

COMPUTATIONAL
RESEARCH
DIVISION

Exploring Post-Moore Microelectronics with HPC

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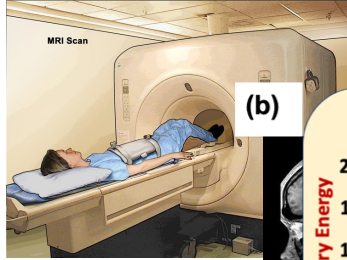
jackie_zhiyao@lbl.gov

Summer Program 2022
Jun 30, 2022

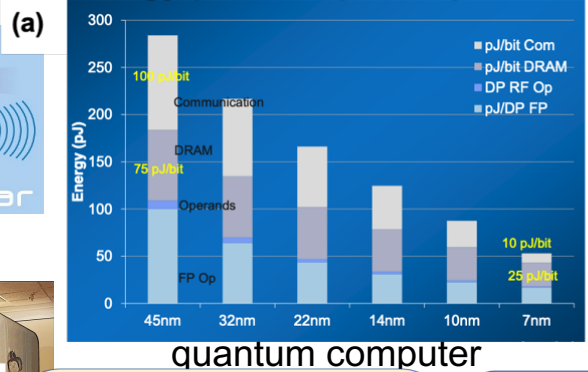
- Continuous miniaturization and integration of modern electronics drives people to utilize novel materials and new techniques



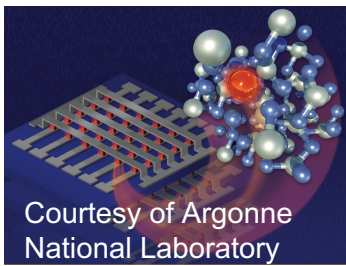
autonomous car



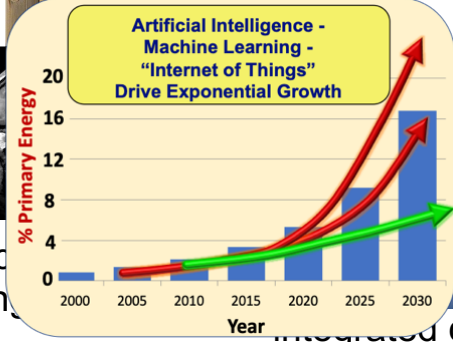
portable magnetic resonance imaging



quantum computer



spintronic memory for computing applications

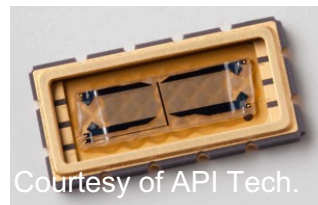


Artificial Intelligence - Machine Learning - "Internet of Things" Drive Exponential Growth

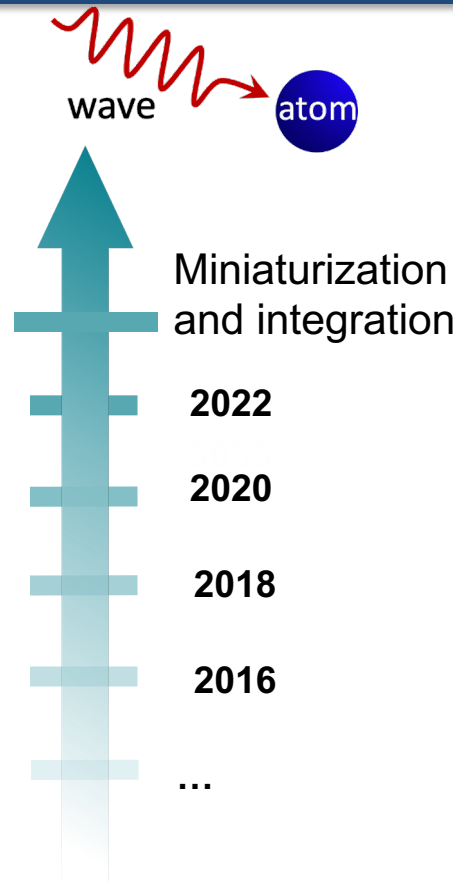
Energy Efficiency
 ~ 100e⁻¹² J/ op
 ~25 % primary energy

Beyond CMOS @ 20e⁻¹⁵ J/ op

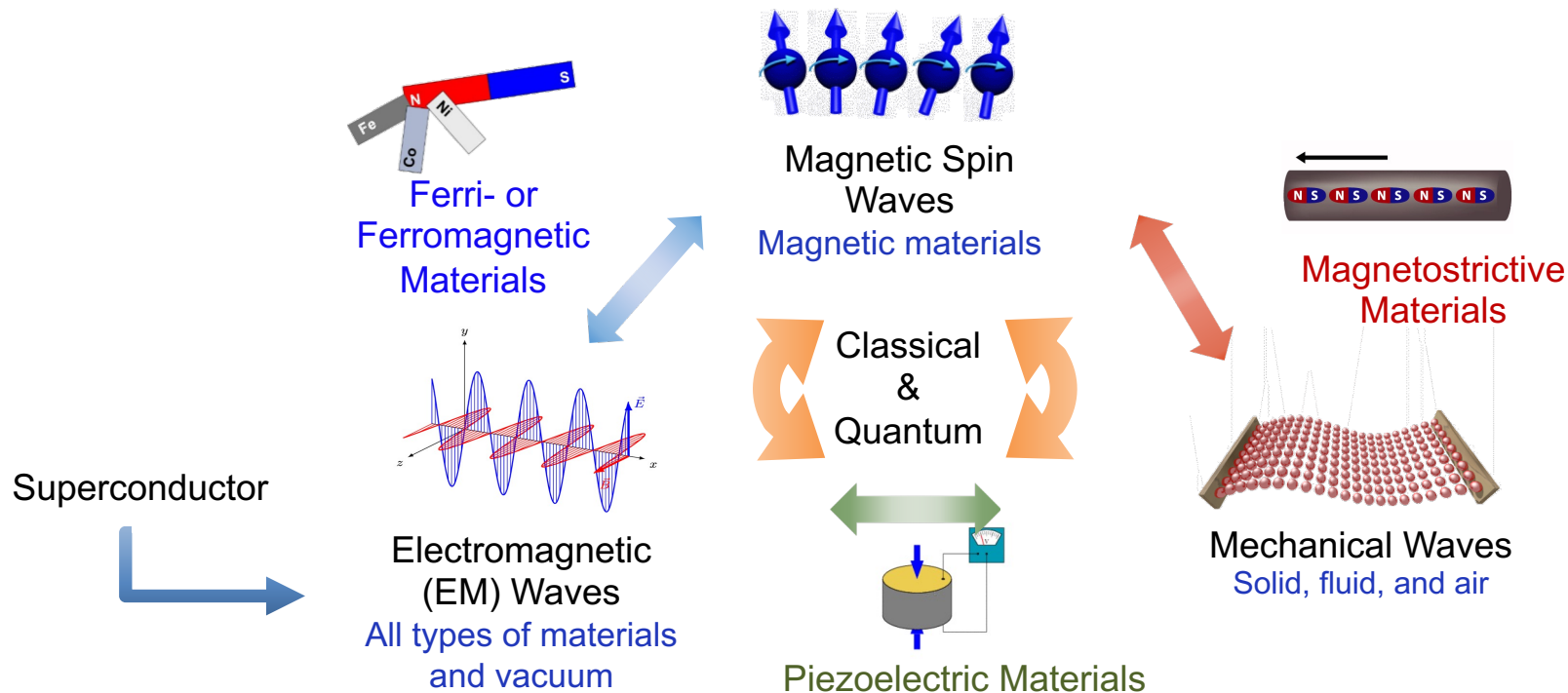
Beyond CMOS @ 1e⁻¹⁸ J/ op

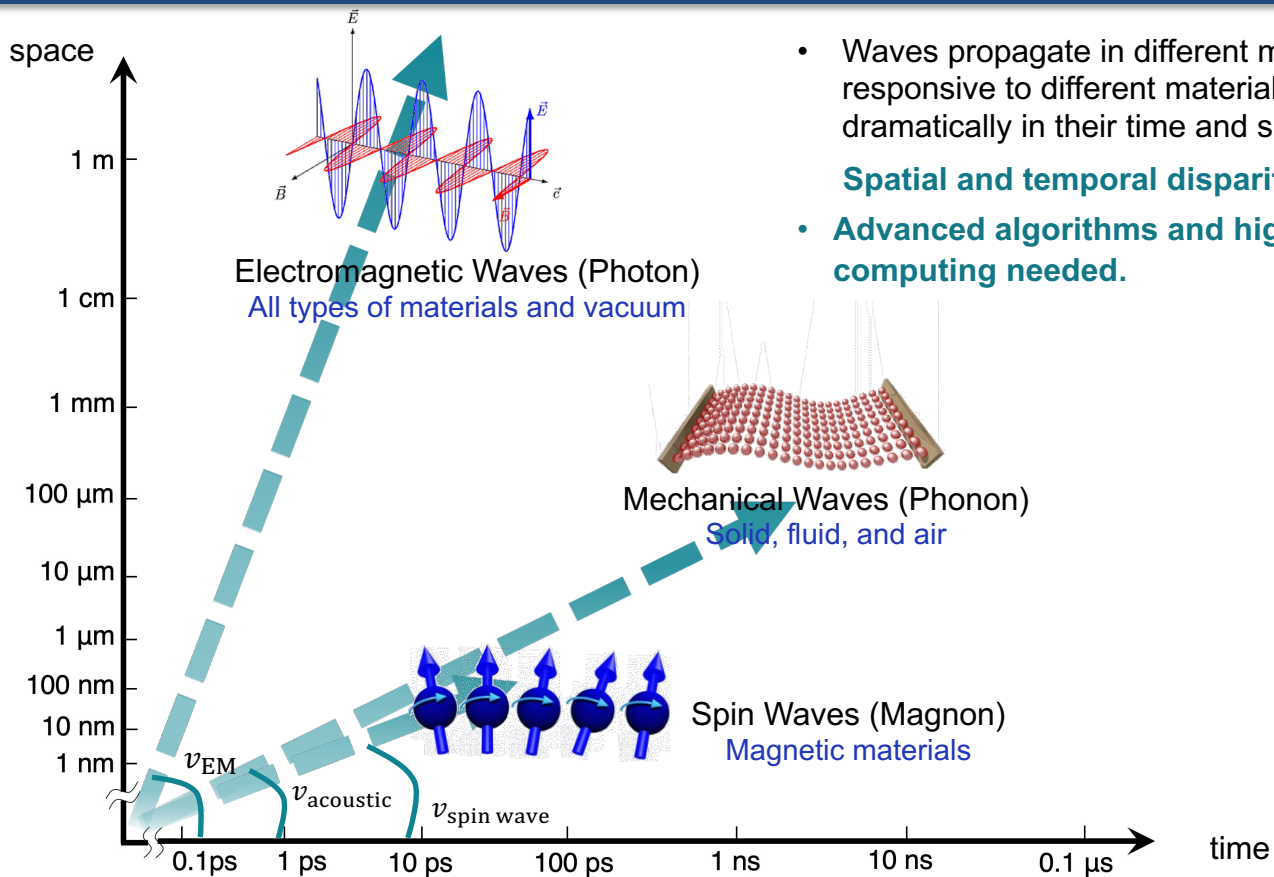


acoustic filter



- Complex physical coupling: Any wave interacting with any material, both classical and quantum
- Currently focused on EM and magnetization, expanding into superconductors and quantum applications





- Waves propagate in different media, are responsive to different materials, and differ dramatically in their time and space dimensions:

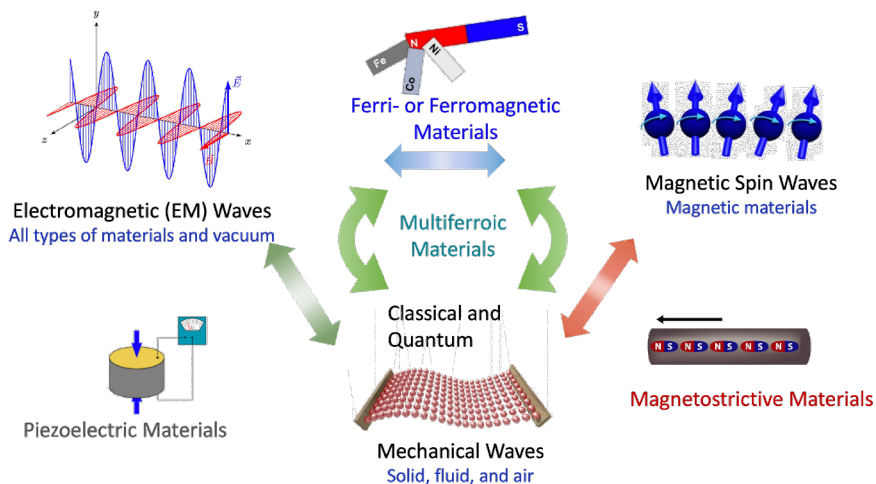
Spatial and temporal disparity

- Advanced algorithms and high-performance computing needed.

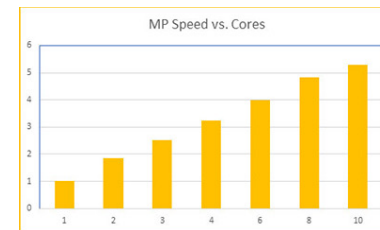
*Axes are not true to scale

Current model limitations:

- Use the same single-physics model ; multiple physical coupling algorithm is not clear
- Treat sub-components as simplified black-boxes
- Typically limited to 1 level of magnification
- Proprietary commercial software (little to no customization of algorithm)
- Poor scaling & no GPU

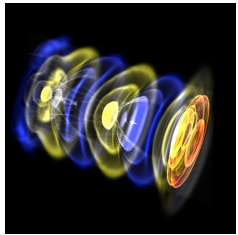


- Maximum scaling 2 nodes, 12 cores/node
- cannot use GPUs

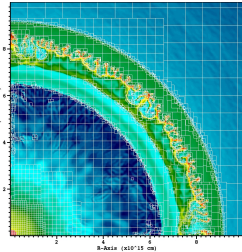


<https://www.ansys.com/blog/how-to-optimize-speed-scalability-ansys-hfss-hpc>

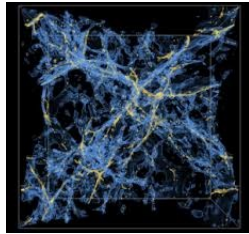
Exascale Computing application projects that partner with AMReX:



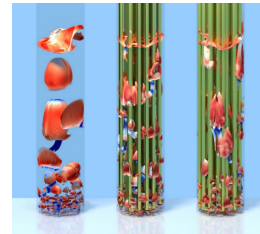
Accelerators



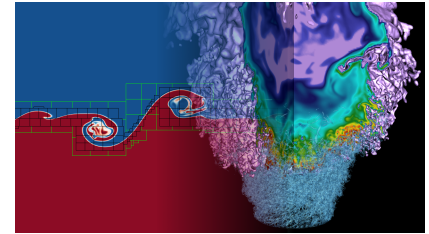
Astrophysics



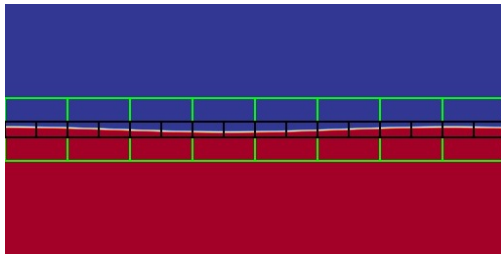
Cosmology



Multiphase flow

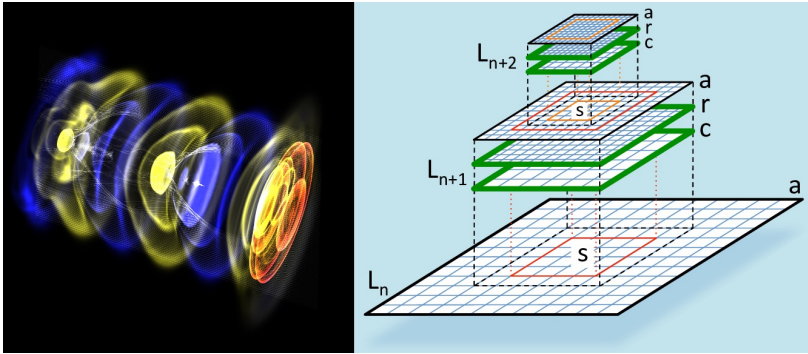


Combustion



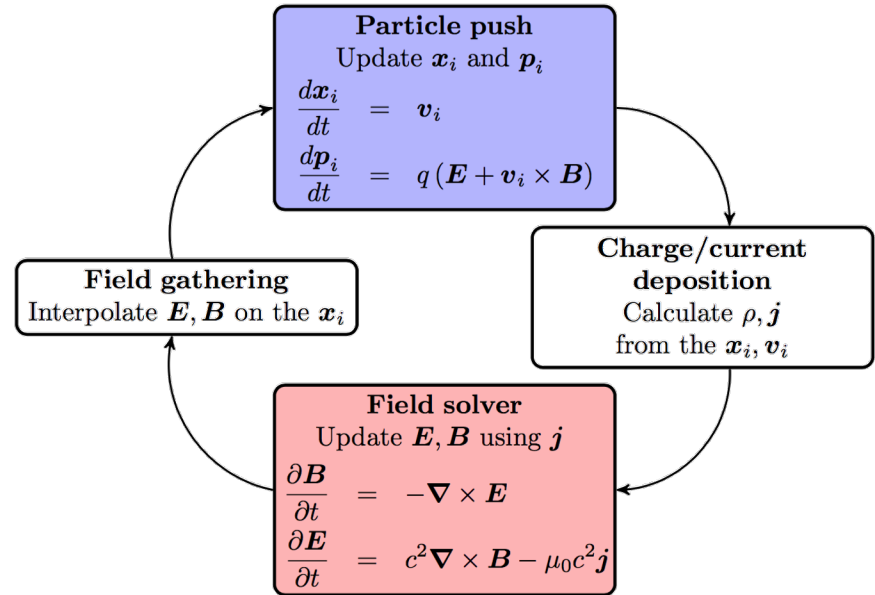
- AMReX is a block-structured Adaptive Mesh Refinement (AMR) framework for solving systems of nonlinear PDEs for a variety of US DOE applications.
 - DOE Exascale Computing Project (ECP) Co-Design Center
 - Performant on full HPC systems (multicore/GPU)
- Provides support for multiphysics modeling for time-dependent PDEs
 - Explicit & implicit mesh operations
 - Multilevel synchronization operations
 - Particle and particle/mesh algorithms
 - Solution of parabolic and elliptic systems using geometric multigrid
 - Embedded boundary (cut-cell) representation of geometry

- WarpX: Accelerator Division ECP application code for accelerators coupling particle-in-cell (PIC) electrons with Maxwell

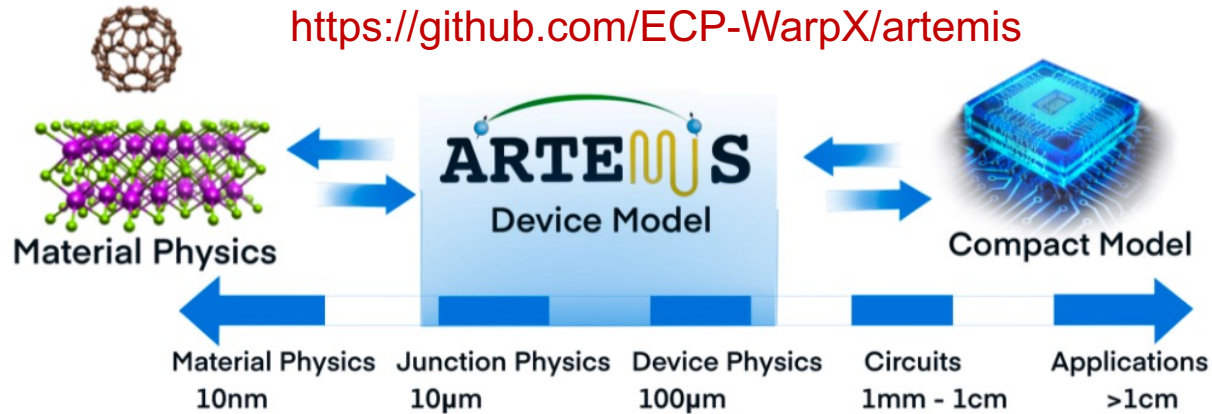


Advanced features:

- Spectral Maxwell solver : High accuracy
- Perfectly-Matched Layers : Wave absorption
- Dynamic load-balancing : Efficient
- Mesh refinement : multi-scale sans artifacts
- Multi-physics: ionization

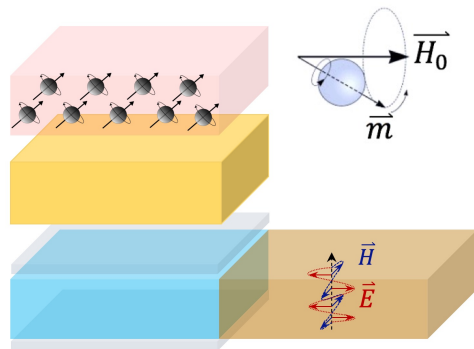


+ **communications between sub-domains:**
guard cell and particle exchanges handled using AMReX



- ARTEMIS bridges the gap between material physics and circuit model of PARADISE, by solving governing PDEs of the physics in devices such as MESO and NCFET.
- Based on two ECP codes, AMReX and WarpX
- Massively parallel, CPU and GPU support

Z. Yao, R. Jambunathan, Y. Zeng, and A. Nonaka, Int. J. High Perform. Comput. Appl., p. 10943420211057906, Jan. 2022.



Maxwell's equation

Landau-Lifshitz-Gilbert (LLG) equation

$$\frac{\partial \mathbf{M}}{\partial t} = \mu_0 \gamma (\mathbf{M} \times \mathbf{H}_{\text{eff}}) - \frac{\alpha}{|\mathbf{M}|} \mathbf{M} \times \frac{\partial \mathbf{M}}{\partial t}$$



$$H_{\text{eff}} = H + H_{\text{bias}} + H_{\text{ani}} + H_{\text{exch}} \dots$$

$$\nabla \times \mathbf{E} = -\frac{\partial \mathbf{B}}{\partial t} \quad \nabla \times \mathbf{H} = \mathbf{J} + \frac{\partial \mathbf{D}}{\partial t}$$

$$\nabla \cdot \mathbf{D} = \rho \quad \nabla \cdot \mathbf{B} = 0$$

- Algorithmically couple PDEs (LLG equation and Maxwell's equations)
- Solution procedure will evolve this coupling at each time step

- \mathbf{M} : volume density of electron spins
- Continuum model that describes the evolution of \mathbf{M} under the effective \mathbf{H} and the torque

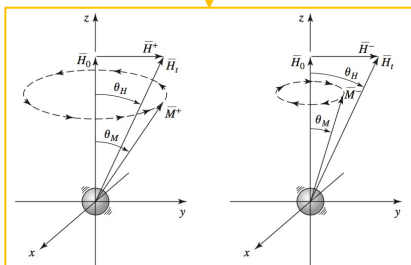
Magnetically Tunable Filter



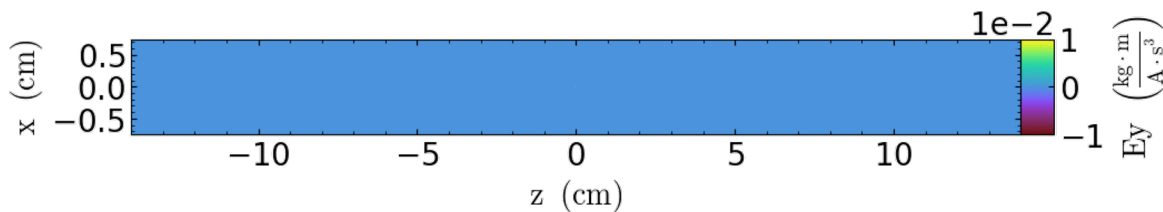
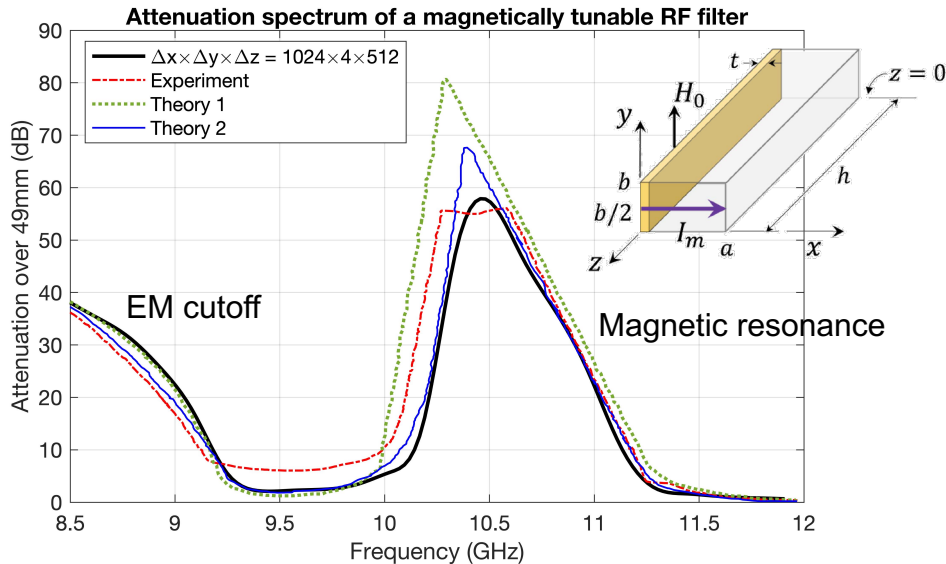
$$t = 0.45 \text{ mm}, a = 14.95 \text{ mm}$$

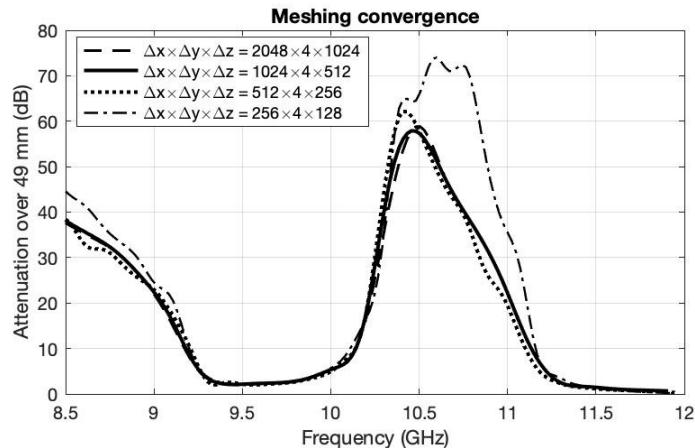
$$4\pi M_S = 1750 \text{ Gauss}$$

$$H_i = 2890 \text{ Oe}, \Delta H = 35 \text{ Oe}$$



- Demonstrated both EM propagation and the coupling between EM and other physical phenomena such as magnetic resonance.





- Convergence on the result is observed in the mesh refinement process
- Adaptive Mesh Refinement (AMR) implementation is under progress

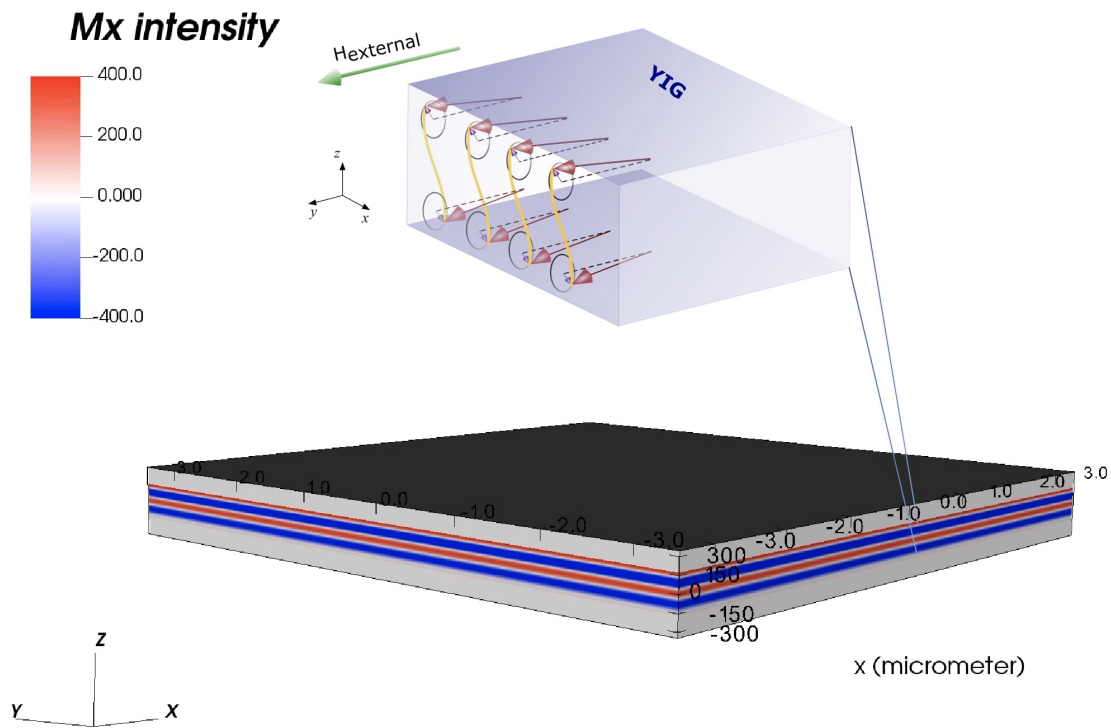
Variable	E_{32}^{64}	E_{64}^{128}	Rate
E_x	1819.4	455.23	2.00
E_y	1820.5	455.04	2.00
E_z	1882.8	470.47	2.00
H_x	9.6659	2.4359	1.99
H_y	9.4752	2.3788	1.99
H_z	9.4590	2.3607	2.00
M_x	5.1343	1.2870	2.00
M_y	4.7851	1.1989	2.00
M_z	4.9642	1.2429	2.00

Table 1: Convergence rates in the L^1 norm for all field variables.

- Second-order accuracy is observed for all field variables

$$L_1 \text{ norm calculation } E_c^f = \frac{1}{N_{\text{pts}}} \sum_{i,j,k} |\phi_f - \phi_c|$$

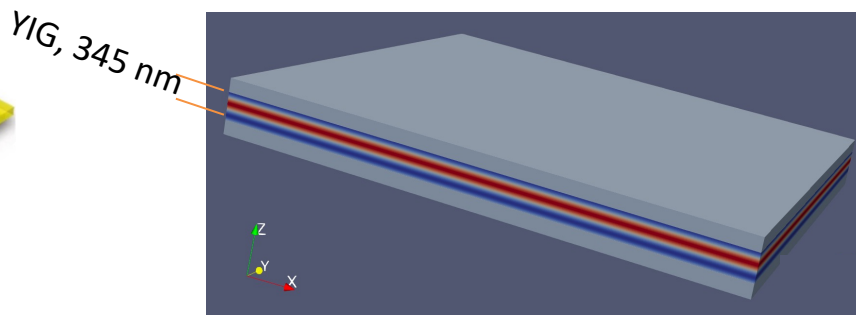
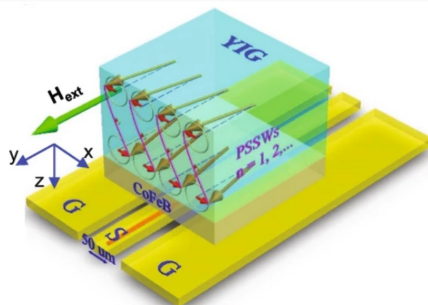
DB: movie.visit
Cycle: 180000 Time:1.61829e-09



$$H_{ex} = \frac{2A_{ex}}{\mu_0 M_S^2} \nabla^2 M$$

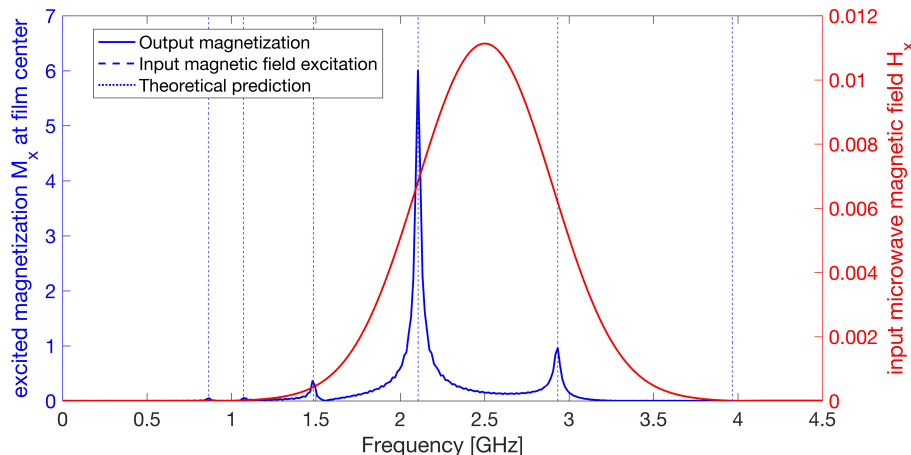
Boundary conditions at magnetic material interfaces:

- Pinned Boundary $M = 0$
- Free Boundary $\frac{\partial M}{\partial n} = 0$



Excitation of perpendicular standing spin waves (PSSW)

$$f_{PSSW} = \frac{\gamma \mu_0}{2\pi} \sqrt{\left[H_{ext} + M_S + \frac{2A_{ex}}{\mu_0 M_S} \left(\frac{(2p+1)\pi}{2d} \right)^2 \right] \times \left[H_{ext} + \frac{2A_{ex}}{\mu_0 M_S} \left(\frac{(2p+1)\pi}{2d} \right)^2 \right]}$$



H. Qin, S. J. Hämäläinen, and S. Van Dijken, *Sci. Rep.*, vol. 8, no. 1, pp. 1–9, 2018.

Resonance frequencies as theoretically predicted

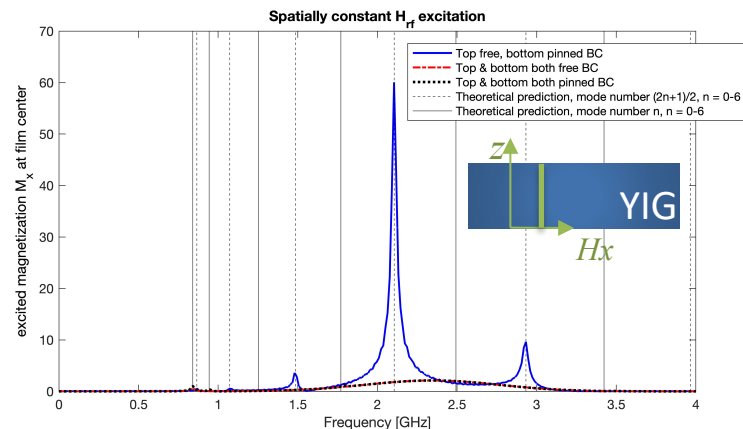
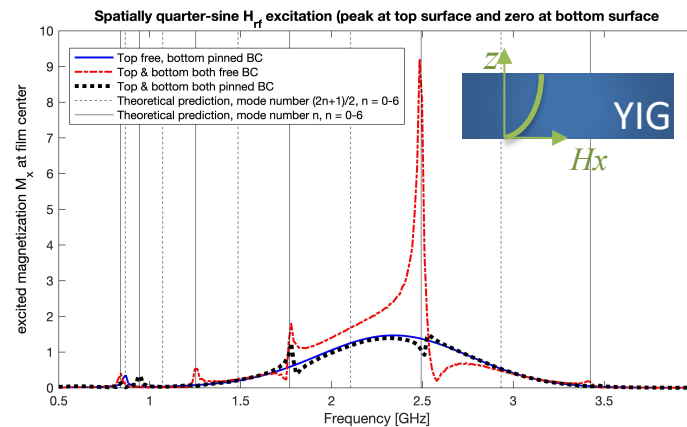
Odd mode:

$$f_{PSSW} = \frac{\gamma\mu_0}{2\pi} \sqrt{\left[H_{ext} + M_S + \frac{2A_{ex}}{\mu_0 M_S} \left(\frac{(2p+1)\pi}{2d} \right)^2 \right] \times \left[H_{ext} + \frac{2A_{ex}}{\mu_0 M_S} \left(\frac{(2p+1)\pi}{2d} \right)^2 \right]}$$

Even mode:

$$f_{PSSW} = \frac{\gamma\mu_0}{2\pi} \sqrt{\left[H_{ext} + M_S + \frac{2A_{ex}}{\mu_0 M_S} \left(\frac{p\pi}{d} \right)^2 \right] \times \left[H_{ext} + \frac{2A_{ex}}{\mu_0 M_S} \left(\frac{p\pi}{d} \right)^2 \right]}$$

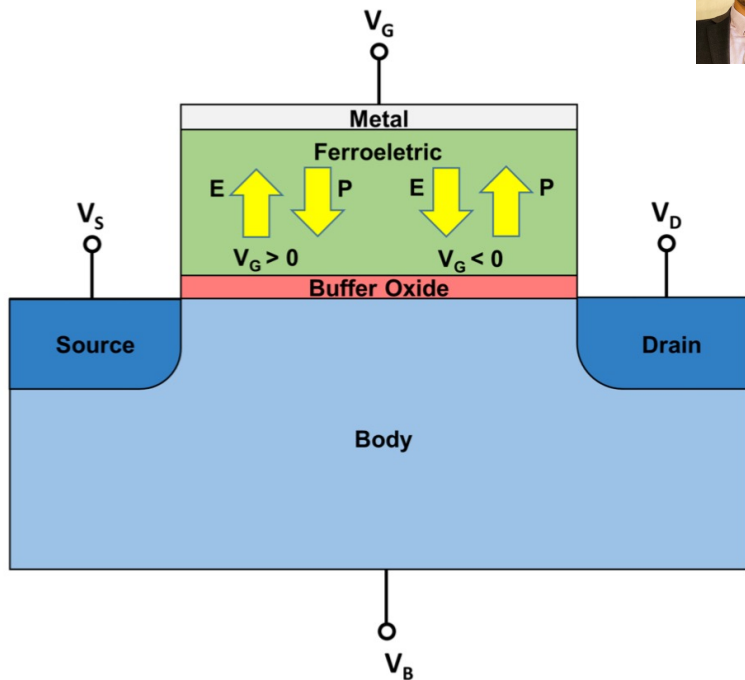
- Engineering EM field distribution in the coupled resonator leads to different magnetic resonance modes



Prabhat Kumar
(AMCR)



NC-FET structure

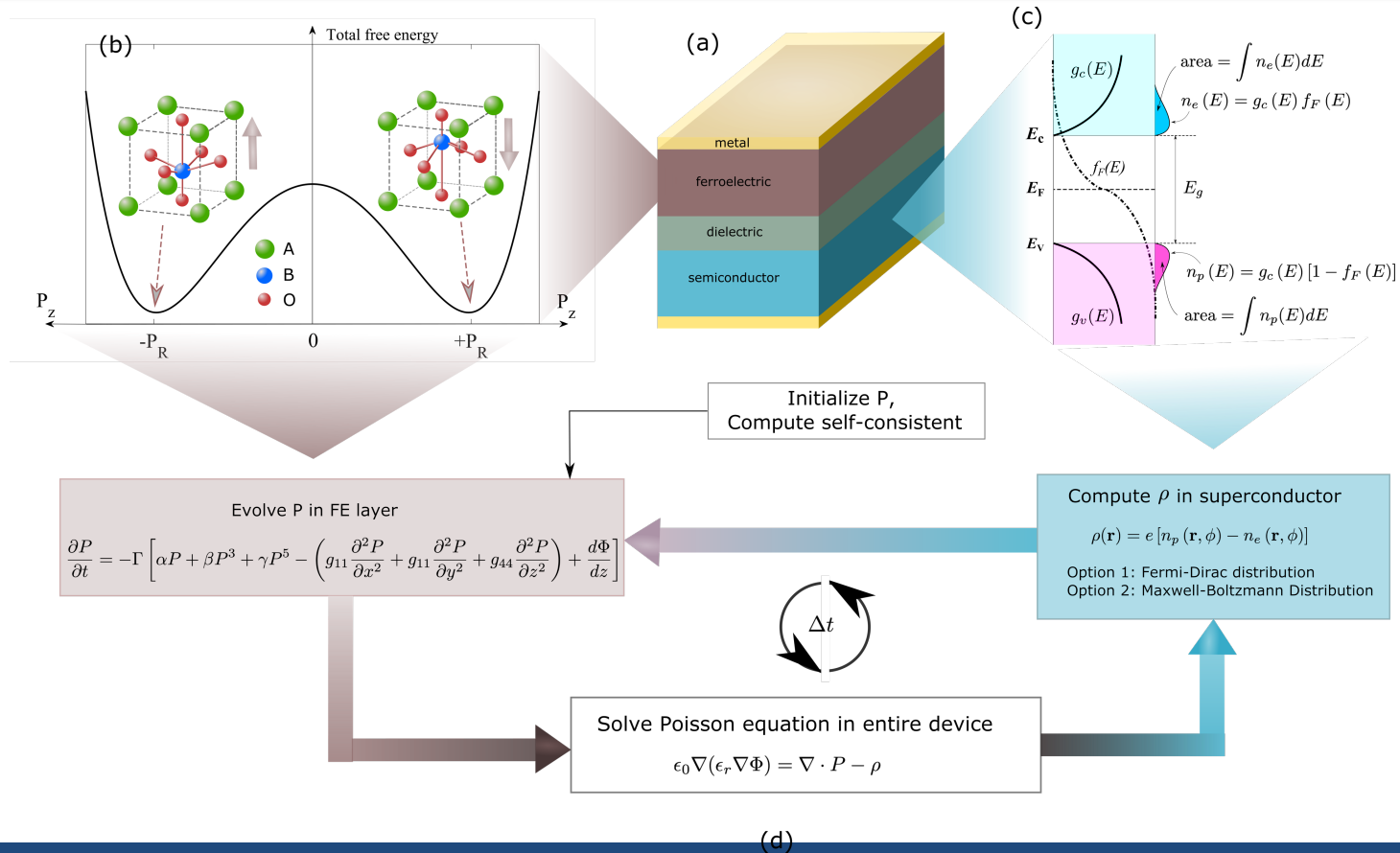


$$S = \frac{\partial V_G}{\partial \log_{10} I_D} = 60mV \left(1 + \frac{C_{dm}}{C_{ox}} \right)$$

- Negative capacitance effect of ferroelectrics can help breach the so-called Boltzmann Tyranny*
- Modeling challenges:
 - Ginzburg-Landau model for FE polarization
 - Poisson's equation across the gate stack
 - Semiconductor charge transport in channel
- Need Boltzmann transport model to capture short-channel effects

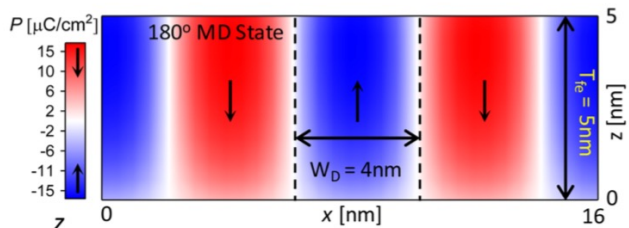
*Salahuddin and Datta, *Nano Lett.* 2008, 8, 2, 405–410

Our model solves self-consistent solutions of coupled ferroelectric systems at every timestep

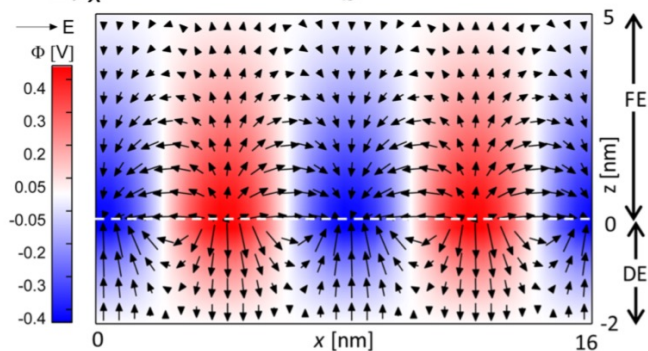


Existing 2D model (Saha & Gupta, Sci Rep 2020)

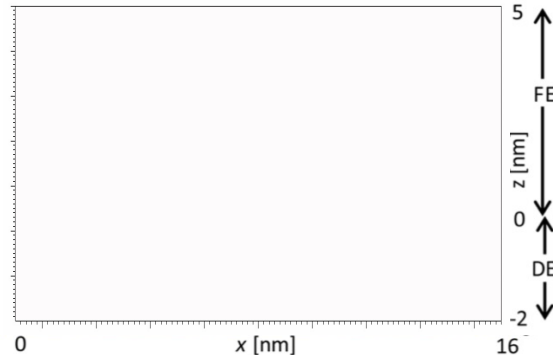
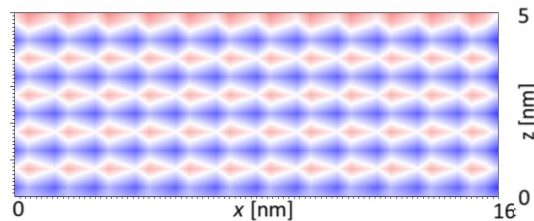
a



b

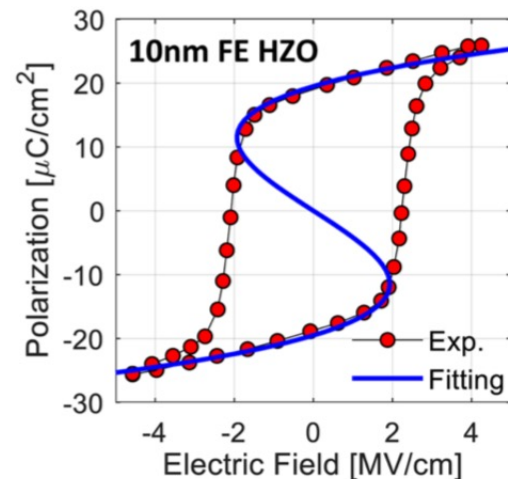


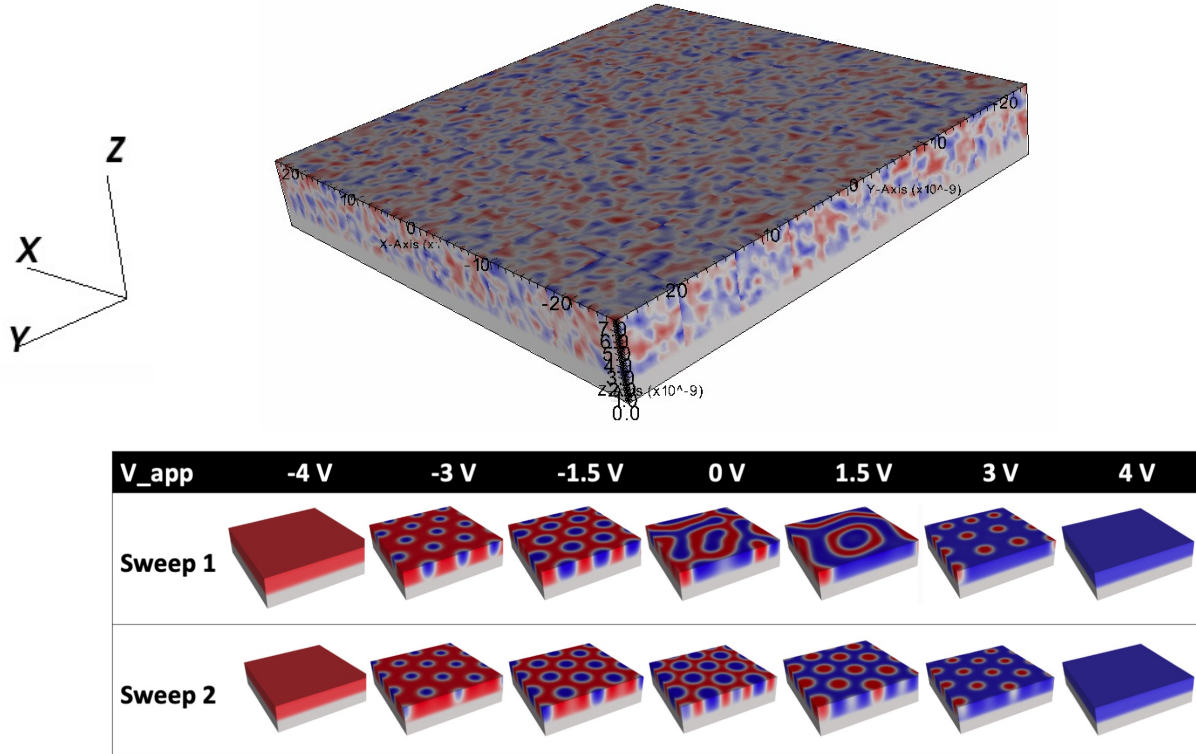
Results Using Our 3D model



Material Parameters

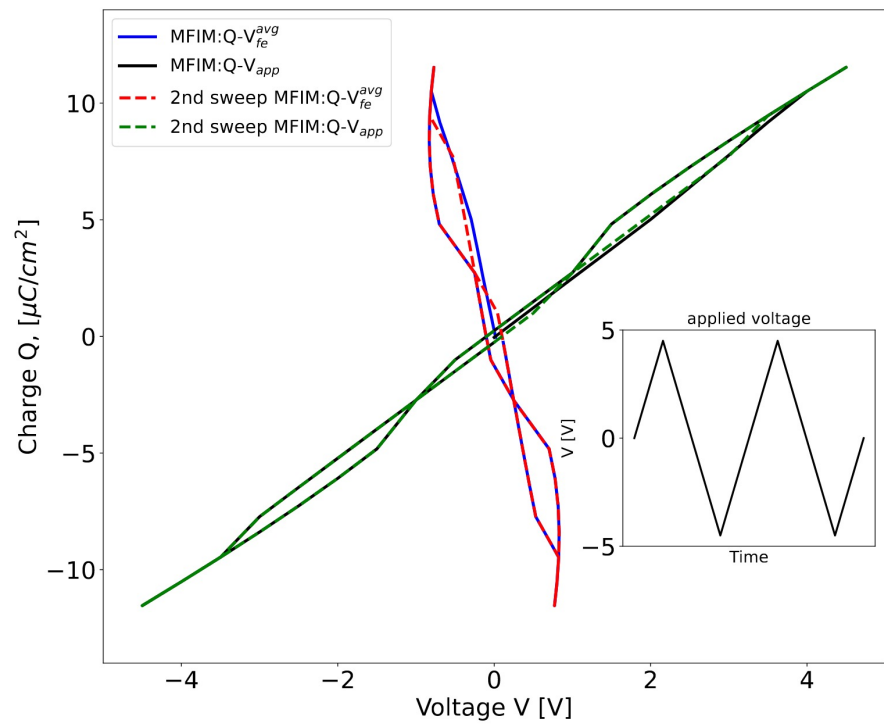
$\alpha, \beta, \gamma, g_{11}, g_{44}, \epsilon_x, \epsilon_y, \epsilon_z$





(b) Polarization switching dynamics

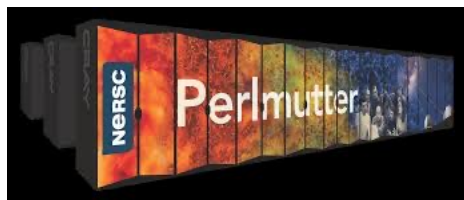
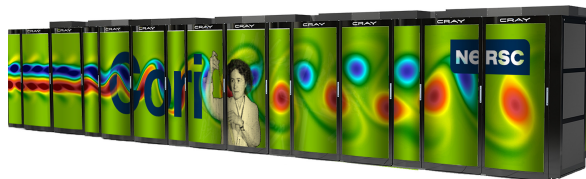
Prabhat Kumar
(AMCR)



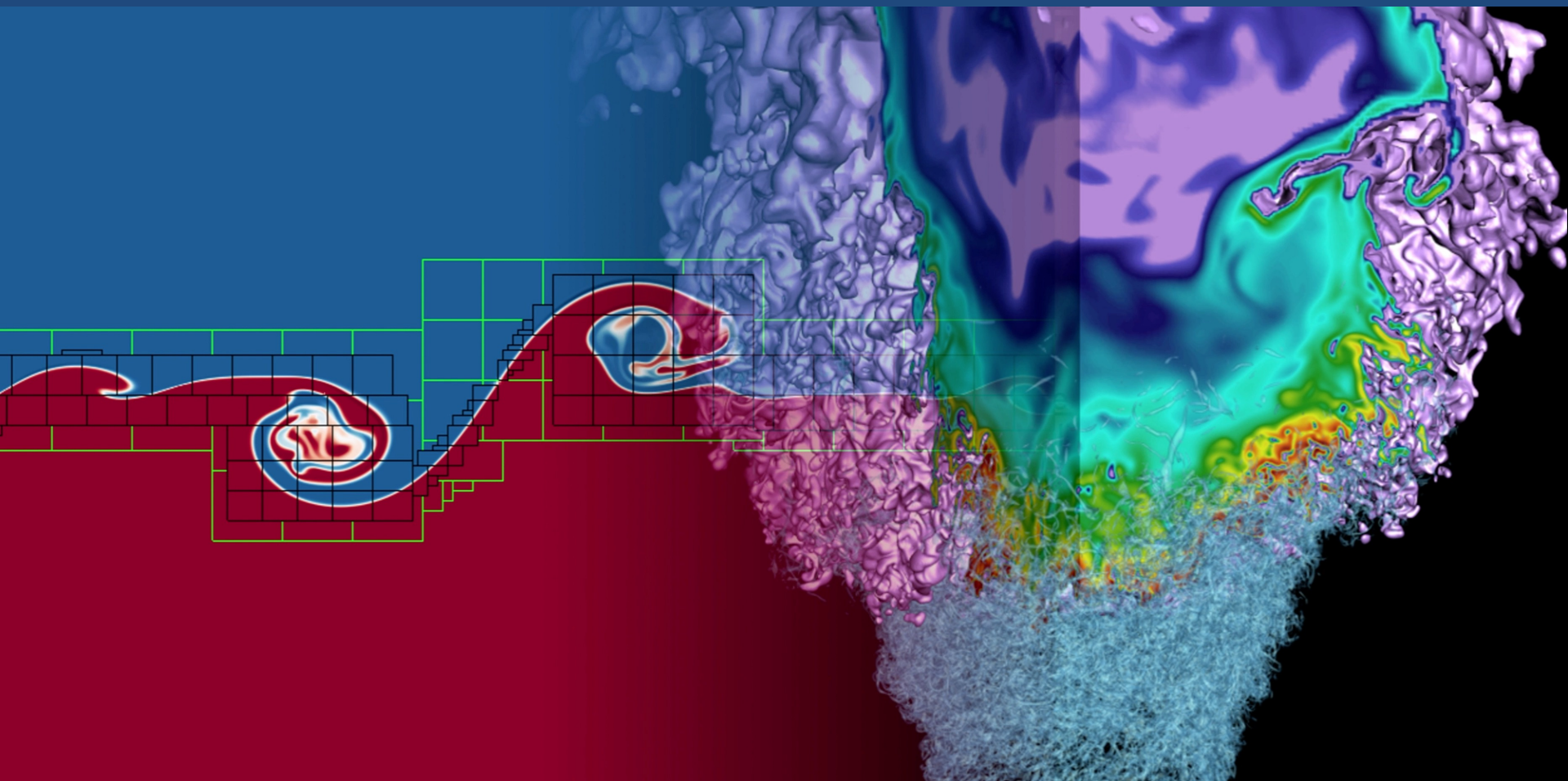
(a) Q-V Characteristics

- The slope of Q-V curve indicates the value of the capacitance
- Hysteresis relation observed, indicating the memory feature of ferroelectric materials
- Except for the 1st sweep, repeated sweeps of voltage can generate identical initial states, and thus resulting in repeated Q-V curves.

- Portable to different platforms
laptops – leadership class multicore/GPU
supercomputers



- Flexible to add “improvised”
physical algorithms
- Adaptive mesh refinement
(AMR) to increase the
scale-disparity



U.S. DEPARTMENT OF
ENERGY

Office of
Science

Motivation

Challenges

Approach

Validations

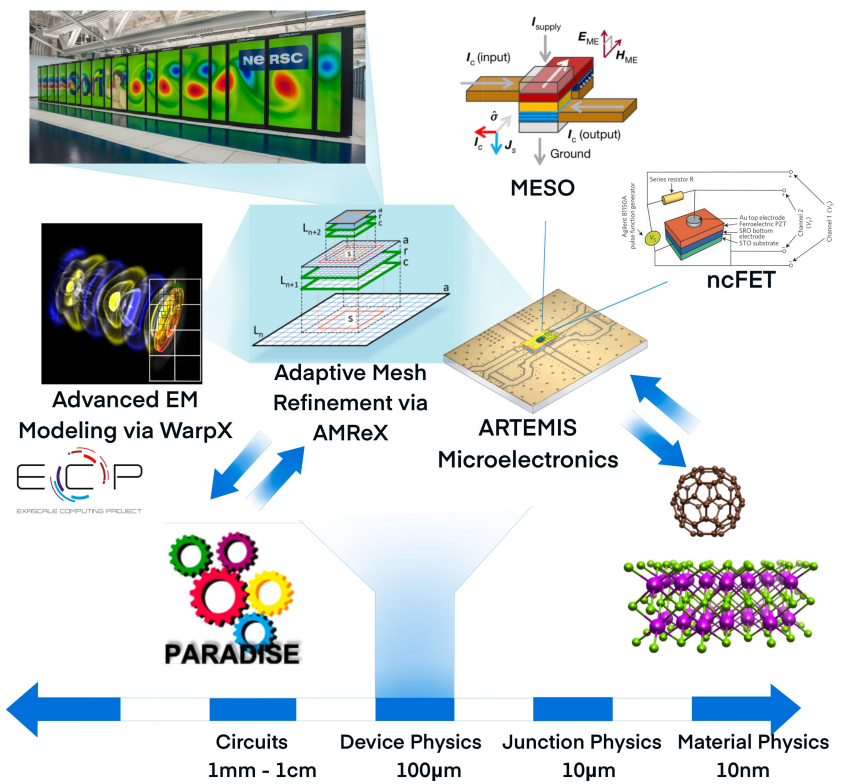
Work in Progress



BERKELEY LAB

Future Works

1. Couple (anti)ferromagnetic and ferroelectric phases for the design of magnetoelectric devices
2. Explore device input/output characteristics for larger-scale circuit design



Thanks to --- AMReX + WarpX + ARTEMIS Team

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Rémi Lehe (ATAP)

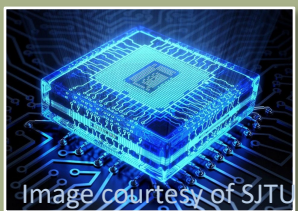
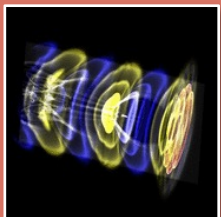


Image courtesy of SJTU

ARTEMIS (Adaptive mesh Refinement Time-domain Electrodynamic Solver)

Revathi Jambunathan (AMCR)



Prabhat Kumar (AMCR)



Saurabh Sawant (AMCR)



Yadong Zeng (U. of Minnesota)



Maurice Garcia-Sciveres (Physics, LBL)



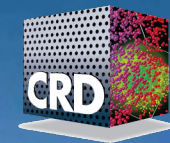
Sinéad Griffin (Molecular Foundry, LBL)



DOE Micro-electronics Co-design Program

Quantum chip design

Thank You !



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Acknowledgment:

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Peter Nugent, CRD Computational Science

Weiqun Zhang, CRD CCSE

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Rémi Lehe, ATAP

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