

# Boulder Summer School:

Lecture #1 → Representing Behavior I  
(or: Philosophy & Worms)

→ What business does a theoretical physicist have studying behavior?

(i.e. if not spherically, how should we model a cow?)

→ To bridge scales & extract important coarse-grained descriptions

Q: Why is this absurd?

→ Many scales

→ Far removed from underlying "mechanism"

→ Apparently qualitative

⇒ Need a quantitative language for behavior that we can take seriously

E.g. DNA → base pairs, amino acids, binding affinities, ...

Neurons → membrane voltage, spikes, firing rates, ...

Behavior → ?

→ How do we ~~represent~~ take a complicated creature like an animal & represent it in a manner conducive to experimental & theoretical interrogation?

→ Greet! But how?

⇒ How have biologists attempted this?

1) Paradigms (e.g. rat in a maze or center-out task)

~~→ embody measurement in the experiment~~  
→ embody measurement in the experiment

2) Coarse Measurements (e.g. velocity, # of beam crosses)

→ pre-planned coarse-graining

3) Intuition (e.g. manual annotation)

→ assumes discrete behavioral types & the form they take

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Q: How can we build on these methods to create a rich, yet comprehensible, measurement of an animal's movements?

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~~→ To start, look at data~~

(e.g. elegans, Stephens et al, PLOS Comp Bio, 2008)

Why is this difficult?

→ continuous or discrete?

→ what length scale?

→ what time scale?

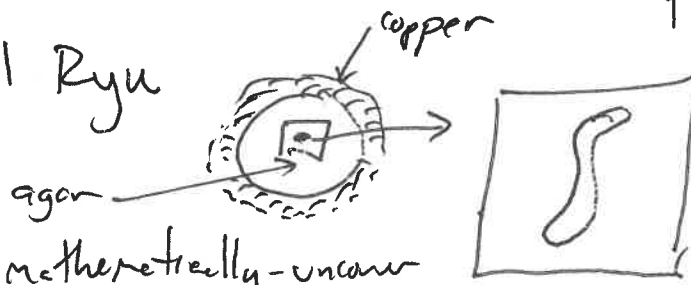
→ How to tell time points apart (distance metric)?

→ start by looking at data

# Crawling Behavior in C. elegans

(Stephens et al, PLoS Comp Bio, 2008)

Experiment: Will Ryu



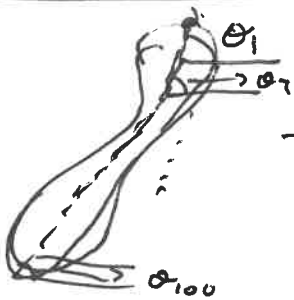
Stephens et al, PNAS 2011

Q: Can we mathematically-uncover the essential dynamics of the worm as it moves?

Choice #1: How to describe the worm?

- options:
- a) center of mass
  - b) mid-point
  - c) centerline
  - d) full 2D area

Try this!  
(connects moment at small ~~body~~ scale to moment at body scale)

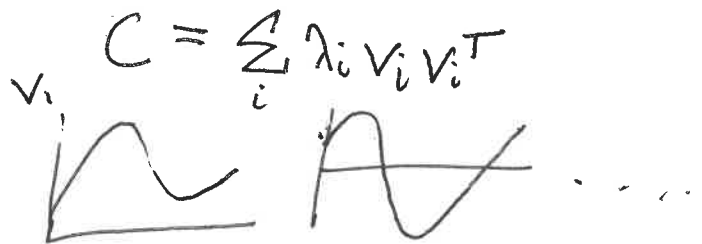
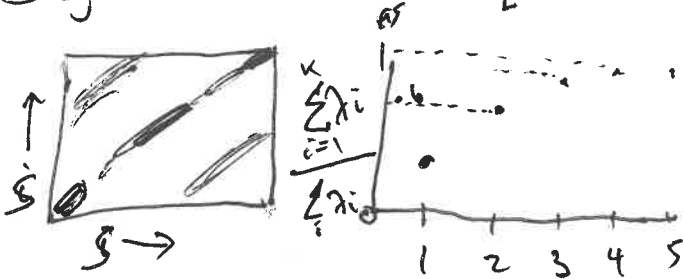


→ set so that  $\langle \theta_i^{(t)} \rangle = 0 \forall t$   
(rotational invariance)

→ another choice

→ Now have 100 time series. Do we need all of them? No!

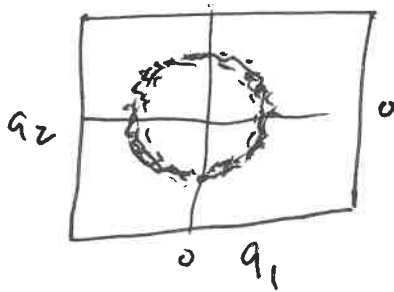
$C_{ij} = \langle \theta_i - \langle \theta_i \rangle \rangle \langle \theta_j - \langle \theta_j \rangle \rangle_T$  → covariance matrix



$$C = \sum_i \lambda_i v_i v_i^T$$

Look at projections onto 1<sup>st</sup> 2 modes:

PPF



$$q_i(t) = (\vec{\sigma}_w - \langle \vec{\sigma}(t) \rangle_t) \cdot \hat{V}_i$$

→ Well-described by a phase variable  $\phi = \tan^{-1} \frac{q_2}{q_1}$

→ One variable describes ~ 1/2 of all postural variance!

Can we write-down a model for the dynamics of this one variable?

Idea: Model as a "Langevin-like" model

$$\frac{d\phi(t)}{dt} = w(t), \quad \frac{dw(t)}{dt} = \overbrace{F[\phi(t), w(t)]}^{\text{Deterministic}} + \overbrace{\sigma[\phi(t), w(t)]\eta(t)}^{\text{Stochastic}}$$

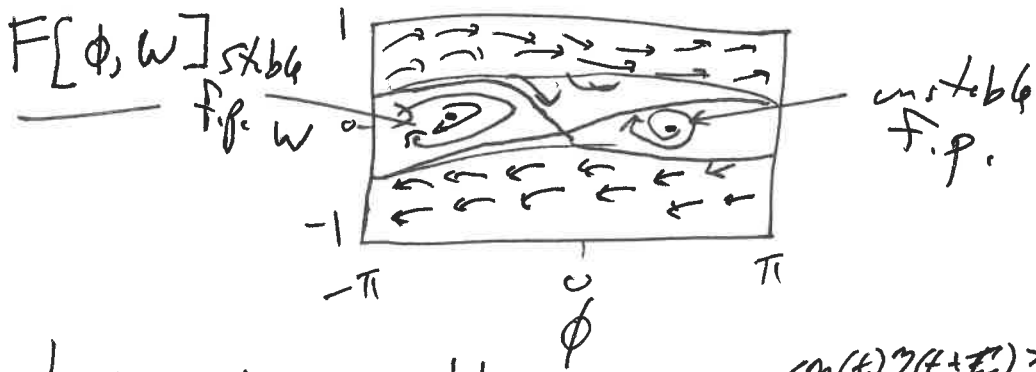
In practice: After some basic filtering & such,  $\langle \eta(t)\eta(t') \rangle = \delta(t-t')$   
calculate  $\phi(t)$  &  $w(t)$  from data

Assume that  $\langle \sigma[\phi(t), w(t)]\eta(t) \rangle = 0$

→ optimal R.M.S. estimate of  $F[\phi(t), w(t)] = \langle \dot{w} | \phi, w \rangle$

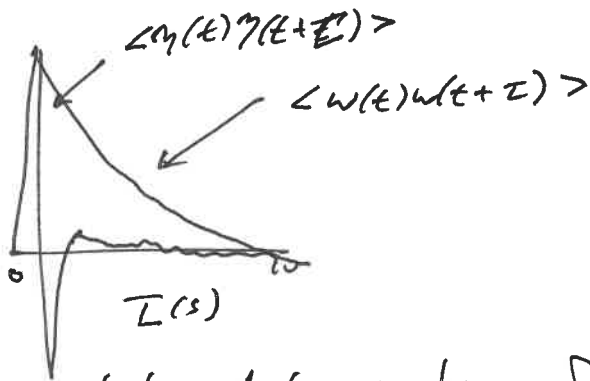
Fit  $F(\phi, w) = \sum_{m=-5}^5 \sum_{p=0}^5 \alpha_{m,p} w^p e^{-im\phi}$  (Fourier series in  $\phi$ , polynomial expansion in  $w$ )

$$\sigma^2(\phi, w) = \langle (\dot{w}(t) - F(\phi(t), w(t)))^2 | \phi, w \rangle_t$$



Look at noise conditions;

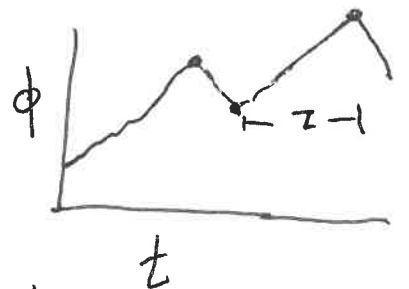
$\tau$  is short  
 $\rightarrow$  we are



happy that we've spread deterministic from stochastic dynamics

But can we predict anything?

1) Reversal Frequency

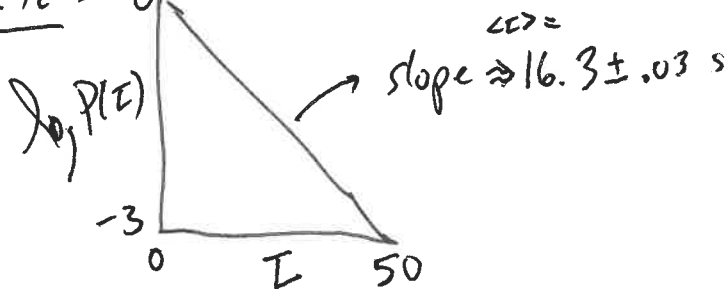


Can we predict the distribution of survival times in the forward state?

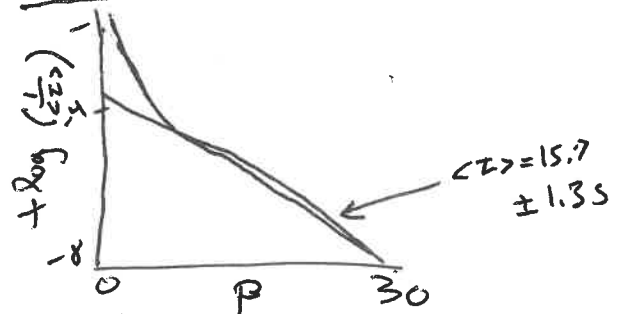
$\rightarrow$  Let's assume that transitions are brief & well-separated in time  $\rightarrow$  should be no memory

$$\Rightarrow P(\tau) \sim \exp\left[-\frac{\tau}{\langle \tau \rangle}\right]$$

Data:



Model: Scale  $\sigma^2(\phi, \omega)$  by  $\beta$



Note: The model didn't even have to have had reversals!

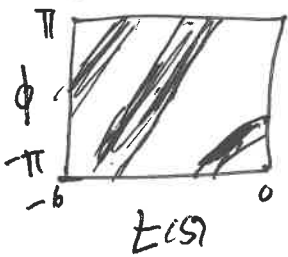
2) Stereotypy  $\rightarrow$  Behaviors performed often & repeatedly

In theories of stochastic barrier crossing, after stereotyped patterns of trajectories occur (if barrier much larger than the noise). We have a "Laguerre" area!

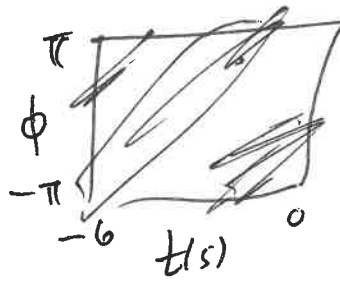
Let  $E = 0 \rightarrow$  time of a trajectory exiting the forward state

Compute:  $p(\phi|t)$  for model & data

Model:



Data:



$\rightarrow$  stereotypy in transition structure emerges naturally from the noise-driven crossing of a dynamical barrier!

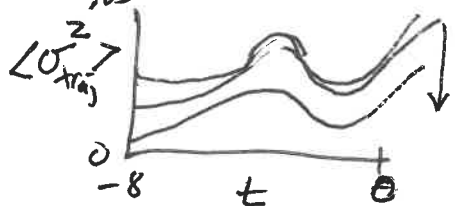
~~Lastly:~~ Lastly: Can we use these measurements to gain insight into biology?

$\Rightarrow$  Worms get hungry. No food on the plate, so more likely to make longer "runs". Break data up into 3 ~10 minute recordings, fit each to  $\frac{dw}{dt} = ( )$

1) All 3 models fit well (i.e. reproduce what we saw before)

2)  $F(\phi, w)$  is roughly unchanged

3)  $\sigma(\phi, w)$  alters



$\rightarrow$  noise decreases with time, implying longer escape times & straighter trajectories

~~Products that~~ Predicts that the worm controls its behavior by controlling noise

# Inspiration & Challenges for Applying to other organisms

## Inspiration

- Use a simple representation of movement ~~the~~ to describe low-d posture
- Reconstructed 2-d dynamical system to describe movements
- Made out-of-sample predictions about behavior & control
- Single time scale emerges

## Challenge

- It's a worm (limbs, other things are more complicated, have complicated units)
- other animals, this is less likely to work (higher-d)
- More difficult because posture ↔ dynamics mapping will be more complicated
- Multiple Time scales in most problems

This is all about how our behavior evolves in time. How does the animal "choose" how to proceed?

→ More next time