



Moment-based identification for wave energy systems

Fundamentals and application

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2. Frequency response fundamentals
3. Moment-based identification
4. Application case: OPT-device
5. Conclusions

Introduction and motivation

Motivation

The equations of motion of wave energy devices, in the LTI case, can be expressed in terms of **Cummins' equation**.

Integro-differential Volterra equation, of the **convolution** class.

This convolution term represents fluid memory effects associated with the dynamics of the **radiation forces**.

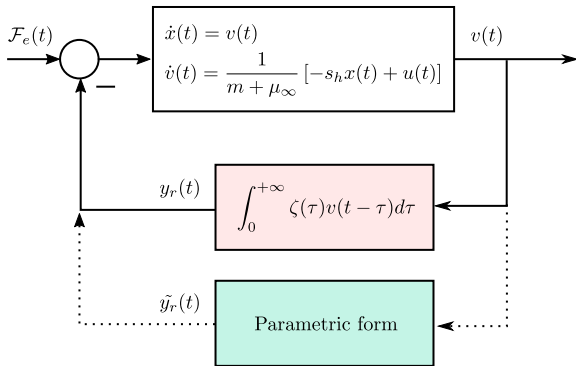
Inconvenient for simulation and analysis and design of control systems!

How can we solve this issue and find a more **tractable** system?

$$\dot{x}(t) = v(t),$$
$$\dot{v}(t) = \frac{1}{m + \mu_\infty} \left[-s_h x(t) - \underbrace{\int_0^{+\infty} \zeta(\tau) v(t - \tau) d\tau}_{y_r(t)} + \mathcal{F}_e(t) \right].$$

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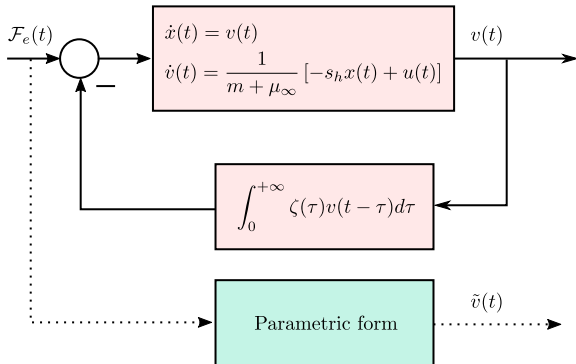
$$\|y_r(t) - \tilde{y}_r(t)\| < \epsilon$$

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$$\|v(t) - \tilde{v}(t)\| < \epsilon$$

These non-parametric models are usually obtained with BEM solvers

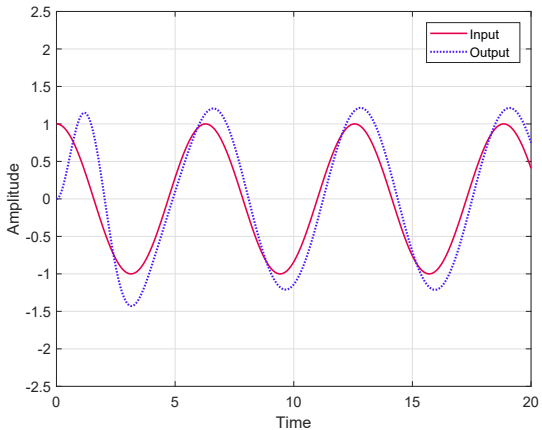
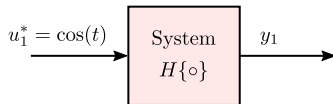
- WAMIT
- NEMOH

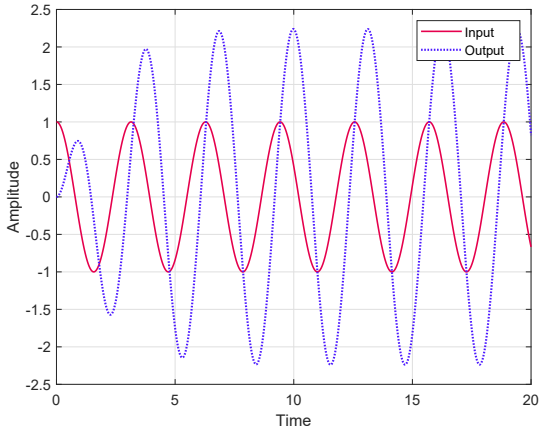
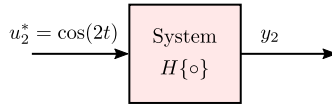
BEM solvers are computationally efficient, but they can only characterise the steady-state characteristics of the WEC.

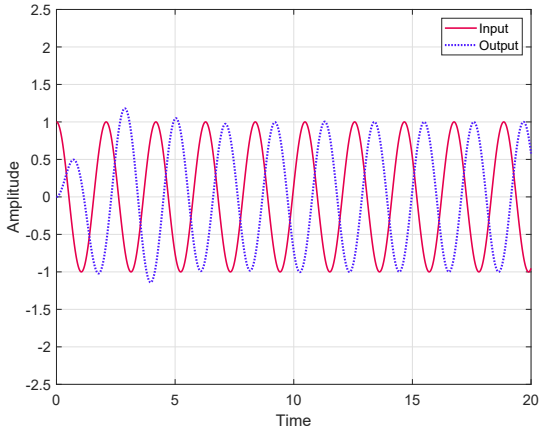
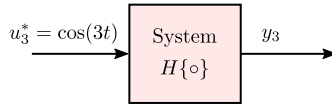
Non-parametric data to identify:

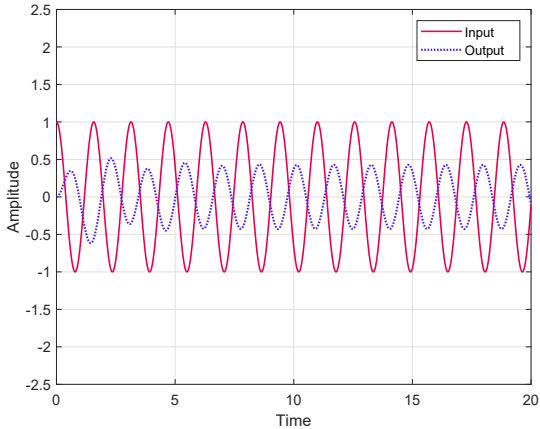
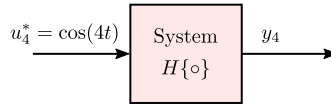
Finite number of points in **frequency-domain** → **frequency response!**

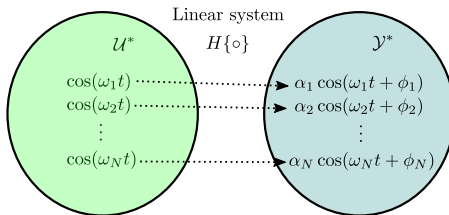
Frequency response fundamentals

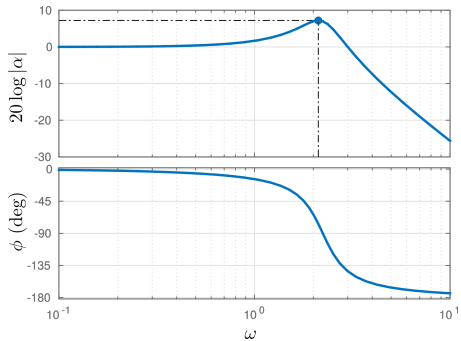
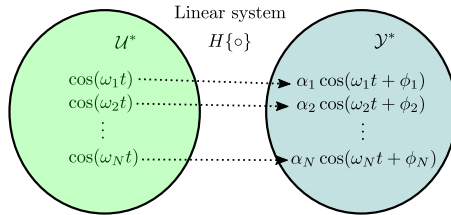












Moment-based identification

Fundamental questions:

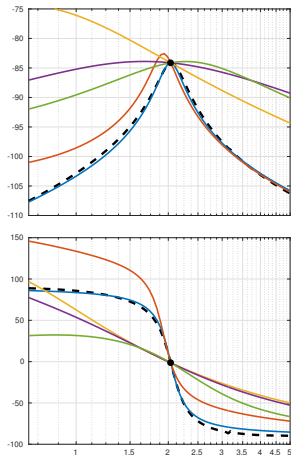
Given a **non-parametric** set of data in **frequency-domain**, how do we select a **family of parametric models** that can represent our data?

Once this family is decided, how do we choose the **best** model inside this set?

An ideal parametric modeling technique should:

- **Preserve** underlying physical characteristics.
- **Preserve** a precise steady-state description for key important frequencies.
- **Minimise** approximation error in a particular frequency range of interest.

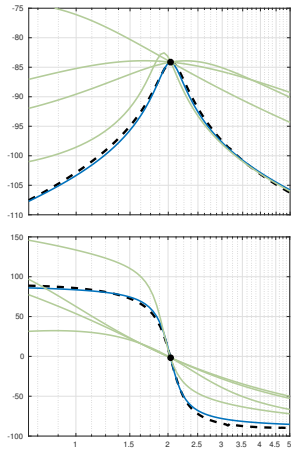
We propose a moment-based approach:



Moment-based family of models:

- LTI models of order ν that preserve an **exact** steady-state response for a **user-defined** set of frequencies $\mathcal{F} = \{\omega_p\}_{i=1}^p$.
- Model order $\nu = 2p$ (**twice** the number of frequencies to interpolate).
- Parameterised in state-space form $(A, B, C, 0)$.

We propose a moment-based approach:



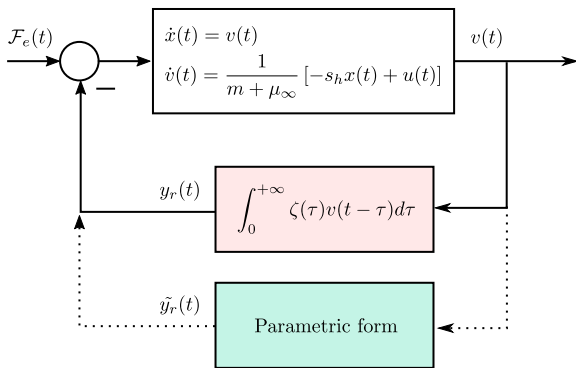
Moment-based optimal model:

- Preserve **internal stability**.
- **Interpolation** in the frequency set \mathcal{F} .
- **Optimal approximation** in Euclidean sense for a *user-defined frequency range*.

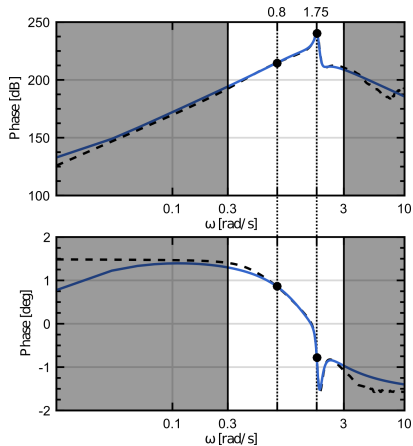
Application case: OPT-device

$$\dot{x}(t) = v(t),$$

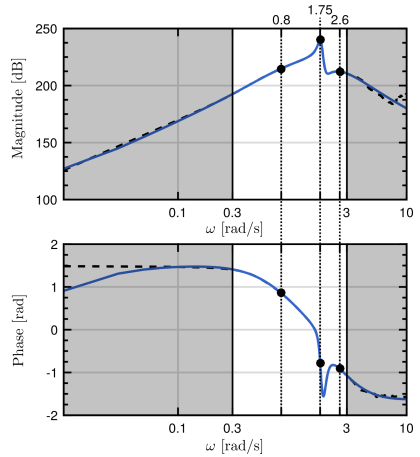
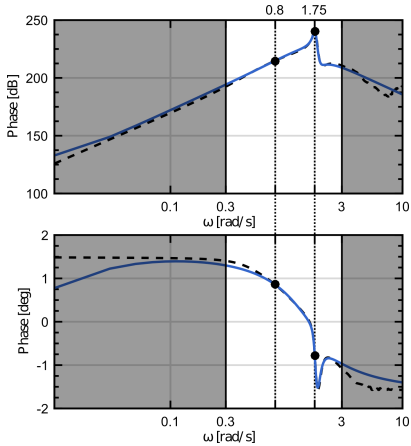
$$\dot{v}(t) = \frac{1}{m + \mu_\infty} \left[-s_h x(t) - \underbrace{\int_0^{+\infty} \zeta(\tau) v(t - \tau) d\tau}_{y_r(t)} + \mathcal{F}_e(t) \right].$$



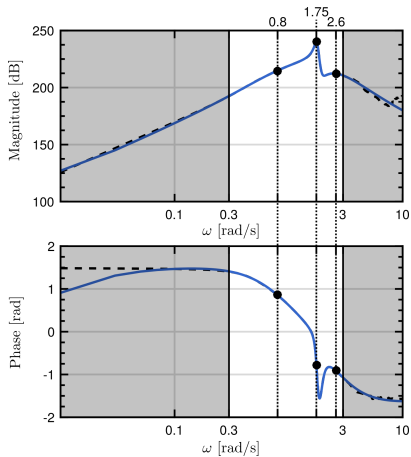
$$\|y_r(t) - \tilde{y}_r(t)\| < \epsilon$$



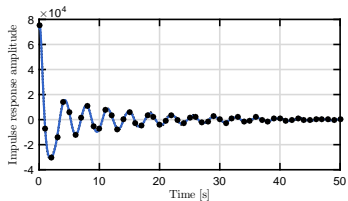
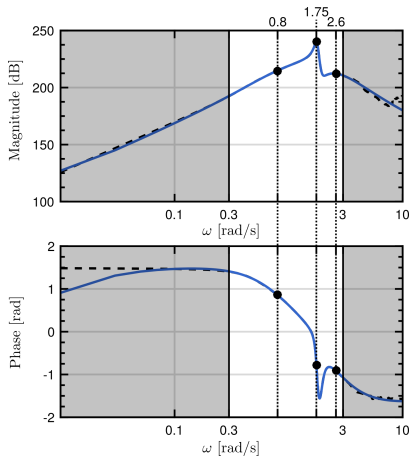
Target frequency-domain data in (dashed-black). Parametric approximation in (solid-blue).



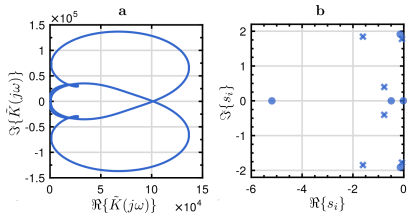
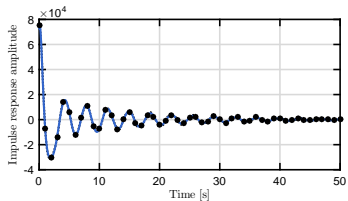
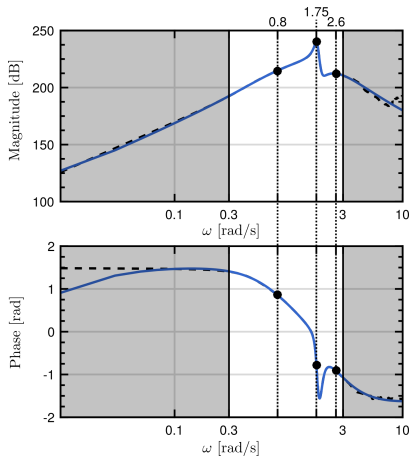
Target frequency-domain data in (dashed-black). Parametric approximation in (solid-blue).



Target non-parametric data in black. Parametric approximation in blue.



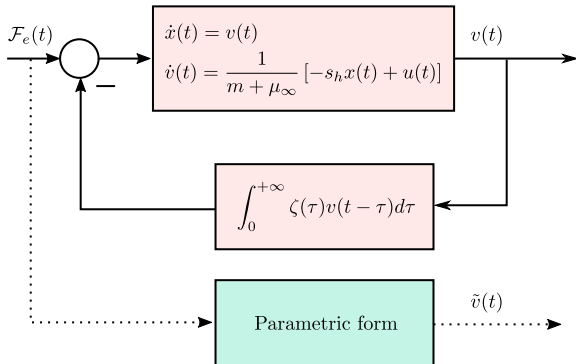
Target non-parametric data in black. Parametric approximation in blue.



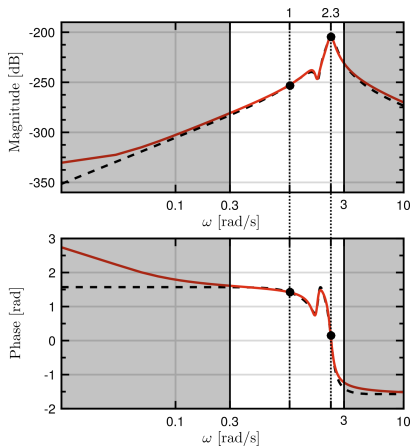
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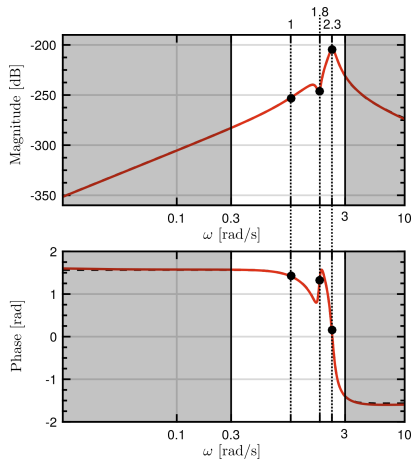
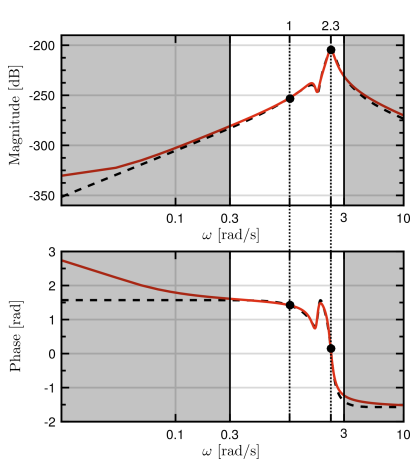
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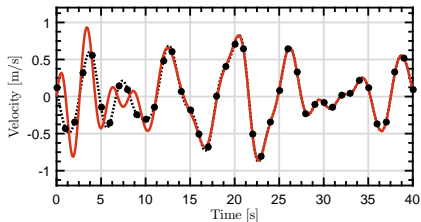
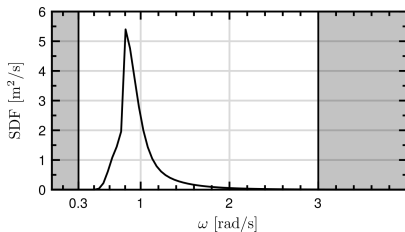
$$\|v(t) - \tilde{v}(t)\| < \epsilon$$



Target frequency-domain data in (dashed-black). Parametric approximation in (solid-red).

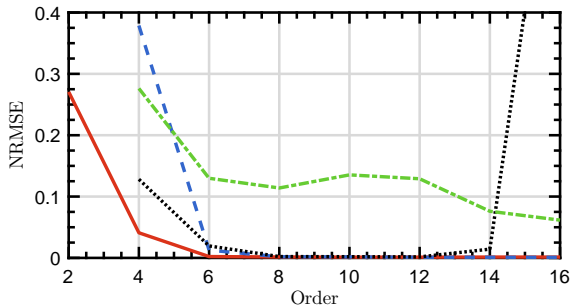


Target frequency-domain data in (dashed-black). Parametric approximation in (solid-red).



Target non-parametric data in black. Parametric approximation in red.

Force-to-velocity response comparison:



Solid-red Force-to-velocity based on moments.

Dashed-blue Cummins' equation + radiation force model based on moments.

Dotted-black Cummins' equation + radiation force model based on *invfreqs* function (NTNU).

Dashed-green Cummins' equation + radiation force model based on Prony's method.

Conclusions

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- We have developed a moment-matching-based identification algorithm that can be applied to obtain a parametric form of both the **force-to-motion** and the **radiation impulse response** dynamics from raw frequency-domain data.
- The strategy provides a parametric model that matches *exactly* the behaviour of the device at key frequencies *that can be selected by the user*.
- The strategy allows for the selection of a **frequency range** to perform the approximation.
- The obtained models inherently respect the physical properties of the device under analysis.

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Finite-Order Approximation by Moment-Matching toolbox → FOAMM

-  Nicolás Faedo, Yeraí Peña-Sanchez, and John V Ringwood. “Finite-order hydrodynamic model determination for wave energy applications using moment-matching”. In: *Ocean Engineering* 163 (2018), pp. 251–263.
-  Nicolás Faedo, Yeraí Peña-Sanchez, and John V Ringwood. “Moment-Matching-Based Identification of Wave Energy Converters: the ISWEC Device”. In: *IFAC-PapersOnLine* 51.29 (2018), pp. 189–194.
-  Nicolás Faedo, Yeraí Peña-Sanchez, and John V. Ringwood. “Parameterisation of Radiation Forces for a Multiple Degree-of-Freedom Wave Energy Converter Using Moment-Matching”. In: *Proceedings of the 2019 Automatic Control Conference, Philadelphia*. 2019.
-  Nicolás Faedo, Yeraí Peña-Sanchez, and John V Ringwood. “Passivity preserving moment-based finite-order hydrodynamic model identification for wave energy applications”. In: *Advances in Renewable Energies Offshore: Proceedings of the 3rd International Conference on Renewable Energies Offshore (RENEW 2018), Lisbon*. 2018, p. 351.



Yerai Peña-Sanchez, Nicolás Faedo, and John V. Ringwood.
“Hydrodynamic Model Fitting for Wave Energy Applications Using
Moment-Matching: A Case Study”. In: *The 28th International Ocean and
Polar Engineering Conference*. International Society of Offshore and Polar
Engineers. 2018.



Yerai Peña-Sanchez, Nicolás Faedo, and John V. Ringwood.
“Moment-Based Parametric Identification of Arrays of Wave Energy
Converters”. In: *Proceedings of the 2019 Automatic Control Conference,
Philadelphia*. 2019.

Thanks!

Any questions?



Centre for Ocean Energy Research