



Modeling, simulation and control of multi-physical systems: A change of paradigm.

Volker Mehrmann
Institute f. Mathematics
TU Berlin

Research Center MATHEON
Mathematics for key technologies





- 1 **General remarks**
- 2 Energy based modeling
- 3 Port-Hamiltonian PDEs
- 4 Model reduction, surrogate models
- 5 Time-discretization for pHDAEs
- 6 Conclusions



- ▶ Modern key technologies require **Modeling, Simulation, and Optimization/control (MSO)** of complex dynamical systems.
- ▶ Most real world systems are **multi-physics systems**, combining components from different physical domains, and with different accuracies and scales in the components.
- ▶ Modeling becomes **exceedingly automatized**, linking subsystems or numerical methods in a network fashion.
- ▶ Models of real world systems have to adapt to changes in the system during life time. **Digital Twins.**
- ▶ Modeling, analysis, numerics, control, optimization, data science techniques **should go hand in hand.**
- ▶ Most real world (industrial) models need **model reduction** for data assimilation, optimization and control.

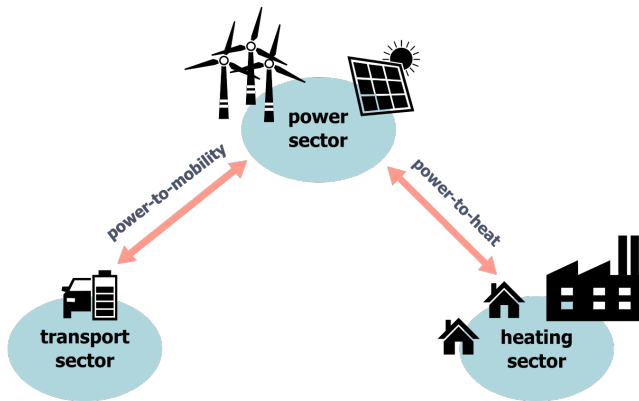
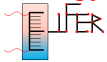


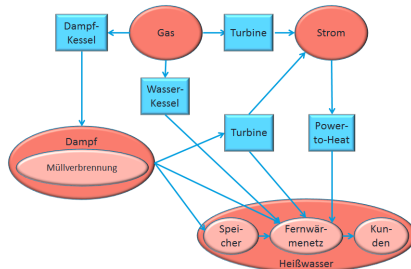
Figure: Sector coupling and the power-to-X concept

German Ministry of Education and Research (BMBF)

Energy efficiency via intelligent district heating networks (EiFer)



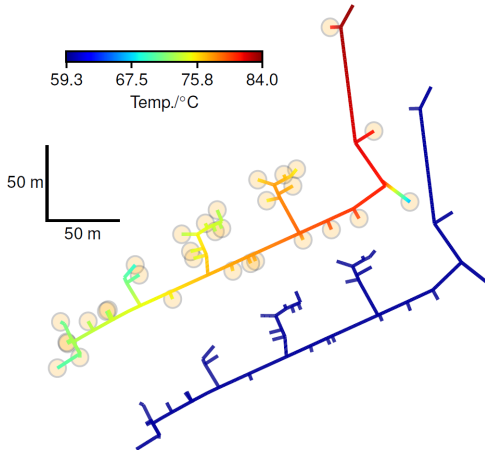
Coupling of heat, electric, waste incineration, and gas.



- ▷ TU Berlin
- ▷ Univ. Trier
- ▷ Fraunhofer ITWM Kaiserslautern
- ▷ Stadtwerke Ludwigshafen.



District Heating network



Simulated heat distribution in local district heating network:
Technische Werke Ludwigshafen. Entry forward flow
temperature 84C, backward flow temperature 60C.



Model: Simplified incompressible 1 D Euler equations.

$$0 = \frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x}(\rho v), \quad \text{Mass conservation,}$$

$$0 = \frac{\partial}{\partial t}(\rho v) + \frac{\partial}{\partial x}(p + \rho v^2) + \frac{\lambda}{2D} \rho v |v| + g \rho \frac{\partial}{\partial x} h, \quad \text{Momentum balance}$$

$$0 = \frac{\partial}{\partial t} \left(\rho \left(\frac{1}{2} v^2 + e \right) \right) + \frac{\partial}{\partial x} (e v) + \frac{k_w}{D} (T - T_w) \quad \text{Energy balance}$$

together with incompressibility condition for water. **Terms for pressure energy and dissipation work have been ignored.**

- ▷ velocity v , density ρ , k_w heat transfer coefficient,
- ▷ temperature T , wall temperature T_w , g gravitational force,
- ▷ λ friction coefficient, e internal energy, pressure p ,
- ▷ h height of pipe, D diameter of pipe.
- ▷ S.-A. Hauschild, N. Marheineke, V. Mehrmann, J. Mohring, A. Moses Badlyan, M. Rein, and M. Schmidt, Port-Hamiltonian modeling of district heating networks, DAE Forum, 333-355, Springer Verlag, 2020.
- ▷ R. Krug, V. Mehrmann, and M. Schmidt, Nonlinear Optimization of District Heating Networks, Optimization and Engineering, Vol. 22, 783-819, 2021.
- ▷ H. Dänschel, V. M., M. Roland, and M. Schmidt, *Adaptive Nonlinear Optimization of District Heating Networks Based on Model and Discretization Catalogs*, In preparation, next week, 2022.



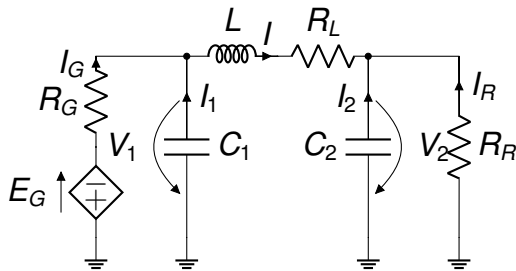
- ▶ Collaborative Research Center CRC910. Stability exponents for *delay differential-algebraic equations in power networks*.
- ▶ DFG priority Programme 1984, Hybrid and multimodal energy systems. Project: Computational Strategies for Distributed Stability Control in Next-Generation Hybrid Energy Systems with Kai Strunz, EE, TU Berlin
- ▶ A. González-Zumba, P. Fernández-de-Córdoba, J. Cortés, and V. M., Stability Assessment of Stochastic Differential-Algebraic Systems via Lyapunov Exponents with an Application to Power Systems. *Mathematics*, Vol. 8, 1393, 2020. doi:10.3390/math8091393
- ▶ D. Hinsén, A Port-Hamiltonian Approach for the Modeling of Power Networks including the Telegraph Equations, Master Thesis, TU Berlin, 2020.
- ▶ V. M. and R. Morandin, Structure-preserving discretization for port-Hamiltonian descriptor systems. *Proceedings of the 58th IEEE Conference on Decision and Control (CDC)*, 9.-12.12.19, Nice, 2019. <https://arxiv.org/abs/1903.10451>.
- ▶ V. M., R. Morandin, S. Olmi, and E. Schöll, Qualitative stability and synchronicity analysis of power network models in port-Hamiltonian form, *Chaos: An Interdisciplinary Journal of Nonlinear Science*, 28, 101-102, 2018.

Toy example: power network

Let $R_G, R_L, R_R > 0$ be resistances, $L > 0$ inductor, $C_1, C_2 > 0$ capacitors and E_G controlled voltage source.

Basic model of a DC generator (E_G, R_G), connected to load (R_R) via transmission line.

$$\begin{aligned}L\dot{I} &= -R_L I + V_2 - V_1, \\C_1\dot{V}_1 &= I - I_G, \quad C_2\dot{V}_2 = -I - I_R, \\0 &= -R_G I_G + V_1 + E_G, \quad 0 = -R_R I_R + V_2.\end{aligned}$$





- ▶ Gas transport CRC TRR 154
- ▶ Werner von Siemens Center Berlin. Electricity generation, gas turbine repair, additive manufacturing of turbine blades.
- ▶ Reactive flow control, new gas Turbine CRC 1029, until 2021.
- ▶ Poro-elastic networks.
- ▶ Multibody dynamics.
- ▶ Survey: V. M. and B. Unger, *Control of port-Hamiltonian differential-algebraic systems and applications*, <http://arxiv.org/abs/2201.06590>, 2022. Acta Numerica submitted.



A Modelling Wish list !

- ▶ Want representations so that **coupling of models works across different scales and physical domains.**
- ▶ Want a representation that is close to the real physics for **open and closed systems.**
- ▶ Model class should have **nice algebraic, geometric, and analytical properties.**
- ▶ Models should be easy to analyze mathematically (**existence, uniqueness, robustness, stability, uncertainty, errors etc**).
- ▶ Invariance under local coordinate transformations (in space and time). Ideally **local normal form.**
- ▶ Model class should allow for easy (**space-time**) **discretization and model reduction.**
- ▶ Class should be good for simulation, control and optimization,

Is there such a Jack of all trades, Eierlegende-Woll-Milch-Sau?



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Energy based network modeling

- ▶ Use **energy/power** as common quantity of different physical systems connected as network via energy transfer.
- ▶ Split components into **energy storage, energy dissipation components, control inputs and outputs, as well as interconnections** and combine via a **Dirac structure**.
- ▶ Allow every submodel to be a **model hierarchy** of fine or course, continuous or discretized, full or reduced models.
- ▶ **A system theoretic way** to realize this are **(dissipative) port-Hamiltonian systems**.
 - ▶ P. C. Breedveld. *Modeling and Simulation of Dynamic Systems using Bond Graphs*, pages 128–173. EOLSS Publishers Co. Ltd./UNESCO, Oxford, UK, 2008.
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Classical nonlinear port-Hamiltonian (pH) ODE/PDE systems

$$\begin{aligned}\dot{x} &= (J(x, t) - R(x, t)) \nabla_x \mathcal{H}(x) + (B(x, t) - P(x, t))u(t), \\ y(t) &= (B(x, t) + P(x, t))^T \nabla_x \mathcal{H}(x) + (S(x, t) - N(x, t))u(t),\end{aligned}$$

- ▷ x is the state, u input, y output.
- ▷ $\mathcal{H}(x)$ is the **Hamiltonian**: it describes the distribution of internal energy among the energy storage elements;
- ▷ $J = -J^T$ describes the **energy flux** among energy storage elements within the system;
- ▷ $R = R^T \geq 0$ describes **energy dissipation/loss** in the system;
- ▷ $B \pm P$: **ports** where energy enters and exits the system;
- ▷ $S - N$, $S = S^T$, $N = -N^T$, direct **feed-through** input to output.
- ▷ In the **infinite dimensional case** J, R, B, P, S, N are operators that map into appropriate function spaces.



Why should this be a good approach?

- ▶ PH systems generalize *Hamiltonian/gradient flow systems*.
- ▶ *Conservation of energy* replaced by *dissipation inequality*

$$\mathcal{H}(x(t_1)) - \mathcal{H}(x(t_0)) \leq \int_{t_0}^{t_1} y(t)^T u(t) dt,$$

- ▶ PH systems are closed under *power-conserving interconnection*. Modularized network based modeling.
- ▶ *Stability and passivity* analysis easy.
- ▶ PH structure allows to preserve physical properties in *Galerkin projection, model reduction*.
- ▶ Physical properties encoded in *algebraic structure* of coefficients and in *geometric structure* associated with flow.
- ▶ Systems are *easily extendable*: Open/closed systems.



Can we add algebraic constraints, like Kirchhoff's laws, position constraints, further conservation laws?

Define port-Hamiltonian partial-differential-algebraic equations (PDAEs) to fulfill as **many points on the wishlist as possible**.

- ▶ C. Beattie, V. M., H. Xu, and H. Zwart, *Linear port-Hamiltonian descriptor systems*. Math. Control Signals and Systems, 30:17, 2018.
- ▶ C. Beattie, V. M., and P. Van Dooren, *Robust port-Hamiltonian representations of passive systems*. Automatica, 100, 182–186, 2019.
- ▶ V. M. and R. Morandin, Structure-preserving discretization for port-Hamiltonian descriptor systems. *Proceedings of the 58th IEEE Conference on Decision and Control (CDC)*, 9.-12.12.19, Nice, 2019. <https://arxiv.org/abs/1903.10451>
- ▶ A. J. van der Schaft, Port-Hamiltonian differential-algebraic systems. In *Surveys in Differential-Algebraic Equations I*, 173-226. Springer-Verlag, 2013.
- ▶ A. van der Schaft and B. Maschke, Generalized Port-Hamiltonian DAE Systems, Systems Control Letters 121, 31-37, 2018.
- ▶ Survey: V. M. and B. Unger, *Control of port-Hamiltonian differential-algebraic systems and applications*, <http://arxiv.org/abs/2201.06590>, 2022. Acta Numerica submitted.



Definition (M./Morandin 2019)

Let $\mathcal{X} \subseteq \mathbb{R}^m$ (state space), $\mathbb{I} \subseteq \mathbb{R}$ time interval, and $\mathcal{S} = \mathbb{I} \times \mathcal{X}$. Consider

$$\begin{aligned} E(t, x)\dot{x} + r(t, x) &= (J(t, x) - R(t, x))e(t, x) + (B(t, x) - P(t, x))u, \\ y &= (B(t, x) + P(t, x))^T e(t, x) + (S(t, x) - N(t, x))u, \end{aligned}$$

Hamiltonian $\mathcal{H} \in C^1(\mathcal{S}, \mathbb{R})$, where $E \in C(\mathcal{S}, \mathbb{R}^{\ell, n})$, $J, R \in C(\mathcal{S}, \mathbb{R}^{n, n})$, $B, P \in C(\mathcal{S}, \mathbb{R}^{\ell, m})$, $S = S^T$, $N = -N^T \in C(\mathcal{S}, \mathbb{R}^{m, m})$ and $e, r \in C(\mathcal{S}, \mathbb{R}^{\ell})$. The system is called *port-Hamiltonian DAE* if

$$\Gamma(t, x) = -\Gamma^T = \begin{bmatrix} J & B \\ -B^T & N \end{bmatrix}, \quad W(t, x) = W^T = \begin{bmatrix} R & P \\ P^T & S \end{bmatrix} \geq 0,$$

$$\frac{\partial \mathcal{H}}{\partial x}(t, x) = E^T(t, x)e(t, x), \quad \frac{\partial \mathcal{H}}{\partial t}(t, x) = e^T(t, x)r(t, x).$$

Definition extends to weak solutions and infinite dimension:



Theorem (M./Morandin 2019)

Consider a pHDAE . Then the power balance equation (PBE)

$$\frac{d}{dt} \mathcal{H}(t, x(t)) = - \begin{bmatrix} e \\ u \end{bmatrix}^T W \begin{bmatrix} e \\ u \end{bmatrix} + y^T u$$

holds along any solution x , for any input u . In particular, the dissipation inequality

$$\mathcal{H}(t_2, x(t_2)) - \mathcal{H}(t_1, x(t_1)) \leq \int_{t_1}^{t_2} y(\tau)^T u(\tau) d\tau$$

holds.



Theorem (M./Morandin 2019)

Consider a pHDAE and another state space $\tilde{\mathcal{X}} \subseteq \mathbb{R}^{\tilde{n}}$, let $\tilde{\mathcal{S}} := \mathbb{I} \times \tilde{\mathcal{X}}$, let $x = \varphi(t, \tilde{x}) \in \mathcal{C}^1(\tilde{\mathcal{S}}, \mathcal{X})$ local diffeomorphism (w.r.t. \tilde{x}) and $U \in \mathcal{C}(\tilde{\mathcal{S}}, \mathbb{R}^{\ell, \ell})$ pointwise invertible. Consider

$$\begin{aligned}\tilde{E}\dot{\tilde{x}} + \tilde{r} &= (\tilde{J} - \tilde{R})\tilde{e} + (\tilde{B} - \tilde{P})u, \\ y &= (\tilde{B} + \tilde{P})^T \tilde{e} + (S - N)u,\end{aligned}$$

with $\tilde{E} = U^T(E \circ \varphi)\partial_{\tilde{x}}\varphi$, $\tilde{J} = U^T(J \circ \varphi)U$, $\tilde{R} = U^T(R \circ \varphi)U$, $\tilde{B} = U^T(B \circ \varphi)$, $\tilde{P} = U^T(P \circ \varphi)$, $\tilde{z} = U^{-1}(z \circ \varphi)$ and $\tilde{r} = U^T(r \circ \varphi + (E \circ \varphi)\partial_t\varphi)$, where $(F \circ \varphi)(t, \tilde{x}) = F(t, \varphi(t, \tilde{x}))$ for any $F \in \mathcal{C}(S, \cdot)$, and let $\tilde{\mathcal{H}}(t, \tilde{x}) := (\mathcal{H} \circ \varphi)(t, \tilde{x})$. Then this is again pHDAE with Hamiltonian $\tilde{\mathcal{H}}$, and to any solution (\tilde{x}, u, y) there corresponds a solution (x, u, y) of the original pHDAE with $x(t) = \varphi(t, \tilde{x}(t))$. Furthermore, if $\varphi(t, \cdot)$ is global diffeomorphism $t \in \mathbb{I}$, then the two systems are equivalent.



Any pHDAE can be made autonomous without destroying the structure by setting

$$\begin{bmatrix} E & r \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \dot{x} \\ \dot{t} \end{bmatrix} = \begin{bmatrix} J - R & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} z \\ 0 \end{bmatrix} + \begin{bmatrix} B - P & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ 1 \end{bmatrix},$$
$$\begin{bmatrix} y \\ 0 \end{bmatrix} = \begin{bmatrix} B + P & 0 \\ 0 & 1 \end{bmatrix}^T \begin{bmatrix} z \\ 0 \end{bmatrix} + \begin{bmatrix} S - N & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} u \\ 1 \end{bmatrix}.$$



Theorem

Consider two autonomous pHDAEs of the form

$$\begin{aligned} E_i \dot{x}_i + r_i &= (J_i - R_i)z_i + (B_i - P_i)u_i, \\ y_i &= (B_i + P_i)^T z_i + (S_i - N_i)u_i, \end{aligned}$$

with Hamiltonians \mathcal{H}_i , for $i = 1, 2$, and assume that aggregated input $u = (u_1, u_2)$ and output $y = (y_1, y_2)$ satisfy $Mu + Ny = 0$ for some $M, N \in \mathbb{R}^{k,m}$. Then interconncted system is pHDAE with Hamiltonian $\mathcal{H} = \mathcal{H}_1 + \mathcal{H}_2$.



Definition (E.g. Van der Schaft/Jeltsema 2014)

Let \mathcal{F} be a linear space and $\mathcal{E} := \mathcal{F}^*$ its dual space. Let $\langle\langle \cdot, \cdot \rangle\rangle$ be the bilinear form on $\mathcal{F} \times \mathcal{E}$ defined as

$$\langle\langle (f_1, e_1), (f_2, e_2) \rangle\rangle := \langle e_1 | f_2 \rangle + \langle e_2 | f_1 \rangle,$$

where $\langle \cdot | \cdot \rangle$ denotes the duality pairing. A *Dirac structure* on $\mathcal{F} \times \mathcal{E}$ is then a linear subspace $\mathcal{D} \subseteq \mathcal{F} \times \mathcal{E}$, such that $\mathcal{D} = \mathcal{D}^\perp$.

In finite dimensions, one only needs to prove $\dim \mathcal{D} = \dim \mathcal{F}$ and $\langle e | f \rangle = 0$ for all $(f, e) \in \mathcal{D}$.

If $(f, e) \in \mathcal{D}$, then f and e are called *flow and effort*, respectively.



Differential geometric viewpoint: A **modulated Dirac structure** over \mathcal{X} is a subbundle of $T\mathcal{X} \oplus T^*\mathcal{X}$, the **Whitney sum** between the tangent and cotangent bundles of \mathcal{X} .

Definition (M./Morandin 2019)

Consider a state space \mathcal{X} and a vector bundle \mathcal{V} over \mathcal{X} with fibers \mathcal{V}_x . A *Dirac structure* over \mathcal{V} is a subbundle $\mathcal{D} \subseteq \mathcal{V} \oplus \mathcal{V}^*$ such that, for all $x \in \mathcal{X}$, $\mathcal{D}_x \subseteq \mathcal{V}_x \times \mathcal{V}_x^*$ is a linear Dirac structure.



Theorem (M./Morandin 2019)

Given autonomous pHDAE, define flow fiber $\mathcal{V}_x = \mathcal{F}_x^s \times \mathcal{F}_x^p \times \mathcal{F}_x^d$ for all $x \in \mathcal{X}$, where $\mathcal{F}_x^s := E(x)T_x\mathcal{X} \subseteq \mathbb{R}^\ell$ is the storage flow, $\mathcal{F}_x^p := \mathbb{R}^m$ is the port flow, and $\mathcal{F}_x^d := \mathbb{R}^{\ell+m}$ is the dissipation flow fiber. Partition $f = (f_s, f_p, f_d) \in \mathcal{V}$ and $e = (e_s, e_p, e_d) \in \mathcal{V}^*$. Then the subbundle $\mathcal{D} \subseteq \mathcal{V} \oplus \mathcal{V}^*$ with

$$\mathcal{D}_x = \left\{ (f, e) \in \mathcal{V}_x \times \mathcal{V}_x^* \mid f + \begin{bmatrix} \Gamma(x) & I_{\ell+m} \\ -I_{\ell+m} & 0 \end{bmatrix} e = 0 \right\}$$

is a Dirac structure over \mathcal{V} . Furthermore, the system $f_s - E(x)\dot{x}$, $e_s = z(x)$, $f_p = y$, $e_p = u$, $e_d = -W(x)f_d$, $(f, e) \in \mathcal{D}_x$ is equivalent to the original pHDAE, and $\langle e \mid f \rangle = 0$ is the PBE.

Geometric interpretation via mix of Lagrangian and Dirac structure [Van der Schaft/Maschke 2018, 2020](#).



- ▶ Structure preserving time-discretization, so that time discrete system has dissipation inequality and power balance equation and has a discretized Dirac structure. (Gauss Legendre collocation, or partitioned implicit Runge-Kutta methods)
- ▶ Use non-uniqueness to obtain robust representations under perturbations.
- ▶ Local and global normal forms.
- ▶ Perturbation analysis, distance to instability, non-passivity, non-regularity.
- ▶ Different port-Hamiltonian PDE formulations, including terms that have been dropped due to model simplifications.



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Different approaches.

- ▷ Operator pH DAE modeling.
- ▷ Differential geometric: Gradient flow, GENERIC.
- ▷ Formal Dirac structures.
- ▷ Structured PDE systems with input and outputs.

References:

- ▷ R. Altmann und P. Schulze A port-Hamiltonian formulation of the Navier-Stokes equations for reactive flows Systems Control Lett., Vol. 100, 2017, pp. 51-55.
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Abstract port-Hamiltonian PDE formulation

$$\begin{aligned}\frac{dz}{dt} &= (\mathcal{J}(z) - \mathcal{R}(z)) \frac{\delta \mathcal{E}(z)}{\delta z} + \mathcal{B}(z)u(z) \quad \text{in } \mathcal{D}_z^*, \\ y(z) &= \mathcal{B}^*(z) \frac{\delta \mathcal{E}(z)}{\delta z} \quad \text{in } \mathcal{D}_u^*,\end{aligned}$$

- ▶ $\mathcal{Z} = \{z \in \mathcal{D}_z \mid \rho \geq \delta, \delta > 0 \text{ a.e.}\} \subset \mathcal{D}_z = W^{1,3}((0, \ell); \mathbb{R}^3)$.
- ▶ For $z \in \mathcal{Z}$, $\mathcal{J}(z)[\cdot], \mathcal{R}(z)[\cdot] : \mathcal{D}_z \rightarrow \mathcal{D}_z^*$ are linear continuous, $\mathcal{J}(z)$ is skew-adjoint, $\mathcal{R}(z)$ is self-adjoint semi-elliptic.
- ▶ The input is given by $u(z) \in \mathcal{D}_u = L^q(\{0, \ell\})$ with linear continuous $\mathcal{B}(z)[\cdot] : \mathcal{D}_u \rightarrow \mathcal{D}_z^*$, $\mathcal{D}_u^* = L^p(\{0, \ell\})$, $1/q + 1/p = 1$.
- ▶ The system theoretic output is denoted by $y(z)$.
- ▶ $\mathcal{E}(z)$ is the relative energy.

- ▶ A. M. Badlyan, B. Maschke, C. Beattie, and V. M., Open physical systems: from GENERIC to port-Hamiltonian systems, Proceedings of MTNS, 2018.
- ▶ A. Moses Badlyan and C. Zimmer. Operator-GENERIC formulation of thermodynamics of irreversible processes. Preprint TU Berlin 2018.



Port-Hamiltonian formulation of incompressible Euler including pressure energy and dissipation work, and entropy balance.

$$0 = \rho \frac{\partial v}{\partial x}, \quad \text{mass conservation}$$

$$0 = \frac{\partial}{\partial t}(\rho v) + v^2 \frac{\partial \rho}{\partial x} + \frac{\partial p}{\partial x} + \frac{\lambda}{2D} \rho v |v| + g \rho \frac{\partial h}{\partial x}, \quad \text{momentum balance}$$

$$0 = \frac{\partial e}{\partial t} + v \frac{\partial e}{\partial x} - \frac{\lambda}{2D} \rho v^2 |v| + \frac{4k_w}{D} (T - T_w), \quad \text{energy balance}$$

$$0 = \frac{\partial s}{\partial t} + v \frac{\partial s}{\partial x} - \frac{\lambda \rho}{2D T} v^2 |v| + \frac{4k_w}{D} \frac{(T - T_w)}{T}, \quad \text{entropy balance}$$

We have to add node conditions, mixing conditions etc.



PH PDE weak form, GENERIC

Variables $z = (\rho, M, \mathbf{e})^T$, $M = \rho v$, energy.

$$\mathcal{E}(z) = \mathcal{H}(z) - T_w \mathcal{S}(z) := \int_0^\ell \left(\frac{|M|^2}{2\rho} + \mathbf{e} + \rho gh \right) dx - T_w \int_0^\ell s(\rho, \mathbf{e}) dx.$$

where T_w is assumed to be constant. Introduce **ballistic free energy** $H(\rho, \mathbf{e}) = \mathbf{e} - T_w s(\rho, \mathbf{e})$, then functional \mathcal{E} and its variational derivatives become

$$\begin{aligned} \mathcal{E}(z) &= \int_0^\ell \left(\frac{|M|^2}{2\rho} + H(\rho, \mathbf{e}) + \rho gh \right) dx \\ \frac{\delta \mathcal{E}(z)}{\delta z} &= \left(\frac{\delta \mathcal{E}(z)}{\delta \rho}, \frac{\delta \mathcal{E}(z)}{\delta M}, \frac{\delta \mathcal{E}(z)}{\delta \mathbf{e}} \right)^T \\ &= \left(\left(-\frac{|M|^2}{2\rho^2} + \frac{\partial H}{\partial \rho} + gh \right), \frac{M}{\rho}, \frac{\partial H}{\partial \mathbf{e}} \right)^T. \end{aligned}$$



The operators are assembled with respect to the (block-) structure of the state z .

Let $\varphi, \psi \in \mathcal{D}_z$ be block-structured $\varphi = (\varphi_\rho, \varphi_M, \varphi_e)^T$. Then

$$\mathcal{J}(z) = \begin{bmatrix} 0 & \mathcal{J}_{\rho,M}(z) & 0 \\ \mathcal{J}_{M,\rho}(z) & \mathcal{J}_{M,M}(z) & \mathcal{J}_{M,e}(z) \\ 0 & \mathcal{J}_{e,M}(z) & 0 \end{bmatrix},$$

is associated with the bilinear form

$$\begin{aligned} \langle \varphi, \mathcal{J}(z)\psi \rangle &= \langle \varphi_\rho, \mathcal{J}_{\rho,M}(z)\psi_M \rangle + \langle \varphi_M, \mathcal{J}_{M,\rho}(z)\psi_\rho \rangle + \langle \varphi_M, \mathcal{J}_{M,M}(z)\psi_M \rangle \\ &\quad + \langle \varphi_M, \mathcal{J}_{M,e}(z)\psi_e \rangle + \langle \varphi_e, \mathcal{J}_{e,M}(z)\psi_M \rangle \end{aligned}$$

$$\langle \varphi_\rho, \mathcal{J}_{\rho,M}(z)\psi_M \rangle = \int_0^\ell \rho(\psi_M \partial_x) \varphi_\rho \, dx,$$

$$\langle \varphi_M, \mathcal{J}_{M,M}(z)\psi_M \rangle = \int_0^\ell M((\psi_M \partial_x) \varphi_M - (\varphi_M \partial_x) \psi_M) \, dx,$$

$$\langle \varphi_e, \mathcal{J}_{e,M}(z)\psi_M \rangle = \int_0^\ell e(\psi_M \partial_x) \varphi_e + (\psi_M \partial_x)(\varphi_e \rho) \, dx$$



The self-adjoint semi-elliptic operator $\mathcal{R}(z)$ has two parts corresponding to the friction in the pipe $\mathcal{R}^\lambda(z)$ and the temperature loss through the pipe walls $\mathcal{R}^{k_w}(z)$.

$$\mathcal{R}(z) = \mathcal{R}^\lambda(z) + \mathcal{R}^{k_w}(z) = \begin{bmatrix} 0 & 0 & 0 \\ 0 & \mathcal{R}_{M,M}^\lambda(z) & \mathcal{R}_{M,e}^\lambda(z) \\ 0 & \mathcal{R}_{e,M}^\lambda(z) & \mathcal{R}_{e,e}^\lambda(z) + \mathcal{R}_{e,e}^{k_w}(z) \end{bmatrix},$$

associated with the bilinear form

$$\begin{aligned} \langle \varphi, \mathcal{R}(z)\psi \rangle &= \langle \varphi_M, \mathcal{R}_{M,M}^\lambda(z)\psi_M \rangle + \langle \varphi_M, \mathcal{R}_{M,e}^\lambda(z)\psi_e \rangle \\ &+ \langle \varphi_e, \mathcal{R}_{e,M}^\lambda(z)\psi_M \rangle + \langle \varphi_e, (\mathcal{R}_{e,e}^\lambda(z) + \mathcal{R}_{e,e}^{k_w}(z))\psi_e \rangle \end{aligned}$$

$$\langle \varphi_M, \mathcal{R}_{M,M}^\lambda(z)\psi_M \rangle = \int_0^\ell \varphi_M \left(\frac{\lambda}{2d} \frac{T}{\vartheta} \rho |v| \right) \psi_M dx,$$

$$\langle \varphi_M, \mathcal{R}_{M,e}^\lambda(z)\psi_e \rangle = \int_0^\ell -\varphi_M \left(\frac{\lambda}{2d} \frac{T}{\vartheta} \rho |v| v \right) \psi_e dx,$$

$$\langle \varphi_e, (\mathcal{R}_{e,e}^\lambda(z) + \mathcal{R}_{e,e}^{k_w}(z))\psi_e \rangle = \int_0^\ell \varphi_e \left(\frac{\lambda}{2d} \frac{T}{\vartheta} \rho |v| v^2 + \frac{4k_w}{d} T \right) \psi_e dx.$$



The input is given as $u(z) \in \mathcal{D}_u$ by $u(z) = [M/\rho]_0^\ell$ and the port operator $\mathcal{B}(z)[\cdot] : \mathcal{D}_u \rightarrow \mathcal{D}_z^*$ via the pairing

$$\langle \varphi, \mathcal{B}(z)u(z) \rangle = - [(\varphi_\rho \rho + \varphi_M M + \varphi_e(\mathbf{e} + \mathbf{p})) u(z)]_0^\ell,$$

coming from the boundary terms via integration by parts.

With the adjoint operator $\mathcal{B}^*(z)[\cdot] : \mathcal{D}_z \rightarrow \mathcal{D}_u^*$, i.e.,

$\langle \varphi, \mathcal{B}(z)u(z) \rangle = \langle \mathcal{B}^*(z)\varphi, u(z) \rangle$, the system theoretic output is

$$y(z) = \mathcal{B}^*(z) \frac{\delta \mathcal{E}(z)}{\partial z} = - \left[\frac{|M|^2}{2\rho} + \mathbf{p} + H(\rho, \mathbf{e}) + \rho g h \right]_0^\ell.$$



The operators \mathcal{J} and \mathcal{R} satisfy **non-interacting conditions**

$$\mathcal{J}(z) \frac{\delta \mathcal{S}(z)}{\delta z} = 0, \quad \mathcal{R}^\lambda(z) \frac{\delta \mathcal{H}(z)}{\delta z} = 0,$$

so flows of Hamiltonian and gradient system do not overlap.

- ▶ In the pH framework the choice of the state variables in the interplay with the energy functional is **crucial for encoding the physical properties**.
- ▶ Asymptotic simplifications (as limit to incompressibility) not straightforward, since they may change underlying structure.
- ▶ In compressible case (gas network) the ideal gas law implies

$$s(\rho, e) = \frac{R}{2} \rho \ln \left(c_p \frac{e^3}{\rho^5} \right), \quad T(\rho, e) = \frac{2}{3R} \frac{e}{\rho}, \quad p(\rho, e) = \frac{2}{3} e,$$

with specific gas constant R and heat capacity c_p .



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Galerkin projection/model reduction

Replace state x in (operator) equation

$$E\dot{x} = (J - R)x + (B - P)u, \quad y = (B + P)^T x + Du, \quad D = S - N$$

by Galerkin projection $x \approx V_r x_r$ with finite (small) dimensional space V_r : **projected (reduced) system**

$$E_r \dot{x}_r = (J_r - R_r)x_r + (B_r - P_r)u, \quad y_r = (B_r + P_r)^T x_r + Du,$$

$E_r = V_r^T E V_r$, $J_r = V_r^T J V_r$, $R_r = V_r^T R V_r$, $B_r = V_r^T B$, $P_r = V_r^T P$.
If V_r forms orthonormal basis, then resulting system is again a pHDAE and all properties are preserved.

- ▶ S. Chaturantabut, C. Beattie and S. Gugercin, Structure-Preserving Model Reduction for Nonlinear Port-Hamiltonian Systems, SIAM J. Scientific Computing, 2016,
- ▶ C. Beattie, S. Gugercin and V. M., *Structure-preserving Interpolatory Model Reduction for Port-Hamiltonian Differential-Algebraic Systems*. <http://arxiv.org/abs/1910.05674>. Festschrift for 70th birthday of A. Antoulas, 2021.
- ▶ H. Egger, T. Kugler, B. Liljegren-Sailer, N. Marheineke, and V. M., *On structure preserving model reduction for damped wave propagation in transport networks*, SIAM Journal Scientific Computing, Vol. 40, A331–A365, 2018.
- ▶ S. Hauschild, N. Marheineke and V. M., Model reduction techniques for linear constant coefficient port-Hamiltonian differential-algebraic systems, <https://arxiv.org/abs/1901.10242>, Control and Cybernetics, 2020.



Compute reduced order pHDAE

$$\begin{aligned} E_r \dot{x}_r &= (J_r - R_r) x_r + (B_r - P_r) u, \quad x_r(t_0) = 0, \\ y_r &= (B_r + P_r)^T x_r + D_r u, \end{aligned}$$

such that $y_r(t)$ is good approximation to $y(t)$ over a wide range of $u(t)$. Let $H(s) = (B^T + P^T)(sE + R - J)^{-1}(B - P) + D$. Given right and left interpolation points $\{\sigma_1, \dots, \sigma_r\}$, $\{\mu_1, \dots, \mu_r\}$ with right and left tangent directions $\{k_1, \dots, k_r\}$, $\{\ell_1, \ell_2, \dots, \ell_r\}$, construct

$H_r(s) = (B_r^T + P_r^T)(sE_r + R_r - J_r)^{-1}(B_r - P_r) + D_r$ such that

$$H(\sigma_i)k_i = H_r(\sigma_i)k_i \quad \text{and} \quad \ell_i^T H(\mu_i) = \ell_i^T H_r(\mu_i), \quad \text{for } i = 1, 2, \dots, r.$$

Interpolation conditions enforced via Petrov-Galerkin projection with

$$\begin{aligned} V_r &= \left[(\sigma_1 E + R - J)^{-1}(B - P)k_1, \quad \dots \quad (\sigma_r E + R - J)^{-1}(B - P)k_r \right], \\ Z_r &= \left[(\sigma_1 E + R - J)^{-T}(B + P)\ell_1, \quad \dots \quad (\sigma_r E + R - J)^{-T}(B + P)\ell_r \right], \\ E_r &= Z_r^T E V_r, \quad J_r = Z_r^T J V_r, \quad R_r = Z_r^T R V_r, \quad B_r = Z_r^T B, \quad P_r = Z_r^T P, \quad D_r = D. \end{aligned}$$



- ▶ The reduced quantities may **no longer have the symmetry structure**. Can be resolved by using a Galerkin projection, i.e., with $Z_r = V_r$. But then only half of the interpolation conditions hold.
- ▶ **The polynomial parts of transfer functions may not match, leading to unbounded errors.**
- ▶ We need to identify the constraints via a condensed form or directly from the structure of the equations.
- ▶ Beattie, Gugercin and V. M., *Structure-preserving Interpolatory Model Reduction for Port-Hamiltonian Differential-Algebraic Systems*. *Festschrift to honor the 70th birthday of A. Antoulas*, 2022. <http://arxiv.org/abs/1910.05674>.



Suppose we know the algebraic constraints explicitly and have the semi-explicit index one pHDAE structure

$$\begin{bmatrix} E_{11} & 0 \\ 0 & 0 \end{bmatrix} \dot{x}(t) = \begin{bmatrix} J_{11} - R_{11} & J_{12} - R_{12} \\ -J_{12}^T - R_{12}^T & J_{22} - R_{22} \end{bmatrix} x(t) + \begin{bmatrix} B_1 - P_1 \\ B_2 - P_2 \end{bmatrix} u(t)$$

$$y(t) = [B_1^T + P_1^T \quad B_2^T + P_2^T] x(t) + Du(t).$$

where $E_{11} = E_{11}^T > 0$ and $J_{22} - R_{22}$ is nonsingular.



Theorem (Beattie, Gugercin, V.M. 2019)

Consider semi-explicit index 1 pHDAE, interpol. points $\{\sigma_1, \sigma_2, \dots, \sigma_r\}$ and tangent dir. $\{k_1, k_2, \dots, k_r\}$. Construct basis $V_r = \begin{bmatrix} V_{r,1}^T & V_{r,2}^T \end{bmatrix}^T$ as $V_r = [(\sigma_1 E + R - J)^{-1}(B - P)k_1, \dots, (\sigma_r E + R - J)^{-1}(B - P)k_r]$ and set $K_r = [k_1 \ \dots \ k_r]$, $D_r = D - (B_2^T + P_2^T)(J_{22} - R_{22})^{-1}(B_2 - P_2)$. Then transfer function $H_r(s)$ of reduced model

$$E_r \dot{x}_r(t) = (J_r - R_r)x_r(t) + (B_r - P_r)u(t), \quad y_r(t) = (B_r + P_r)x_r(t) + D_r u(t)$$

with $E_r = V_{r,1}^T E_{11} V_{r,1}$, $J_r - R_r = V_r^T (J - R) V_r + K_r^T (D_r - D) K_r$, $(B_r + P_r)^T = (B + P) V_r + (B_2^T + P_2^T)(J_{22} - R_{22})^{-1}(B_2 - P_2) K_r$, matches polyn. part of $H(s)$ and tangentially interpolates it. The reduced system is again a pHDAE if the reduced passivity matrix

$w_r = \begin{bmatrix} R_r & P_r \\ P_r^T & S_r \end{bmatrix}$ is positive semidefinite.



Numerical example

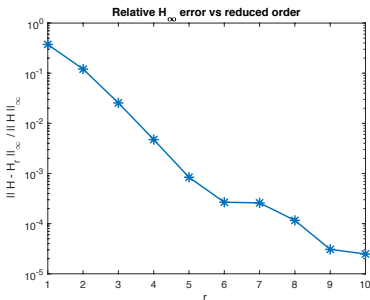
pHDAE formulation of incompressible Oseen equations,

$$\begin{aligned} \partial_t \mathbf{v} &= -(\mathbf{a} \cdot \nabla) \mathbf{v} + \mu \Delta \mathbf{v} - \nabla p + \mathbf{f} & \text{in } \Omega \times (0, T], & \quad \mathbf{v} = \mathbf{0}, \\ \mathbf{0} &= -\operatorname{div} \mathbf{v}, & \text{in } \Omega \times (0, T], & \quad \mathbf{v} = \mathbf{v}^0, \end{aligned}$$

with velocity \mathbf{v} and pressure p , $\mu > 0$ is the viscosity, and

$\Omega = (0, 1)^2$. $\mathbf{f} = b(x)u(t)$ is an externally body force.

FD/FEM discretization gives pHDAE with $n = 7399$, $n_v = 4900$, and $n_p = 2499$.





- ▶ Generate projection spaces via POD or shifted POD approaches in transport dominant case.
- ▶ Combine with Empirical Interpolations Methods. (D)EIM.
- ▶ Incorporate information from physical system.
- ▶ Much to do, in particular if the system has many transports.
 - ▶ Barrault, Maxime, et al. *An empirical interpolation method: application to efficient reduced-basis discretization of partial differential equations*. Comptes Rendus Mathematique 2004.
 - ▶ Chaturantabut, Beattie, and Gugercin. Structure-preserving model reduction for nonlinear port-Hamiltonian systems. SIAM Journal Scientific Computing, 2016.
 - ▶ Chaturantabut, Sorensen, Nonlinear Model Reduction via Discrete Empirical Interpolation, SIAM Journal Scientific Computing, 2010.
 - ▶ Reiss, Schulze, Sesterhenn, and V.M. *The shifted proper orthogonal decomposition: A mode decomposition for multiple transport phenomena*. SIAM Journal Scientific Computing, 2018.



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- ▷ Most classical ODE/DAE methods do not preserve the energy or dissipation inequality.
- ▷ We need classes of integrators that do.
- ▷ Want integrators that lead to discrete-time pH systems.
- ▷ Preservation of constraints.

Idea: Use Dirac Structure

- ▷ E. Celedoni and E.H. Høiseith, *Energy-Preserving and Passivity-Consistent Numerical Discretization of Port-Hamiltonian Systems*, *arXiv:1706.08621v1*
- ▷ Kotyczka, Lefèvre, *Discrete-Time Port-Hamiltonian Systems Based on Gauss-Legendre Collocation*, *IFAC-PapersOnLine* 51, no. 3 (2018): 125–30.



Collocation methods for pHDAEs

Given input $u : \mathbb{I} \times \mathcal{X} \rightarrow \mathbb{R}^m$, interval $\mathbb{I} = [t_0, t_f]$ of length $h = t_f - t_0$, consistent initial condition x_0 at time t_0 , we approximate the solution $x(t)$ with polynomial $\tilde{x}(t)$ of degree s , so that with $\tilde{x}(t_0) = x_0$, \tilde{x} satisfies the pHDAE in s collocation points $t_i = t_0 + h\gamma_i$ with $\gamma_i \in [0, 1]$ for $i = 1, \dots, s$.

Let l_i denote the i -th Lagrange interp. poly. with nodes $\gamma_1, \dots, \gamma_s$, i.e., $l_i(\tau) := \prod_{\substack{j=1 \\ j \neq i}}^s \frac{\tau - \gamma_j}{\gamma_i - \gamma_j}$. Then for certain $\dot{x}_i = \dot{\tilde{x}}(t_i)$,

$$\dot{\tilde{x}}(t_0 + \tau h) = \sum_{i=1}^s \dot{x}_i l_i(\tau), \quad \tilde{x}(t_0 + \tau h) = x_0 + h \sum_{j=1}^s \dot{x}_j \int_0^\tau l_j(\sigma) d\sigma,$$

$$x_i := \tilde{x}(t_i) = x_0 + h \sum_{j=1}^s \alpha_{ij} \dot{x}_j, \quad x_f := \tilde{x}(t_f) = x_0 + h \sum_{j=1}^s \beta_j \dot{x}_j,$$

$\alpha_{ij} := \int_0^{\gamma_i} l_j(\sigma) d\sigma$, $\beta_j := \int_0^1 l_j(\sigma) d\sigma$, $i, j = 1 \dots s$ associated Runge-Kutta coefficients.



Dirac structure of discretization

Let \mathcal{D}_x be the Dirac structure associated to the pHDAE and define Dirac structure of discretization as $\{\mathcal{D}_{x_i} : i = 1, \dots, s\}$, i.e.,

$$\mathcal{D}_{x_i} = \left\{ (f^i, e^i) \in \mathcal{V}_{x_i} \times \mathcal{V}_{x_i}^* \mid f^i + \begin{bmatrix} \Gamma(t_i, x_i) & l_{\ell+m} \\ -l_{\ell+m} & 0 \end{bmatrix} e^i = 0 \right\},$$

with $f^i = (f_s^i, y_i, f_r^i)$ and $e^i = (e_s^i, u_i, e_r^i)$.

Taking $(f^i, e^i) \in \mathcal{D}_{x_i}$, together with

$$x_f = x_0 + h \sum_{i=1}^s \beta_i \dot{x}_i, \quad x_i = x_0 + h \sum_{j=1}^s \alpha_{ij} \dot{x}_j,$$

$$f_s^i = -E(x_i) \dot{x}_i, \quad e_s^i = e(x_i), \quad e_r^i = -W(x_i) f_r^i, \quad u_i = u(x_i),$$

the system is equivalent to applying the collocation method and computing the *discrete input* and *output* u_i, y_i , for $i = 1, \dots, s$.



Define the *collocation flows, efforts, input and output* as

$$\begin{aligned}\tilde{f}_s(t_0 + h\tau) &= \sum_{i=1}^s f_s^i l_i(\tau), & \tilde{e}_s(t_0 + h\tau) &= \sum_{i=1}^s e_s^i l_i(\tau), \\ \tilde{f}_r(t_0 + h\tau) &= \sum_{i=1}^s f_r^i l_i(\tau), & \tilde{e}_r(t_0 + h\tau) &= \sum_{i=1}^s e_r^i l_i(\tau), \\ \tilde{y}(t_0 + h\tau) &= \sum_{i=1}^s y_i l_i(\tau), & \tilde{u}(t_0 + h\tau) &= \sum_{i=1}^s u_i l_i(\tau).\end{aligned}$$

Then, $(\tilde{f}_s, \tilde{y}, \tilde{f}_r, \tilde{e}_s, \tilde{u}, \tilde{e}_r) \in \mathcal{D}_{\tilde{x}}$ in all collocation points t_j .

- ▶ E. Hairer, C. Lubich and G. Wanner, *Structure-Preserving Algorithms for Ordinary Differential Equations*, Springer, Berlin, Germany: Springer-Verlag, 2006.
- ▶ Kotyczka, Lefèvre, *Discrete-Time Port-Hamiltonian Systems Based on Gauss-Legendre Collocation*, IFAC-PapersOnLine 51, no. 3 (2018): 125–30.



With $H(t) := \mathcal{H}(\tilde{x}(t))$ we get $H(t) - H(t_0) = \int_{t_0}^t \dot{H}(s) ds$ and in collocation points, the Power Balance Equation is satisfied.

$$\begin{aligned}\dot{H}(t_i) &= \nabla \mathcal{H}(x_i)^T \dot{x}_i = \mathbf{e}(x_i)^T E(x_i) \dot{x}_i = \\ &= -\langle \mathbf{e}_s^i | \mathbf{f}_s^i \rangle = \langle \mathbf{e}_r^i | \mathbf{f}_r^i \rangle + \langle \mathbf{y}_i | \mathbf{u}_i \rangle, i = 1, \dots, s\end{aligned}$$

Apply collocation method of order p

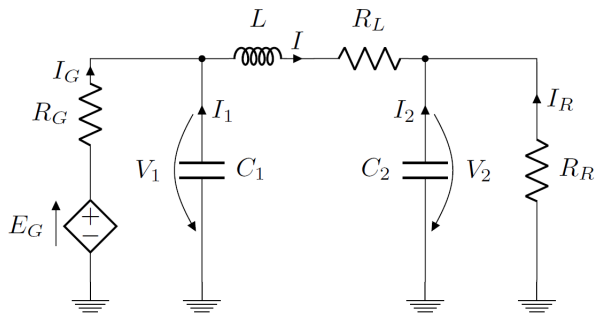
$$H(t_f) - H(t_0) = \int_{t_0}^{t_f} \left(\langle \tilde{\mathbf{e}}_r(s) | \tilde{\mathbf{f}}_r(s) \rangle + \langle \tilde{\mathbf{y}}(s) | \tilde{\mathbf{u}}(s) \rangle \right) ds + \mathcal{O}(h^{p+1}).$$

If the Runge-Kutta conditions satisfy $\beta_j \geq 0$ for $j = 1, \dots, s$ then $h \sum_{j=1}^s \langle \mathbf{e}_r^j | \mathbf{f}_r^j \rangle \leq 0$, thus discrete system stays dissipative.

Exact if Hamiltonian is quadratic.



Toy example of DC power network:



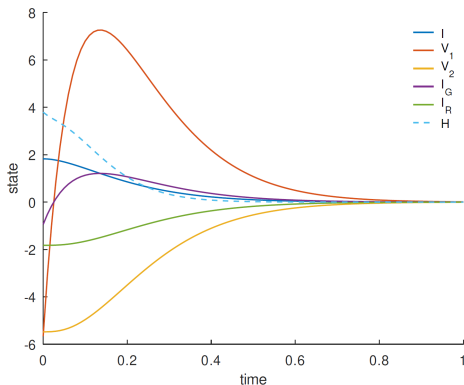
The energy of the system is stored in the inductor and in the two capacitors, giving the Hamiltonian

$$\mathcal{H}(I, V_1, V_2) = \frac{1}{2}LI^2 + \frac{1}{2}C_1V_1^2 + \frac{1}{2}C_2V_2^2.$$



Simulation Toy example

Choose $L = 2$, $C_1 = 0.01$, $C_2 = 0.02$, $R_L = 0.1$, $R_G = 6$, $R_R = 3$.



Implicit midpoint rule, (Gauss-Legendre) $s = 1$ stages, order $p = 2$ for ODEs and semi-explicit DAEs of index 1.

After discretization with implicit Gauss-Legendre, linear systems have positive (semi)-definite symmetric part.

For $(M + N)x = b$ with $M = M^T > 0$ $N = -N^T$ use symmetric part as preconditioner, and solve the equivalent system

$$(I - K)x = \hat{b}, \quad \text{where } K = M^{-1}N, \quad \hat{b} = M^{-1}b.$$

Then K is M -normal i.e., $M^{-1}K^T M = -K$. This is necessary and sufficient for K to admit an optimal 3-term recurrence for generating an M -orthogonal basis of the Krylov subspace $\mathcal{K}_k(K, v)$ for each k and initial vector v .

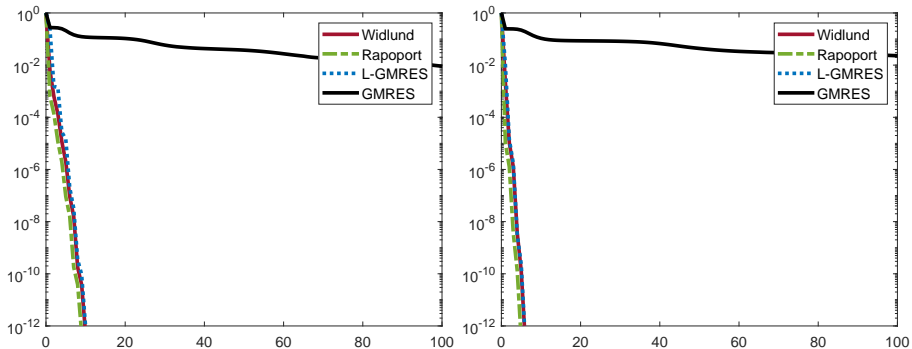
Oblique projection method with Galerkin projection property:

$$x_k \in \mathcal{K}_k(K, \hat{b}) \quad \text{s.t.} \quad r_k = b - (M + N)x_k \perp \mathcal{K}_k(K, \hat{b}).$$

- ▶ O. Widlund, A Lanczos method for a class of nonsymmetric systems of linear equations. *SIAM J. Numer. Anal.*, 15(4):801–812, 1978.
- ▶ M. Manguoğlu and V. Mehrmann, A two-level iterative scheme for general sparse linear systems based on approximate skew-symmetrizers. <http://arxiv.org/abs/2009.06954>, Electronic Transactions Numerical Analysis, Vol. 54, 370–391, 2021.
- ▶ C. Güdücü, J. Liesen, V. M., and D. Szyld, *On non-Hermitian positive (semi)definite linear algebraic systems arising from dissipative Hamiltonian DAEs*, <http://arxiv.org/abs/2111.05616>, 2021.



Numerical example, convergence



Stokes equation. Relative residual norms with $\tau = 0.001$ and $\tau = 0.0001$ (left and right).

Run times differences drastic if step-size is decreased.



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What about our wish list?

- ▶ Want representations so that **coupling of models works across different scales and physical domains.**
- ▶ Want a representation that is close to the real physics for **open and closed systems.**
- ▶ Model class should have **nice algebraic, geometric, and analytical properties.**
- ▶ Models should be easy to analyze mathematically (**existence, uniqueness, robustness, stability, uncertainty, errors etc**).
- ▶ Invariance under local coordinate transformations (in space and time). Ideally **local normal form.**
- ▶ Model class should allow for easy (**space-time**) **discretization and model reduction.**
- ▶ Class should be good for simulation, control and optimization, **pH DAE systems are ideal, almost all wishes are fulfilled.**



But there are many things to do

- ▶ Real time control, optimization.
- ▶ Other physical domains.
- ▶ Incorporate stochastics in models.
- ▶ Stability analysis.
- ▶ Function Spaces.
- ▶ Error estimates.
- ▶ Preconditioning.
- ▶ Data based realization.
- ▶ Uncertainty quantification.
- ▶ Software.
- ▶ Digital twins
- ▶ ...



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Details: <http://www.math.tu-berlin.de/?id=76888>