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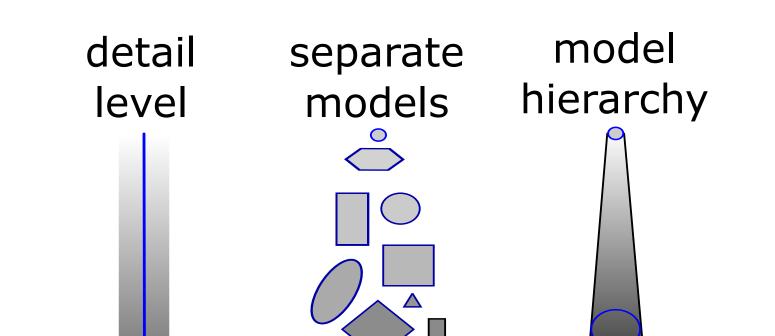
# Model Reduction and Simulation using Hierarchical Moment Models

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#### **Application Background**

Inaccurate classical fluid models in (1) atmospheric reentry of spacecraft (2) fusion reactor and plasma simulations (3) shallow free surface flows



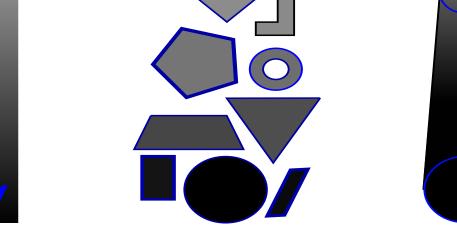
#### **Model Hierarchies**

+ preserve structure of the model

- + allow for efficient numerics
- + enable adaptivity within simulation

#### Challenges

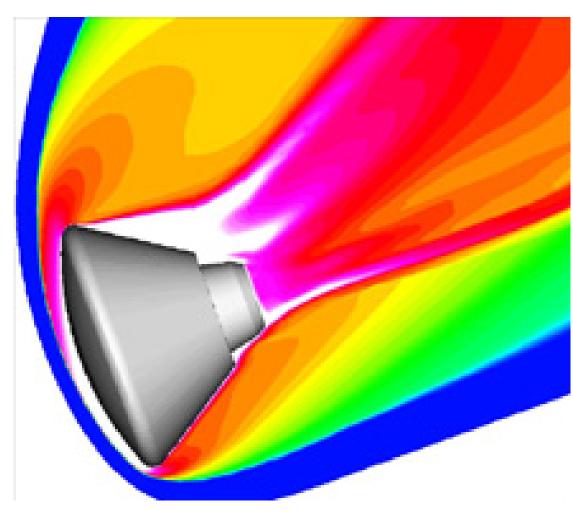
Modeling, analysis and numerics need to consider high-dimensionality
multiple scales
adaptivity



#### **Goal: Self-learning models**

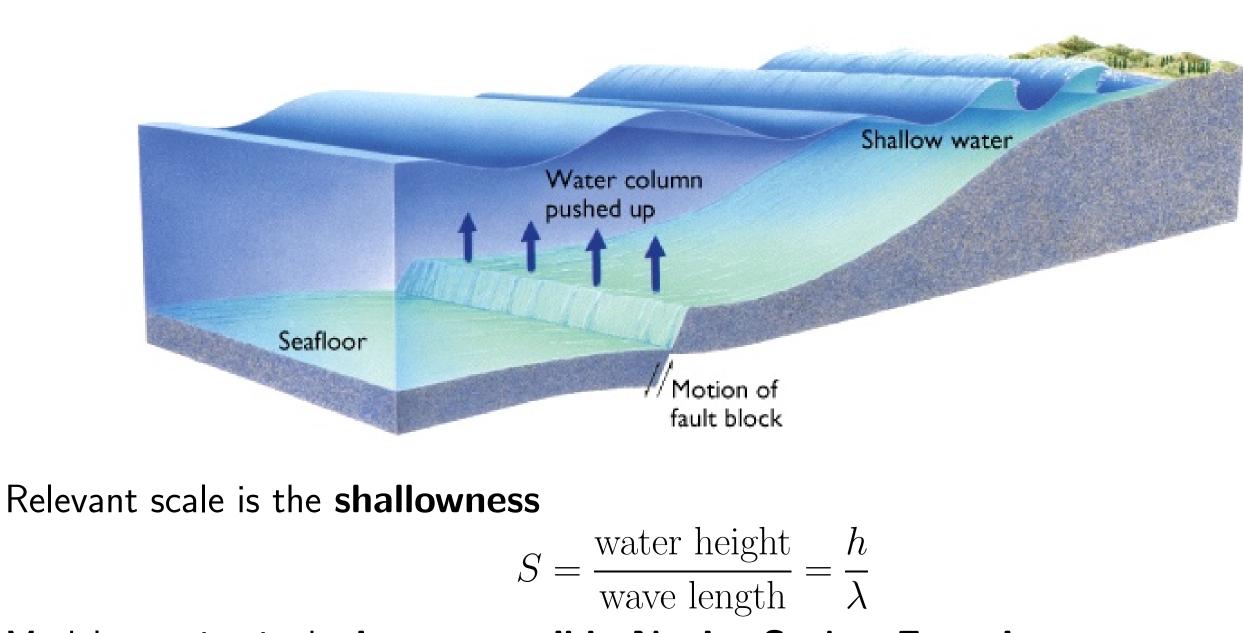
 $\Rightarrow$  moment models as hierarchical PDEs

#### **Application 1: Atmospheric Reentry**

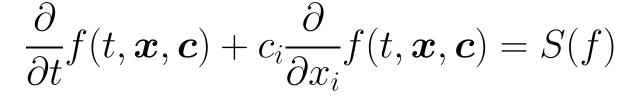


Relevant scale is the **Knudsen number**  $Kn = \frac{\text{mean free path length}}{\text{reference length}} = \frac{\ell}{L}$ Model equation is the **Boltzmann Transport Equation** 

#### **Application 2: Shallow Flows**



Model equation is the **incompressible Navier-Stokes Equation** 



#### **Ansatz: Hermite Expansion** [1, 2]

Expand unknown distribution function in Hermite series around equilibrium Maxwellian

$$f(t, x, c) = \sum_{i=0}^{M} f_i(t, x) \phi_i\left(\frac{c-v}{\sqrt{\theta}}\right)$$

Leads to hyperbolic moment model

 $\partial_t \boldsymbol{u}_M + \boldsymbol{A} \partial_x \boldsymbol{u}_M = \boldsymbol{S}, \quad \boldsymbol{u}_M = (\rho, v, \theta, f_3, f_4, \dots, f_M)^T$ 

#### **Results: Hypersonic channel flow** [2, 5]



## $\nabla \cdot \boldsymbol{u} = 0, \quad \partial_t \boldsymbol{u} + \boldsymbol{u} \cdot \nabla \boldsymbol{u} = -\frac{1}{\rho} \nabla p + \frac{1}{\rho} \nabla \cdot \boldsymbol{\sigma} + g$

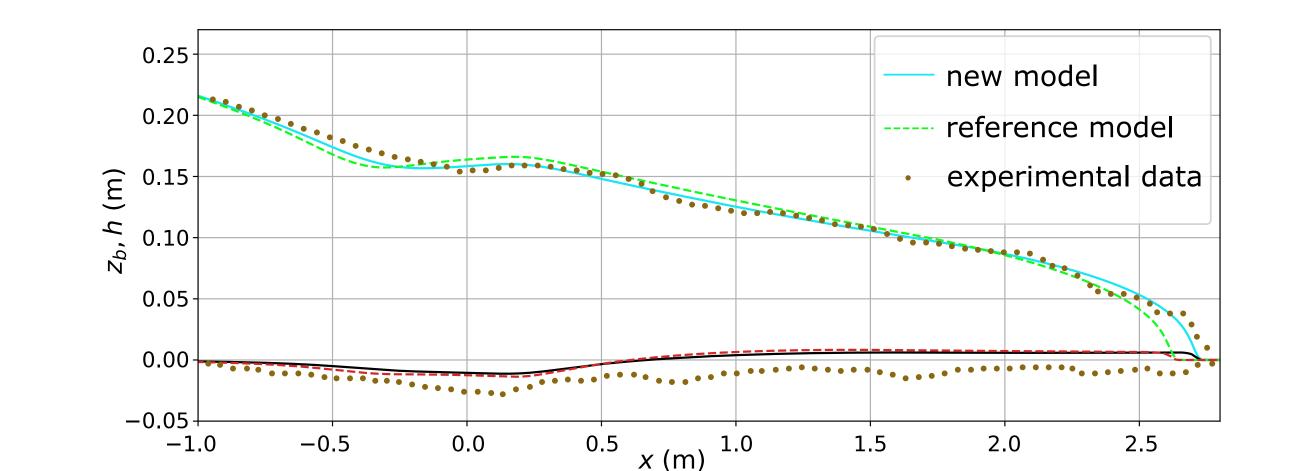
#### **Ansatz: Legendre Expansion** [3]

Expand unknown horizontal velocity profile in Legendre series around mean velocity

$$u(t, x, z) = u_m(t, x) + \sum_{i=1}^{M} \alpha_i(t, x)\phi_i\left(\frac{z - h_b}{h}\right)$$

Leads to hyperbolic Shallow Water Moment Model [4]  $\partial_t \boldsymbol{u}_M + \boldsymbol{A}_M \partial_x \boldsymbol{u}_M = \boldsymbol{S}, \quad \boldsymbol{u}_M = (h, u_m, \alpha_1, \alpha_2, \dots, \alpha_M)^T \in \mathbb{R}^{M+2}$ 

### **Results: Shallow sediment transport** [6]



#### **Next challenges:**

(1) derivation of **hybrid moment models** (2) development of **adaptive numerical schemes** (3) simulation of **application tests** 

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#### References

[1] Y. Fan, et al. Model Reduction of Kinetic Equations by Operator Projection, J. Stat. Phys., (2016)

[2] J. Koellermeier. Derivation and numerical solution of hyperbolic moment equations for rarefied gas flows, RWTH Aachen University, (2017)

[3] J. Kowalski, M. Torrilhon. Moment Approximations and Model Cascades for Shallow Flow, Commun. Comp. Phys., (2019) [4] J. Koellermeier, M. Rominger. Analysis and Numerical Simulation of Hyperbolic Shallow Water Moment Equations, *Commun. Comp. Phys.*, (2020)

[5] J. Koellermeier, G. Samaey. Projective Integration Schemes for Hyperbolic Moment Equations, *Kinet. Relat. Mod.*, (2021) [6] J. Garres-Díaz, et al. Shallow Water Moment models for bedload transport problems, Commun. in Comput. Phys., (2021)

