

# PKU-HUST Lectures (2020-2021)

QV

- 'Modeling Dynamical Systems' for MFE, C.C. & after Biophysics)
- a useful (I hope) methods

mini-course { minimal practical knowledge }

- Emphasis on nonlinear dynamics, space-time scales
- Topics:

- i.) Feedback loops, (time) predator-prey
- ii.) transition evolution and propagation Fisher, FN  
(space-time) traffic flow
- iii.) avalanches, bursty flux  
(cavenges)  $\rightarrow$  (aval/solutions)
- iv.) nonlinear diffusion and spreading.
- v.) Negative viscosity and Cahn-Hilliard eqn

vd)

Model reduction and renormalization  
- a look at the theory

+

History of Nonlinear Plasma Theory  
(1 → 2 lectures)

E P J H

P.D. Frisch, Y. Pomeau

Sources:

- J.D. Murray, "Mathematical Biology"
- R. May, "Stability and Complexity in Model Ecosystems"
- F. Morrison, "The Art of Modelling Dynamic Systems"
- + more coming
- selected papers.

# PKU-HUST Lectures I

## (1) Feedback Loops and Predator-Prey Systems

Plan:

- OV and Motivation system
- Constructing the System - Solving

Drift Waves + Shear Flow
- Lessons from Ecology (c.f. May)
  - General Structure ; constraint
  - Aside: Bigger Models, Stability & Complexity
  - General structure ?
  - Kolmogorov Theorem ; Fixed pts. vs. cycles ; Feedback
  - Time Delays → Cycles

## - Fluctuating Environments and Static vs Approaches.

→ Physical Motivation

→ Fluctuating Logistics - a simple example.

→ Applications



- looking Ahead.

I) QV and Basins

→ Fundamental Problem(s)

- Drift waves + zonal shear flow

- L → H + transition

- Natural candidate for Predator - Prey Model

## 2 animal

- what?

$$\frac{dH}{dt} = H F(H, P)$$

↑ rates

$$\frac{dP}{dt} = P G(H, P)$$

P(t), H(t)

Hare

H = prey

P = predator

(CD)



can extend to  $2m \times 2m$ .

Lohse - Volterra

- What does it mean?

$H \rightarrow \boxed{\varepsilon}$  - fluctuation intensity,

$P \rightarrow U = \boxed{V^2}$  - mean square  $V_E^2$

$\boxed{CD}$  ?  $\varepsilon = \langle \varepsilon \rangle$  } avg. over thin layer  
 $V^2 = \langle V^2 \rangle$  } [ignore optical evaporation]

- Why?

2 populations

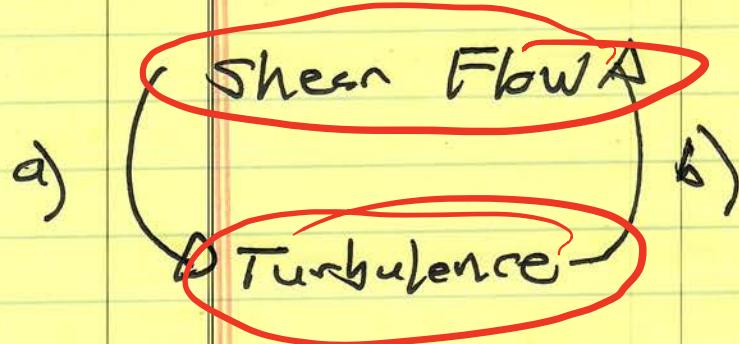
→ self-regulation (feedback)

→ symmetry constraints

$\langle \varepsilon \rangle$   
 $\varepsilon$

## TIME

## 1) Self-regulation



a) shear stabilizes, controls turbulence

$\rightarrow$  everyone knows ...

b) turbulence drives flow via Reynolds stress

$$\underline{J} = \frac{\partial}{\partial t} \underline{J}_{p-1} + \underline{J}_{ps} + \underline{J}_4$$

$$\partial_t \langle \underline{v}_z \rangle = - \nabla_z \langle \bar{v}_r \bar{v}_z \rangle$$

$$\underline{D} \cdot \underline{J} = 0$$

A.b. More precisely:

from vorticity eqn. ( $\underline{D} \cdot \underline{J} = 0$ )

$$\frac{\partial}{\partial t} \left( \frac{\partial^2}{\partial z^2} \partial_r \langle \phi \rangle \right) = - \partial_r \langle \bar{v}_r \frac{\partial^2}{\partial z^2} \bar{\phi} \rangle + \dots$$

omogeneity

Flux of polarization charge

$\langle \bar{v}_r \bar{n}_i \rangle_{go} - \langle \bar{v}_r \bar{n}_e \rangle$

NOIR

Design

Net polarization charge

( $\bar{n}_i$  generalized  
- Rosenbluth,  
Hinton 77)

ExB shear flow

1 sym.

Reynolds force



but:  $\langle \tilde{U}_r \nabla_L^2 \phi \rangle = \partial_r \langle \tilde{U} \phi \rangle$

(HW): Show this!

(G.I.) Taylor's identity (100+ yrs old)

Upshot:

- Reynolds stress, force  $\leftrightarrow$   
polarization charge flux

\* - Thm: McIntyre + Wood:

R. Wood

"PV mixing" [i.e. polarization charge flux]  
+ I direction of symmetry

$\Rightarrow$  Z.-F. generation.

∴ How reconcile?  $\Rightarrow$  Feedback loop

2) **Symmetry \*** (already encountered)

Why is Z.F. 'special'?

(P.O., I<sup>2</sup>, H  
'05)

Z.F. is mode  
of:



(k<sub>r</sub> ≠ 0)

(n, m → 0)

- ~~minimal~~ minimal inertia - easily excited

k<sub>z</sub> → 0

$$(1 + k_z^2 \rho^2)$$

$$k_z^2 \rho^2$$

\*

H.W.

H = M

2D fluid

- ~~minimal transport~~  
(zero!)

(none!)

\*

$$\cdot n = 0 \Rightarrow \tilde{v}_r = 0$$

$$m = 0$$

$$\tilde{v}_r = 0$$

can't relax Dn, DT, etc.

- ~~minimal damping~~

\*

(no Landau damping - R.H., '97)

Symmetry → P-P.

- ⇒ Symmetry constraints + consideration of self-regulation
- ⇒ Z. F. can grow/he excited only via 'feeding off' of turbulence, yet shear turbulence, thus regulating it.

⇒ natural problem for predator-prey formulation!

→ What does it look like?

Cf. PD, et. al. '94) Mixing length process

$$\frac{dE}{dt} = \gamma_0 E - \alpha_1 E^2 - \alpha_2 U E$$

$\frac{dU}{dt} = -\mu_1 U + \alpha_2 U E$

Fluxes  
growth turb.  
self sett.  
coupling key  
Flow damping

see table in paper for coeff

→ N.B.: - simplest

→ An infinity of extensions:

most used.

$$\sim \bar{f}_0 \rightarrow \bar{f}_0(\nabla P) \rightarrow \bar{f}_0(a) L_p = a / L_{\text{unit}}$$

↑  
coupling to profile evolution

$$\langle \nabla f L_P \rangle + \partial_n \langle \tilde{V}_n \tilde{P} \rangle = n Q = n (\bar{Q} + \tilde{Q})$$

nose

$$\underbrace{\nabla f L_P + \partial_n \chi_0 \sum \partial_n L_P}_{\text{OD}} = n (\bar{Q} + \tilde{Q})$$

$$\text{OD} \quad \nabla f L_P + D_0 \sum L_P = n (\bar{Q} + \tilde{Q})$$

Melkony, P.D. 2002

→ Kum, P.D. b3

$$\sim V_E = \frac{\nabla P}{n} + V_0 \quad \text{(Corrsen, Melk.)}$$

1D toro col. flow, particles,  
stoch. field ...  
User's guide to Arel-Prey extension

$$u = \langle v_i \rangle^2$$

B-) How is it constructed?

Key term:  $\lambda_2 u E$  → Reynolds Coupling

Point: Reynolds work transferred energy / flow ↔ fluctuations

$\Delta w$ : 2 ways

$$\text{Now, } \partial_t \langle v_i \rangle = - \partial_r \langle \tilde{v}_r v_i \rangle$$

$$\partial_t \int dr \frac{\langle v_i \rangle^2}{2} = \int \langle \tilde{v}_r v_i \rangle \partial_r \langle v_i \rangle$$

but

$$\partial_t \sum_k + \partial_t \int dr \frac{\langle v_i \rangle^2}{2} = 0$$

flctn

layer

$$\text{so } \partial_t \sum_k = - \int dr \langle \tilde{v}_r v_i \rangle \partial_r \langle v_i \rangle$$

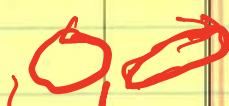
$$\text{Formally } \partial_t \langle \sum_k \rangle = - \langle \tilde{v}_r v_i \rangle \langle v_i \rangle'$$

Layer average

What of stress?

→ 'Shear - Eddy Tilting Feedback Loop'

$$\langle \tilde{v}_r \tilde{v}_t \rangle = - \frac{C^2}{B^2} \sum_k \underbrace{\langle k_x k_y \rangle}_{\text{eddy orientation!}} l_{thk}^{-2}$$



$\sim \langle k_x k_y \rangle \rightarrow \left\{ \begin{array}{l} \text{eddy} \\ \text{orientation!} \end{array} \right.$

i.e.  $l_{thk}^{-2} = f(k_x) g(k_y)$

$f, g$  even functions

$$\langle k_x k_y \rangle = 0 \quad \Downarrow \quad \begin{array}{l} \text{no stress,} \\ \text{mode structure} \end{array}$$

but

→ Shear flow can induce correlation!

i.e.

$$\frac{d k_x}{dt} = - \frac{\partial}{\partial x} (\omega + k_x V) \quad (Smeul's Law)$$

$$\frac{dk_x}{dt} = - \frac{\partial}{\partial x} (k_x V_E) = - k_x \langle V_E \rangle'$$

$$k_x = k_x^{(0)} - k_0 \langle v_E \rangle' t$$

$\sim$  [shearing coordinate]

Goldreich  
L-B '65

$$t \sim \tilde{T}_G$$



$$k_x = k_x^{(0)} - k_0 \langle v_E \rangle' \tilde{T}_G$$

$$\langle k_x k_y \rangle = \langle k_x^{(0)} k_y \rangle - \{ k_0 \langle v_E \rangle' \tilde{T}_G \}$$

intrinsic correlation

tilting induces correlation

$\Delta$  mode structure

(esp. radial propagation!)

Pearlstein  
— Berke

H W: Calculate for slab DW.

$\approx$

$$\langle \tilde{v}_r \tilde{v}_\perp \rangle = t \frac{c^2}{B_0^2} \sum_k k_\theta \langle v_E \rangle' \tilde{T}_G \cancel{\frac{1}{2} \rho_a l^2}$$

$$\approx \langle \tilde{v}_r^2 \rangle \langle v_E \rangle' \tilde{T}_G \frac{k}{l}$$

and

$$\partial_t \langle \Sigma_w \rangle \approx - \langle \tilde{v}^2 \rangle \tilde{\tau}_{\text{ex}} \langle v_{\perp}^{\prime 2} \rangle$$

$\downarrow$

$$\sim -D \langle v_{\perp}^{\prime 2} \rangle$$

note  $\sim \langle v_{\perp}^{\prime 2} \rangle^2$ !  
dependence

-Necessarily, flow eqn. must have same form, opposite sign  
 to conserve energy

-  $\langle v_{\perp}^{\prime 2} \rangle$  is natural  $v$  variable.

$\Rightarrow$  basic predator-prey structure,

$\Rightarrow$  flow - fluctuation kinetic energy exchange

Alternative Perspective:

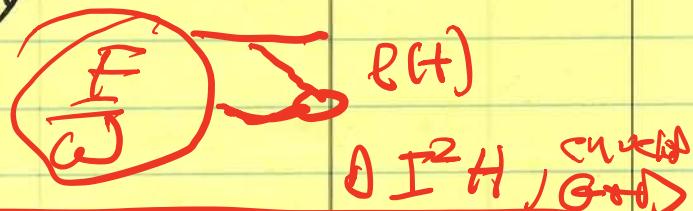
$\Rightarrow$  Spectral Energy Evolution

- Recall fluctuations respond adiabatically to Z.F.

c.e.  $\omega \approx \omega_*$  vs.  $\omega \sim 0$

then, have for wave action density

$$\textcircled{N} = \sum_k \frac{\rightarrow \text{energy density}}{\omega_k}$$



$$\partial_t N + (\underline{v}_{grt} \cdot \nabla) N - \frac{\partial}{\partial x} (\omega + k \cdot \underline{v}) \cdot \frac{\partial N}{\partial k} = CCN$$

refraction

$$\partial_t \langle N \rangle + \frac{\partial}{\partial k_r} \left\langle -\frac{\partial}{\partial x} (k \omega \tilde{v}_E) \tilde{N} \right\rangle = CCN$$

If  $\omega \rightarrow \infty$

and usual meridional crank  $\Rightarrow$   
see ( $\Delta I^2 H$  '05)

$$\partial_t \langle N \rangle - \frac{\partial}{\partial k_r} D_{kr} \frac{\partial}{\partial k_r} \langle N \rangle = \langle C(CV) \rangle$$

$$D_{kr} = \sum_l \frac{k \omega^2}{2} \left[ \frac{N \sigma^2 l^2}{2} \right] \frac{C_{kl}}{k_l}$$

induced diffusion  
random shearing

## Ray theory

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Note:  $\frac{\partial \omega}{\partial k_r}$  wave addt

$$\rightarrow \frac{1}{I}$$

$$\frac{\partial}{\partial k_r} \left[ \frac{\omega}{\omega - k_r v} + i\delta \right] \leftrightarrow \frac{\partial}{\partial k_r} \frac{\omega}{\omega - k_r v}$$

$$\partial_{k_r} D_{kr} \partial_{k_r} \sim$$

$$\frac{k_r^2}{k_r^2} \sum \left| V_{EZ} \right|^2 I^2$$



as before.

HW  
work out  
correspondence

then:

$$\langle \epsilon \rangle = \omega \langle N \rangle$$

$$\partial_t \langle \epsilon \rangle = - \int dk_r D_N \frac{\partial \langle N \rangle}{\partial k_r} \frac{\partial \omega}{\partial k_r}$$

$$= - \int dk_r \frac{\partial \omega}{\partial k_r} D_N \frac{\partial \langle N \rangle}{\partial k_r}$$

$$\frac{\omega}{1 + k_r^2 T^2}$$

backward wave

$\frac{\partial \langle N \rangle}{\partial k_r}$  fundamental

→ Same structure

$$\rightarrow \partial_t \epsilon \sim - \langle v^2 \rangle \epsilon^2$$

Answers

Lessons From Ecology

(c.f. May)

→ What can be said about this type of system?

1 predator - 1 prey

Prop.?  
Solve  
=

General:

$$\rightarrow \frac{dH}{dt} = H F(H, P)$$

$$\frac{dP}{dt} = P G(H, P)$$

H → prey

P → predator.

→ Lotka-Volterra  
killing

(linear)

Mathes

$$\frac{dH}{dt} = \gamma H - \alpha HP$$

conservation

$$\frac{dP}{dt} = -\frac{dP}{t} + \alpha HP$$

depletion

linear osc

LCO

→ Linear oscillations / growth damping → Show

$\rightarrow \underline{U_E} \rightarrow DW - ZF$

$\downarrow$  departure, L-V.

$$\frac{1}{2} \frac{dF}{dt} = \delta_0 E - \alpha_1 \underline{E^2} - \alpha_2 U E$$

$DW \rightarrow$  ~~predator-prey~~ prey

$$\frac{1}{2} \frac{dU}{dt} = -\mu U + \alpha_2 U E$$

Shear Flow  $\rightarrow$  predator.

$\rightarrow$  For 1 pred, 1 prey:

$$\frac{dH}{dt} = H \circ F(H, P)$$

\*\*

$$\frac{dP}{dt} = P G(H, A)$$



Kolmogorov Thm. (1936), based on  
Poincaré-Bendixson Theorem.

→ Predator - Prey systems of

form  $\begin{cases} F \\ G \end{cases}$  have either:

- a stable equilibrium point
- a stable limit cycle

$F, G$  continuous, with  
continuous first derivatives

characterize  
scorable syst.

and

$$(i) \frac{\partial F}{\partial P} < 0 \quad \rightarrow \text{rate of prey } \xrightarrow{\text{growth}} \text{decrease as pred } \uparrow$$

or reneway

$$(ii) H \frac{\partial F}{\partial H} + P \frac{\partial F}{\partial P} < 0 \quad \xrightarrow{\text{rate}} \text{prey growth decreases}$$

with population size

$$\frac{\partial G}{\partial P} < 0 \quad \rightarrow \text{rate of increase of predators decreases with}$$

population size of pred.

$$\frac{\partial G}{\partial P} < 0 \quad \rightarrow \text{rate of increase of predators decreases with}$$

population size of pred.

U.i.)  $H(\partial G/\partial t) + P(\partial G/\partial P) > 0$

CS → # predators is increasing function  
of population size

V.i.)  $F(0, 0) > 0$

$$\frac{dH}{dt} = FH$$

CS Prey have positive growth  
rate (~ linear), for small  
population

and

$$\exists A, B, C \quad \frac{\partial f}{\partial t}$$

$$\begin{aligned} \frac{dH}{dt} &= HF \\ &= H[\gamma - \alpha \underbrace{P}_{\gamma} - \beta F] \end{aligned}$$

Vii.)  $F(0, A) = 0$ , with  $A > 0$

$$\boxed{P = \gamma / \alpha}$$

→ can have predator population  
to stop further prey increase,  
for low levels prey

→ predators induced death

Viii.)  $F(B, 0) = 0$ , with  $B > 0$

→ critical prey population beyond  
which prey cannot increase,  
even absent predators!

$$\frac{dH}{dt} = H F$$

$$= H [\gamma - \alpha P - \beta H]$$

$\cancel{H = \gamma/P}$

(\*)  $\rightarrow$  state prey must be able to self-sustain, as for logistic competition, absent predators

$\boxed{B = \gamma \cdot \delta_0 / \delta_1}$

viii)  $\sigma(C, 0) = 0$ , with  $C > 0$

There exists a critical prey size  $C$  that stops further increase in predators.

$\boxed{C \rightarrow M/\delta_2}$

(ix)  $B > C$ , or system collapses.

$\boxed{\text{need } \gamma_0 > \alpha_1 M / \delta_2}$

on flows collapse.

$c_0 = i_{x_1}$  satisfied  $\Rightarrow$

system will have fixed point on LCO solution structure!

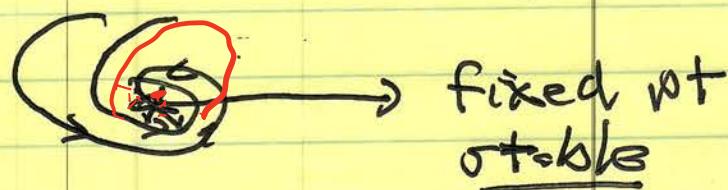
N.B.:

→ (c) → ix.) are criteria for  
sustainable, 'sensible', system

→ some can be  $\geq$ ,  $\leq$

→ Criteria for (stationary state)  
attractor

i.e. Fixed pt



fixed pt  
stable



LCO



F.P.

fixed pt is

unstable.

but

but LCO  
encircles ct

e.g. 'Mexican hat' potential

∴ if system satisfies h-thy and  
fixed pt unstable, then LCO

# LCO $\rightarrow$ time delay

21.

→ Sense of effect - Model variations  
starting

From Lotka-Volterra:

$$\frac{dH}{dt} = H [a - \alpha P]$$

- all  $> 0$

$$\frac{dP}{dt} = P [-b + \beta H]$$

- often  $\kappa = \beta$

1)  $a \rightarrow r(1 - H/k)$   $\leftarrow$  Logistic

carrying capacity  $(\delta_{LF} - \delta_{FH})$

stabilizing

2)  $- \alpha HP \rightarrow$  prey removal by predator

$$\rightarrow kP(1 - e^{-\alpha t})$$
 destabilizing  
(oscillating)

$$\rightarrow \frac{kHP}{H+D}$$
 saturation of predation  
at 0

$\rightarrow$  Transitions  $\rightarrow$  IC Thm. Conditions

- back to PD 7/4 model:

Fixed pts:

- trivial:  $E = U = 0$

- [No flow]:  $E = \gamma_0/d_1$   
 $\{$  mode  $\}$   $U = 0$

- Flow:  $E = U/d_2$   $\rightarrow$  turbulence set by flow damping

QCs:

$$U = (\gamma_0 - \frac{\alpha_1 M}{d_2}) / d_2$$

$B \gg C$

$$\frac{\gamma_0}{d_1} > \frac{M}{d_2}$$

$\rightarrow$  flow set by  
turb growth

Superficially counterintuitive

$\Rightarrow$  signature of pred-prey system,

$\Rightarrow$  observed, in GL simulations  
Liu et al. '98

$$E \sim \sqrt{ } \quad \text{v}$$

$$\text{Fluctn} \rightarrow F_L + F_{\text{fl}}$$

Clearly:

- transition for:

$\kappa \text{ thin} \Rightarrow$  cond. melt. F.R. transitions

$\frac{\gamma_0}{d_1, d_2} > \frac{C_B}{C}$  (B > C criterion)

Power Threshold

$\sim P_{\text{thresh}} \approx \gamma_0 / (D P) \sim Q / D$   
 $\Rightarrow$  gets critical power, flux

- can linearize about 2 fixed pts (non-trivial) HW

No flow:

$$(\gamma_0/d_1, 0)$$

"Mode"  $\rightarrow$  eigenmode of P-Psyat.

flow

~~B > C~~

$$\begin{cases} \gamma = -C_U - d_2 \gamma_0 / d_1 \rightarrow U \\ \gamma = -\gamma_0 \rightarrow E \end{cases}$$

Observe:

- E mode always heavily damped.

- U mode [soft] near transition

$$\gamma \rightarrow 0$$

slow

ext. fibo

- $\gamma_u \rightarrow 0$  at threshold
  - $\Rightarrow T_{\text{transition}} \rightarrow \infty$
  - (this is a bifurcation)
- 2<sup>nd</sup> order transition, soft mode is signature.

Now, since  $|\gamma_E| \gg |\gamma_u|$ ,

- $E$  will relax to equilibria / fixed point much faster than  $U$
- $\Rightarrow$  - 'Slave' fluctuates to flow, i.e.

$$\frac{1}{2} \frac{dE}{dt} \rightarrow 0 = (\gamma_0 - \alpha_1 E - \alpha_2 U) E$$

$$E = (\gamma_0 - \alpha_2 U) / \alpha_1$$

$$\frac{1}{2} \cdot \alpha_1 \frac{dU}{dt} = - \mu U + \alpha_2 U E$$

$$= - \mu U + \alpha_2 U (\gamma_0 - \alpha_2 U) / \alpha_1$$

$\Sigma$ , system described by flow:

$$\frac{1}{2} \frac{dU}{dt} = \left( \frac{\alpha_2}{\alpha_1} \gamma_0 - 1 \right) U - \frac{\alpha_2^3}{\alpha_1} U^2$$

$U_1^{1/2}$

$$\frac{1}{2} \frac{dU_1}{dt} = \left( \frac{\alpha_2}{\alpha_1} \gamma_0 - 1 \right) U_1 - \frac{\alpha_2^3}{\alpha_1} U_1^2$$

Logistic Eqn.

→ Logistic eqn, with growth threshold

+ Diffn → [Fisher Eqn.] → [propagating front]

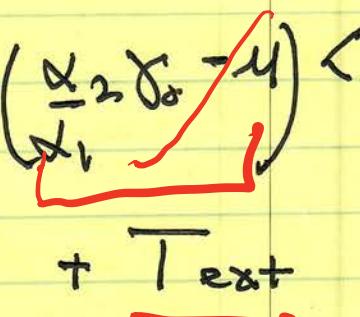
→ If write for  $\langle U_1' \rangle$

$$\frac{d\langle U_1' \rangle}{dt} = \left( \frac{\alpha_2}{\alpha_1} \gamma_0 - 1 \right) \langle U_1' \rangle - \frac{\alpha_2^3}{\alpha_1} \langle U_1' \rangle^3$$

→ T O GL

→ 2nd order transition

and can add external torque/drive:

$$\frac{d}{dt} \langle V_+ \rangle = \left( \frac{\alpha_2 \delta_0 - u}{\alpha_1} \right) \langle V_+ \rangle - \frac{\alpha_2^2}{\alpha_1} \langle V_+ \rangle^3 + \overline{T_{ext}}$$


- Bias? (aka J-TEXT experiments)

Here

- after response in permanent near criticality / away from criticality

L&L stat. Mech

⇒ Topic for serious further work...

K Thm → transitions

$B > G$

→ Further:

- LCO's - time delays  
next class.
- Model developments

→ Near criticality → Noise

$$\frac{1}{2} \frac{d^2 u}{dt^2} = \left( \frac{\chi_2}{\chi_1} \gamma_0 - u \right) u - \frac{\chi_2^2}{\chi_1} u^2$$

$$Q = \bar{Q} + \tilde{Q}$$

$$\gamma_0 \sim \gamma_{CP} \sim \gamma_0 (\tilde{Q}/Q)$$

$$\sim \bar{\gamma} + \tilde{\gamma}$$

as  $Q = \bar{Q} + \tilde{Q}$

Mean  
heat flux

↳ bursts, noise  
i.e. pdf edge  
heat flux

$P_{\text{d}f}[v^2]$

Versatility 28

"Const"  
are  
not const.

→ Begs the question of  
how deal with noise ?

✓

⇒ Stochastic mode) → PdF  
↓  
Fokker-Planck Eqn.

$$D_v = \frac{\beta}{m} v_{th}^2 \sim \frac{|\vec{f}_0|^2 T_{ac}}{m^2}$$

### - Multiplicative Noise (Simple)

Consider logistic Eqn  $\rightarrow$  populations

$$\frac{dN}{dt} = N(k - N)$$

↓      ↪ saturation  
 Malthusian      by competition       $N = \# \text{ of population}$   
 growth       $\sim N^2$   
 (exponential)

$$x_{n+1} = \alpha x_n(1 - x_n)$$

Logistic Map

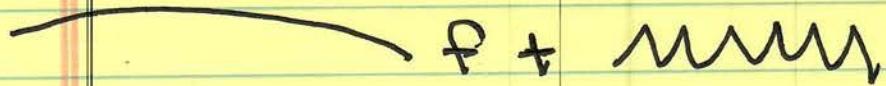
$N=0, N=k$  are fixed pts

Now, could consider variability in  $k$ , and treat as stochastic variable

$$\frac{dN}{dt} = N(k_0 + \tilde{\delta}(t) - N) + \tilde{x}(t)$$

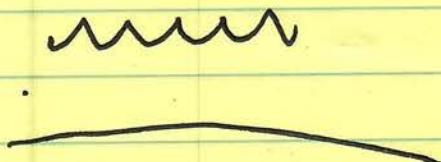
↓  
 variability in resources      ↓  
 ⇒ multiplicative noise      external input variability  
 → rate      ⇒ additive noise.

c.e. additive:

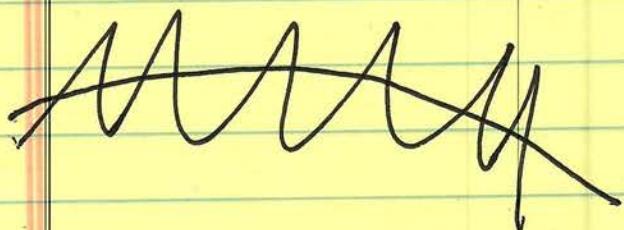


noise on top  
deterministic  
base

=



multiplicative:



multiples by  
fast, random  
quantity

How treat?

$f(N, t) \rightarrow$  population  
pdf

→ Fokker-Planck Equation  $f_t f(N)$

→ here  $\langle \tilde{y}(t) \tilde{y}(t') \rangle = k_B T_{\text{ao}} \delta(t - t')$

Delta correlated for simplicity.

N.B. This is a "textbook model".

→ additive, as usual

$$\langle \tilde{x} \tilde{f} \rangle = 0$$

Then :  $\frac{dN}{dt} = N(k_0 + \tilde{f}(t) - N) + \tilde{x}$

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$$\frac{\partial f(N, t)}{\partial t} = -\frac{\partial}{\partial N} \left[ (k_0 N - N^2) f(N, t) - \frac{\partial}{\partial N} (D f(N)) \right]$$

For  $D_j$ :

$$\langle \Delta N \Delta N \rangle = \int_{-\infty}^t dt' \int_{-\infty}^t dt'' \langle \tilde{f}(t') \tilde{f}(t'') \rangle > N^2$$

↑ !

$$+ \int_{-\infty}^t dt' \int_{-\infty}^t dt'' \langle \tilde{x}(t') \tilde{x}(t'') \rangle >$$

$$= k_0 l^2 T_{ac} N^2 + + D l^2 T_{ac}$$

$\overset{\text{if}}{\text{Nonlinearity in } D}$

→ one trademark feature  
of multiplicative noise

→ Note:  $N \rightarrow \sigma \Rightarrow D \rightarrow 0$

Rate variation  $\Rightarrow$  Pd f spread  
in proportion to population.

→ Additive correction significant at low  $N$ .

Now, ignoring additive correction,

$$\partial_t f(N) = -\frac{\partial}{\partial N} \left\{ (k_0 N - N^2) f(N) \right. \\ \left. - \frac{\alpha}{2N} \left( \frac{180 P_{ac}}{2} N^2 f(N, t) \right) \right\}$$

is Fokker-Planck Equation

and stationarity:

$$N(k_0 - N) f(N) = \frac{\partial}{\partial N} \left( \frac{180 P_{ac}}{2} N^2 f(N) \right)$$

Norm

$$\delta CN = \frac{1}{C \cdot n} [2(k_0/r^2) - 2] e^{-2N/\tau^2}$$

$$r^2 = \frac{180f^2 T_{ac}}{2}$$

↑  
Power

{ exponential  
tail }

Need  $k_0^2 > (f^2/2)^2 \Leftrightarrow f > 1/k_0$

i.e.  $\left\{ k_0 > \frac{180f^2 T_{ac}}{2} \right\}$

$\tau \rightarrow 0$   
to avoid log.  
singularity

Physics of  $k_0 > \frac{180f^2 T_{ac}}{2}$  ?

Convenient to linearize around  
fixed point:

$$\frac{dN}{dt} = (k + \tilde{f} - N)N$$

Validity ?

$$N = k_0 + \tilde{n}$$

$$\frac{d\tilde{n}}{dt} = (k_0 + \tilde{n})(k_0 + \tilde{f} - k_0 - \tilde{n})$$

$$\cong k_0 \tilde{f} - k_0 \tilde{n} + O(n^2)$$

$$\partial_r f(n) = -\frac{\partial}{\partial n} \left[ -k_0 n f(n) - \frac{\partial}{\partial n} \left( \frac{k_0^2 \delta^2 \gamma_{\text{sc}}}{2} f(n) \right) \right]$$

$\downarrow$   
linearize about fixed pt.

$$= -\frac{\partial}{\partial n} \left[ -k_0 n f(n) - \frac{\partial}{\partial n} \left( \frac{k_0^2 r^2}{2} f(n) \right) \right]$$

$\Rightarrow$  zero flux / stationarity:

$$f(n) = \underset{\text{const}}{\sum} \exp \left[ -n^2 / k_0 r^2 \right]$$

Valid for:  $\langle (\tilde{r}/N_0)^2 \rangle = \langle \tilde{r}^2 / k_0 \rangle < 1$

Now  $\langle \tilde{r}^2 \rangle = \frac{4}{2} k_0$

$\therefore \langle (\tilde{r}/k_0)^2 \rangle < 1 \Rightarrow \left\{ \frac{r^2}{2k_0} < 1 \right.$

→ again ;  $\sigma^2 < 2 k_B$

i.e. fluctuations small compared  
to logistic growth.

N.B. :

- can determine time evolution
- can get moments
- spatio-temporal dynamics,