Iterated Function Systems and Randomly Forced PDEs.

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A Simple Example

• First order system driven by a clocked random pulse sequence:

$$\frac{dv}{dt} = -v + \xi(t)$$
$$\xi(t) = \sum_{p} a_{p}g(t - p\tau)$$

 $\{a_p\} \in \{0,1\}^{\mathbb{Z}^+}$ are the input symbols and g is supported on $(0,\tau)$.

• Symbols input at constant rate, τ^{-1} .



• In the examples, g is a raised cosine.

Randomly Forced First Order ODE (cont.)

- Can be solved using elementary undergraduates.
- The important parameter is au .



A typical response v(t), when $\tau = \log 3$.



Input and output: comparing $\xi(t)$ and v(t).

Sampling the Output

- When processing signals, it is usual to sample the output.
- A simple picture emerges if we sample at the symbol input rate.



The response v(t) (curve) and samples $v(p\tau)$ (dots).



More samples—can you see a pattern?

Sampling the Output

- A pattern becomes clear with a longer time series.
- The samples $\{v(p\tau)\}$ seem to be distributed as a middle thirds Cantor set.



A longer time series—red shows data of previous figure.

From Another Point of View

- Integrate the ODE for one sample period, τ .
- Depending on the input symbol the output changes according to:

$$\begin{array}{rcl} v & \mapsto & \lambda v \\ & v & \mapsto & \lambda v + b \end{array}$$
 where $\lambda = e^{-\tau}$ and $b = e^{-\tau} \int_0^\tau e^t g(t) dt$

- The sampled output is given by random composition of these maps.
- A skew product system: base dynamics a full shift on {0,1}^{ℤ+}; fibre dynamics given by ODE.
- An Iterated Function System (IFS)

Iterated Function Systems

- Let C be the set of nonempty compact subsets of a complete metric space, (X, d)
- Equipped with the Hausdorff metric, C is a complete metric space
- Define the map $F: \mathcal{C} \to \mathcal{C}$

$$FU = \bigcup_{a=1}^{|\mathcal{A}|} f_a U$$

were \mathcal{A} is a finite alphabet and the maps $\{f_a : a \in \mathcal{A}\}$ act on **X**.

Theorem 1 Given that the $\{f_a\}$ are contraction maps, $F : C \to C$ has a unique fixed point K, and for any $A \in C$, the sequence $F^n A \to K$ in the Hausdorff metric.

Parameter Dependence of Attractor



The Basic Model

• The cable equation:

$$RC\partial_t v = \partial_x^2 v - RAv + RI_e$$

with R the resistance per unit length of the conductor, C and A the capacitance and conductance per unit length of the insulation and I_e is the input current.

- IFS model of digital channels.
- The cable equation is used to model (passive) sections of axon or dendrite: where the term Av is a linear approximation of the membrane current.
- Within each τ second interval, input one of a finite number of possible finiteduration pulse—assume no overlap of inputs.

Specifying the Model

• Finite cable $x \in \Omega = [0, l]$ with zero current boundary conditions:

 $\partial_x v(0,t) = 0$ and $\partial_x v(l,t) = 0 \quad \forall t$

- Rescale time $t \mapsto t/RC$ and introduce the dimensionless parameter $\rho = RA$.
- The cable equation (with $J_e(x,t) = RI_e(x,t)$):

$$\partial_t v = \partial_x^2 v - \rho v + J_e \tag{1}$$

• The input $J_e(x,t)$ is supplied as a spatiallycoded, pulse sequence

The IFS Consists of Contractions

- IFS consists of a finite set of maps $\{f_a : L^2(\Omega) \to L^2(\Omega), a \in \mathcal{A}\}$
- Given $v(x,0) \in L^2(\Omega)$ integrate for time τ with input $J_e^a(x,t), t \in (0,\tau)$ corresponding to a symbol $a \in \mathcal{A}$ to obtain $v(x,\tau) \in L^2(\Omega)$ and write

$$v(x,\tau) \stackrel{\text{def}}{=} f_a v$$

Lemma 2 If $\rho > 0$, the $\{f_a : a \in A\}$ are contractions in the $L^2(\Omega)$ norm.

 For two arbitrary states v₁, v₂; by linearity of PDE:

 $\partial_t ||v_1 - v_2||^2 = -2\rho ||v_1 - v_2||^2 - 2||\partial_x (v_1 - v_2)||^2$

from which it follows that:

 $||v_1(\tau) - v_2(\tau)|| = ||(f_a v_1 - f_a v_2)|| \le e^{-\rho\tau} ||(v_1 - v_2)||$

Existence of a Unique Attractor

- Let B be the set of nonempty, closed bounded subsets of (X, d) and extend F to act on B
- Equipped with the Hausdorff metric, \mathcal{B} is a complete metric space and $\mathcal{C} \subseteq \mathcal{B}$

Theorem 3 Given hypotheses in Theorem 1, for any $A \in \mathcal{B}$, $F^n A \to K$ in the Hausdorff metric.

• Proof by approximating $A \in \mathcal{B}$ by an *r*-net $P_r(A)$ and showing that

 $FP_r(A) = P_{r\lambda_0}(FA)$

where $\lambda_0 = e^{-\rho\tau} < 1$ is the contractivity of *F*. Note that by Theorem 1,

$$F^n P_r(A) \to K$$

The Cable Equation Attractor

- So we have a compact attractor, *K*, for the forced cable equation.
- A simple corollary to Theorem 3 shows the attractor is finite-dimensional

Corollary 4 The box-counting dimension and hence the Hausdorff dimension—of K is bounded

$$\dim_{H}(K) \leq \dim_{B}(K) \leq \frac{\log |\mathcal{A}|}{|\log \lambda_{0}|}$$

 This bound is linear in the symbol input rate:

 $\dim_H(K) \le \dim_B(K) \le \frac{\log |\mathcal{A}|}{\rho \tau}$

Being More Explicit

• Write the input as:

$$J_e(x,t) = \frac{1}{2}J_0(t) + \sum_{k=1}^{\infty} J_k(t)\cos(\frac{\pi kx}{l})$$

• with spatio-temporal symbols:

$$J_k(t) = \sum_p J_k^{(a_p)} g(t - p\tau)$$

• Introduce a Fourier series solution:

$$v(x,t) = \frac{1}{2}v_0(t) + \sum_{k=1}^{\infty} v_k(t)\cos(\frac{\pi kx}{l})$$

• Evolution of Fourier coefficients:

$$\dot{v}_k(t) = -(\rho + (\pi k/l)^2)v_k(t) + J_k(t)$$

An Infinite-dimensional Affine IFS

- Gives an infinite-dimensional IFS: $\{v_k \mapsto \lambda_k v_k + b_k J_k^{(a)} : k \in \mathbb{Z}^+, a \in \mathcal{A}\}$
- Where the contractions are given by:

$$\lambda_k = \exp[-\tau(\rho + (\pi k/l)^2)]$$

- and the offsets $b_k J_k^{(a)}$ contain the term: $b_k = e^{\left[-\tau(\rho + (\pi k/l)^2)\right]} \int_0^\tau e^{\left[t(\rho + (\pi k/l)^2)\right]} g(t) dt$
- Later we truncate to get IFSs of affine maps $\tilde{f}_a: \mathbb{R}^n \to \mathbb{R}^n$

$$\tilde{f}_a x = Tx + \beta_a$$

- T is diagonal with elements $\{\lambda_k\}_{k=0}^{n-1}$
- Note that T is independent of the symbol being input.

Dimension of Self-affine Attractors

Theorem 5 (Falconer,Solomyak) Let $\{T_a : a \in A\}$ be linear contractions such that $\max\{||T_a|| : a \in A\} < 1/2$ and let $\{\beta_a \in \mathbb{R}^n : a \in A\}$ be vectors. If K is an affine invariant set satisfying:

$$K = \bigcup_{a=1}^{|\mathcal{A}|} (T_a(K) + \beta_a)$$

Then $\dim_H K = \dim_B K = d(T_1, T_2, \dots, T_{|\mathcal{A}|})$ for almost all $(\beta_1, \beta_2, \dots, \beta_{|\mathcal{A}|}) \in \mathbb{R}^{|\mathcal{A}|n}$ in the sense of $|\mathcal{A}|n$ -dimensional Lebesgue measure.

• $d(T_1, T_2, \dots, T_{|\mathcal{A}|})$ is the singularity dimension (next slide).

Theorem 6 (Solomyak) Let $T_a = T \forall a \in \mathcal{A}$. If the eigenvalues of T are such that all the images $TK + \beta_a$ of the attractor K are pairwise disjoint, then Falconer's formula holds for almost all $(\beta_1, \beta_2, \dots, \beta_{|\mathcal{A}|}) \in \mathbb{R}^{|\mathcal{A}|n}$.

The Singularity Dimension

• Think of the following as a multiplicative interpolation:

 $\phi^s(T) = \sigma_1 \sigma_2 \dots \sigma_{r-1} \sigma_r^{s-r+1}$

where the σ_i are the ordered singular values of T and $r \in \mathbb{Z}^+$ is such that $r-1 < s \leq r$.

• ϕ^s is strictly decreasing and continuous.

• Now we sum over words of length q:

$$\Sigma_q^s = \sum_{a_1,\dots,a_q} \phi^s(T_{a_1} \circ T_{a_2} \circ \dots T_{a_q})$$

- Submultiplicative: $\Sigma_{q_1+q_2}^s \leq \Sigma_{q_1}^s \Sigma_{q_2}^s$.
- Therefore, $\Sigma_{\infty}^{s} = \lim_{q \to \infty} (\Sigma_{q}^{s})^{\frac{1}{q}}$ exists.
- If $\sum_{\infty}^{n} \leq 1$, there is a unique value of s called d—such that $\sum_{\infty}^{d} = 1$.

The Cable Equation Attractor Dimension

- Use an *n*-dimensional truncation (*n* arbitrarily large).
- Then: $\Sigma_q^s = |\mathcal{A}|^q \phi^s(T^q)$
- Where $\phi^s(T^q) = (\phi^s(T))^q$ can be written explicitly.
- It follows taking the limiting geometric mean gives: $\Sigma_{\infty}^{s} = |\mathcal{A}|\phi^{s}(T)$
- The required value of d is the solution of $|\mathcal{A}|\phi^d(T)=1$
- Linear interpolation of the cumulative sum of the ordered list of Lyapunov exponents and $\log |\mathcal{A}|$ (the maximal entropy of the shift invariant measures on $\mathcal{A}^{\mathbb{Z}^+}$).

Testing the Hypotheses

Proposition 7 Consider the IFS consisting of $|\mathcal{A}| = 2$ maps $\tilde{f}_a : \mathbb{R}^n \to \mathbb{R}^n$ where $\tilde{f}_a x = Tx + \beta_a$ and T is diagonal with nonvanishing elements $T_{kk} = \lambda_{k-1}$. Then for sufficiently large n, all the images $TK + \beta_a$ of the attractor K are pairwise disjoint for almost all $(\beta_1, \beta_2) \in \mathbb{R}^{2n}$ in the sense of 2n-dimensional Lebesgue measure.

- For a proof, take the *n*th component of the IFS.
- Problem reduces to case at very beginning of talk
- So no overlap iff $\lambda_n < 1/2$.
- This can always be arranged for large enough *n*.

Numerical Values of the Attractor Dimension

• For almost all forms of input current distribution, we can, therefore, find $\dim_H K$ using a simple numerical root finder.



Red: dim_{*H*} *K* vs τ in the case $\rho = 2.0$ **Blue**: The upper bound from Corollary 4



Red: dim_{*H*} *K* vs τ in the case $\rho = 0.5$ Blue:: The upper bound from Corollary 4

The Effect of Noise

- Things are easier with noise.
- Add i.i.d. random shifts $y_i \in D \subset \mathbb{R}^n$ at each application of a map from the IFS.
- Where the y_i have a.c. distribution with bdd density, supported on an arbitrarily small disc D at the origin.
- For sample path y the attractor is K^y

Theorem 8 (Jordan, Pollicott, Simon) Given a contracting self-affine IFS of the form assumed in Theorem 5. For \mathbb{P} -almost all $y \in \mathcal{D}^{\infty}$ then:

1. if $d(T_1, \ldots, T_{|\mathcal{A}|}) \leq n$ then $\dim_H(K^y) = d(T_1, \ldots, T_{|\mathcal{A}|})$

2. if $d(T_1, ..., T_{|\mathcal{A}|}) > n$ then $m(K^y) > 0$.

Concluding Remarks

- Finite-dimensional attractors for randomly forced, noisy, extended systems.
- Generalise the noise model?
- Generalise to neural systems where the timing is random?
- Also can introduce nonlinearities—maybe need to extend Theorems 1 and 3 to allow for non-contracting flows.
- Information theory—channel capacity?
- Delay embedding for IFSs—results of Robinson.