



# Property-preserving model reduction for conservative and dissipative systems

Anthony Gruber

Center for Computing Research, Sandia National Laboratories, Albuquerque, NM



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# Full-Order Model

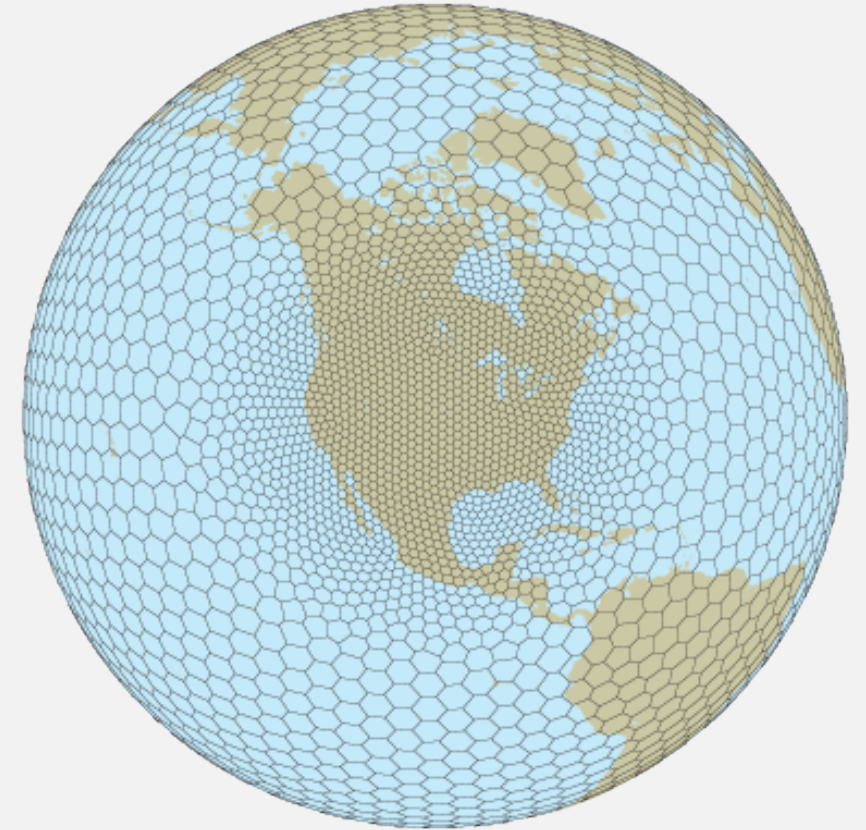
- Consider state variable  $\mathbf{x}(t, \boldsymbol{\mu}) \in \mathbb{R}^N$ .
  - $N$  ranges from “large” to “very large” ( $10^6+$  not uncommon).
  - $\boldsymbol{\mu}$  is vector of parameters.
- Dynamics take the form

$$\dot{\mathbf{x}}(t, \boldsymbol{\mu}) = \mathbf{A}\mathbf{x}(t, \boldsymbol{\mu}) + \mathbf{f}(\mathbf{x}(t, \boldsymbol{\mu})) .$$

- Can be expensive to solve. How to reduce cost?

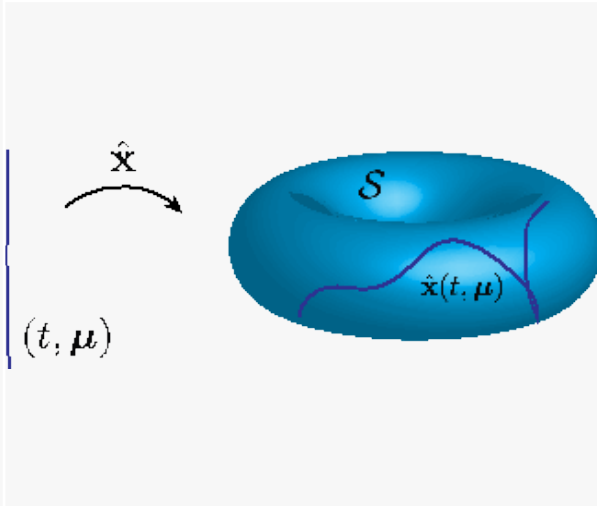
# Reduced-Order Models

- High-fidelity PDE simulations are **expensive**.
  - semi-discretization blows up dimensionality.
- Good results possible without solving full PDE?
- Standard is to **encode** -> **solve** -> **decode**.
  - Linear: **POD**, RBM, etc.
  - Nonlinear: kernel methods, neural networks, etc.



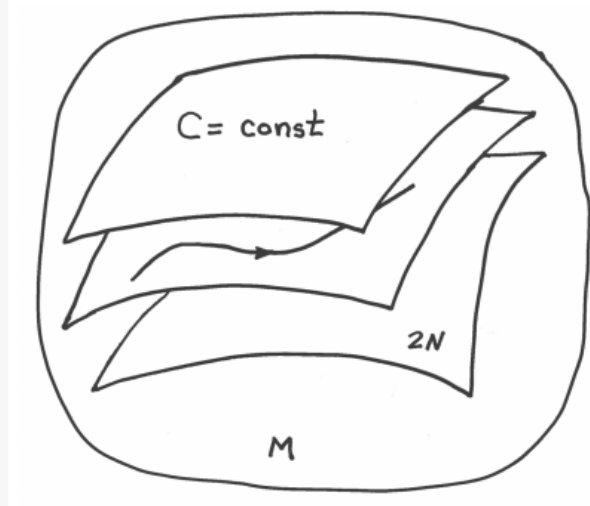
<https://mpas-dev.github.io/atmosphere/atmosphere.html>

# Outline



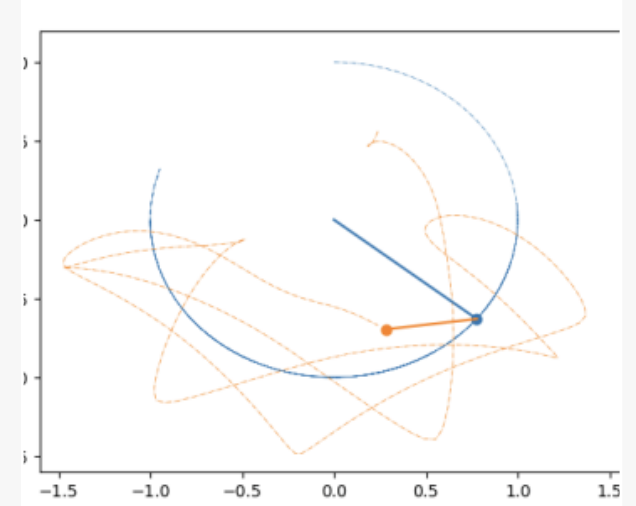
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Introduction to Model Reduction



2

Hamiltonian Model Reduction



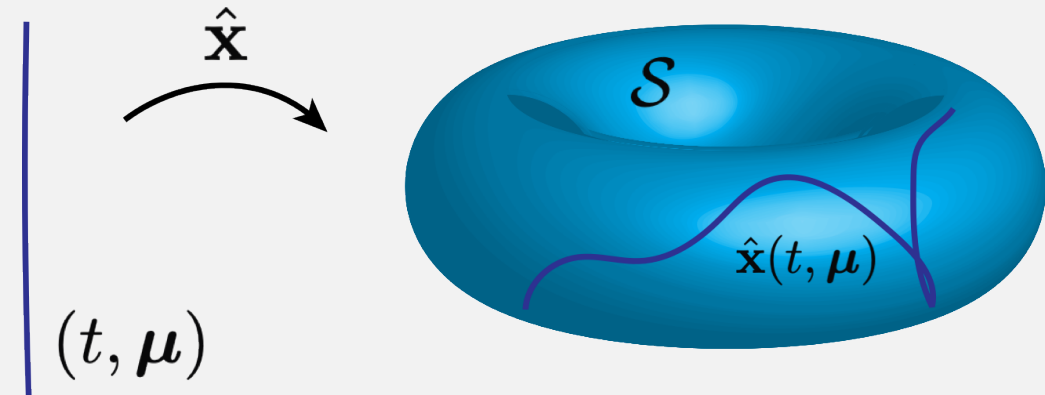
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Metriplectic Model Reduction

- Joint work with Max Gunzburger (UT Austin), Lili Ju (U of SC), Zhu Wang (U of SC), and Irina Tezaur (Sandia CA).

# Idea Behind ROM

- Do we really need all  $10^6$  dimensions?
  - No, if  $(t, \boldsymbol{\mu}) \mapsto \mathbf{x}(t, \boldsymbol{\mu})$  is unique.
- $\mathcal{S} = \{\mathbf{x}(t, \boldsymbol{\mu}) \mid t \in [0, T], \boldsymbol{\mu} \in D\} \subset \mathbb{R}^N$ ,  
solution manifold.
  - $(n_\mu + 1)$  dimensions enough for  
loss-less representation of  $\mathcal{S}$ .
- How can we approximate  $\mathcal{S}$  efficiently?

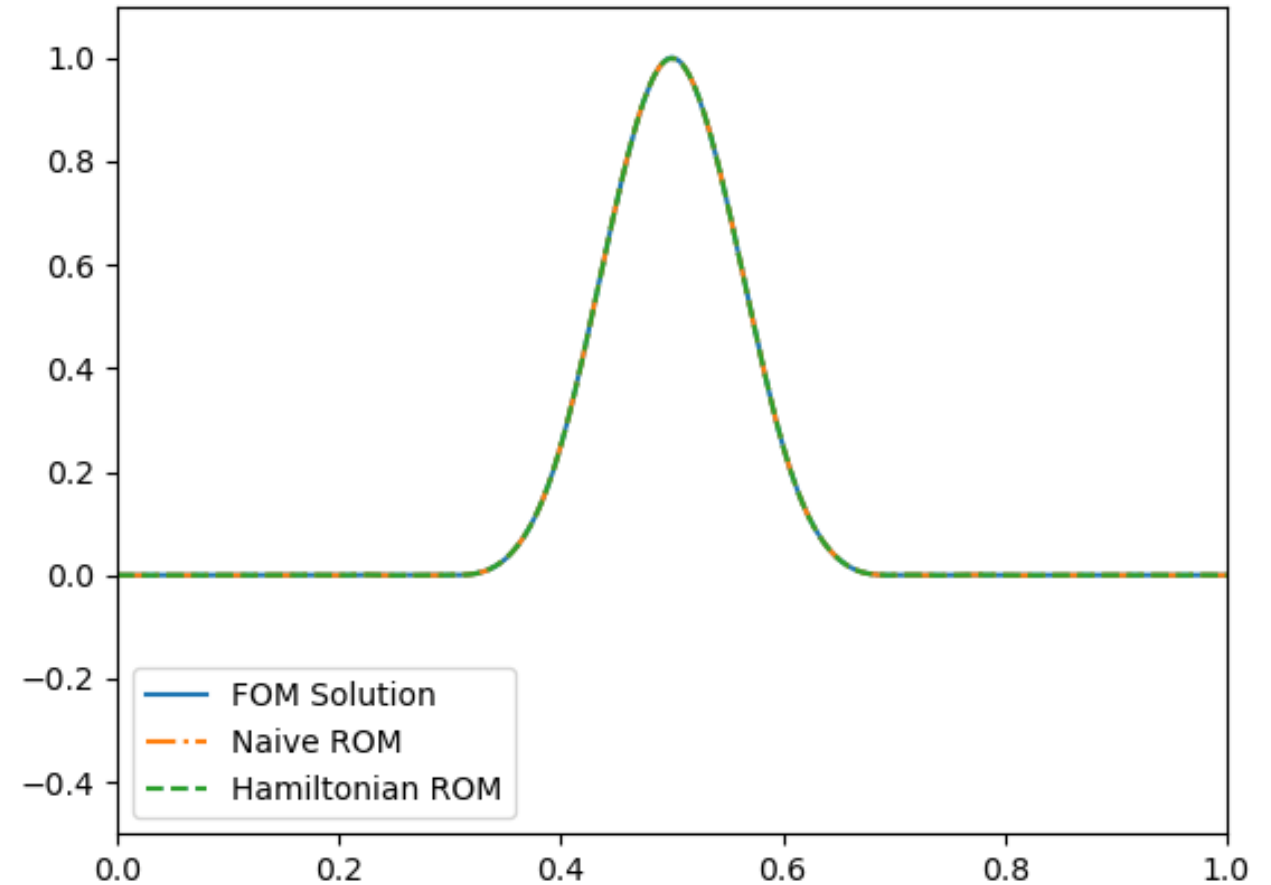


# Proper Orthogonal Decomposition

- Do PCA on solution snapshots  $\mathbf{X} = \mathbf{x}(t_j, \boldsymbol{\mu}_j)$ ,  $1 \leq j \leq N_t$ .
  - Yields  $\mathbf{X} \approx \mathbf{U}\boldsymbol{\Sigma}\mathbf{V}^\top$ , Galerkin projection  $\tilde{\mathbf{x}} = \mathbf{U}\hat{\mathbf{x}}$ .
- Orthogonality of POD basis  $\mathbf{U}$  implies, for  $\hat{\mathbf{x}} \in \mathbb{R}^n$ ,  $\hat{f} = f \circ \tilde{\mathbf{x}}$ ,
$$\dot{\hat{\mathbf{x}}} = \mathbf{U}^\top \mathbf{A} \mathbf{U} \hat{\mathbf{x}} + \mathbf{U}^\top \mathbf{f}(\mathbf{U} \hat{\mathbf{x}}) := \hat{\mathbf{A}} \hat{\mathbf{x}} + \hat{\mathbf{f}}(\hat{\mathbf{x}})$$
- ODE of size  $N$  converted to ODE of size  $n$ .

# Does this work?

- Trained on one period
  - Tested on five



# Hamiltonian Systems

- Archetype for conservative systems:  $\dot{\mathbf{x}} = \{\mathbf{x}, H\} = \mathbf{L}\nabla H$ .
  - Governed by scalar potential function  $H$  and SS matrix  $\mathbf{L}$ .
- $\mathbf{L}$  defines (potentially degenerate) Poisson bracket  $\{F, G\} = \nabla F \cdot \mathbf{L}\nabla G$ .
  - Satisfies Jacobi identity  $\{F, \{G, H\}\} + \{G, \{H, F\}\} + \{H, \{F, G\}\} = 0$ .
- Guarantees that flow is  $\perp \nabla H$  and energy is conserved:

$$\dot{H}(\mathbf{x}) = \dot{\mathbf{x}} \cdot \nabla H = \mathbf{L}\nabla H \cdot \nabla H = -\mathbf{L}\nabla H \cdot \nabla H = 0.$$



# Examples) Hamiltonian Systems

- Undamped simple harmonic oscillator:  $m\ddot{x} = -kx$

$$H = \frac{1}{2m} (p^2 + q^2) \quad \mathbf{L} = \mathbf{J} = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

$$p = m\dot{x}, \quad q = m\sqrt{\frac{k}{m}}x$$

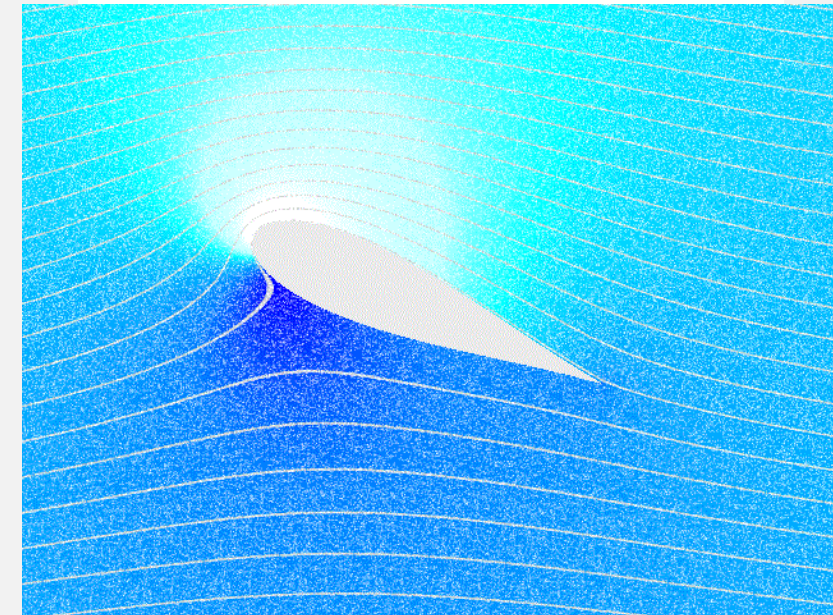
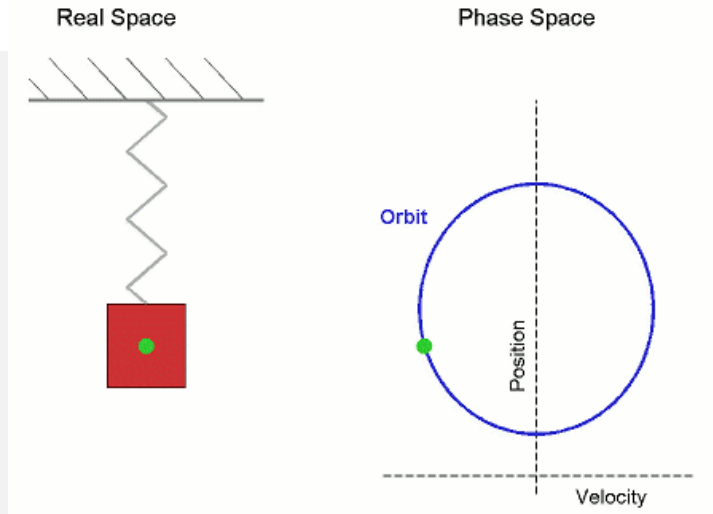
- Incompressible Euler:  $\dot{\omega} = \omega \cdot \nabla \mathbf{u} - \mathbf{u} \cdot \nabla \omega$

$$H = \frac{1}{2} \int |\mathbf{u}|^2 dx. \quad L(\omega) = (\omega \cdot \nabla - \nabla \omega) \nabla \times$$

- Warning! Vorticity is the Hamiltonian variable!

$$\omega = \nabla \times \mathbf{u}$$

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# What about ROM?

- Naïve Galerkin projection yields  $\dot{\hat{\mathbf{x}}} = \mathbf{U}^\top \mathbf{L} \nabla H(\tilde{\mathbf{x}})$ .
  - Not Hamiltonian,  $\mathbf{U}^\top \mathbf{L} \neq -\mathbf{L}^\top \mathbf{U}$ .
- One solution due to Y. Gong, Q. Wang, Z. Wang (2017):
  - Recall that  $\nabla \hat{H}(\hat{\mathbf{x}}) = \tilde{\mathbf{x}}' \cdot \nabla H(\tilde{\mathbf{x}}) = \mathbf{U}^\top \nabla H(\tilde{\mathbf{x}})$ .
- We want  $\mathbf{U}^\top \mathbf{L} \nabla H(\tilde{\mathbf{x}}) = \hat{\mathbf{L}} \mathbf{U}^\top \nabla H(\tilde{\mathbf{x}}) = \hat{\mathbf{L}} \nabla \hat{H}(\hat{\mathbf{x}})$ .
  - Implies the overdetermined system  $\mathbf{U}^\top \mathbf{L} = \hat{\mathbf{L}} \mathbf{U}^\top$ .
  - Solution is  $\hat{\mathbf{L}} = \mathbf{U}^\top \mathbf{L} \mathbf{U}$  (antisymmetry is obviously inherited).

# Hamiltonian ROM

- Energy conservation is retained:

$$\dot{\hat{H}}(\hat{\mathbf{x}}) = \dot{\hat{\mathbf{x}}} \cdot \nabla \hat{H} = \hat{\mathbf{L}} \nabla \hat{H} \cdot \nabla \hat{H} = -\hat{\mathbf{L}} \nabla \hat{H} \cdot \nabla \hat{H} = 0$$

- Can prove convergence to FOM solution with increasing POD basis.
- Need a symplectic time integrator (many choices available).
  - For "easy" nonlinearities, AVF is a good choice.

$$\frac{\mathbf{x}^{k+1} - \mathbf{x}^k}{\Delta t} = \int_0^1 \mathbf{L} \nabla H \left( t \mathbf{x}^{k+1} + (1 - t) \mathbf{x}^k \right) dt$$

# Nonintrusive Hamiltonian ROMs

- What happens if no access to FOM code? *Operator inference*.

- Partial solution for canonical systems (Sharma, Kramer, Wang 2022):

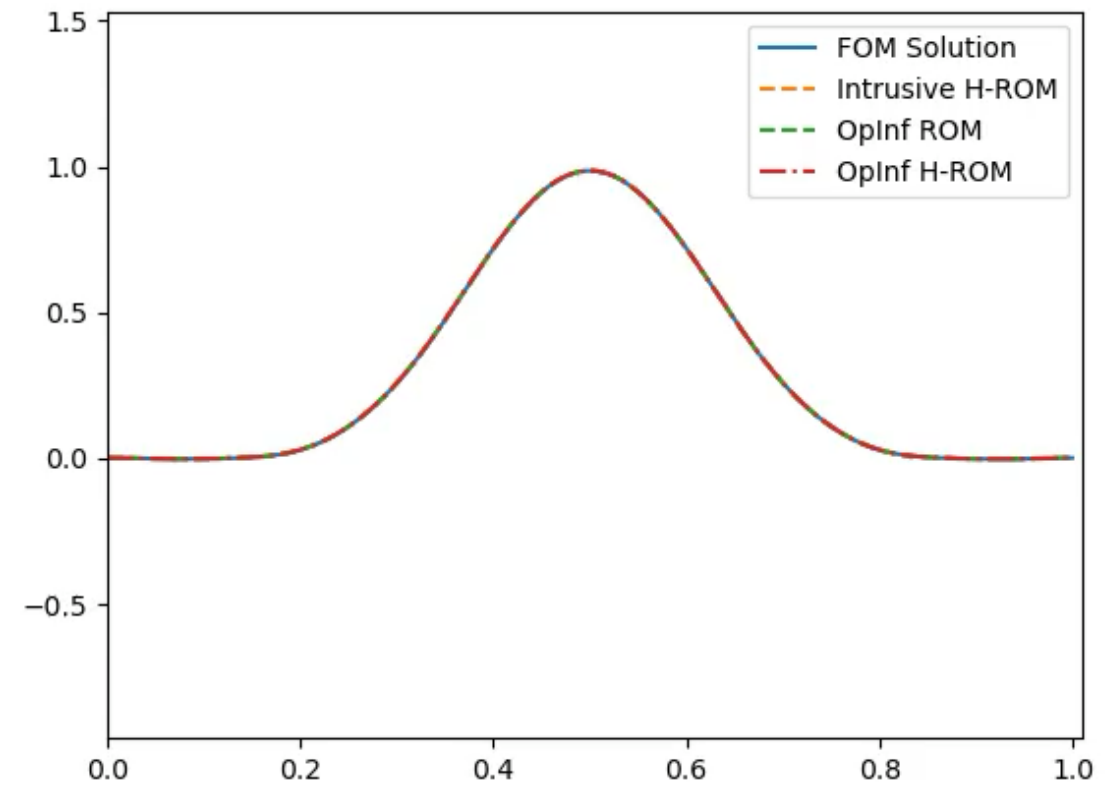
- Postulate a reduced Hamiltonian  $\hat{H}(\hat{\mathbf{q}}, \hat{\mathbf{p}}) = \hat{\mathbf{q}}^\top \hat{\mathbf{A}}_{qq} \hat{\mathbf{q}} + \hat{\mathbf{p}}^\top \hat{\mathbf{A}}_{pp} \hat{\mathbf{p}}$ .

- Dynamical system becomes  $\begin{pmatrix} \dot{\hat{\mathbf{q}}} \\ \dot{\hat{\mathbf{p}}} \end{pmatrix} = \begin{pmatrix} 0 & \mathbf{I} \\ -\mathbf{I} & 0 \end{pmatrix} \begin{pmatrix} \hat{\mathbf{A}}_{qq} & \mathbf{0} \\ \mathbf{0} & \hat{\mathbf{A}}_{pp} \end{pmatrix} \begin{pmatrix} \hat{\mathbf{q}} \\ \hat{\mathbf{p}} \end{pmatrix}$ .

- Infer  $\hat{\mathbf{A}}_{qq} = \operatorname{argmin}_{\hat{\mathbf{A}}=\hat{\mathbf{A}}^\top} \left| \hat{\mathbf{X}}_{p,t} + \hat{\mathbf{A}} \hat{\mathbf{X}}_q \right|^2$      $\hat{\mathbf{A}}_{pp} = \operatorname{argmin}_{\hat{\mathbf{A}}=\hat{\mathbf{A}}^\top} \left| \hat{\mathbf{X}}_{q,t} - \hat{\mathbf{A}} \hat{\mathbf{X}}_p \right|^2$

# Does it work?

- Yeah! But...
  - Relies on  $\mathbf{U}^\top \mathbf{J} = \mathbf{J}_r \mathbf{U}^\top$ .
  - Needs a block-diagonal  $\nabla \hat{H}$
- How to extend to more general systems?



# Hamiltonian Operator Inference

- Recognize special case of more general OpInf procedure:
  - Joint w. Irina Tezaur (Sandia CA)
  - Can solve  $\operatorname{argmin}_{\hat{\mathbf{L}} \text{ or } \hat{\mathbf{A}}} \left| \hat{\mathbf{X}}_t - \hat{\mathbf{L}} \hat{\mathbf{A}} \hat{\mathbf{X}} \right|^2$ ,  $\hat{\mathbf{L}}^\top = -\hat{\mathbf{L}}, \hat{\mathbf{A}}^\top = \hat{\mathbf{A}}$ .
- If  $\mathbf{L}$  is known, this is “canonical” inference!
- If  $\nabla H$  is known, this is *noncanonical inference*.

# Hamiltonian Operator Inference

- Need to solve minimization

$$\operatorname{argmin}_{\mathbf{D} \in \mathbb{R}^{N \times N}} \left( \|\mathbf{C} - \mathbf{A} (\mathbf{D} \pm \mathbf{D}^\top) \mathbf{B}\|^2 + \eta \|\mathbf{D}\|^2 \right),$$

- Boils down to *unconstrained* linear system:

$$((\mathbf{A}^\top \mathbf{A} \bar{\oplus} \mathbf{B} \mathbf{B}^\top) (\mathbf{I} \pm \mathbf{K}) + \eta \mathbf{I}) \operatorname{vec} \mathbf{D} = \operatorname{vec} (\mathbf{A}^\top \mathbf{C} \mathbf{B}^\top \pm \mathbf{B} \mathbf{C}^\top \mathbf{A}),$$

where  $\mathbf{A} \bar{\oplus} \mathbf{B} = \mathbf{A} \otimes \mathbf{B} + \mathbf{B} \otimes \mathbf{A}$ ,  $\operatorname{vec} \mathbf{D}^\top = \mathbf{K} \operatorname{vec} \mathbf{D}$ .

# KdV Equation

- Consider solving  $u_t = \alpha u u_x + \rho u_x + \gamma u_{xxx}, \quad [-L, L] \times [0, T]$

- Recast as  $u_t = \mathcal{D} \frac{\delta \mathcal{H}}{\delta u}, \quad \mathcal{H} = \int_0^L \left( \frac{\alpha}{6} u^3 + \frac{\rho}{2} u^2 - \frac{\nu}{2} u_x^2 \right) dx, \quad \mathcal{D} = \partial_x$

- Discretizing with periodic BCs yields

$$\mathbf{A} = \frac{1}{2\Delta x} \begin{pmatrix} 0 & 1 & 0 & 0 & \cdots & -1 \\ -1 & 0 & 1 & 0 & \cdots & 0 \\ & & \ddots & \ddots & \ddots & \\ 0 & \cdots & 0 & -1 & 0 & 1 \\ 1 & \cdots & 0 & 0 & -1 & 0 \end{pmatrix}, \quad \mathbf{B} = \frac{1}{\Delta x^2} \begin{pmatrix} -2 & 1 & 0 & 0 & \cdots & 1 \\ 1 & -2 & 1 & 0 & \cdots & 0 \\ & & \ddots & \ddots & \ddots & \\ 0 & \cdots & 0 & 1 & -2 & 1 \\ 1 & \cdots & 0 & 0 & 1 & -2 \end{pmatrix}$$

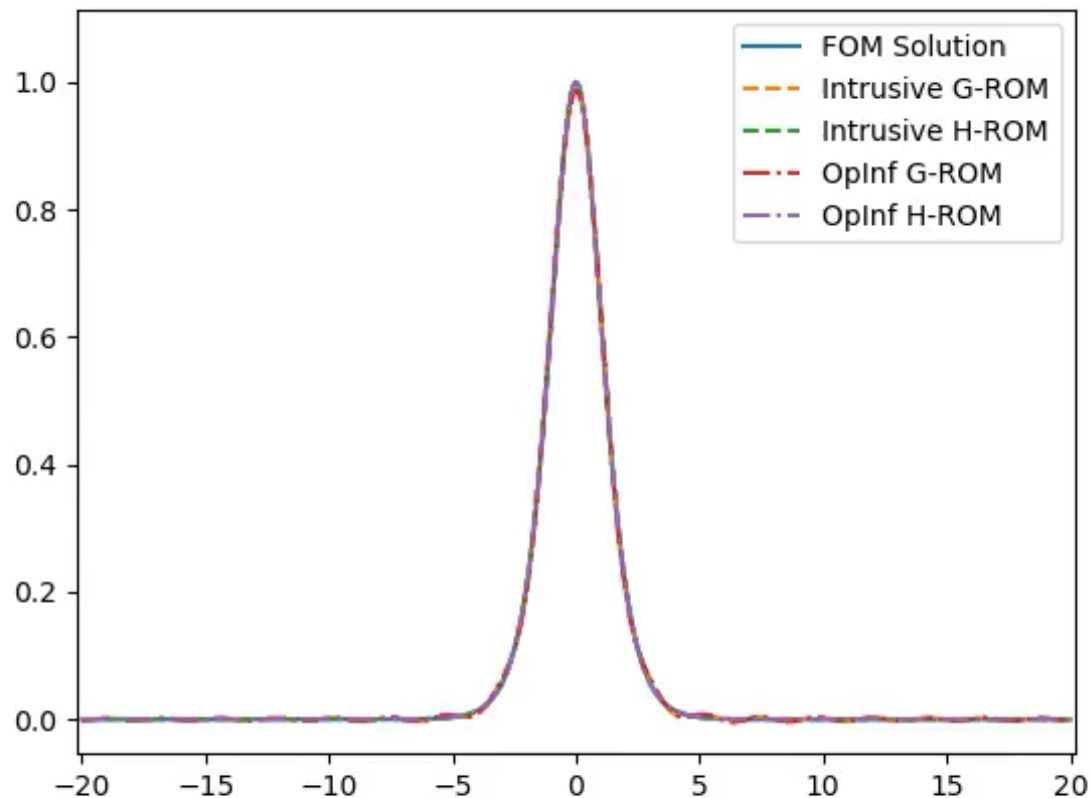
$$\begin{aligned} \frac{d\mathbf{u}}{dt} &= \mathbf{A} \nabla_{\mathbf{u}} H(\mathbf{u}) \\ &= \mathbf{A} \left( \frac{\alpha}{2} \mathbf{u}^2 + \rho \mathbf{u} + \nu \mathbf{B} \mathbf{u} \right) \end{aligned}$$

$$u_0(x) = \text{sech}^2 \left( \frac{x}{\sqrt{2}} \right)$$

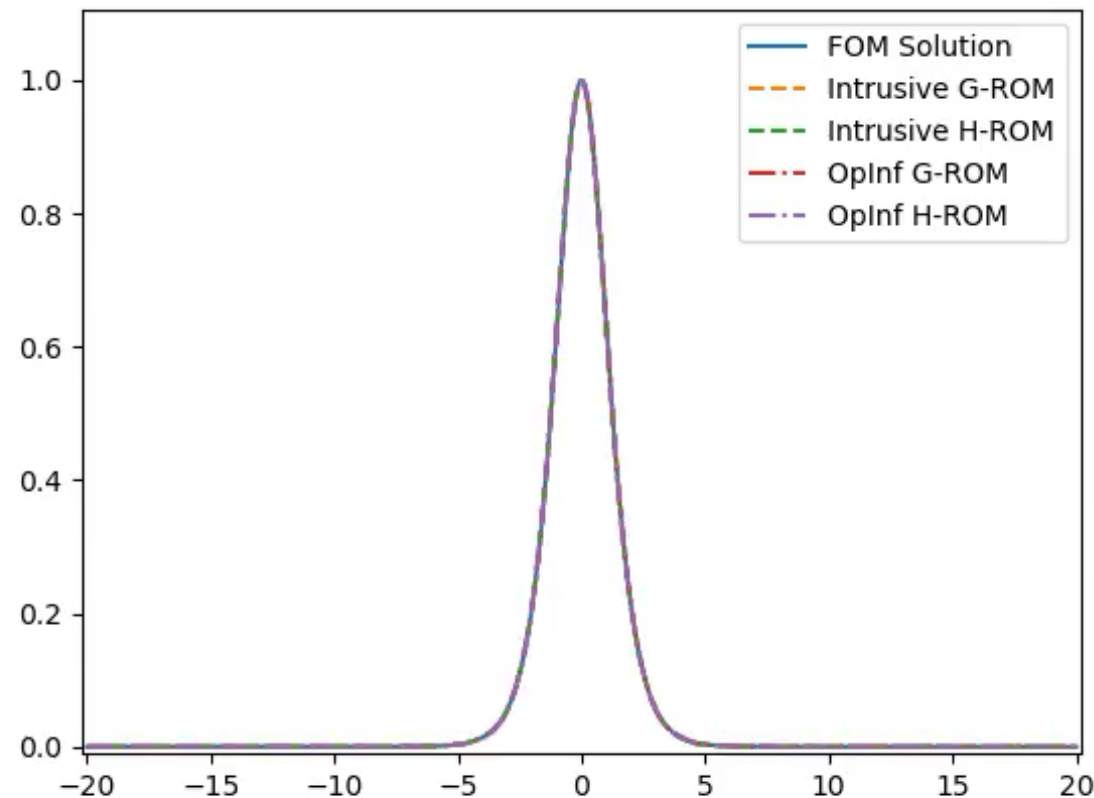
- $\mathbf{A} = \mathbf{L}$  is non-canonical!



# KdV Equation

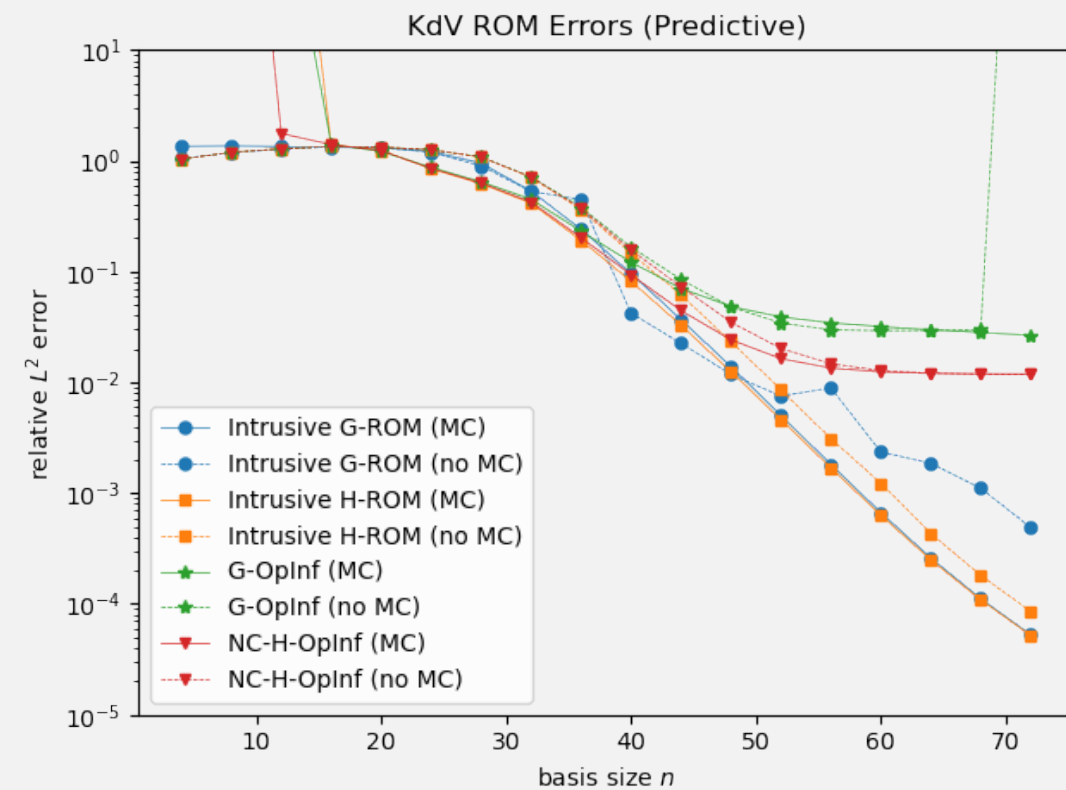
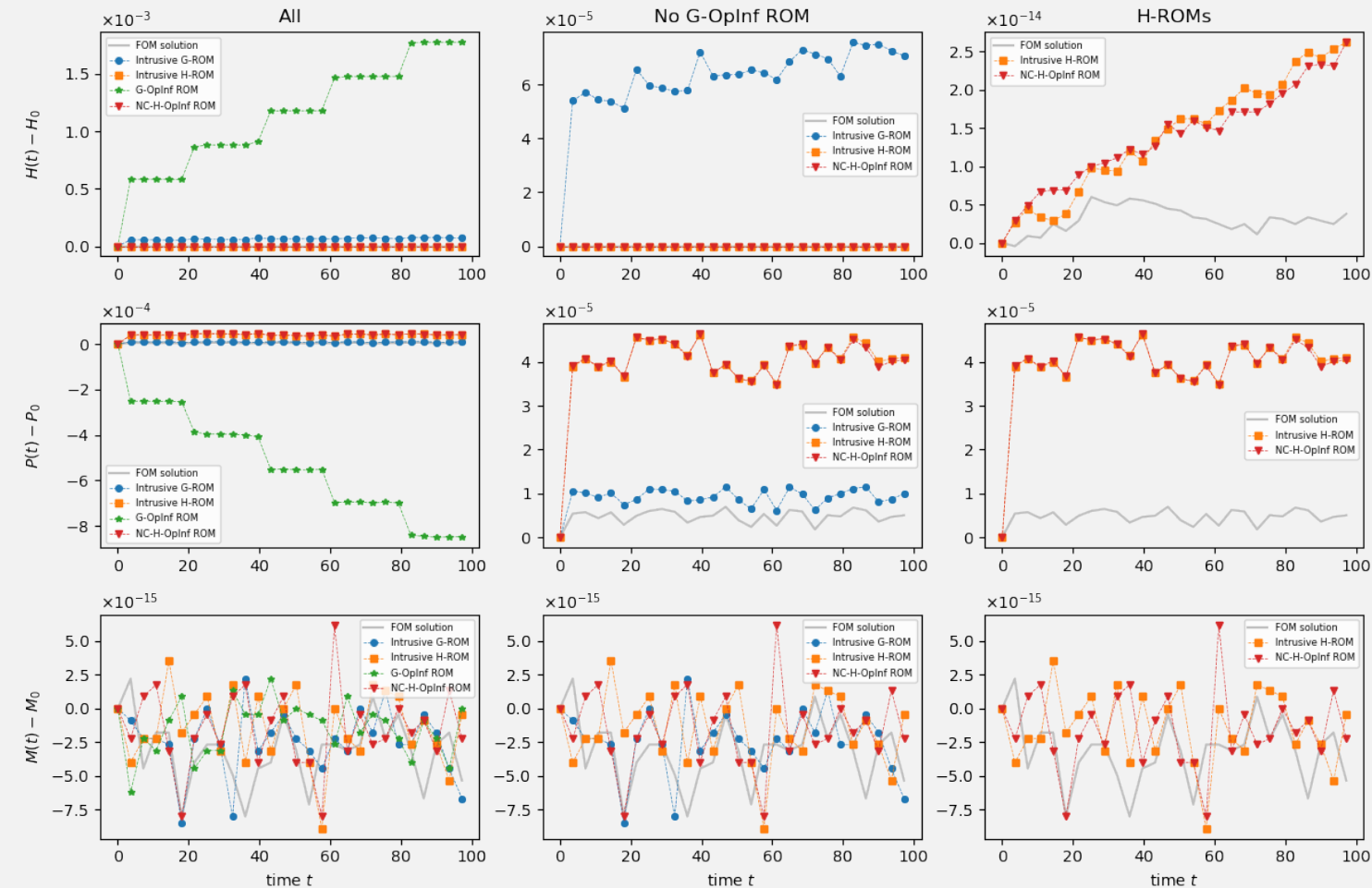


$N = 500, n=36$



$N = 500, n=48$

# KdV Equation



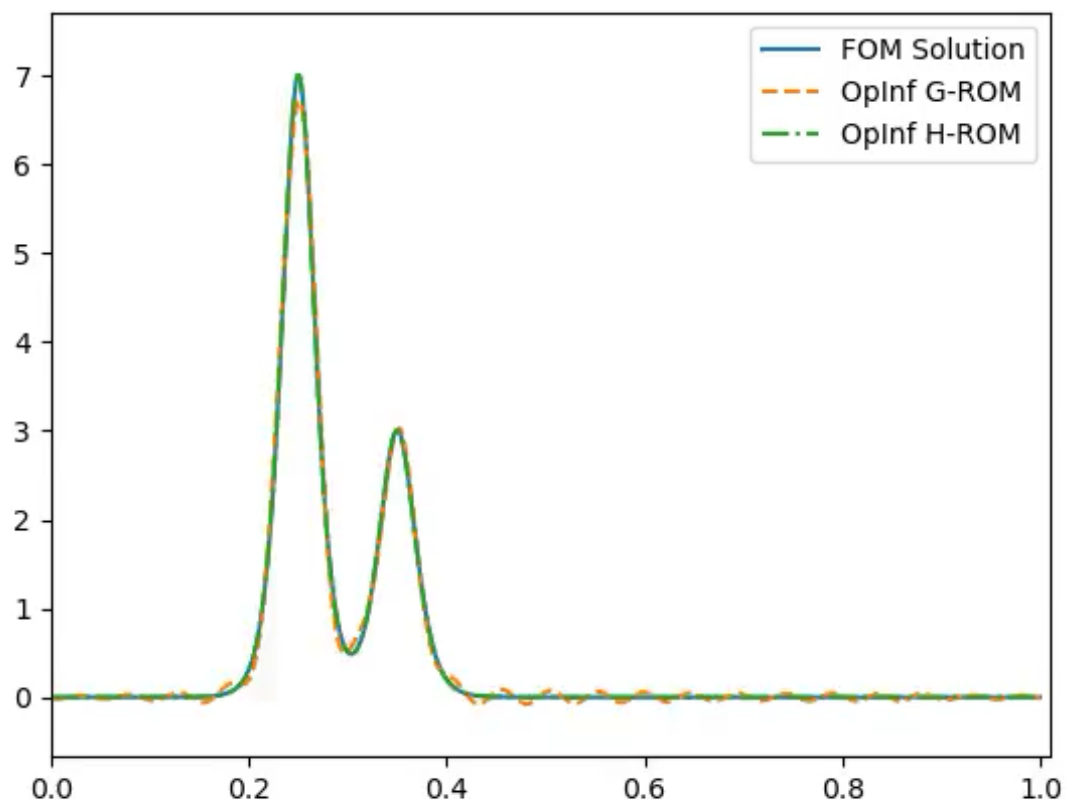
# Convergence with increasing data

- If  $\lim_{n \rightarrow N} |(\mathbf{I} - \mathbf{U}\mathbf{U}^\top) \mathbf{x}| = 0, \forall \mathbf{x},$   
 $\lim_{\Delta t \rightarrow 0} \max_i |\mathbf{x}_t(t_i) - \dot{\mathbf{x}}(t_i)| = 0,$   
 $\mathbf{X}, H(\mathbf{X})$  have maximal rank.
- Theorem:  $\hat{\mathbf{L}} \rightarrow \mathbf{U}^\top \mathbf{L} \mathbf{U}, \hat{\mathbf{A}} \rightarrow \mathbf{U}^\top \mathbf{A} \mathbf{U}$   
as  $n \rightarrow N$  and  $\Delta t \rightarrow 0.$

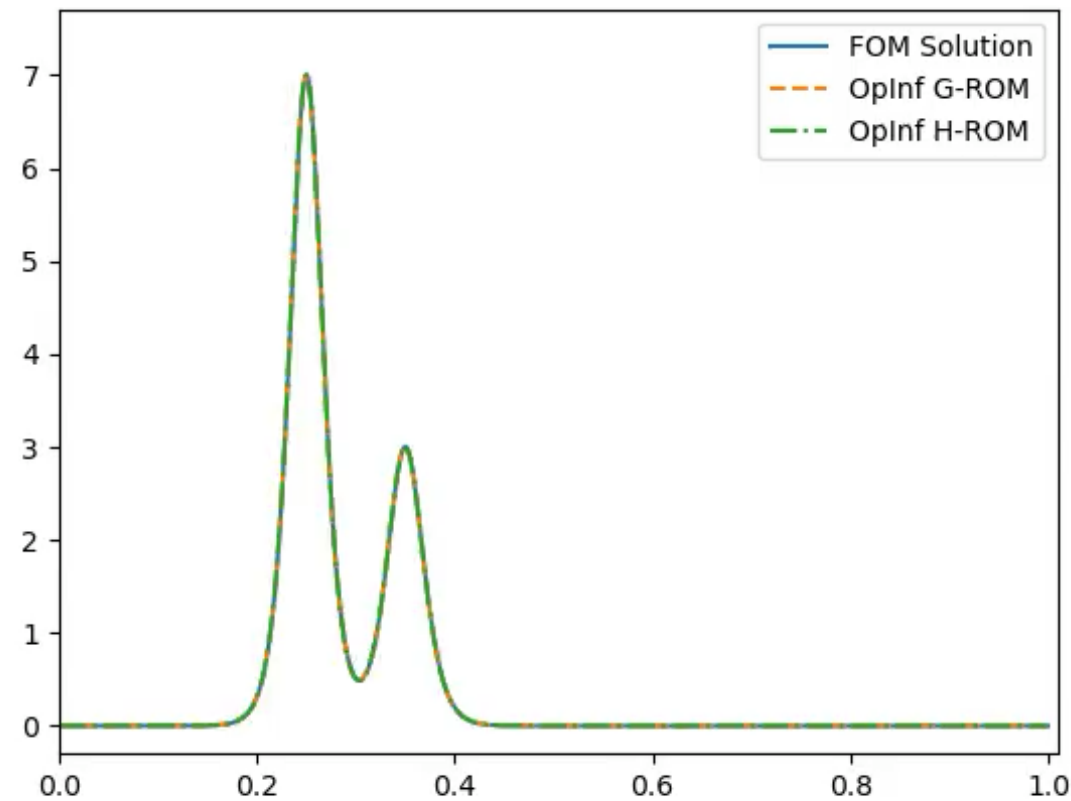
# BBM Equation

- Benjamin-Bona-Mahoney equation:  $\dot{x} = \alpha x_s + \beta x x_s - \gamma \dot{x}_{ss}.$
- Hamiltonian:  $H(x) = \frac{1}{2} \int_0^\ell \alpha x^2 + \frac{\beta}{3} x^3 ds,$
- Poisson structure:  $L = - (1 - \partial_s^2)^{-1} \partial_s,$
- Intrusive H-ROM not feasible
  - Can we still get a good OpInf H-ROM?

# BBM Equation

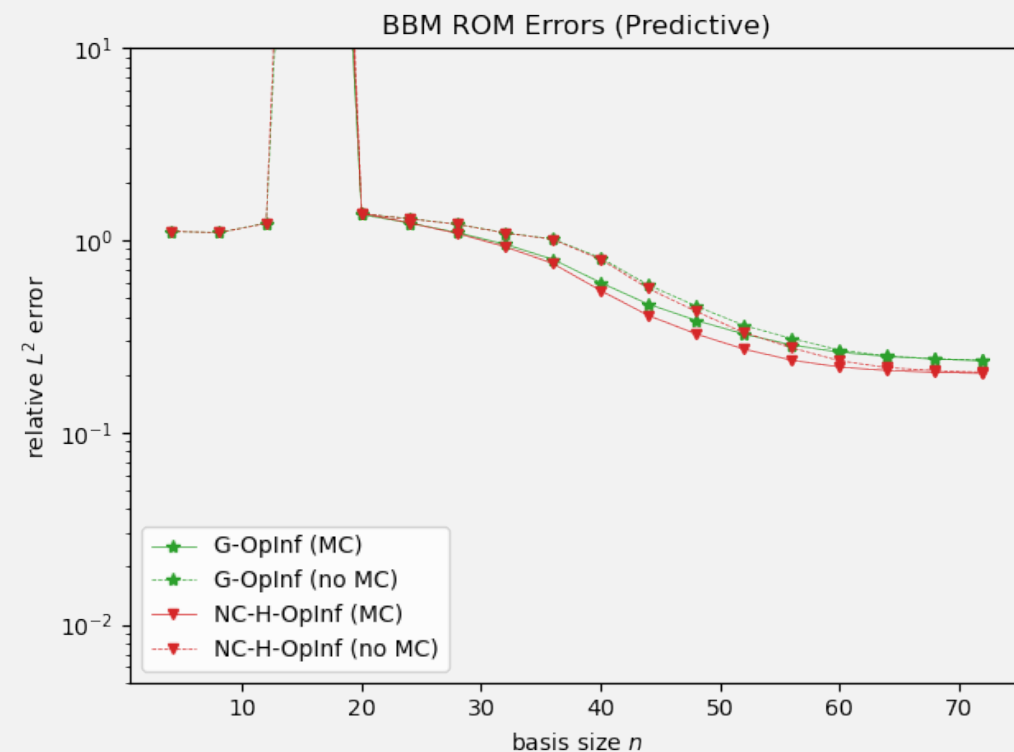
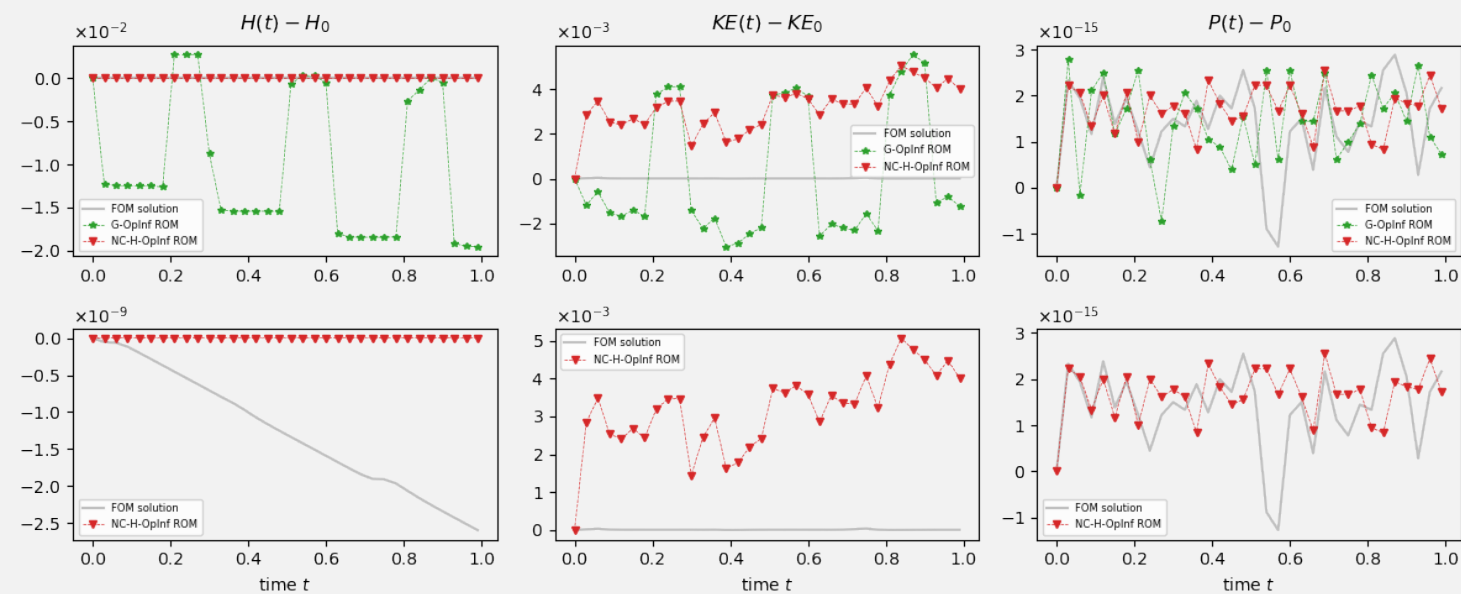


$N = 1024, n=36$



$N = 1024, n=72$

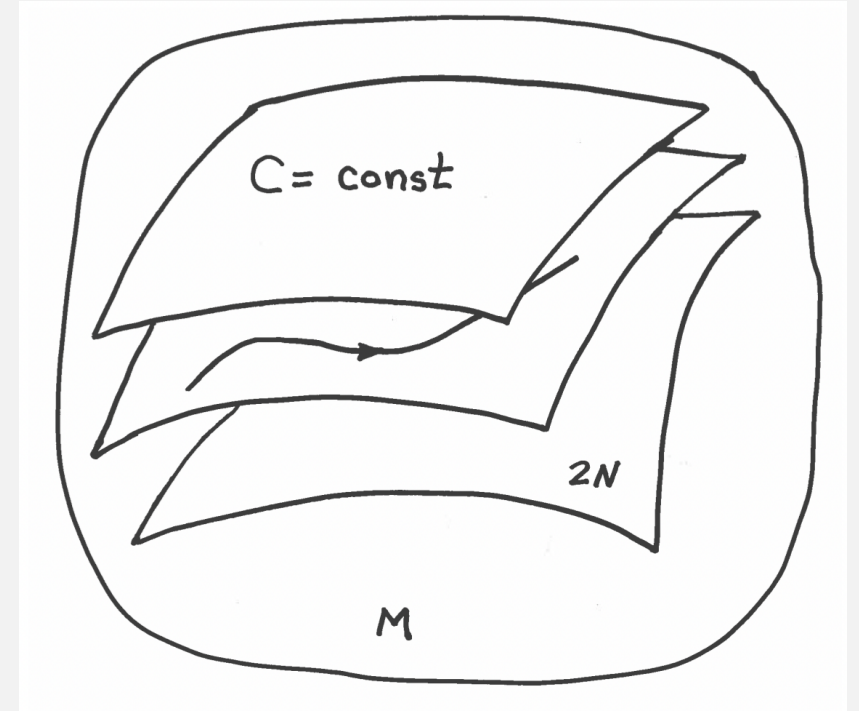
# BBM Equation



# Beyond Hamiltonian systems

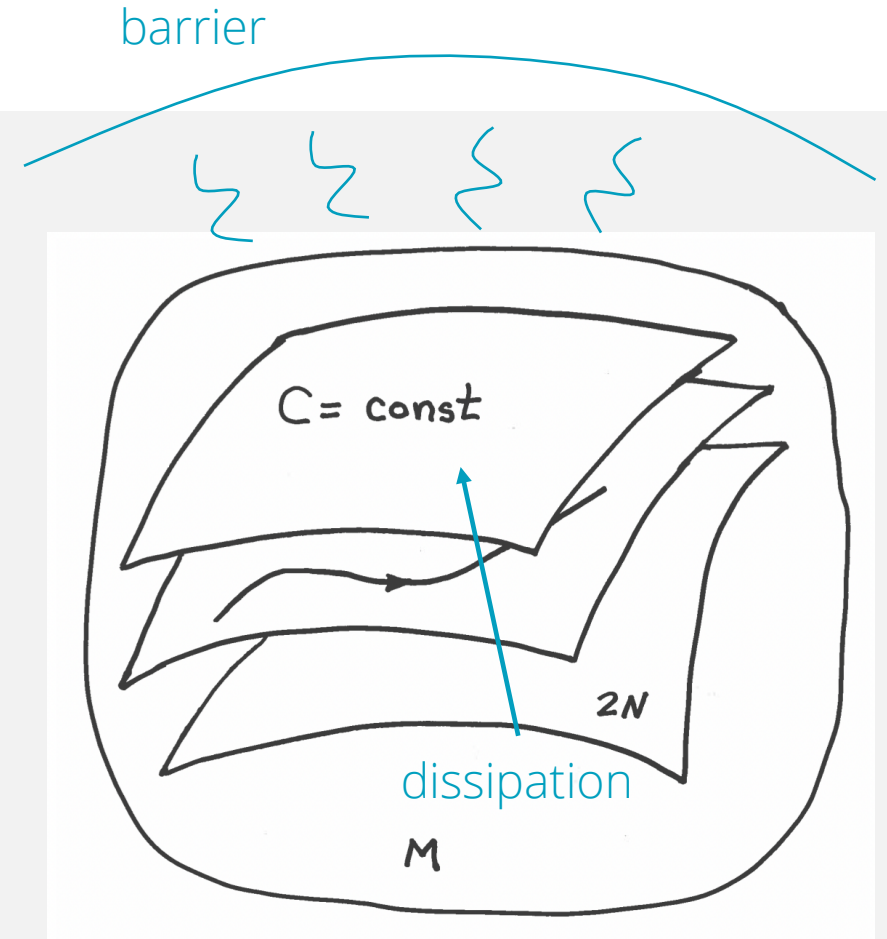
- Generalizes first law of thermodynamics.
  - Only useful for conservative systems!
- What about, e.g., dissipation?
  - Need *two* laws....
- Recall Casimir invariants:  $\{\cdot, C\} = \mathbf{L} \nabla C = \mathbf{0}$ .
  - Casimirs are potential **entropy** functions.

Illustration courtesy of P. J. Morrison



# Beyond Hamiltonian systems

- Notice,  $H' = H + \lambda^i C_i$  generates same dynamics as  $H$ .
  - But relaxes to different equilibria!
- Can we maintain a complete picture of the dynamics?
  - Choose  $S = C$  for some Casimir  $C$ .
- Analogue of *isolated* systems in thermodynamics.
  - Examples: Boltzmann equation, Vlasov with collisions.





# Metriplectic Systems

- Consider a system  $\dot{\mathbf{x}} = \{\mathbf{x}, E\} + [\mathbf{x}, S] = \mathbf{L}\nabla E + \mathbf{M}\nabla S$ .
  - Hamiltonian + Gradient:  $\mathbf{L}^\top = -\mathbf{L}, \mathbf{M}^\top = \mathbf{M}$ .
  - Energy and Entropy function(al)s  $E, S$ .
- Want to capture  $\dot{E} = 0, \dot{S} \geq 0$ .
  - How to appropriately “stitch together” reversible and irreversible parts?

# Metriplectic Systems

- Note,

$$\dot{E}(\mathbf{x}) = \dot{\mathbf{x}} \cdot \nabla E = \mathbf{L} \nabla E \cdot \nabla E + \mathbf{M} \nabla S \cdot \nabla E = \nabla S \cdot \mathbf{M} \nabla E,$$

$$\dot{S}(\mathbf{x}) = \dot{\mathbf{x}} \cdot \nabla S = \mathbf{L} \nabla E \cdot \nabla S + \mathbf{M} \nabla S \cdot \nabla S = -\nabla E \cdot \mathbf{L} \nabla S + |\nabla S|_{\mathbf{M}}^2$$

- Solution? Prescribed degeneracies!

- Choose  $\mathbf{L} \nabla S = \mathbf{M} \nabla E = \mathbf{0}$ .

# Example: Thermoelastic Double Pendulum

$$\begin{pmatrix} \dot{\mathbf{q}}_1 \\ \dot{\mathbf{q}}_2 \\ \dot{\mathbf{p}}_1 \\ \dot{\mathbf{p}}_2 \\ \dot{S}_1 \\ \dot{S}_2 \end{pmatrix} = \begin{pmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \\ -\frac{\partial}{\partial \mathbf{q}_1}(E_1 + E_2) \\ -\frac{\partial}{\partial \mathbf{q}_2}(E_1 + E_2) \\ T_1^{-1}T_2 - 1 \\ T_1T_2^{-1} - 1 \end{pmatrix}$$

- State variable

$$\mathbf{x} = (\mathbf{q}_1 \quad \mathbf{q}_2 \quad \mathbf{p}_1 \quad \mathbf{p}_2 \quad S_1 \quad S_2)^\top$$

$$L = \begin{pmatrix} \mathbf{0}_{4 \times 4} & \mathbf{S} \\ -\mathbf{S}^\top & \mathbf{0}_{6 \times 6} \end{pmatrix}, \quad M = \begin{pmatrix} \mathbf{0}_{8 \times 8} & \mathbf{0}_{8 \times 2} \\ \mathbf{0}_{2 \times 8} & \mathbf{T} \end{pmatrix}$$

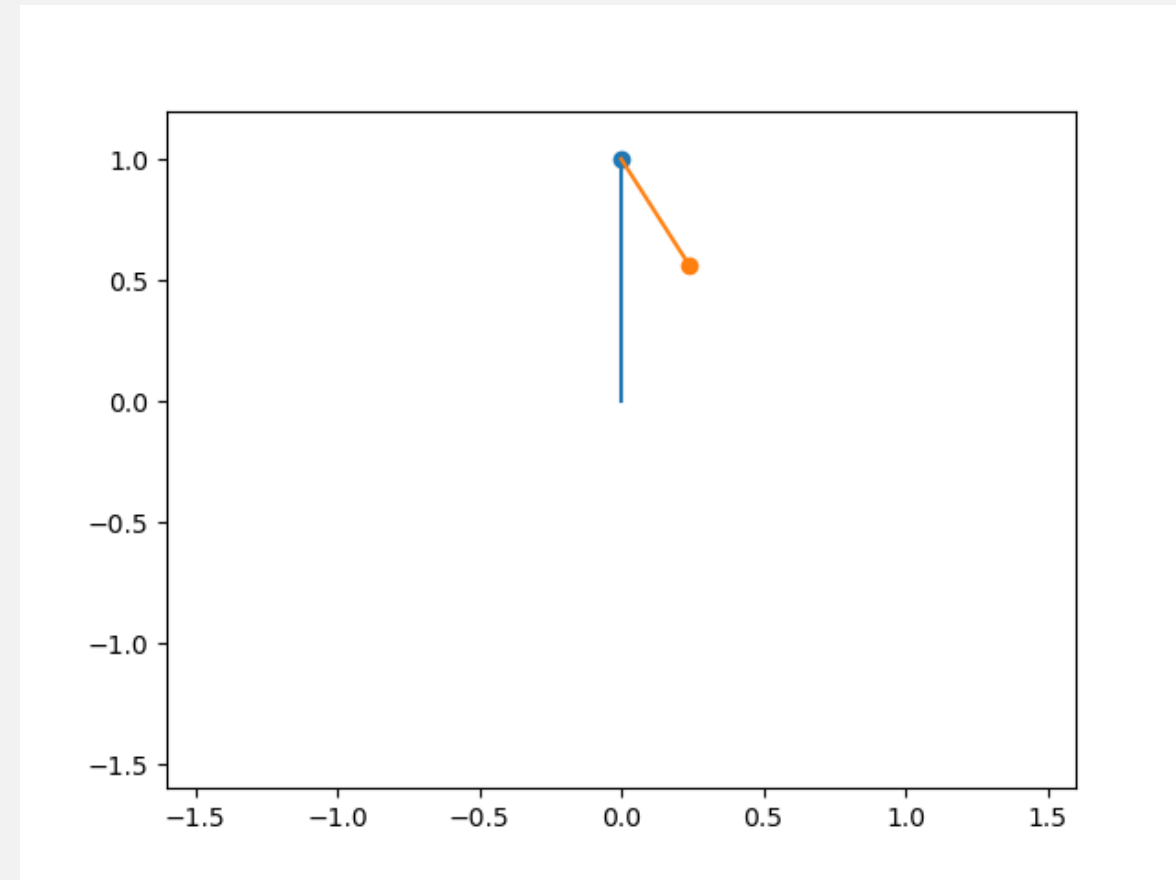
$$\mathbf{S} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, \quad \mathbf{T} = \begin{pmatrix} \frac{T_2}{T_1} & -1 \\ -1 & \frac{T_1}{T_2} \end{pmatrix} \quad T_i = \partial_{S_i} E_i$$

- Energy/Entropy:  $S = S_1 + S_2$

$$E = \frac{1}{2} (|\mathbf{p}_1|^2 + |\mathbf{p}_2|^2) + E_1 + E_2$$

$$E_i = \frac{1}{2} (\log \lambda_i)^2 + \log \lambda_i + e^{S_i - \log \lambda_i} - 1$$

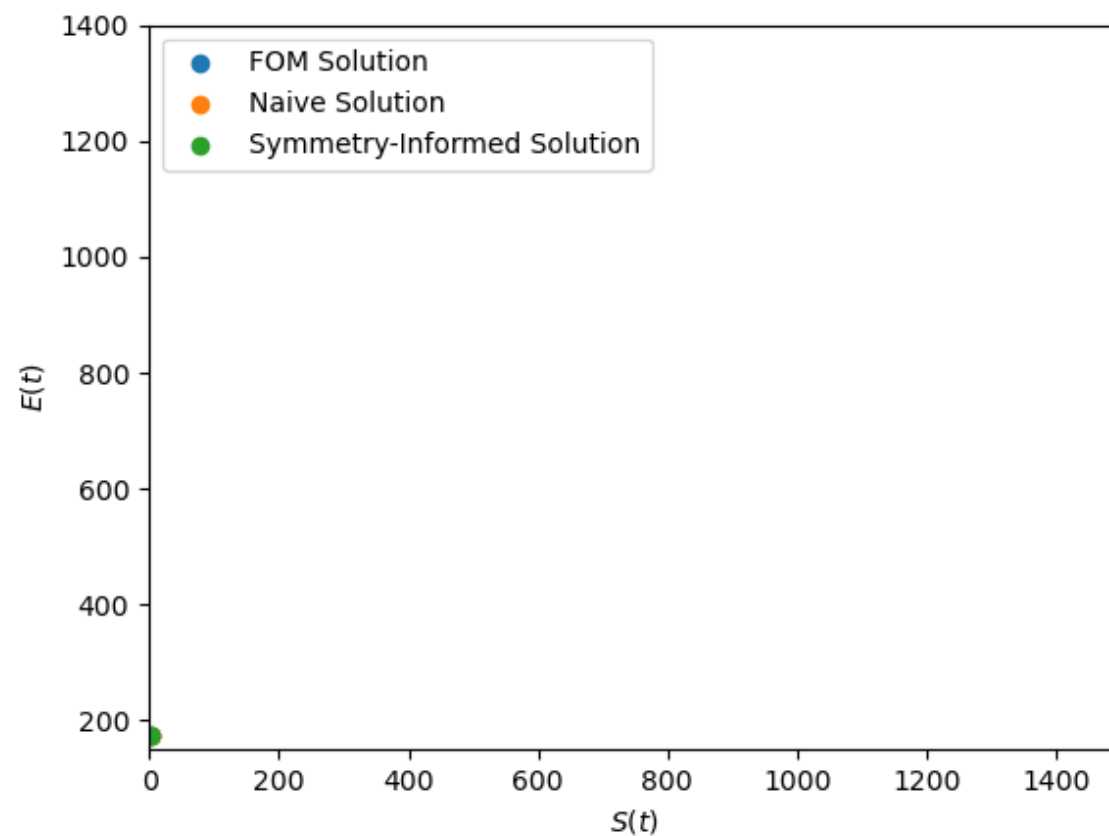
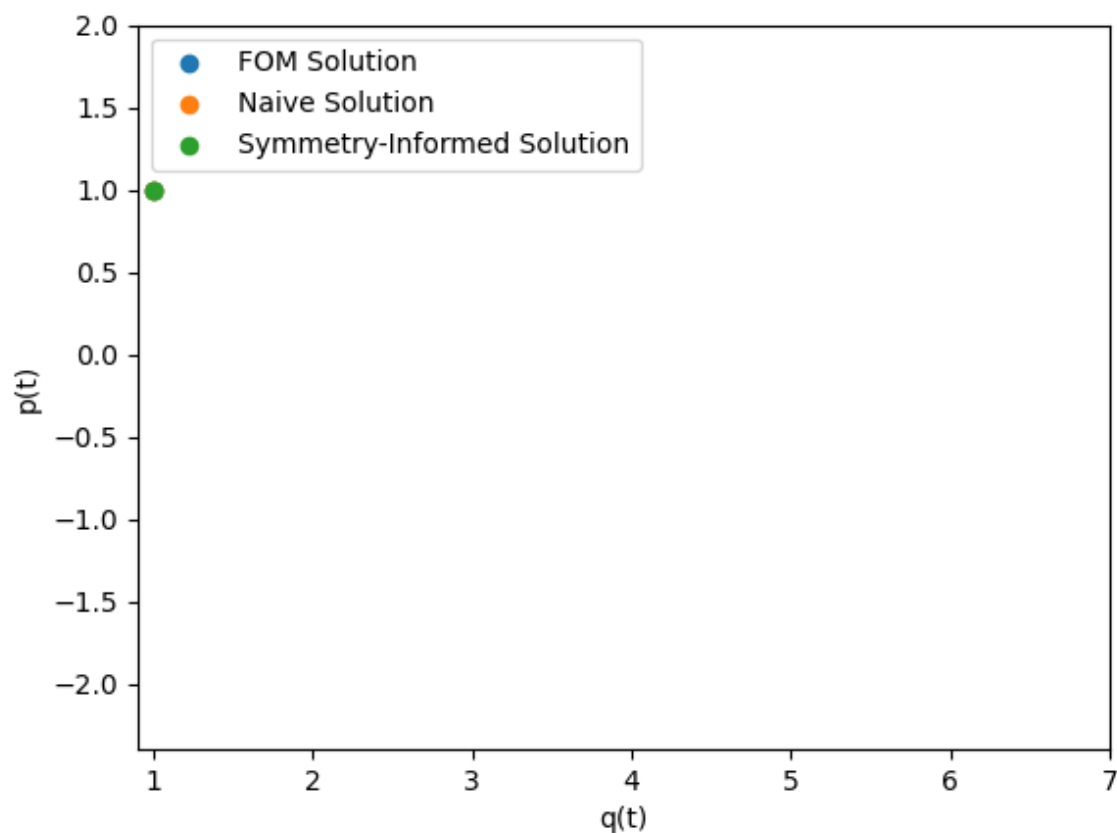
$$\lambda_1 = |\mathbf{q}_1|, \lambda_2 = |\mathbf{q}_2 - \mathbf{q}_1|$$



# Metriplectic SP-ROM

- Want to reduce  $\dot{\mathbf{x}} = \{\mathbf{x}, E\} + [\mathbf{x}, S] = \mathbf{L}\nabla E + \mathbf{M}\nabla S$ .
  - $\hat{\mathbf{M}} = \mathbf{U}^\top \mathbf{M} \mathbf{U}$  and  $\hat{\mathbf{L}} = \mathbf{U}^\top \mathbf{L} \mathbf{U}$  satisfy symmetries, but....
- Clearly  $\hat{\mathbf{L}}\nabla\hat{S} = \mathbf{U}^\top \mathbf{L} \mathbf{U} \mathbf{U}^\top \nabla S \neq \mathbf{0}$  (same for  $\hat{\mathbf{M}}\nabla\hat{E}$ ).
  - Metriplectic structure is not preserved!
  - No separation between reversible and irreversible parts.

# Does this matter?



# Metriplectic SP-ROM

- How to ensure  $\hat{\mathbf{L}}\nabla\hat{S} = \hat{\mathbf{M}}\nabla\hat{E} = \mathbf{0}$ .
  - Constrained optimization? Too expensive.
  - Penalty method? Too loose.
- Build in symmetries directly!
  - Can always parameterize at expense of **increasing tensor degree**.

# Metriplectic SP-ROM

- If  $\mathbf{L} \nabla S = \mathbf{0}$  and  $\mathbf{L}^\top = -\mathbf{L}$ , then  $\mathbf{L} = \boldsymbol{\xi}(\nabla S)$ .
  - $\boldsymbol{\xi}$  is order 3 totally antisymmetric tensor field  $\xi_{jk}^i \partial_i \otimes dx^j \otimes dx^k$ .
- If  $\mathbf{M} \nabla E = \mathbf{0}$  and  $\mathbf{M}^\top = \mathbf{M}$ , then  $\mathbf{M} = \boldsymbol{\zeta}(\nabla E, \nabla E)$ .
  - $\boldsymbol{\zeta}$  is rank 4 tensor field  $\zeta_{kjl}^i \partial_i \otimes dx^k \otimes dx^j \otimes dx^l$ .
  - Satisfies  $\zeta_{ikjl} = -\zeta_{kijl} = -\zeta_{iklj} = \zeta_{jlik}$ .

# Metriplectic SP-ROM

- Need to solve underdetermined systems!
  - For  $L$ :  $\xi_{jk}^i \partial^k S = L_j^i$ .
  - For  $M$ :  $\zeta_{kjl}^i (\partial^k E)(\partial^l E) = M_j^i$ .
- These are design decisions.



# Why does this help?

- Galerkin projection  $\tilde{\mathbf{x}} = \mathbf{U}\hat{\mathbf{x}}$  yields:

$$\dot{\tilde{\mathbf{x}}} = \mathbf{U}^\top \boldsymbol{\xi} (\nabla S) \nabla E + \mathbf{U}^\top \boldsymbol{\zeta} (\nabla E, \nabla E) \nabla S$$

- If instead

$$\dot{\tilde{\mathbf{x}}} = \overset{\hat{\mathbf{L}}}{\boldsymbol{\xi} \left( \nabla \hat{S} \right)} \nabla \hat{E} + \overset{\hat{\mathbf{M}}}{\boldsymbol{\zeta} \left( \nabla \hat{E}, \nabla \hat{E} \right)} \nabla \hat{S}$$

- For some  $\hat{\boldsymbol{\xi}}, \hat{\boldsymbol{\zeta}}$  with the right symmetries, then things will work!

# Simplifying M

- Consider the eigenvalue decomposition  $\mathbf{M} = \sum_{\alpha=1}^r \lambda_{\alpha} \mathbf{m}^{\alpha} \otimes \mathbf{m}^{\alpha}$ .
  - Suppose  $\mathbf{m}^{\alpha} = \mathbf{A}^{\alpha} \nabla E$  where  $A_{ij}^{\alpha} = -A_{ji}^{\alpha}$ .
- Can choose  $\zeta = \sum_{\alpha=1}^r \lambda_{\alpha} \mathbf{A}^{\alpha} \otimes \mathbf{A}^{\alpha}$ .
- Automatically preserves symmetries.
  - Requires solving  $A_{ij}^{\alpha} \partial^j E = m_i^{\alpha}$  (design decision).

# Metriplectic ROM

- All together:  $\dot{\hat{\mathbf{x}}} = \hat{\mathbf{L}} \nabla \hat{E} + \hat{\mathbf{M}} \nabla \hat{S}$   
$$= \hat{\xi} \left( \nabla \hat{S} \right) \nabla \hat{E} + \hat{\zeta} \left( \nabla \hat{E}, \nabla \hat{E} \right) \nabla \hat{S}$$
- Theorem: if  $\mathbf{U}$  is a POD basis and  $\xi, \zeta, \nabla E, \nabla S$  are regular,

$$\int_{t=0}^T \|\mathbf{x}(t) - (\mathbf{x}_0 + \mathbf{U} \hat{\mathbf{x}}(t))\|^2 \leq C \sum_{j>n} \sigma_j^2 \rightarrow 0$$

# Recap: what's necessary?

- Given:  $\mathbf{L}, \mathbf{M}, \nabla E, \nabla S$  defining metriplectic system.
- Compute eigenvalue decomposition  $\mathbf{M} = \sum_{\alpha=1}^r \lambda_{\alpha} \mathbf{m}^{\alpha} \otimes \mathbf{m}^{\alpha}$ .
- Solve  $\mathbf{L} = \boldsymbol{\xi}(\nabla S)$  and  $\mathbf{m}^{\alpha} = \mathbf{A}^{\alpha} \nabla E$  (freedom here).
- Compute  $\hat{\boldsymbol{\xi}} = \mathbf{U}^{\top} \boldsymbol{\xi}(\mathbf{U}) \mathbf{U}$  and  $\hat{\mathbf{A}}^{\alpha} = \mathbf{U}^{\top} \mathbf{A}^{\alpha} \mathbf{U}$ .
- \*Assemble RO quantities:  $\hat{\mathbf{L}} = \hat{\boldsymbol{\xi}}(\nabla \hat{S})$ ,  $\hat{\mathbf{M}} = \sum_{\alpha=1}^r \lambda_{\alpha} \hat{\mathbf{A}}^{\alpha} \nabla \hat{E} \otimes \hat{\mathbf{A}}^{\alpha} \nabla \hat{E}$ .

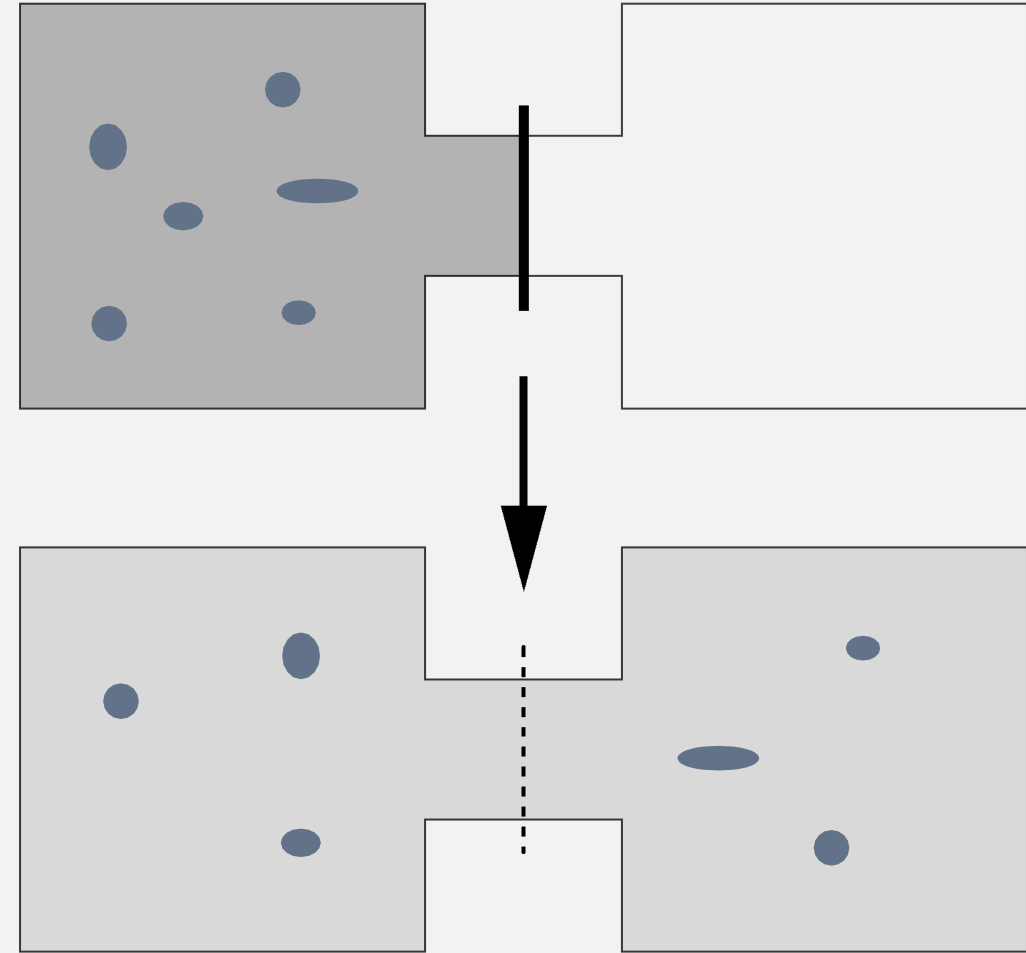
# Toy example: two gas containers

- Variables  $q, p, S_1, S_2$  .

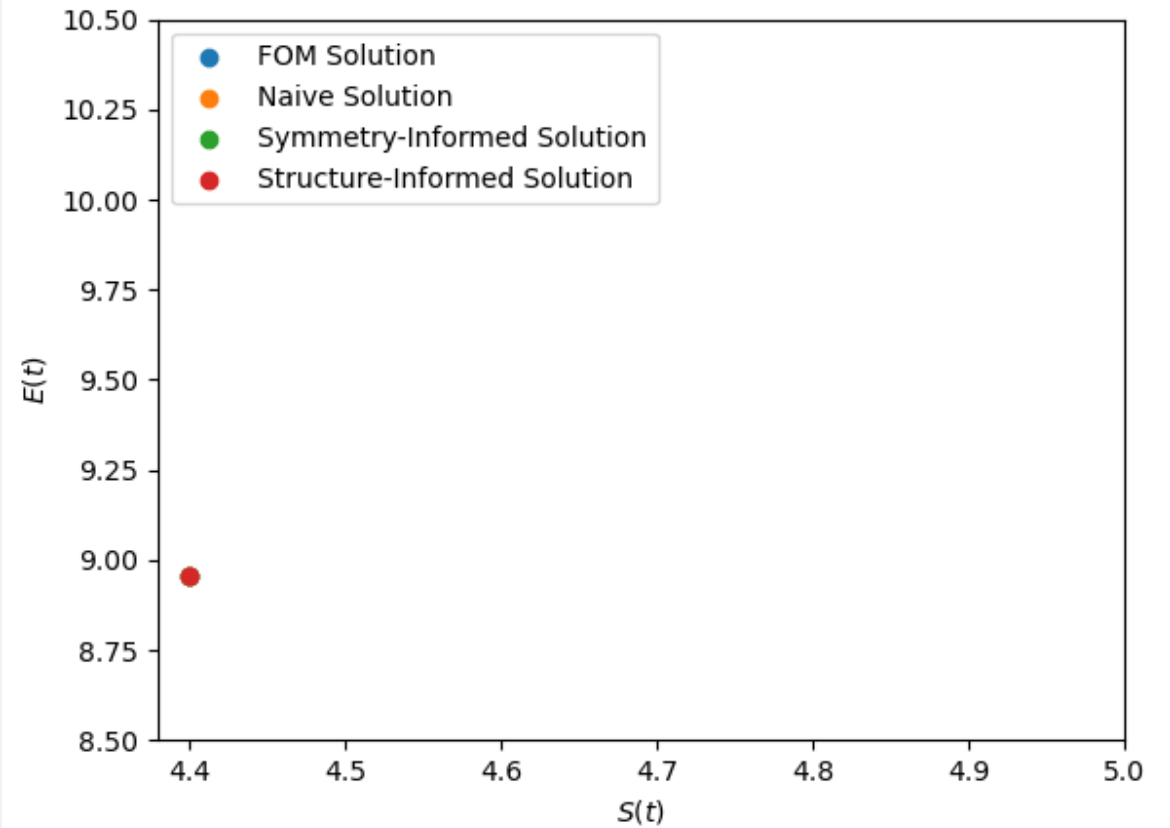
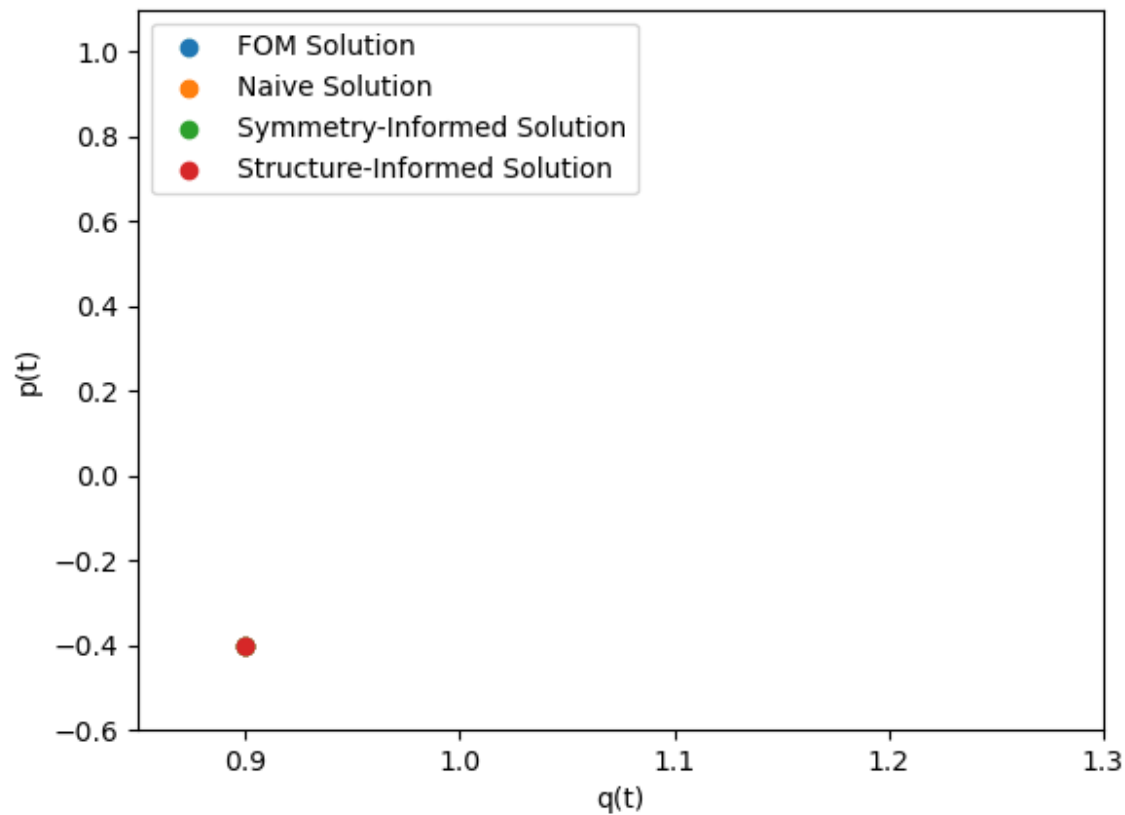
- Energy:  $E(\mathbf{x}) = \frac{p^2}{2m} + \underbrace{\left( \frac{e^{\frac{S_1}{Nk_B}}}{\hat{c}q} \right)^{\frac{2}{3}}}_{E_1} + \underbrace{\left( \frac{e^{\frac{S_2}{Nk_B}}}{\hat{c}(2-q)} \right)^{\frac{2}{3}}}_{E_2}$  .

- Entropy:  $S(\mathbf{x}) = S_1 + S_2$  .  $T_i = \partial_{S_i} E_i$

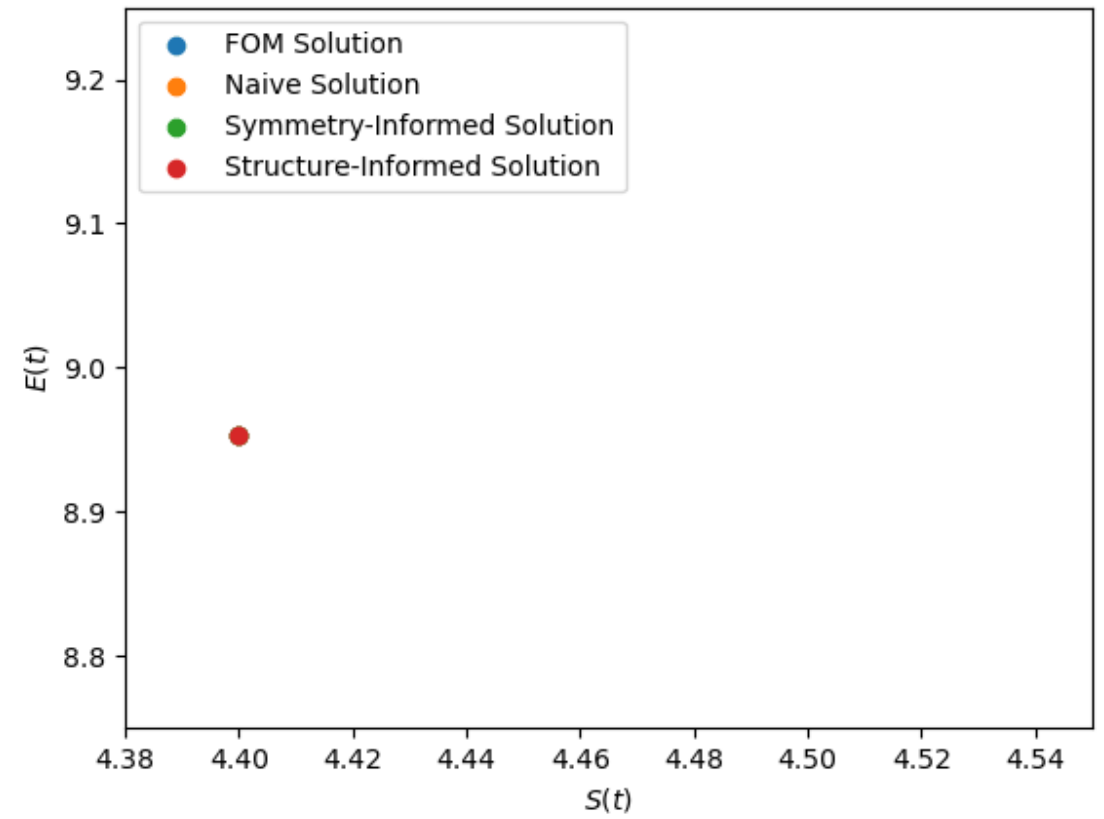
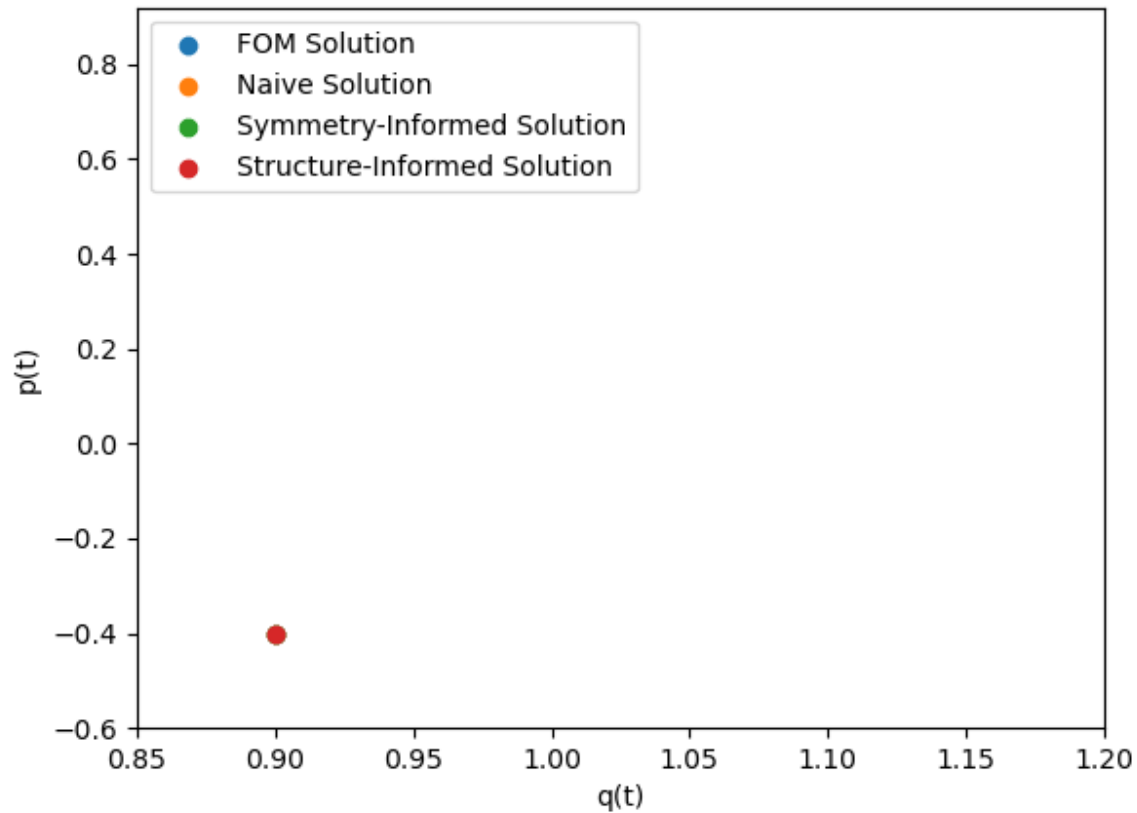
$$\mathbf{L} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad \mathbf{M} = \gamma \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & T_1^{-2} & -(T_1 T_2)^{-1} \\ 0 & 0 & -(T_1 T_2)^{-1} & T_2^{-2} \end{pmatrix}$$



# Results: two gas containers (3 modes)



# Results: two gas containers (4 modes)



# Is this enough?

- **Good:** Explicit recipe, exact structure preservation, convergence.
- **Bad:** Requires storage of two **sparse degree 3 tensors**.
  - $\mathbf{\xi}$  contains  $\binom{N}{3} \ll N^3$  independent components.
  - $\mathbf{A}$  contains  $r \binom{N}{2} \ll N^3$  independent components.
- Still too expensive. Any way to reduce cost?



# Exterior algebra (EA)

- Recall the EA on  $V$ :  $\Lambda(V) = T(V) / \{\mathbf{v} \otimes \mathbf{v} \mid \mathbf{v} \in V\}$ 
  - For example:  $\mathbf{v} \wedge \mathbf{w} = \frac{1}{2} (\mathbf{v} \otimes \mathbf{w} - \mathbf{w} \otimes \mathbf{v})$  (DON'T store this)
- Advantage? Can multiply analytically first.
  - $\mathbf{v} \cdot (\mathbf{w}_1 \wedge \mathbf{w}_2 \wedge \dots \wedge \mathbf{w}_n) = \sum_{i=1}^n (-1)^{i-1} (\mathbf{v} \cdot \mathbf{w}_i) \mathbf{w}_1 \wedge \dots \wedge \widehat{\mathbf{w}_i} \wedge \dots \wedge \mathbf{w}_n$
  - So,  $\mathbf{v} \cdot (\mathbf{u} \wedge \mathbf{W}^k) = (\mathbf{v} \cdot \mathbf{u}) \mathbf{W}^k - \mathbf{u} \wedge (\mathbf{v} \cdot \mathbf{W}^k)$ .

# Why is this useful?

- Identify  $\mathbf{L}$  with the bivector sum  $\mathbf{L} = \sum_{j < i} L^{ij} \mathbf{e}_i \wedge \mathbf{e}_j$ .
- We have  $\mathbf{L} \cdot \mathbf{v} = (-1)^{1(2+1)} \mathbf{v} \cdot \mathbf{L} = \sum_{j < i} L^{ij} v_j \mathbf{e}_i - \sum_{j > i} L^{ji} v_j \mathbf{e}_i = \mathbf{L} \mathbf{v}$ .
- So, define  $\boldsymbol{\xi} = \mathbf{L} \wedge \mathbf{s}_{k_0}$ .  $L^{ii} = 0$ 
  - Because of our design decision,  $\mathbf{s}_{k_0} = \mathbf{e}_{k_0} / S^{k_0}$ .
- $\boldsymbol{\xi} (\nabla S) = (\nabla S \cdot \mathbf{L}) \wedge \mathbf{s}_{k_0} + \underbrace{(\mathbf{s}_{k_0} \cdot \nabla S)}_{=1} \mathbf{L} = -\mathbf{L} \nabla S \wedge \mathbf{s}_{k_0} + \mathbf{L} = \mathbf{L},$

# Why is this useful?

- Now, structure preservation is guaranteed by

$$\hat{\mathbf{L}} = \hat{\xi} \left( \nabla \hat{S} \right) = (\mathbf{U}^\top \mathbf{L} \mathbf{U} \wedge \mathbf{U}^\top \mathbf{s}_{k_0}) \cdot \nabla \hat{S} = \boxed{\bar{\mathbf{L}} \nabla \hat{S} \wedge \mathbf{s}_{k_0}} + \left( \hat{\mathbf{s}}_{k_0} \cdot \nabla \hat{S} \right) \bar{\mathbf{L}}$$

since

$$\hat{\mathbf{L}} \nabla \hat{S} = \boxed{(0) \hat{\mathbf{s}}_{k_0} + \left( \hat{\mathbf{s}}_{k_0} \cdot \nabla \hat{S} \right) \bar{\mathbf{L}} \nabla \hat{S}} - \left( \hat{\mathbf{s}}_{k_0} \cdot \nabla \hat{S} \right) \bar{\mathbf{L}} \nabla \hat{S} .$$

- Only need to store  $\bar{\mathbf{L}}$  and  $\hat{\mathbf{s}}_{k_0} = \mathbf{U}^{k_0} / S^{k_0} !!$

# Savings?

$$\mathbf{m}^\alpha \leftarrow \sqrt{\lambda_\alpha} \mathbf{m}^\alpha$$

- Similarly, choose  $\mathbf{A}^\alpha = \mathbf{a}_{k_1}^\alpha \wedge \mathbf{e}_{k_1}$  where  $\mathbf{a}_{k_1} = \mathbf{m}^\alpha / E^{k_1}$ .
- Then,  $\hat{\mathbf{A}}^\alpha = \hat{\mathbf{a}}_{k_1}^\alpha \wedge \mathbf{U}^{k_1}$  where  $\hat{\mathbf{a}}_{k_1}^\alpha = \mathbf{U}^\top \mathbf{m}^\alpha / E^{k_1}$ .
- Leads to  $\hat{\mathbf{A}}^\alpha \nabla \hat{E} = \left( \nabla \hat{E} \cdot \mathbf{U}^{k_1} \right) \hat{\mathbf{a}}_{k_1}^\alpha - \left( \nabla \hat{E} \cdot \hat{\mathbf{a}}_{k_1}^\alpha \right) \mathbf{U}^{k_1}$
- No need to store SS matrices.

# Savings?

- We have  $\hat{\xi} : n^3 \rightarrow n + \binom{n}{2},$

$$\hat{\mathbf{A}} : rn^2 \rightarrow n(r + 1)$$

- Makes metriplectic ROM feasible for larger problems.
  - *Make the algebra work for you!*

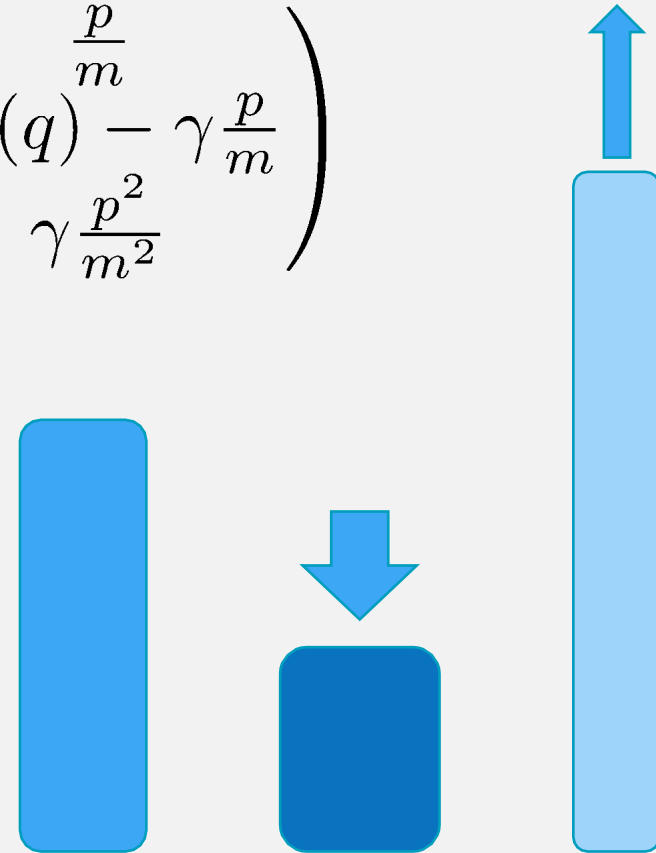
# Example: Damped Thermoelastic Rod

- 1-D elastic rod with coordinate  $s$ .
  - damped Hamiltonian system with friction.

$$\begin{pmatrix} \dot{q} \\ \dot{p} \\ \dot{S} \end{pmatrix} = \begin{pmatrix} \frac{p}{m} \\ V'(q) - \gamma \frac{p}{m} \\ \gamma \frac{p^2}{m^2} \end{pmatrix}$$

$$\begin{aligned} E(p, q, e) &= H(p, q) + S(e) \\ &= \int_0^\ell \left( \frac{p(s)^2}{2m(s)} + V(q(s)) \right) + \int_0^\ell e(s) \end{aligned}$$

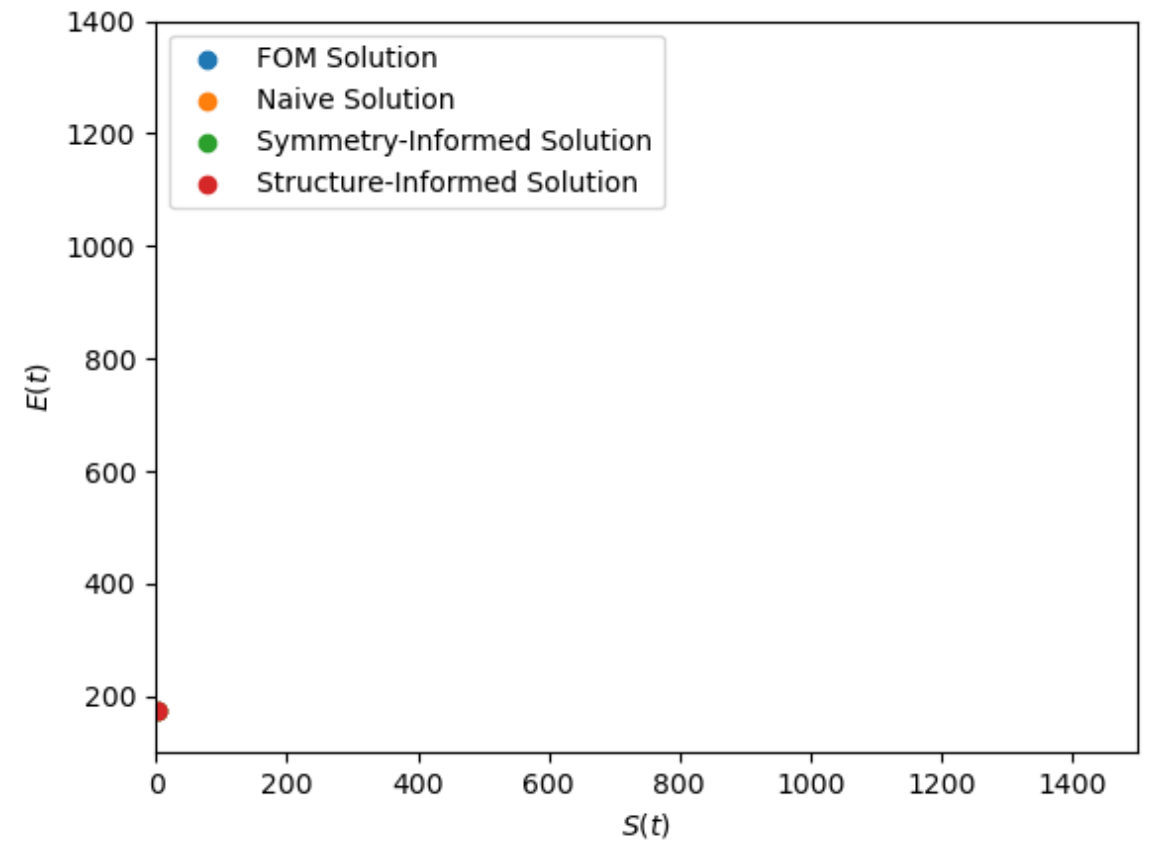
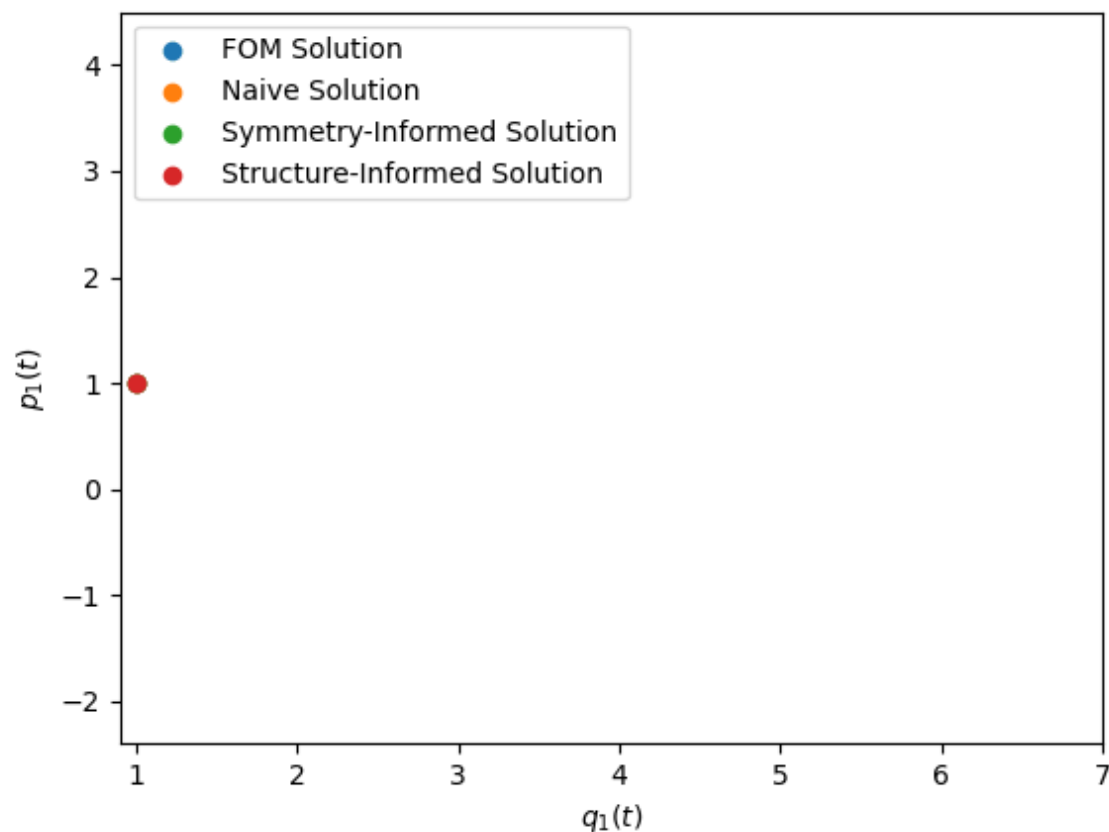
- $V$  a given potential function.



# Example: Damped Thermoelastic Rod

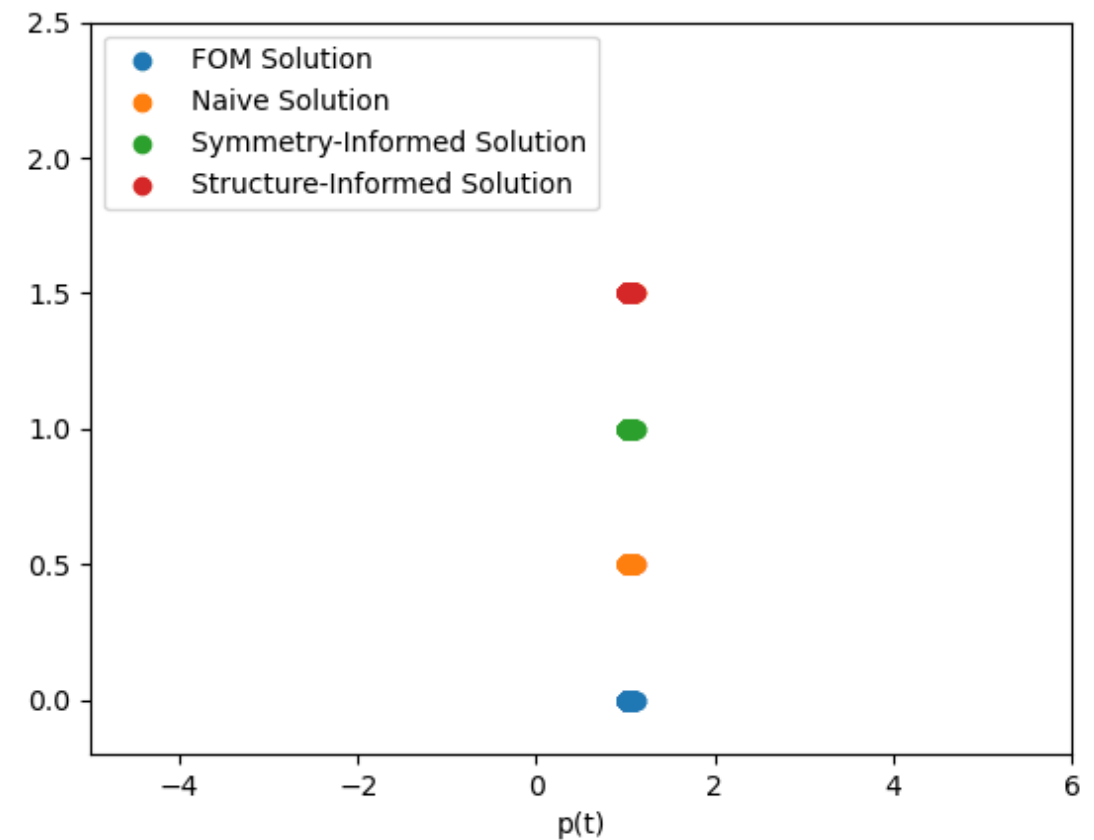
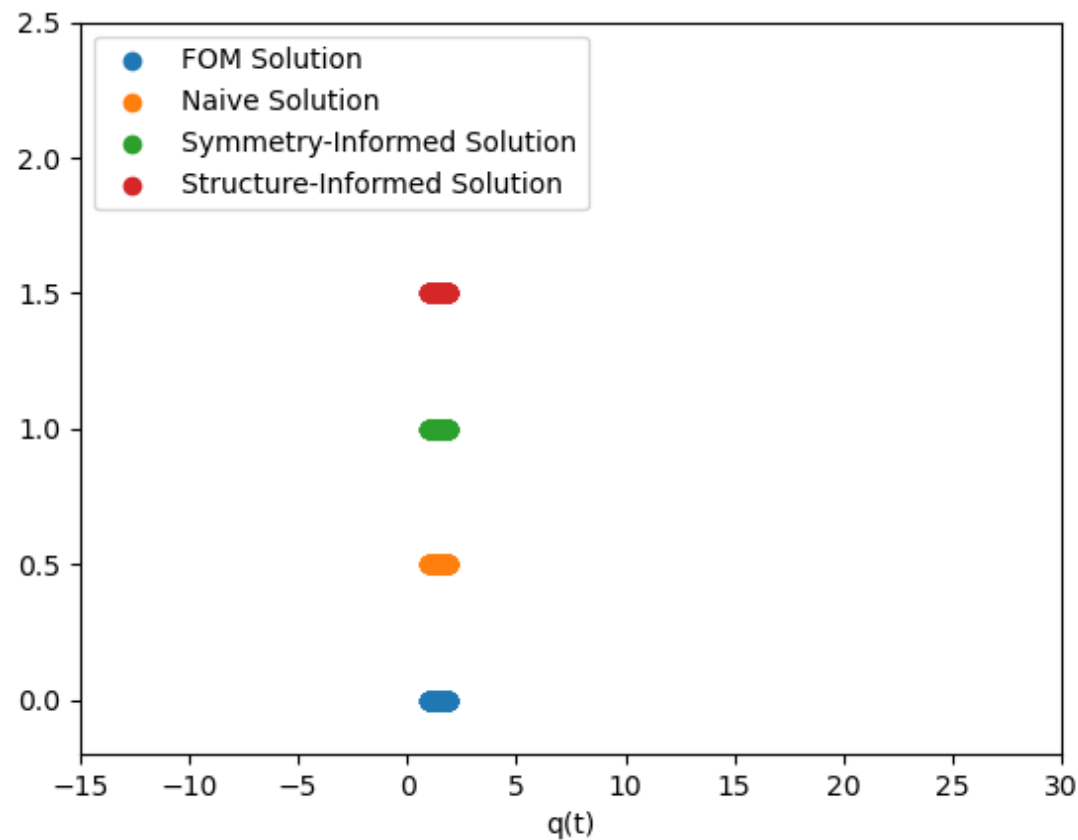
- Discretization yields  $\mathbf{x} = (\mathbf{q} \quad \mathbf{p} \quad S)^\top \in \mathbb{R}^{2N+1}$ .  $\nabla S = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$
- $$\mathbf{L} = \begin{pmatrix} \mathbf{0}_{N \times N} & \mathbf{I} & \mathbf{0} \\ -\mathbf{I} & \mathbf{0}_{N \times N} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & 0 \end{pmatrix}, \quad \mathbf{M}(\mathbf{x}) = \gamma \begin{pmatrix} \mathbf{0}_{N \times N} & \mathbf{0}_{N \times N} & \mathbf{0} \\ \mathbf{0}_{N \times N} & \mathbf{I} & -\frac{\mathbf{p}}{m} \\ \mathbf{0} & -\frac{\mathbf{p}^\top}{m} & \left(\frac{|\mathbf{p}|}{m}\right)^2 \end{pmatrix} \quad \nabla E(\mathbf{x}) = \begin{pmatrix} \mathbf{V}'(\mathbf{q}) \\ \frac{\mathbf{p}}{m} \\ 1 \end{pmatrix}$$
- Notice,  $\mathbf{M} = \sum_{\alpha=1}^N \mathbf{m}^\alpha \otimes \mathbf{m}^\alpha$ ,  $\mathbf{m}^\alpha = \sqrt{\gamma} \left( \mathbf{0} \quad \mathbf{e}_\alpha \quad -\frac{p_\alpha}{m} \right)^\top$ ,
- Can choose  $\hat{\boldsymbol{\xi}} = \bar{\mathbf{L}} \wedge \mathbf{U}^{2N+1}$ ,  $\hat{\mathbf{A}}^\alpha(\tilde{\mathbf{x}}) = \mathbf{U}^\top \mathbf{m}^\alpha(\tilde{\mathbf{x}}) \wedge \mathbf{U}^{2N+1}$

# Results: Damped Thermoelastic Rod (10 modes)

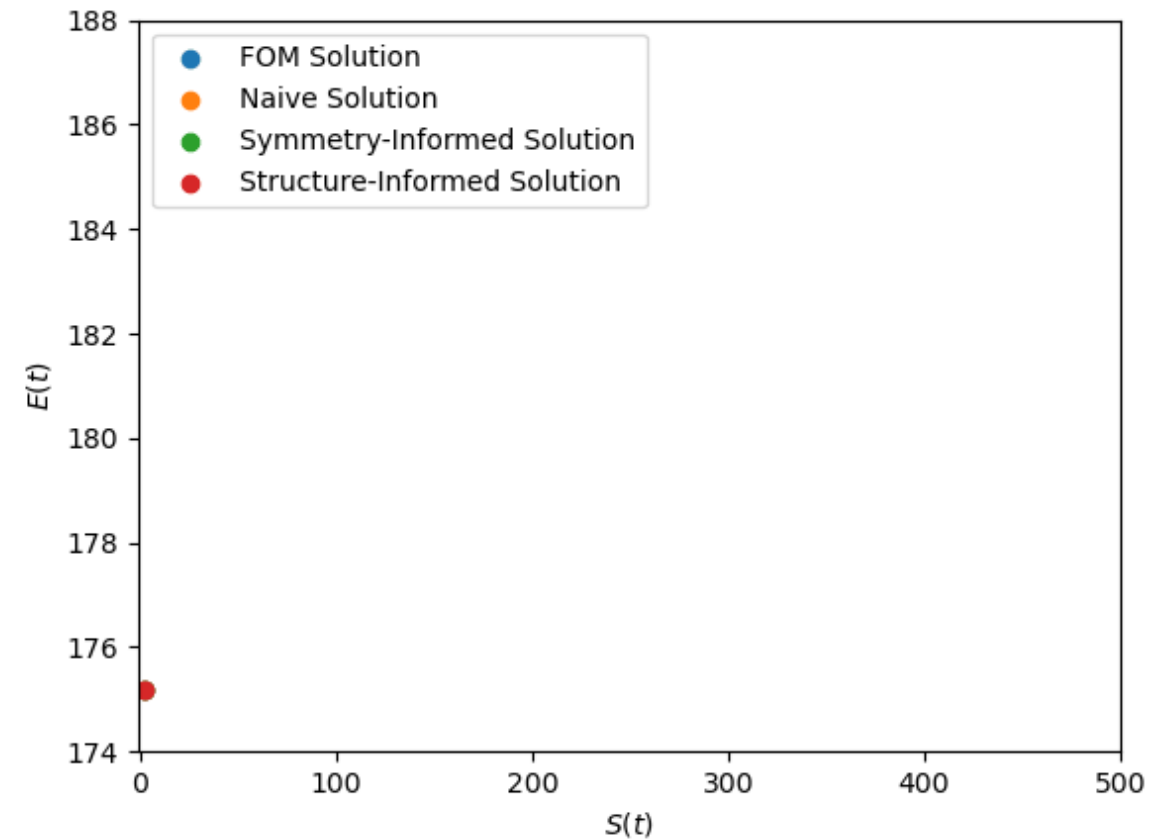
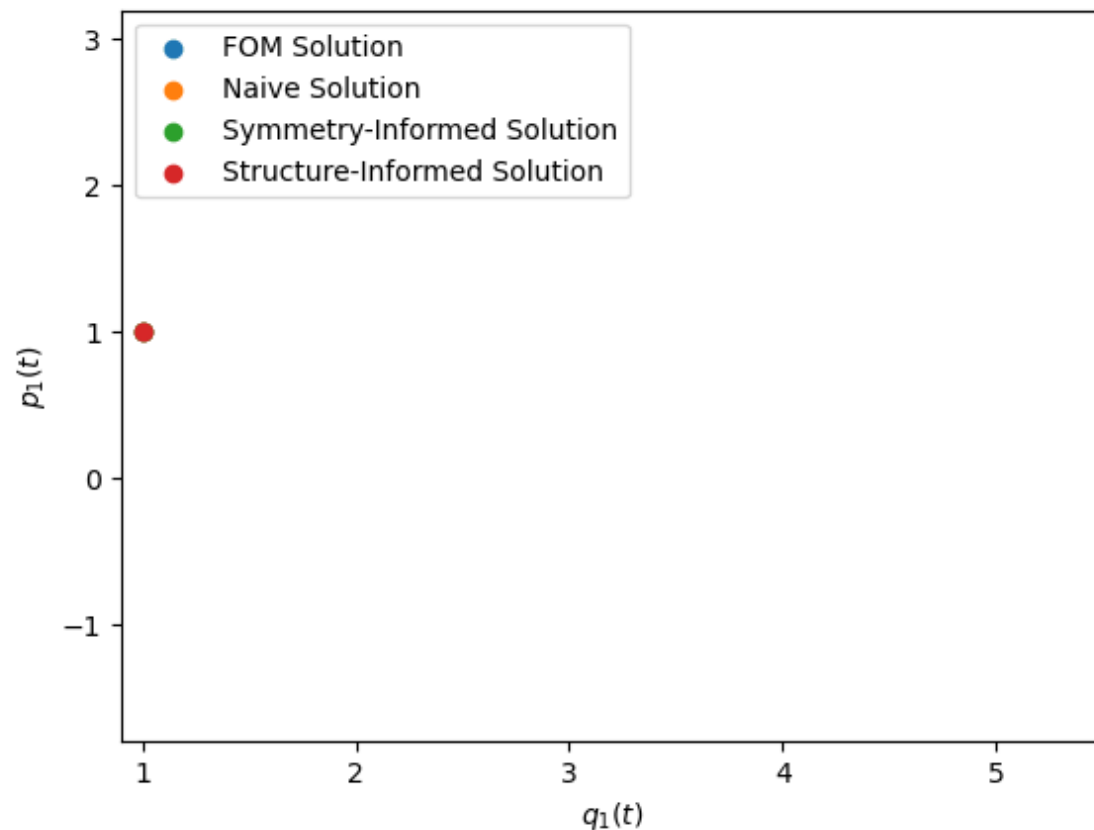




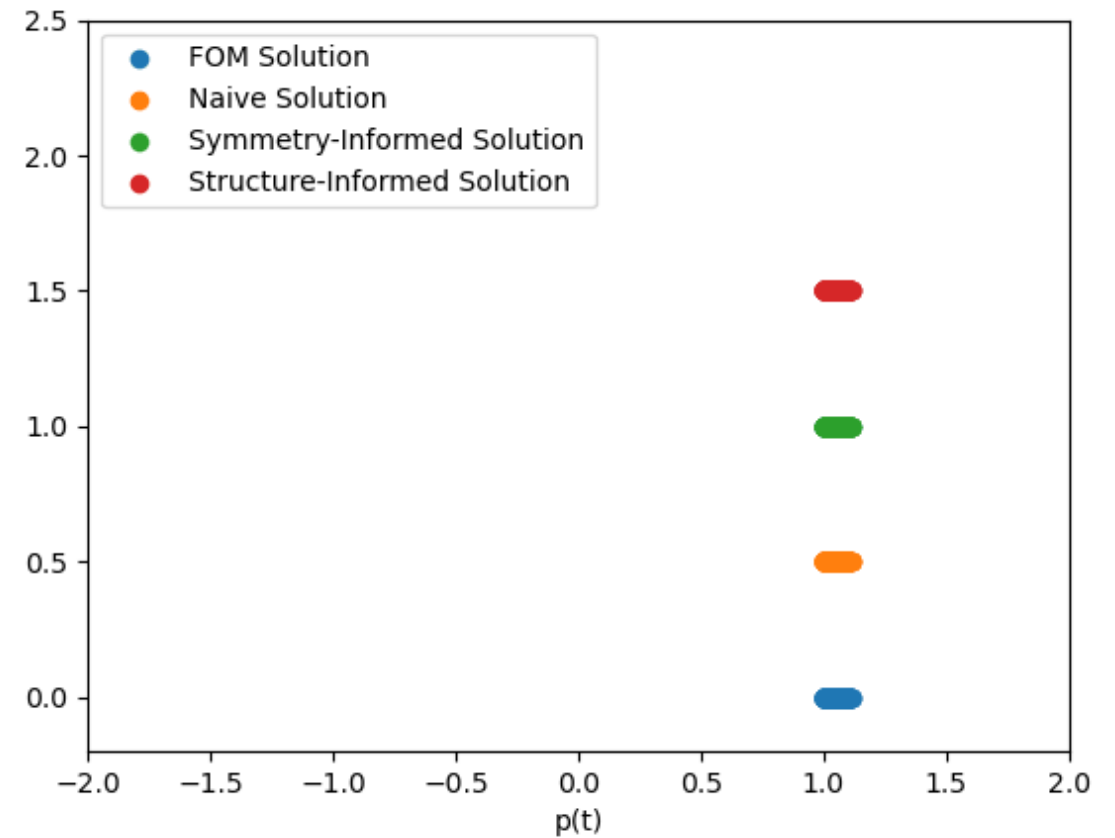
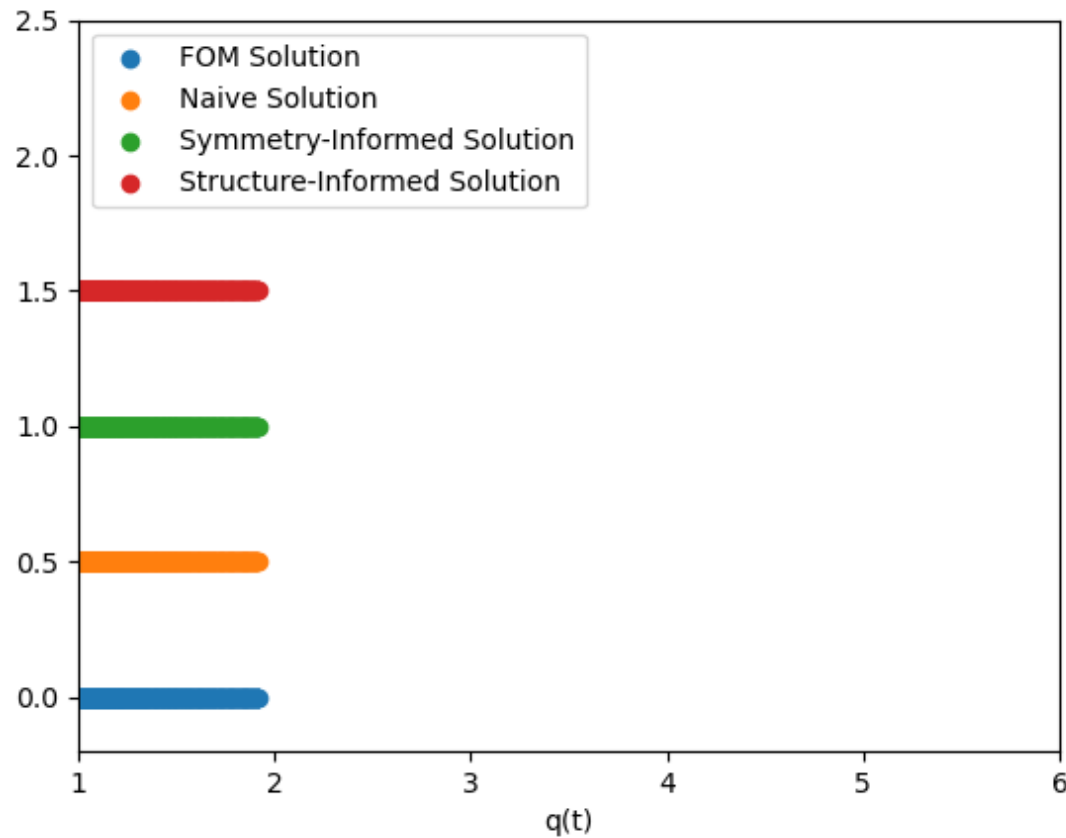
# Results: Damped Thermoelastic Rod (10 modes)



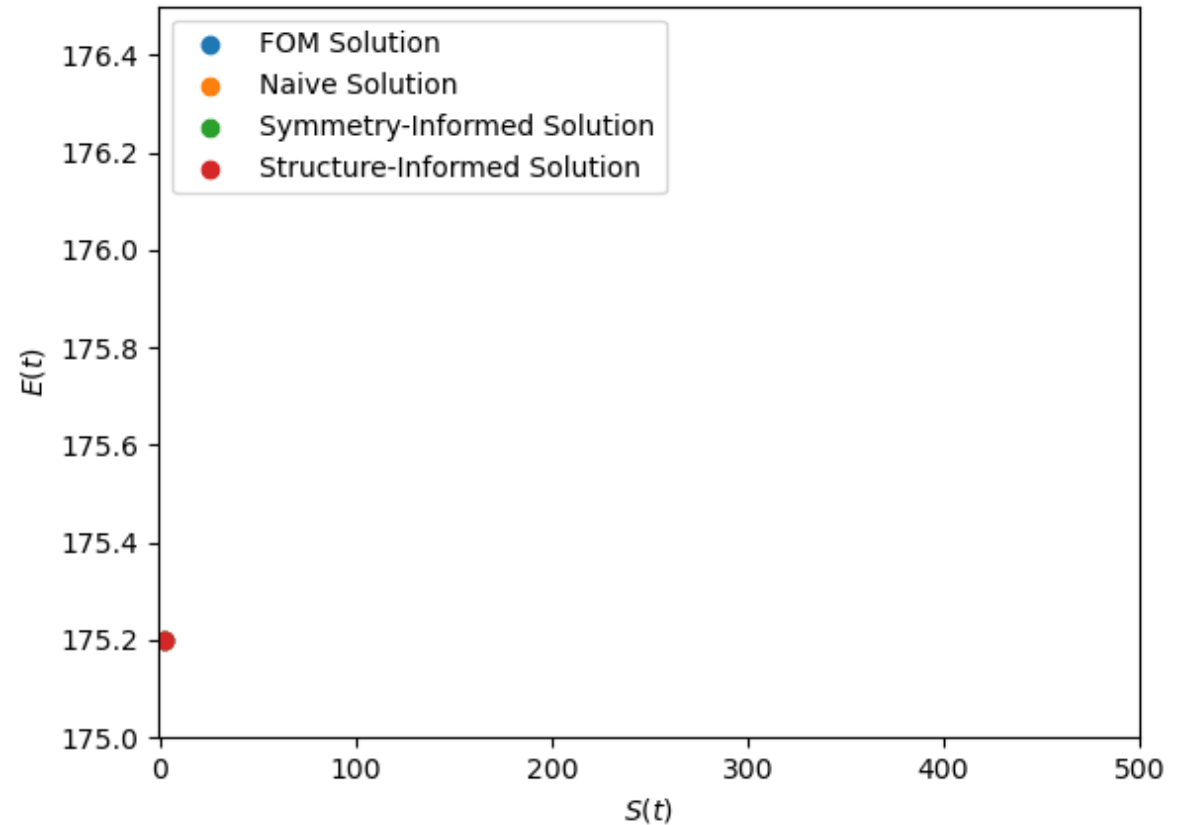
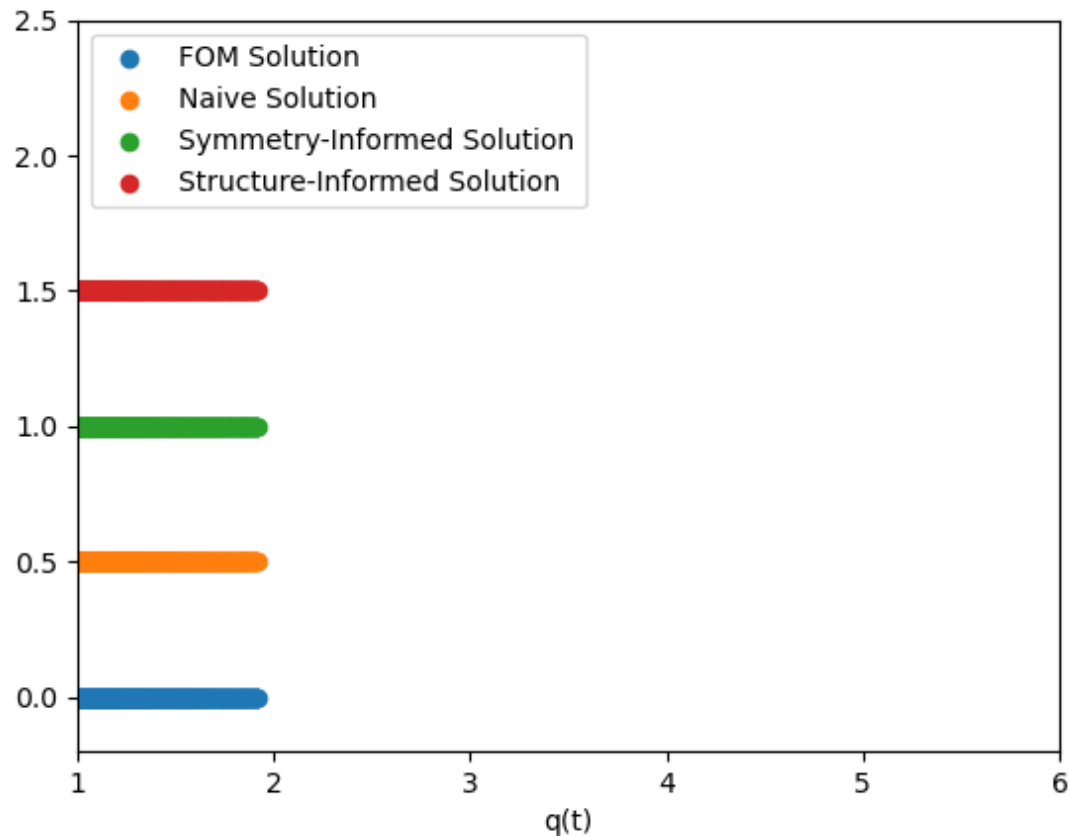
# Results: Damped Thermoelastic Rod (60 modes)



# Results: Damped Thermoelastic Rod (60 modes)



# Results: Damped Thermoelastic Rod (120 modes)



# What's next?

- Structure-preserving hyper-reduction for nonlinearities.
  - Need non-intrusive version!!
- Good methods for Lagrangian systems.
  - Extension to Euler-Poincare, Lie-Poisson.

$$\frac{\partial f}{\partial t} = -\mathcal{J}^{ij} H_j f_i + \frac{1}{2} D \mathcal{J}^{ik} \frac{\partial}{\partial x^i} \left[ f \mathcal{J}^{jk} \frac{\partial}{\partial x^j} (\log f + \beta H) \right],$$

$$\{F, G\} = \int_{\Omega} f \frac{\partial}{\partial x^i} \left( \frac{\delta F}{\delta f} \right)_{\beta} \mathcal{J}^{ij} \frac{\partial}{\partial x^j} \left( \frac{\delta G}{\delta f} \right)_{\beta} dV,$$

$$[F, G] = \frac{D}{2} \int_{\Omega} f \frac{\partial}{\partial x^i} \left( \frac{\delta F}{\delta f} \right)_{\beta} \mathcal{J}^{ik} \mathcal{J}^{jk} \frac{\partial}{\partial x^j} \left( \frac{\delta G}{\delta f} \right)_{\beta} dV,$$

Naoki Sato, Dissipative brackets for the Fokker-Planck equation in Hamiltonian systems and characterization of metriplectic manifolds, Physica D: Nonlinear Phenomena, Volume 411, 2020.

# Thank you!

Contact: [adgrube@sandia.gov](mailto:adgrube@sandia.gov)

## References:

A. Gruber, M. Gunzburger, L. Ju, Z. Wang, “Energetically Consistent Model Reduction for Metriplectic Systems”, CMAME, 2023

A. Gruber, I. Tezaur, “Canonical and Noncanonical Hamiltonian Operator Inference”, (coming soon!)

## Codes:

<https://github.com/agrubertx/metriplectic> POD-ROM

<https://github.com/ikalash/HamiltonianOpInf>