

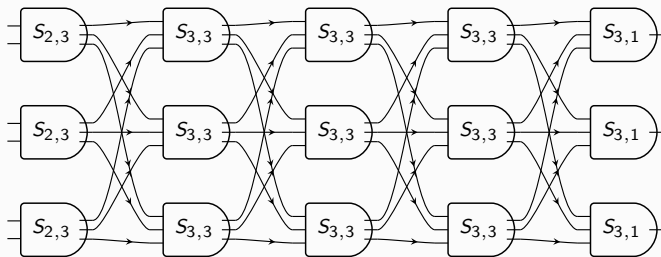
# Dynamic PROPs for Networked Learners

Brandon T. Shapiro\* and David I. Spivak

CALCO 2023



- 1 Nested dynamic structures
- 2 Polynomials and wiring diagrams
- 3 Dynamics: polynomial coalgebras
- 4 Nesting and networks: operad and PROP structure
- 5 A dynamic PROP for deep learning
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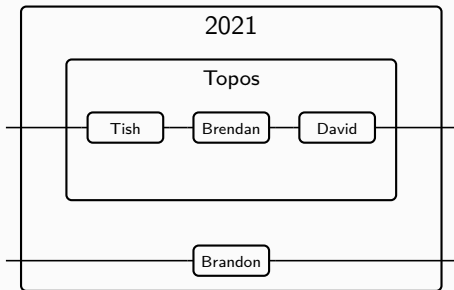


# Nested dynamic structures

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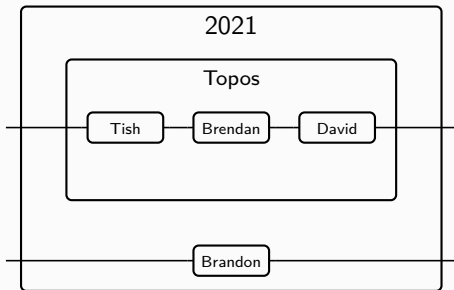
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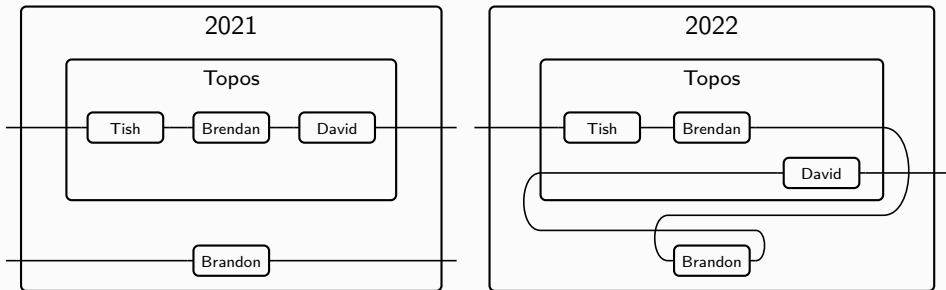
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(Not an accurate representation of Topos Institute's internal structure)

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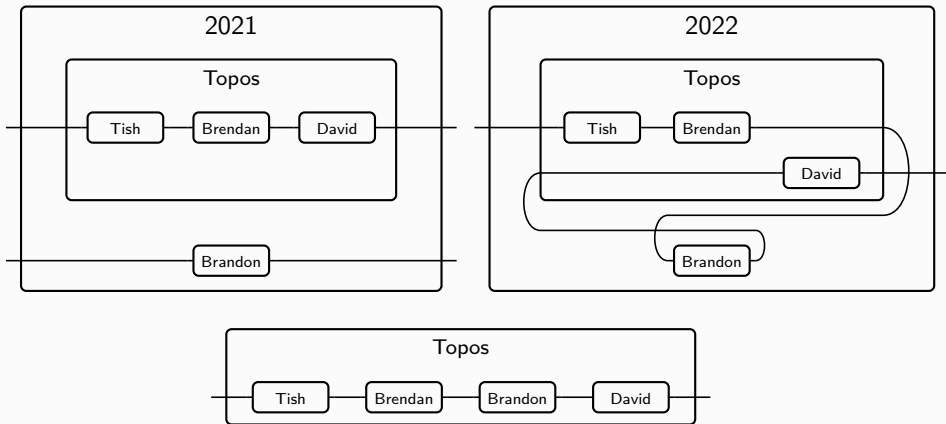
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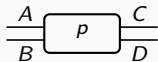
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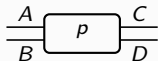
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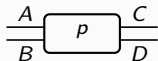
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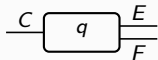
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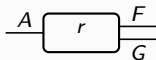
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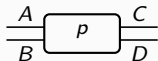
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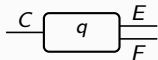
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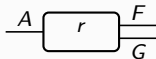
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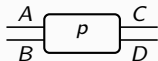


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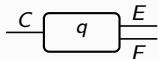
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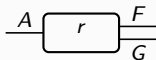
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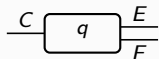
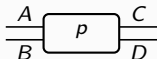
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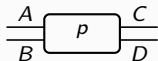
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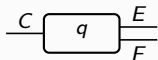
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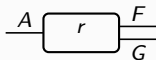
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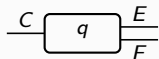
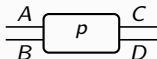
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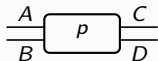


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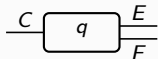
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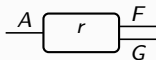
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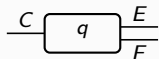
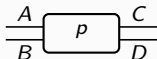
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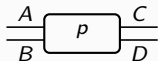
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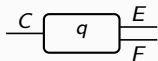
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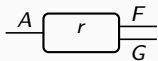
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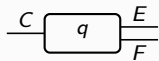
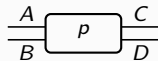
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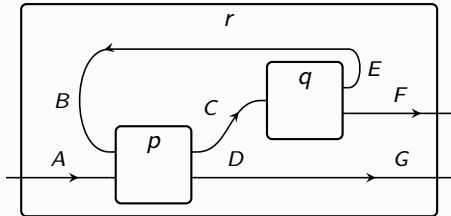
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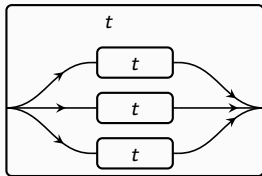
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- the feedback  $x \in \mathbb{R}^n$  causes  $(k, f, r)$  to update to

$$(k, f, r + \epsilon \pi_1 Df^T x)$$

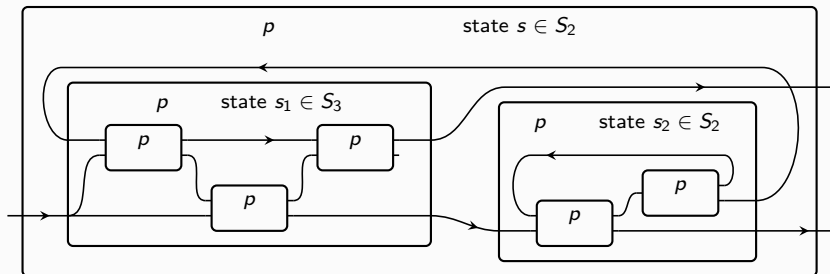
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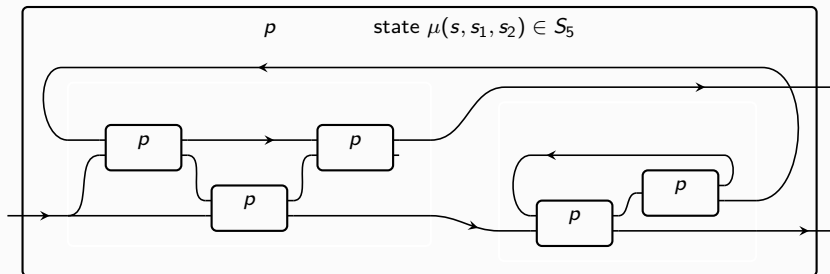
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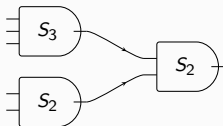
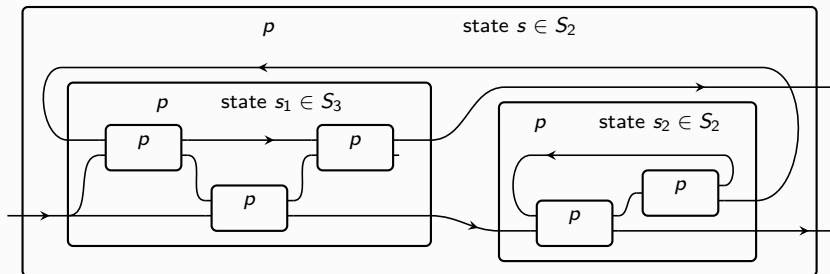
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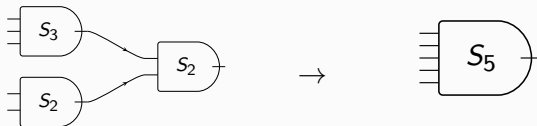
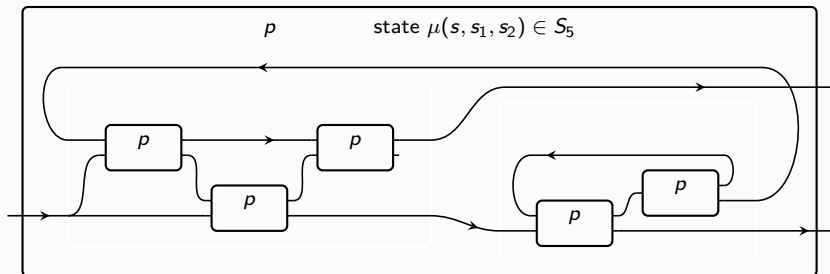
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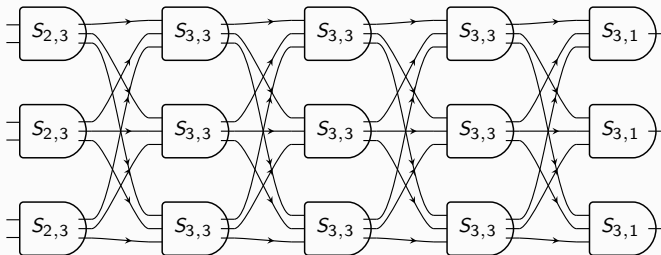
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$$S_{\ell,m} \times S_{m,n} \rightarrow S_{\ell,n}, \quad \text{and} \quad S_{m_1,n_1} \times S_{m_2,n_2} \rightarrow S_{m_1+m_2,n_1+n_2}$$

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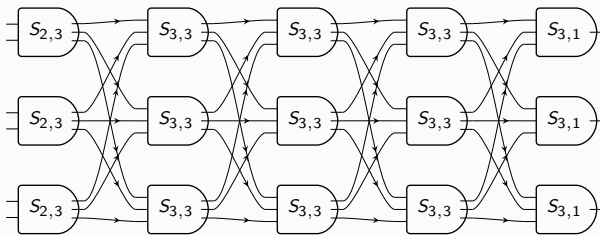
- $(\sigma, \tau, (k, f, r)) \mapsto$

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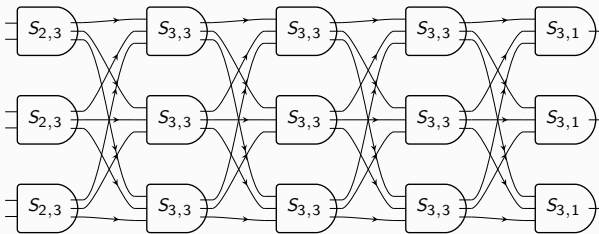
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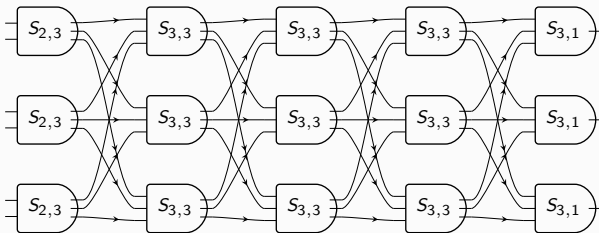
- $((k_1, f_1, r_1), (k_2, f_2, r_2)) \mapsto (k_1 + k_2, f_1 \times f_2, (r_1, r_2))$



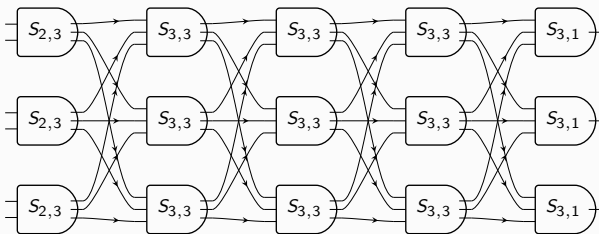
- Model other types of learning systems compositionally as dynamic categorical structures



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- As a well behaved categorical structure, dynamic PROPS can be easily compared, combined, or generalized
- Implementation in computing frameworks based on category theory or polynomials ([algebraicjulia.org](http://algebraicjulia.org))



- Brandon T. Shapiro and David I. Spivak, “Dynamic operads, dynamic categories: From deep learning to prediction markets” arXiv:2205.03906
- David I. Spivak, “Learners’ Languages” arXiv:2103.01189
- Sophie Libkind and David I. Spivak, “When you light up, I light up: A dynamical monoidal category of Hebbian learners” Topos Institute Blog

Thanks!