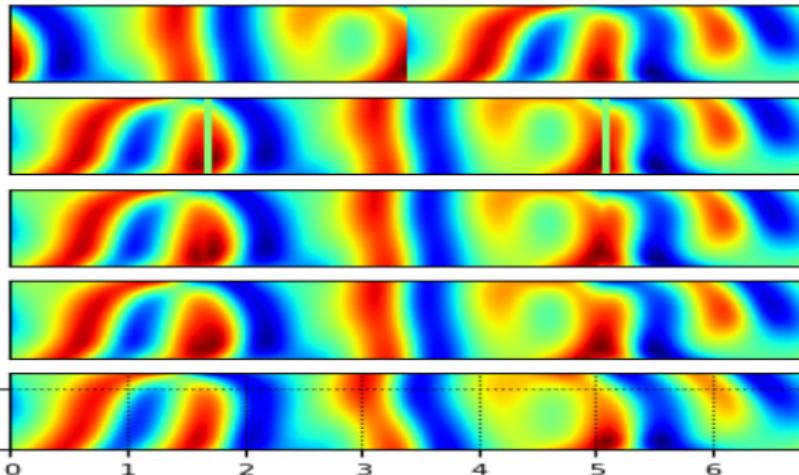
$$(L_f, T_f) = (43.066, 105.08) = (L_0 - 1.363, T_0 + 5.08)$$

Adjoint descent took only 7 laptop CPU seconds

temporally glued Frankenstein

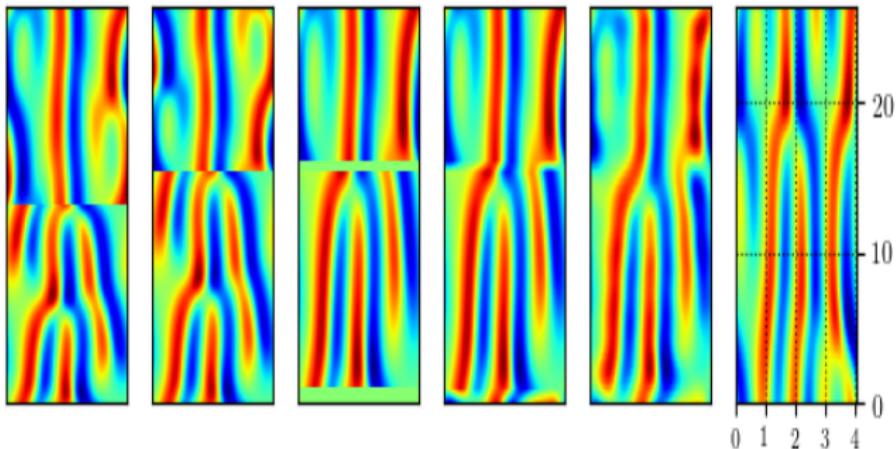


spatial gluing of two invariant 2-tori



- 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

15

the initial guess

the converged solution $u(x, t)$

¹⁵Q. Wang et al., Phys. Fluids **25**, 110818 (2013).