



$$-i\beta\omega\left(\mathsf{T}-\frac{\gamma-1}{\gamma\alpha}\mathsf{P}\right)-\beta\ell_{\mathsf{h}}\mathsf{c}\Delta\mathsf{T}=\mathsf{S},\\-\gamma(\omega^{2}-i\ell_{\mathsf{v}}\mathsf{c}\omega\Delta)(\mathsf{P}-\alpha\mathsf{T})-\mathsf{c}^{2}\Delta\mathsf{P}=$$

- **Theorem :** Optimal  $L^2$  and  $H^1$  error estimates.

# **Block Preconditioners of Trace Gas Sensor Models**

# Automating the Finite Element Method

- We employ the Python interface to FEniCS:
- Construct or import a mesh (Dolfin).
- Efficient assembly of global system matrices (FFC).
- Calculate finite element solution (NumPy, SciPy, PyAMG).

# Linear System

The system	$A_h u = f$	f becomes,
------------	-------------	------------

(	$a_1K$	$a_2M$	0	$-a_3M$	
	$-a_2M$	$a_1K$	$a_3M$	0	
	$-a_4K$	$-a_3M$	$a_5K$	-H	
	a <sub>3</sub> M	$-a_4K$	Н	$a_5K$	

Where K and M are the stiffness and mass matrices, respectively, and  $\mathbf{H} = \mathbf{K} - \kappa^2 \mathbf{M}$  is the Helmholtz matrix.

# **Classical Preconditioning Techniques**

Ν	GMRES	ILU(3)	Block Jacobi	Block Gauss-Seidel
32	1156	12	269	141
64	4320	14	291	154
128	10000 +	44	257	143
256		235	245	134
512		429	231	127

Table: Iteration counts for the unpreconditioned GMRES method and GMRES coupled with the block Jacobi, block Gauss Seidel and Incomplete LU factorization preconditioners using realistic physical data.

# **Custom Block Preconditioner**

The effects of fluid viscosity and thermal conduction described by the characteristic lengths  $\ell_v$  and  $\ell_h$ , respectively, are particularly small. Ignoring terms containing these values suggests the following block preconditioner.

$$P = \begin{pmatrix} 0 & a_{2}M & 0 & -a_{3}M \\ -a_{2}M & 0 & a_{3}M & 0 \\ 0 & -a_{3}M & 0 & -H \\ a_{3}M & 0 & H & 0 \end{pmatrix}.$$
 (4)  
Clustering :  $\lambda(P^{-1}A_{h}) = 1 + Ch^{-2}$ , but

• Eigenvalue  $C = O(10^{-10}).$ 





$$\begin{pmatrix} \mathbf{T}_1 \\ \mathbf{T}_2 \\ \mathbf{P}_1 \\ \mathbf{P}_2 \end{pmatrix} = \begin{pmatrix} \mathbf{S} \\ \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \end{pmatrix}$$
(3)

### **Complex Shifted Laplacian**

Inverting **P** can be reduced to solving a block upper triangular system where we must solve the two Helmholtz sub-problems. For this we apply the now common complex shifted Laplacian preconditioner.  $\mathsf{P}_{\mathsf{H}} = \mathsf{K} - (\alpha + \mathsf{i}\beta)\kappa^{2}\mathsf{M}$ (5)

where lpha and eta are real numbers chosen such that multigrid methods are known to behave better on  $P_H$  than on H itself.

### **Numerical Results**



Figure: Preconditioned residual norms at the **k**<sup>th</sup> GMRES iteration for the 2D problem on various  $\mathbf{N} \times \mathbf{N}$  meshes.

### **Future Work**

- boundary.

### References

- Waves in a Trace Gas Sensor," PyHPC 2013.
- J. Scientific Computing.



Test with HPC parallel libraries such as Trilinos or PETSc. Experiment with alternative methods for Helmholtz problems. Couple current problem with an elasticity model for tuning fork

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