Generating functional approach to dynamical systems with colored-noise interactions

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References

- S. Suweis, F. Ferraro, C. Grilletta, S. Azaele, A. Maritan Interactions Physical Review Letters 133 (16), 167101
- F. Ferraro^{*}, C. Grilletta^{*}, A. Maritan, S. Suweis, S. Azaele annealed disorder arXiv:2405.05183

Generalized Lotka-Volterra Systems with Time Correlated Stochastic

Exact solution of Dynamical Mean-Field Theory for a linear system with

Non-linear model for dynamics of ecological community

$$\dot{x}_i = x_i \left(1 \right)$$

- x_i : number of individuals of species i = 1, ..., N
- With N species, $\sim N^2$ interaction parameters
- Popular approach: disordered interactions



$$\dot{x}_i = x_i \left(1 - x_i - \sum_{j \neq i} \alpha_{ij} x_j \right)$$

$$mean(\alpha_{ij}) = \mu/N$$
$$var(\alpha_{ij}) = \sigma^2/N$$

[Bunin, PRE 2017]



- Predicted equilibrium distribution is Gaussian
- But not observed experimentally! Instead power-law, gamma, log-normal, ...
- Lotka-Volterra equations are reasonable.
 Discrepancy might lie in assumptions on interactions:
 - random
 - two-body
 - Gaussian
 - fixed in time ("quenched")
 - instantaneous



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"Annealed" interactions

Interactions as Ornstein-Uhlenbeck processes

$$\alpha_{ij}(t) = \frac{\mu}{N} + \frac{\sigma}{\sqrt{N}} z_{ij}(t)$$

- Correlation time: τ
- Amplitude is chosen to have limits
 - $\tau \rightarrow \infty$: quenched disorder

- $\tau \rightarrow 0$: white noise

 $\langle z_{ij}(t) \rangle = 0$ $\langle z_{ij}(t)z_{ij}(t')\rangle = Q(t-t')$

 $Q(\Delta t) = \frac{1 + 2\tau}{2} \exp(-|\Delta t|/\tau)$ 2τ

Annealed disorder fits data better!



Linear system with annealed disorder

$$\dot{x}_i(t) = h - kx$$



 $x_i(t) + \sum \alpha_{ij}(t) x_j(t)$ j≠i

Dynamical Mean-Field Theory

$$\dot{x}_i(t) = h - kt$$

- Idea:
 - all degrees of freedom are statistically equivalent
 - write a closed equation for a representative one
 - interaction term is random, Gaussian by CLT, time-dependent \Rightarrow Gaussian process

 $\alpha_{ij}(t) + \sum \alpha_{ij}(t) x_j(t)$ i≠i

Generating functional

$$Z[\psi] = \int D[x] \delta(x - x^*) e^{i\psi \cdot x}$$
$$D[x] = \prod_{i,t} dx_i(t) \qquad \delta(x - x^*) = \prod_{i,t} \delta(x_i(t) - x_i^*(t)) \qquad \psi \cdot x = \sum_i \int dt \psi_i(t) x_i(t)$$

- $x_i^*(t)$ are solutions of linear system for a given realization of $\alpha_{ii}(t)$
- If $Z[\psi]$ is known, all information about $x_i^*(t)$ can be obtained taking derivatives w.r.t to ψ_i



Generating functional

$$Z[\psi] = \int D$$

- Take average over disorder $\alpha_{ii}(t)$
- Result is $\langle Z[\psi] \rangle = Z_{eff}^N[\psi]$ with Z_{eff} generating functional of $\dot{x}(t) = h$ -

where $\eta(t)$ is Gaussian noise with

- $\langle \eta(t) \rangle = \mu \langle x(t) \rangle$



$\mathcal{V}[x]\delta(x-x^*)e^{i\psi\cdot x}$

$$-kx(t) + \eta(t)$$

 $\langle \eta(t)\eta(t')\rangle_c = \sigma^2 Q(t-t')\langle x(t)x(t')\rangle$

Dynamical Mean-Field Theory

$$\dot{x}_i(t) = h - kx$$

$$\dot{x}(t) = h$$

- DMFT equation is same as Ornstein-Uhlenbeck SDE
- ...but noise is self-consistent! Much more complicated

 $x_i(t) + \sum \alpha_{ij}(t) x_j(t)$ j≠i $-kx(t) + \eta(t)$

DMFT process is Gaussian

 $\dot{x}(t) = h -$

• Explicit solution is

$$x(t) = x_0 e^{-kt} + \int_0^t ds e^{-k(t-s)} \left(h + \eta(t)\right)$$

- x(t) is Gaussian, since it is a linear combination of Gaussian $\eta(t)$
- So we only need to find mean $\langle x(t) \rangle$ and autocorrelation $\langle x(t)x(t') \rangle$
- Mean is trivial. For autocorrelation...

$$-kx(t) + \eta(t)$$

Autocorrelation: PDE

- Rewrite DMFT equation as $\eta(t) = \dot{x}(t)$
- Take the product $\eta(t)\eta(t')$ and average
- Result is a PDE for the autocorrelation

$$\left[\partial_t \partial_{t'} + k(\partial_t + \partial_{t'}) + k^2 - \sigma^2 Q(t - t')\right] C(t, t') = f(t, t')$$

where

$$f(t,t') = \sigma^2 Q(t-t') \langle x(t) \rangle \langle x(t') \rangle$$



$$(t) + kx(t) - h$$

ion
$$C(t, t') = \langle x(t)x(t') \rangle - \langle x(t) \rangle \langle x(t') \rangle$$

Autocorrelation: Riemann method

$$\left[\partial_t \partial_{t'} + k(\partial_t + \partial_{t'}) + k^2 - \sigma^2 Q(t - t')\right] C(t, t') = f(t, t')$$

- With the transformation $C(t, t') = e^{-k(t+t')}D(t, t')$ this is equivalent to
- Linear hyperbolic PDE: Riemann method Find Riemann function A(s, s'; t, t') which solves

$$\begin{cases} \left[\partial_s \partial_{s'} - \sigma^2 Q(s) \\ A(s, t'; t, t') = 1 \\ A(t, s'; t, t') = 1 \end{cases} \end{cases}$$

 $\left|\partial_t \partial_{t'} - \sigma^2 Q(t-t')\right| D(t,t') = e^{k(t+t')} f(t,t'),$

$$[-s'] A(s, s'; t, t') = 0$$

Autocorrelation: Riemann function

- Make change of variables $u = e^{-s/\tau}$
- With some manipulations Riemann function is

$$A(s, s'; t, t') = \begin{cases} A_0(s, s'; t, t') & s - s' > 0\\ A_0(-s, -s'; -t, -t') & s - s' < 0 \end{cases}$$

with

 $A_0(s, s'; t, t') = J_0 \left(\lambda \sqrt{1 + 1} \right)$

$$\tau, u' = e^{s'/\tau}$$

$$\left(e^{-s/\tau}-e^{-t/\tau})(e^{s'/\tau}-e^{t'/\tau})\right)$$

Autocorrelation: exact solution

• Riemann method gives the autocorrelation as

$$C(t, t') = e^{-k(t+t')} \int_0^t ds \int_0^$$

- We were not able to compute the integral...
- $\int_{0}^{t'} ds' e^{k(s+s')} A(s, s'; t, t') f(s, s')$

Autocorrelation: numerics





Stationary autocorrelation

- At long times DMFT process becomes stationary
- Autocorrelation depends only on difference of times: $C(t, t') = C_{ct}(t t')$
- PDE becomes ODE

$$-C_{st}''(t) + \left[k^2 - \sigma^2 Q(t)\right] C_{st}(t) = \sigma^2 Q(t) \langle x \rangle_{st}^2,$$

which can be solved with standard methods

Stationary variance

• Since
$$C_{st}(t) = \langle x(t)x(0) \rangle - \langle x(t) \rangle \langle x(t) \rangle$$

It turns out that

$$\sigma_{st}^{2} = \left[\frac{{}_{1}F_{2}\left(n;n+1,2n+1;-\lambda^{2}/4\right)}{{}_{0}F_{1}\left(2n;-\lambda^{2}/4\right)-{}_{0}F_{1}\left(2n+1;-\lambda^{2}/4\right)}-1\right]\langle x\rangle_{st}^{2},$$

with

$$n = k\tau$$
 $\lambda = \sqrt{2\tau(1+2\tau)}\sigma$

 $\langle x(0) \rangle$, stationary variance is $C_{st}(0)$

Stationary variance



Stationary variance



non-monotonic in general!

Phase diagram

Variance diverges

Stationary state is reached



Critical variance

 $2nJ_{2n}(\lambda_c) - \lambda_c J_{2n-1}(\lambda_c) = 0$

$$n = k\tau$$
 $\lambda_c = \sqrt{2\tau(1+2\tau)}\sigma_c$



Critical variance

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Critical variance

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Conclusions

- We studied a multidimensional linear system with colored-noise interactions
- Using the generating functional we derived equation for a representative d.o.f.
- We showed that the solution is a Gaussian process
- We derived and solved equations for the mean and autocorrelation, solving exactly the system
- We found some unexpected features!
 - non-monotonic variance
 - re-entrant phase transition

Perspectives

- Possible extensions
 - hierarchical structure in interactions
 - finite connectivity
 - non-Gaussian interactions
- What happens with a combination of quenched and annealed disorder?
- stability of complex systems to time-dependent equilibria?

Can the framework of annealed disorder be used to extend May's result on

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