



# On Space-Time Approaches in Reduced Basis Methods

Time-Periodic Problems

Kristina Steih

Institute for Numerical Mathematics

**Joint work with:**

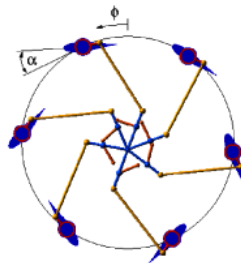
Karsten Urban, Ulm University

## Motivation

### Optimization of a ship propeller

- ▶ parameter studies
- ▶ long simulation times

⇒ Model reduction in parameter space



## Challenges for Reduced Basis Methods

- ▶ **Time-periodic problem**
- ▶ Parameter-dependent domain
- ▶ Time-dependent parameter function
- ▶ Nonlinearities
- ▶ ⋮

[CTU]

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## Mathematical Problem Formulation

### Parameter-dependent time-periodic problem

Let  $V \hookrightarrow H \hookrightarrow V'$  a Gelfand triple,  $\mu \in \mathcal{D} \subset \mathbb{R}^p$ ,  $\mathcal{A}(t; \mu) : V \rightarrow V'$ .

$$J(u(\mu)) = \int_0^T \ell(u(\mu); \mu) dt,$$

$$u_t + \mathcal{A}(t; \mu)u = g(t; \mu) \quad \text{on } \Omega,$$

$$u(0) = u(T) \quad \text{in } H,$$

$$u(t, x; \mu) = h_D(x; \mu) \text{ on } \Gamma_D, \quad \frac{\partial}{\partial n} u(t, x; \mu) = h_N(x; \mu) \text{ on } \Gamma_N.$$

**Example:**

$$\mathbf{V} = H_{(0)}^1(\Omega) \hookrightarrow \mathbf{H} := L_2(\Omega) \hookrightarrow \mathbf{V}' := H^{-1}(\Omega).$$

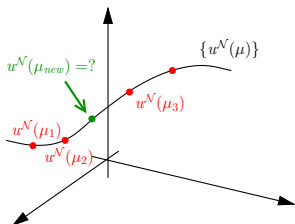
## Introduction to RBM

### Space-Time in RBM

### Why Space-Time ?

### Adaptive Space-Time Methods

## The idea behind RBM



### Assumption:

- ▶ High-dimensional discretization space  $\mathcal{X}^{\mathcal{N}}$  (FEM, FV, wavelets, ...)
- ▶  $w^{\mathcal{N}}(\mu)$  smooth in  $\mu$
- ▶ Low-dimensional solution space  $\mathcal{M}^{\mathcal{N}} = \{w^{\mathcal{N}}(\mu) : \mu \in \mathcal{D}\}$

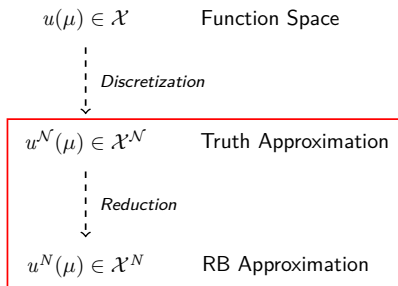
### Procedure: Approximation of $\mathcal{M}^{\mathcal{N}}$

1. *Basis construction:*  $N$  solutions  $w^{\mathcal{N}}(\mu_1), \dots, w^{\mathcal{N}}(\mu_N)$ ,  $N \ll \mathcal{N}$ .
2. *Projection:* Compute  $w(\mu_{new})$  on  $\mathcal{X}^N := \text{span}\{w^{\mathcal{N}}(\mu_1), \dots, w^{\mathcal{N}}(\mu_N)\}$ .

### Application:

- ▶ Many-query context (e.g. optimization)
- ▶ Real-time applications

## Reliable and efficient error bounds



$$e_{\mathcal{N}}(\mu) = w^{\mathcal{N}}(\mu) - u_{\mathcal{N}}(\mu)$$

### Error Bound

$$\|w^{\mathcal{N}}(\mu) - u_{\mathcal{N}}(\mu)\| \leq \Delta_{\mathcal{N}}(\mu)$$

- ▶ **rigorous:** strict upper bound
- ▶ **efficient:**  $\mathcal{N}$ -independent
- ▶ **sharp:**  $\eta(\mu) := \frac{\Delta_{\mathcal{N}}(\mu)}{\|e_{\mathcal{N}}(\mu)\|} \approx 1$

- ▶ *Offline Phase:* Parameter choice for basis construction
- ▶ *Online Phase:* Verification of reduced solutions

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## Standard Approaches

### Initial Value Problems

- ▶ Method of lines + time-stepping schemes (implicit Euler/Crank-Nicolson)

For  $\{t_k\}_{k=1,\dots,K}$ , find  $u^k(\mu) = u(t_k; \mu) \in V$  with  $u^0(\mu) = u_0(\mu)$  and

$$\langle u^k, v \rangle + \Delta t a(t_k, u^k, v; \mu) = \langle u^{k-1}, v \rangle + \Delta t g(t_k, v; \mu) \quad \forall v \in V.$$

#### Note:

A Crank-Nicolson time-stepping scheme can be formulated as a space-time problem for the special choice by choosing a finite tensorized basis with

- piecewise linear trial functions in time,
- piecewise constant test functions in time.

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M. Grepl and A.T. Patera. *A Posteriori Error Bounds for Reduced-Basis Approximations of Parametrized Parabolic Partial Differential Equations*. M2AN, 2005.

B. Haasdonk and M. Ohlberger. *Reduced Basis Method for Finite Volume Approximations of Parameterized Linear Evolution Equations*. M2AN, 2008.



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### Periodic Problems

- ▶ Method of lines + time-stepping + **fixed-point iterations**:  $u^0(\mu) = u^K(\mu)$

#### Note:

Consequently, such a solution of a periodic problem requires

- an a-priori unknown number of repetitions of an initial value problem.

Usually, these approaches only yield a spatial reduced basis  $V^N \subset V^{\mathcal{N}_x}$  and thus also require time-stepping methods in the online phase.

## RB Error Bounds in time-stepping approaches

### Time-discrete error bound for periodic problems

The reduction error for a reduced basis based on an implicit Euler scheme can be bounded as follows:

$$\|e_N(\mu)\|_{\mathcal{Y}} \approx \left( \Delta t \sum_{k=1}^K \|e_N^k(\mu)\|_V^2 \right)^{\frac{1}{2}} \leq \left( \frac{\Delta t}{\alpha^2(\mu)} \sum_{k=1}^K \|r_N^k(\cdot; \mu)\|_{V'}^2 \right)^{\frac{1}{2}} =: \Delta_N^{\text{FP}, \mathcal{Y}}(\mu),$$

where

- ▶  $r_N^k(\mu) : V \rightarrow \mathbb{R}$  is the residual at time step  $t_k$
- ▶  $\alpha(\mu)$  is the coercivity constant of  $a(t, \cdot, \cdot; \mu)$
- ▶  $\mathcal{Y} := L_2(0, T; V)$

Here:

- ▶ Calculation of  $\|r_N^k(\cdot; \mu)\|_{V'}$ 
  - ▶ using Riesz representors  $\hat{r}_N^k(\mu) \in V$
  - ▶ offline-online decomposition

## Space-Time Formulation

### Notation:

$$\mathcal{Y} := L_2(0, T; V),$$

$$\mathcal{X}^{per} := L_2(0, T; V) \cap H_{per}^1(0, T; V') := \{u \in L_2(0, T; V) : u_t \in L_2(0, T; V'), u(0) = u(T) \text{ in } H\}$$

### Well-posedness

Let  $\mathcal{A}(t)$  be uniformly bounded and coercive. Then the problem

$$\text{Find } u(\mu) \in \mathcal{X}^{per} : \underbrace{\int_0^T \langle u_t, v \rangle dt + \int_0^T a(t, u, v; \mu) dt}_{=: b(u, v; \mu)} = \underbrace{\int_0^T g(t, v; \mu) dt}_{=: f(v; \mu)} \quad \forall v \in \mathcal{Y}.$$

is well-posed.

**Proof:** Babuška-Aziz, cf. [Schwab/Stevenson 2009]. □

⇒ **Additional dimension**

## Space-Time RB Error Bound

### Error bounds for a periodic space-time problem

The reduction error for a space-time reduced basis can be bounded as follows:

$$\|e_N(\mu)\|_{\mathcal{Y}} \leq \frac{\|r_N(\cdot; \mu)\|_{\mathcal{Y}'}}{\alpha(\mu)} =: \Delta_N^{\text{ST}, \mathcal{Y}}(\mu),$$

$$\|e_N(\mu)\|_{\mathcal{X}} \leq \frac{\|r_N(\cdot; \mu)\|_{\mathcal{Y}'}}{\beta(\mu)} =: \Delta_N^{\text{ST}, \mathcal{X}}(\mu),$$

where

- ▶  $r_N(\mu) : \mathcal{Y} \rightarrow \mathbb{R}$  is the space-time residual,
- ▶  $\beta(\mu) = \inf_{0 \neq u \in \mathcal{X}^{\text{per}}} \sup_{0 \neq v \in \mathcal{Y}} \frac{|b(u, v; \mu)|}{\|u\|_{\mathcal{X}^{\text{per}}} \|v\|_{\mathcal{Y}}}$  is the inf-sup constant of  $b(\cdot, \cdot; \mu)$ .

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- ▶ Calculation of  $\|r_N(\cdot; \mu)\|_{\mathcal{Y}'}$ :
  - ▶ using Riesz representors  $\hat{r}_N(\mu) \in \mathcal{Y}$
  - ▶ offline-online decomposition
- ▶ Both  $\alpha(\mu)$  and  $\beta(\mu)$  are usually replaced by efficient (i.e.  $\mathcal{N}$ -independent) lower bounds  $\alpha_{\text{LB}}(\mu)$ ,  $\beta_{\text{LB}}(\mu)$ .

## Short comparison

Space-Time	Fixed-Point Iterations
<p><i>Solution procedure offline:</i></p> <ul style="list-style-type: none"> <li>▶ Periodic basis functions</li> <li>▶ 1 large system (<math>\mathcal{O}(\mathcal{N}_t \mathcal{N}_x)</math>)</li> </ul>	<p><i>Solution procedure offline:</i></p> <ul style="list-style-type: none"> <li>▶ Time-stepping with unknown number of iterations <math>M</math></li> <li>▶ Many small systems (<math>MK\mathcal{O}(\mathcal{N}_x)</math>)</li> </ul>
<p><i>Basis construction:</i></p> <ul style="list-style-type: none"> <li>▶ Space-time basis functions</li> <li>▶ Space-time offline quantities</li> <li>→ high memory requirements</li> <li>▶ Time-dependent operators: ✓</li> </ul>	<p><i>Basis construction:</i></p> <ul style="list-style-type: none"> <li>▶ Spatial basis functions (POD)</li> <li>▶ Spatial offline quantities</li> <li>→ low memory requirements</li> <li>▶ Time-dependent operators: ?</li> </ul>
<p><i>Solution procedure online:</i></p> <ul style="list-style-type: none"> <li>▶ 1 reduced system (<math>\mathcal{O}(N_{ST})</math>)</li> </ul>	<p><i>Solution procedure online:</i></p> <ul style="list-style-type: none"> <li>▶ <math>MK</math> reduced systems (<math>\hat{a} \mathcal{O}(N_{FP})</math>)</li> </ul>

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## Model Problem

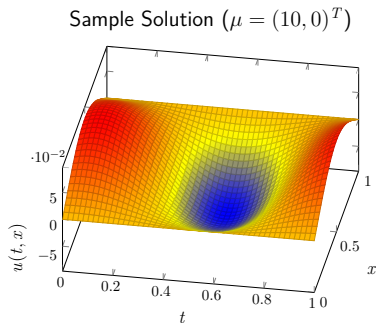
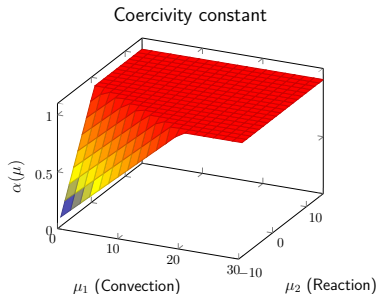
### Convection-Diffusion-Reaction Example

$$u_t - u_{xx} + \mu_1 \left(\frac{1}{2} - x\right) u_x + \mu_2 u = \cos(2\pi t) \quad \text{on } \Omega = (0, 1),$$

$$u(t, 0) = u(t, 1) = 0,$$

$$u(0, x) = u(T, x), \quad (T = 1).$$

Parameter Domain:  $\mu \in \mathcal{D} := [0, 30] \times [-9, 15]$ .





## Why Space-Time? – Effective error bounds

Space-Time inf-sup constant  $\beta(\mu)$

Consider **non-coercive initial-value** systems.

- ▶ Case 1: Diffusion-Convection ( $\mu_2 = 0$ ) – system asymptotically stable
- ▶ Case 2: Diffusion-Reaction ( $\mu_1 = 0$ ) – system asymptotically unstable

## Why Space-Time? – Effective error bounds

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Standard Crank-Nicolson error bound contains

- ▶ stability constant  $\rho_N(t^k, \mu)$  (energy estimates),
- ▶ with long-term behaviour  $e^{-\mu_i T}$ ,  $i = 1, 2$ .

⇒ **exponentially growing** error bounds.

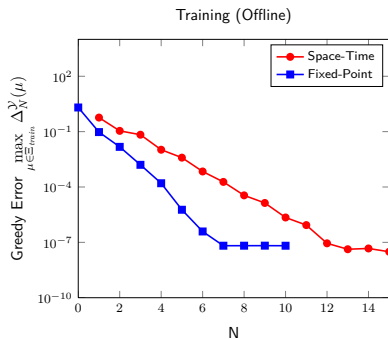
Space-time inf-sup constant:

- ▶ Case 1: long-term behaviour  $(\mu_1 T)^{-1}$
- ▶ Case 2: long-term behaviour  $e^{-\mu_2 T}$

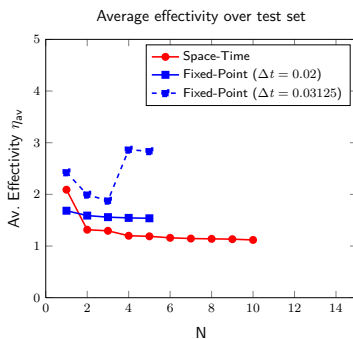
⇒ reflects **true system behaviour**.

## Why Space-Time? – Effective error bounds

$$\|r_N(\cdot; \mu)\|_{\mathcal{Y}'} \leftrightarrow \left( \sum_{k=1}^K \|r_N^k(\cdot; \mu)\|_{V'}^2 \right)^{1/2}$$



(a) Training ( $n_{\text{train}} = 400$ )



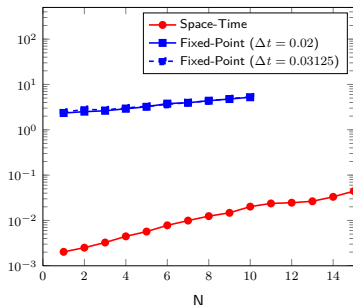
(b) Average Effectivity ( $n_{\text{test}} = 225$ )

$$\eta_{\text{av}} := \frac{1}{n_{\text{test}}} \sum_{\mu \in \Xi_{\text{test}}} \frac{\Delta_N^{\mathcal{Y}}(\mu)}{\|e_N(\mu)\|}$$

## Why Space-Time? – Online speed-up

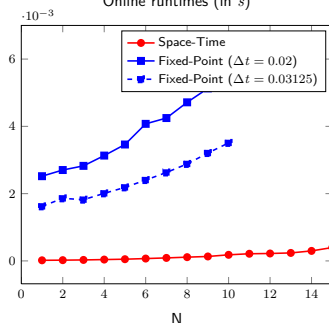
$$\mathcal{O}(N_{ST}) \leftrightarrow MK \mathcal{O}(N_{FP})$$

Online runtimes (in % of offline times)



(a) Runtime reduction

Online runtimes (in s)



(b) Online runtimes

### Implementations:

- ▶ Space-Time: LAWA
- ▶ Fixed-Point Iterations: `rb00mit/libmesh`

	Time Truth (in s)
Space-Time	0.8965
Fixed-Point ( $\Delta t = 0.02$ )	0.1076
Fixed-Point ( $\Delta t = 0.03125$ )	0.0662

LAWA: Library for Adaptive Wavelet Algorithms, [lawa.sourceforge.net](http://lawa.sourceforge.net), 2001

libmesh: Kirg, Stogner and Carey, *A C++ Library for Parallel Adaptive Mesh Refinement / Coarsening Simulations*, Eng. with Comp., 2006

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# Adaptive Space-Time Methods

## Initial Value Problems:

- ▶ “Right” choice of space-time basis leads to Crank-Nicolson scheme.
- ▶ Calculations decouple.

## Periodic Problems:

- ▶ Crank-Nicolson scheme has to be combined with fixed-point iterations.
- ▶ Calculations do not decouple.

### Idea

Adaptive space-time algorithm.

## Wavelets:

Variational formulation equivalent to bi-infinite matrix-vector problem:

$$b(u, v; \mu) = f(v; \mu), \quad u \in \mathcal{X}, v \in \mathcal{Y} \quad \iff \quad \mathbf{B}u = \mathbf{f}, \quad u, f \in \ell_2$$

## Conclusion

Space-Time methods can lead to

- ▶ inf-sup constants that reflect the true behaviour of the underlying systems,
- ▶ more effective error bounds,
- ▶ faster reduced models.

The drawback of the additional dimension can be circumvented:

- ▶ Initial value problems: specific basis yields Crank-Nicolson scheme. [YPU]
- ▶ Periodic problems: adaptive space-time methods.

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M. Yano, A.T. Patera and K. Urban *A Space-Time Certified RBM for Burgers' Equation. Submitted to Math. Models and Methods in Appl. Sc.* , 2012.

M. Yano. *A Space-Time Petrov-Galerkin Certified RBM – Application to the Boussinesq Equation. Submitted to SIAM J. Sc. Comp.* , 2012.

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**Thank you for your attention**

