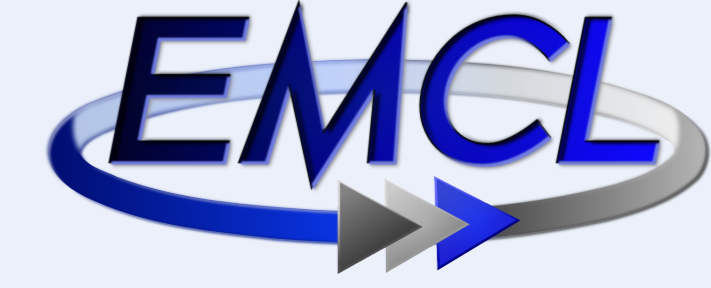


# Model development for meteorological applications

Martin Baumann, Simon Gawlok, Philipp Gerstner, Vincent Heuveline, Nils Schween, Chen Song, Peter Zaspel  
Engineering Mathematics and Computing Lab, Heidelberg University



UNIVERSITÄT  
HEIDELBERG  
ZUKUNFT  
SEIT 1386



## Low-Mach (LM) and Compressible Navier-Stokes (CNS) Model for Cyclone-Cyclone Interaction

CNS vs. LM:

- Variables:
  - $v$ : Velocity field
  - $\rho^*$ : density of fluid
  - $\theta^*$ : temperature of density of fluid
  - $p^*$ : pressure within fluid
- Modelization of fluid's behavior:
  - Conservation of momentum
  - Conservation of mass
  - Conservation of energy/temperature
  - Ideal gas law

► Atmosphere: many relevant flows in LM regime

Key observation for LM:

- Split total pressure  $p$  in three parts:

$$p(x, t) := p_{th}(t) + p_0(x) + p^*(x, t)$$

- $p_{th}$ : thermodynamic pressure (const. in space),
- $p_0$ : hydrostatic pressure (const. in time),
- $p^*$ : hydrodynamic pressure

► In LM regime:

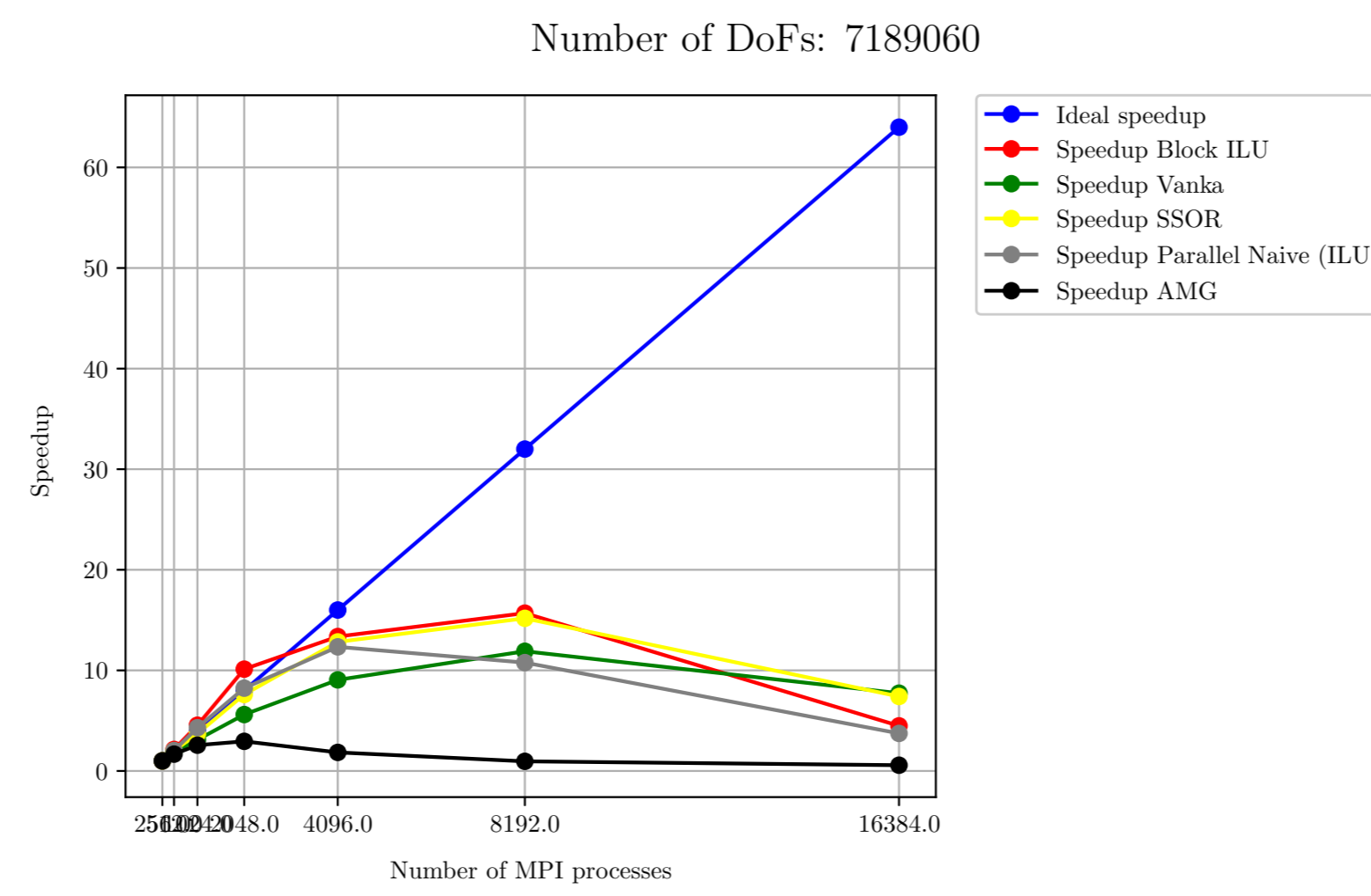
$$|p^*(x, t)| \ll |p_{th}(t) + p_0(x)|$$

⇒ Neglect  $p^*$  in ideal gas law

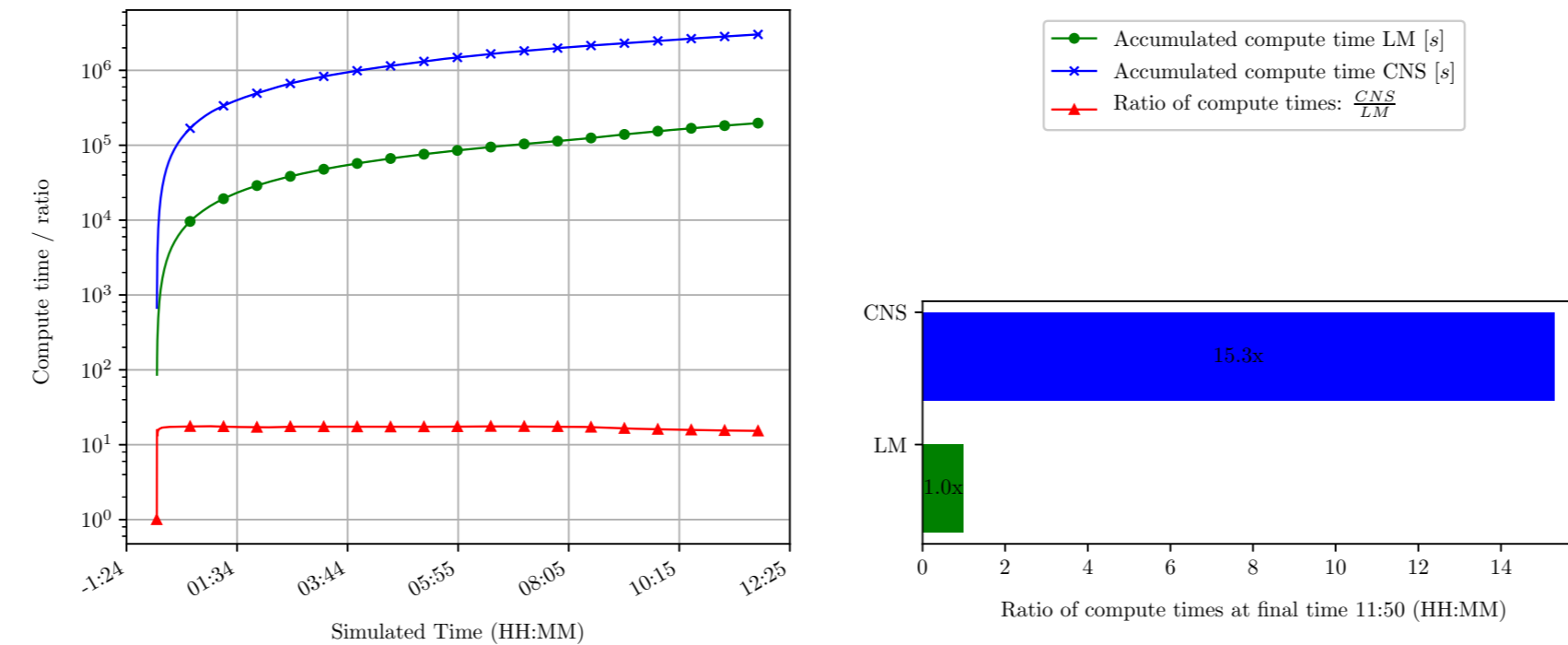
⇒ Damping of acoustic modes in LM

Numerical Solver:

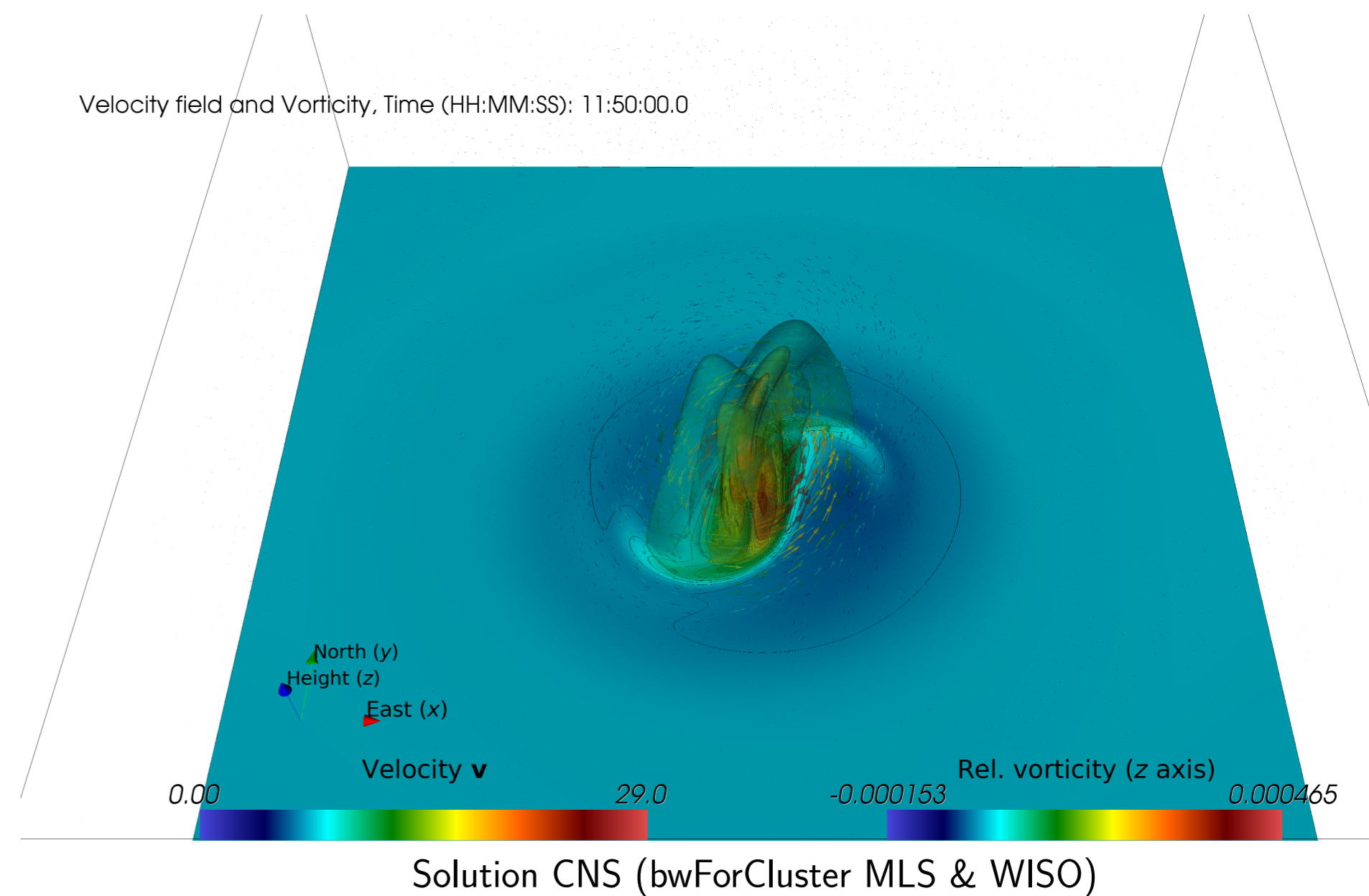
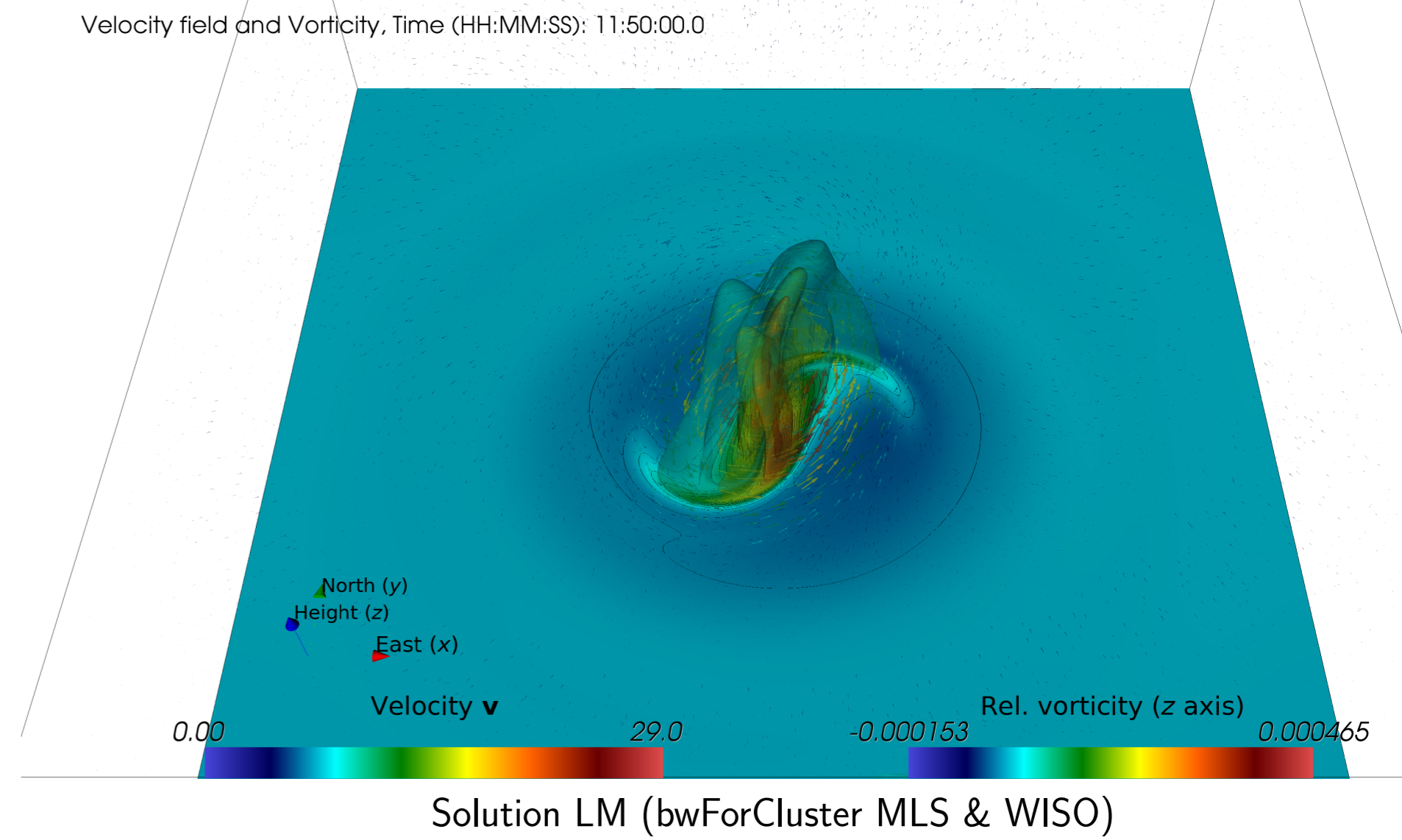
- Finite element discretization in space, finite differences in time
- Nonlinear solver: Newton's method
- Linear solver CNS:
  - Generalized Minimum Residual (GMRES) method
  - Preconditioner: BoomerAMG of hypre library
- Linear solver LM:
  - Flexible Generalized Minimum Residual (FGMRES) method
  - Preconditioner: nested Schur complement approach



Scaling for Poisson's equation on JUQUEEN



Runtime CNS vs. LM (bwForCluster MLS & WISO)



## Thermal Electro-Hydrodynamical Boussinesq Equations

Boussinesq approximation, augmented by dielectrophoretic force and Gauss law

$$\partial_t u + (u \cdot \nabla) u - \nu \Delta u + \frac{1}{\rho} \nabla p = \alpha_f (\nabla \Phi)^2 \nabla \theta - \alpha_g (\theta - \theta_r) \vec{g}$$

$$\nabla \cdot u = 0$$

$$\partial_t \theta + (u \cdot \nabla) \theta - \kappa \Delta \theta = 0$$

$$-\nabla \cdot (\epsilon (1 + \epsilon(\theta_r - \theta)) \nabla \Phi) = 0$$

$$E = \nabla \Phi$$

System matrix after FEM discretization

$$\mathcal{K} = \begin{pmatrix} \mathcal{A} & \mathcal{B} \\ \mathcal{C} & \mathcal{D} \end{pmatrix}$$

Temperature-potential matrix

$$\mathcal{A} = \begin{pmatrix} A_\theta & 0 \\ * & A_\Phi \end{pmatrix}$$

Incompressible N-S matrix

$$\mathcal{D} = \begin{pmatrix} A_v + N & B^T \\ -B & 0 \end{pmatrix}$$

Optimal preconditioner

$$\mathcal{P} = \begin{pmatrix} \mathcal{A} & \mathcal{B} \\ 0 & \mathcal{S} \end{pmatrix}$$

with spectrum

$$\sigma(\mathcal{P}^{-1} \mathcal{K}) = \{1\}$$

Computable preconditioner

$$\hat{\mathcal{P}} = \begin{pmatrix} \mathcal{A} & \mathcal{B} \\ 0 & \hat{\mathcal{S}} \end{pmatrix}$$

Exact SC

$$\mathcal{S} = \mathcal{D} - \mathcal{C} \mathcal{A}^{-1} \mathcal{B}$$

$$= \begin{pmatrix} A_v + N + Q(A) & B^T \\ -B & 0 \end{pmatrix}$$

Approximate SC

$$\hat{\mathcal{S}} = \begin{pmatrix} A_v & B^T \\ -B & 0 \end{pmatrix}$$

→ Instantaneous Stokes matrix

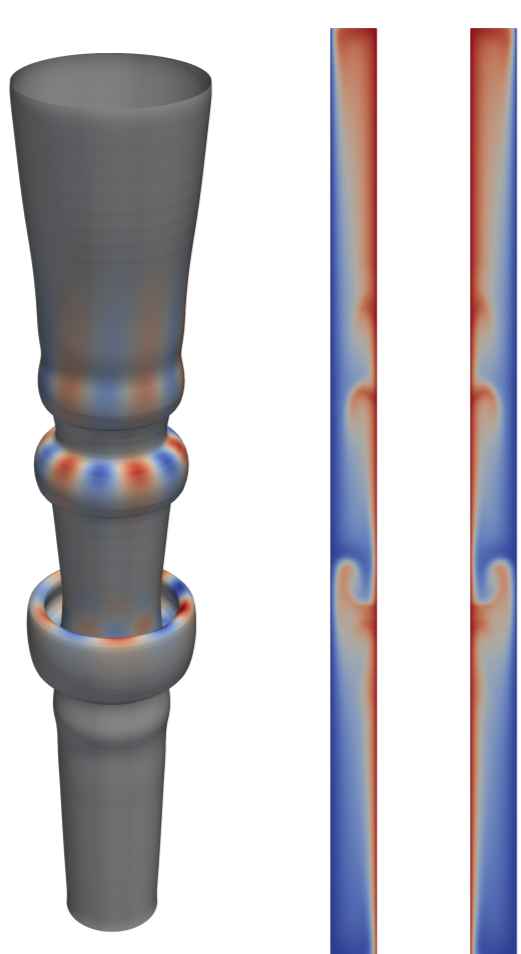
Involved sub solvers

- Pressure convection diffusion preconditioner for Stokes system
- Algebraic Multigrid for elliptic sub matrices

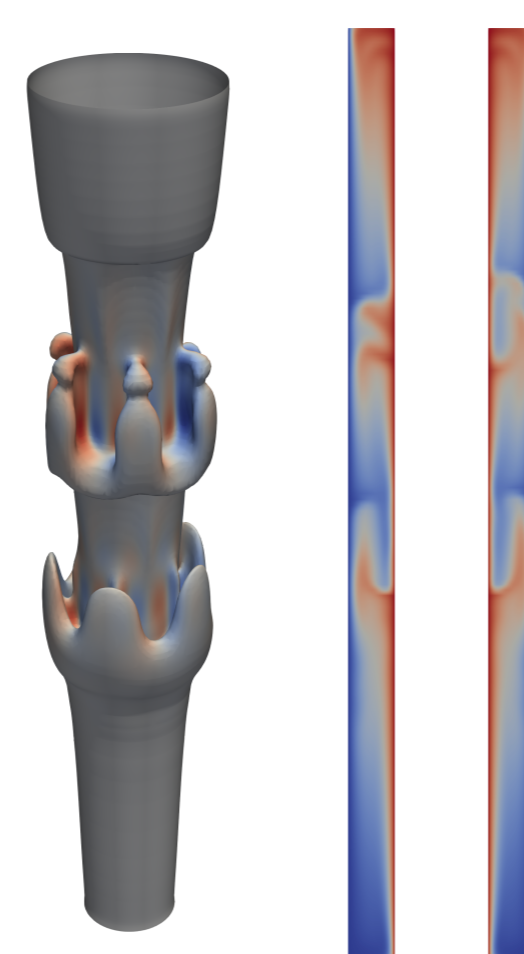
Parameters

$\nu$	$5 \cdot 10^{-6}$
$\kappa$	$7.74 \cdot 10^{-8}$
$\rho$	$9.23 \cdot 10^2$
$e$	$1.07 \cdot 10^{-3}$
$\alpha_f$	$1.4 \cdot 10^{-17}$
$\alpha_g$	$1.08 \cdot 10^{-3}$
$\delta\theta$	5 K
$V$	10 kV
$H$	100 mm
$r_i$	5 mm
$r_o$	10 mm
DOFs	$2.6 \cdot 10^7$

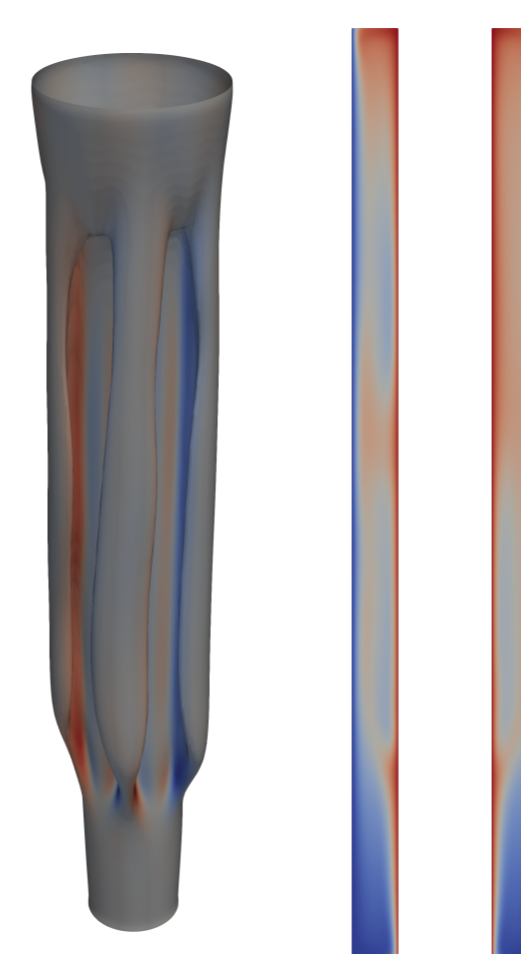
Axial symmetric phase (25s)



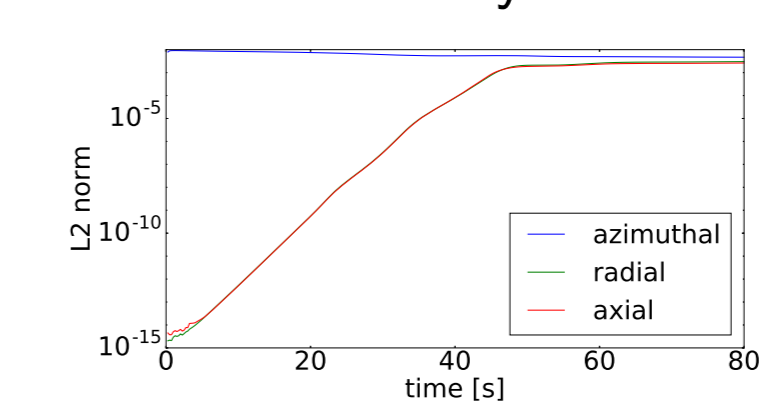
Transition phase (47.5s)



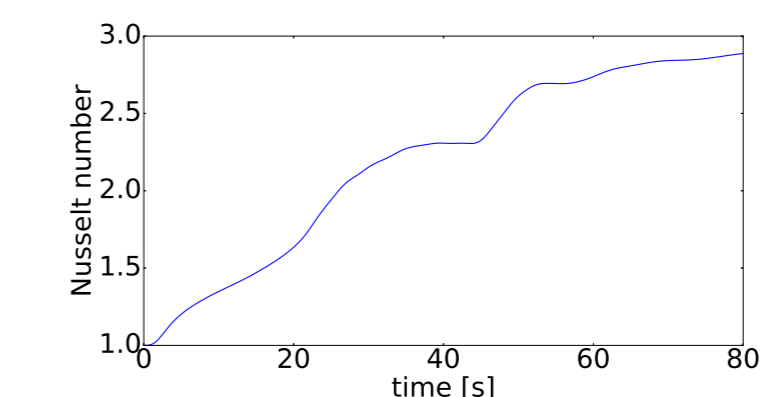
Stationary phase (80s)



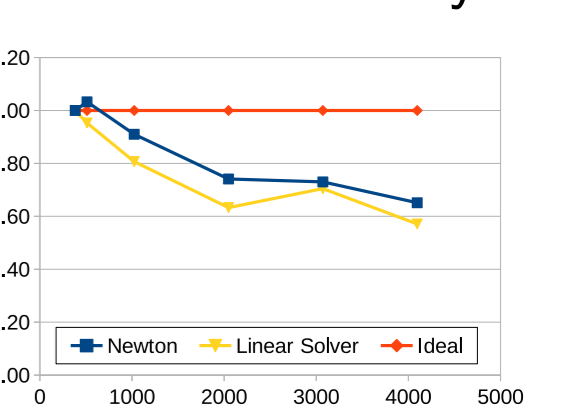
Vorticity



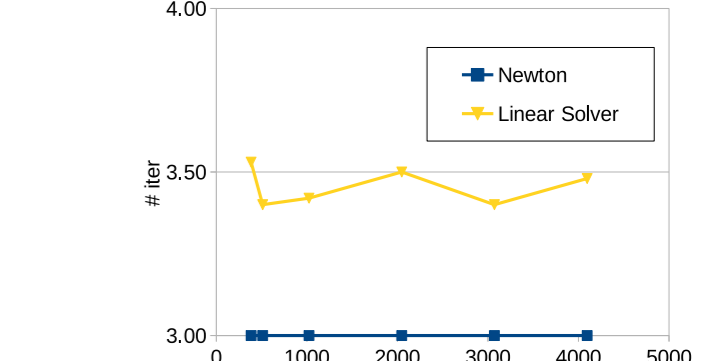
Heat transfer



Parallel efficiency



Average number of iterations



Figures: isosurface ( $\theta = 25^\circ\text{C}$ ) with axial vorticity,  $\theta$  on cutplane

## Uncertainty Quantification for Blood Pump Simulation

Three sources of uncertainty are considered:

- inflow boundary condition  $g(x)$ ,
- angular speed  $\omega$ ,
- dynamic viscosity  $\mu$ .

Three random variables are modeled as Uniform distribution:

$\xi_i \sim U(-1, 1), i = 1, 2, 3$ .

$$g(x) := g_0(x) + \sigma_1 g_0(x) \xi_1,$$

$$\omega := \omega_0 + \sigma_2 \omega_0 \xi_2,$$

$$\mu := \mu_0 + \sigma_3 \mu_0 \xi_3.$$

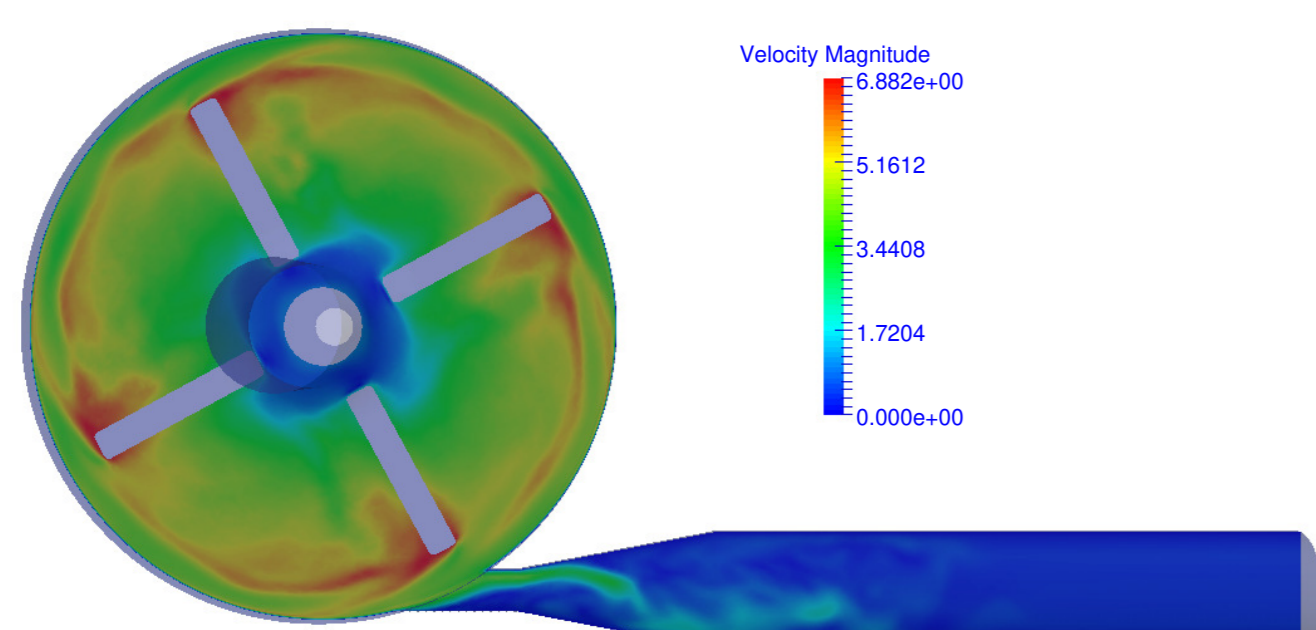
$$\left( \frac{u_{h,k}}{\partial t}, v_h \right) + \sum_{i=0}^P \sum_{j=0}^P ((\hat{u}_{h,i} - u_{h,i}^r) \cdot \nabla u_{h,j}, v_h) C_{ijk} + \sum_{i=0}^P \sum_{j=0}^P \frac{\mu_i}{\rho} (\nabla u_{h,j}, \nabla v_h) C_{ijk} - \frac{1}{\rho} (p_{h,k}, \nabla \cdot v_h) - (f_{h,k}, v_h)$$

$$+ (\tau_M \frac{\partial u_{h,k}}{\partial t} + \sum_{i=0}^P \sum_{j=0}^P (u_{h,i} - u_{h,i}^r) \cdot \nabla u_{h,j} C_{ijk} - \sum_{i=0}^P \sum_{j=0}^P \frac{\mu_i}{\rho} \Delta u_{h,j} C_{ijk} + \frac{1}{\rho} p_{h,k} - f_{h,k})$$

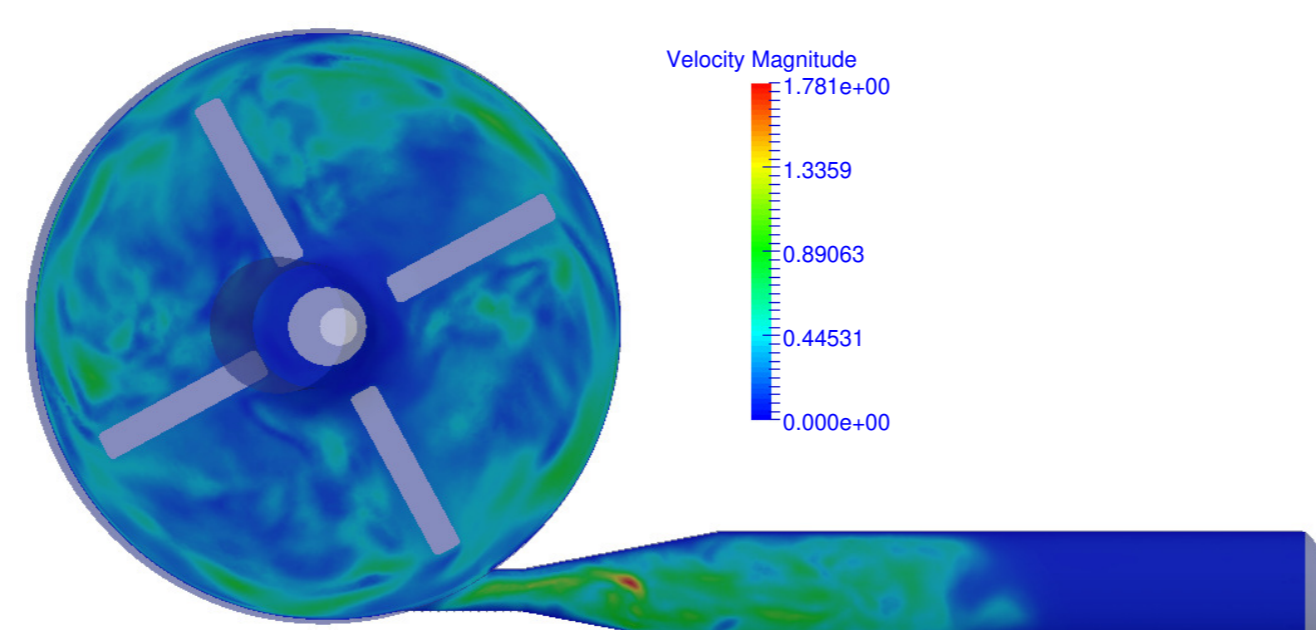
$$, (\hat{u}_{h,0} - u_{h,0}^r) \cdot \nabla v_h + (\tau_C \nabla \cdot u_{h,k}, \nabla \cdot v_h) = 0,$$

$$\left( \frac{1}{\rho} \nabla u_{h,k}, q_h \right) + (\tau_M \frac{\partial u_{h,k}}{\partial t} + \sum_{i=0}^P \sum_{j=0}^P (u_{h,i} - u_{h,i}^r) \cdot \nabla u_{h,j} C_{ijk} - \sum_{i=0}^P \sum_{j=0}^P \frac{\mu_i}{\rho} \Delta u_{h,j} C_{ijk} + \frac{1}{\rho} \nabla p_{h,k} - f_{h,k}), \nabla q_h = 0,$$

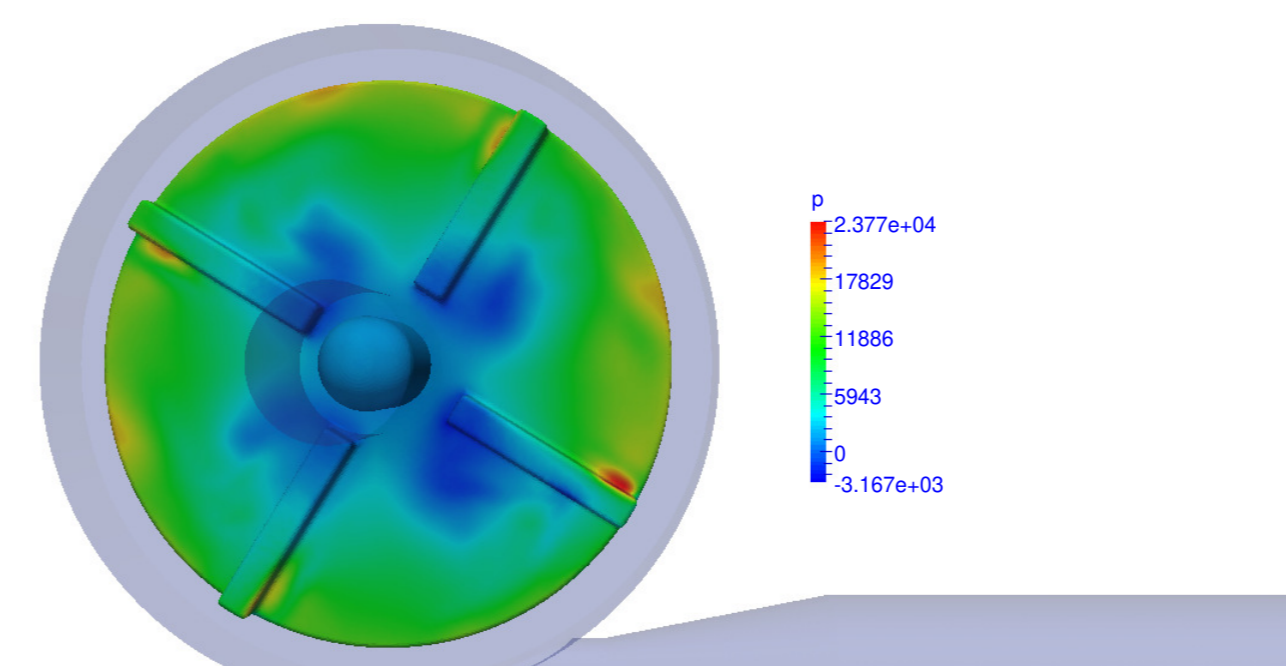
$$\forall v_h, q_h$$



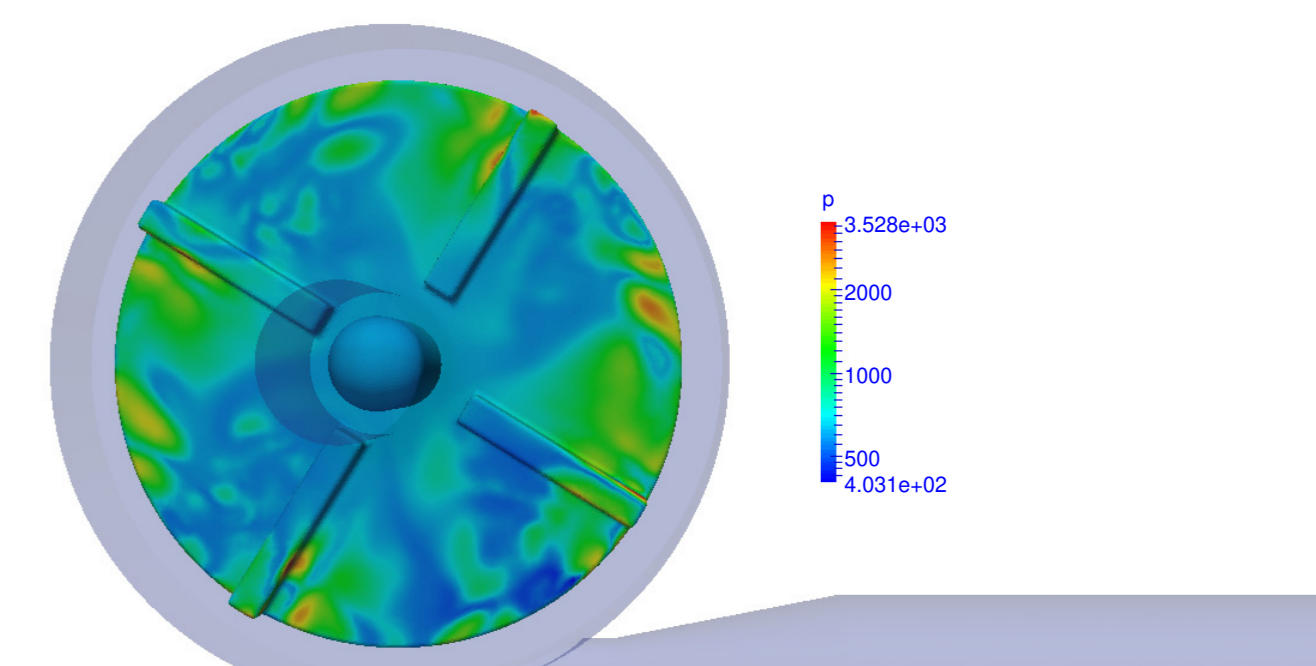
Mean value of velocity



Standard deviation of velocity



Mean value of pressure



Standard deviation of pressure

## Acknowledgements

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