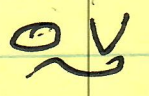


PKU - HUST Lectures (2020-2021)



→ Modeling Dynamical Systems, for MFE, C.C.F. after Biophysics Applications

→ a useful (I hope) methods

mini-course (minimal practical knowledge)

→ Emphasis on non-linear dynamics, space-time scales

→ Topics: ⇒ MFE

i.) Feedback loops, predator-prey (time)

ii.) transition evolution and propagation (space-time) Fisher, FN
traffic flow

iii.) avalanches, kursty flux (Burgers; → wav/solitons)

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 O J H

P.D., Frisch, γ, 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

(a) Feedback loops and Predator-Prey Systems

Plan:

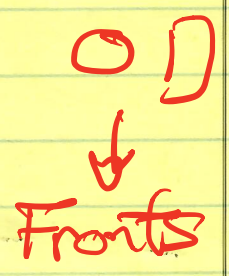
- OV and Motivation } Drift Waves + Shear Flow
- system
- Constructing the System - { Very Simple Examples }
- Solving
- Lessons from Ecology (c.f. May)
 - General structure; constraint
 - Aside: Bigger Models, stability & complexity
General structure }
 - Kolmogorov Theorem; → feasible structure
 - { Fixed pts. vs. Cycles
 - { Feedback
 - Time Delays → Cycles

- Fluctuating Environments and Statistical Approaches.

→ Physics Motivation

→ Fluctuating Logistic - a simple example.

→ Applications



- Looking Ahead.

I) QV and Basis

→ Fundamental Problem(s)

- Drift waves + zonal shear flow

- L → H transition

- Natural candidate for Predator - Prey Model

2 animal

- what?

hares

H = prey

P = predator

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

↑ rates

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

P(t), H(t)

(00)

~~00~~

can extend to 2m x 2m.

Lotka-Volterra

- what does it mean?

H → Σ - fluctuation intensity, energy

P → $U = \dot{V}^2$ - mean square \dot{V}_E^2

00? $\Sigma = \langle \Sigma \rangle$
 $\dot{V}^2 = \langle \dot{V}^2 \rangle$ } avg. over thin layer.
[ignores optical evolution]

- why?

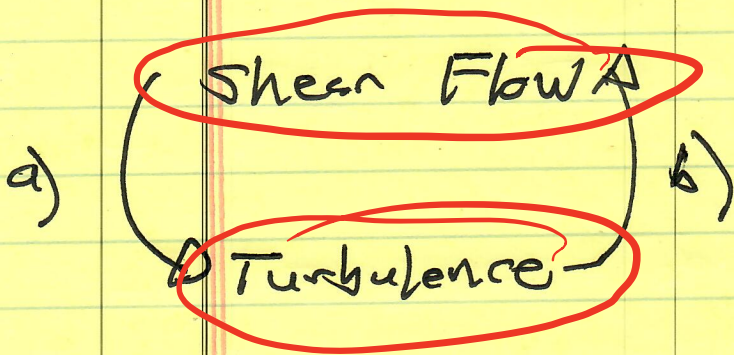
2 populations

$\langle \dot{V} \rangle$
 \dot{V}

→ self-regulation (feedback)

→ symmetry constraints Δ constructed

1) self-regulation



a) shear stabilizes, controls turbulence
 → everyone knows ...

b) turbulence drives Flow, via Reynolds stress

$$\underline{\underline{J}} = \overset{\uparrow}{J_{2-1}} + \overset{\uparrow}{J_{35}} + \overset{\uparrow}{J_{4}}$$

$$\partial_t \langle v_{\perp} \rangle = - \nabla_{\perp} \langle \tilde{v}_r \tilde{v}_{\theta} \rangle \quad \rightarrow \quad \underline{\underline{\nabla \cdot J}} = 0$$

n.b. more precisely:

from vorticity eqn. ($\nabla \cdot J = 0$)

$$\overset{\text{Deriv}}{=} \partial_t \left(\frac{1}{r^2} \partial_r^2 \langle \phi \rangle \right) = - \partial_r \langle \tilde{v}_r \overset{\text{smirnovity}}{\partial_r^2} \tilde{\phi} \rangle + \dots$$

net polarization change

Flux of polarization charge

$$\langle \tilde{v}_r \tilde{\eta}_{i\theta} \rangle - \langle \tilde{v}_r \tilde{\eta}_{e\theta} \rangle$$

(generalizes Rosenbluth Hinton 77)

NOTE

ExB shear flow

1 sym.

Reynolds force



but: $\langle \tilde{u}_r \tilde{v}_z^2 \rangle = \partial_n \langle \tilde{u}_r \tilde{v}_z \rangle$

HW:

show this!
(G.I.) Taylor identity (100+ yrs old)

Upshot:

- Reynolds stress, force ↔ polarization charge flux

* - thm: McIntyre & Wood:

R. Wood

"PV mixing" [ie. polarization charge flux] + 1 direction of symmetry ⇒ Z.F. generation.

How reconcile? ⇒ feedback loop

2) Symmetry * (already encountered)

Why is ZF 'special'?

(P.O., I², H '05)

ZF is mode ($k_{\parallel} \neq 0$, $\eta, \mu \rightarrow 0$)
 of:

~~minimal~~ minimal inertia = easily excited

$k_{\parallel} \rightarrow 0$

HW

$(1 + k_{\perp}^2 \rho^2)$

\downarrow
 H-M

\rightarrow $k_{\perp}^2 \rho^2$

\downarrow
 2D Fluid

*

Minimal transport (none!) *

(zero)

$\eta = 0$

$\mu = 0$

\Rightarrow

$\tilde{V}_r = 0$

can't relax $\partial n, \partial T$, etc.

Minimal damping *

(no Landau damping - R.H., 47)

Symmetry \rightarrow P-P

⇒ Symmetry, constraints + consideration
of self-regulation

⇒ Z. F. can grow/be excited only
via 'Feeding off' of turbulence, yet
shears turbulence, thus regulating
it.

⇒ natural problem for predator-prey
formulation!

→ What does it look like? ~~E + U~~
Cef. PD, et. al. '94) Mixing length processes

$$\frac{1}{2} \frac{dE}{dt} = \underbrace{\alpha_0 E}_{\substack{\uparrow \\ \text{growth} \\ \text{turb.}}} - \underbrace{\alpha_1 E^2}_{\substack{\uparrow \\ \text{self} \\ \text{setn.}}} - \underbrace{\alpha_2 U E}_{\substack{\uparrow \\ \text{coupling} \\ \text{key}}}$$

$$\frac{1}{2} \frac{dU}{dt} = -U U \downarrow \text{Flow damping} + \alpha_2 U E$$

see table in paper for coeffs

→ N.B. - simplest

→ An infinity of extensions:
most inst.

~ δ_0 → $\bar{\delta}_0$ (DP) → $\bar{\delta}_0$ ($a/L_p = a/L_{part}$)

↑
coupling to profile evolution

$\partial_t \langle \rho \rangle + \partial_n \langle \tilde{v}_n \tilde{\rho} \rangle = nQ = n(\bar{\Phi} + \tilde{\Phi})$
↑
noise

$\partial_t \langle \rho \rangle + \partial_n \chi_0 \Sigma \partial_n \langle \rho \rangle = n(\bar{Q} + \tilde{Q})$

OD $\partial_t \langle \rho \rangle + D_0 \Sigma \langle \rho \rangle = n(\bar{\Phi} + \tilde{\Phi})$

Melroy, P.D. 2002 → Kim, P.D. 03

~ $V_E = \frac{DIP}{n} + V_0$ (Carreras, McKi)

~ IPD, turbulent flow, particles, stoch. field ...

uses guide to Arad-Arey extension

$$u = \langle v_{\perp}^2 \rangle$$

B.) How is it constructed?

Key term: $\nabla_{\perp} u \cdot \mathbf{E} \rightarrow$ Reynolds Coupling

Point: Reynolds work transfers energy Flow \leftrightarrow Fluctuations

HW: 2 ways

$$\text{Now, } \partial_t \langle v_{\perp} \rangle = - \nabla_n \langle \tilde{v}_n \tilde{v}_{\perp} \rangle$$

$$\partial_t \int d\mathbf{r} \frac{\langle v_{\perp}^2 \rangle}{2} = \int \langle \tilde{v}_n \tilde{v}_{\perp} \rangle \nabla_n \langle v_{\perp} \rangle$$

but

$$\partial_t \sum_{\text{fluctu}} \epsilon_k + \partial_t \int d\mathbf{r} \frac{\langle v_{\perp}^2 \rangle}{2} = 0$$

so

$$\partial_t \epsilon_k = - \int d\mathbf{r} \langle \tilde{v}_n \tilde{v}_{\perp} \rangle \nabla_n \langle v_{\perp} \rangle$$

$$\text{Formally } \partial_t \langle \epsilon_k \rangle = - \langle \tilde{v}_n \tilde{v}_{\perp} \rangle \langle v_{\perp} \rangle$$

Layman average

What of stress?

→ 'Shear - Eddy Tilting Feedback Loop'

$\langle \tilde{v}_r \tilde{v}_l \rangle = -\frac{c^2}{B^2} \sum_{\underline{k}} \underline{k}_x \underline{k}_y |\tilde{\phi}_k|^2$ zip

~ $\langle k_x k_y \rangle$ → eddy orientation!

i.e. $|\tilde{\phi}_k|^2 = F(\cancel{k_x}) g(\cancel{k_y})$

f, g even functions

$\langle k_x k_y \rangle = 0$ → no stress
mode structure

but → Shear flow can induce correlation!

i.e. $\frac{d\underline{k}}{dt} = -\frac{\partial}{\partial \underline{x}} (\omega + \underline{k} \cdot \underline{v})$ (Snell's Law)

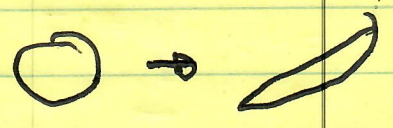
$\frac{dk_x}{dt} = -\frac{\partial}{\partial x} (\omega + \underline{k} \cdot \underline{v}_E) = -k_y \langle v_E \rangle$

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

~ shearing coordinate

Goldreich
L-B '65

$$t \sim \tau_0$$



$$k_x = k_x^{(0)} - k_0 \langle v_E \rangle' \tau_0$$

$$\langle k_x k_y \rangle = \langle k_x^{(0)} k_y \rangle - k_0^2 \langle v_E \rangle' \tau_0$$

intrinsic correlation

turbulence induces correlation

mode structure (esp. radial propagation!)

Peardstein
- Benf

HW: Calculate for slab DW.

15

$$\langle \tilde{u}_r \tilde{u}_r \rangle = t \frac{c^2}{B_0^2} \sum_k k_0^2 v_E \tau_{0k} |\Phi_k|^2$$
$$\approx \langle v_r^2 \rangle v_E \tau_0$$

and

$$D_t \langle \Sigma_k \rangle \approx - \frac{\langle \dot{v}^2 \rangle}{2} \tau_{\text{cor}} \langle v_{\perp} \rangle^2$$

\downarrow note $\sim \langle v_{\perp} \rangle^2$ dependence
 $\sim -D \langle \dot{v} \rangle^2$

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

- $\langle v_{\perp} \rangle^2$ is natural U 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.i.e. $\omega \sim \omega_*$ vs. $\omega \sim \omega$

then, have for wave action density

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

$\left(\frac{E}{\omega}\right) \rightarrow R(H)$
 $\rightarrow \partial I^2 H, \text{ (Good)}$

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

\uparrow
relaxation

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

\rightarrow HW \rightarrow show

and usual quadratic crank \Rightarrow
 see (DI² H '05)

$\partial_t \langle N \rangle - \frac{\partial}{\partial k_n} D_k \frac{\partial \langle N \rangle}{\partial k_n} = \langle C(N) \rangle$

$D_k = \sum_l \frac{k \omega^2 N_{E_2 l^2} \tau}{\omega_{E_1} \omega_{E_2}} \rightarrow$ induced diffusion random shearing

Ray chaos

Note: ω wave packet

$$\frac{c}{\Omega - \gamma(\omega) + i\delta} \leftrightarrow \frac{c}{\omega - kv}$$

$\rightarrow \Omega/\gamma$

$$\partial_{k_r} \rho_k \partial_{k_r} \sim \frac{k_0^2}{k_r^2} \sum_{\mathbf{z}} |V_{Ez}|^2 \rho^2$$

$$\sim \frac{k_0^2}{k_r^2} \sum_{\mathbf{z}} |V_{Ez}|^2 \rho^2$$

$$\langle V^2 \rangle$$

as before.

HW

work out commutator

then:

$$\langle \mathcal{E} \rangle = \omega \langle N \rangle$$

$$\partial_t \langle \mathcal{E} \rangle = - \int dk \rho_k \partial_{k_r} \langle N \rangle \frac{\partial \omega}{\partial k_r}$$

$$= - \int dk \frac{\partial \omega}{\partial k_r} \rho_k \partial_{k_r} \langle N \rangle$$

ω
1st order

$\langle \mathcal{E} \rangle \sim \langle \mathcal{E} \rangle$

backward wave

$$\frac{\partial \langle N \rangle}{\partial k_r} \langle \mathcal{E} \rangle$$

fundamental

Same structure

$$\partial_t \mathcal{E} = - \langle v \rangle^2 \mathcal{E}$$

Aside

C. ↓

Lessons From Ecology (cf. May)

→ What can be said about this type of system?
 ↓
 1 predator - 1 prey

Prop.?
Solns

General:

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

H → prey

P → predator

→ Lotka-Volterra breeding (linear)

Methods

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

conservation

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

linear osc
 LCO

→ Linear oscillations / growth-decay → show

$$\rightarrow \underline{U_0} \rightarrow DW - ZF$$

↓ departure, L-V.

$$\frac{1}{2} \frac{dE}{dt} = \gamma_0 E - \alpha_1 E^2 - \alpha_2 UE$$

DW → ~~prey~~ prey

$$\frac{1}{2} \frac{dU}{dt} = -u u + \alpha_2 UE$$

Shear Flow → predator.

→ For 1 pred, 1 prey:

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

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

**

⇒

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

→ Predator - Prey systems of form KK have either:

- a stable equilibrium point
- a stable limit cycle

ills

LF

F, G continuous, with continuous first derivatives

characterize
oscillate syst.

and

CS

(i) $\partial F / \partial P < 0$

→ rate of prey ^{growth} decrease as pred ↑

- or renewal

(ii) $H (\partial F / \partial H) + P (\partial F / \partial P) < 0$

→ prey growth decreases with population size

CS
of resource depletion

(iii) $\partial G / \partial P < 0$

→ rate of increase of predators decreases with population size of pred.

CS
of collapse

cu.) $H(\partial G/\partial H) + P(\partial G/\partial P) > 0$

CS
maintain

→ # predators is increasing function of population size

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

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

CS

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

$\gamma = f$

and

$F(A, B, C) \text{ o/t}$

$\frac{dH}{dt} = H F$
 $= H [\gamma - \alpha P - \beta H]$

vii) $F(0, A) = 0$, with $A > 0$

$P = \sigma/\alpha$

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

(*)

→ Predation induced satiation

viii.)

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

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

(*)

$$\frac{\partial H}{\partial t} = H F$$

$$= H [\alpha - \alpha_1 P - \beta H]$$

$$H = \alpha / \rho$$

(*) → state prey must be able to self-saturate, as for logistic competition, about predators $[B = \alpha_0 / \alpha_1]$

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

There exists a critical prey size C that stops further increase in predators. $\frac{\partial P}{\partial t} = P \theta = P [-\mu + \alpha_2 H]$

~~...~~ $[C \rightarrow \mu / \alpha_2]$

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

$[need \alpha_0 > \alpha_1 \mu / \alpha_2]$ → on flows collapse.

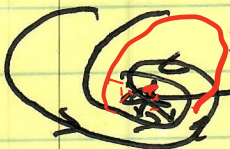
(i) = (x) satisfied ⇒
system will have fixed point on
LCO solution structure!

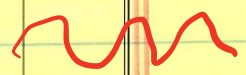
N.B.:

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

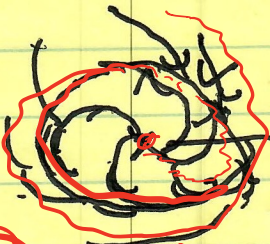
→ some can be \geq , \leq

→ Criteria for (stationary state) attractor

c.e. Fixed pt  → fixed pt stable



LCO

 → Fixed pt is unstable. but LCO encircles it.

but

F.P. ~~unstable~~

c.e. 'Mexican hat' potential

∴ if system satisfies k-Thm and LCO fixed pt. unstable, then

LCO \rightarrow time delay

21.

\rightarrow Sense of effect - Model variations starting from Lotka-Volterra:

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

- all > 0

- often $\alpha = \beta$

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

1) $a \rightarrow r (1 - H/K) + \dots$ Logistic
 \downarrow
carrying capacity (self-stn.)

stabilizing

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

$\rightarrow kP(1 - e^{-\alpha H})$ destabilizing (costly)

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

→ Transitions → K Thm. Conditions

- back to PD 94 model:

Fixed pts:

- travel: $E = U = 0$

- No flow: $E = \gamma_0 / \alpha_1$
 $U = 0$ → self-satn.
 NO Flow

- Flow: $E = U / \alpha_2$ → turbulence set by flow damping

$U = (\gamma_0 - \frac{\alpha_1 U}{\alpha_2}) / \alpha_2$

$\beta > \gamma_0$
 $\frac{\gamma_0}{\alpha_1} > \frac{U}{\alpha_2}$ → Flow set by turb growth

Superficially counterintuitive

⇒ signature of pred-prey system

⇒ observed in GK simulations
 Lm, et. al. '98

$E \sim \sqrt{\quad}$

Flux $\rightarrow F_1 + F_{20}$

Clearly:

- transition for:

$k_{TUM} \Rightarrow$ cond. melt. f.r.
transition

$\gamma_0 > d_1 \mu / d_2$ (circled)
Power Threshold (circled)
CB > C (criterion)

$\sim P_{thres}$, as $\gamma = -1/DP \sim Q/D$ (circled)
 \Rightarrow sets critical power, flux

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

No flow:

$(\gamma_0/d_1, 0)$

"Mode" \rightarrow eigenvalue of P-System

$\gamma = -(\mu - d_2 \gamma_0/d_1) \rightarrow U$ (circled)
 $\gamma = -\gamma_0 \Rightarrow E$ (circled)

Observe:

- E mode always heavily damped.

- U mode 'soft' near transition

$\gamma \rightarrow 0$

slower

expt. τ_{00}

- $\gamma_u \rightarrow 0$ at threshold

$\Rightarrow \tau_{\text{transition}} \rightarrow \infty$ (into a bifurcation).

\sim 2nd order transition, soft mode
is signature.

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

- E will relax to equilibria/fixed point much faster than U

- 'slave' follows to flow, c.e.

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

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

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

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

system described by flow:

$$\frac{1}{2} \frac{dU}{dt} = \left(\frac{\alpha_2}{\alpha_1} \delta_0 - \mu \right) U - \frac{\alpha_2^2}{\alpha_1} U^2$$

$$\frac{1}{2} \frac{dU}{dt} = \left(\frac{\alpha_2}{\alpha_1} \delta_0 - \mu \right) U - \frac{\alpha_2^2}{\alpha_1} U^2$$

Logistic Eqn.

→ Logistic eqn, with growth threshold

+ Diffn → Fisher Eqn. → propagating frontiers

→ if write for $\langle U \rangle'$

$$\frac{d\langle U \rangle'}{dt} = \left(\frac{\alpha_2}{\alpha_1} \delta_0 - \mu \right) \langle U \rangle' - \frac{\alpha_2^2}{\alpha_1} \langle U \rangle'^2$$

→ TOGL

→ 2nd order transition

and can add external torque/drive :

$$\frac{d}{dt} \langle \dot{U}_1 \rangle = \left(\frac{\alpha_2 \gamma_0 - \mu}{\alpha_1} \right) \langle \dot{U}_1 \rangle - \frac{\alpha_2^2}{\alpha_1} \langle \dot{U}_1 \rangle^3 + \text{Torque}$$

- Bias (aka J-TEXT experiments)

- also response in ferromagnet near criticality / away from criticality
Look at next

⇒ Topic for serious further work...

h₂ Thm → transitions B > C

→ Further:

- LCO's - time delays
 next class.

- model developments

→ Near criticality → noise

$$\frac{1}{2} \frac{dU}{dt} = \left(\frac{\alpha_2}{\alpha_1} \delta_0 - \mu \right) U - \frac{\alpha_2^2}{\alpha_1} U^2$$

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

$$\delta_0 \sim \delta(OP) \sim \delta_0 \left(\frac{Q}{D} \right)$$

$$\sim \overline{\delta} + \tilde{\delta}$$

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

↓
 mean
 heat flux

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

$P_{LS}[v^2]$

variability 28

↳
"const"

case
not const.

→ Begs the question of
how deal with noise?

↳

⇒ Stochastic model → Pdf
↓
Fokker-Planck Eqn.

$$D_v = \frac{\beta}{m} v_{th}^2 \sim \frac{|\tilde{f}_0|^2 \tau_{sc}}{m^2}$$

- Multiplicative Noise (Simple)

Consider logistic Eqn \rightarrow Populations

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

\downarrow Malthusian growth (exponential)
 \downarrow saturation by competition $\sim N^2$

$N \equiv$ # of population

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

Logistic Map

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

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

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

\downarrow
 variability in resources

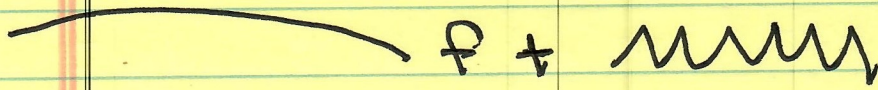
\Rightarrow multiplicative

noise
 \Rightarrow rate

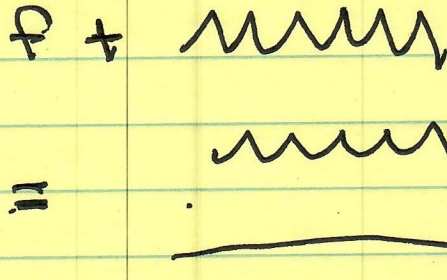
\downarrow
 external input variability

\Rightarrow additive noise.

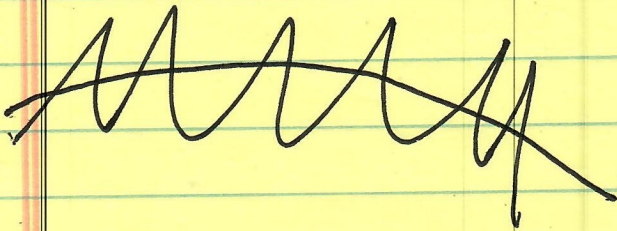
i.e. additive:



noise on top
deterministic
base



Multiplicative:



multiplies by
fast, random
quantity

How treat?

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

\rightarrow Fokker-Planck Equation for $f(N)$

\rightarrow here $\langle \tilde{\gamma}(t) \delta(t') \rangle = \tilde{\gamma}_0^2 \tilde{\gamma}_{00} \delta(t-t')$

Delta correlated for simplicity.

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

\rightarrow additive, e.g. $u(t)u(s)$

$$\langle \tilde{\alpha} \tilde{\alpha} \rangle = 0$$

12.

Then : $\frac{dN}{dt} = N(k_0 + \tilde{\alpha}(t) - N) + \tilde{\alpha}^2$

18

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

For D_j :

$$\langle \Delta N \Delta N \rangle = \int_{dt'}^t \int_{dt''}^t \langle \tilde{\alpha}(t') \tilde{\alpha}(t'') \rangle N^2 + \int_{dt'}^t \int_{dt''}^t \langle \alpha(t') \alpha(t'') \rangle$$

$$= |\sigma_0|^2 \tilde{\tau}_{ac} N^2 + |\alpha_0|^2 \tilde{\tau}_{ac}$$

Nonlinearity in D

→ one trademark feature of multiplicative noise

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

Rate variation \Rightarrow Pdf spread
in proportion to population.

→ Additive correction significant at
low N .

Now, ignoring additive correction,

$$\partial_t F(N) = - \frac{\partial}{\partial N} \left\{ (k_0 N - N^2) F(N, t) - \frac{\partial}{\partial N} \left(\frac{1}{2} \sigma_0^2 \tau_{ac} N^2 F(N, t) \right) \right\}$$

is Fokker-Planck Equation

and stationarity:

$$N(k_0 - N) F(N) = \frac{\partial}{\partial N} \left(\frac{1}{2} \sigma_0^2 \tau_{ac} N^2 F(N) \right)$$

Norm

$$FCM = C \cdot n \left[2(k_0/\sigma) - 2 \right] e^{-2N/\sigma^2}$$

$$\sigma^2 = |\tilde{w}_0|^2 \tau_{av}$$

↑
Power

exponential tail

Need $k_0^2 > (\sigma^2/2)^2 \iff f > 1/n$

i.e. $k_0 > \frac{|\tilde{w}_0|^2 \tau_{av}}{2}$

$n \rightarrow \infty$
to avoid log. singularity

Physics of $k_0 > \frac{|\tilde{w}_0|^2 \tau_{av}}{2}$?

Convenient to linearize around fixed point:

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

Validity ?

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

$$\begin{aligned} \frac{d\tilde{n}}{dt} &= (k_0 + \tilde{n}) (k_0 + f - k_0 - \tilde{n}) \\ &\approx k_0 f - k_0 \tilde{n} + O(\tilde{n}^2) \end{aligned}$$

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$$\partial_t F(n) = -\frac{\partial}{\partial n} \left[-k_0 n F(n) - \frac{\partial}{\partial n} \left(\frac{k_0^2 \sigma^2 \tau^2}{2} F(n) \right) \right]$$

linearize abt fixed pt.

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

⇒ Zero flux / stationarity:

$$F(n) = \sum_{const} C \exp \left[-n^2 / k_0 \tau^2 \right]$$

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

Now $\langle \tilde{n}^2 \rangle = \frac{\tau^2 k_0}{2}$

$$\sigma_0 \langle (\tilde{n}/k_0)^2 \rangle < 1 \Rightarrow \left\{ \frac{\tau^2}{2k_0} < 1 \right.$$

→ again ; $\underbrace{\sigma^2 < 2k_0}$

i.e. fluctuations small compared
to logistic growth.

N.B.:

- can determine time evolution

- can get moments

- spatio-temporal dynamics.