Bistable model

A teaser: machine learning model?

Subcritical turbulence spreading and avalanche birth

R.A. Heinonen and P.H. Diamond

CASS and Department of Physics University of California, San Diego

Festival de Théorie 2019, Aix-en-Provence

Supported by the Department of Energy under Award Number DE-FG02-04ER54738

Introduction

- Turbulence spreading is an important nonlinear phenomenon in drift wave turbulence
- Challenge the conventional wisdom on spreading and point out issues with the supercritical Fisher equation paradigm
- Suggest a new model based on subcritical turbulence, which features avalanche-like spatiotemporal intermittency
- We make testable predictions which distinguish it from Fisher
- I might say the words 'phase' and 'dynamics' at some point, but probably not consecutively

Bistable model

A teaser: machine learning model?

Outline



2 Bistable model



Bistable model

A teaser: machine learning model?

Background: turbulence spreading and avalanching

A teaser: machine learning model?

What the Fick?: turbulence spreading

- Spreading is important because it spells doom for local Fickian transport models
- Turbulence can radially self-propagate (even into linearly stable zones!) via nonlinear coupling

$$\partial_t \varepsilon_{\mathbf{k}} \sim -\sum_{\mathbf{k}'} (\mathbf{k} \cdot \mathbf{k}' \times \hat{z})^2 |\tilde{\phi}_{\mathbf{k}'}|^2 R(\mathbf{k}, \mathbf{k}') I_{\mathbf{k}} \to \frac{\partial}{\partial x} D_x(I_{\mathbf{k}}) \frac{\partial}{\partial x} I_{\mathbf{k}} - \mathbf{k}\mathbf{k} : \mathbf{D}I_{\mathbf{k}}$$

$$D_{\mathrm{x}} = \sum_{\mathbf{k}'} k_{\mathrm{y}}^{\prime 2} |\phi_{\mathbf{k}'}|^2 R(\mathbf{k}, \mathbf{k}')$$

• Decouples flux-gradient relation: local turbulence intensity now depends on global properties of the profiles

Bistable model

A teaser: machine learning model? 0000000

Depiction of spreading



Figure: Spatiotemporal evolution of flux-surface-averaged turbulence intensity in toroidal GK simulation. Linearly unstable region is 0.42 < r < 0.76. From [Wang et al., 2006]

Background: turbulence spreading and avalanching $\tt 000\bullet00000$

Bistable model

A teaser: machine learning model?

Avalanches

- Bursty, intermittent transport events associated with SOC. Account for a large percentage of total flux!
- Initially localized fluctuation cascades through neighboring regions via gradient coupling
- Closely related to spreading: both result in fast, mesoscopic turb front propagation. Unified model?



Figure: Heat flux spectrum from GK simulation showing 1/f scaling

Background: turbulence spreading and avalanching oooooooo

Bistable model

A teaser: machine learning model?

Depiction of avalanching



Figure: Pressure (left) and potential (right) contours for simuliations of resistive drift interchange turbulence [Carreras et al., 1996]

Fisher model

• Conventional wisdom for spreading is Fisher-type equation for turbulence intensity:



• For $\gamma_0 > 0$, dynamics characterized by traveling fronts connecting unstable "laminar root" I = 0 and saturated "turbulent root" $I = \gamma_0 / \gamma_{nl}$ with speed $c = \sqrt{\frac{D_0 \gamma_0^2}{2\gamma_{nl}}}$

Bistable model

A teaser: machine learning model?

Depiction of Fisher evolution



Figure: Evolution of traveling turbulence front in Fisher model. From [Gürcan and Diamond, 2006]

Bistable model

A teaser: machine learning model?

How does Fisher do?

- Propagation speed and characteristic front size $\ell \sim \sqrt{D/\gamma_0}$ in reasonable agreement with simulation
- Can be derived with some rigor from Fokker-Planck approach or renormalization of Hasegawa-Wakatani [Gürcan and Diamond, 2005, Gürcan and Diamond, 2006]
- But: weak spreading into stable zone. Dubiously consistent with experiment?



Figure: Experiment by Nazikian et al 2005 clearly showing fluctuations in stable zone

When does Fisher even make sense?

- Fisher model purports to describe spreading of a patch of turbulence in linearly unstable zone
- Begs the question: why didn't noise already excite the whole system to turbulence?
- Only relevant if $\gamma_0 \ll c/\Delta x$ i.e. $\Delta x^2 \gamma_{nl} \ll D_0$
- Otherwise, physical fronts separating laminar/turbulent domains generally require *bistability* à la [Pomeau, 1986]

Bistable model •••••••••

A teaser: machine learning model?

Bistable model

Bistable model ••••••••• A teaser: machine learning model? 0000000

A new(ish) model is born

 Heinonen and Diamond 2019: propose phenomenological model of form

$$\partial_t I = \gamma_1 I + \gamma_2 I^2 - \gamma_3 I^3 + \partial_x (D(I)\partial_x I)$$

• take
$$D(I) = D_0 I$$

- New physics: nonlinear turbulence drive
 x I². Can sustain sufficiently large fluctuations even when linearly damped
- Bistable in weak damping regime
- Estimate $\gamma_1 \sim \epsilon \omega_*, \ \gamma_{2,3} \sim \omega_*, \ D_0 \sim \chi_{GB}$
- But is MF plasma actually subcritically unstable?

Bistable model

A teaser: machine learning model?

Evidence for subcriticality

- [Inagaki et al., 2013]: experiments demonstrate hysteresis between fluctuation intensity and driving gradient (no TB present). Suggests bistable S-curve relation?
- Turbulence subcritical in presence of strong perpendicular flow shear [Carreras et al., 1992, Barnes et al., 2011, van Wyk et al., 2016] or in the presence of magnetic shear [Biskamp and Walter, 1985, Drake et al., 1995]
- Profile corrugations [Waltz, 1985, Waltz, 2010] and phase space structures [Lesur and Diamond, 2013] can drive nonlinear instability



Figure: Hysteresis between intensity and gradient, flux and gradient

Background: turbulence spreading and avalanching ${\tt ooooooooo}$

Bistable model

A teaser: machine learning model?

Cousin models

- Compare to bistable models for subcritical transition to fluid turbulence [Barkley et al., 2015, Pomeau, 2015].
- Compare to [Gil and Sornette, 1996] model for sandpile avalanches

$$\partial_t S = \gamma \left(|\partial_x h| / g_c - 1 \right) S + \beta S^2 - S^3 + \partial_x (D_S S \partial_x S) \\ \partial_t h = \partial_x (D_h S \partial_x h).$$

- $S \leftrightarrow I$, $h \leftrightarrow p$
- Weak gradient coupling limit $D_p \ll D_I \Rightarrow$ our model
- Strong gradient coupling limit: I slaved to p. ∂_xp ∝ I⁻¹ ⇒ linear term is c − γI, where c is a constant which depends on BCs. Bistable again!

Bistable model

A teaser: machine learning model?

Model analysis I

$$\partial_t I = \gamma_1 I + \gamma_2 I^2 - \gamma_3 I^3 + \partial_x (D(I)\partial_x I)$$

- Qualitatively similar to Fisher EXCEPT in weak damping case $\gamma_1 < 0$ and $\gamma_2^2 > 4|\gamma_1|\gamma_3$
- Can then transform to Zel'dovich/Nagumo equation

$$\partial_t I = f(I) + \partial_x (DI\partial_x I)$$
$$f(I) \equiv \gamma I (I - \alpha) (1 - I)$$



where
$$\alpha \equiv I_{-}/I_{+}, \ \gamma \equiv I_{+}^{2}\gamma_{3}, \ D \equiv I_{+}D_{0}, \ I_{\pm} \equiv (\gamma_{2} \pm \sqrt{\gamma_{2}^{2} - 4|\gamma_{1}|\gamma_{3}})/2\gamma_{3}$$

Background: turbulence spreading and avalanching 00000000

Bistable model

A teaser: machine learning model?

Model analysis II

• Can write in variational form

$$D(I)\partial_t I = -rac{\delta \mathcal{F}}{\delta I}$$

with free energy functional



and $d\mathcal{F}/dt \leq 0$

Bistable model

A teaser: machine learning model?

Model analysis III

- I = 0 metastable for $\alpha < \alpha^* = 3/5$, abs. stable for $\alpha > \alpha^*$
- "Potential barrier" at $I = \alpha$: threshold for onset of nonlinear instability



Figure: "Potential" part of \mathcal{F}

Bistable model

A teaser: machine learning model? 0000000

Model analysis IV

- Unlike Fisher, traveling fronts admitted in marginal/weak damping case!
- Propagation speed $c \sim \sqrt{D\gamma}$ (depends on α), characteristic scale $\ell \sim \sqrt{D/\gamma}$
- "Maxwell construction" for speed

$$c \int_{-\infty}^{\infty} D(I(z))I'(z)^2 dz = \int_{0}^{1} D(I)f(I) dI$$

z = x - ct

• Thus turbulence spreads if $\alpha < \alpha^*$, recedes if $\alpha > \alpha^*$. Corresponds to (meta)stability of fixed points

Bistable model

A teaser: machine learning model?

Penetration into stable zone I

- Consider spreading of turbulence from lin. unstable to lin. stable zone
- Simple model: $\gamma_1 = \gamma_g > 0$ for x < 0, $\gamma_1 = -\gamma_d < 0$ for x > 0
- Allow turbulent front to form in lefthand region and propagate
- In Fisher model, penetration is *weak*: forms stationary, exponentially-decaying profile with $\lambda \sim \sqrt{D_0/\gamma_{nl}} \sim \Delta_c$. Dubiously consistent with observation





Bistable model

A teaser: machine learning model? 0000000

Penetration into stable zone II

- However, in our model, a new front with reduced speed/amplitude forms in second region if weakly damped (i.e. γ_d is small enough that α < α^{*})
- Hence: can have ballistic propagation even in stable zone!
- More strongly delocalizing effect on the flux-gradient relation, compared to Fisher





Background: turbulence spreading and avalanching 00000000

Bistable model

A teaser: machine learning model?

Penetration into stable zone III



Figure: Spreading into stable zone in GK simulation with magnetic shear [Yi et al., 2014]. Ballistic propagation???

Bistable model

A teaser: machine learning model?

Local threshold behavior

- In contrast to Fisher, sufficiently large localized puff of turbulence will grow into front and spread. Suggestive of an avalanche triggered by initial seed
- How to determine threshold?



Two puffs differing only in spatial size are initialized; one grows and spreads, other collapses

Bistable model

A teaser: machine learning model?

Avalanche threshold

- Obviously puff amplitude must exceed $I_0 = \alpha$ or else $\gamma_{eff} = (I \alpha)(1 I) < 0$
- Consider "cap" of puff (part exceeding $l = \alpha$)
- Size threshold governed by competition between diffusion of turbulence out of cap and total nonlinear growth in cap (suggested by free energy functional)
- Sets scale $\sqrt{D/\gamma}$



Avalanche threshold (details)

- Strategy: assume initial puff is symmetric, has single max *l*₀ and single lengthscale *L*
- Expand intensity curve about max to quadratic order, plug into dynamical equation, integrate over extent of cap
- Result: growth if

$$L > L_{\min} = \sqrt{\frac{D(\alpha)I_0}{f(I_0) - \frac{1}{3}(I_0 - \alpha)f'(I_0)}} = \sqrt{\frac{3D\alpha I_0}{\gamma(I_0 - \alpha)((1 - 2\alpha)I_0 + \alpha)}}$$

• Power law
$$L_{min} \sim (I_0 - lpha)^{-1/2}$$

Background: turbulence spreading and avalanching ${\scriptstyle 00000000}$

Bistable model

A teaser: machine learning model?

Avalanche threshold: analytical vs. simulation



Figure: Numerical result for threshold at $\alpha = 0.3$ for three types of initial condition (Gaussian (I_1), Lorentzian (I_2), parabola (I_3)), compared with analytical estimate

Triggering an avalanche

- How might a puff of sufficient size form?
- Near linear marginality, threshold is weak:

$$I_- \sim rac{|\gamma_1|}{\gamma_2} \ll 1, \; L_{
m min} \sim \left(rac{\chi_{\it GB}}{\omega_*}
ight)^{1/2} \sim \Delta_c \, .$$

- Suggests threshold can be triggered by noise
- Simulations of model with appropriate choice of noise (multiplicative + small additive background) show that front propagation events will be intermittently excited

Bistable model: conclusions

- Natural extension of Fisher model that allows for coexistence of laminar/turbulent domains
- Supported by substantial evidence for subcritical turbulence
- Provides simple framework for understanding avalanching: local exceedance of nonlinear instability by turbulent puffs
- Key testable predictions: ballistic spreading into weakly linearly damped regions, power-law threshold for spreading of puffs

Background: turbulence spreading and avalanching 00000000

Bistable model

References I

Barkley, D., Song, B., Mukund, V., Lemoult, G., Avila, M., and Hof, B. (2015).
The rise of fully turbulent flow.
Nature, 526:550–553.
Barnes, M., Parra, F. I., Highcock, E. G., Schekochihin, A. A., Cowley, S. C., and Roach, C. M. (2011).
 Turbulent transport in tokamak plasmas with rotational shear.
Phys. Rev. Lett., 106:175004.
Biskamp, D. and Walter, M. (1985).
Suppression of shear damping in drift wave turbulence.
Phys. Lett., 109A(1,2).
Carreras, B. A., Newman, D., Lynch, V. E., and Diamond, P. H. (1996).
A model realization of self-organized criticality for plasma confinement.
Physics of Plasmas, 3(8):2903–2911.
Carreras, B. A., Sidikman, K., Diamond, P. H., Terry, P. W., and Garcia, L. (1992).
Theory of shear flow effects on long?wavelength drift wave turbulence.
Physics of Fluids B: Plasma Physics, 4(10):3115–3131.
Drake, J. F., Zeiler, A., and Biskamp, D. (1995).
Nonlinear self-sustained drift-wave turbulence.
Phys. Rev. Lett., 75:4222–4225.
Cit L. and Samatha D. (1006)
GII, L. and Sornette, D. (1990).
Landau-Ginzburg theory of self-organized criticality.

Phys. Rev. Lett., 76:3991-3994.

Bistable model A teaser: machine learning model?

References II



Physics of Plasmas, 13(5):052306. Inagaki, S., Tokuzawa, T., Tamura, N., Itoh, S.-I., Kobavashi, T., Ida, K., Shimozuma, T., Kubo, S.,

Tanaka, K., Ido, T., et al. (2013).

How is turbulence intensity determined by macroscopic variables in a toroidal plasma? Nuclear Fusion, 53(11):113006.

Lesur, M. and Diamond, P. H. (2013).

Nonlinear instabilities driven by coherent phase-space structures. Phys. Rev. E, 87:031101.



Pomeau, Y. (1986).

Front motion, metastability and subcritical bifurcations in hydrodynamics. Physica D: Nonlinear Phenomena, 23(1):3 – 11.



Pomeau, Y. (2015).

The transition to turbulence in parallel flows: A personal view. Comptes Rendus Mecanique, 343:210-218.



van Wyk, F., Highcock, E. G., Schekochihin, A. A., Roach, C. M., Field, A. R., and Dorland, W. (2016). Transition to subcritical turbulence in a tokamak plasma. Journal of Plasma Physics, 82(6):905820609.

Background: turbulence spreading and avalanching 00000000

Bistable model

A teaser: machine learning model? 0000000

References III



Waltz, R. E. (1985).

Subcritical magnetohydrodynamic turbulence. Phys. Rev. Lett., 55:1098-1101.



Waltz, R. E. (2010).

Nonlinear subcritical magnetohydrodynamic beta limit. *Physics of Plasmas*, 17(7):072501.



T. S., and Manickam, J. (2006). Gyro-kinetic simulation of global turbulent transport properties in tokamak experiments. *Physics of Plasmas*, 13(9):092505.



Yi, S., Kwon, J. M., Diamond, P. H., and Hahm, T. S. (2014).

Effects of q-profile structure on turbulence spreading: A fluctuation intensity transport analysis. *Physics of Plasmas*, 21(9):092509. Bistable model

A teaser: machine learning model? •000000

A teaser: machine learning model?

Bistable model

A teaser: machine learning model?

Towards a complete model

- A realistic model should include coupling to zonal flow and pressure profile
- Start with Hasegawa-Wakatani:

$$\partial_t n + \{\phi, n\} = \alpha(\phi - n) + \text{diss.}$$

 $\partial_t \nabla^2_{\perp} \phi + \{\phi, \nabla^2_{\perp} \phi\} = \alpha(\phi - n) + \text{diss.}$

with $\alpha = -\eta {\partial_z}^2$ the adiabatic operator representing parallel electron response

• Take zonal averages:

$$\partial_t \langle n \rangle + \partial_x \langle \tilde{n} \tilde{v}_x \rangle = \text{diss.}$$

 $\partial_t \langle \zeta \rangle + \partial_x \langle \tilde{\zeta} \tilde{v}_x \rangle = \text{diss.}$
 $\partial_t \langle \varepsilon \rangle + \langle (\tilde{n} - \tilde{\zeta}) \tilde{v}_x \rangle \partial_x \langle n - \zeta \rangle + \partial_x \langle \varepsilon \tilde{v}_x \rangle = \text{diss.}$
where $\zeta = \nabla_{\perp}^2 \phi$, $\varepsilon = \frac{1}{2} (\tilde{n} - \tilde{\zeta})^2$

Learning mean field theory

- How to proceed? Need model for turbulent fluxes $\Gamma_q = \langle \tilde{q} \tilde{v}_x \rangle$ but hard to calculate
- Idea: use simulations to train machine learning model that maps mean profiles to local fluxes
- Here ML is just a form of nonparametric regression: no need to impose a model
- One approach: local model

$$\Gamma_q(x) = f(\partial_x n|_x, \partial_x^2 n|_x, \dots, \zeta|_x, \partial_x \zeta|_x, \dots, \varepsilon|_x, \partial_x \varepsilon|_x, \dots)$$

• Challenges: feature selection, noise suppression. Also is local model even valid?

A teaser: machine learning model?

Preliminary results: particle flux

• Training on \sim 20 simulations of 2D Hasegawa-Wakatani at $\alpha = 2$ and constraining the model with symmetries of HW, a simple neural network learns a reasonable model for the particle flux



Learned turbulent particle flux as function of density gradient at zero vorticity gradient (left) and vice versa (right.)

Bistable model

A teaser: machine learning model?

Preliminary results: particle flux

- Flux is approximately linear combination of terms prop. to $\partial_x \langle n \rangle$ and $\partial_x \langle \zeta \rangle$. First is obvious, latter less so!
- No clear dependence on shear itself



Figure: Dependence of particle flux on both density and vorticity gradients

A teaser: machine learning model?

Preliminary results: particle flux

 Results can be explained by simple quasilinear theory. However, must include effects of mean vorticity gradient on dispersion relation! Ignored in most studies

$$\partial_t \tilde{n} + V(x) \partial_y \tilde{n} + \partial_x \langle n \rangle \partial_y \tilde{\phi} = \alpha (\tilde{\phi} - \tilde{n})$$
$$\partial_t \tilde{\zeta} + V(x) \partial_y \tilde{\zeta} - \frac{V''(x)}{\psi} \partial_y \tilde{\phi} = \alpha (\tilde{\phi} - \tilde{n})$$
$$\omega = \frac{k_y}{1 + k^2} (\partial_x \langle n \rangle + \frac{V''}{\psi}) + k_y V$$

for $\alpha \gg 1$

• Real part of frequency proportional to PV gradient, not density gradient! This has lots of interesting consequences, just one of which is effect on particle flux

Preliminary results: particle flux

• Quasilinear flux in adiabatic limit can be computed as

$$\Gamma_{n} \simeq \frac{1}{\alpha} \sum_{\mathbf{k}} -\frac{k_{y}^{2}}{1+k^{2}} \left(k^{2}\kappa - V''\right) |\tilde{\phi}_{\mathbf{k}}|^{2}$$

- Good agreement with ML
- Vorticity gradient term can result in staircasing
- This project very much a work in progress (vorticity and enstrophy flux are harder), but this simple result shows its potential to elucidate new physics