Reading for this lecture: CMR articles Yeung, TBA, & PRATISP

Projects!

Proposal due in one week.

Projects!

3. Write up your Project Proposal with the following sections. The result should be 2-3 pages long.

- 3a. Goal:What is your primary project goal? What you would like to learn?
- 3b. System: Describe how the dynamical system is nonlinear and time-dependent.
 - What's the state space?
 - What's the dynamic?
 - Why is the system behavior interesting?
- 3c. Dynamical properties: What dynamical properties are you going to investigate?
- 3d. Intrinsic computation properties: What information processing properties are you going to investigate?
- 3e. Methods: What methods will you use? Why are they appropriate?
- 3f. Hypothesis: What is your current guess as to what you will find?
- 3g. Steps: List the appropriate steps for your investigation; for example, read literature, write simulator, do mathematical analysis, estimate properties from simulation, write up report, and so on. 3h. Time: Estimate how long each step will take. Can you complete the project within one month?

Project:

- 1. Topic: Information processing and computation in natural or engineered dynamical system.
- 2. Select in consultation with us.
- 3. Project report presented to class at the end of the term.
- 4. Written report (including code and documentation) due then.
- 5. Website on project with the report, code, and documentation preferred.

Possible topics:

- Estimate information quantities for complex system of your choice
- Survey structural complexity versus entropy for a class of dynamical systems
- Implement CA pattern analysis
- Analyze patterns generated by variant of 1D or 2D spin systems
- Relationship between energy and information
- Maxwell's demon: Energy versus information
- Relationship between intrinsic computation and phase transitions
- Novel computation:
 - Quantum
 - \circ DNA
 - Analog/continuous
 - Stochastic
 - Evolutionary
 - Neural

Possible topics ...

- Review current research on a complex system of your choice; such as,
 - Complex materials, self-assembly in nanotechnology
 - Chemical pattern formation
 - Biological morphogenesis
 - Bioinformatics
 - Economics: Game dynamical systems
- Simulate a self-organizing system:
 - Statistical mechanical model:
 - Ising, Potts, Heisenberg, X-Y, spin glass, and the like.
 - Cellular automata, map lattice, and the like.
 - Population dynamics: Ecological or evolutionary
 - Networks: Neural, Internet, WWW, social, gene expression, ...
 - Transportation networks: Traffic flow, power grid, world trade, ...
- Build an experimental chaotic or pattern-forming system:
 - Electronic circuit
 - Mechanical device
 - Chemical oscillator
 - Video feedback (see JPC articles)

Possible topics ...

- Effect of external noise on
 - Chaotic behavior
 - This or that kind of bifurcation
 - Routes to chaos
- Probability densities:
 - $^{\circ}$ Time evolution of densities for 1D and 2D maps
 - Approximate invariant distributions
 - Convergence to invariant distributions
 - Intrinsic computation analysis of density evolution
- Transform-based analysis of chaos:
 - Fourier analysis
 - Wavelet analysis
- Chaotic encryption

Possible topics ...

- Philosophical review of:
 - Causality
 - \circ Teleology
 - Randomness, including human perception of (cf. Amos Tversky papers)
 - Coincidence, including human perception of (cf. Persi Diaconis papers)
 - Prediction
 - Cybernetics (cf. Wiener biography "Dark Hero of the Information Age")

Examples from past years:

•2012:

- Christina Cogdell & Paul Reichers: Nonlinear Dynamics of Passionflower Tendril Free Coils
- Nichole Sanderson: "Killing" and "Collapsing"—How varying transition probabilities between states alters the statistical and topological properties of probabilistic ε-machines
- Vikram Vijayaraghavan: Complexity and Critical Behavior

•2010:

- Charles Brummitt: Networks Of Chaotic Maps—A New Network Growth Model, Inferring Topology From Symbolic Dynamics
- Luke Grecki: An Algebraic View of Topological ε-Machines
- Paul Riechers: Spatiotemporal Computational Mechanics

• 2009:

- Ryan James: Block Entropy in 1+1 Dimensions
- Ben Johnson: An Introduction to Quantum Computation
- Richard Watson: Information Theoretic Approaches to Spiking Models
- Paul Smaldino: Does Learning Mean a Decrease in Entropy?
- Nicholas Travers: Computational Mechanics of ECAs, and Machine Metrics

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Time (a)symmetry of processes:

Process: $\mathcal{P} \sim \Pr(\overleftrightarrow{X})$

$$\overrightarrow{X} = \dots X_{-2} X_{-1} X_0 X_1 X_2 \dots$$

$$\overrightarrow{X} = \dots X_{-2} X_{-1}$$

$$\overrightarrow{X} = X_0 X_1 X_2 \dots$$

Forward process: Same

$$\overrightarrow{\mathcal{P}} \sim \Pr(\sigma^+ \circ \overleftrightarrow{X})$$

with +t shift: $\sigma^+ \circ \overleftrightarrow{X} = \dots X_{-2} X_{-1} X_0 X_1 X_2 \dots$

Reverse process: Scan measurements in reverse order

$$\overleftarrow{\mathcal{P}} \sim \Pr(\sigma^- \circ \overleftarrow{X})$$

with -t shift: $\sigma^- \circ \overleftarrow{X} = \dots X_2 X_1 X_0 X_{-1} X_{-2} \dots$

Time (a)symmetry of processes ...

Prediction:

Forward entropy rate: $h_{\mu}(\overrightarrow{\mathcal{P}}) = H[X_0|\overleftarrow{X}]$

Retrodiction:

Reverse entropy rate:
$$h_{\mu}(\overleftarrow{\mathcal{P}}) = H[X_{-1}|\overrightarrow{X}]$$

Which time direction is most unpredictable?

Theorem: Entropy rate is time symmetric

$$h_{\mu}(\overrightarrow{\mathcal{P}}) = h_{\mu}(\overleftarrow{\mathcal{P}})$$

Time (a)symmetry of processes ...

Proof sketch:

$$\begin{split} h_{\mu}(\overrightarrow{\mathcal{P}}) &= H[X_{0}|\overleftarrow{X}] \\ &= \lim_{L \to \infty} H[X_{0}|X_{-L+1}, \dots, X_{-1}] \\ &= \lim_{L \to \infty} \left[H[X_{-L+1}, \dots, X_{0}] - H[X_{-L+1}, \dots, X_{-1}] \right] \\ \mathbf{Stationarity:} \ H[X_{-L+1}, \dots, X_{-1}] &= H[X_{-L+2}, \dots, X_{0}] \\ h_{\mu}(\overrightarrow{\mathcal{P}}) &= \lim_{L \to \infty} \left[H[X_{-L+1}, \dots, X_{0}] - H[X_{-L+2}, \dots, X_{0}] \right] \\ &= \lim_{L \to \infty} H[X_{-L+1}|X_{-L+2}, \dots, X_{0}] \\ &= H[X_{-L+1}|\overrightarrow{X}_{-L+2}] \end{split}$$

$$= H[X_{-1} | \overrightarrow{X}]$$
$$= h_{\mu} (\overleftarrow{\mathcal{P}})$$

Time (a)symmetry of processes ...

Theorem: Entropy rate is time symmetric

$$h_{\mu}(\stackrel{\rightarrow}{\mathcal{P}}) = h_{\mu}(\stackrel{\leftarrow}{\mathcal{P}})$$

Both directions are equally (asymptotically) unpredictable.

Time (a)symmetry of processes ...

Excess entropy is time symmetric.

 $\mathbf{E}(\stackrel{
ightarrow}{\mathcal{P}}) = \mathbf{E}(\stackrel{
ightarrow}{\mathcal{P}})$

Proof sketch:

$$I[\overleftarrow{X}; \overrightarrow{X}] = I[\overrightarrow{X}; \overleftarrow{X}]$$

Conclusion: Neither entropy rate or excess entropy detect temporal asymmetry

Time (a)symmetry of processes ...

Forward ϵM :

$$\vec{M} = \vec{\mathcal{P}} / \sim$$

Reverse ϵM :

$$\stackrel{\leftarrow}{M} = \stackrel{\leftarrow}{\mathcal{P}} / \sim$$

Is a process differently structured in forward or reverse time?

Theorem: ϵM need not be time symmetric

$$\stackrel{\leftarrow}{M} \neq \stackrel{\rightarrow}{M} \\ C_{\mu}(\stackrel{\leftarrow}{M}) \neq C_{\mu}(\stackrel{\rightarrow}{M})$$

Time (a)symmetry of processes ... Proof by example:

Misiurewicz parameter in the Logistic map:

First root, r < 4, where critical point is periodic and

 $f^4(\frac{1}{2}) = f^5(\frac{1}{2})$

Find

 $r\approx 3.9277370017867516$

Use binary generating partition.

Time (a)symmetry of processes ...

Example: Misiurewicz parameter for Logistic map ...

Forward machine:



Time (a)symmetry of processes ...

Example: Misiurewicz parameter for Logistic map ...



Time (a)symmetry of processes ...

Measure of time asymmetry: Causal Irreversibility

$$\Xi(\mathcal{P}) = C_{\mu}(\vec{M}) - C_{\mu}(\vec{M})$$

Misiurewicz process irreversibility:

 $\Xi(\mathcal{P}) \approx 0.36$ bits

Time (a)symmetry of processes ...

Causal Irreversibility:

Information to remember for optimally predicting and optimally retrodicting can differ.

Even though the degree of unpredictability is the same in both time directions.



The I-Diagram ...

Three random variables:

 $X \sim \Pr(x)$ $Y \sim \Pr(y) \qquad (X, Y, Z) \sim \Pr(x, y, z)$ $Z \sim \Pr(z)$

Information measures:

 $H[X] \quad H[Y] \quad H[Z] \quad \cdots \quad I[X;Y;Z] \quad \cdots \quad H[X,Y,Z]$

7 atomic information measures.

Information Diagrams for Processes Information diagram for three random variables:



Information Diagrams for Processes Information diagram for three random variables ...



Information Diagrams for Processes Information diagram for three random variables ...



Markov chain: $X \to Y \to Z$ H(Y)H[Y|X, ZH[Z]H(X)H[X|Y,Z]H[Z|X,Y]I[Y; Z|X]I[X;Y|Z]I[X;Y;Z]

Process I-diagrams: Process has an infinite number of RVs! $Pr(\overleftrightarrow{X}) = Pr(\dots, X_{-2}, X_{-1}, X_0, X_1, X_2, \dots)$

Rather: $\Pr\left(\overleftrightarrow{X}\right) = \Pr\left(\overleftarrow{X}\ \overrightarrow{X}\right)$

Start with 2-variable I-diagram and whittle down: Past as composite random variable: XFuture as composite random variable: XInformation measures:

$$\begin{split} H[\overleftarrow{X}] & H[\overrightarrow{X}] & H[\overrightarrow{X},\overleftarrow{X}] \\ H[\overleftarrow{X}|\overrightarrow{X}] & H[\overrightarrow{X}|\overleftarrow{X}] & I[\overrightarrow{X};\overleftarrow{X}] & H[\overrightarrow{X}|\overleftarrow{X}] + H[\overleftarrow{X}|\overrightarrow{X}] \\ \end{split}$$

There are 3 = 2²-1 atomic information measures:

$$\begin{split} H[\overrightarrow{X}|\overleftarrow{X}] & H[\overleftarrow{X}|\overrightarrow{X}] & I[\overrightarrow{X};\overleftarrow{X}] \\ \end{split}$$



Process I-diagram using E-machine:

Start with 3-variable I-diagram and whittle down: Past as composite random variable: XFuture as composite random variable: XCausal states: $S \in S$

Information measures:

$$H[\overleftarrow{X}] \hspace{0.1cm} H[\overrightarrow{X}] \hspace{0.1cm} H[\mathcal{S}] \hspace{0.1cm} \cdots \hspace{0.1cm} I[\overrightarrow{X};\overleftarrow{X};\mathcal{S}] \hspace{0.1cm} \cdots \hspace{0.1cm} H[\overrightarrow{X},\overleftarrow{X},\mathcal{S}]$$

There are 7 (= $2^3 - 1$) atomic information measures.



ε-Machine I-diagram:



Information Diagrams for Processes What is $H[\overrightarrow{X}|\mathcal{S}]$? Unpredictability: $H[\overrightarrow{X}^{L}|\mathcal{S}] = Lh_{\mu}$ **Proof Sketch:** $H[\overrightarrow{X}^L|\mathcal{S}] = H[\overrightarrow{X}^L|\overleftarrow{X}]$ $= H[X_0X_1\ldots X_{L-1}|\overleftarrow{X}]$ $= H[X_1 \dots X_{L-1} | \overleftarrow{X} X_0] + H[X_0 | \overleftarrow{X}]$ $= H[X_1 \dots X_{L-1} | \overleftarrow{X}] + H[X_0 | \overleftarrow{X}]$ $= H[X_{L-1}|\overleftarrow{X}] + \dots + H[X_1|\overleftarrow{X}] + H[X_0|\overleftarrow{X}]$ $= LH[X_0|\overleftarrow{X}]$ $= Lh_{\mu}$

ε-Machine I-diagram:



Information Diagrams for Processes What is Mystery Wedge? $H[\mathcal{S}|\vec{X}]$ Uncertainty of causal state given future. Implications? Recall Bound on Excess Entropy: $\mathbf{E} \leq C_{\mu}$ Proof sketch: $\mathbf{E} = I[\overleftarrow{X}; \overrightarrow{X}]$ $= H[\overrightarrow{X}] - H[\overrightarrow{X}|\overleftarrow{X}]$ $= H[\overrightarrow{X}] - H[\overrightarrow{X}|\mathcal{S}]$ am the (Mystery Wedge! $= I[\overrightarrow{X}; \mathcal{S}]$ $= H[\mathcal{S}] - H[\mathcal{S}|\overrightarrow{X}]$ $\leq H[\mathcal{S}]$ $= C_{\mu}$

Wedge is the inaccessibility of hidden state information! $H[S|\overrightarrow{X}] = C_{\mu} - \mathbf{E}$ Wedge controls Internal - Observed

Information Diagrams for Processes What is $H[\overleftarrow{X}|\overrightarrow{X}] + H[\overrightarrow{X}|\overleftarrow{X}]$?

Recall distance: d(X, Y) = H[X|Y] + H[Y|X]

So
$$H[\overrightarrow{X}|\overleftarrow{X}] + H[\overleftarrow{X}|\overrightarrow{X}]$$

is the distance between the past and the future!

How related to eM?
From I-Diagram:
$$H[\overleftarrow{X}^{L}|\overrightarrow{X}^{L}] = H[\overleftarrow{X}^{L}|\mathcal{S}] + H[\mathcal{S}|\overrightarrow{X}^{L}]$$

 $H[\overrightarrow{X}^{L}|\overleftarrow{X}^{L}] = H[\overrightarrow{X}^{L}|\mathcal{S}] = h_{\mu}L$
So: $d(\overleftarrow{X}^{L}, \overrightarrow{X}^{L}) = H[\overleftarrow{X}^{L}|\overrightarrow{X}^{L}] + H[\overrightarrow{X}^{L}|\overleftarrow{X}^{L}]$
 $= H[\overleftarrow{X}^{L}|\mathcal{S}] + H[\mathcal{S}|\overrightarrow{X}^{L}] + h_{\mu}L$
Retrodiction Mystery Prediction
Wedge Error

Information Diagrams for Processes Mystery wedge? $H[S|\overrightarrow{X}]$

Retrodiction?

Need extended framework for analyzing: Temporal asymmetry Reverse time view of process

Next week: Mixed states! A key tool.

Reading for next lecture: CMR articles TBA PRATISP IACP IACPLCOCS