

# Cosmological N-Body Simulations and Galaxy Surveys

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ANL: HEP (Salman Habib, Katrin Heitmann), ALCF, MCS

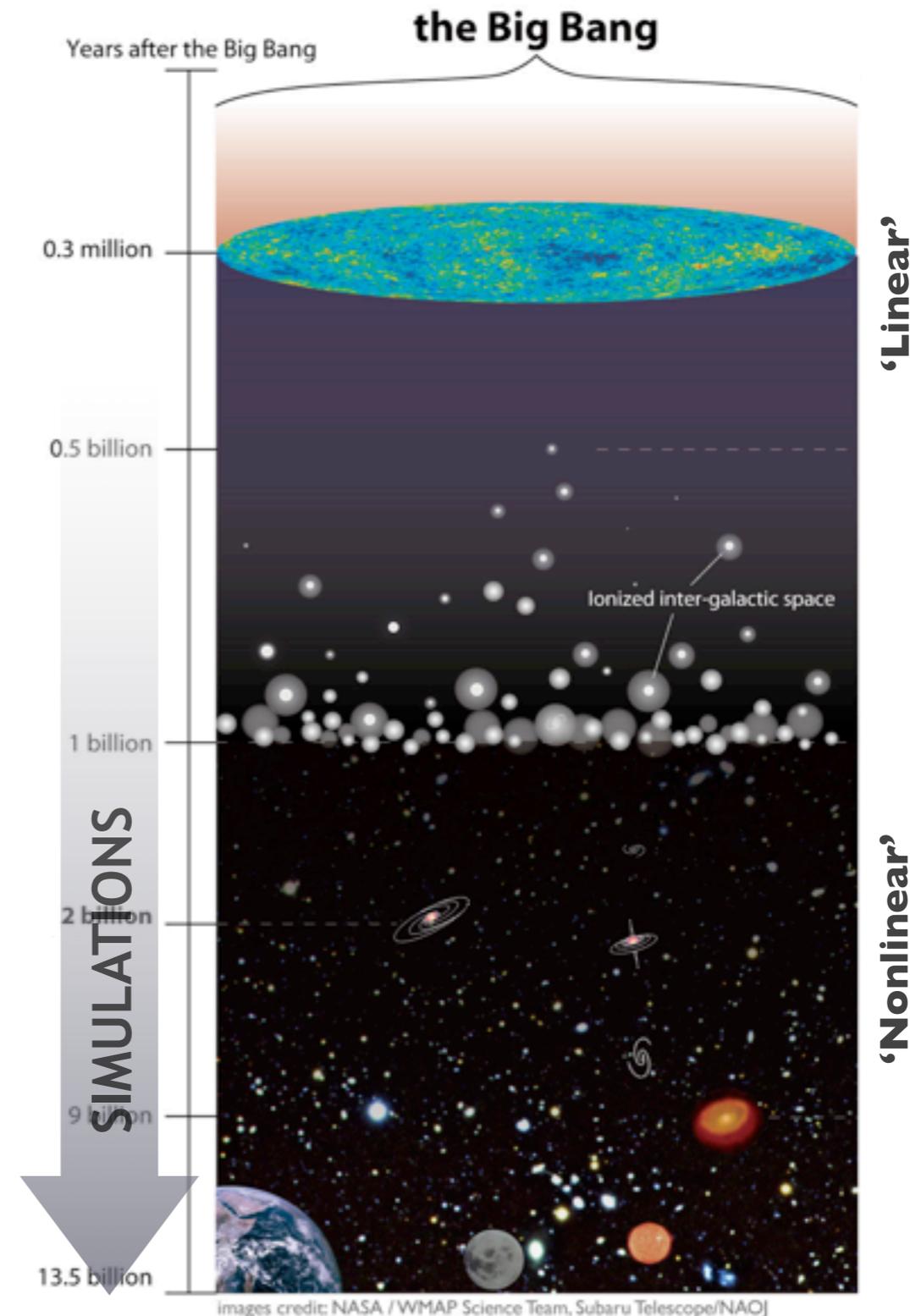
LANL: CCS-6, CCS-7

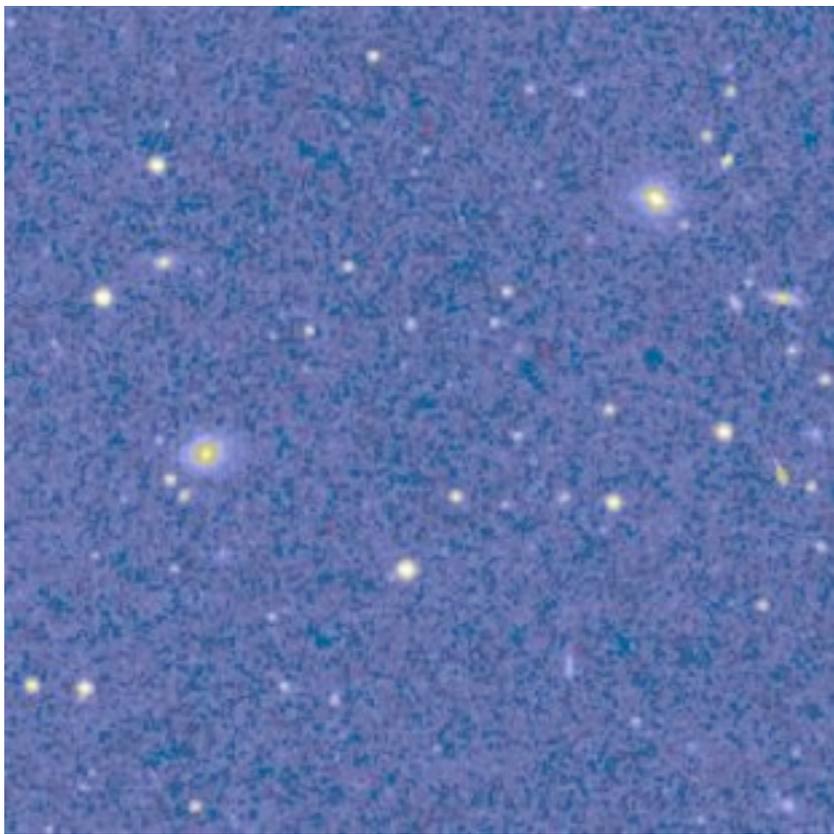
LBL: P, CR

UC Berkeley, Northwestern, Rutgers, VA Tech

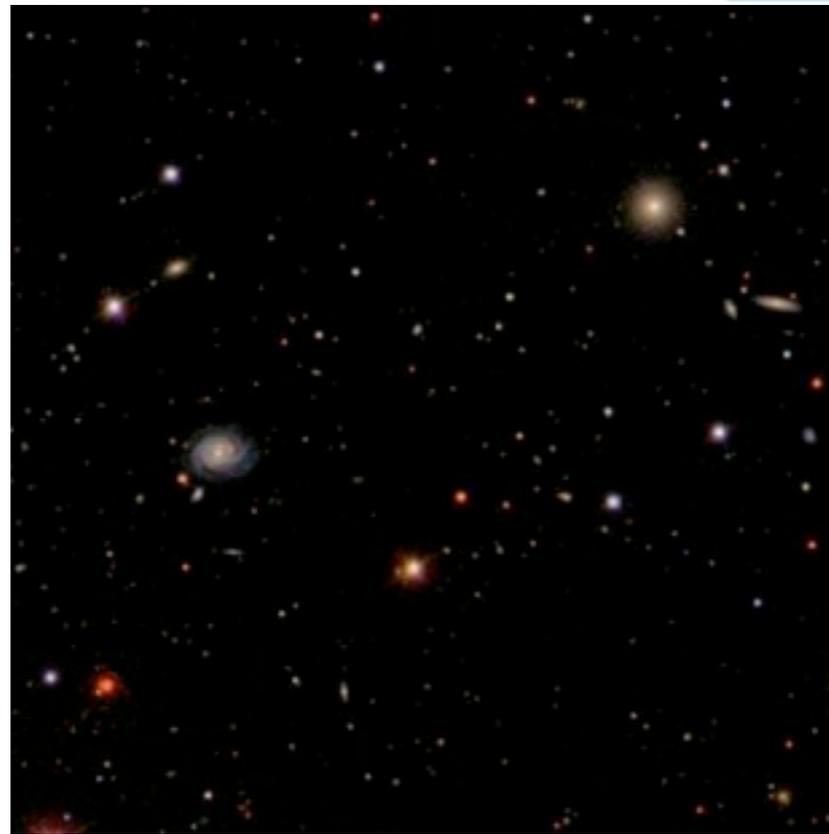
# Structure Formation: The Basic Paradigm

- ▶ **Solid understanding of structure formation; success underpins most cosmic discovery**
  - Initial conditions laid down by inflation
  - Initial perturbations amplified by gravitational instability in a dark matter-dominated Universe
  - Relevant theory is gravity, field theory, and atomic physics ('first principles')
- ▶ **Early Universe:**
  - Linear perturbation theory very successful (CMB)
- ▶ **Latter half of the history of the Universe:**
  - Nonlinear domain of structure formation, impossible to treat without large-scale computing





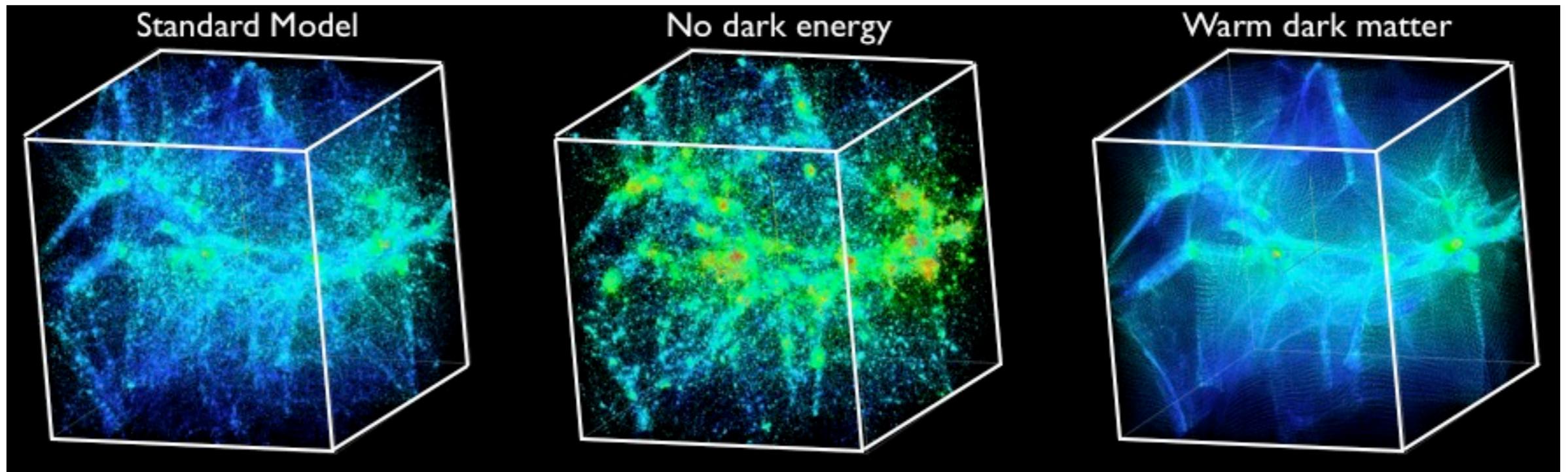
Digitized Sky Survey  
1950s-1990s



Sloan Digital Sky Survey  
2000-2008



Large Synoptic Survey Telescope  
2020-2030  
(Deep Lens Survey image)



# Gravity in an Expanding Universe

## ▶ Evolution

- Initial conditions: Gaussian random field of small density contrast
- Structure formation: gravitational collapse
  - Vlasov-Poisson equation in expanding space
  - Expansion slows collapse from exponential to power law, structures on many scales

## ▶ Accuracy

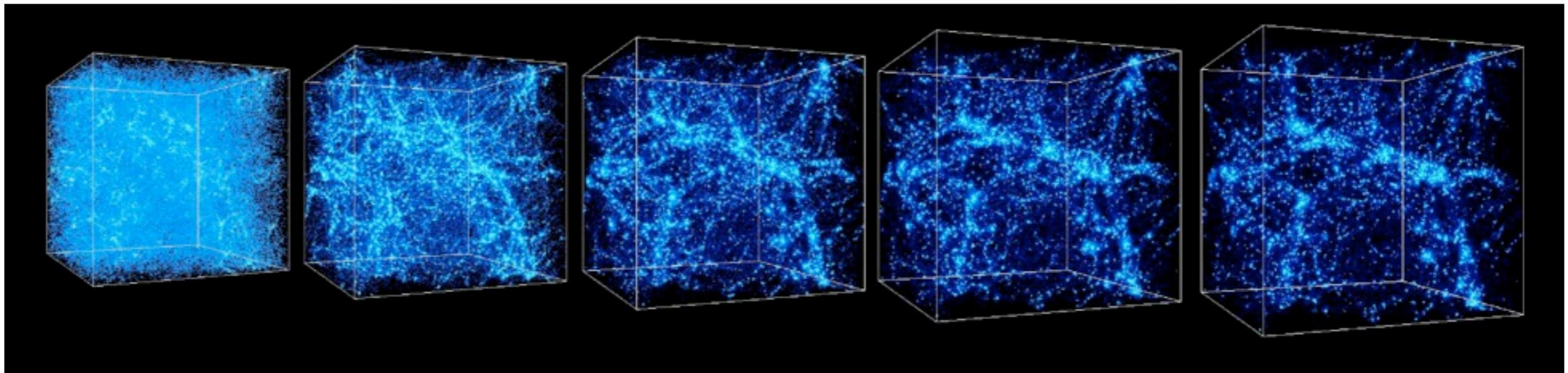
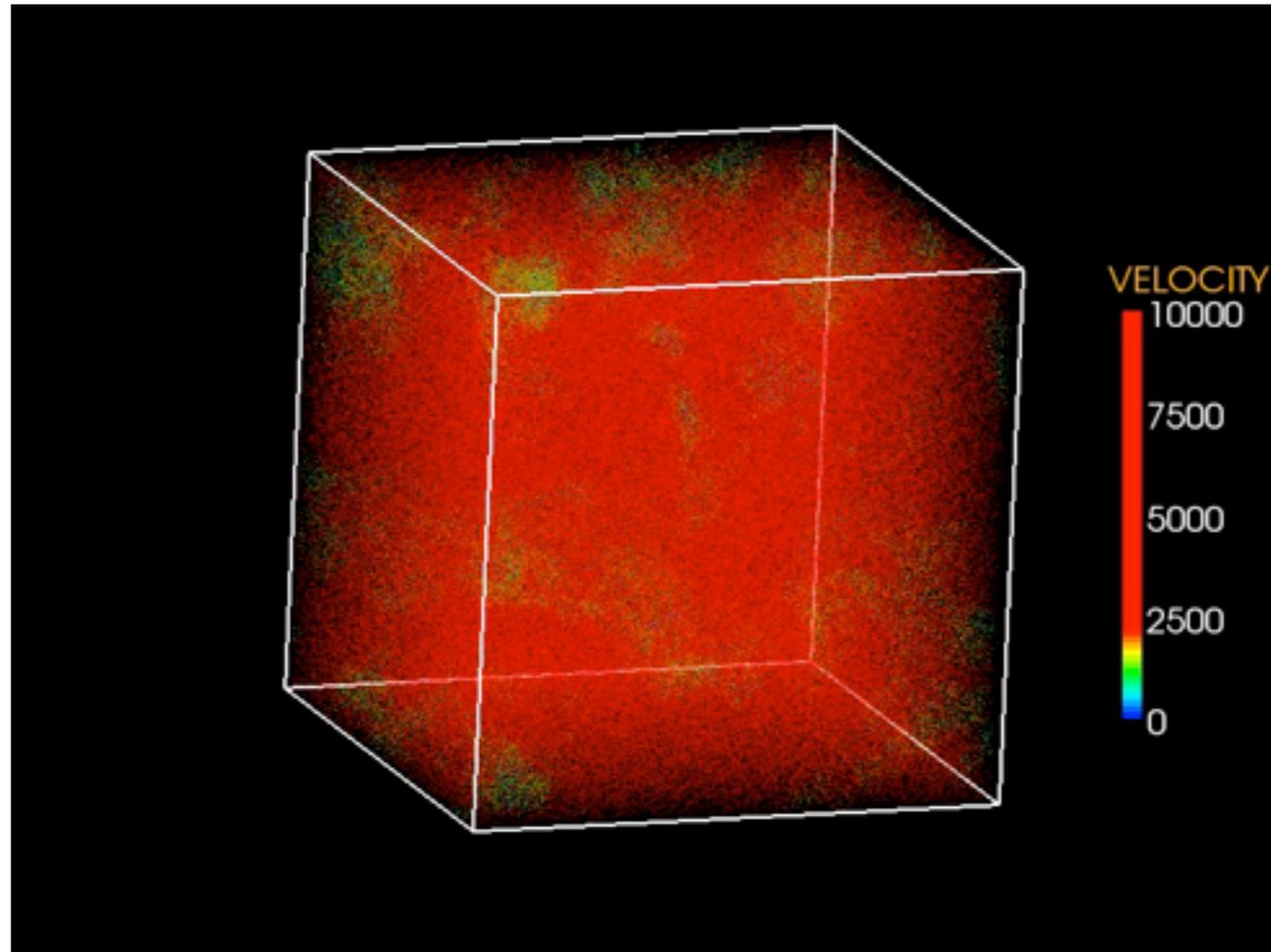
- Observations will soon require percent-level theoretical predictions
- N-body methods use tracer particles to evolve phase-space distribution (no 6D PDE)

## ▶ Inverse problem

- Compare forward model of structure formation with galaxy surveys
- Build emulators from limited number of simulations in parameter space

$$\begin{aligned}\frac{\partial f_i}{\partial t} + \dot{\mathbf{x}} \frac{\partial f_i}{\partial \mathbf{x}} - \nabla \phi \frac{\partial f_i}{\partial \mathbf{p}} &= 0, & \mathbf{p} &= a^2 \dot{\mathbf{x}}, \\ \nabla^2 \phi &= 4\pi G a^2 (\rho(\mathbf{x}, t) - \langle \rho_{\text{dm}}(t) \rangle) = 4\pi G a^2 \Omega_{\text{dm}} \delta_{\text{dm}} \rho_{\text{cr}}, \\ \delta_{\text{dm}}(\mathbf{x}, t) &= (\rho_{\text{dm}} - \langle \rho_{\text{dm}} \rangle) / \langle \rho_{\text{dm}} \rangle, \\ \rho_{\text{dm}}(\mathbf{x}, t) &= a^{-3} \sum_i m_i \int d^3 \mathbf{p} f_i(\mathbf{x}, \dot{\mathbf{x}}, t).\end{aligned}$$

# Simulations of Structure Formation in the Universe



# HACC: Hybrid Accelerated Cosmology Code

## ▶ Requirements

- Particles:  $10^{10}$  standard,  $10^{11}$  state-of-the-art,  $10^{12}$  very soon
- Throughput: many runs for inverse problems, days for each run

## ▶ Supercomputers

- Parallelism: code must scale
  - Many nodes: message passing, weak scaling, Mira = 48k nodes
  - Many cores: local threading, strong scaling, Mira = 16 cores x 4 hardware threads per node
- Architectures: code must adapt
  - CPU, IBM Cell, IBM Blue Gene, GPU, Intel MIC
  - Different programming paradigms and optimizations

## ▶ Design

- Split gravitational force calculation
  - Long-range: spectral (FFT) particle-mesh (PM) methods using MPI, portable
  - Short-range: algorithm choice and optimizations for particular architectures, modular, high intensity

## ▶ Versions

- P<sup>3</sup>M (particle-particle, particle-mesh): IBM Cell, GPU
- TreePM: CPU, IBM Blue Gene

# Data Challenges: At Runtime

## ▶ Checkpoints

- Up to (roughly) half of machine memory to disk, Mira = 100s TB (without compression)
- Must be fast to reduce time taken from allocation
- Stressful for network-attached storage
- Robust persistent local storage would be nice

## ▶ Data reduction

- Cannot store enough full particle outputs to do all analysis in post-processing
- On-the-fly analysis to produce reduced outputs
  - Halo finding (galaxies, clusters), 2D mass projection (weak lensing), sub-samples of particles, etc.
- Reduced outputs must be sufficient for scientific analysis

## ▶ Architectures

- More varied analysis code is more difficult to optimize for different HPC systems



# Data Challenges: Sharing Data Products

- ▶ Rich simulation outputs
  - Many different scientific queries on the same simulations
  - Insufficient bandwidth to disperse complete (reduced) data products
- ▶ How should scientists interact with the outputs?
  - Move analysis local to data
  - Further reductions before network transfer
- ▶ What will that require?
  - What kind of hardware?
    - HPC, DB, DISC (Data-Intensive SuperComputer)
  - What kind of software? Who will write/maintain? What can we build on?
    - SQL, NoSQL
  - Have worked with Globus Online
  - Looking at Galaxy (computational biology)

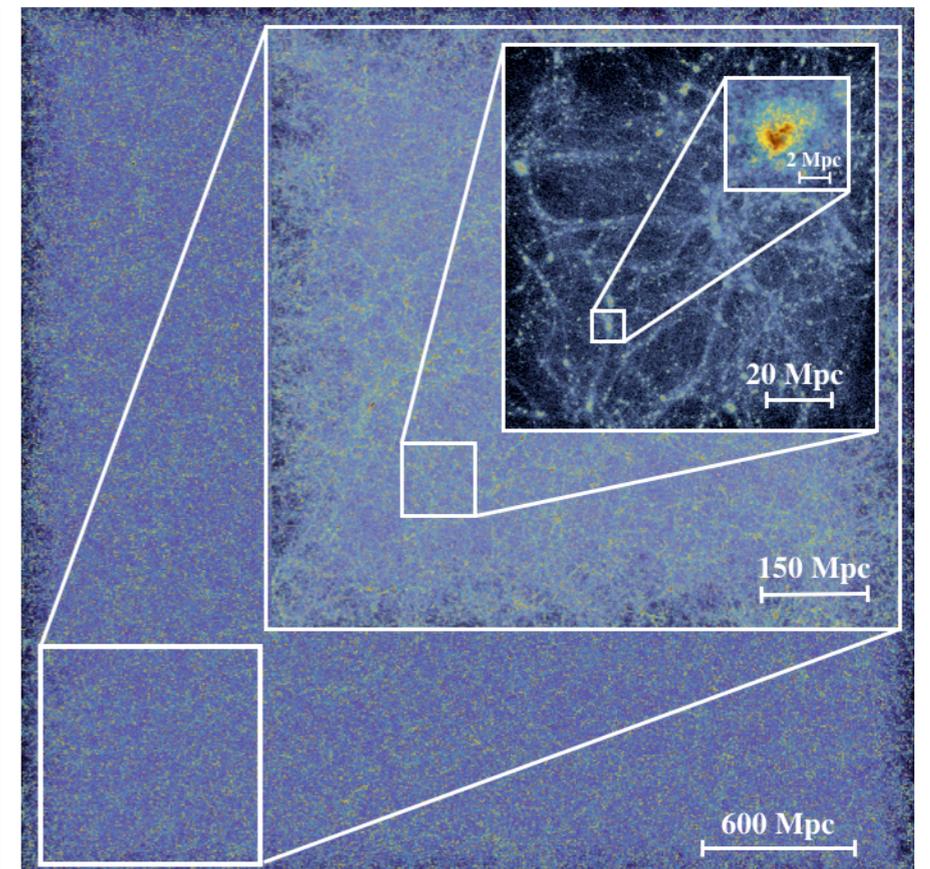
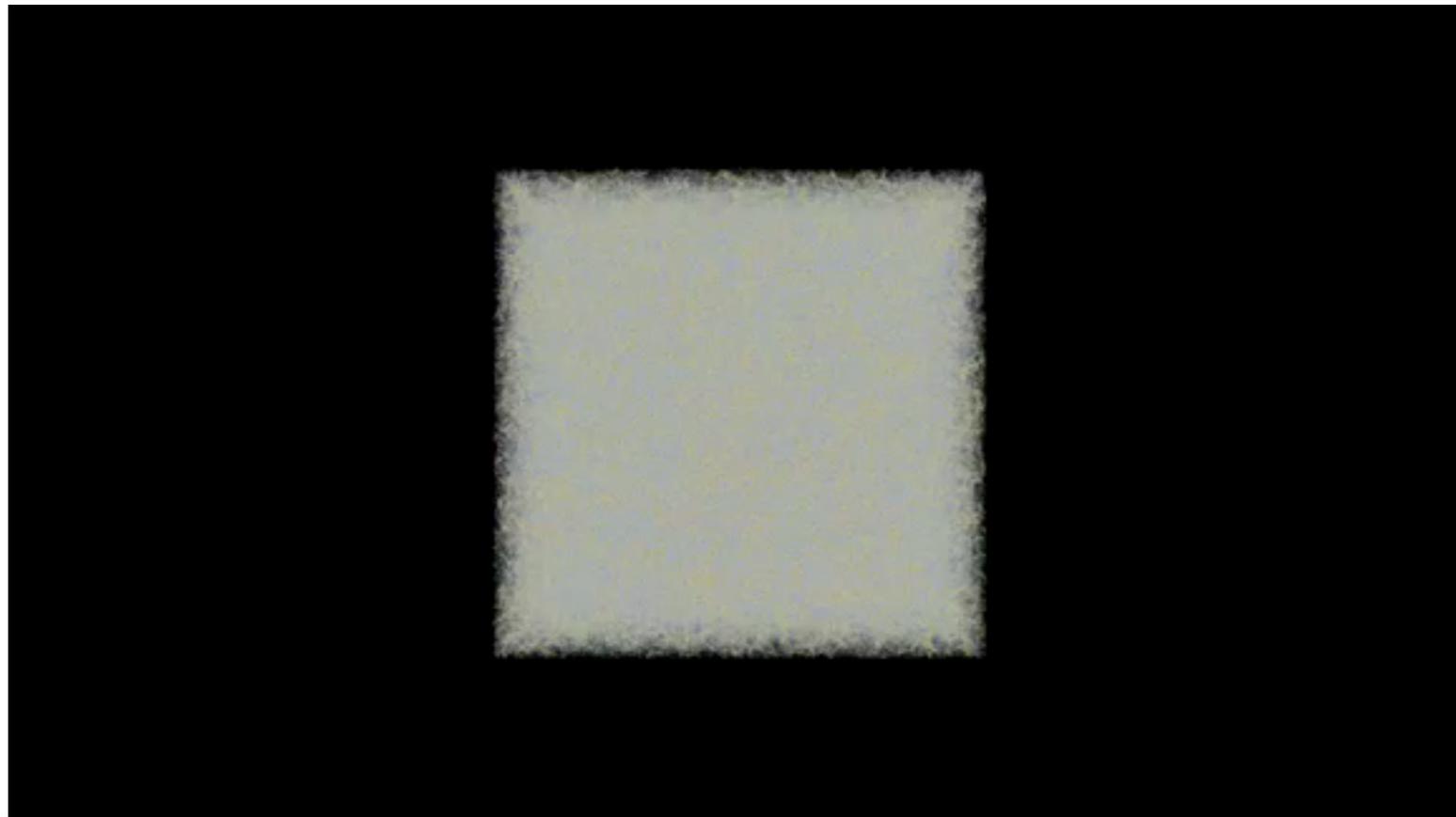
# Data Challenges: Visualization

## ▶ Particle number

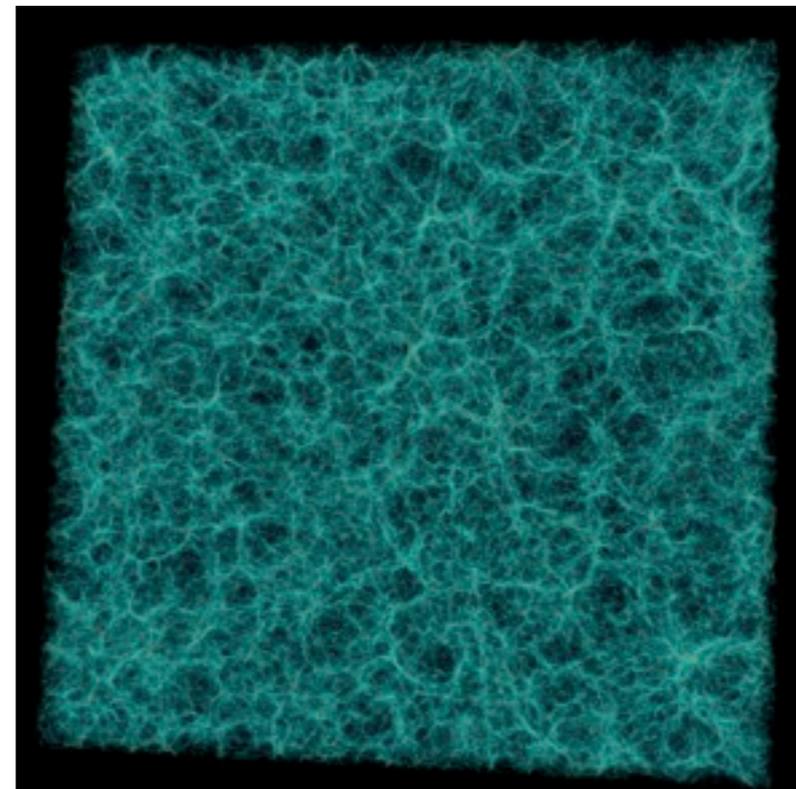
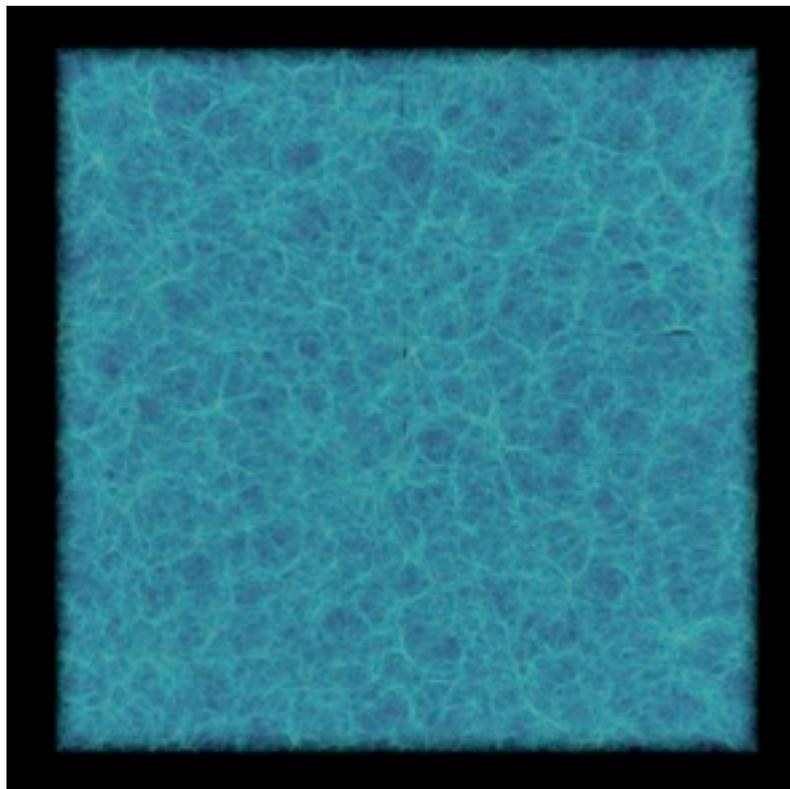
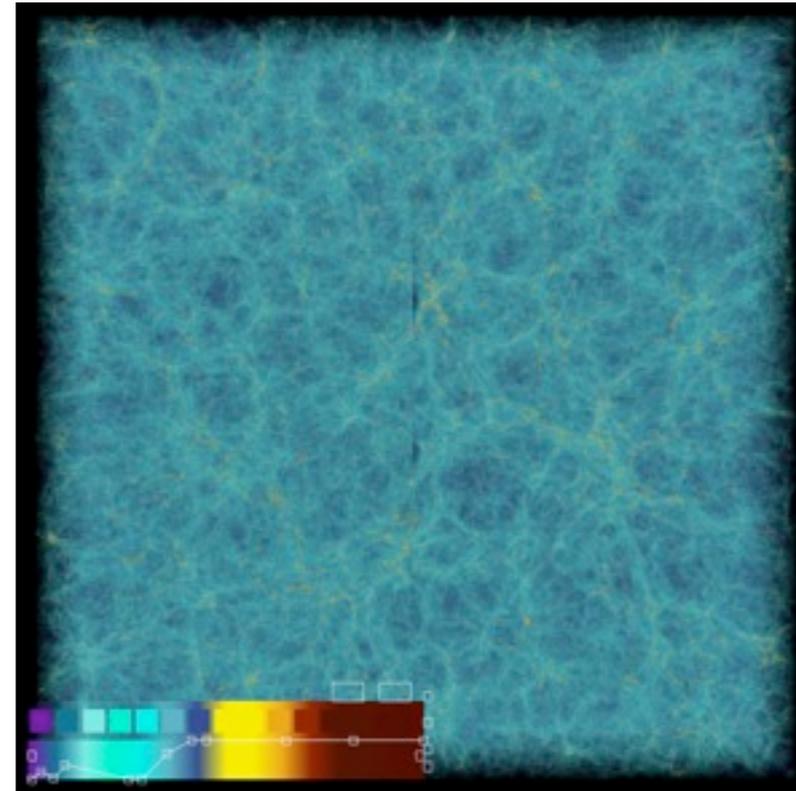
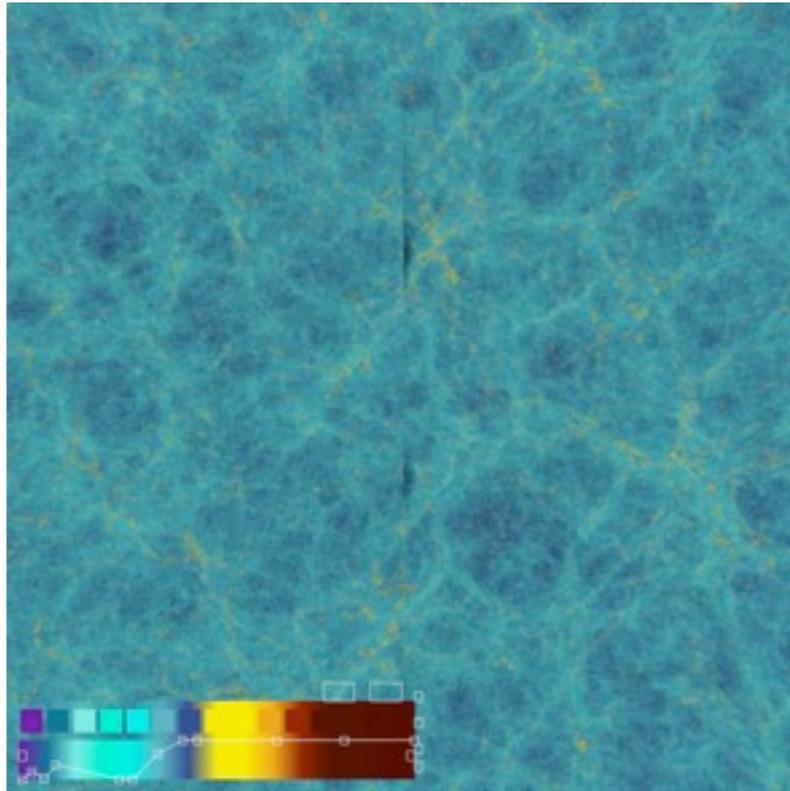
- Far more particles than pixels
  - Some experiments with particle sampling methods
- Volume rendering seems to be more common than direct particle methods
- Visualize reduced outputs

## ▶ Dynamic range

- Scales: homogeneous on large scales, complex structures everywhere on small scales
- Huge density contrasts



# Preliminary Slice of a 230 Billion Particle Simulation



# Data Challenges: Putting It All Together

- ▶ Observational data volume will not be the challenge
  - But these data sets are very rich and scientists will want to query it in many different ways
- ▶ Numerical theory can produce (almost arbitrarily) large volumes of data
  - Also rich
- ▶ How do we put these together?
  - Data-Intensive SuperComputers (DISC)

