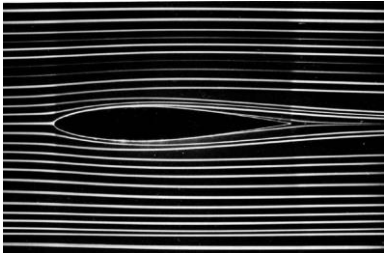


Lecture 2

Blackboard interlude 2

- Laminar-turbulent transition in convection
 - Linear stability analysis
 - Supercritical transition
- Laminar-turbulent transition in pipe
 - Linear stability analysis
 - Subcritical transition

What is turbulence?

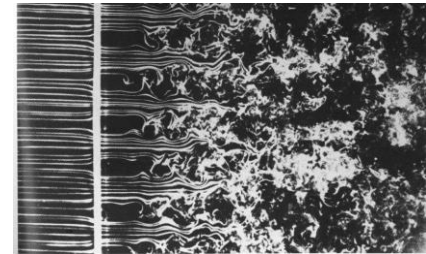


laminar flow

steady
predictable



laminar-turbulent transition critical behavior



fully-developed turbulence

fluctuating
unpredictable

Re < 1600

Re ~ 2000

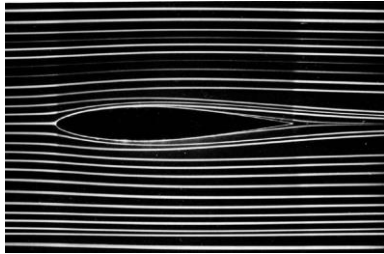
Re > 10⁵

$$\nabla \cdot \mathbf{u} = 0$$
$$\rho \left(\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} \right) = -\nabla p + \mu \nabla^2 \mathbf{u}$$

$$\text{Re} = \frac{UL}{\nu}$$

$$\nu = \mu/\rho$$

What is turbulence?



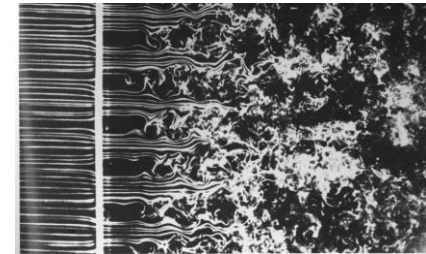
laminar flow

steady
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laminar-turbulent transition

critical behavior



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fluctuating
unpredictable

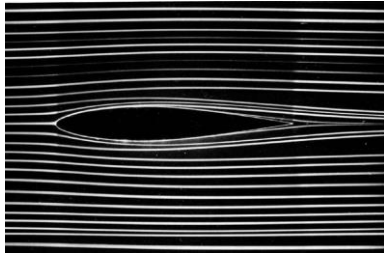
$Re < 1600$

$Re \sim 2000$

$Re > 10^5$

- Boundary conditions not periodic
- Turbulence generated by instabilities
 - Linear and long-wavelength
 - **Nonlinear and spatially-localized**
 - Not artificial noise
- Turbulence interacts with mean flow
 - Turbulence can generate emergent mean flows

What is turbulence?



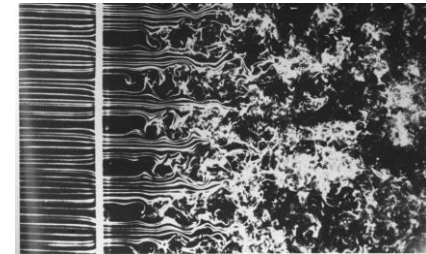
laminar flow

steady
predictable



laminar-turbulent transition

critical behavior



fully-developed turbulence

fluctuating
unpredictable

Re < 1600

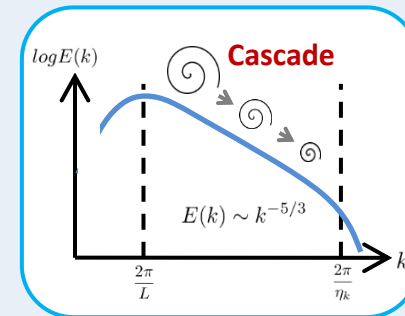
Re ~ 2000

Re > 10⁵

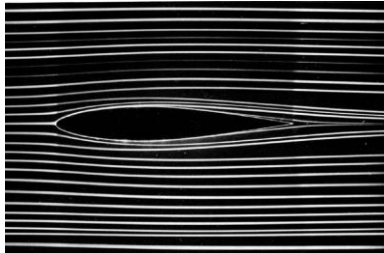
Fully-developed turbulence

- **Energy cascade: Non-equilibrium steady state**
Energy transfers step by step → power law
- **Dissipative anomaly:**
energy dissipation rate = 0 when viscosity = 0
energy dissipation rate ≠ 0 when viscosity → 0

$$\epsilon = \nu |\nabla u|^2 \neq 0 \text{ as } \nu \rightarrow 0$$



What is turbulence?



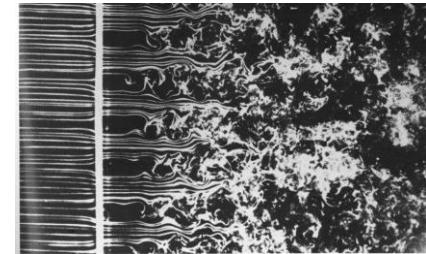
laminar flow

steady
predictable



laminar-turbulent transition

critical behavior



fully-developed turbulence

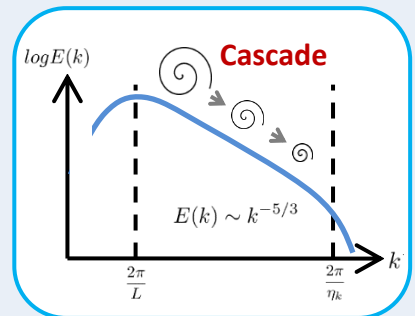
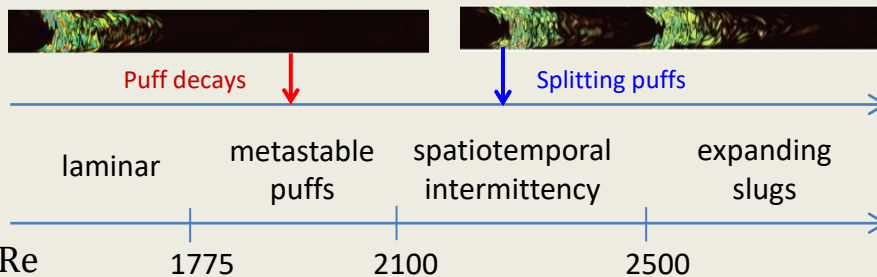
fluctuating
unpredictable

$Re < 1600$

$Re \sim 2000$

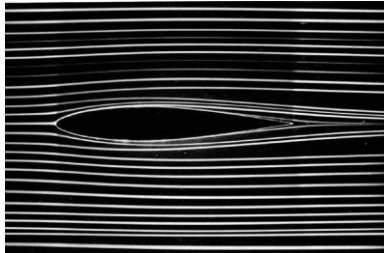
$Re > 10^5$

Transitional pipe flow



$$\epsilon = \nu |\nabla u|^2 \neq 0 \text{ as } \nu \rightarrow 0$$

What is turbulence?



laminar flow

steady
predictable

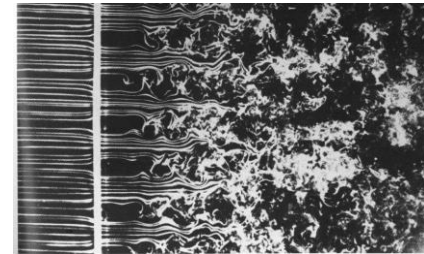
$Re < 1600$



laminar-turbulent transition

critical behavior

$Re \sim 2000$



fully-developed turbulence

fluctuating
unpredictable

$Re > 10^5$

Transitional pipe flow



Puff decays

Splitting puffs

laminar

metastable
puffs

spatiotemporal
intermittency

expanding
slugs

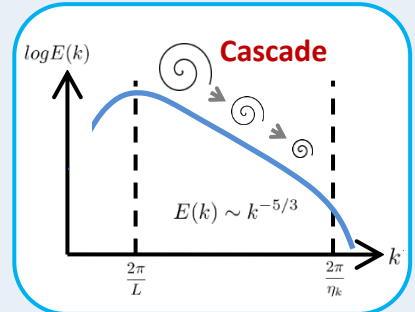
1775

2100

2500

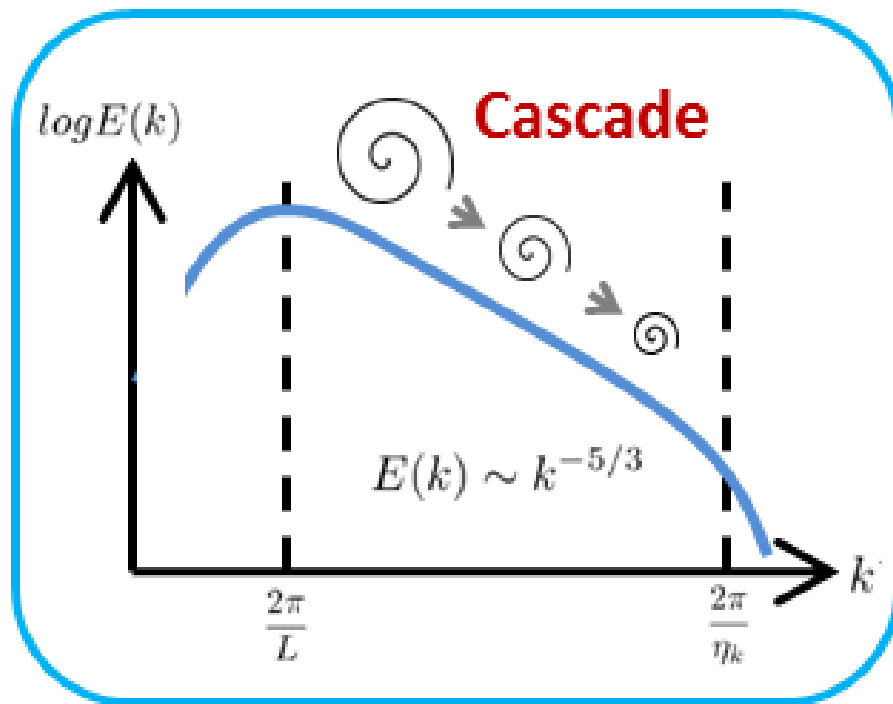
Re

Turbulence & mean flows



$$\epsilon = \nu |\nabla u|^2 \neq 0 \text{ as } \nu \rightarrow 0$$

Energy cascade



- Eddies spin off other eddies in a Hamiltonian process.
 - Does not involve friction!
 - Hypothesis due to Richardson, Kolmogorov, ...
- Implication: viscosity will not enter into the equations

Kolmogorov's similarity hypotheses

- Dimensional analysis
- $E(k) = E(k, \varepsilon, \nu, L)$
 - Kolmogorov length $\eta_K = (\nu^3/\varepsilon)^{1/4}$
- $E(k) = (\nu^2/\eta_K) F(k\eta_K, kL)$
 - Complete similarity as $kL \rightarrow \infty$
- $E(k) = (\nu^2/\eta_K) F(k\eta_K, \infty)$
 - Complete similarity as $k\eta_K \rightarrow 0$
- $E(k) = F(0, \infty) \varepsilon^{2/3} k^{-5/3}$



Kolmogorov's similarity hypotheses

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Kolmogorov's similarity hypotheses

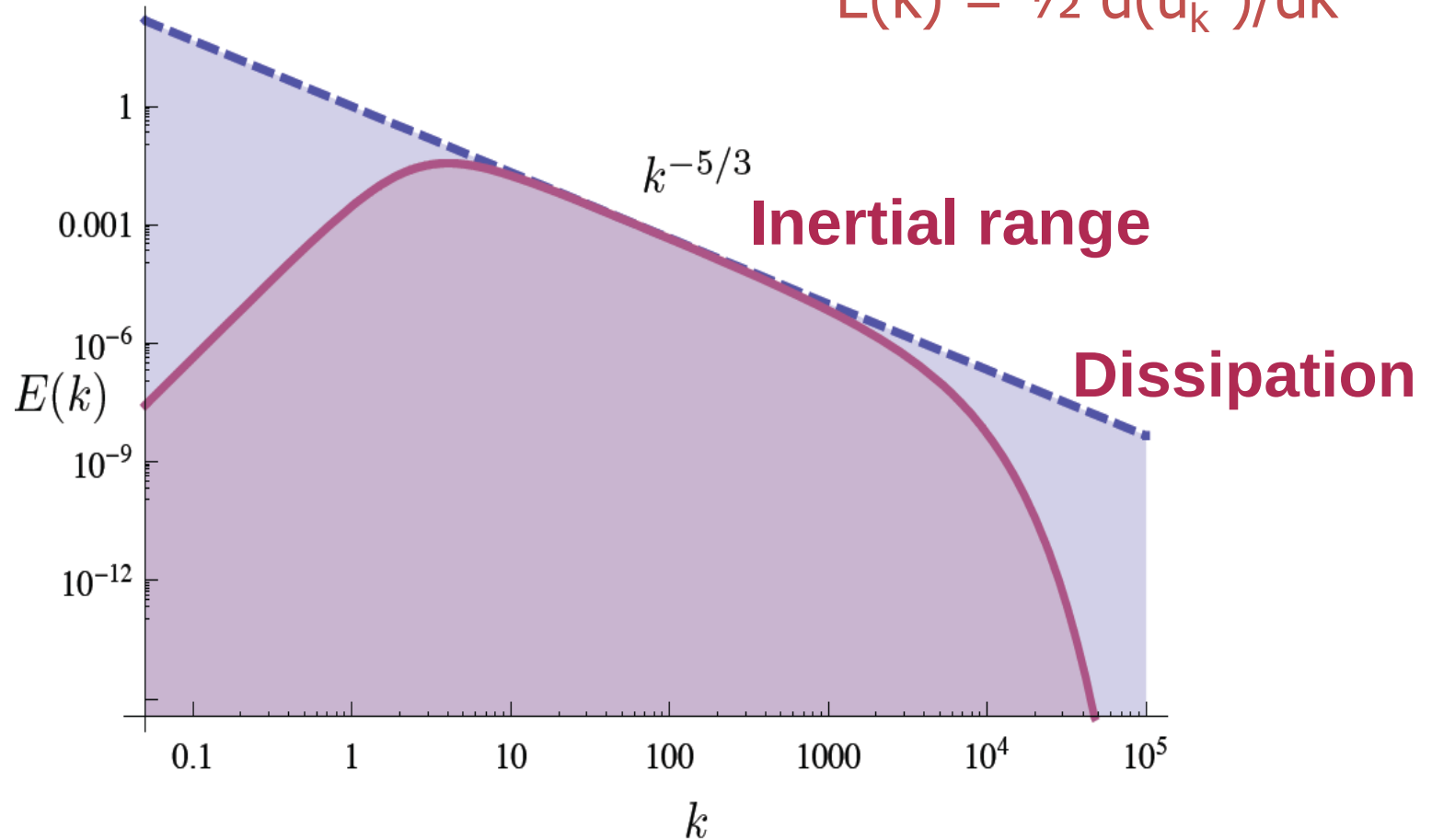
- Dimensional analysis
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 - Complete similarity as $k\eta_K \rightarrow 0$
- $E(k) = F(0, \infty) \varepsilon^{2/3} k^{-5/3}$



The energy spectrum

Integral scale

$$E(k) = \frac{1}{2} \frac{d(u_k^2)}{dk}$$



Kolmogorov's similarity hypotheses

- Dimensional analysis
- $E(k) = E(k, \varepsilon, \nu, L)$
 - Kolmogorov length $\eta_K = (\nu^3/\varepsilon)^{1/4}$
- $E(k) = (\nu^2/\eta_K) F(k\eta_K, kL)$
 - Incomplete similarity as $kL \rightarrow \infty$
- $E(k) = (\nu^2/\eta_K) F_1(k\eta_K) (kL)^\eta$
 - Complete similarity as $k\eta_K \rightarrow 0$
- $E(k) = F_1(0) \varepsilon^{2/3} k^{-5/3} (kL)^\eta$

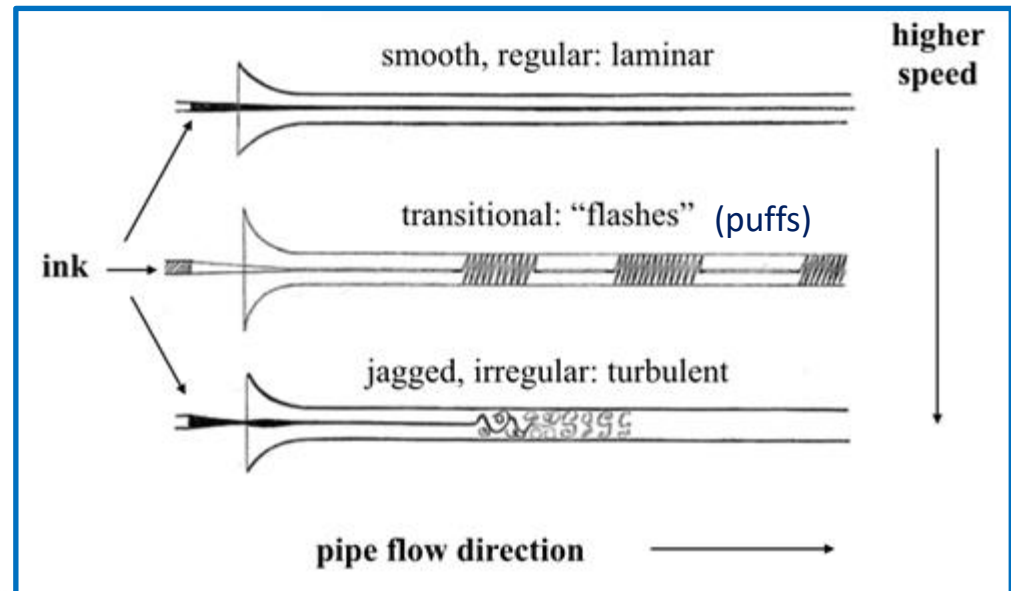
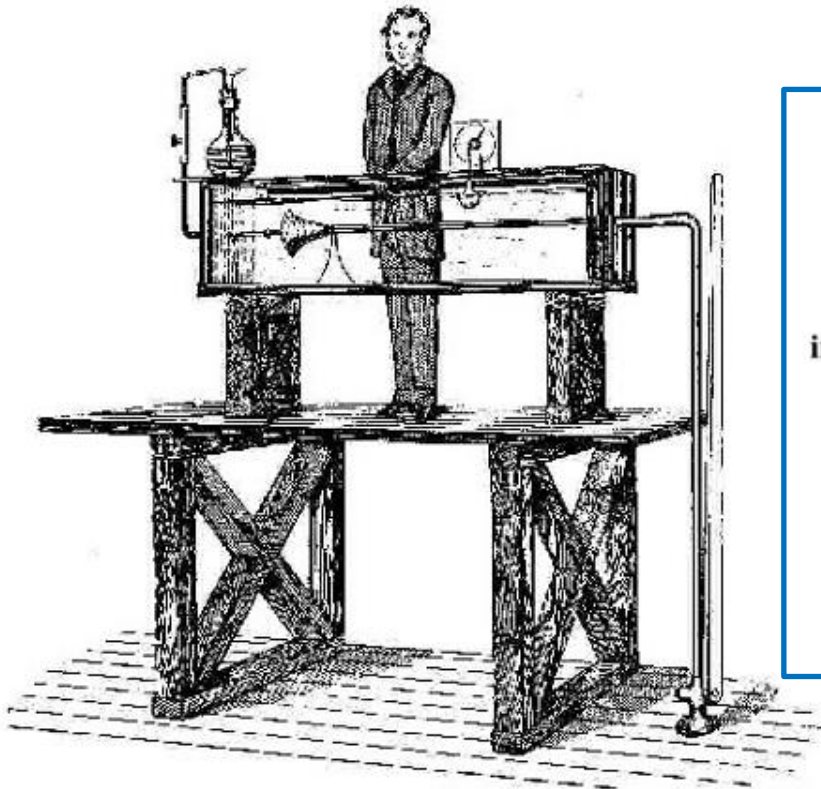
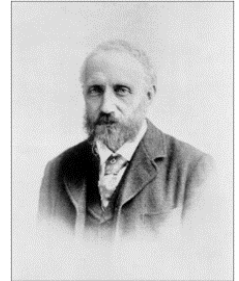
Intermittency exponent



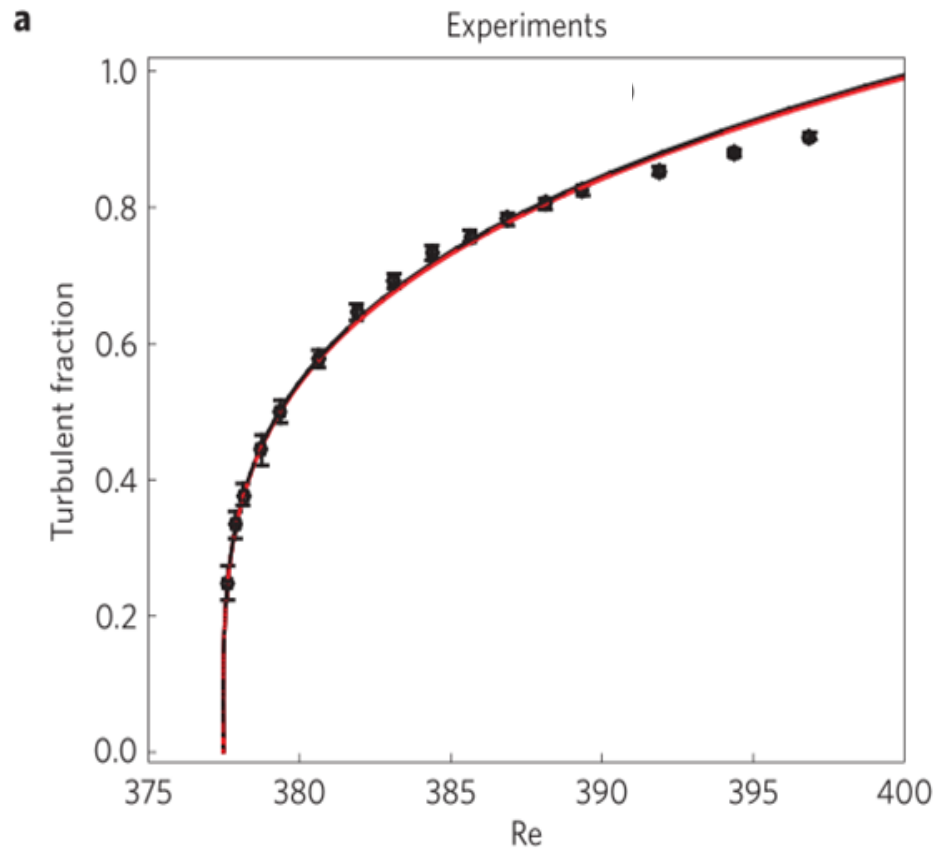
2. Puffs

Transitional turbulence in pipe flow: puffs

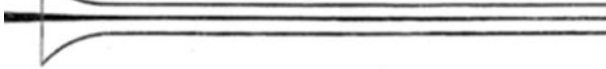
- Reynolds' original pipe turbulence (1883) reports on the transition
- Defined Reynolds' number $Re = \frac{UL}{\nu}$



How much turbulence is in the pipe?



smooth, regular: laminar



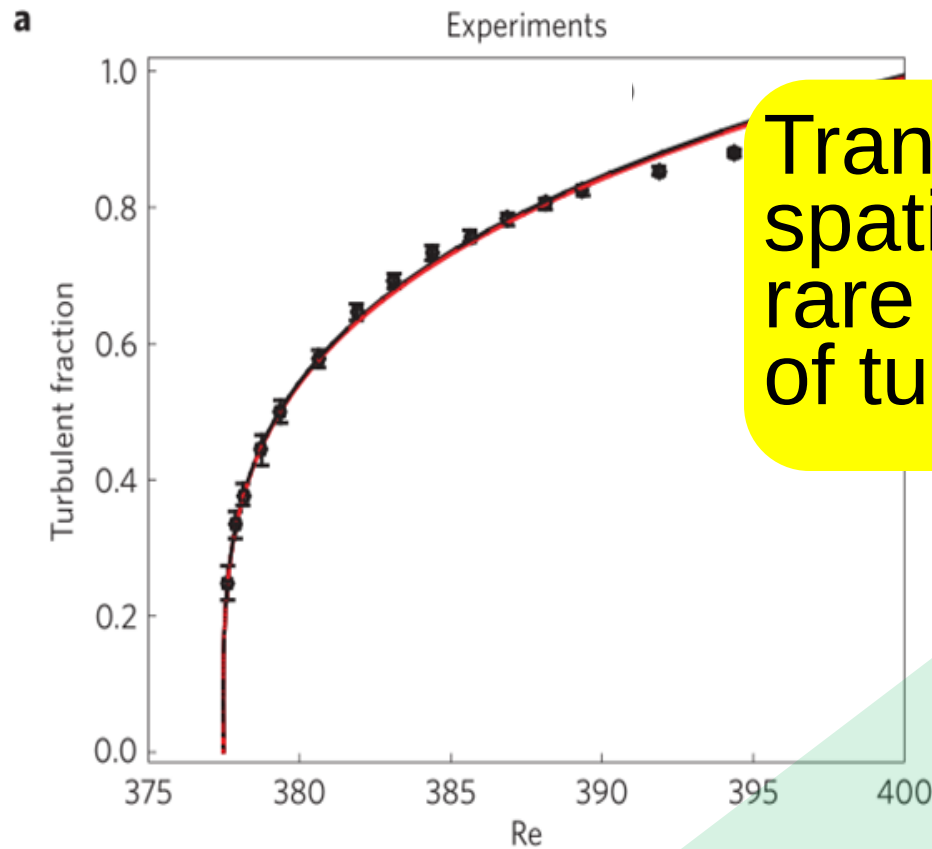
transitional: “flashes”



jagged, irregular: turbulent

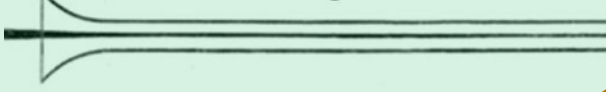


How much turbulence is in the pipe?

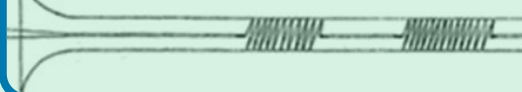


Transition not spatially uniform: rare sharp bursts of turbulence

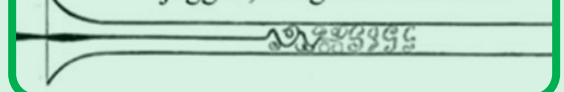
smooth, regular: laminar



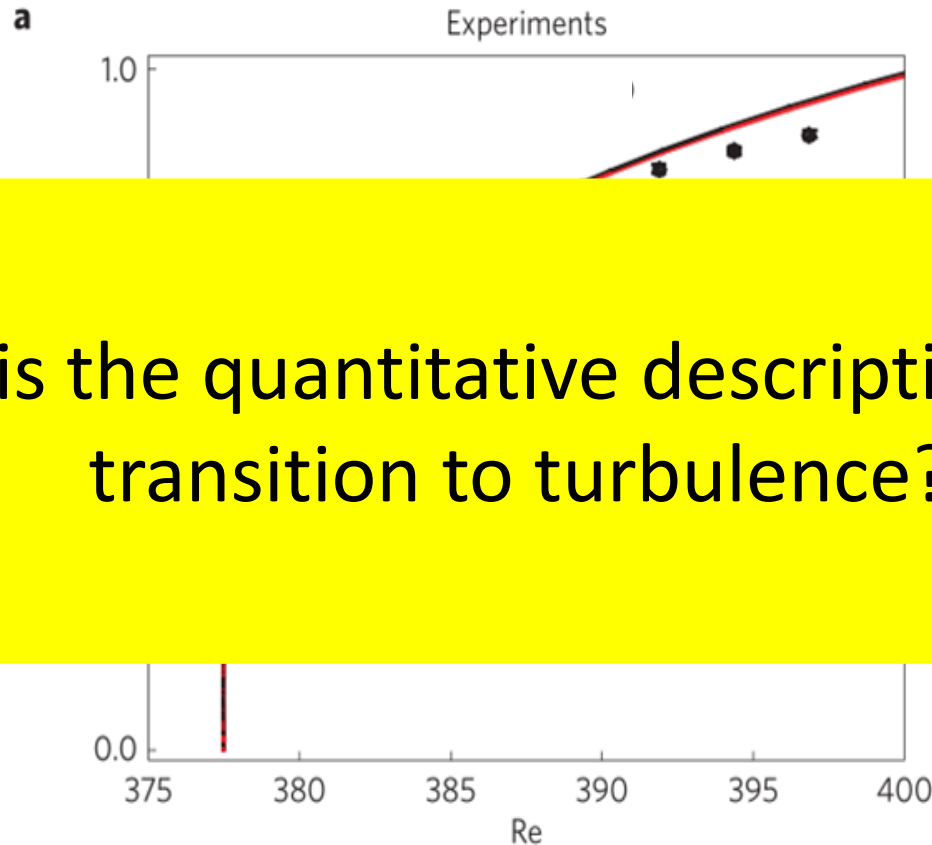
transitional: “flashes”



jagged, irregular: turbulent

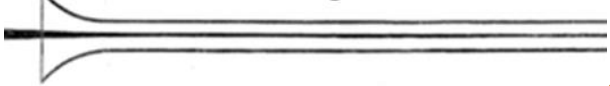


How much turbulence is in the pipe?



What is the quantitative description of the transition to turbulence?

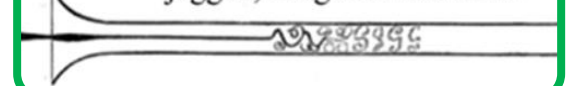
smooth, regular: laminar



transitional: "flashes"

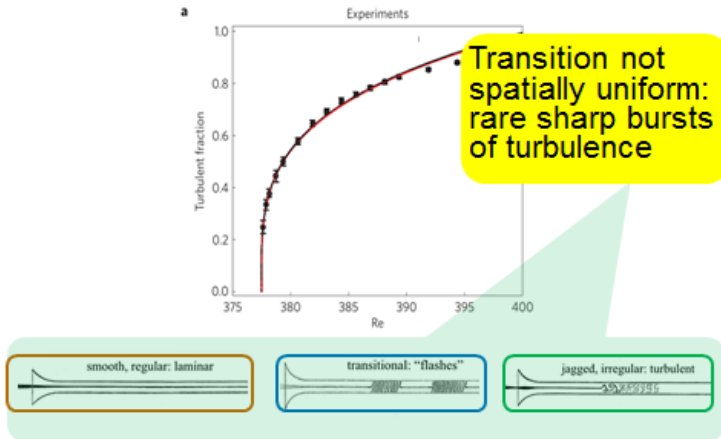


jagged, irregular: turbulent



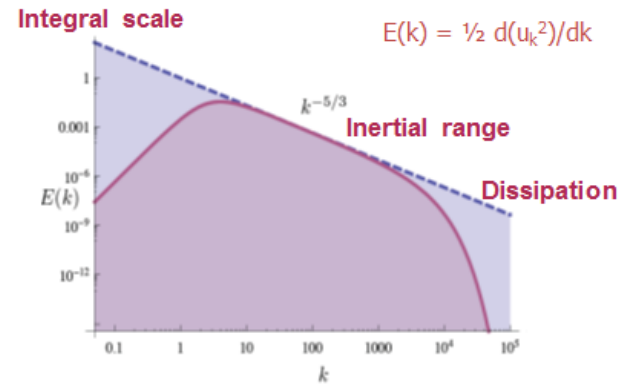
Turbulence & Phase Transitions

How much turbulence is in the pipe?



- Turbulent fraction as a function of Re is reminiscent of a continuous phase transition

The energy spectrum



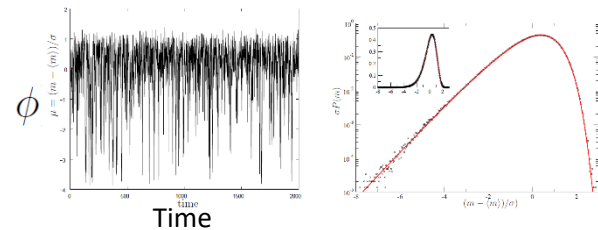
- Strong power-law correlated fluctuations is reminiscent of a continuous phase transition

**Why is fully-developed
turbulence hard?**

Why is turbulence unsolved?

Why are phase transitions hard?

- **Strong interactions + fluctuations** in the order parameter ϕ :
 - Very non-Gaussian
 - Intermittency



- **No usable small parameter!**

Landau free energy

$$\mathcal{H} = \int d^d \mathbf{r} \left[\frac{1}{2} \gamma (\nabla \phi)^2 + \frac{1}{2} r_0 \phi^2 + \frac{1}{4} u_0 \phi^4 \right]$$

$$t = (T - T_c) / T_c \quad r_0 \sim t$$

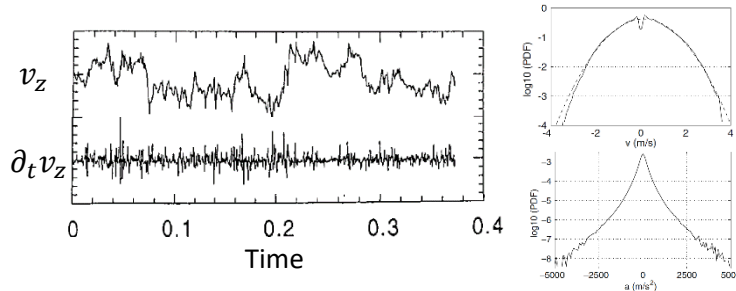
$$\text{rescaling} \quad \bar{u}_0 \sim u_0 t^{(d-4)/2} \begin{cases} d < 4 : \bar{u}_0 \rightarrow \infty \\ d > 4 : \bar{u}_0 \rightarrow 0 \end{cases}$$

The coefficient of the **interaction** becomes relatively large at **phase transition** $t \rightarrow 0$

Why is turbulence unsolved?

Why is turbulence hard?

- **Strong interactions + fluctuations** in the velocity derivatives $\partial_i v_j$:
 - Very non-Gaussian
 - Intermittency: intervals of weak fluctuations interspersed with bursts of strong fluctuations



- **No usable small parameter!**

The Navier-Stokes equation

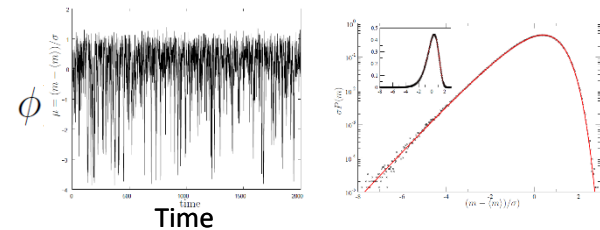
$$\partial_t \mathbf{v} + (\mathbf{v} \cdot \nabla) \mathbf{v} = -\frac{1}{\rho} \nabla p + \nu \nabla^2 \mathbf{v} + \mathbf{f}$$

Turbulence: $\text{Re} \rightarrow \infty$, $\nu \rightarrow 0$

The coefficient of the **nonlinearity** becomes relatively large as the **viscosity** $\nu \rightarrow 0$

Why are phase transitions hard?

- **Strong interactions + fluctuations** in the order parameter ϕ :
 - Very non-Gaussian
 - Intermittency



- **No usable small parameter!**

Landau free energy

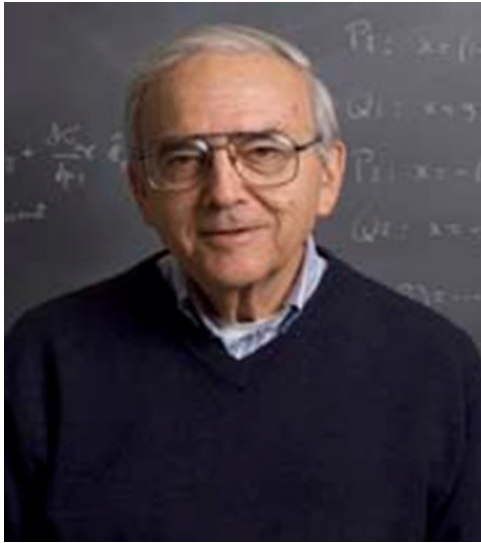
$$\mathcal{H} = \int d^d \mathbf{r} \left[\frac{1}{2} \gamma (\nabla \phi)^2 + \frac{1}{2} r_0 \phi^2 + \frac{1}{4} u_0 \phi^4 \right]$$

$$t = (T - T_c)/T_c \quad r_0 \sim t$$

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The coefficient of the **interaction** becomes relatively large at **phase transition** $t \rightarrow 0$

How was critical phenomena solved?



Ben Widom discovered “data collapse” (1965)



Leo Kadanoff explained data collapse, with scaling concepts (1966)



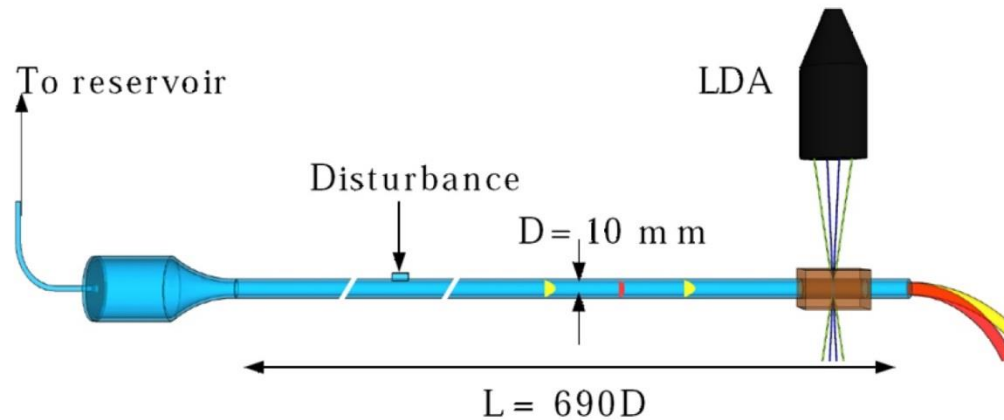
Ken Wilson developed the RG based on Kadanoff’s scaling ideas (1970)

- **Common features**
 - Strong fluctuations
 - Power law correlations
- **Can we solve turbulence by following critical phenomena?**
- **Does turbulence exhibit critical phenomena at its onset?**

Transition to turbulence

Precision measurement of turbulent transition

Q: will a turbulent puff survive to the end of the pipe?



Many repetitions → **Survival probability = $P(\text{Re}, t)$**

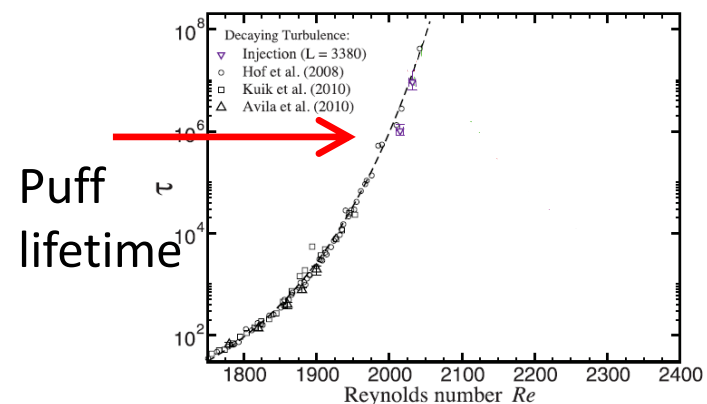
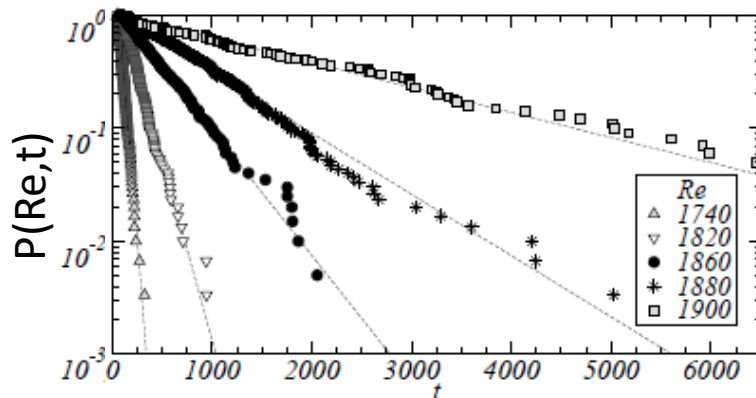
Pipe flow turbulence



Decaying single puff



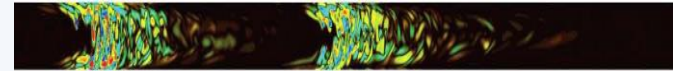
Survival probability $P(Re, t) = e^{-\frac{t-t_0}{\tau(Re)}}$



Pipe flow turbulence



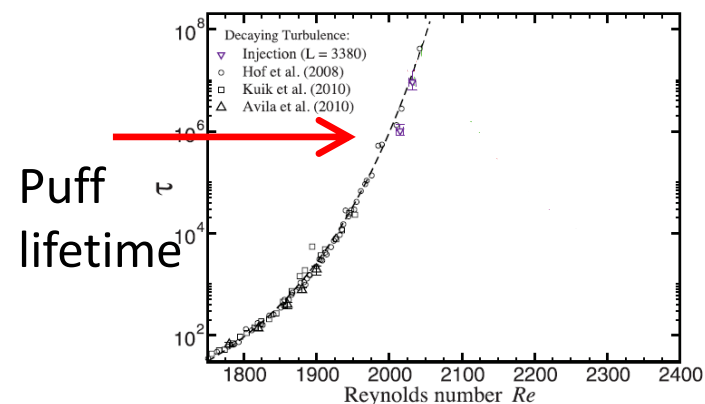
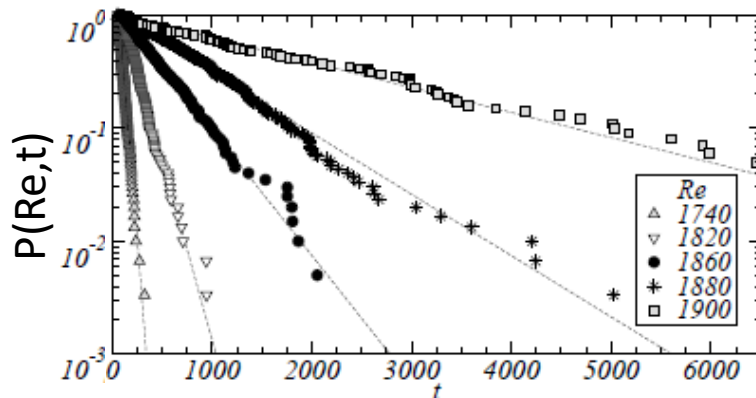
Decaying single puff



Splitting puffs



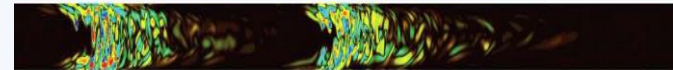
Survival probability $P(Re, t) = e^{-\frac{t-t_0}{\tau(Re)}}$



Pipe flow turbulence



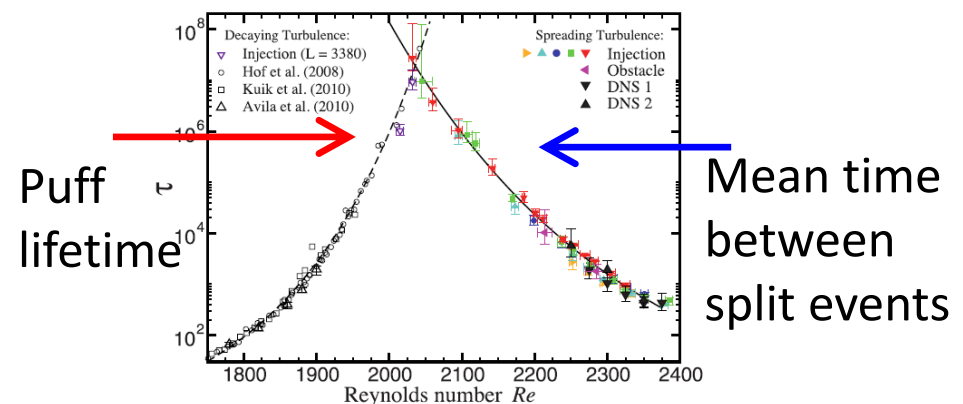
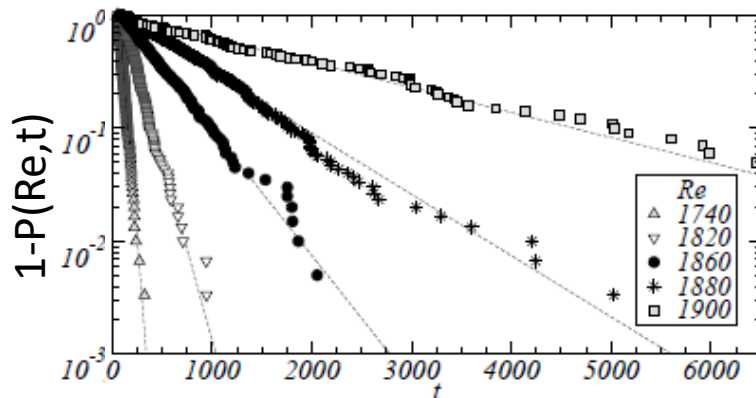
Decaying single puff



Splitting puffs



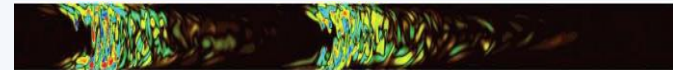
Splitting probability $1 - P(\text{Re}, t) = e^{-\frac{t-t_0}{\tau(\text{Re})}}$



Pipe flow turbulence



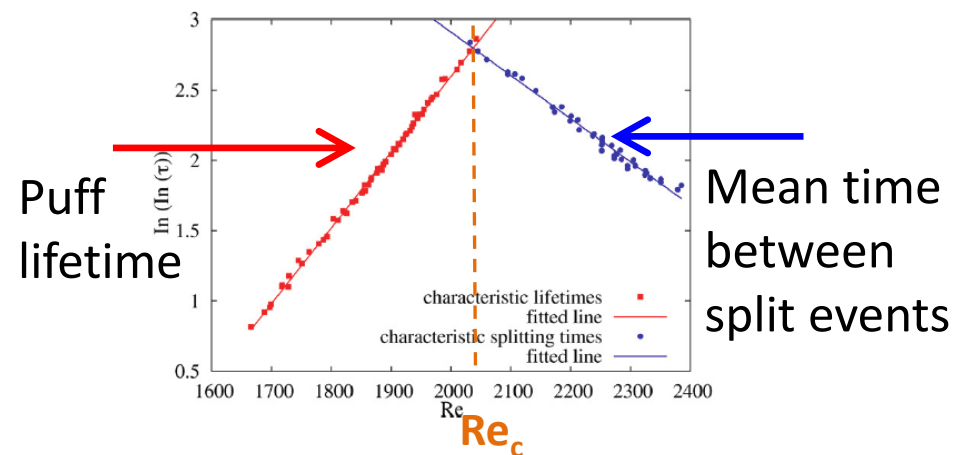
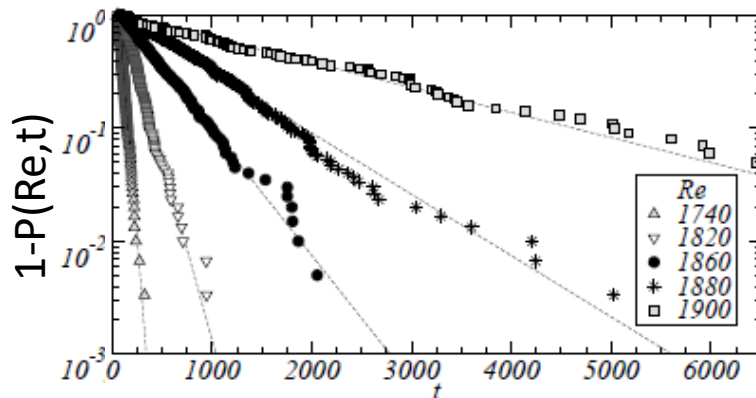
Decaying single puff



Splitting puffs



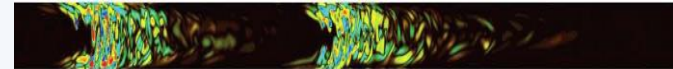
Splitting probability $1 - P(\text{Re}, t) = e^{-\frac{t-t_0}{\tau(\text{Re})}}$



Pipe flow turbulence



Decaying single puff



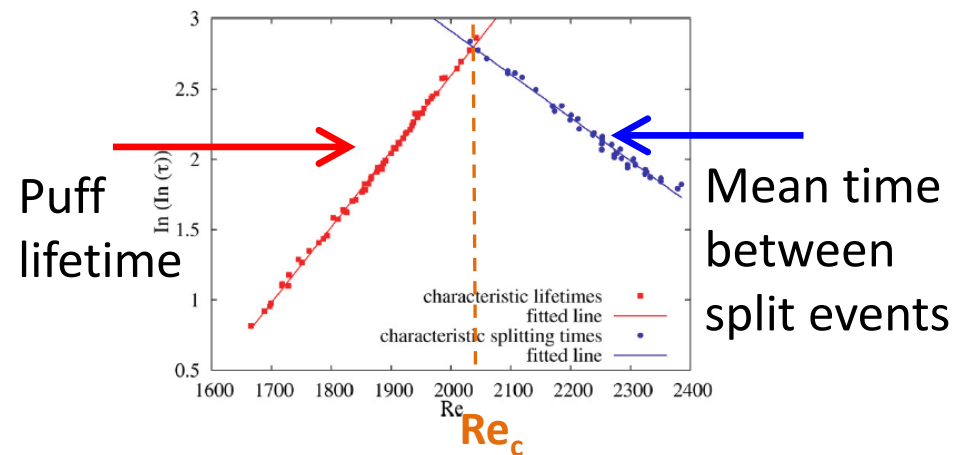
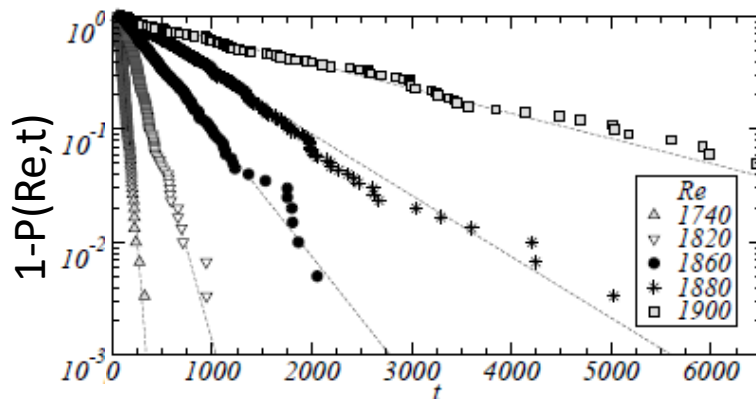
Splitting puffs



Super-exponential scaling:

$$\frac{\tau}{\tau_0} \sim \exp(\exp Re)$$

Splitting p



Theory for the laminar-turbulent transition in pipe flow

Logic of modeling phase transitions

Magnets

Electronic structure



Ising model

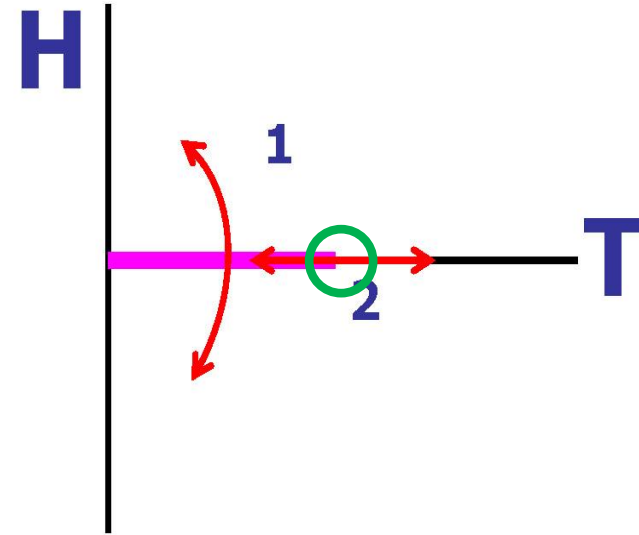
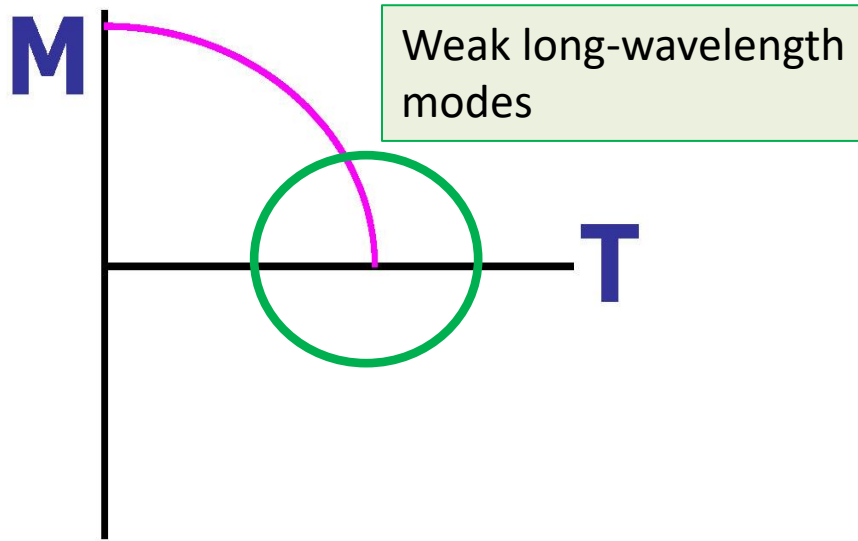


Landau theory



Renormalization group
universality class
(Ising universality class)

Critical phenomena in magnets



$$M \sim M_0[|T - T_c|/T_c]^\beta \text{ for } H = 0 \text{ as } T \rightarrow T_c$$

$$\text{Critical isotherm: } M \sim H^{1/\delta} \text{ for } T = T_c$$

- Widom (1963) pointed out that both these results followed from a *similarity formula*:

$$M(t, h) = |t|^\beta f_M(h/t^\Delta)$$

where $t \equiv (T - T_c)/T_c$ for some choice of exponent Δ and scaling function $f_M(x)$

Critical phenomena in magnets

$$M(t, h) = |t|^\beta f_M(h/t^\Delta)$$

where $t \equiv (T - T_c)/T_c$ for some choice of exponent Δ and scaling function $f_M(x)$

- **To determine the properties of the scaling function and unknown exponent, we require:**
 - $f_M(z) = \text{const.}$ for $z = 0$
 - This gives $M \sim M_0 [|T - T_c|/T_c]^\beta$ for $T < T_c$
 - For large values of z , i.e. non-zero h , and $t \rightarrow 0$, t dependence must cancel out so that $M \sim H^{1/\delta}$

Thus $f_M(z) \sim z^{1/\delta}, z \rightarrow \infty$.

Calculate Δ : t dependence will only cancel out if $\beta - \Delta/\delta = 0$

$$M = |t|^\beta f_M(h/|t|^{\beta\delta})$$

- **This data collapse formula connects the scaling of correlations with the thermodynamics of the critical point**

Universality at a critical point

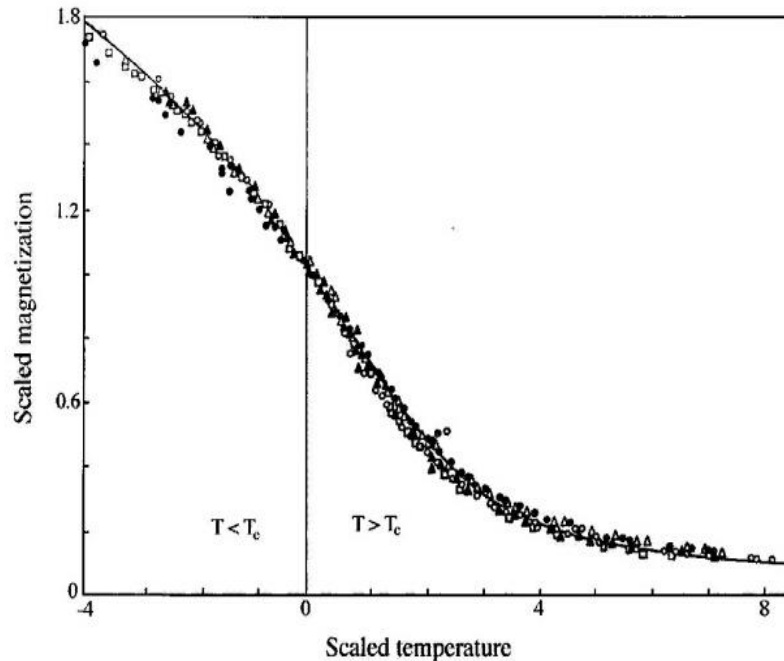


FIG. 1. Experimental MHT data on five different magnetic materials plotted in scaled form. The five materials are CrBr_3 , EuO , Ni , YIG , and Pd_3Fe . None of these materials is an idealized ferromagnet: CrBr_3 has considerable lattice anisotropy, EuO has significant second-neighbor interactions. Ni is an itinerant-electron ferromagnet, YIG is a ferrimagnet, and Pd_3Fe is a ferromagnetic alloy. Nonetheless, the data for all materials collapse onto a single scaling function, which is that calculated for the $d=3$ Heisenberg model [after Milošević and Stanley (1976)].

Stanley (1999)

- Magnetization M of a material depends on temperature T and applied field H
 - $M(H,T)$ ostensibly a function of two variables
- Plotted in appropriate scaling variables get ONE universal curve
- Scaling variables involve critical exponents

A theoretical physics success

A model ...

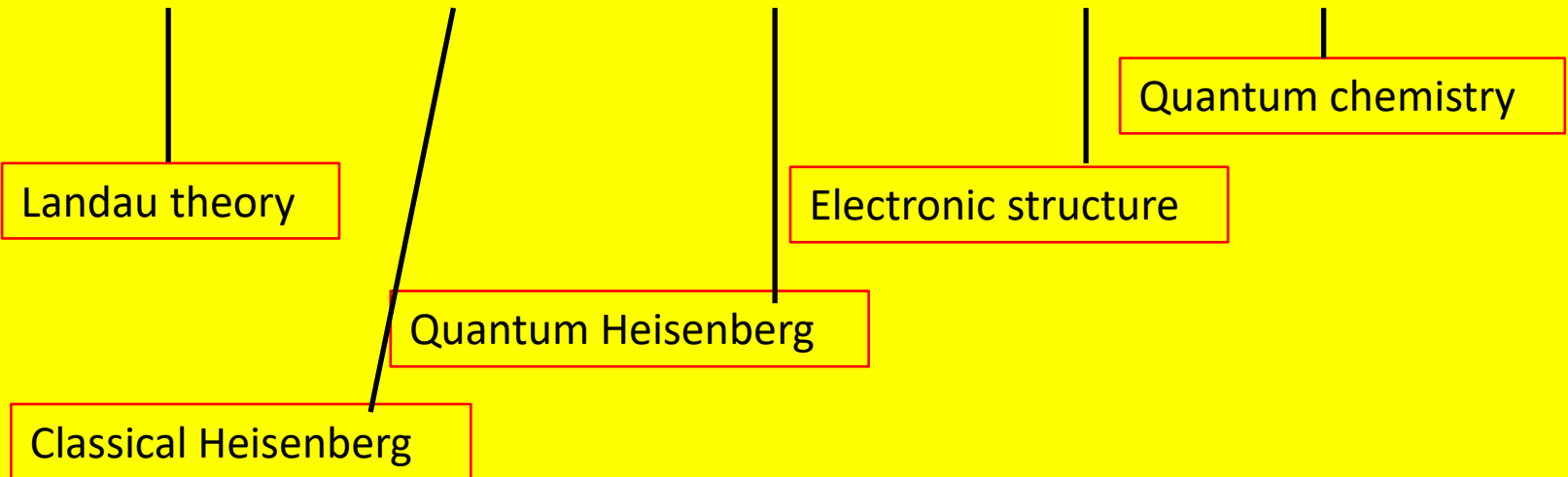
Gives a precise prediction in agreement with experiment!

materials collapse onto a single scaling function, which is that calculated for the $d=3$ Heisenberg model [after Milošević and Stanley (1976)].

Stanley (1999)

A theoretical physics success

A model of a model of a model of a model of a model



Gives a precise prediction in agreement with experiment!

Non-systematic approximations

materials collapse onto a single scaling function, which is that calculated for the $d=3$ Heisenberg model [after Milošević and Stanley (1976)].

Stanley (1999)

Logic of modeling phase transitions

Magnets

Electronic structure



Ising model



Landau theory



Renormalization group
universality class
(Ising universality class)

Turbulence

Kinetic theory



Navier-Stokes eqn



?



?

Logic of modeling phase transitions

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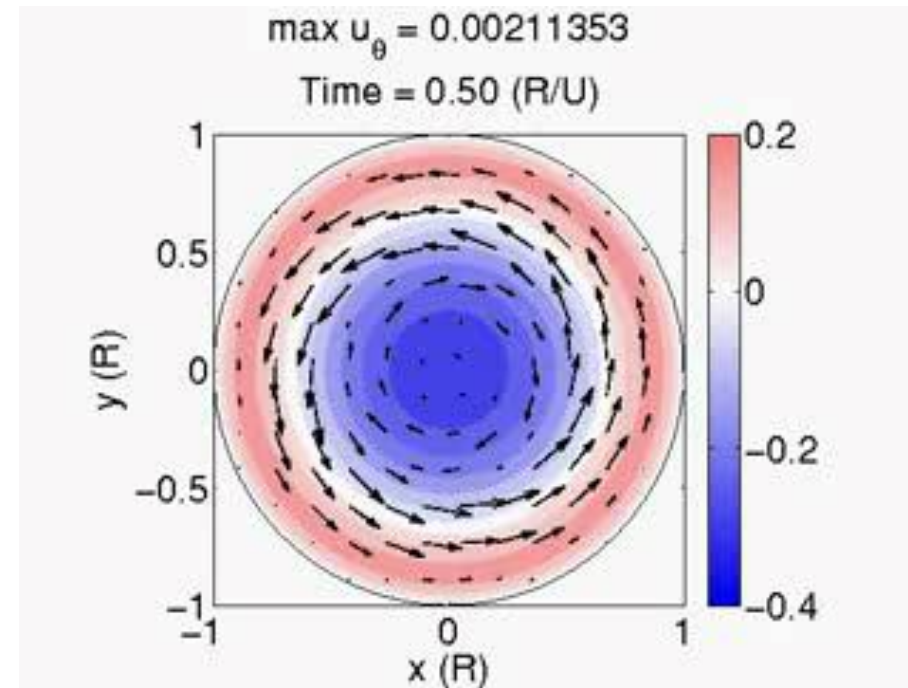
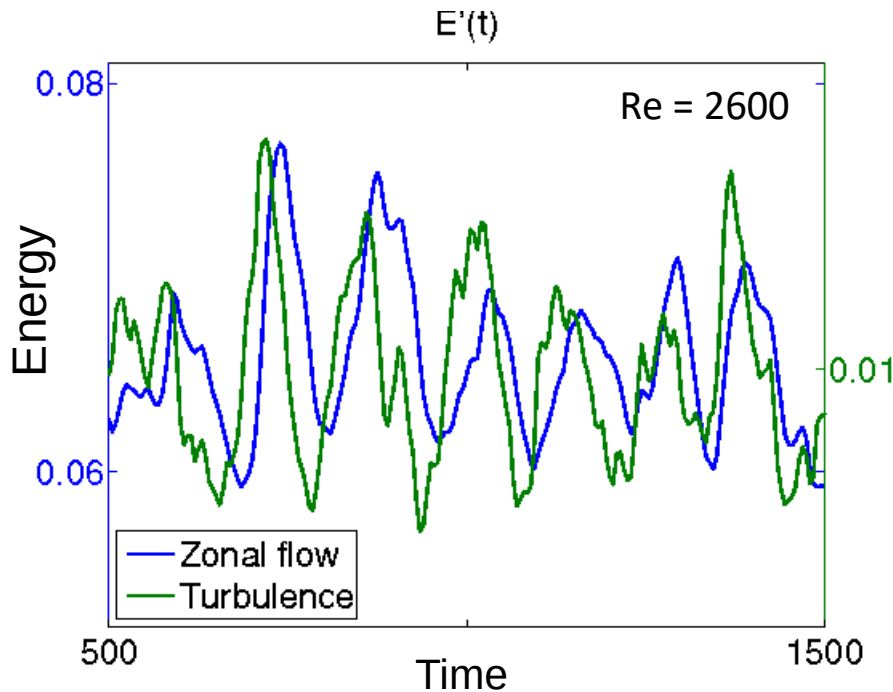
Identification of long-wavelength collective modes at the laminar-turbulent transition

To avoid uncontrolled approximations, we use **direct numerical simulation** of the **Navier-Stokes equations**

DNS of 3D Navier-Stokes equations

- Use pseudospectral method to deal with both derivatives and nonlinear terms in streamwise direction and azimuthal directions
- Use Chebyshev polynomials to resolve large gradients in the boundary layer, in the radial direction
- 60 grid points in the radial (r) direction,
- 32 Fourier modes in the azimuthal (θ) direction and 128 modes in the axial (z) direction
- The spatial resolutions were chosen such that the resolvable power spectra span over six orders of magnitude.
- The pipe length L is 10 times its diameter D , with periodic boundary conditions in the z direction

Predator-prey oscillations in pipe flow



Decomposition into large & small scales

- Use pseudospectral method to deal with both derivatives and nonlinear terms in streamwise direction and azimuthal directions
- Use Chebyshev polynomials to resolve large gradients in the boundary layer, in the radial direction
- Zonal flow velocity $\mathbf{u}_{ZF} \equiv (\bar{u}_z, \bar{u}_\theta, \bar{u}_r)$, where $\bar{u}_r = 0$
- Zonal flow energy is defined as

$$E_{ZF}(t) \equiv (1/2) \int |\tilde{\mathbf{u}}(0, 0, r)|^2 dV \quad \text{with} \quad \tilde{\mathbf{u}}_{ZF} \equiv \tilde{\mathbf{u}}(k=0, m=0, r)$$

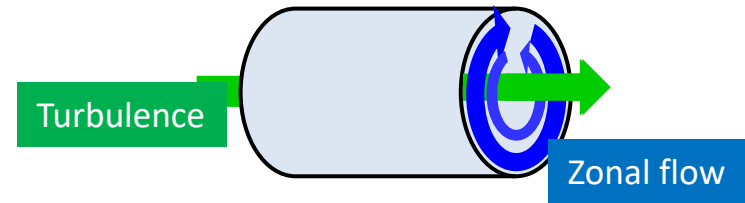
- Turbulence energy is defined as

$$E_T(t) \equiv (1/2) \sum_{|k| \geq 1, |m| \geq 1} \int |\tilde{\mathbf{u}}(k, m, r)|^2 dV$$

What drives the zonal flow?

- Interaction in two fluid model

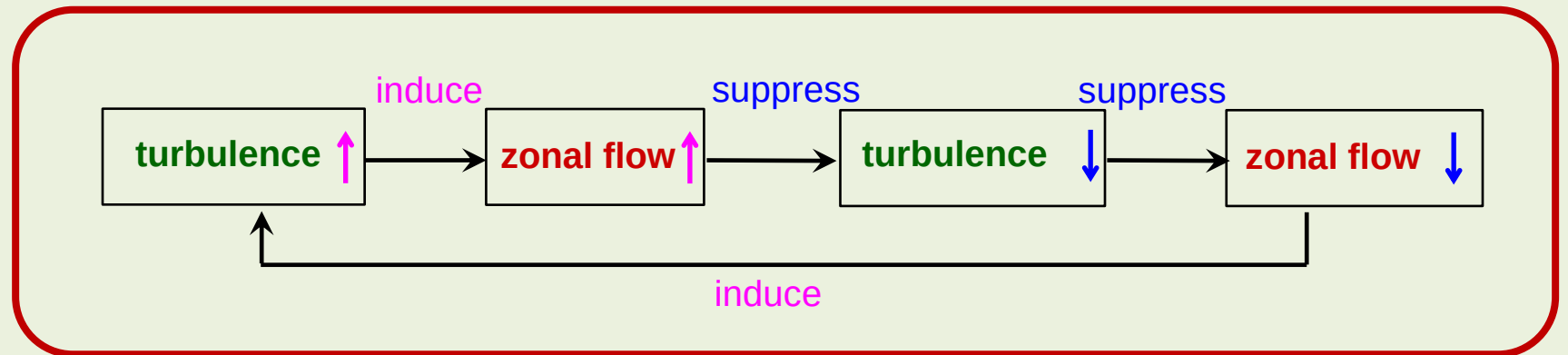
- Turbulence, small-scale ($k>0$)
- Zonal flow, large-scale ($k=0, m=0$): induced by turbulence and creates shear to suppress turbulence



- 1) Anisotropy of turbulence creates Reynolds stress which generates the mean velocity in azimuthal direction

$$\partial_t \langle v_\theta \rangle = -\partial_r \langle (\tilde{v}_\theta \cdot \tilde{v}_r) \rangle - \mu \langle v_\theta \rangle$$

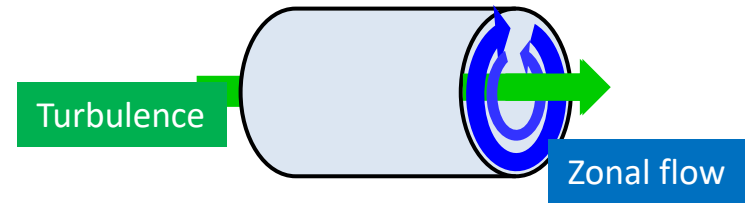
- 2) Mean azimuthal velocity decreases the anisotropy of turbulence and thus suppress turbulence



What drives the zonal flow?

- Interaction in two fluid model

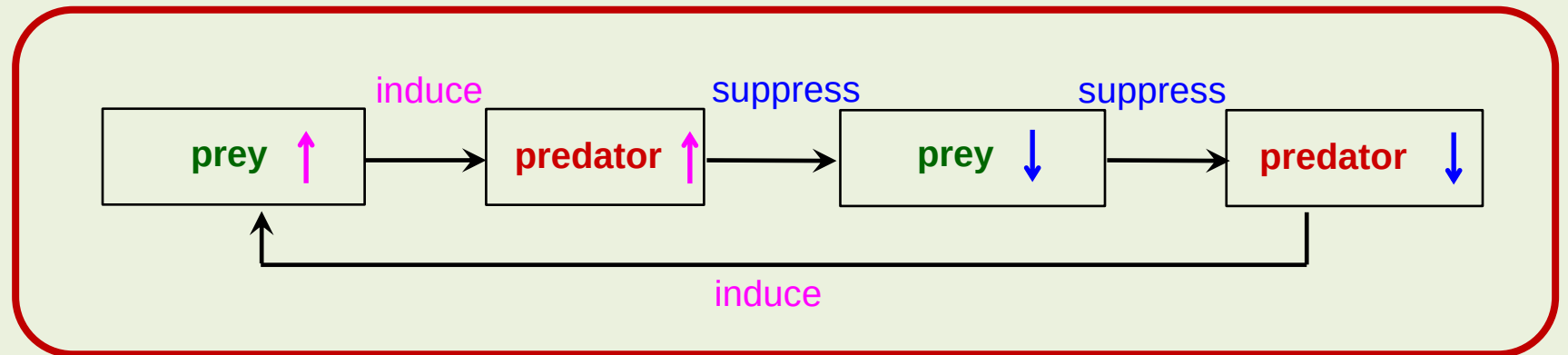
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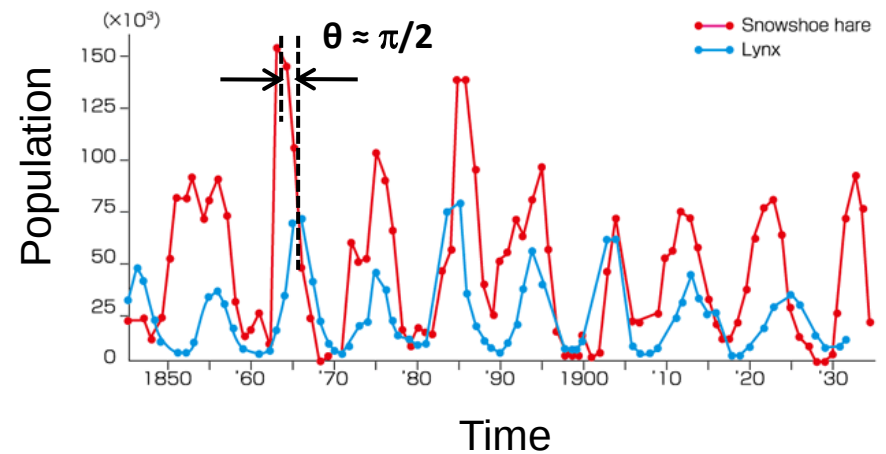
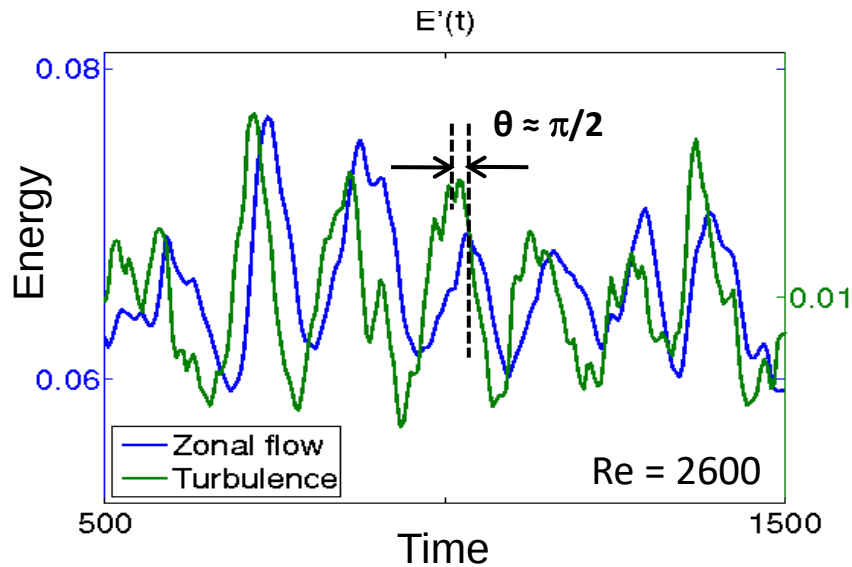
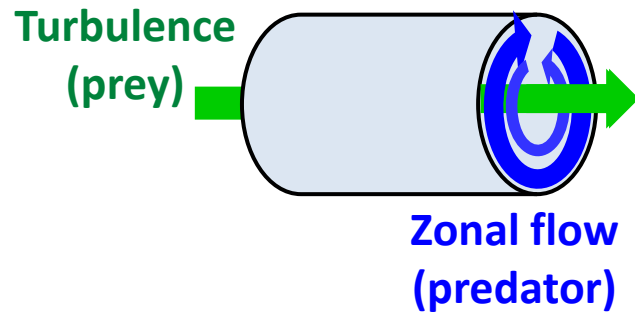
- 2) Mean azimuthal velocity decreases the anisotropy of turbulence and thus suppress turbulence



Pipe flow near transition to turbulence



Predator-prey ecosystem



Blackboard interlude 3

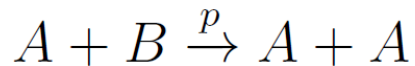
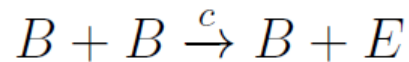
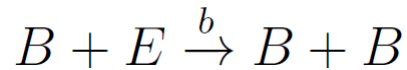
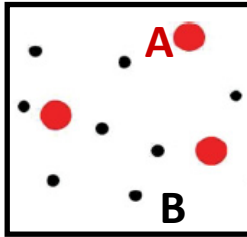
- Predator-prey equations description of cyclical behavior is strange, because a deterministic description based on the physics does not work! Let's see this ...
- Intrinsically stochastic description is required

Stochastic model of predator-prey dynamics

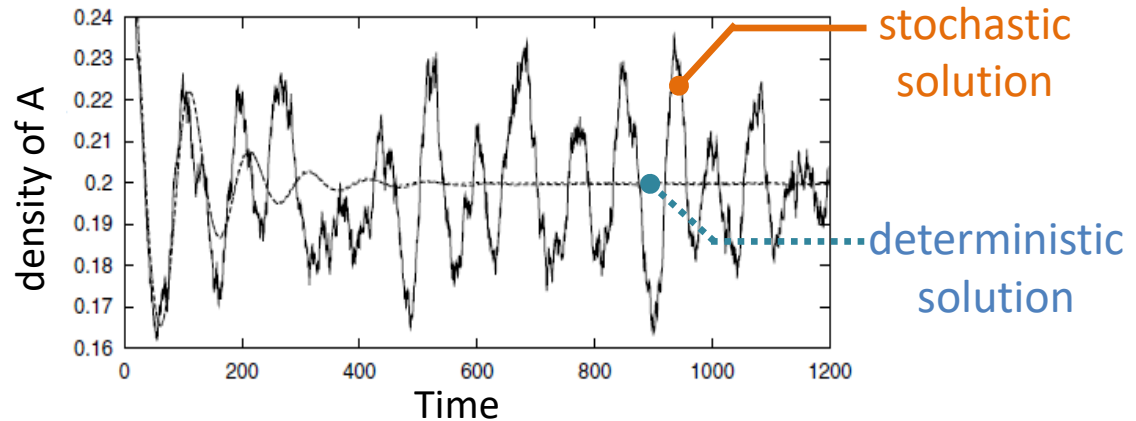
- Stochastic individual-level model

fluctuations in number → **demographic stochasticity** that induces **quasi-cycles**

A = predator
B = prey
E = food or
available space



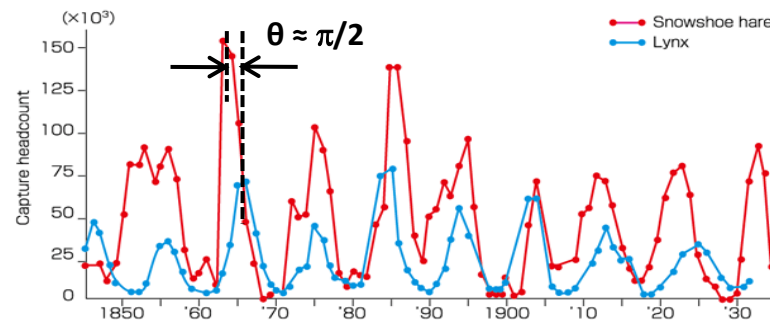
Stochastic individual-level simulation



Normal population cycles in a predator-prey system



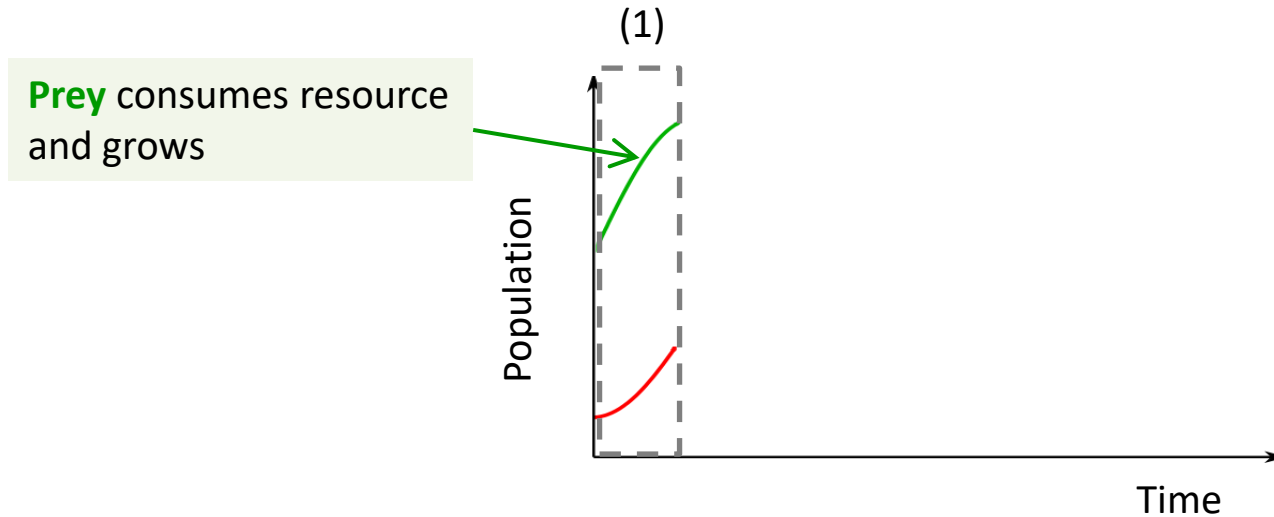
$\pi/2$ phase shift between prey and predator population



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**Persistent oscillations
+
Fluctuations**

Cartoon picture for normal cycles ($\pi/2$ phase shift)

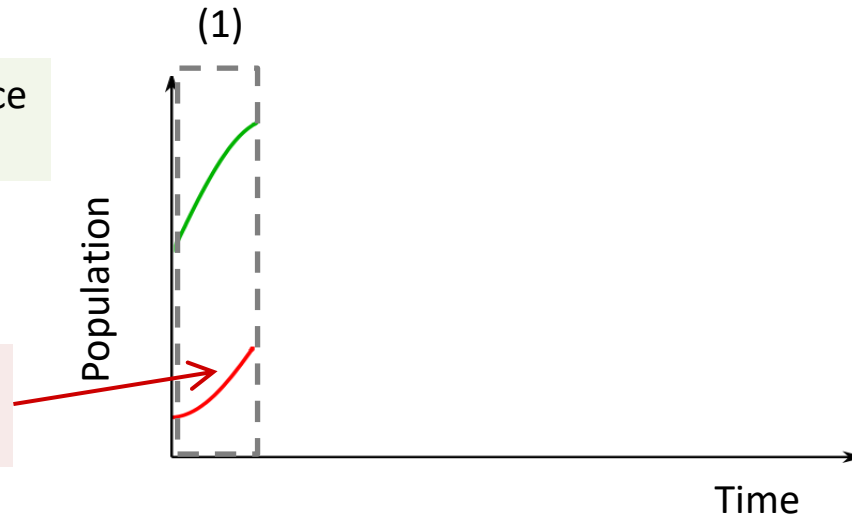


Cartoon picture for normal cycles ($\pi/2$ phase shift)



Prey consumes resource
and grows

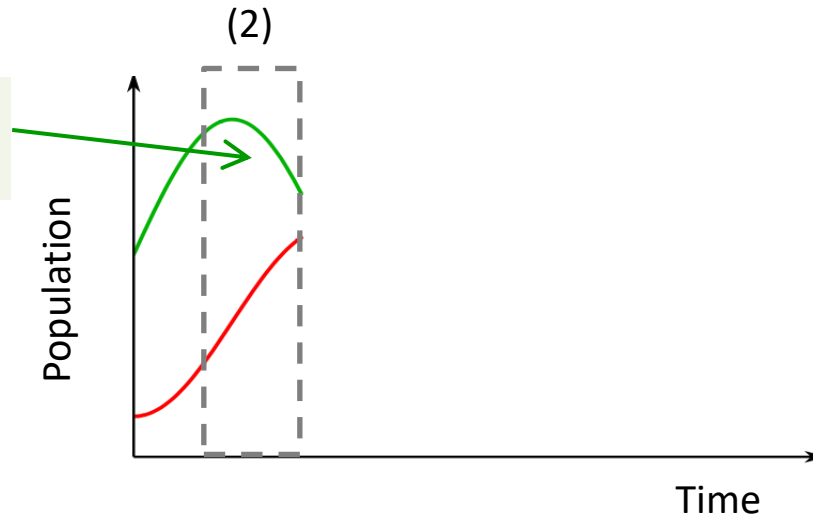
Predator eats prey
and grows



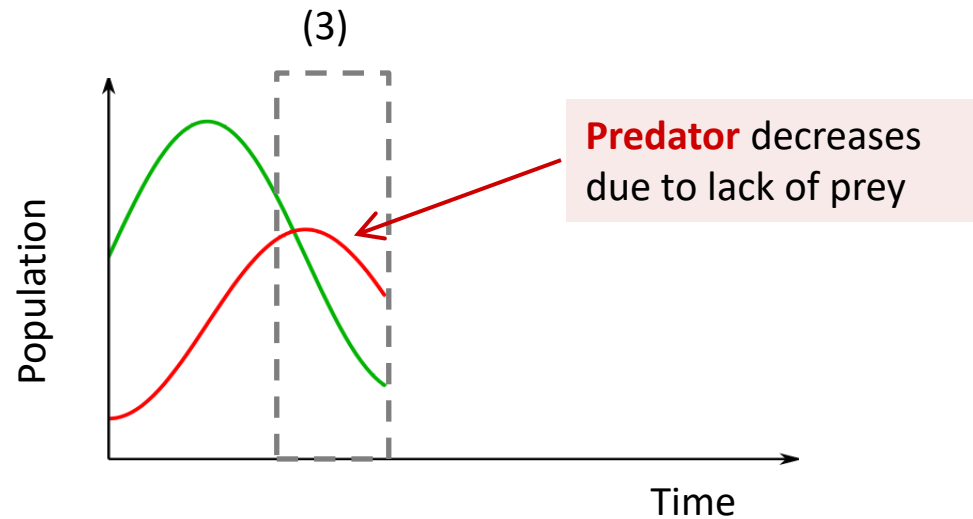
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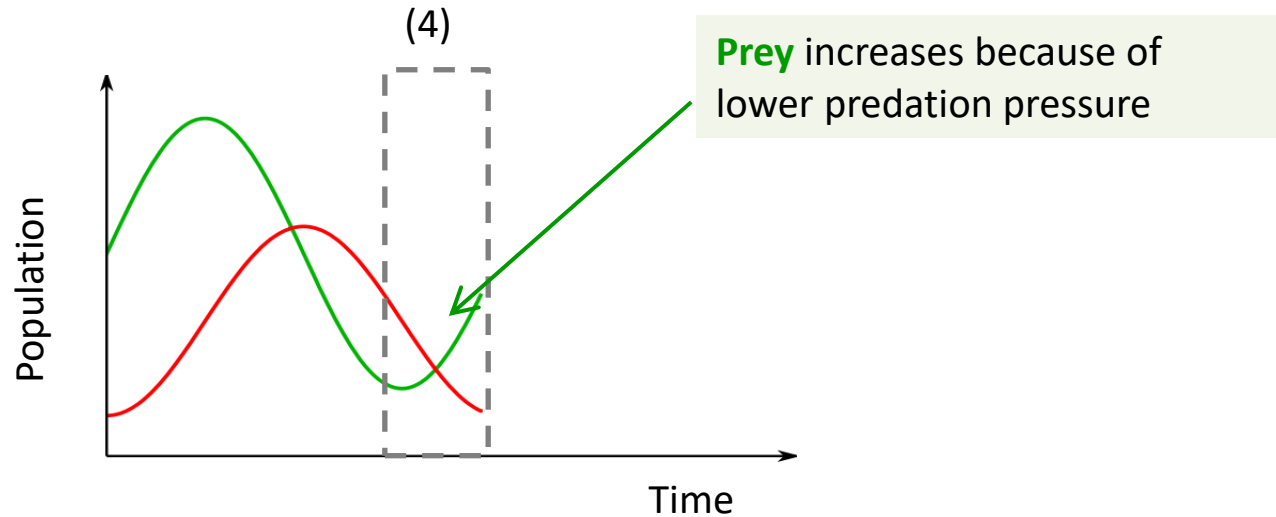
Prey decreases due to predation by predator



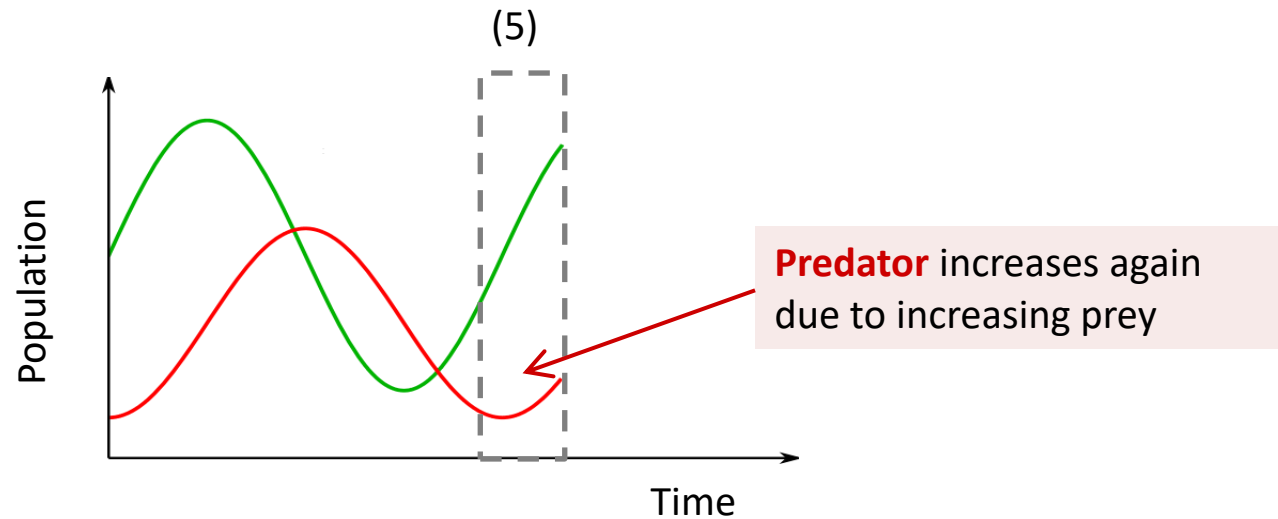
Cartoon picture for normal cycles ($\pi/2$ phase shift)



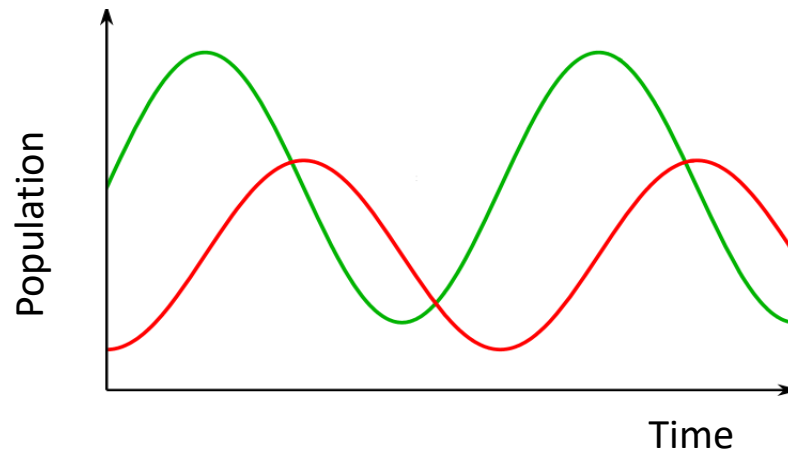
Cartoon picture for normal cycles ($\pi/2$ phase shift)



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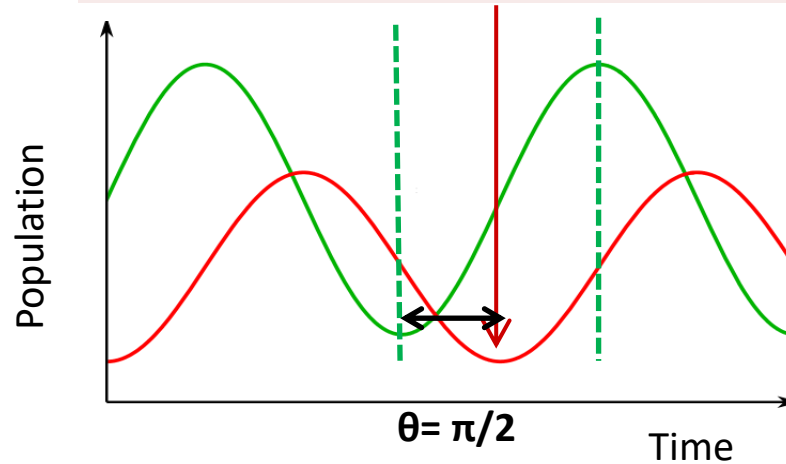
Cartoon picture for normal cycles ($\pi/2$ phase shift)



Cartoon picture for normal cycles ($\pi/2$ phase shift)



Predator can only start to grow after **prey** grows and before **prey** declines



Phase shift is a quarter period

Lotka-Volterra equations for predator-prey dynamics

- Lotka-Volterra eqn: conventional model for population dynamics
- L-V for prey-predator system:

Lotka-Volterra equations for predator-prey dynamics

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 - Reproduction of prey proportional to prey density → exponential growth

$$\frac{du}{dt} = bu$$

Lotka-Volterra equations for predator-prey dynamics

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 - Limited food resource → consider carrying capacity of prey, K_u

$$\frac{du}{dt} = bu\left(1 - \frac{u}{K_u}\right)$$

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 - Reproduction of prey proportional to prey density → exponential growth
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 - Predator hunts prey → predation proportional to prey & predator densities

$$\frac{du}{dt} = bu\left(1 - \frac{u}{K_u}\right) - puv$$

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$$\frac{du}{dt} = bu\left(1 - \frac{u}{K_u}\right) - puv$$
$$\frac{dv}{dt} = puv - dv$$

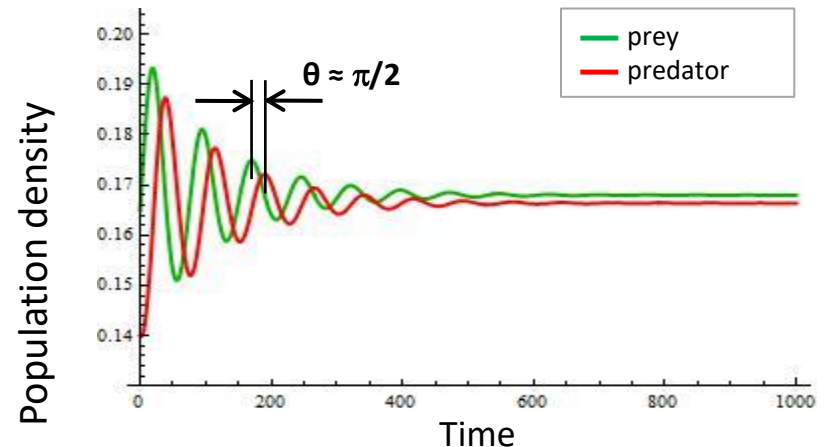
u: prey v: predator b: prey metabolic rate
 K_u : prey carrying capacity
p: predation rate d: predator death rate

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u: prey v: predator b: prey metabolic rate
 K_u : prey carrying capacity
p: predation rate d: predator death rate



- Predicts $\pi/2$ phase shift between prey and predator
- **Problems:** No oscillations → **Contrary to experiments!**

Stochasticity can qualitatively change the predictions of ecological models

Noise can stabilise persistent cycles
in time and patterns in space

Predator-prey models

Deterministic

$$\text{prey} \quad \frac{dx}{dt} = bx - ex^2 - pxy$$

$$\text{predator} \quad \frac{dy}{dt} = pxy - dy$$

Predator-prey models

Deterministic

$$\text{prey} \quad \frac{dx}{dt} = \overset{\text{growth}}{bx} - \overset{\text{competition}}{ex^2} - \overset{\text{predation}}{pxy}$$

$$\text{predator} \quad \frac{dy}{dt} = \underset{\text{predation}}{pxy} - \underset{\text{death}}{dy}$$

Predator-prey models

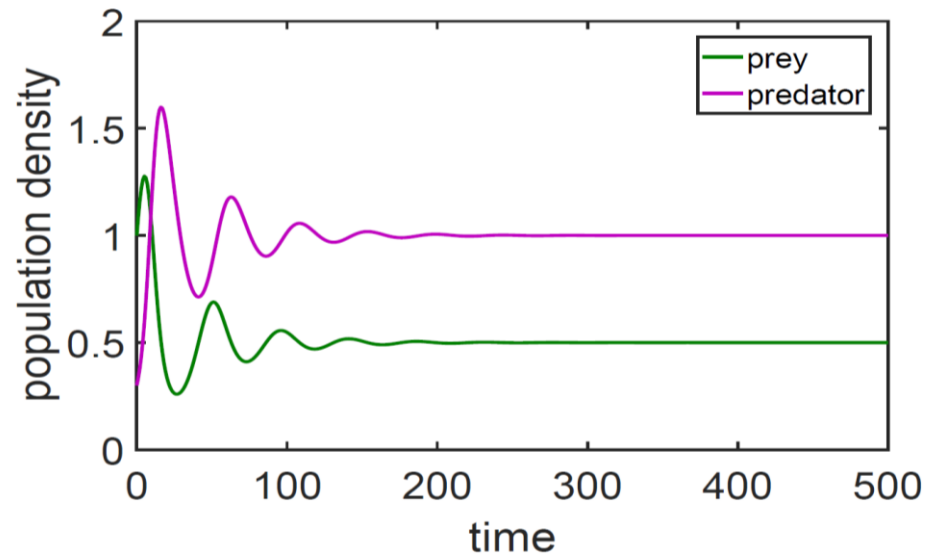
Deterministic

$$\text{prey} \quad \frac{dx}{dt} = bx - ex^2 - pxy$$

$$\text{predator} \quad \frac{dy}{dt} = pxy - dy$$

Oscillatory damping toward
a steady state.

Cycles are not persistent.



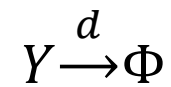
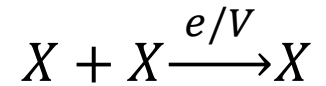
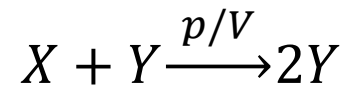
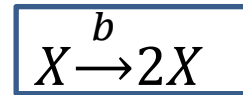
Predator-prey models

Deterministic

$$\text{prey} \quad \frac{dx}{dt} = \boxed{bx} - ex^2 - pxy$$

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Stochastic



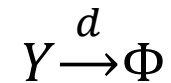
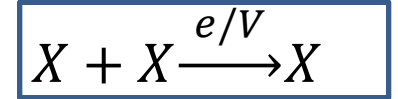
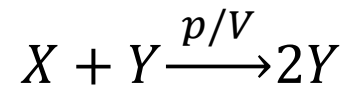
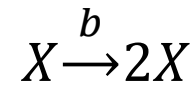
Predator-prey models

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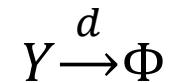
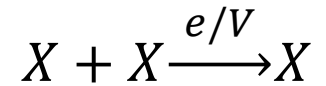
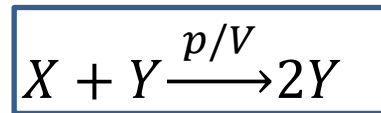
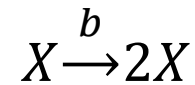
Predator-prey models

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predator $\frac{dy}{dt} = \boxed{pxy} - dy$

Stochastic



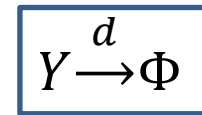
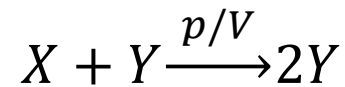
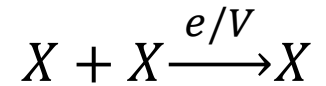
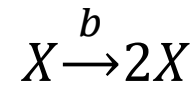
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Predator-prey models

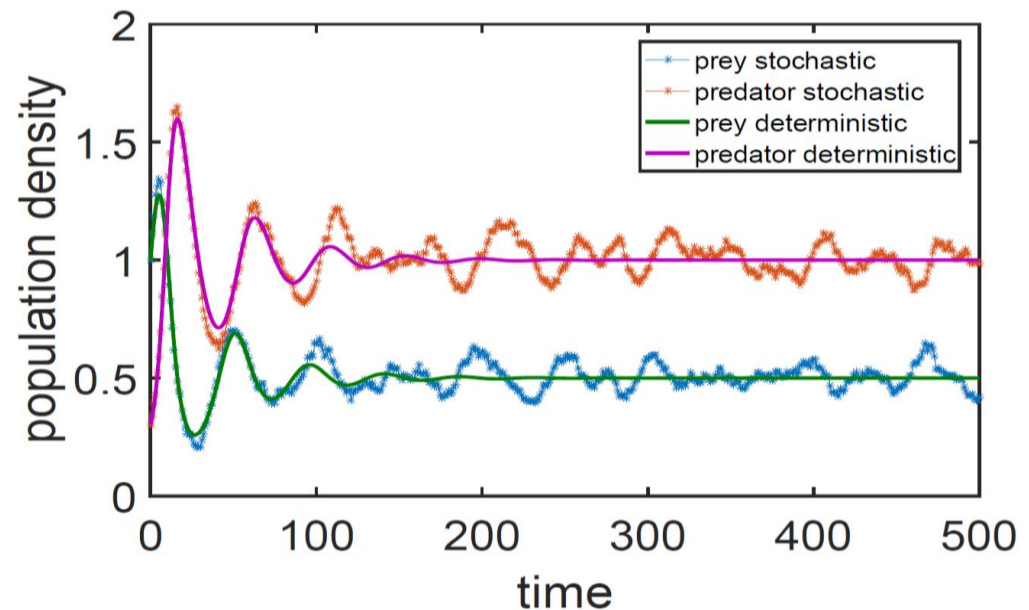
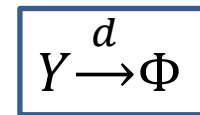
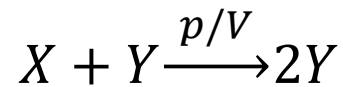
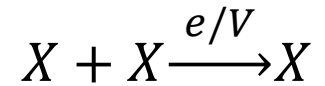
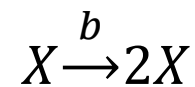
Deterministic

prey $\frac{dx}{dt} = bx - ex^2 - pxy$

predator $\frac{dy}{dt} = pxy - \boxed{dy}$

Noisy persistent
quasi-cycles

Stochastic



Predator-prey models

Deterministic

Stochastic

prey $\frac{dx}{dt} = r - \frac{e}{V} X$

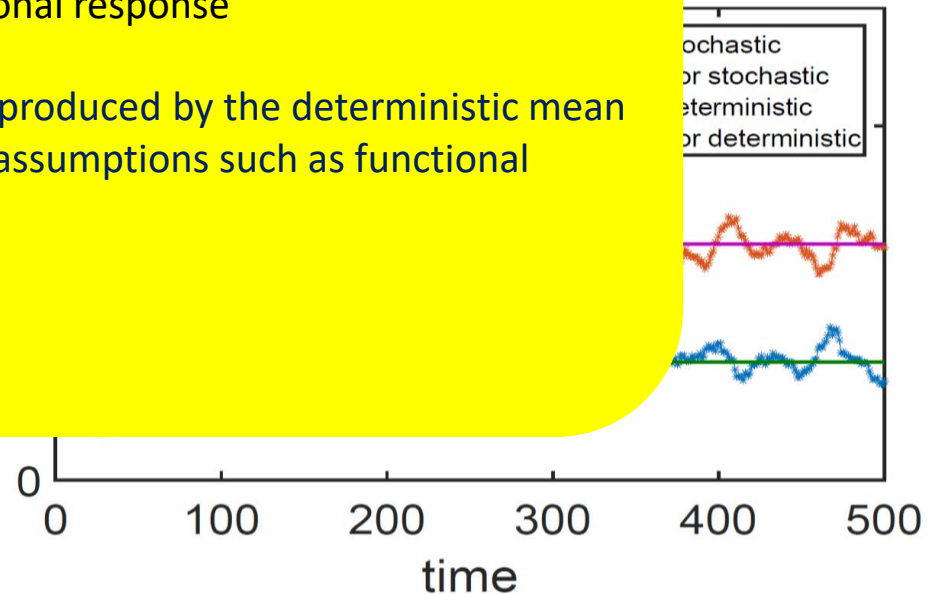
predator $\frac{dy}{dt} = \frac{e}{V} X - d$

Stochasticity recapitulates the population cycles without needing to invoke additional levels of realism such as predator satiation or other assumptions about functional response

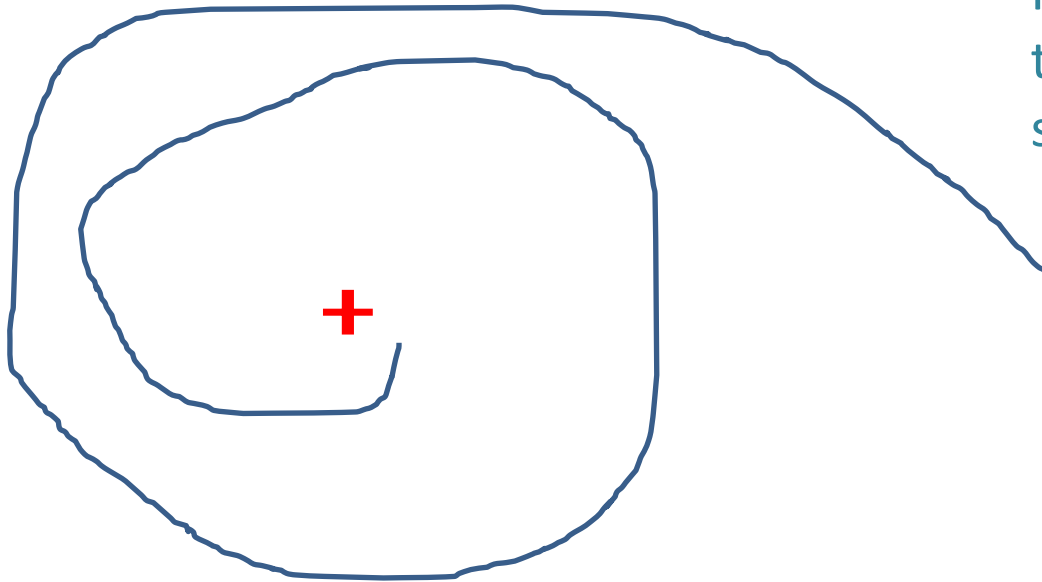
Persistent oscillations are not reproduced by the deterministic mean field theory, without additional assumptions such as functional response

Noise
qual

$e/V \rightarrow X$

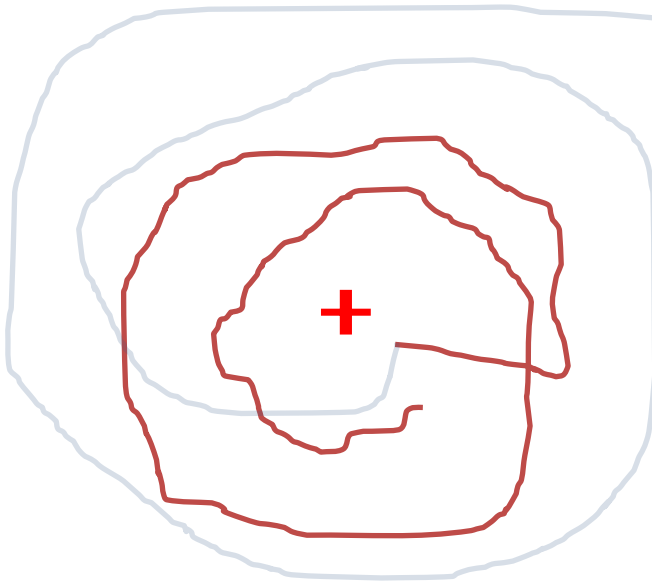


Noise-induced quasi-cycles



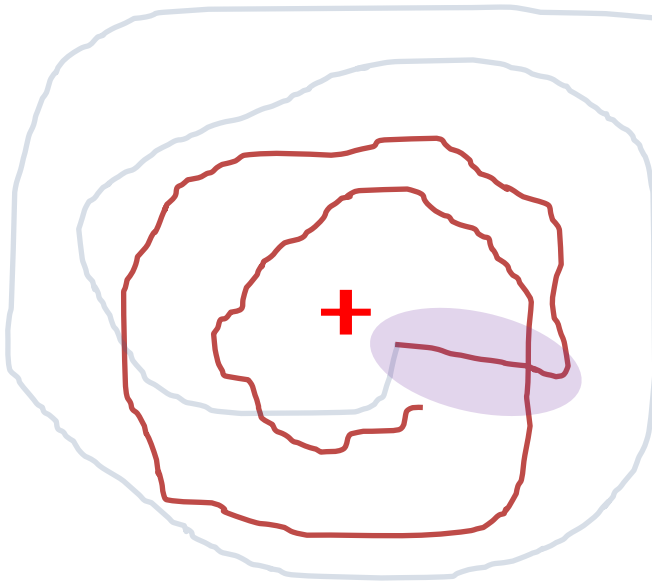
Populations decay
toward the steady
state.

Noise-induced quasi-cycles



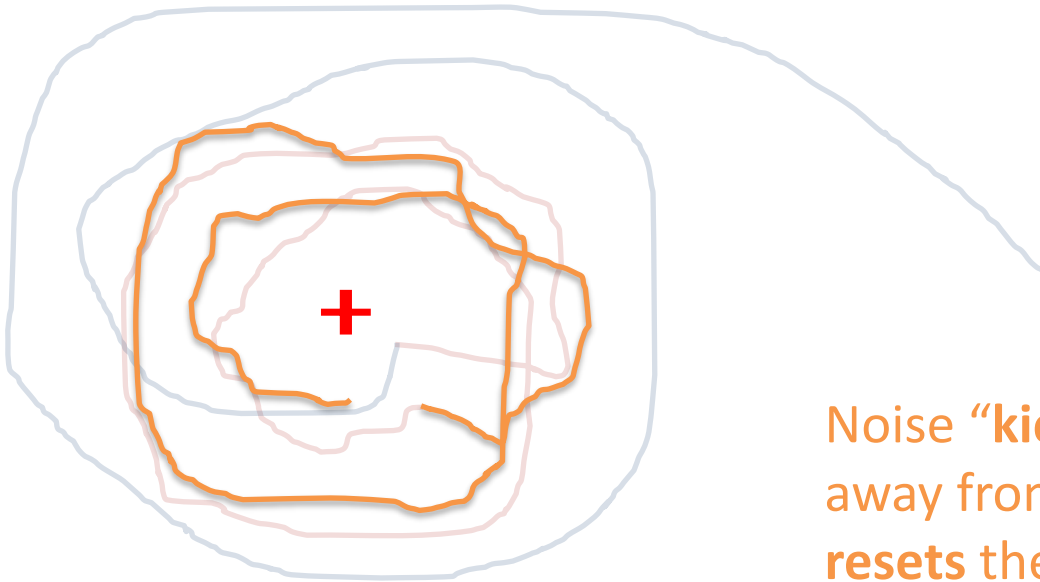
Noise “**kicks**” the trajectory away from the original path, and **resets** the decay.

Noise-induced quasi-cycles



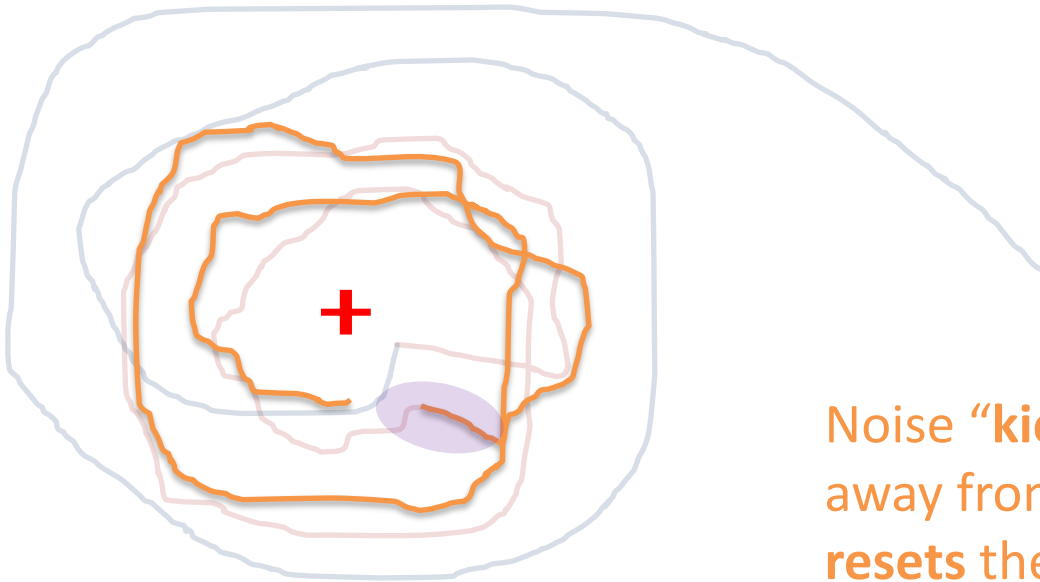
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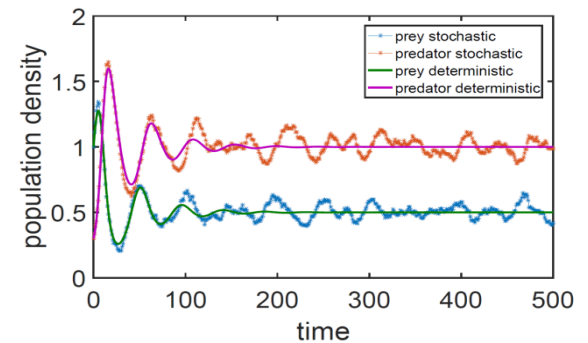
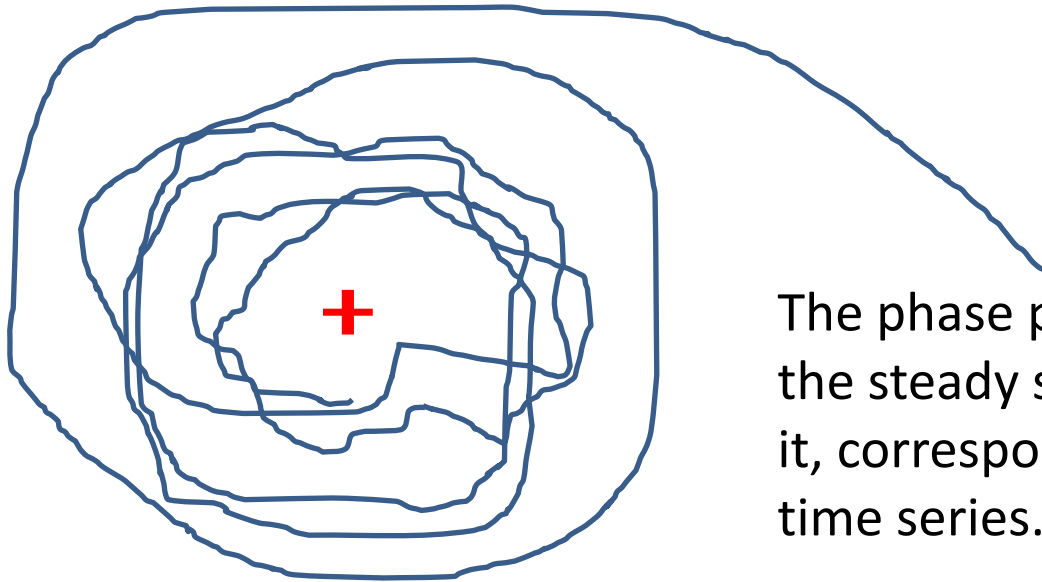
Noise “**kicks**” the trajectory away from the original path, and **resets** the decay, again.

Noise-induced quasi-cycles



Noise “**kicks**” the trajectory away from the original path, and **resets** the decay, again.

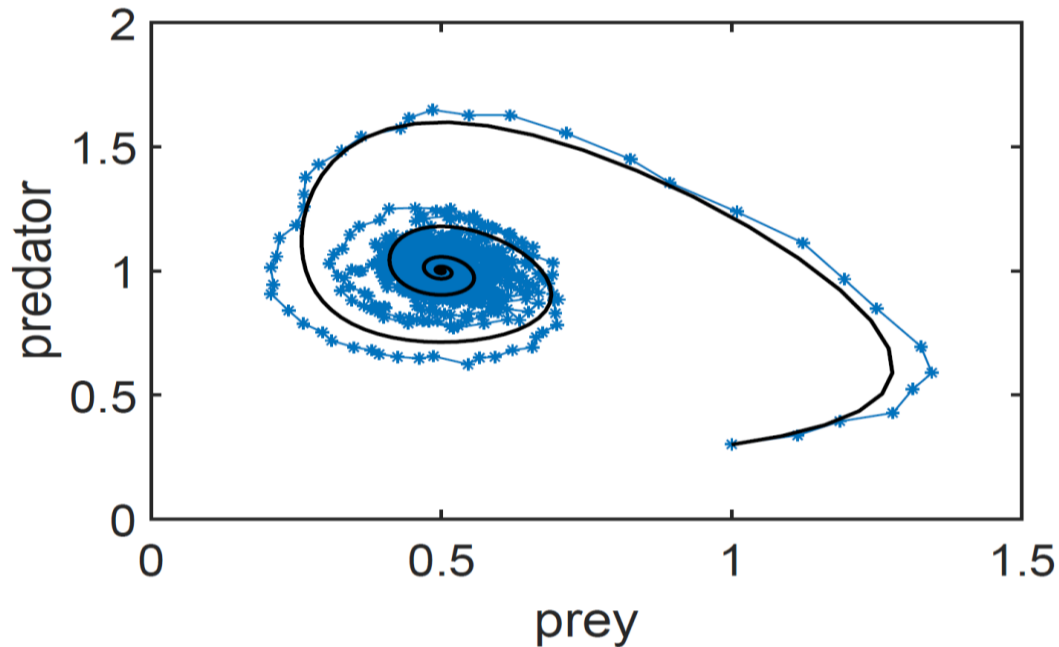
Noise-induced quasi-cycles



The phase portrait winds around the steady state but never stays at it, corresponding to quasi-cycles in time series.

The winding period is close to the period of stable complex eigenvalue about the fixed point.

Phase portrait



Deterministic

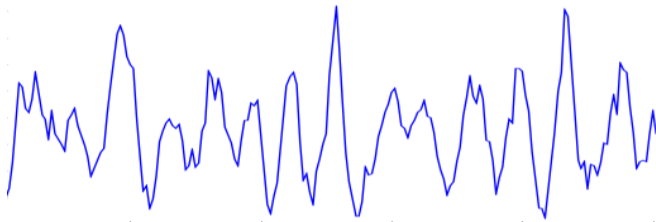
Damped motion
toward the fixed point

Stochastic:

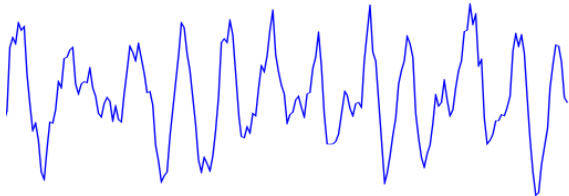
Noise stimulates
deviation from fixed
point, followed by
relaxation to fixed
point, followed by
noise ...

What is the problem with Lotka-Volterra?

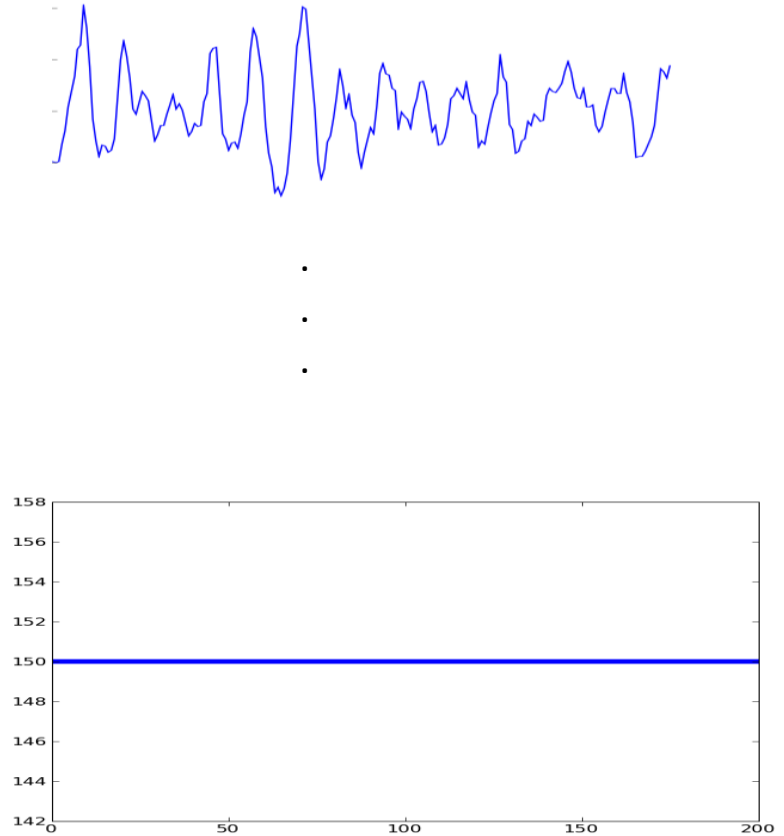
Averaging used to derive Lotka-Volterra removes cycles or spatial structure



+



=



Comparison of Models

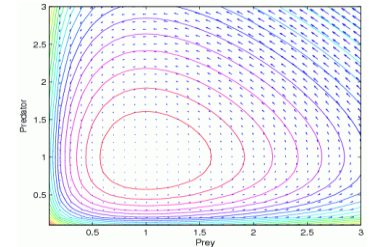
- **Lotka-Volterra equations** (Holling type I response function):

$$\frac{du}{dt} = bu - puv$$

$$\frac{dv}{dt} = puv - dv$$

the fixed point is a center
existence of a conserved quantities

Initial cond. dependent



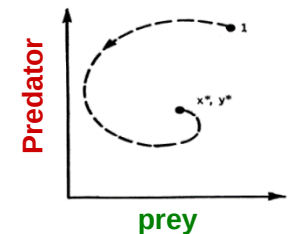
- **Volterra equations:**

$$\frac{du}{dt} = bu\left(1 - \frac{u}{K_u}\right) - puv$$

$$\frac{dv}{dt} = puv - dv$$

the fixed point is a stable focus

No persistent oscillations



- **Satiation model (Holling type II response function):**

$$\frac{du}{dt} = bu\left(1 - \frac{u}{K_u}\right) - p\frac{uv}{K_s + u}$$

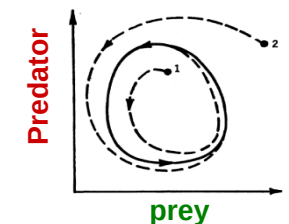
$$\frac{dv}{dt} = p\frac{uv}{K_s + u} - dv$$

the fixed point is a limit cycle

Persistent oscillations

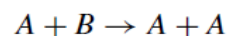
No fluctuations

Not robust



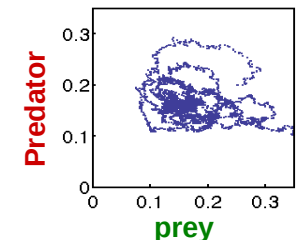
- **Individual level model** (Volterra equations + demographic stochasticity)

$$A \rightarrow \emptyset$$



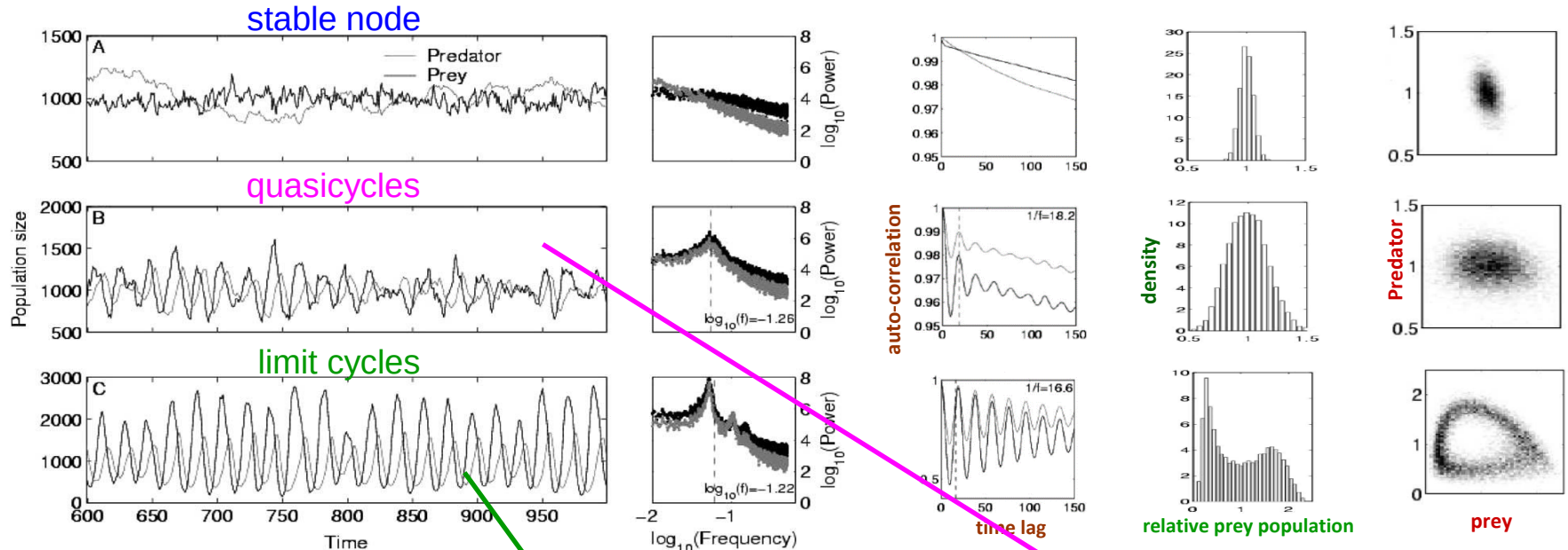
demographic stochasticity drives the trajectory
out from the fixed point (a stable focus)

Persistent random oscillations
Generic

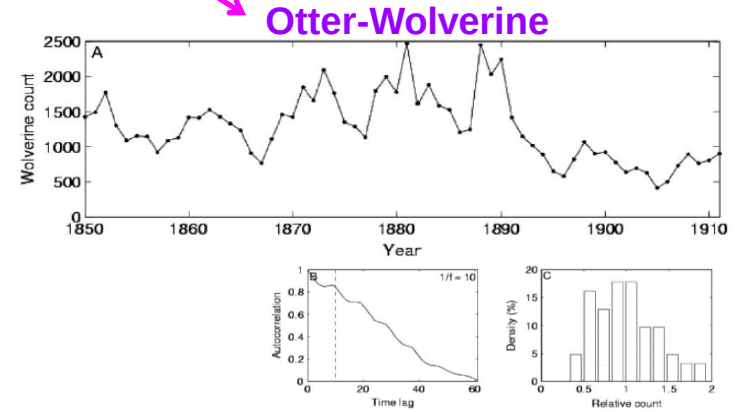
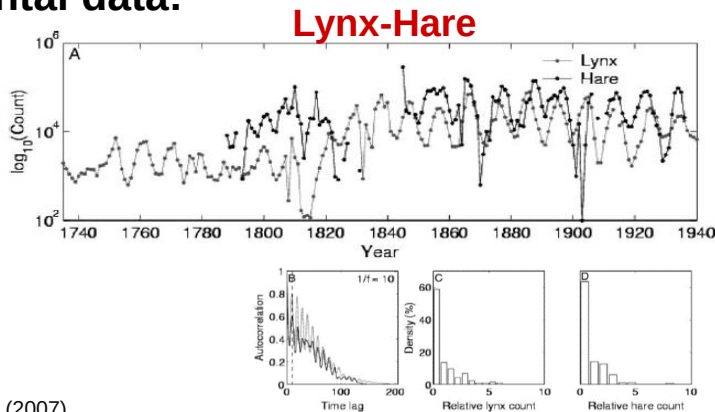


Quasi-cycles vs. Limit cycles in Experiments

- Qualitative differences predicted by stochastic simulations:



- Experimental data:



**Landau theory for transitional pipe
turbulence is stochastic predator-prey**

Bill Wyld (1928-2013)



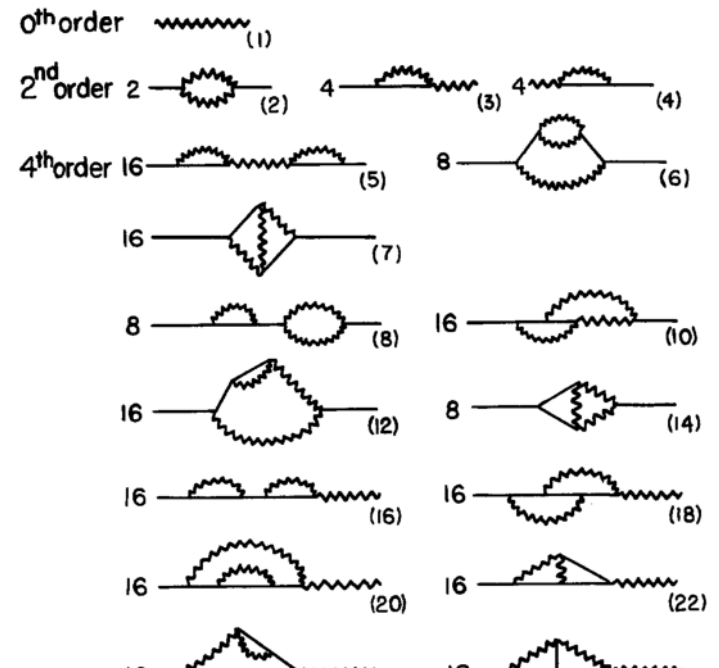
Formulation of the Theory of Turbulence in an Incompressible Fluid

H. W. WYLD, JR.

Physics Department, University of Illinois, Urbana, Illinois and Space Technology Laboratories, Los Angeles, California

The theory of turbulence in an incompressible fluid is formulated using methods similar to those of quantum field theory. A systematic perturbation theory is set up, and the terms in the perturbation series are shown to be in one to one correspondence with certain diagrams analogous to Feynman diagrams. From a study of the diagrams it is shown that the perturbation series can be rearranged and partially summed in such a way as to reduce the problem to the solution of three simultaneous integral equations for three functions, one of which is the second order velocity correlation function. The equations have the form of infinite power series integral equations, and the first few terms in the power series are derived from an analysis of the diagrams to sixth order. Truncation of the integral equations at the lowest nontrivial order yields Chandrasekhar's equation, and truncation at a higher order yields the equations discussed by Kraichnan.

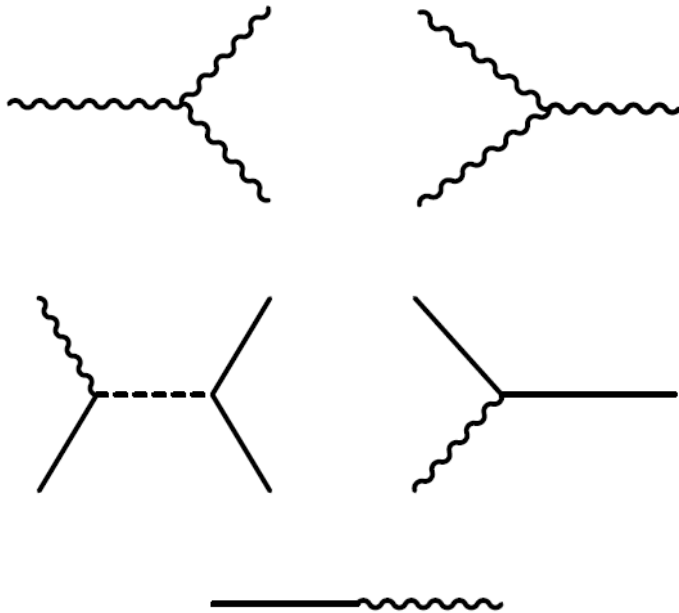
THEORY OF TURBULENCE



Derivation of predator-prey equations

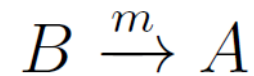
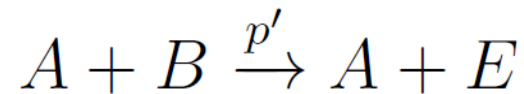
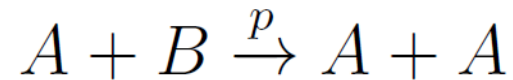
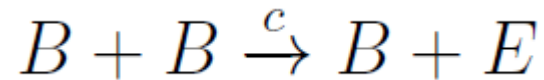
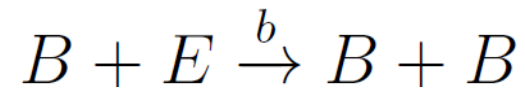
 Zonal flow  Turbulence
 Vacuum = Laminar flow

Zonal flow-turbulence



A = predator B = prey
 E = food/empty state

Predator-prey



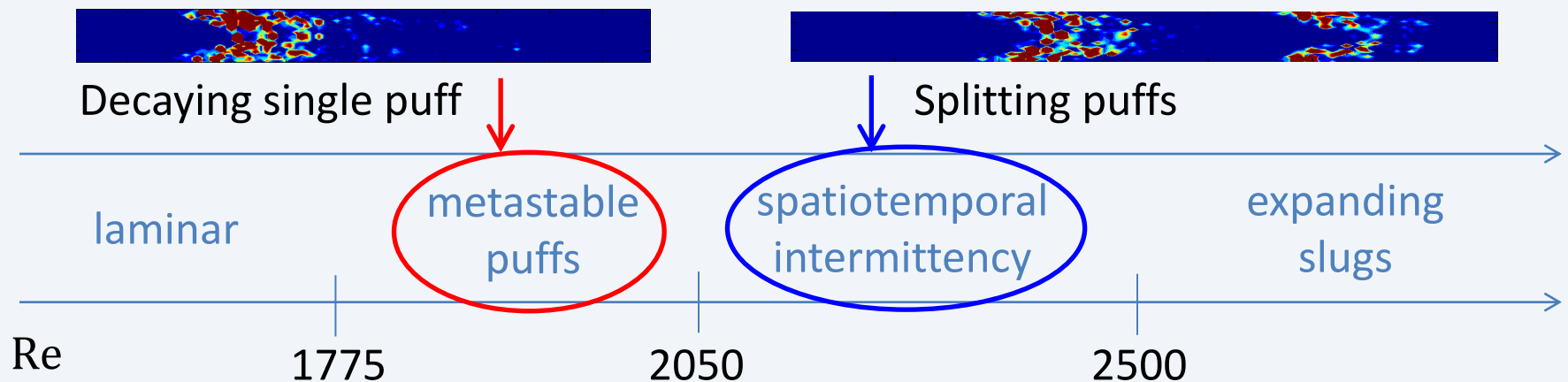
Stochastic predator-prey recapitulates turbulence data

Phase diagram

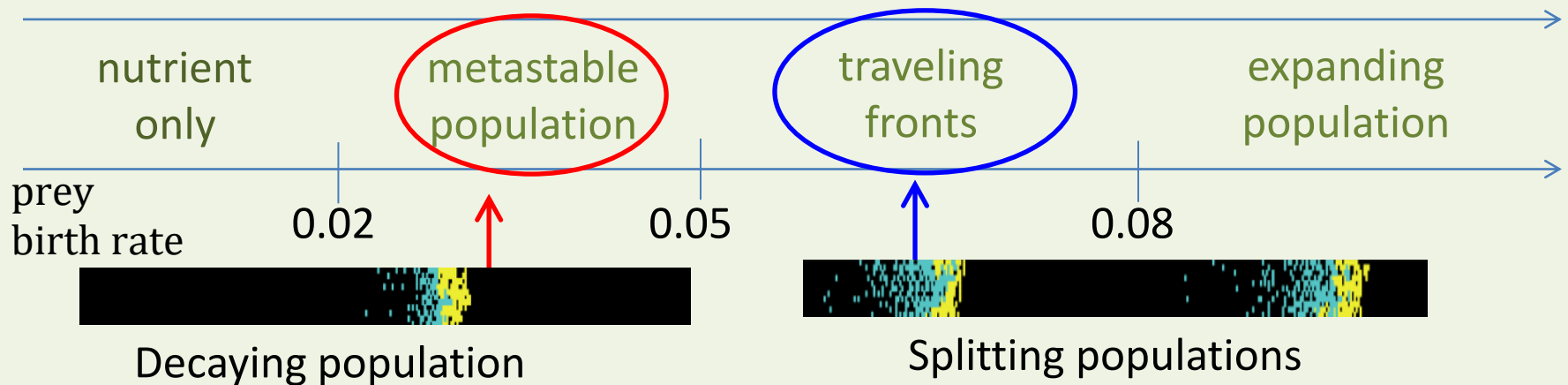
Lifetime statistics

Universality class prediction

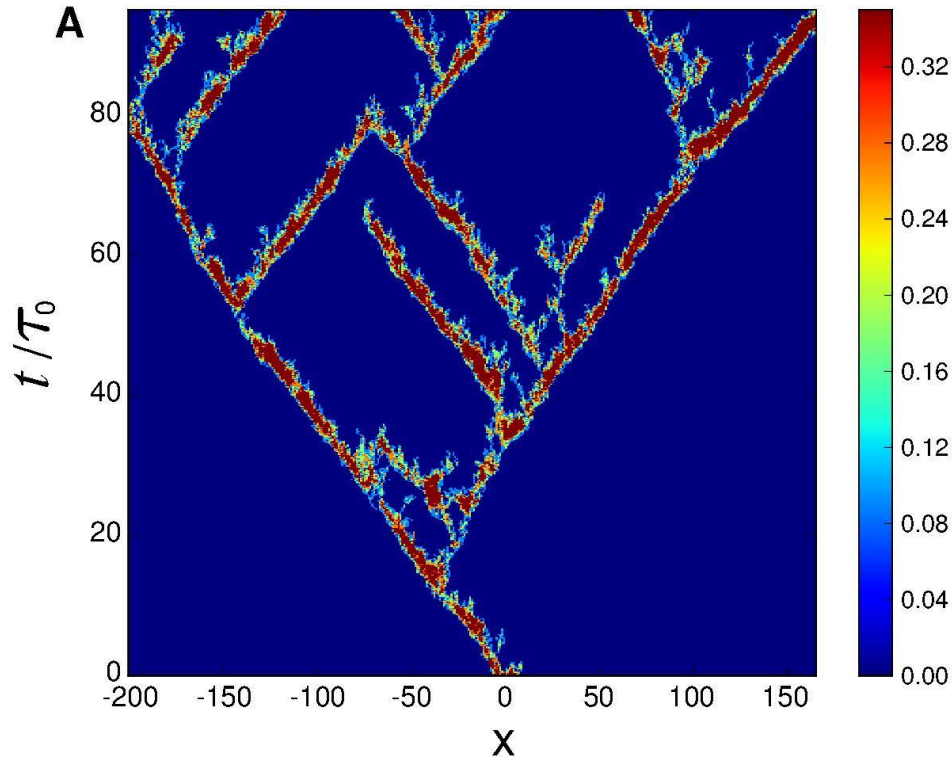
Pipe flow turbulence



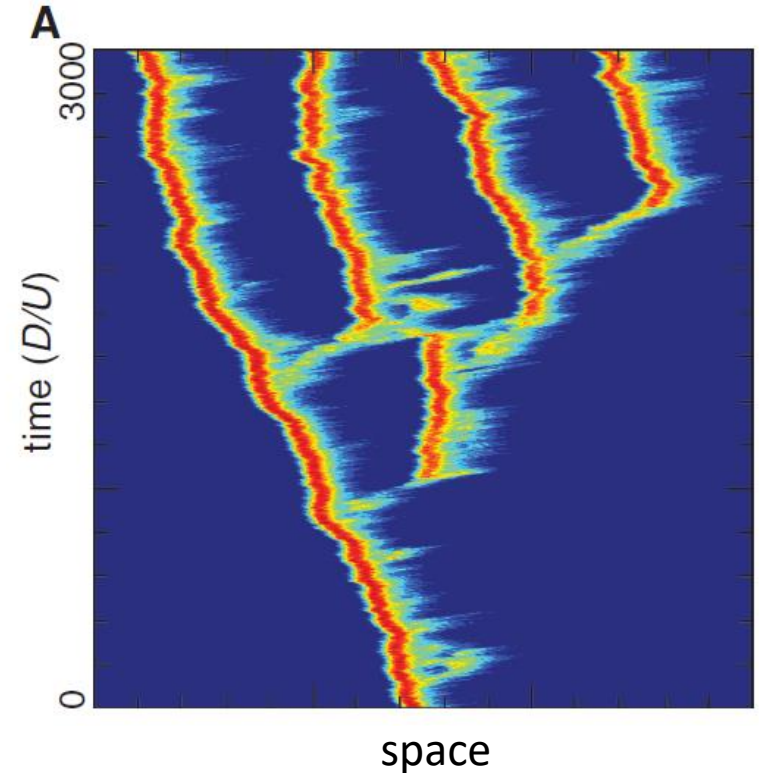
Predator-prey model



“Puff splitting” in predator-prey systems

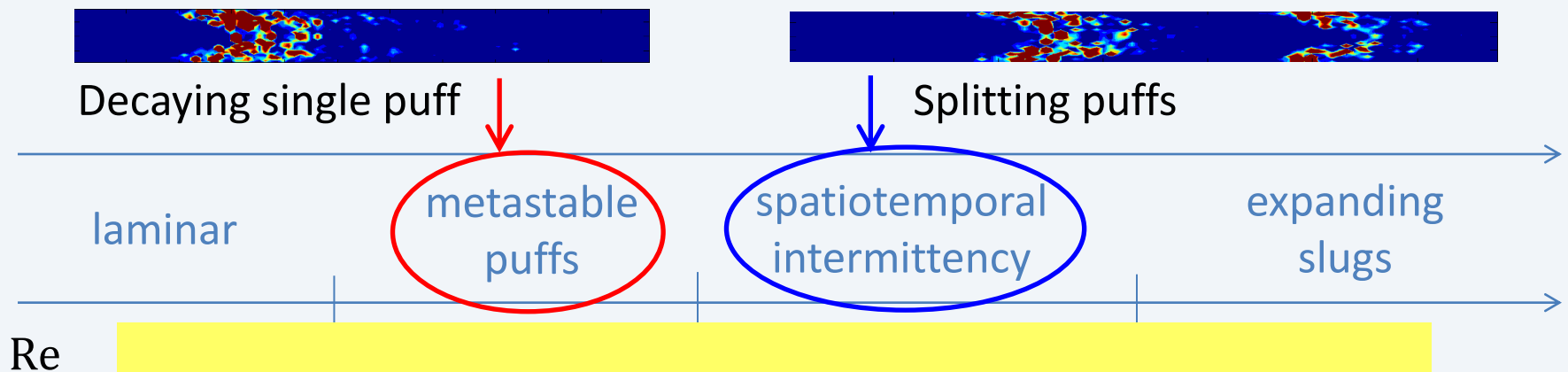


Population-splitting in
predator-prey ecosystem

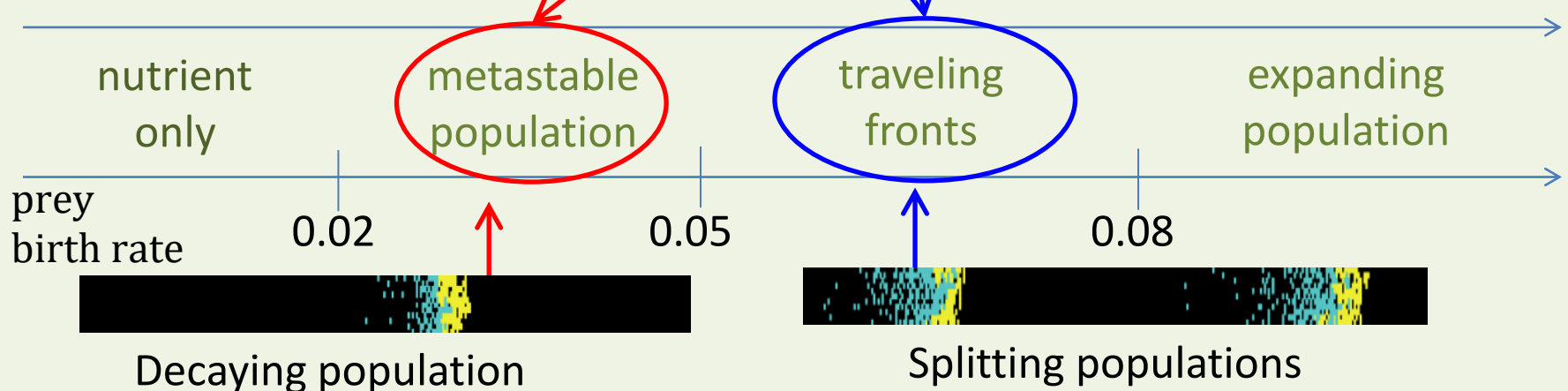


Puff-splitting in pipe turbulence
Avila et al., Science (2011)

Pipe flow turbulence

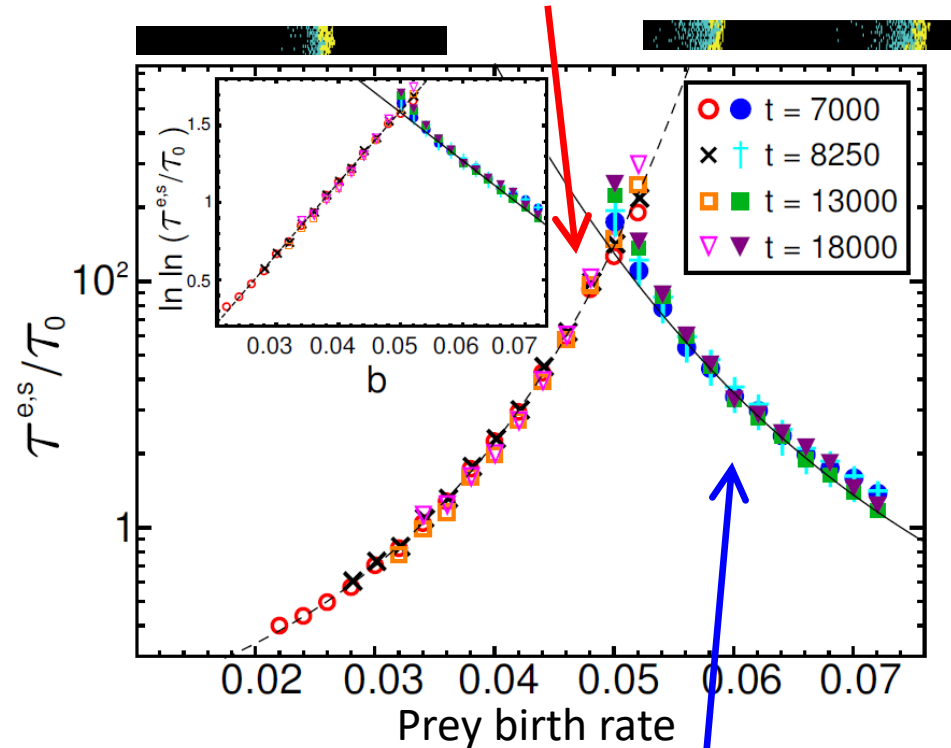


Measure the statistics of the **extinction time** and the **time between population split events** in predator-prey system.



Predator-prey vs. transitional turbulence

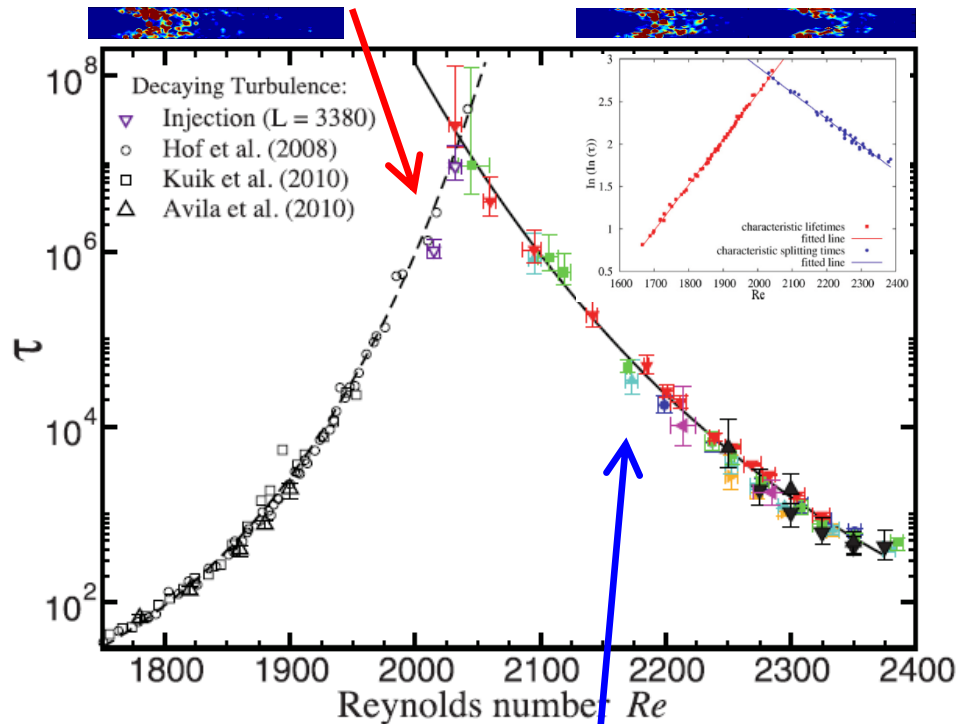
Prey lifetime



Mean time between
population split events

Shih, Hsieh and Goldenfeld
Nature Physics **12**, 245 (2016)

Turbulent puff lifetime

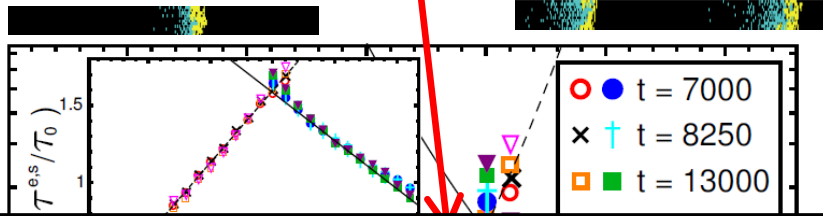


Mean time between
puff split events

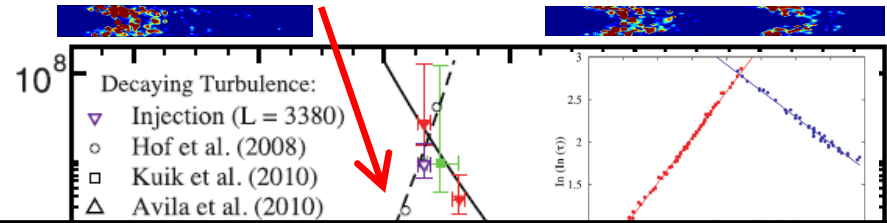
Avila et al., *Science* **333**, 192 (2011)
Song et al., *J. Stat. Mech.* 2014(2), P020010

Predator-prey vs. transitional turbulence

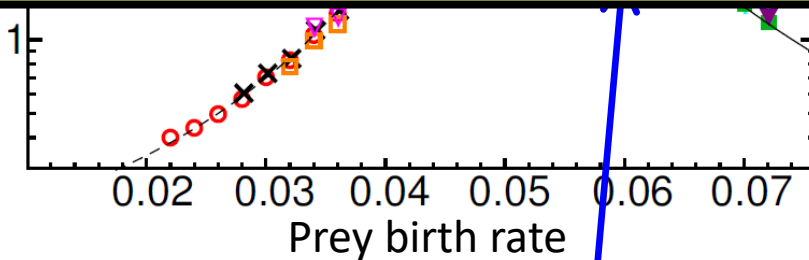
Prey lifetime



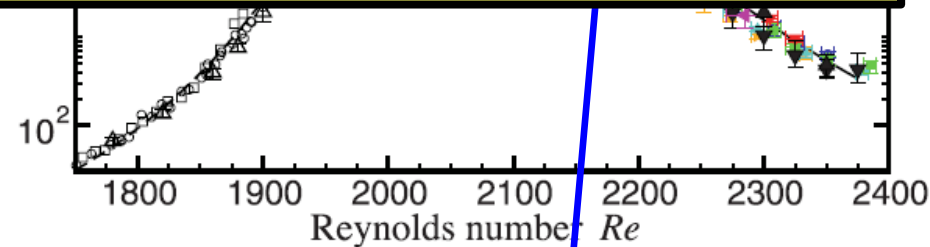
Turbulent puff lifetime



Extinction in Ecology = Death of Turbulence



Mean time between
population split events

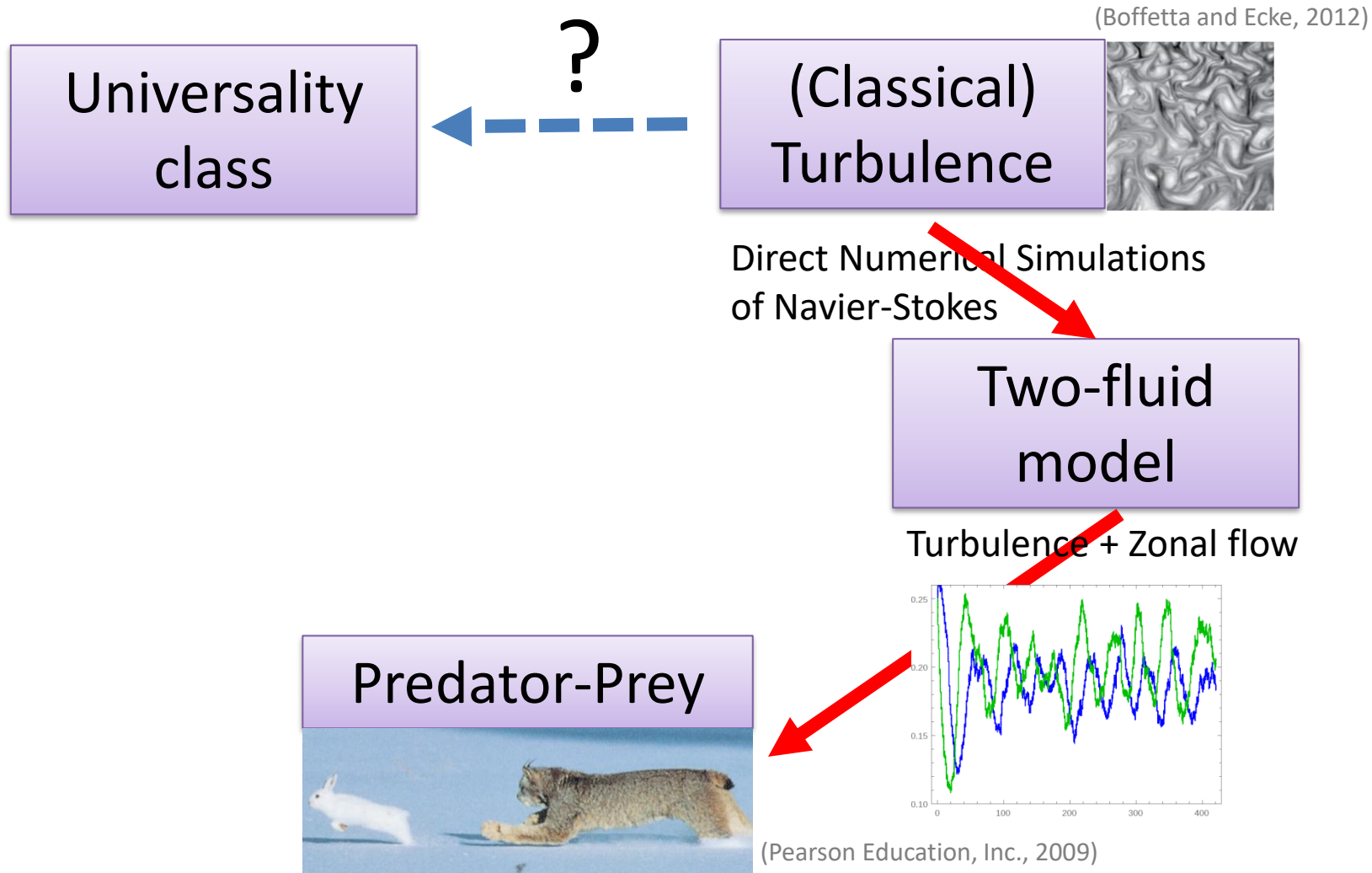


Mean time between
puff split events

Shih, Hsieh and Goldenfeld
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Avila et al., *Science* **333**, 192 (2011)
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Roadmap: Universality class of laminar-turbulent transition



Roadmap: Universality class of laminar-turbulent transition

