Linear response for macroscopic observables in high-dimensional systems

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Joint work with Georg Gottwald

### Linear response theory

Consider a smooth family of deterministic dynamical systems  $x_n = T^{\varepsilon}(x_{n-1})$ , which are mixing with physical invariant measures  $\mu^{\varepsilon}$ .:

$$\mathbb{E}^{arepsilon}[\Psi] := \int \Psi(x) \, \mathrm{d} \mu^{arepsilon}(x)$$

**Linear response theory (LRT):** What is  $\frac{d}{d\varepsilon}\mu^{\varepsilon}\mathbb{E}^{\varepsilon}[\Psi]$ ? (e.g. for Taylor approximations)

 $\ldots$  supposing  $\mathbb{E}^{\varepsilon}[\Psi]$  is differentiable

## LRT in practice

The application of linear response theory to climate systems has met with some success:

- Toy models: Majda et al '07, '10, Lucarini & Sarno '11
- Barotropic models: Bell '80, Gritsun & Dymnikov '99, Abramov & Majda '09
- Quasi-geostrophic models: Dymnikov & Gritsun '01
- Atmospheric models: North et al '04, Cionni et al '04, work of Gritsun and others '02, '07, '10, Ring & Plumb '08
- Coupled climate models: Langen & Alexeev '05, Kirk & Davidoff '09, Fuchs et al '14, Ragone et al '15

# LRT in practice



Chekroun et al., 2014



However:

- Rough responses are known in atmospheric and ocean dynamics (e.g. Chekroun et al. '14)
- The failure of linear response needs very long time series to be visible (Gottwald, W. & Wouters '17)

# LRT in theory

Analytically, we know LRT works in

- Statistical mechanics: Kubo '66
- Stochastic dynamical systems: Hänggi '78, Hairer & Majda '10
- Axiom A (uniformly hyperbolic dissipative chaos): Ruelle '97-8
- Other dissipative systems. . . ?

Baladi and others ('08, '10, '14, '15) proved there is no linear response for quadratic maps, even Whitney differentiability.



#### The question

In this talk we will address the following question:

When and why does linear response occur (for all practical purposes) at macroscopic scales in high-dimensional systems?



### The model

We study simple complex systems:



### The model

We study slightly more complex complex systems:



# The model

We study slightly more complex complex systems:



		macroscopic observables	
microscopic subsystem		uncoupled	coupled
f satisfies LRT	finite $M$		1
	$M \to \infty$		*
$f$ violates LRT with smooth $\frac{\mathrm{d}\nu}{\mathrm{d}a}$	finite $M$	(1)	( <b>イ</b> )
	$M \to \infty$		*
$f$ violates LRT with non-smooth $\frac{\mathrm{d}\nu}{\mathrm{d}a}$	finite $M$	X	( <b>イ</b> )
	$M \to \infty$	X	X



We will derive reductions for mean-field dynamics and discuss (very rich) LRT properties.

### Uncoupled case

System parameters:  $a^{(j)}, j = 1, ..., M$  sampled from measure  $\nu$ Dynamics:

$$q_n^{(j)} = f(q_{n-1}^{(j)}; a^{(j)}, \varepsilon), \ j = 1, \dots, M$$

Observable:

$$\Psi_n = \frac{1}{M} \sum_{j=1}^M \psi(q_n^{(j)})$$



Each subsystem  $q^{(j)}$  evolves independently: suppose they have physical measures  $\mu^{a^{(j)},\varepsilon}$  and are mixing.

#### Uncoupled case: expectations

relevant for LRT

Two (nested) ways to take expectations:

- Over dynamics, i.e. initial conditions:  $\mathbb{E}^{\varepsilon}[\cdots]$
- Over dynamical systems, i.e. selection of parameters a<sup>(j)</sup> (if relevant): ⟨ 𝔼<sup></sup>[···] ⟩

# LRT of mean-field $\Psi$

We are interested in behaviour with respect to  $\varepsilon$  of

$$\mathbb{E}^arepsilon \Psi = rac{1}{M}\sum_{j=1}^M \mathbb{E}^arepsilon[\psi(q^{(j)})]$$

The  $q^{(j)}$  evolve independently of each other so at statistical equilibrium,

$$\mathbb{E}^{arepsilon}[\psi(q^{(j)})] = \int \psi(q) \mathrm{d} \mu \int^{\mathsf{a}^{(j)},arepsilon}(q)$$
 only depends on  $a^{(j)}$ 

### LRT of mean-field $\Psi$

Because the  $a^{(j)}$  are randomly selected, a CLT in  $\langle \cdot \rangle$  gives

$$\mathbb{E}^{arepsilon}\Psi=rac{1}{M}\sum_{j=1}^{M}\mathbb{E}^{arepsilon}[\psi(q^{(j)})]=ar{\Psi}^{arepsilon}+rac{1}{\sqrt{M}}\eta^{arepsilon}+o(1/\sqrt{M})$$

where  $\eta^{\varepsilon}$  is a mean-zero Gaussian process in  $\varepsilon\textsc{,}$  and

$$ar{\Psi}^arepsilon = \langle \mathbb{E}^arepsilon [\psi(q)] 
angle = \iint \psi(q) \, \mathrm{d} \mu^{m{a},arepsilon}(q) \, \mathrm{d} 
u(m{a})$$

So response of mean-field  $\Psi$  is  $\bar{\Psi}^{\varepsilon}$  plus small correction for finite ensemble size.

$$ar{\Psi}^arepsilon = \langle \mathbb{E}^arepsilon [\psi(q)] 
angle = \iint \psi(q) \, \mathrm{d} \mu^{m{a},arepsilon}(q) \, \mathrm{d} 
u(m{a})$$

- Clearly if microscopic subsystems satisfy LRT then so does  $\bar{\Psi}^{\varepsilon}$ .
- On the other hand if the microscopic subsystems violate LRT and ν is discrete (e.g. ν = δ<sub>a0</sub>), then Ψ<sup>ε</sup> will not have LRT.

If  $\nu$  is smooth (e.g.  $\frac{d\nu}{da} \in BV$ ), then averaging over  $d\nu(a)$  can give "collective" linear response of microscopic systems that may violate LRT:

• Easy case: If  $f(\cdot; a, \varepsilon) = f(\cdot; a + K\varepsilon)$ :

$$\frac{\mathrm{d}\bar{\Psi}^{\varepsilon}}{\mathrm{d}\varepsilon} = \int \frac{\mathrm{d}}{\mathrm{d}\varepsilon} \int \psi(q) \,\mathrm{d}\mu^{a+\varepsilon}(q) \,\mathrm{d}\nu(a)$$
$$= \int \frac{\mathrm{d}}{\mathrm{d}a} \int \psi(q) \,\mathrm{d}\mu^{a+\varepsilon}(q) \,\mathrm{d}\nu(a)$$
$$= -\iint \psi(q) \,\mathrm{d}\mu^{a+\varepsilon}(q) \,\mathrm{d}\left(\frac{\mathrm{d}\nu}{\mathrm{d}a}\right)$$

 $\implies$  LRT holds

- If  $f(\cdot; a, \varepsilon)$  is a family of (analytic) unimodal maps:
  - These maps obey LRT along topological conjugacy classes (Ruelle '09);
  - Avila *et al* ('03) conjectured that topological conjugacy classes of these maps have a uniformly analytic codimension-one lamination.

This may imply  $\bar{\Psi}^{\varepsilon}$  has linear response.



Smooth family of unimodal maps:

$$f(q; a, \varepsilon) = (a + 4\varepsilon q(1 - q))q(1 - q),$$
  
 $u \sim \operatorname{Cosine}(3.75, 0.05)$ 





# LRT of $\eta^{\varepsilon}$

What about finite M correction  $\eta^{\varepsilon}$ ? Suppose the microscopic variables violate LRT with  $C^{\alpha}$  ( $\alpha < 1$ ) response but  $\bar{\Psi}^{\varepsilon}$  satisfies LRT. Then

$$\begin{split} \left\langle (\eta^{\varepsilon} - \eta^{\varepsilon_0})^2 \right\rangle &= \left\langle (\mathbb{E}^{\varepsilon} [\psi(q^{(j)})] - \mathbb{E}^{\varepsilon_0} [\psi(q^{(j)})])^2 \right\rangle - (\bar{\Psi}^{\varepsilon} - \bar{\Psi}^{\varepsilon_0})^2 \\ &= \mathcal{O}(|\varepsilon - \varepsilon_0|^{2\alpha}) - \mathcal{O}(|\varepsilon - \varepsilon_0|^2) = \mathcal{O}(|\varepsilon - \varepsilon_0|^{\alpha})^2. \end{split}$$

Hence  $\eta^{\varepsilon}$  is  $C^{\alpha}$  a.s., so violates LRT.

# LRT of $\eta^{\varepsilon}$

Thus, for finite M we only get "approximate" LRT (when microscopic subunits do not have LRT).



#### Macroscopic reduction

What about the *dynamics* of  $\Psi_n$ ? The  $q^{(j)}$ s are independent of each other, so for any n

$$\Psi_n = rac{1}{M} \sum_{j=1}^M \psi(q_n^{(j)})$$

is a sum of independent random variables. Thus

$$\Psi_n = \mathbb{E}^{\varepsilon} \Psi + rac{1}{\sqrt{M}} \zeta_n + o(1/\sqrt{M})$$

where  $\zeta_n, n \in \mathbb{N}$  are mean-zero Gaussian random variables.

### Macroscopic reduction

When  $M \gg 1$ ,  $\zeta$  appears to converge to a stationary Gaussian process.

The autocorrelation function is given by the microscopic subsystems:

$$\mathsf{Cov}[\zeta_m,\zeta_n] = \langle \mathsf{Cov}[\psi(q_m),\psi(q_n)] \rangle$$

so  $\zeta$  has decay of correlations and can be approximated by e.g. an AR process.

Side note: as with mean-fields, variability observables such as  $M(\Psi_n - \mathbb{E}^{\varepsilon}[\Psi])^2$  also have (approximate) LRT.

# Non-coupling system conclusions

- Response of mean-field is at least as smooth as that of microscopic dynamics
- Possible to get LRT (for all intents and purposes) at macroscopic level with microscopic dynamics that violate LRT
- Mean-field dynamics are  $\mathcal{O}(M^{-1/2})$  Gaussian fluctuations about expectation value

### Mean-field coupled case

System parameters:  $a^{(j)}, j = 1, ..., M$  sampled from measure  $\nu$ Dynamics:

$$q_n^{(j)} = f(q_{n-1}^{(j)}; \Phi_{n-1}, a^{(j)}, \varepsilon), \ j = 1, \dots, M$$
$$\Phi_n = \frac{1}{M} \sum_{j=1}^M \phi(q_n^{(j)})$$

Observable:

$$\Psi_n = \frac{1}{M} \sum_{j=1}^M \psi(q_n^{(j)})$$



# Externally-coupled system

System parameters:  $a^{(j)}, j = 1, ..., M$  sampled from measure  $\nu$ External driver:  $d_n$ Dynamics:

$$q_n^{(j)} = f(q_{n-1}^{(j)}; d_{n-1}, a^{(j)}, \varepsilon), \ j = 1, \dots, M$$
$$\Phi_n = \frac{1}{M} \sum_{j=1}^M \phi(q_n^{(j)})$$



Observable:

$$\Psi_n = \frac{1}{M} \sum_{j=1}^M \psi(q_n^{(j)})$$

Suppose  $q^{(j)}$  have time-dependent physical measures  $\mu_n^{d,a^{(j)},\varepsilon}$  with decay of correlations.

#### Externally-coupled system

We can apply the same CLT ideas, so e.g.

$$\langle \mathbb{E}^{\varepsilon}[\Phi_n|d] \rangle = \iint \phi(q) \, \mathrm{d} \mu_n^{d, a^{(j)}, \varepsilon}(q) \, \mathrm{d} \nu(a)$$

which only depends on  $(d_m)_{m < n}$ . We have

$$\Phi_n = \langle \mathbb{E}^{\varepsilon}[\Phi_n|d] \rangle + \frac{1}{\sqrt{M}} \tilde{\eta}_n^{d,\varepsilon} + \frac{1}{\sqrt{M}} \tilde{\zeta}_n^d + o(1/\sqrt{M})$$

where the process  $\tilde{\zeta}$  is now non-stationary, and  $\tilde{\eta}^{\varepsilon}$  depends on time.

**Ansatz:** if  $M \gg 1$ , the coupled system can be approximated by setting  $d_n \equiv \Phi_n$ .



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This gives the macroscopic reduction:

$$\Phi_{n} = \langle \mathbb{E}^{\varepsilon}[\Phi_{n}|\Phi] \rangle + \frac{1}{\sqrt{M}} \tilde{\eta}_{n}^{\Phi,\varepsilon} + \frac{1}{\sqrt{M}} \tilde{\zeta}_{n}^{\Phi} + o(1/\sqrt{M})$$
  
=:  $F(\Phi_{n-1}, \Phi_{n-2}, \dots; \varepsilon)$  self-generated noise  
usually smaller than  $\tilde{\zeta}$ 

The macroscopic reduction

$$\Phi_n = F(\Phi_{n-1}, \Phi_{n-2}, \dots; \varepsilon) + \frac{1}{\sqrt{M}} \tilde{\eta}_n^{\Phi, \varepsilon} + \frac{1}{\sqrt{M}} \tilde{\zeta}_n^{\Phi} + o(1/\sqrt{M})$$
$$\Psi_n = G(\Phi_{n-1}, \Phi_{n-2}, \dots; \varepsilon) + \frac{1}{\sqrt{M}} \eta_n^{\Phi, \varepsilon} + \frac{1}{\sqrt{M}} \zeta_n^{\Phi} + o(1/\sqrt{M})$$

defines a stochastic dynamical system.



The macroscopic reduction

$$\Phi_n = F(\Phi_{n-1}, \Phi_{n-2}, \dots; \varepsilon) + \frac{1}{\sqrt{M}} \tilde{\eta}_n^{\Phi, \varepsilon} + \frac{1}{\sqrt{M}} \tilde{\zeta}_n^{\Phi} + o(1/\sqrt{M})$$
$$\Psi_n = G(\Phi_{n-1}, \Phi_{n-2}, \dots; \varepsilon) + \frac{1}{\sqrt{M}} \eta_n^{\Phi, \varepsilon} + \frac{1}{\sqrt{M}} \zeta_n^{\Phi} + o(1/\sqrt{M})$$

defines a stochastic dynamical system. Modulo  $\eta$ 's:

- The noise  $\tilde{\zeta}^{\Psi}$  generates (annealed) LRT in the microscopic particles, so this noisy system is  $\sim$ smooth in  $\Phi$  and  $\varepsilon$ .
- So  $\Phi$  obeys LRT for finite *M*.
- Thus so does Ψ.

LRT for unimodal microscopic components,  $\nu = \frac{1}{3}(\delta_{a_1} + \delta_{a_2} + \delta_{a_3})$ :



LRT for unimodal microscopic components,  $\frac{d\nu}{dx} \in C^3$ :



As  $M \to \infty$  we have macroscopic reduction

$$\Phi_n = F(\Phi_{n-1}, \Phi_{n-2}, \dots; \varepsilon)$$
  
$$\Psi_n = G(\Phi_{n-1}, \Phi_{n-2}, \dots; \varepsilon)$$

defines a (smooth) stochastic dynamical system. In particular external forcing washes out over time because of microscopic mixing, so

$$\Phi_n \approx F(\Phi_{n-1}, \Phi_{n-2}, \ldots, \Phi_{n-K}; \varepsilon),$$

i.e. emergent dynamics of  $\Phi_n$  are low-dimensional.

If dynamics converges to equilibrium  $\Phi_n \equiv \bar{\Phi}^{\varepsilon}$  we have

$$\bar{\Phi}^{\varepsilon} = F(\bar{\Phi}^{\varepsilon}, \bar{\Phi}^{\varepsilon}, \dots; \varepsilon) := F_0(\bar{\Phi}^{\varepsilon}; \varepsilon)$$

which is a smooth function if the microscopic subsystems have "collective" LRT. Then,

$$\frac{\mathrm{d}\bar{\Phi}^{\varepsilon}}{\mathrm{d}\varepsilon} = \left(1 - \frac{\partial F_0}{\partial\bar{\Phi}^{\varepsilon}}\right)^{-1} \frac{\partial F_0}{\partial\varepsilon}$$

(+ stability) and hence  $\Phi$  has LRT.

For unimodal microscopic component example,  $\frac{d\nu}{dx} \in C^3$ , we see saddle-node bifurcation:



What are the other possible macroscopic dynamics and do they obey LRT?

- LRT in thermodynamic limit is difficult to study accurately using naive methods: need both long time series *and* very large microscopic ensembles.
- However, suppose a<sup>(j)</sup> ≡ a<sub>0</sub>. We can write system in terms of measures μ<sup>d,ε</sup><sub>n</sub> and Perron-Frobenius operators L:

$$\mu_n^{\Phi,\varepsilon} = \mathcal{L}_{f(\cdot;\Phi_{n-1},\varepsilon)}\mu_{n-1}^{d,\varepsilon},$$
  
 $\Phi_n = \int \phi(q) \,\mathrm{d}\mu_n^{\Phi,\varepsilon}(q),$ 

• For uniformly expanding *f* these equations can be very efficiently approximated with spectral methods (W. '19).

Consider a mean-field-coupled system

$$q_n^{(j)} = g(q_{n-1}^{(j)}; \varepsilon \Phi_{n-1})$$
  
 $\Phi_n = \frac{1}{M} \sum_{j=1}^M \phi(q_n^{(j)}).$ 

In a few lines of code, the limiting macroscopic dynamics can be simulated very accurately using Poltergeist.jl.



For large  $\varepsilon$  we see period doubling bifurcation to chaos:



The attracting  $\Phi$  dynamics look unimodal:



# LRT in thermodynamic limit

We have breakdown of LRT in the thermodynamic limit:



*Side question*: are the structure of the dynamics in the thermodynamic limit hyperbolic? *Answer*: No. There are homoclinic tangencies.

How do we know? Continuation, making use of Poltergeist.jl.







## Conclusions

Various mechanisms by which linear response may emerge *and/or break down* in large coupled chaotic systems:

- Inhomogeneous collections of microscopic subsystems may have a differentiable average response despite individually violating LRT
- Self-generated noise can induce LRT in large but finite systems
- In thermodynamic limit LRT depends on collective microscopic LRT *and* structure of macroscopic dynamics
- Macroscopic dynamics may be non-hyperbolic chaos, violate LRT

#### Further directions

- Study of networks beyond big mean-field couplings
- More rigorous study of some of these phenomena would be very interesting.

### Further details

Wormell, C.L. and Gottwald, G.A., 2019. Linear response for macroscopic observables in high-dimensional systems. arXiv:1907.13490.

### Aside on periodic windows

Unimodal maps have periodic dynamics on a dense (but not full measure) parameter set—i.e., non-mixing. To keep things simple, we avoid this by adding "hidden" dynamics  $r_n^{(j)} \in [0, 1]$ :

 $f(q, r; \boldsymbol{a}, \varepsilon) = \begin{cases} (\tilde{f}(q; \boldsymbol{a}, \varepsilon), 2r), & r \leq 1/2\\ (q, 2r - 1), & r > 1/2. \end{cases}$ 

This makes the unimodal  $q^{(j)}$  dynamics mixing while retaining the same invariant measures.

(N.B. at statistical equilibrium,  $\{r_n \ge 1/2\}_{n \in \mathbb{N}}$  are *i.i.d.* Bernoulli.)

# "Mixing"

If dynamical system  $x_n = f(x_{n-1})$  is mixing with respect to measure  $\mu$  then for all  $w \in L^2(\mu)$  with  $\mathbb{E}[w] = 1$ ,

$$\mathbb{E}[\psi(x_n)w(x_0)] = \int \psi(x_n)w(x_0) \,\mathrm{d}\mu(x_0) \xrightarrow{n \to \infty} \mathbb{E}[\psi]$$

More generally, are going to assume that if  $\tilde{\mu}$  is a "nice" measure,

$$\int \psi(x_n) \,\mathrm{d}\tilde{\mu}(x_0) \xrightarrow{n \to \infty} \mathbb{E}[\psi]$$