

Welcome!

The program will begin soon.
You will not hear audio until we begin.

A LEAP IN SCIENTIFIC COMPUTING FOR RESERVOIR SIMULATION

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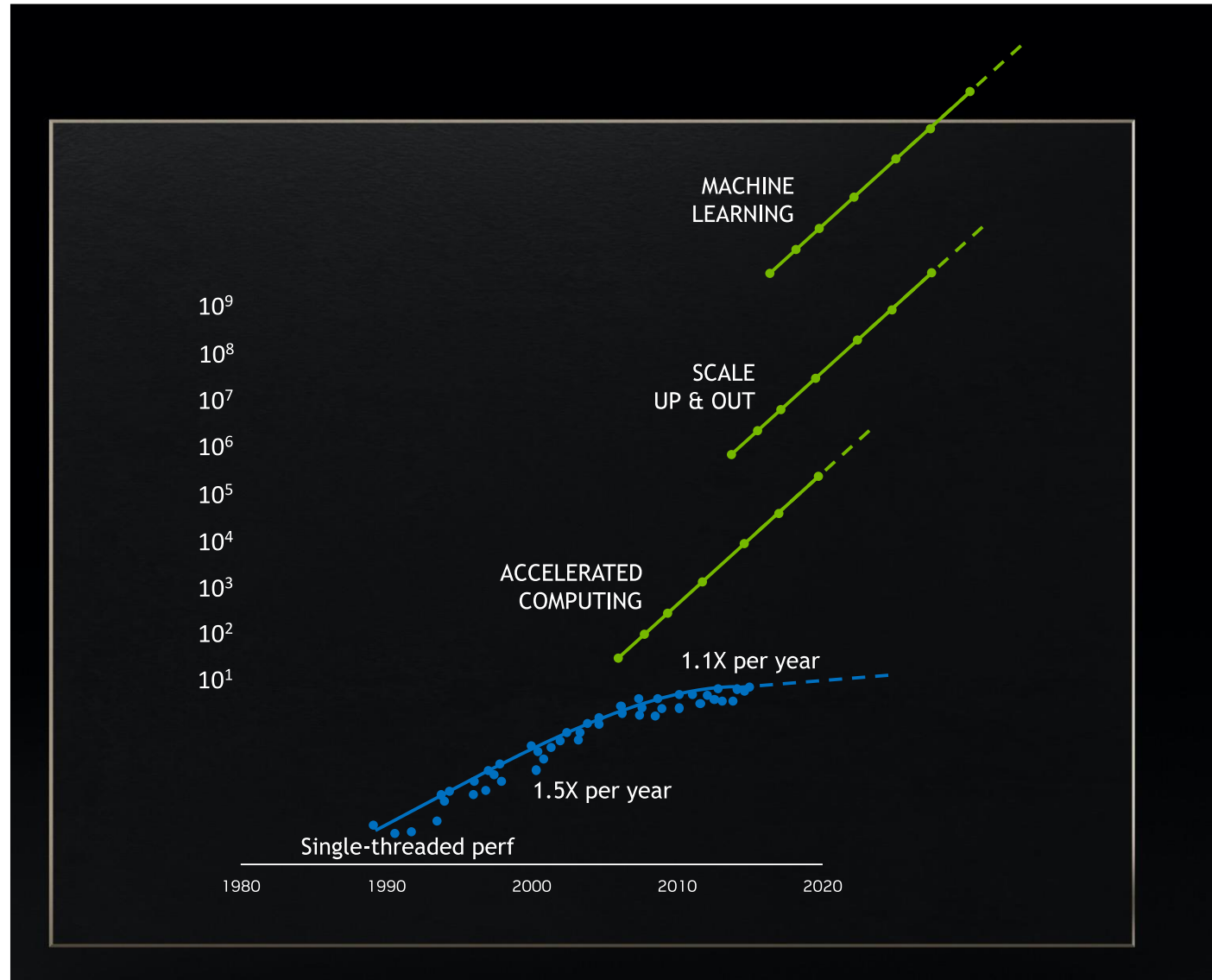
A Leap in Scientific Computing for Reservoir Simulation



- Many current reservoir simulation solutions
 - Computationally, resource, and time expensive
- High performance computing solution
 - GPU-accelerated reservoir simulation
- Machine learning solutions
 - What's next for reservoir simulation?



Advances in High-Performance Computing and Machine Learning



GPU-Accelerated Reservoir Simulation

Challenges of Traditional Reservoir Simulation

- Time and computationally expensive
- Reduced workforce that is required to produce more results with fewer resources (time, people)

Current Industry Solutions

- Company A built reservoir simulation software on GPUs from the ground up
- Company B porting reservoir simulation software to be GPU-accelerated

Performance Benchmarks

- Company A: 11.2-million cell model with 9 components forecast for 20 years:
 - 30 minutes on 8 GPUs; 45 minutes on 4 GPUs
- Company B black oil:
 - 2.3-million cell model forecast for 30 years: 4.5x performance increase; 7.5x throughput with MPS; 5.7x throughput with MIG
 - 600-thousand cell model: 3.7x performance increase; 11x throughput with MPS; 9x throughput with MIG

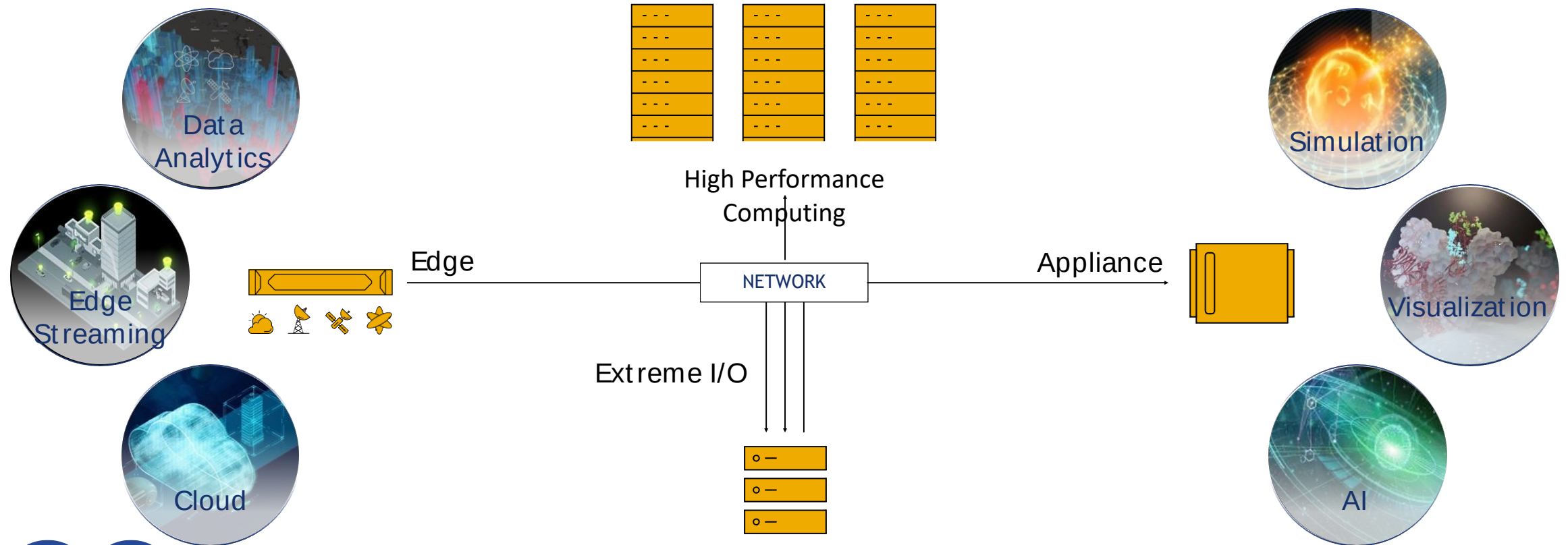
GPU-Accelerated Reservoir Simulation



- Benefits
 - Significant acceleration
 - GPU technology allows for several users per GPU or ensemble modeling
- Challenges
 - Many solutions currently single GPU, user limited to memory of the GPU used
 - Multi-GPU solutions bring an additional layer of complexity when porting from CPU to GPU



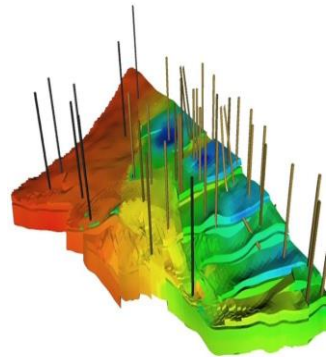
Hybrid AI-HPC for Scientific Computing



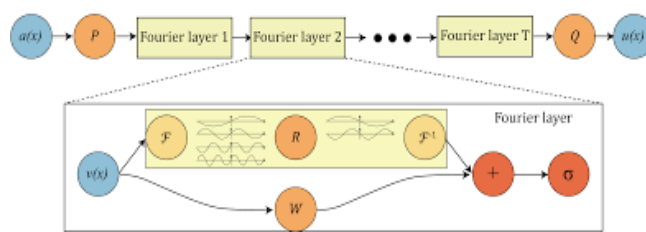
Reservoir Simulation with AI



Obtain few N data points from finite volume solver



FNO (Fourier Neural operators)



Build surrogate model with *PINO* (FNO + Physics)



Physics

$$\phi \frac{\partial S_\ell}{\partial t} = \nabla \cdot \left(-K \frac{k_{r\ell}}{\mu_\ell} [\nabla p_\ell - \rho_\ell \hat{g} \nabla z] \right) + f_\ell \dots \text{Eqn. (2)}$$

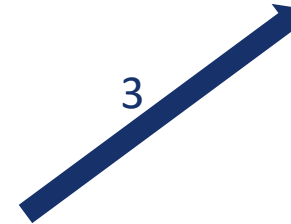
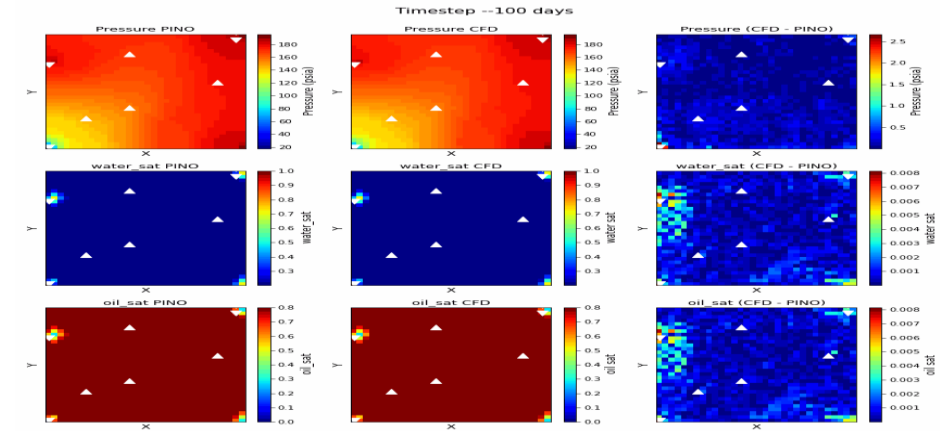
ϕ = effective porosity, S_ℓ = phase saturation for water (w) and oil (o), K = absolute permeability,

$k_{r\ell}$ = phase relative permeability and a function of the phase saturation, ρ_ℓ = phase density,

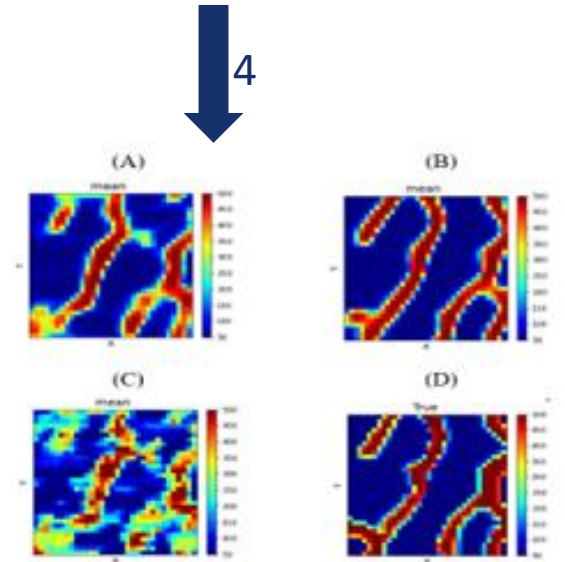
μ_ℓ = phase viscosity, $\hat{g} \nabla z$ = acceleration due to gravity. f_ℓ = source and sink terms representing the phase production and injection. The equation is closed by adding auxiliary equations for capillary pressure between oil and water ($p_{cow} = p_o - p_w$) and ($S_o + S_w = 1$).



PINO surrogate for solving the forward problem



Use *PINO* surrogate forward model in an inverse problem workflow



Reservoir Simulation AI



- Benefits
 - Additional acceleration compared to HPC results
- Challenges
 - Understanding and proper application of AI models
 - Still in R&D phase, not commercially available

