

# Intrusive Reduced Order Modeling for Inverse Problems

AS5401: Data-driven modeling of complex aerospace systems and fluid flows

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# Agenda

- 1 Aim and Motivation
- 2 Transient Heat Conduction
- 3 Coupled Reaction- Diffusion
- 4 Wave Equation
- 5 Electrohydrodynamics
- 6 Conclusions

## Aim and Motivation

# Aim and Motivation

## Aim

- Inverse Problems are the tasks of using observed data and estimating the values of unknown parameters that characterize the system.
- The aim of this project is to **formulate and demonstrate the inverse problem through Reduced order models (ROM)** for various partial equations.

## Motivation

- How effectively one can leverage ROM's for such problems. **Example:** Calculation of thermal diffusivity from set(s) of temperature profile data.
- Transient Heat Conduction (1D **linear** PDE)- **Inverse problem**
- Reaction- Diffusion (1D **linear coupled** PDE)- **Coupling systems**
- Wave (1D **nonlinear** PDE)- **POD-DEIM**
- Electrohydrodynamics (1D **nonlinear coupled** PDE)- **Actual system of interest**

# Transient Heat Conduction

# Transient Heat Conduction

## Intrusive ROM

- One-dimensional transient temperature variation (first order linear PDE) with thermal diffusivity ( $\alpha$ ),

$$\frac{\partial T}{\partial t} = \alpha \frac{\partial^2 T}{\partial x^2}, \quad 0 < \alpha \leq 1, \quad 0 \leq x \leq 1$$

- For snapshot matrix,  $\mathcal{T} = [T(t_1), T(t_2), \dots, T(t_n)]$  modes ( $\mathbf{U}_T$ ) are extracted through SVD. Truncation rank is chosen for energy of atleast 99.9% ( $\rho_T = 3$ ) and reduced order model (ROM) is formulated as,

$$\frac{\partial \hat{T}}{\partial x} = \mathbf{U}_T^\top \alpha \mathbf{U}_T \frac{\partial^2 \hat{T}}{\partial x^2} = \hat{\mathbf{A}} \hat{T}$$

# Transient Heat Conduction

## Validation

### ■ Boundary lifting,

decomposing solution as sum of lifting function (satisfy BC's) and ROM expansion [1].

### ■ Golden search technique has been implemented for convex optimization. The objective $\mathcal{J}$ is written as,

$$\mathcal{J}_\alpha = \frac{1}{2} \|T - \bar{T}\|^2$$

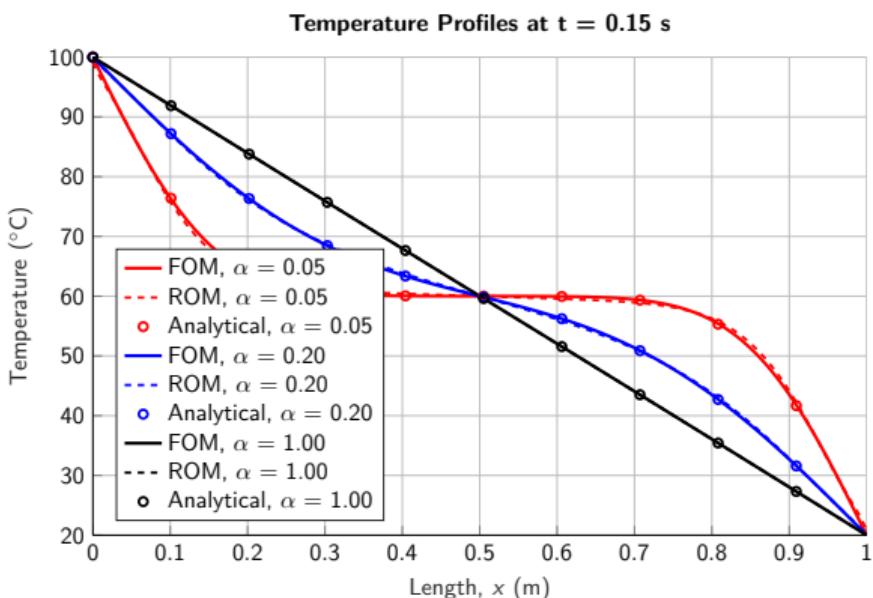
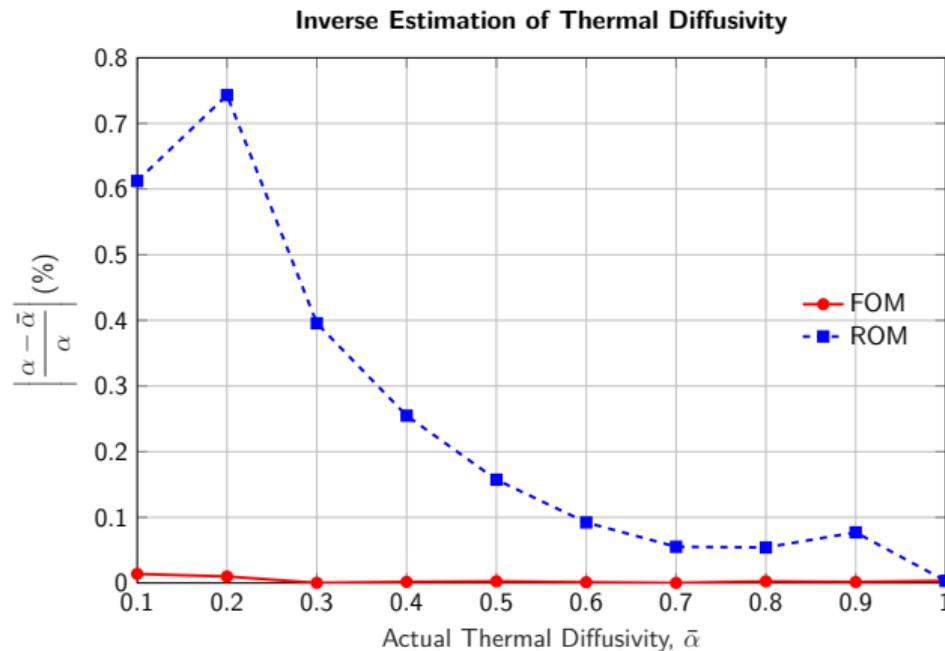


Figure: FOM and ROM validation

## Transient Heat Conduction

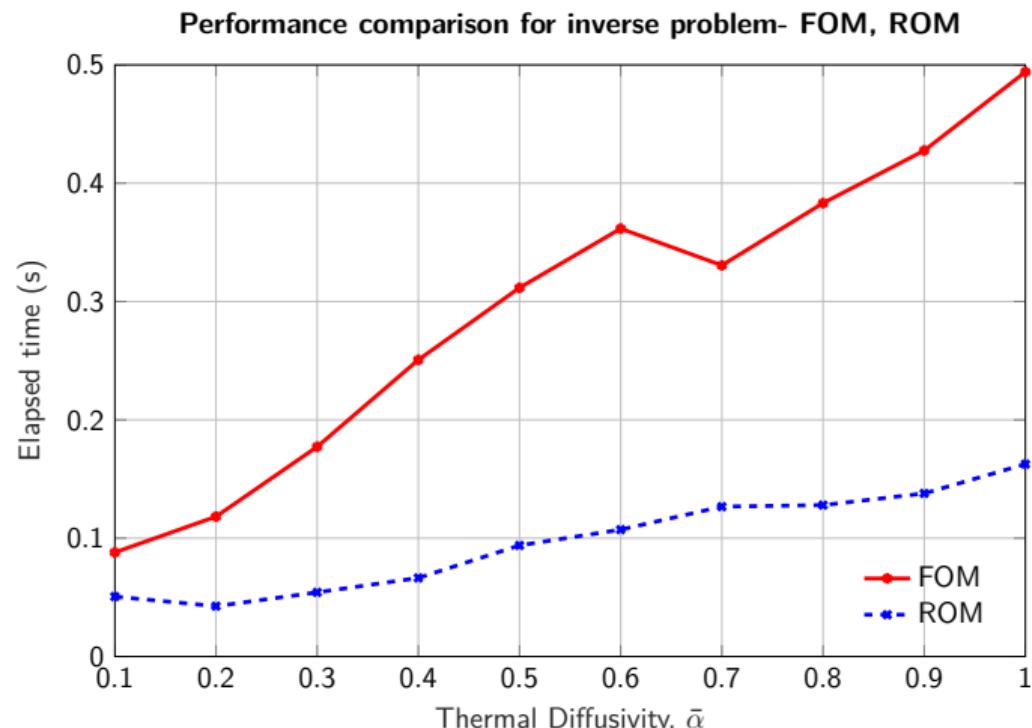
## Inverse Problem- Golden Search



**Figure:** Prediction of thermal diffusivity

# Transient Heat Conduction

Inverse Problem- Golden Search



# Coupled Reaction- Diffusion

# Coupled Reaction- Diffusion

## Intrusive ROM

- One-dimensional coupled reaction- diffusion equation with system constants,  $a, b, \alpha, \beta$

$$\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2} + \alpha(v - u) + F_u, \quad u(0, t) = u_L, u(1, t) = u_R$$

$$\frac{\partial v}{\partial t} = b \frac{\partial^2 v}{\partial x^2} + \beta(u - v) + F_v, \quad v(0, t) = v_L, v(1, t) = v_R$$

$$x \in [0, 1], \quad t > 0$$

- The **methods of manufactured solution** was implemented for validation of FOM.

# Coupled Reaction- Diffusion

## Intrusive ROM

- For snapshot matrix,  $\mathcal{U}, \mathcal{V}$ , modes  $(\mathbf{U}_u, \mathbf{U}_v)$  for  $u, v$  truncated to 99.9% ( $\rho_u = \rho_v = 1$ ) of energy and ROM can be formulated as,

$$\frac{\partial \hat{u}}{\partial t} = \mathbf{U}_u^\top a \mathbf{U}_u \frac{\partial^2 \hat{u}}{\partial x^2} + \mathbf{U}_u^\top \alpha (\mathbf{U}_v \hat{v} - \mathbf{U}_u \hat{u}) + \mathbf{U}_u^\top F_u + \textcolor{blue}{\alpha} \mathbf{U}_u^\top (\tilde{v} - \tilde{u})$$

$$\frac{\partial \hat{v}}{\partial t} = \mathbf{U}_v^\top b \mathbf{U}_v \frac{\partial^2 \hat{v}}{\partial x^2} + \mathbf{U}_v^\top \beta (\mathbf{U}_u \hat{u} - \mathbf{U}_v \hat{v}) + \mathbf{U}_v^\top F_v + \textcolor{blue}{\beta} \mathbf{U}_v^\top (\tilde{u} - \tilde{v})$$

$$\tilde{u} = u_L \frac{1-x}{L} + u_R \frac{x}{L}, \quad \tilde{v} = v_L \frac{1-x}{L} + v_R \frac{x}{L}$$

- Coupling of ROM through **transformation of subspace from  $\tilde{u} \leftrightarrow \tilde{v}$**  is due to boundary lifting.

# Coupled Reaction- Diffusion

## Validation

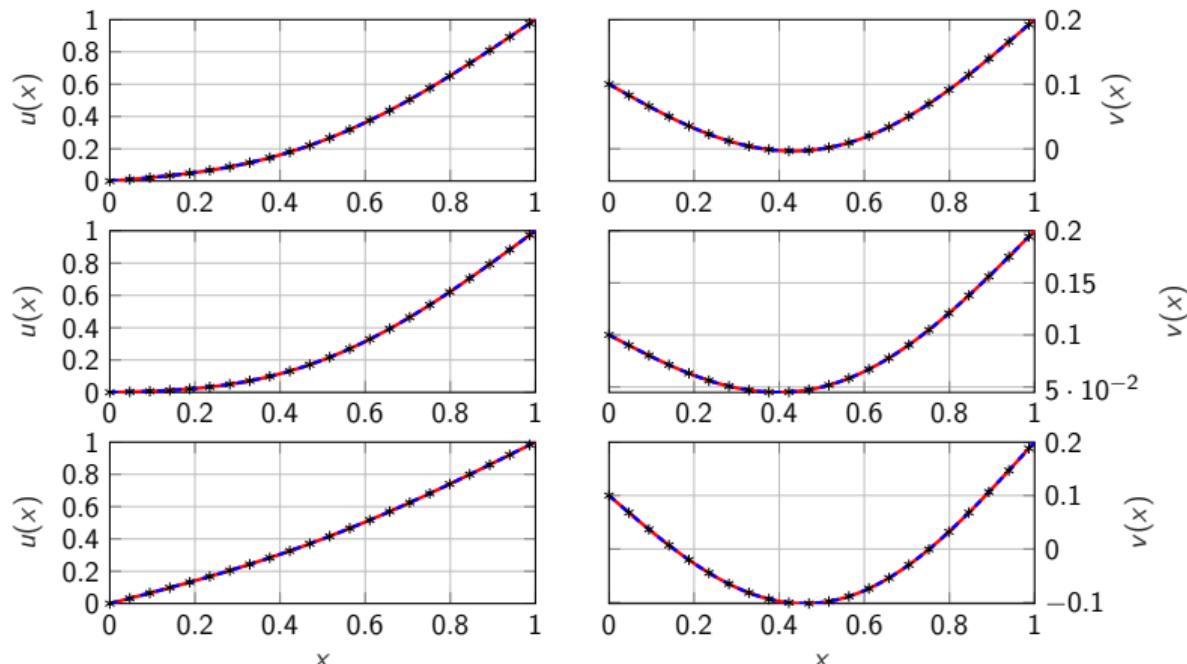
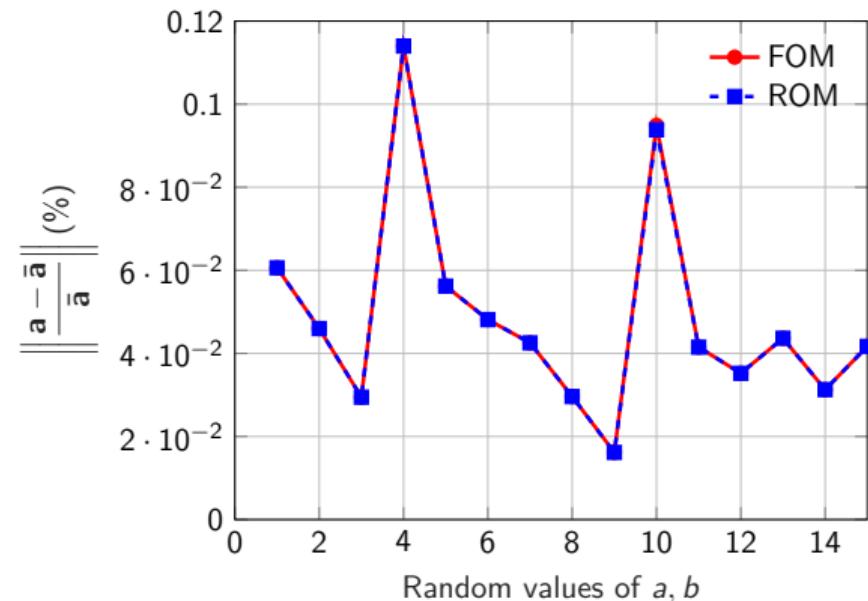


Figure: Validation for three  $U, V$  values. (FOM,ROM, \* Analytical)

# Coupled Reaction- Diffusion

Inverse Problem- Golden Search

**Inverse Estimation Error for  $a = (a \in [0.1, 1], b \in [0.1, 2])$**



**Figure:** Inverse Error estimation

# Coupled Reaction- Diffusion

Inverse Problem- Golden Search

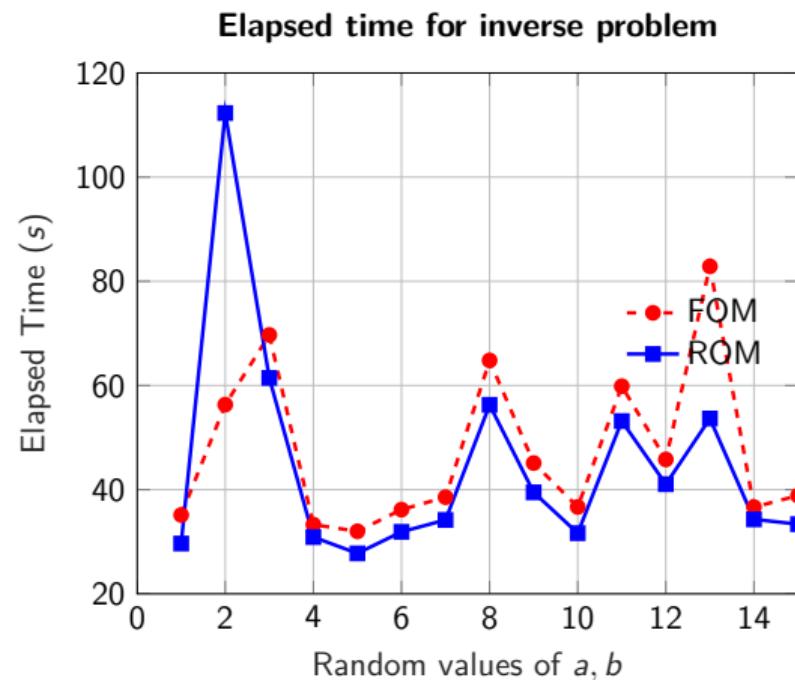


Figure: Elapsed time for inverse problem

# Wave Equation

# Wave Equation

## Intrusive ROM

- Formulation of one-dimensional non-linear wave equation for FOM, ROM such that snapshot matrix,  $\mathcal{U}$  has modes ( $\mathbf{U}_u$ ) with 99.9% ( $\rho_u = 3$ ) energy are,

$$\frac{\partial u}{\partial t} = -u \frac{\partial u}{\partial x} = -\frac{\partial(u^2/2)}{\partial x} = -N, \quad x \in [0, 1], t \in (0, 0.3], \quad u(x, 0) = \sin(\pi x/L)$$

$$\frac{\partial \hat{u}}{\partial t} = -\mathbf{U}_u^\top \left[ \frac{\partial (\mathbf{U}_u \hat{u})^2/2}{\partial x} \right]$$

- Modes of non linear term,  $\mathbf{N} = [N(t_1), N(t_2), \dots, N(t_n)]$  is  $\mathbf{U}_N$  with 99% ( $\rho_N = 3$ ) energy. Reduced non-linear model through DEIM [2] is,

$$\frac{\partial \hat{u}}{\partial t} = -\mathbf{U}_u^\top \hat{N}$$

$$\hat{N} = \mathbf{U}_N^\top (\mathbf{P} \mathbf{U}_N)^\dagger \mathbf{P}^\top N, \quad \mathbf{P} \leftarrow \text{DEIM}(\mathbf{U}_N)$$

# Wave Equation

## Intrusive ROM

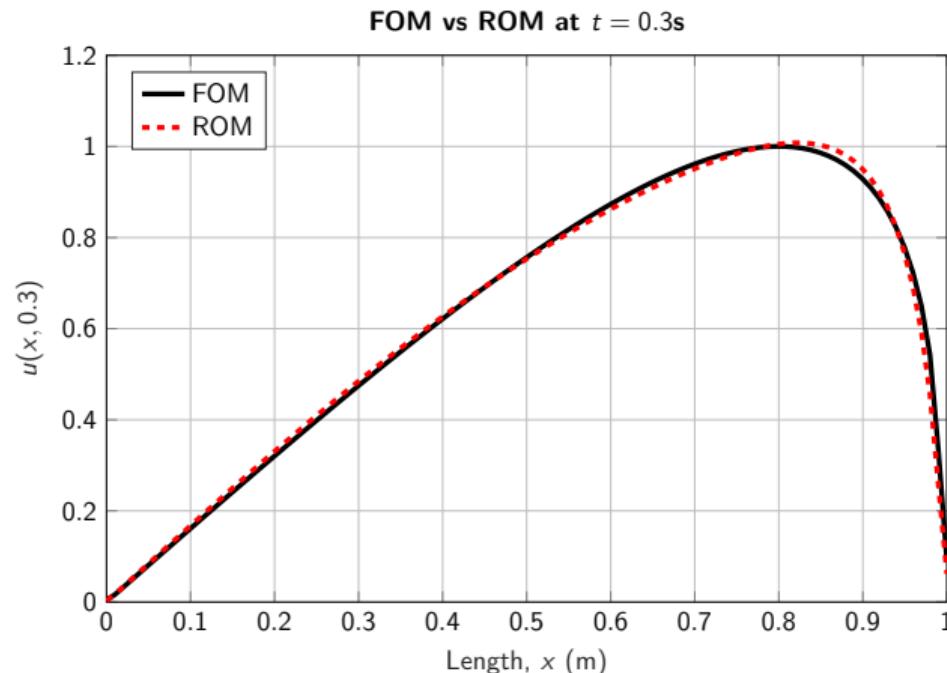


Figure: Velocity distribution at  $t = 0.3\text{s}$

# Wave Equation

## Intrusive ROM

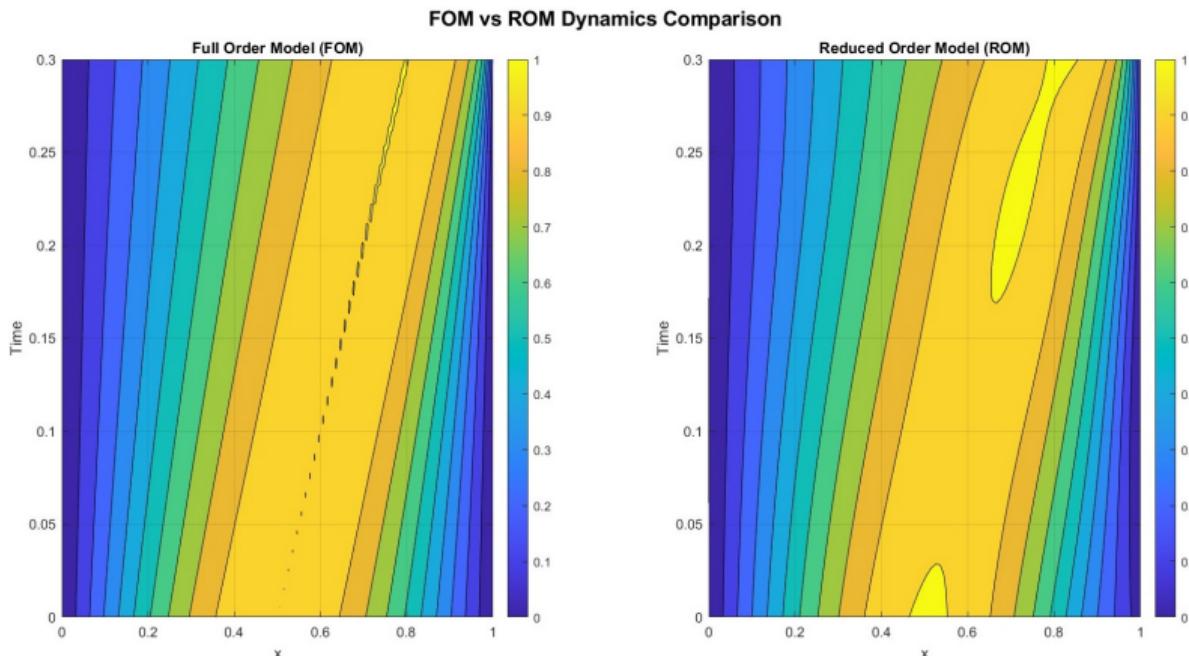


Figure: Transient Velocity distribution for FOM, ROM

# Electrohydrodynamics

# Electrohydrodynamics

## Intrusive ROM

- EHD is one-dimensional coupled non-linear equation but expressed as,

$$\frac{1}{\mu_e} \frac{\partial}{\partial t} \left( \frac{\partial^2 \phi}{\partial x^2} \right) = \frac{\partial^3 \phi}{\partial x^3} \frac{\partial \phi}{\partial x} + \frac{\partial^2 \phi}{\partial x^2} \frac{\partial^2 \phi}{\partial x^2}$$

- Snapshot modes,  $\mathbf{U}_\phi$  have 99.9% of energy,

$$\frac{\partial}{\partial t} \left( \frac{\partial^2 \hat{\phi}}{\partial x^2} \right) = \mu_e \mathbf{U}_\phi^\top \left( \mathbf{U}_\phi \frac{\partial^3 \hat{\phi}}{\partial x^3} \mathbf{U}_\phi \frac{\partial \hat{\phi}}{\partial x} + \mathbf{U}_\phi \frac{\partial^2 \hat{\phi}}{\partial x^2} \mathbf{U}_\phi \frac{\partial^2 \hat{\phi}}{\partial x^2} \right)$$

- Currently, ROM model is **numerically unstable** and are under research.

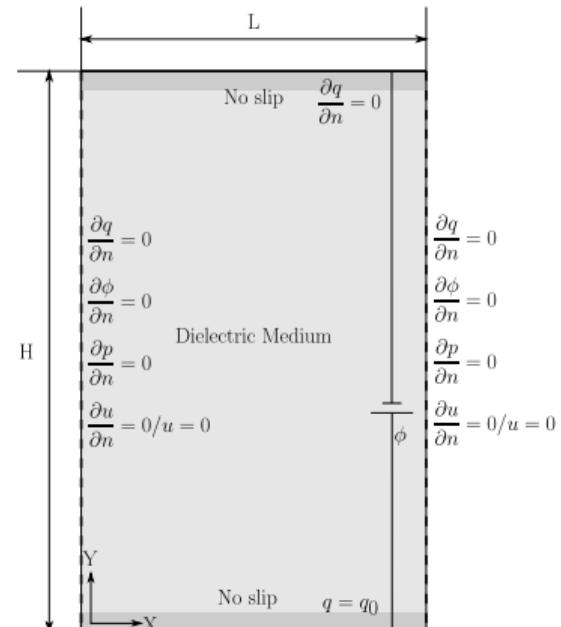


Figure: Schematic of the EHD [3] 21 / 30

# Conclusions

# Conclusions

- Inverse search problems are **faster** in execution than full order model
- Coupled diffusion-reaction ROM model is an **surrogate model** with minimal error and computing time
- Boundary lifting often **enhances ROM accuracy**
- Currently for EHD there is **numerical instability** and being researched upon.

## Takeaway

- Gradient based solver with ROM lays a groundwork for **real time inverse prediction**
- One must **account human hours** into making ROM for digital twins.

## References

- [1] Fabian Key, Max von Danwitz, Francesco Ballarin, and Gianluigi Rozza. Model order reduction for deforming domain problems in a time-continuous space-time setting. *International Journal for Numerical Methods in Engineering*, 124(23): 5125–5150, August 2023. ISSN 1097-0207. doi: 10.1002/nme.7342.
- [2] M. Rojas M.M. Baumann, M.B. Van Gijzen. Nonlinear model order reduction using pod/deim for optimal control of burgers' equation. *mathesis*, TU Delft, 2013. URL <https://github.com/ManuelMBaumann/MasterThesis>.
- [3] P.A. Vázquez and A. Castellanos. Numerical simulation of ehd flows using discontinuous galerkin finite element methods. *Computers & Fluids*, 84:270–278, September 2013. ISSN 0045-7930. doi: 10.1016/j.compfluid.2013.06.013.

# Validation

# Heat Equation

## Analytical Solution

- For 1D transient heat conduction, analytical solution employed for validation is,

$$T(x, t) = T_s(x) + \theta(x, t)$$

$$T_s(x) = T_L \frac{1-x}{L} + T_R \frac{x}{L}$$

$$\theta(x, t) = \sum_{n=1}^{\infty} B_n \sin(n\pi x/L) \exp(-\alpha(n\pi/L)^2 t)$$

$$B_n = \frac{2}{L} \int_0^L (T_0(x, t) - T_s(x)) \sin(n\pi x/L)$$

# Reaction- Diffusion Equation

## Methods of Manufactured Solutions

- Solution of  $u, v$  are assumed to have form, where  $U, V$  are unknown constants. Further,  $F_u, F_v$  are calculated as which satisfy the governing equation

$$u = \tilde{u} + U \sin(\pi x) \cos(2\pi t), \quad \tilde{u} = u_L \frac{1-x}{L} + u_R \frac{x}{L}$$

$$v = \tilde{v} + V \sin(\pi x) \cos(2\pi t), \quad \tilde{v} = v_L \frac{1-x}{L} + v_R \frac{x}{L}$$

$$F_u = \frac{\partial u}{\partial t} - a \frac{\partial^2 u}{\partial x^2} - \alpha(v - u)$$

$$F_v = \frac{\partial v}{\partial t} - b \frac{\partial^2 v}{\partial x^2} - \beta(u - v)$$

- The whole idea is to assume any constant values for  $a, b, \alpha, \beta, U, V$  compute the  $F_u, F_v$  then solve the FOM for validation against analytical expression.

# Reaction- Diffusion Equation

Coupling term- Boundary lifting

$$\begin{aligned} u - v &= \tilde{u} - \tilde{v} + U \sin(\pi x) \cos(2\pi t) - V \sin(\pi x) \cos(2\pi t) \\ v - u &= \tilde{v} - \tilde{u} + V \sin(\pi x) \cos(2\pi t) - U \sin(\pi x) \cos(2\pi t) \end{aligned}$$

- $u - v$  and  $v - u$  introduces the error of  $\tilde{u} - \tilde{v}$  and  $\tilde{v} - \tilde{u}$  and corrected with  $\alpha(\tilde{u} - \tilde{v})$  and  $\beta(\tilde{u} - \tilde{v})$ . For the ROM, the subspace has to be transformed.

$$\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2} + \alpha(v - u) + F_u + \alpha(\tilde{u} - \tilde{v})$$

$$\frac{\partial v}{\partial t} = b \frac{\partial^2 v}{\partial x^2} + \beta(u - v) + F_v + \beta(\tilde{u} - \tilde{v})$$

# Electrohydrodynamics

## Numerical validation

- The coupled EHD system is formulated as,

$$\nabla \cdot (\epsilon \nabla \phi) = -q$$

$$\mathbf{E} = -\nabla \phi$$

$$\frac{\partial q}{\partial t} + \nabla \cdot (\mu_e q \mathbf{E}) = 0$$

- These set of equations are coupled as,

$$\frac{1}{\mu_e} \frac{\partial}{\partial t} \left( \frac{\partial^2 \phi}{\partial x^2} \right) = \frac{\partial^3 \phi}{\partial x^3} \frac{\partial \phi}{\partial x} + \frac{\partial^2 \phi}{\partial x^2} \frac{\partial^2 \phi}{\partial x^2}$$

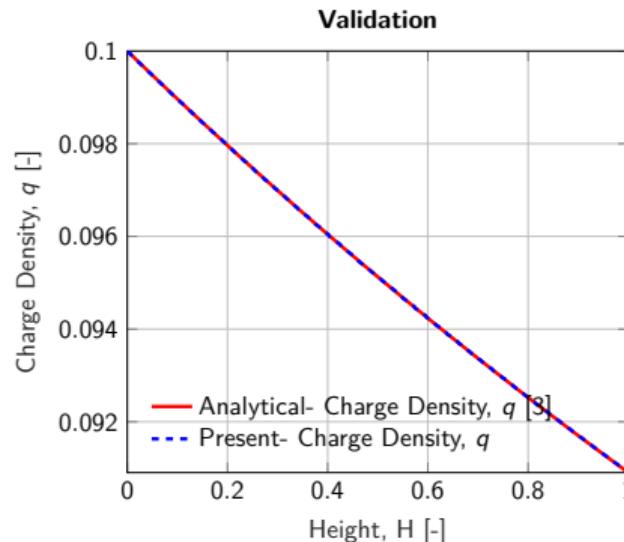


Figure: Validation of weak injection ( $C = 0.1$ )

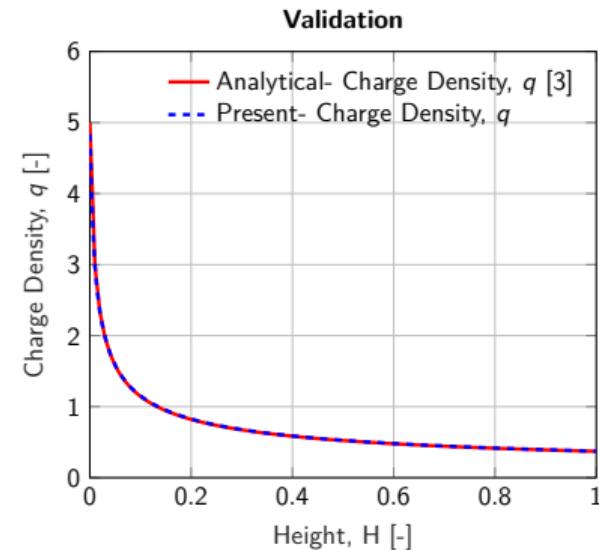


Figure: Validation of strong injection ( $C = 10$ )