

Interpolated models for non-intrusive affinization of reduced basis methods

Eivind Fonn

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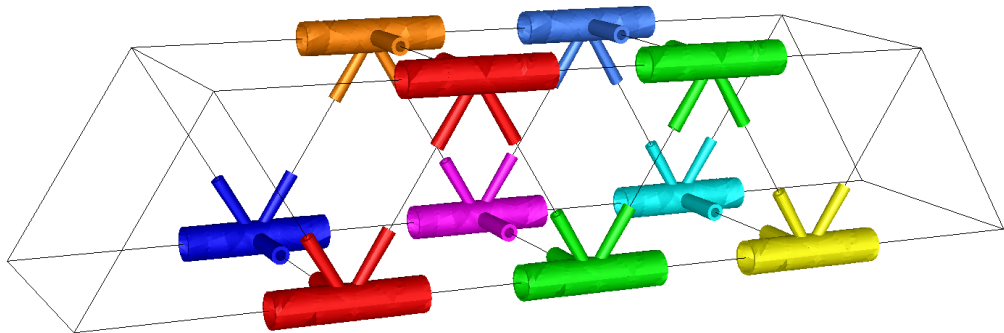
- The what: jacket joint components
- The plan (part 1): port condensation ROM technique
- The plan (part 2): thoughts about intrusiveness
- Methodology
- Results

Jacket joints

- Jacket structures used in marine operations are subject to rigorous compliance standards
- Much time is spent solving linear elasticity problems on such structures
- Beams are modeled cheaply using beam elements but the joints are not
- Want a joint ROM parametrized on material properties and geometry (plate thickness, radius, beam angles etc)

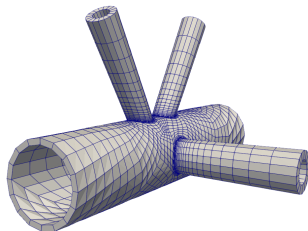


Example



Static condensation

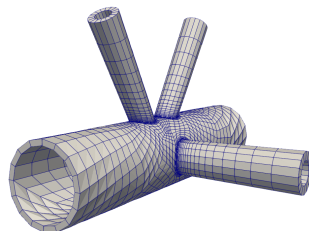
- Static condensation ‘compresses’ each component to a $6n \times 6n$ superelement that acts as coupling terms between otherwise disconnected beam elements.
- Works well if each component is *identical*.
- Still requires the solution of a nontrivial system.
- Not parametrized over geometry, material, etc.



Port condensation

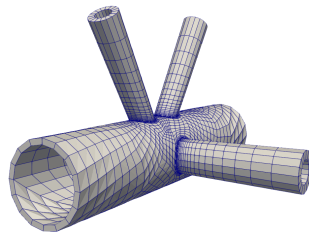
The technique follows in brief Eftang and Patera (2013).

- For each *port*, six ROMs are constructed - each evaluating the response on the structure from one degree of freedom on the boundary (three translations and three rotations)
- In addition, one ROM is constructed with *homogeneous* boundary conditions, evaluating the response from other sources, e.g. gravity

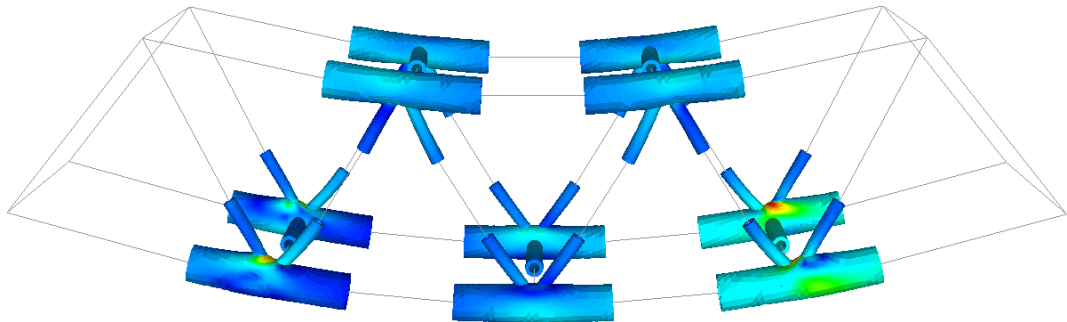


Port condensation

- In the online stage, each of the $6n$ ROMs are queried and the solutions form the internal basis functions for a $6n \times 6n$ superelement matrix
- The solution to the homogenous ROM enters the right hand side as a load term
- *Parametrized* and *faster* static condensation



Example



Intrusiveness

- Many decisions made in the following are motivated by non-intrusiveness
- RBMs are typically quite intrusive: to reduce a FOM requires expert knowledge of it
- We aim to deliver a generalized RBM software package (AROMA) that is as non-intrusive as possible
 - snapshots as fully-featured solution vectors
 - stiffness matrix and load vector in restricted sense (i.e. with fixed DoFs removed and lift applied) — available as debug output in most FOM software
 - knowledge of DoF classification (fixed/free) and lift — fairly mundane
- Aim is to view the FOM as a nearly opaque black box

Reduced basis methods

Substitute the solution of a large system

$$\mathbf{A} \mathbf{u} = \mathbf{f}$$

with a smaller one

$$\underbrace{\mathbf{V}^\top \mathbf{A} \mathbf{V}}_{\mathbf{A}^r} \underbrace{\mathbf{u}}_{\mathbf{u}^r} = \underbrace{\mathbf{V}^\top \mathbf{f}}_{\mathbf{f}^r}$$

Solution and assembly

- Extremely fast *solutions* in the online stage: $\mathbf{A}^r(\mu)$ frequently on the order of $\sim 10 \times 10$ or so.
- Assembly of $\mathbf{A}^r(\mu)$ remains an open issue. On the face of it,

$$\mathbf{A}^r(\mu) = \mathbf{V}^\top \mathbf{A}(\mu) \mathbf{V}$$

requires the construction of the full-order matrix $\mathbf{A}(\mu)$ in the online stage.

- Many solutions have been proposed, most of which relates to...

The assumption of affine parametric dependence

If $\mathbf{A}(\mu)$ takes the form

$$\mathbf{A}(\mu) = \sum_i \xi_i(\mu) \mathbf{A}_i$$

then $\mathbf{A}^r(\mu)$ takes the form

$$\mathbf{A}^r(\mu) = \mathbf{V}^\top \mathbf{A}(\mu) \mathbf{V} = \sum_i \xi_i(\mu) \mathbf{V}^\top \mathbf{A}_i \mathbf{V} = \sum_i \xi_i(\mu) \mathbf{A}_i^r$$

Each \mathbf{A}_i^r is small and easily storeable. Assuming the ξ_i are not pathological and the sum not too long, assembly in this form is competitive with time spent solving the system.

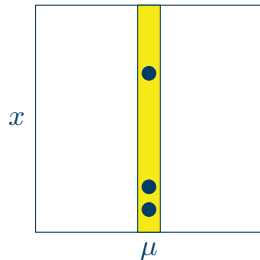
The assumption of affine parametric dependence

- Affinity tends to hold well with
 - explicit material parameters: elastic properties, viscosities, thermal conductivities
 - load and boundary data
- But not with geometric variability
 - even simple variation manifests in the integrands (for FEM) or fluxes (for FVM) in nontrivial ways (Fonn et. al., 2018)
- Approximate affine parametric dependence can be *forced* using various interpolation methods

Empirical Interpolation

The technique follows in brief Eftang and Patera (2013).

- EIM requires the online stage to evaluate a function $g(x, \mu)$ at certain points x in the physical domain, or a vector-valued function $g_i(\mu)$ at certain entries i .
- Here, g is the Finite Element variational form or similar for FVM
- In the black-box FOM context we operate in, such restricted evaluation of the integrand is not possible (too intrusive)
- Instead, we resort to Matrix Least Squares



Matrix Least Squares

Given reasonable guesses for $\xi_i(\mu)$, compute corresponding coefficient matrices \mathbf{A}_i by minimizing

$$\int_{\mu} \left\| \mathbf{A}(\mu) - \sum_i \xi_i(\mu) \mathbf{A}_i \right\|_2^2 d\mu.$$

In practice, this is implemented the same was as conventional L^2 -fit on a sampling over μ :

$$\begin{aligned} \mathbf{B}_{ji} &= \xi_i(\mu_j) \\ \mathbf{A}_i &= \sum_{jk} (\mathbf{B}^\top \mathbf{B})_{ik}^{-1} \mathbf{B}_{jk} \mathbf{A}(\mu_j) \end{aligned}$$

Matrix Least Squares

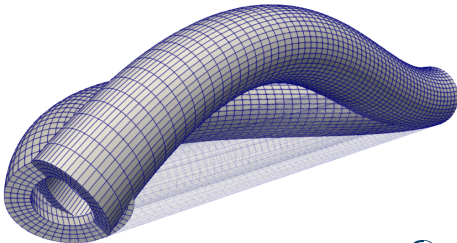
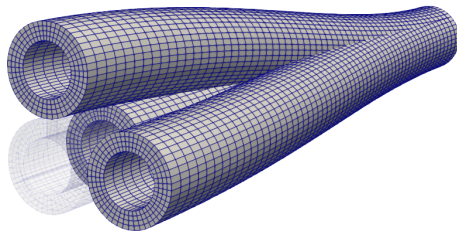
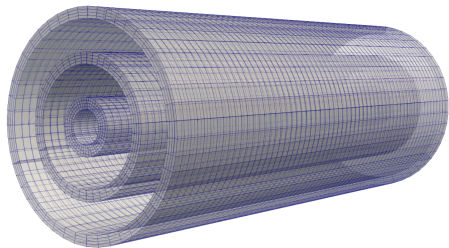
- In practice, sampling \mathbf{A} , \mathbf{f} is done at the same time, and in the same parameter values as the snapshots.
- If the sampling strategy is known in advance, the contribution from $\mathbf{A}(\mu_j)$ can be immediately applied and the matrix then discarded.
- Much leeway is admitted in the choice of ξ_i
 - known affinities may be exploited, e.g.

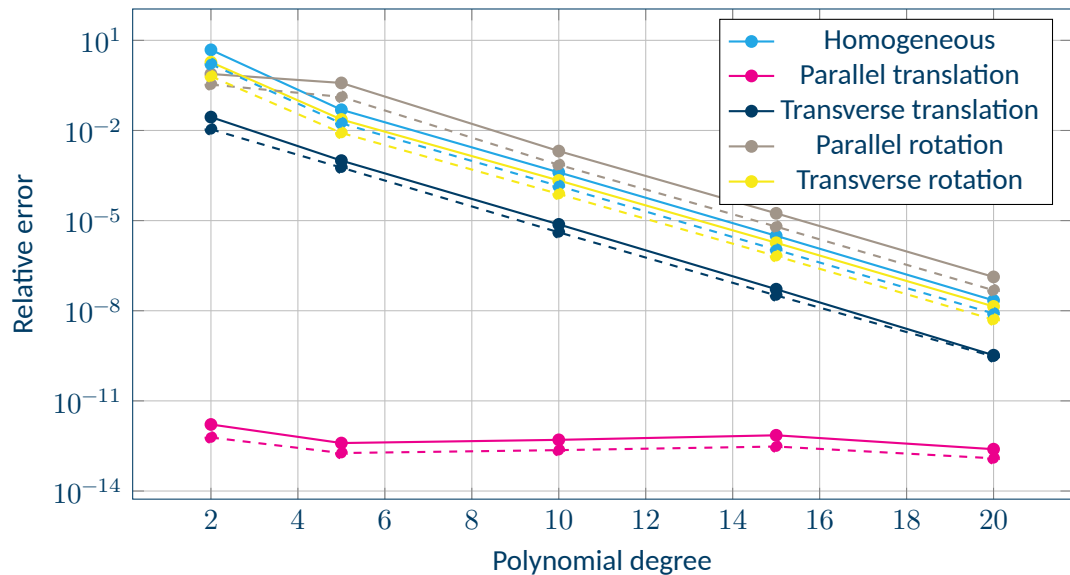
$$\xi_1(E, \nu) = \frac{E}{2(1 + \nu)}, \quad \xi_2(E, \nu) = \frac{E}{(1 - 2\nu)(1 + \nu)}$$

- orthogonal polynomial families as a last resort, $\xi_i(r) = P_i(r)$
- a combination, $\xi_i(E, \nu) = EP_i(\nu)$

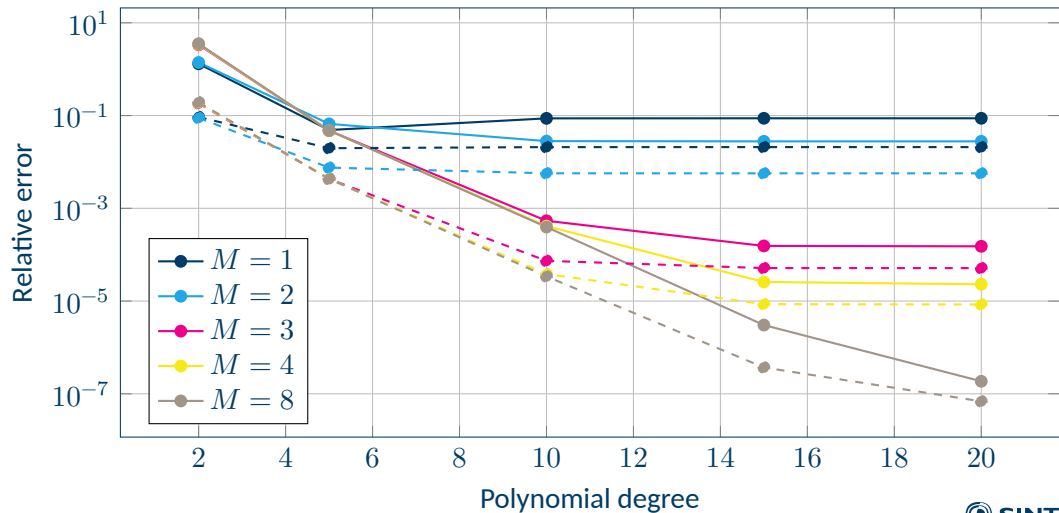
Example: two-port cylinder

Minimal example with variable radius

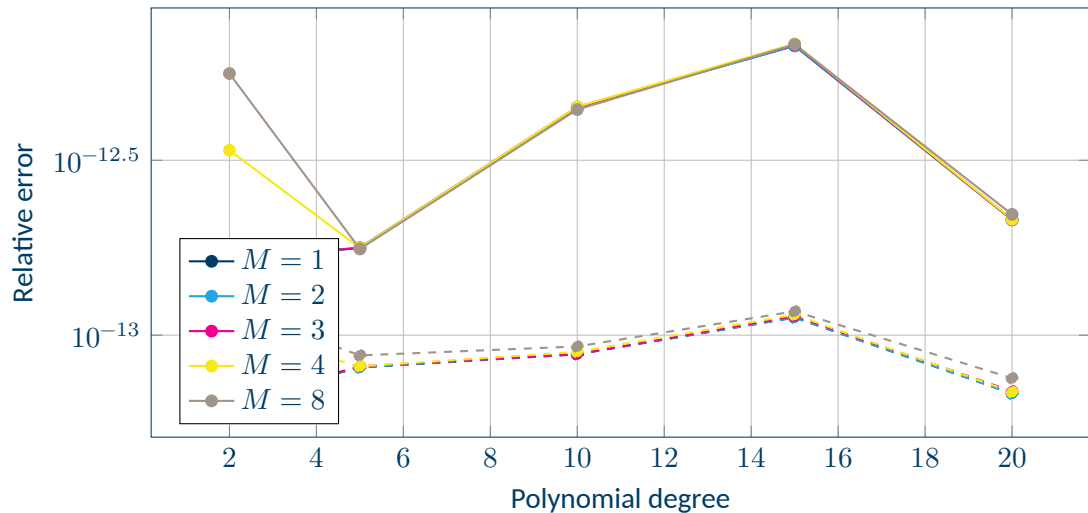




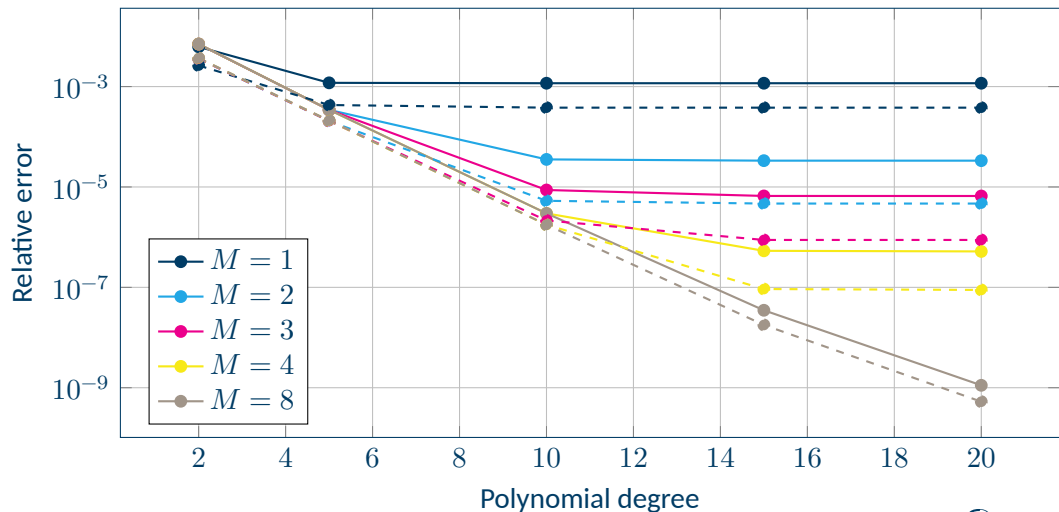
Homogenous case



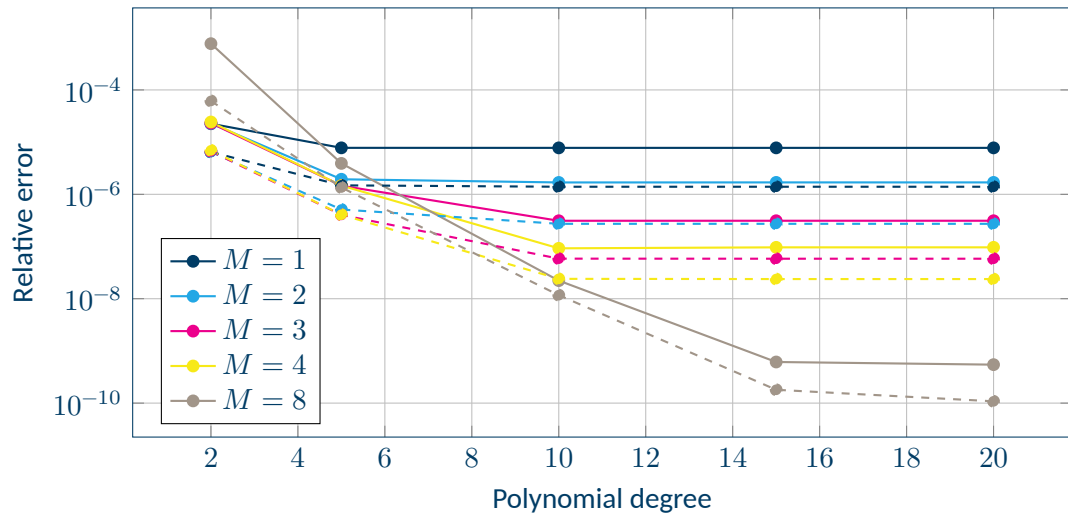
Parallel translation case



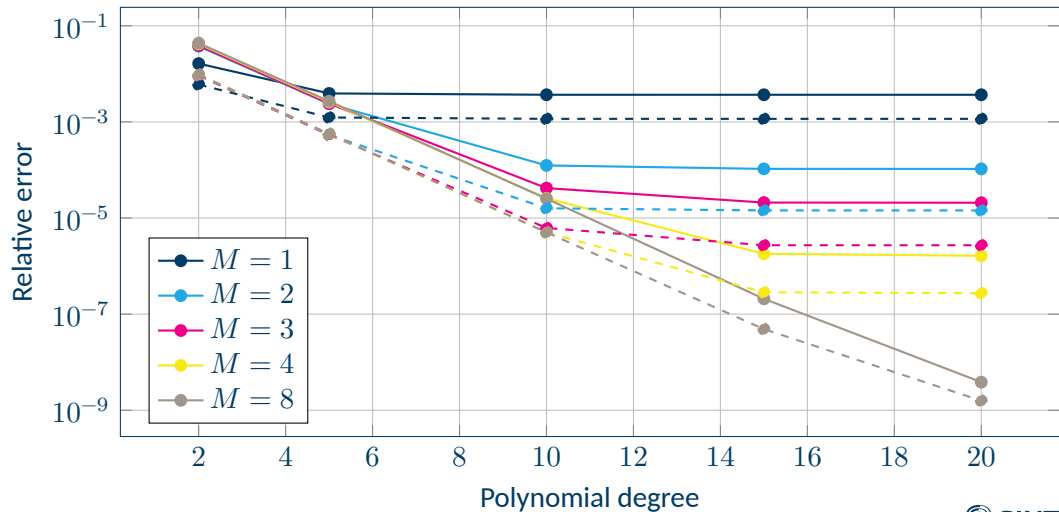
Transverse translation case



Parallel rotation case



Transverse rotation case





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