

Dynamic operads: Compositional adaptive systems

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TOPOS
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Outline

1 Introduction

- Contextualizing this work
- Compositional adaptive systems
- Plan for the talk

2 Interfaces and arrangements

3 Dynamic operads

4 Conclusion

Categorical systems theory

I've been funded by DoD since 2008 to develop *Applied Category Theory*.

- CT is the mathematics of compositional structures and relationships.
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The idea is to separate structure and content, letting content vary.

- The algebraic notion of *group* is a structure: the group itself can vary.
- Structure-preserving maps and operations let you probe and build.
- These are often “unreasonably effective” for implementation and reuse.

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Today: discuss categorical systems theory. What *structure* is prevalent?

Compositional adaptive systems

A system is a collection of interacting parts that form an integrated whole.

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I'm interested in systems that adapt to circumstances, based on feedback.

- Adaptive control is achieved by adjusting control param's in real-time.
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- We also see it in living systems: they adapt to feedback.
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Rather than “complex” adaptive systems: *compositional* ones.

- Baking complexity into the name precludes studying simple systems.
- It'd be like studying $\mathbb{N} - \{0, \dots, 100\}$. It's good to include 0.
- We see compositionality everywhere. Let's bake that in instead.

Plan for the talk

For the remainder of the time I will:

- Explain key notions: interfaces and dynamic arrangements;
- Hint at mathematical definition: dynamic operads;
- Show examples: neural nets and compositional classical mechanics;
- Conclude.

Outline

1 Introduction

2 Interfaces and arrangements

- Formulating systems
- General theory of interfaces
- Arrangements: static and dynamic

3 Dynamic operads

4 Conclusion

How are we conceptualizing systems

I want a formalism that bakes in compositionality.

- We often see systems of systems of systems.
- The behavior of the whole is not determined by that of the parts, but...
- ...it is determined by the behavior of the parts and *how they interact*.
- New collectively-reinforcing patterns can *emerge* as the parts interact.
- What emerges becomes basic behavior at the next compositional level.

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- They interact with other systems through some sort of *interface*.
- This offers a low-dimensional projection of their state space.
- The interface not only describes how information leaves the system,...
- ...but also how information enters the system, parameterizing behavior.
- What can the system express, and what can impress upon the system?

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What's a broad working definition for the general structure of interfaces?

Mathematics of interfaces

For this talk, we will formalize interfaces as “bundles” in a general sense.

- Pick a category $\mathcal{M}$ with products, e.g. sets or smooth manifolds.
- Pick a notion of bundle $V: \mathcal{M}^{\text{op}} \rightarrow \mathbf{Cat}$.
- E.g. for V you could use “vector bundles on manifolds” $E \rightarrow B$.
- Or you could use submersions of manifolds or maybe maps in $\mathcal{M}$.

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The theory is pretty agnostic about how you choose $\mathcal{C}, V$.

- Rule 0: given $f: M' \rightarrow M$, need “pullback” map $f^*: V(M) \rightarrow V(M')$
- Rule 1: need $V(M_1) \times V(M_2) \xrightarrow{\otimes} V(M_1 \times M_2)$...i.e. V lax monoidal.
- An *interface* is just a pair (M, V) with $M: \mathcal{M}$ and $V: V(M)$.

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This interface tells us how information will go to and from the system.

- Suppose a system is housed in an interface (M, V) .
- Think of M as all the outputs, the ways the system expresses its state.
- For each expression, $m : M$, think of the fiber m^*V as all inputs.

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Example: $\mathcal{M} = \mathbf{Mfd}$, $V(M) = \text{“vector bundles on } M\text{”}$.

- Example interfaces: (T_M^*M) and $(T_M M)$, (co)tangent bundles.
- The system outputs points $m: M$ and inputs (co)tangent vectors at m .

Arrangements 1 of 3

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- An *interface* is a pair $\binom{V}{M}$ where $M : \mathbf{Mfd}$ and $V \rightarrow M$ is a v.bundle
- Output space is M ; given $m : M$, input space is fiber $V_m := m^*(V)$.
- E.g. for $V = \mathbb{R}^m \times \mathbb{R}^n$, we're outputting $x : \mathbb{R}^m$ and inputting $y : \mathbb{R}^n$.
- For arb. $M : \mathbf{Mfd}$ and $V := T^*M$, output m , input covector $x : T_m^*M$.
- Later we will put open dynamical systems in these interfaces $\binom{V}{M}$.

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We will consider static arrangements now; later: dynamic arrangements.

- We'll make a system of systems by arranging interfaces in an interface.
- That is, we'll define a notion of map $\binom{V_1}{M_1} \otimes \cdots \otimes \binom{V_k}{M_k} \begin{smallmatrix} \longleftarrow \\ \longrightarrow \end{smallmatrix} \binom{W}{N}$.
- First, let $\binom{V_1}{M_1} \otimes \binom{V_2}{M_2} := \binom{V_1 \times V_2}{M_1 \times M_2}$, e.g. $\binom{TM}{M} \otimes \binom{TN}{N} \cong \binom{T(M \times N)}{M \times N}$.

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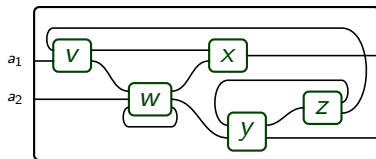
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So what do we take to be a map $\binom{V}{M} \begin{smallmatrix} \longleftarrow \\ \longrightarrow \end{smallmatrix} \binom{W}{N}$?

- Define it to be a pair $(f, f^\sharp)$ where $f : M \rightarrow N$ smooth and...
- ... $f^\sharp : f^*W \rightarrow V$. That is, pullback W to M and then map to V .
- Next we'll explain what this means and give an example.
- Then we'll discuss dynamic arrangements.

Arrangements 2 of 3

How should we think of an arrangement $(\begin{smallmatrix} V_1 \\ M_1 \end{smallmatrix}) \otimes \cdots \otimes (\begin{smallmatrix} V_k \\ M_k \end{smallmatrix}) \begin{smallmatrix} \hookleftarrow \\ \hookrightarrow \end{smallmatrix} (\begin{smallmatrix} W \\ N \end{smallmatrix})$?



The above is an arrangement of type $(\begin{smallmatrix} \mathbb{R}^2 \\ \mathbb{R}^2 \end{smallmatrix}) \otimes (\begin{smallmatrix} \mathbb{R}^3 \\ \mathbb{R}^3 \end{smallmatrix}) \otimes (\begin{smallmatrix} \mathbb{R}^2 \\ \mathbb{R}^1 \end{smallmatrix}) \otimes (\begin{smallmatrix} \mathbb{R}^2 \\ \mathbb{R}^2 \end{smallmatrix}) \otimes (\begin{smallmatrix} \mathbb{R}^1 \\ \mathbb{R}^2 \end{smallmatrix}) \begin{smallmatrix} \hookleftarrow \\ \hookrightarrow \end{smallmatrix} (\begin{smallmatrix} \mathbb{R}^2 \\ \mathbb{R}^2 \end{smallmatrix})$

- Here, we write $(\begin{smallmatrix} \mathbb{R}^n \\ \mathbb{R}^m \end{smallmatrix})$ to denote the trivial bundle on $\mathbb{R}^m$ with fiber $\mathbb{R}^n$.
- The map comes in two pieces: first, $f: \mathbb{R}^2 \times \mathbb{R}^3 \times \mathbb{R}^1 \times \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}^2$.
- Check the diagram: $f(v_1, v_2, w_1, w_2, w_3, x_1, y_1, y_2, z_1, z_2) := (x_1, y_2)$.
- Second, a map $f^\sharp: f^*(\mathbb{R}^2) \rightarrow \mathbb{R}^2 \times \mathbb{R}^3 \times \mathbb{R}^2 \times \mathbb{R}^2 \times \mathbb{R}^1$.
- Check the diagram: $f^\sharp(v_1, v_2, w_1, w_2, w_3, x_1, y_1, y_2, z_1, z_2, a_1, a_2) := (z_2, a_1, v_2, a_2, w_3, v_1, w_1, z_1, w_2, y_1)$
- Inner systems trade information between themselves and outside world.

We'll gener'ze in two ways: non-trivial bundles and dynamic arrangements.

Arrangements 3 of 3

Even static arrangements are much more general than wiring diagrams.

- Any map of manifolds $f: M_1 \times \cdots \times M_k \rightarrow N$ induces

$$(T_{M_1}^*) \otimes \cdots \otimes (T_{M_k}^*) \xleftarrow{\quad} (T_N^*)$$

- Reason: given f , we get $T^*f: T^*N \rightarrow T^*(M_1) \times \cdots \times T^*(M_k)$.

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But we want dynamic arrangements: arrangements that can change.

- Want that what flows through the system can change the arrangement.
- Companies: communication among people can change the structure.
- Cyberphysical systems: signals can change the way energy flows.
- Deep learning: training data can change the weights.
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Let's consider the running example of manifolds and vector bundles.

- Define a dynamic arrangement from $\binom{V}{M}$ to $\binom{W}{N}$ to consist of:
- ...a choice of manifold X and a map $\binom{V}{M} \otimes \binom{TX}{X} \begin{matrix} \longleftarrow \\ \longrightarrow \end{matrix} \binom{W}{N}$.
- Given a curve in M and a field $N \rightarrow W$, get time-varying v.field on X .
- Why call this dynamic? Ans. 1: if $X = \mathbb{R}^0$ recover static arrangement.
- Ans. 2: The arrangement is param'd by $x: X$, and this changes.

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3 **Dynamic operads**

- Dynamic operads
- Examples
- Compositional adaptive systems

4 Conclusion

Dynamic operads

Now we have interfaces and arrangements; how do they compose?

- A (typed) operad $\mathcal{O}$ is an algebraic structure for multi-ary operations.
- They come up in logic, programming, tensor networks, grammars, etc.
- To define an $\mathcal{O}$: specify its types, operations, and composition rules.

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When $\mathcal{O}$'s operations form a “blah”, it's called a “blah”-enriched operad.

- Unadorned, assume $\mathcal{O}$ has a *set* of operations, so it's **Set**-enriched.
- But $\mathcal{O}$ could have operations that vector spaces.
- For any mon'l cat'y $\mathcal{D}$ (e.g., vector spaces), have $\mathcal{D}$ -enriched operads.

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In our case, we will be interested in dynamically-enriched operads.

- Rather than a set of operations, we have a dynamical system...
- ...each of whose states forms an operation.
- What we've been calling “arrangements” are the operations.
- Dynamic arrangements $\binom{V}{M} \rightleftarrows \binom{W}{N}$ change in time.
- That is, there is a time-action on the space of operations.

Examples

We've proven that this sort of structure comes up in diverse applications.

- One must prove many coherences to ensure things “work as expected”.
- CT suggests the right definitions, and they hold in our examples.
- Examples: neural networks, Hebbian learning, prediction markets.

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The neurons in a neural network form a dynamic operad.

- Interfaces: $\left(\begin{smallmatrix} T^*\mathbb{R}^n \\ \mathbb{R}^n \end{smallmatrix} \right)$. Meaning, output a vector, receive a gradient.
- Dynamic arrangements $\left(\begin{smallmatrix} T^*\mathbb{R}^{m_1} \\ \mathbb{R}^{m_1} \end{smallmatrix} \right) \otimes \cdots \otimes \left(\begin{smallmatrix} T^*\mathbb{R}^{m_k} \\ \mathbb{R}^{m_k} \end{smallmatrix} \right) \otimes \left(\begin{smallmatrix} T\mathbb{R}^d \\ \mathbb{R}^d \end{smallmatrix} \right) \xleftarrow{\quad} \left(\begin{smallmatrix} T^*\mathbb{R}^n \\ \mathbb{R}^n \end{smallmatrix} \right)$.
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Where does this come from? Often use $d := (m_1 + \cdots + m_k) \times n$.

- Given a map $\mathbb{R}^{m_1} \times \cdots \times \mathbb{R}^{m_k} \times \mathbb{R}^d \rightarrow \mathbb{R}^n$, e.g. “activated matmul”, ...
- ...one obtains a map backwards on cotangent vectors.
- Finish by using the canonical Riemannian structure $T^*\mathbb{R}^d \xrightarrow{\cong} T\mathbb{R}^d$.
- Generalize by replacing $\mathbb{R}^{m_i}$ and $\mathbb{R}^n$ with arb. manifolds and $\mathbb{R}^d$...
- ...with any $X : \mathbf{Mfd}$ having a “feedback response” map $T^*X \rightarrow TX$.
- Example: Riemannian and symplectic mfd's X have such structures.

Compositional physics

What makes artificial networks work?

- It's unknown why they're so good at their job.
- But what makes them feasibly fast is their compositional structure.
- Gradient back-propagation is compositional by the chain rule.
- But gradient *descent* is extra! It's comes from Riemann $T^*X \rightarrow TX$.
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Compositional physics is formally analogous.

- Hamiltonian dynamics replaces gradient descent dynamics.
- I'm not a physicist nor geometer; joint with Owen Lynch.
- Think of network architecture as a causal factoring of your system.
- In a dynamic arrangement $\left(\begin{smallmatrix} T^*M \\ M \end{smallmatrix} \right) \otimes \left(\begin{smallmatrix} TX \\ X \end{smallmatrix} \right) \xleftarrow{\quad} \left(\begin{smallmatrix} T^*N \\ N \end{smallmatrix} \right), \dots$
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Example: a system of colliding balls, each with springs or balls inside.

- The compositional physics allows computation by message passing.
- Forces on boundary are transferred inside, momentum is output.
- Physics sim'n & ANN training can be combined as one procedure. 11/13

Putting things together

Compositional adaptive systems in nature.

- Prediction markets, Hebbian learning, and ANNs have a similar flavor.
- We can see the system “adapting” to feedback.
- Wealths grow and shrink, cells wire together, weights change.
- The current arrangement dictates the input-output behavior, yet...
- ...what goes on throughout the system changes this arrangement.
- All of the above fit into a single formalism: dynamic operads.

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- All of the above fit into a single formalism: dynamic operads.

Classical (Hamiltonian) mechanics fits right in as well.

- We just had to replace the Riemannian with symplectic $T^*X \rightarrow TX$.
- Rather than “adapting” by descending, “adapt” by minimizing action.

Putting things together

Compositional adaptive systems in nature.

- Prediction markets, Hebbian learning, and ANNs have a similar flavor.
- We can see the system “adapting” to feedback.
- Wealths grow and shrink, cells wire together, weights change.
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One can thus combine neural nets and Hamiltonian physics engines.

- Backprop in ANNs is formally analogous to force transfer via b'ndary.
- One software architecture includes both simulation and learning.
- Structural thinking lets us find conceptual neighbors, generic theorems.

Outline

- 1 Introduction
- 2 Interfaces and arrangements
- 3 Dynamic operads
- 4 Conclusion**

Summary

We want a very general approach to systems that adapt to feedback.

- Category theory is all structure, agnostic as to content.
- It helps you articulate the types and how they relate.
- We want compositional open dynamical systems.

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We articulated the structure of interfaces and dynamic arrangements.

- Very general setup: lax monoidal $V : \mathcal{M}^{\text{op}} \rightarrow \mathbf{Cat}$.
- Interfaces are pairs (X, V) , denoted $\binom{V}{X}$, where $X : \mathcal{M}$ and $V : V(X)$.
- Example: $\binom{T^*X}{X}$. Output an element of X , receive a cotangent vector.
- Arrangements of little systems in a big system define inform'n flow.
- Arrangements are dynamic: they change based on what flows.

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Thanks; comments and questions welcome!