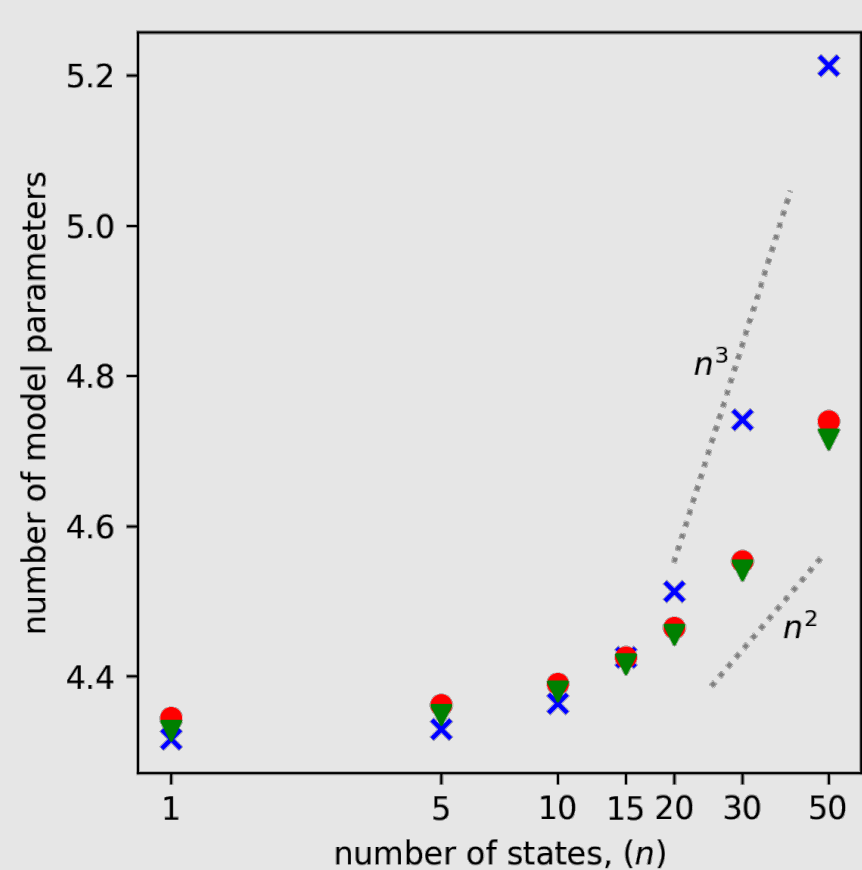
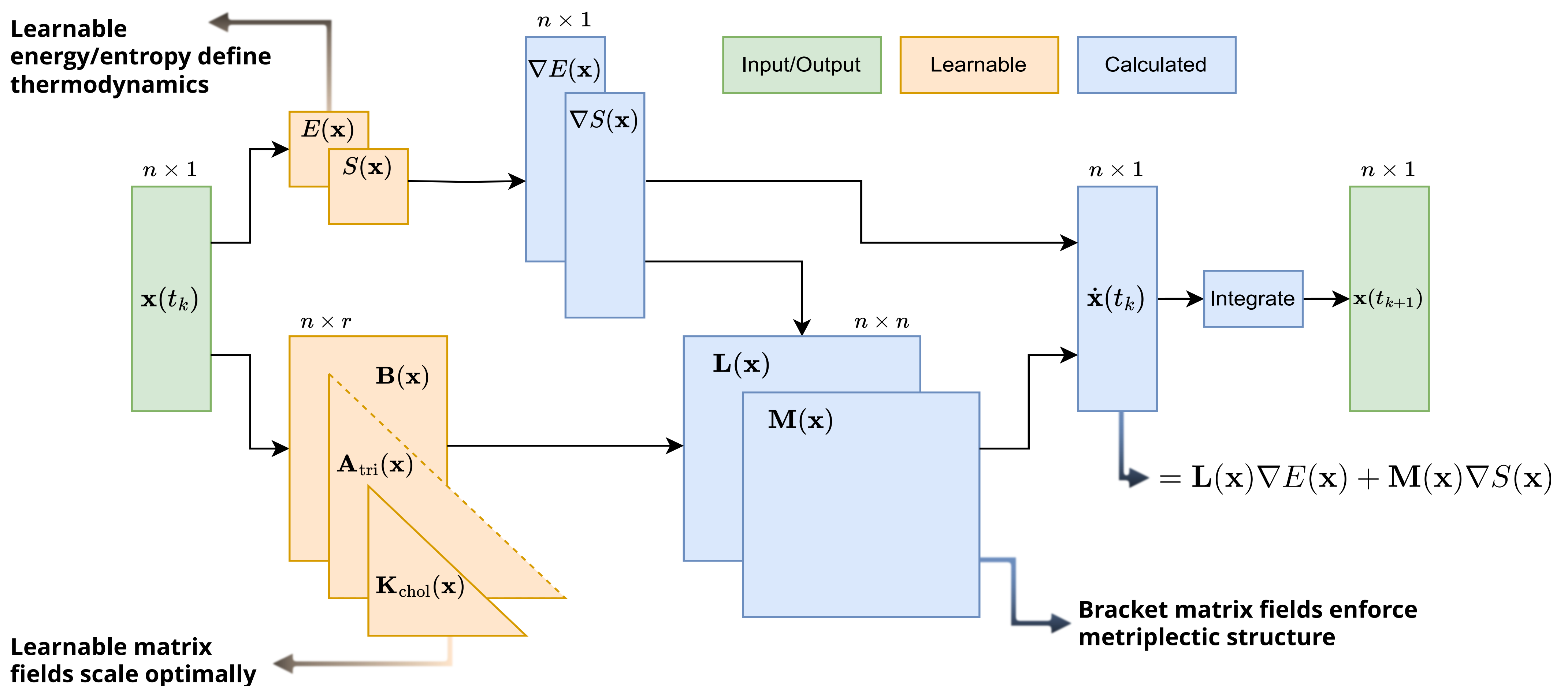


Efficiently Parameterized Neural Metriplectic Systems

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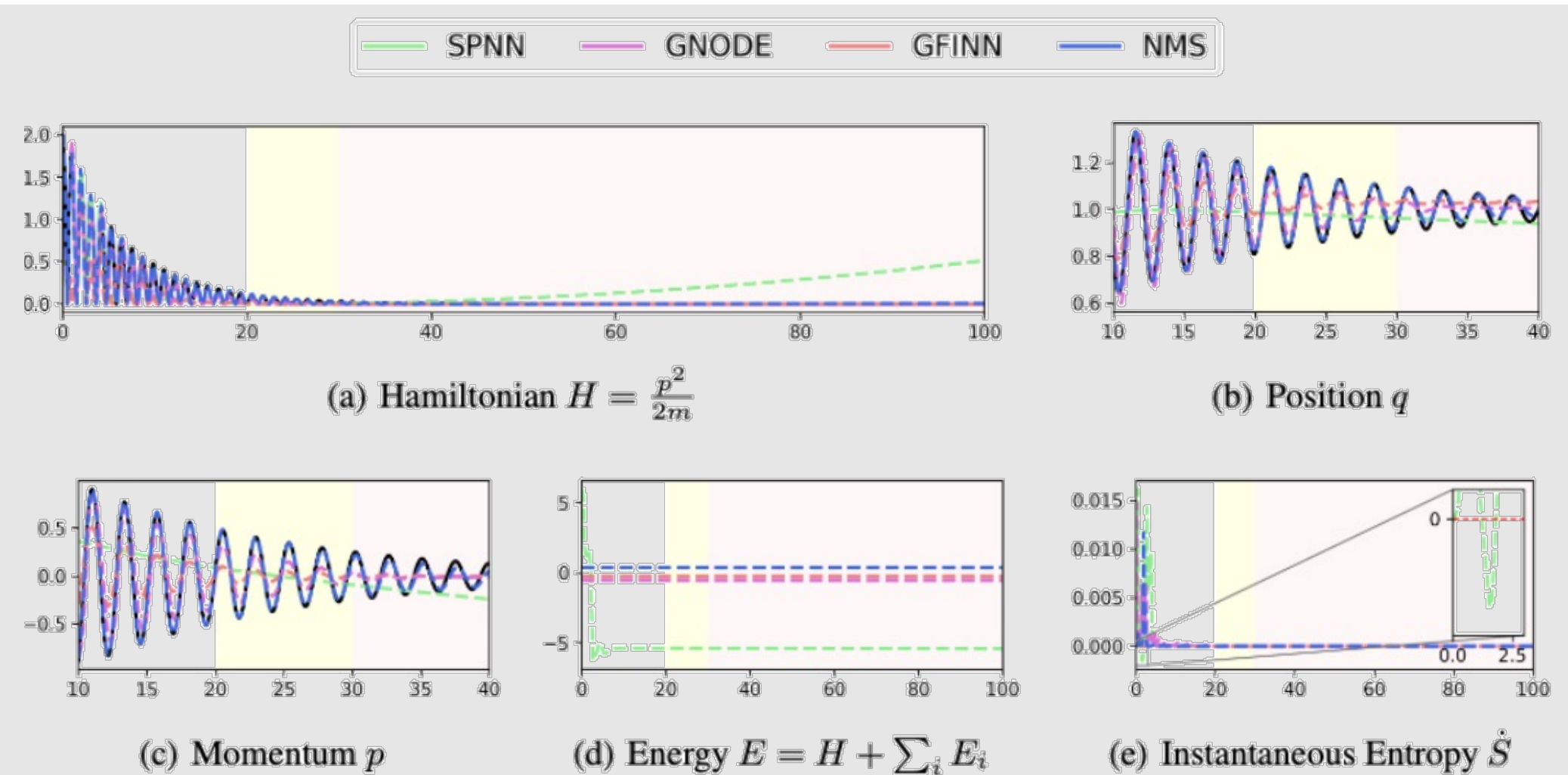
Given simulated or experimental data, infer a metriplectic model which **respects fundamental thermodynamic principles**.

- Guaranteed energy conservation and entropy generation laws.
- Universal approximation properties.
- Rigorous bounds on generalization error.
- Better predictive performance.



Our method (NMS) exhibits better predictive performance at reduced cost.

Optimal (quadratic) scaling in the number of states (n) is retained.



How this works: *Exterior algebra*

- Metriplectic structure parameterized via multivector fields with symmetries.
- Energy conservation and entropy generation guaranteed **independent of training quality**.

Provable guarantees on network performance

- Universal approximation of any nondegenerate metriplectic system.
- Computable *a posteriori* error estimate bounding generalization error in time.

$$\left(\int_0^T |\mathbf{x} - \tilde{\mathbf{x}}|^2 dt \right)^{\frac{1}{2}} \leq \varepsilon \left| \frac{b}{a} \right| \sqrt{e^{2aT} - 2e^{aT} + T + 1},$$

Rigorous statements on energy/entropy stability and learnable conservation laws.

Well-behaved and **stable** feature propagation.

Better data efficiency, interpretability, and predictive performance.



Check out our paper here!

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👤 <https://www.agrubertx.github.io>