

# Star-by-star simulations of the Milky Way accelerated by AI surrogate modeling

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The 13th "FugakuNEXT" Application Seminar

"Colors of Liberty" project, Eduardo Kobra

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1. What are galaxy simulations?
2. Simulation with AI surrogate modeling (Hirashima et al. SC'25)
3. Porting our code to GPUs using CUDA and SYCL

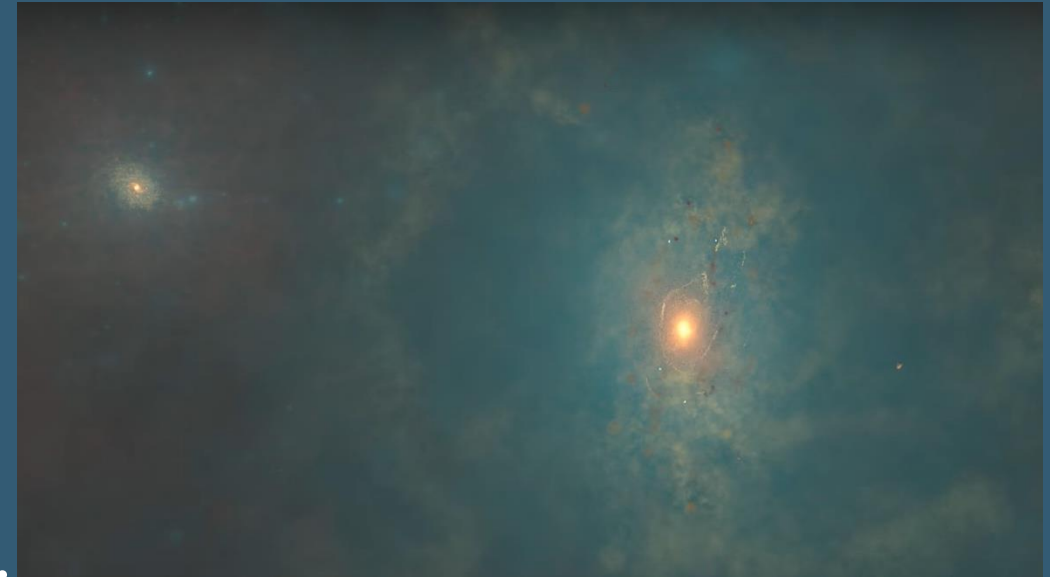
# What are Galaxy Simulations?

- The system with dark matter, stars, and gas.
- Many particle system ( $10^{11}$  stars in the Milky Way)
- Multi-timescale ( $\sim 10^8$  yr for galactic rotation v.s.  $10^2$  yr for supernovae)
- Track the time evolution of kinematics and thermodynamics by integrating ordinary equations for each pair of particles (Poisson equations and Euler equations).

e.g., self-gravity

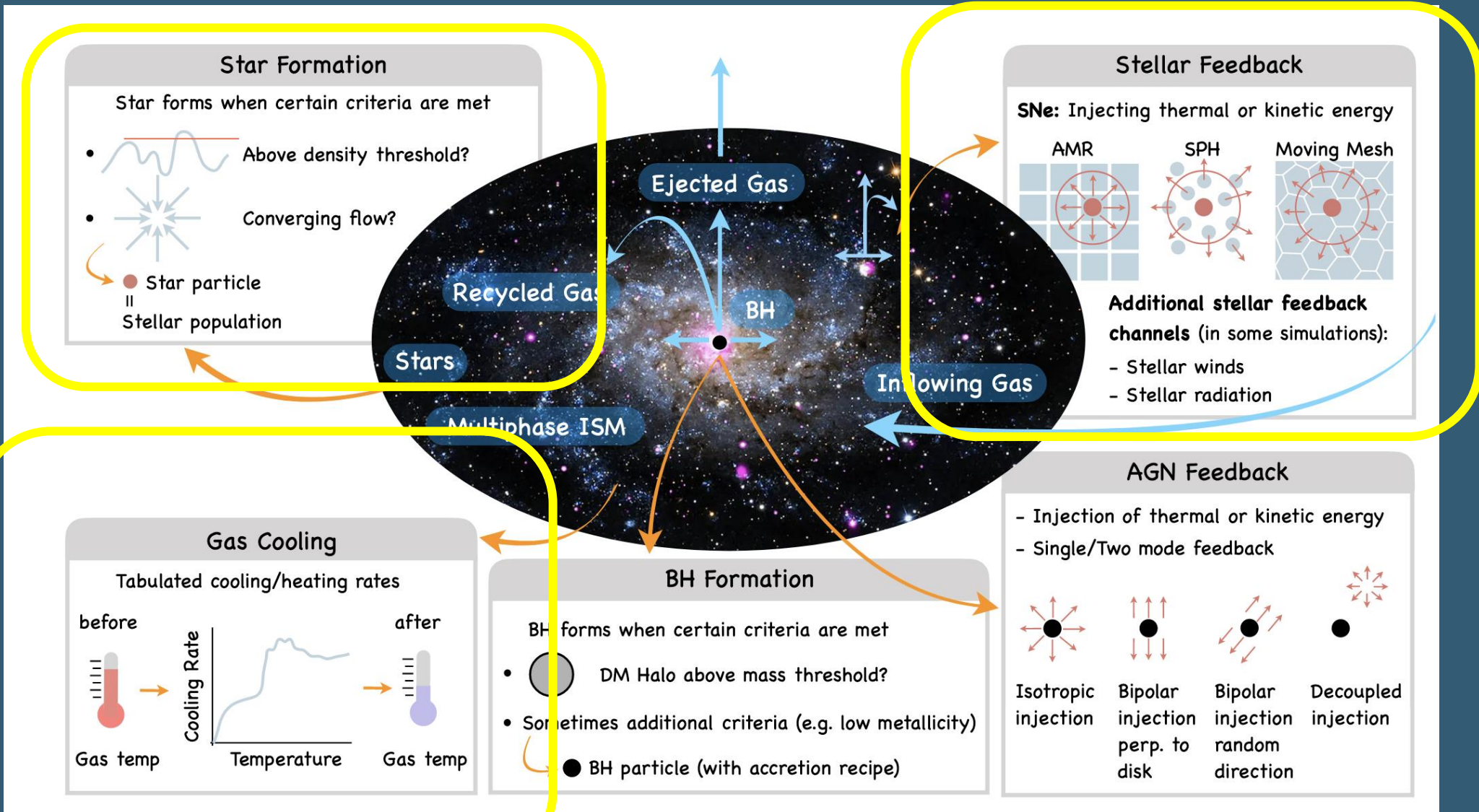
$$\frac{d^2 \mathbf{x}_i}{dt^2} = \sum_{j \neq i, 1 \leq j \leq N} G m_j \frac{\mathbf{x}_j - \mathbf{x}_i}{|\mathbf{x}_j - \mathbf{x}_i|^3}$$

- Other physics
  - Supernova, radiation, star-formation...



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# Models for Galaxy (Formation) Sims.



# Models for Galaxy (Formation) Sims.

### Star

Star forms when

- Ab...
- Col...

Star particle  
Stellar populati...

### Dark matter + baryons (hydrodynamical)

### Stellar Feedback

jecting thermal or kinetic energy

Additional stellar feedback channels (in some simulations):

- Stellar winds
- Stellar radiation

### Gas C

Tabulated cooling

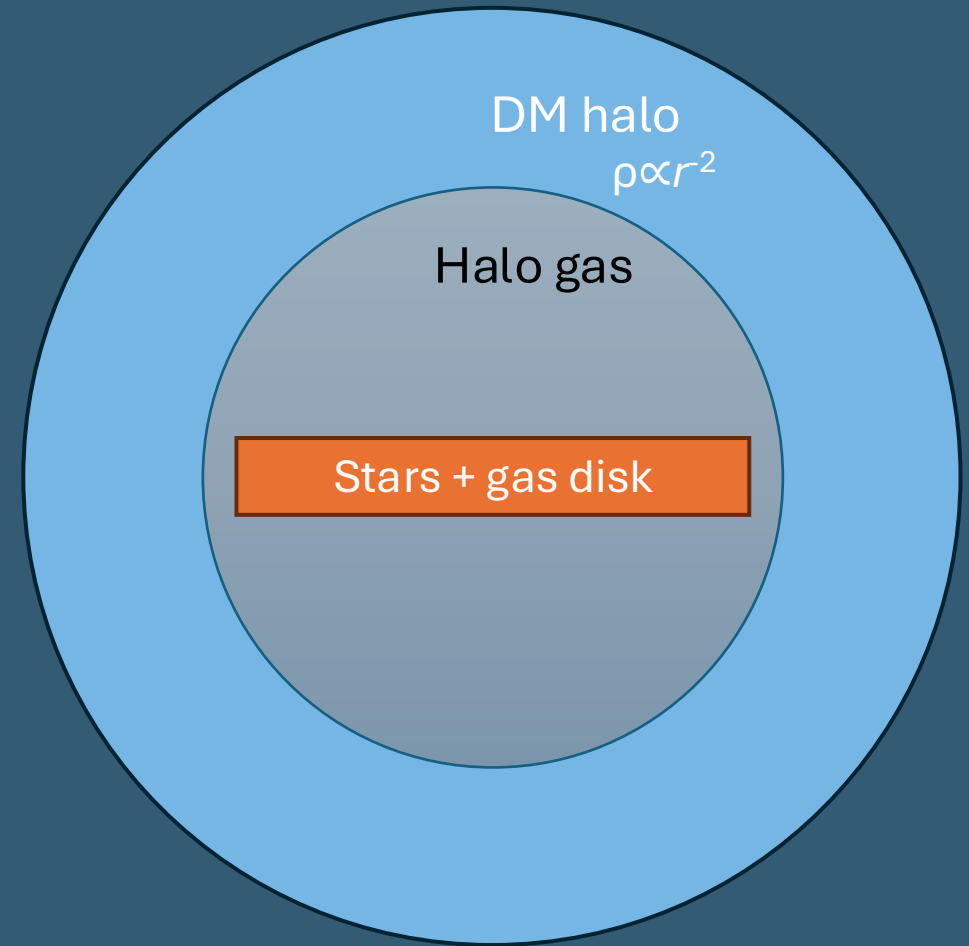
### AGN Feedback

n of thermal or kinetic energy  
Two mode feedback

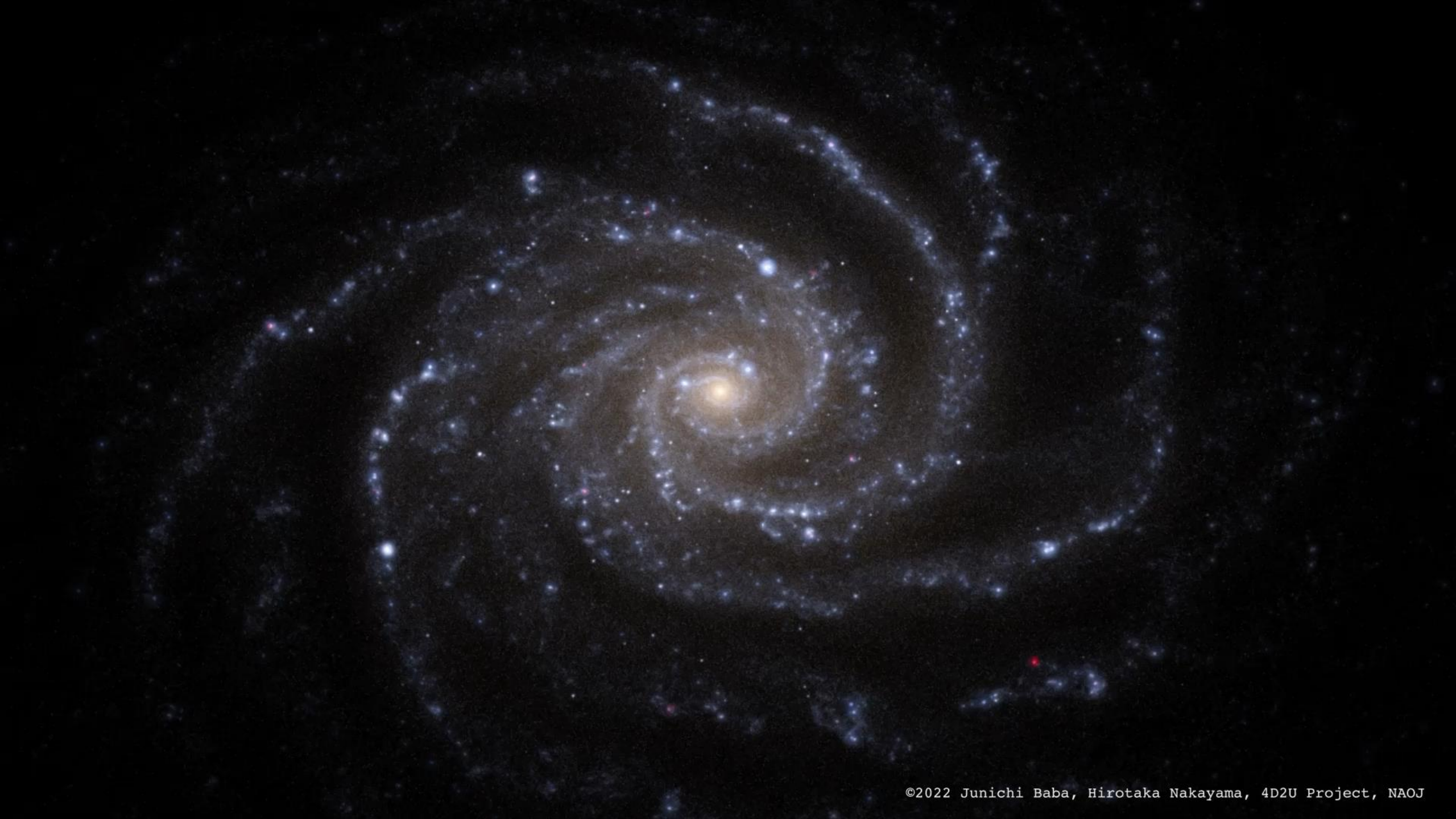
Bipolar injection perp. to disk    Bipolar injection random direction    Decoupled injection

# Idealized Galaxy Model

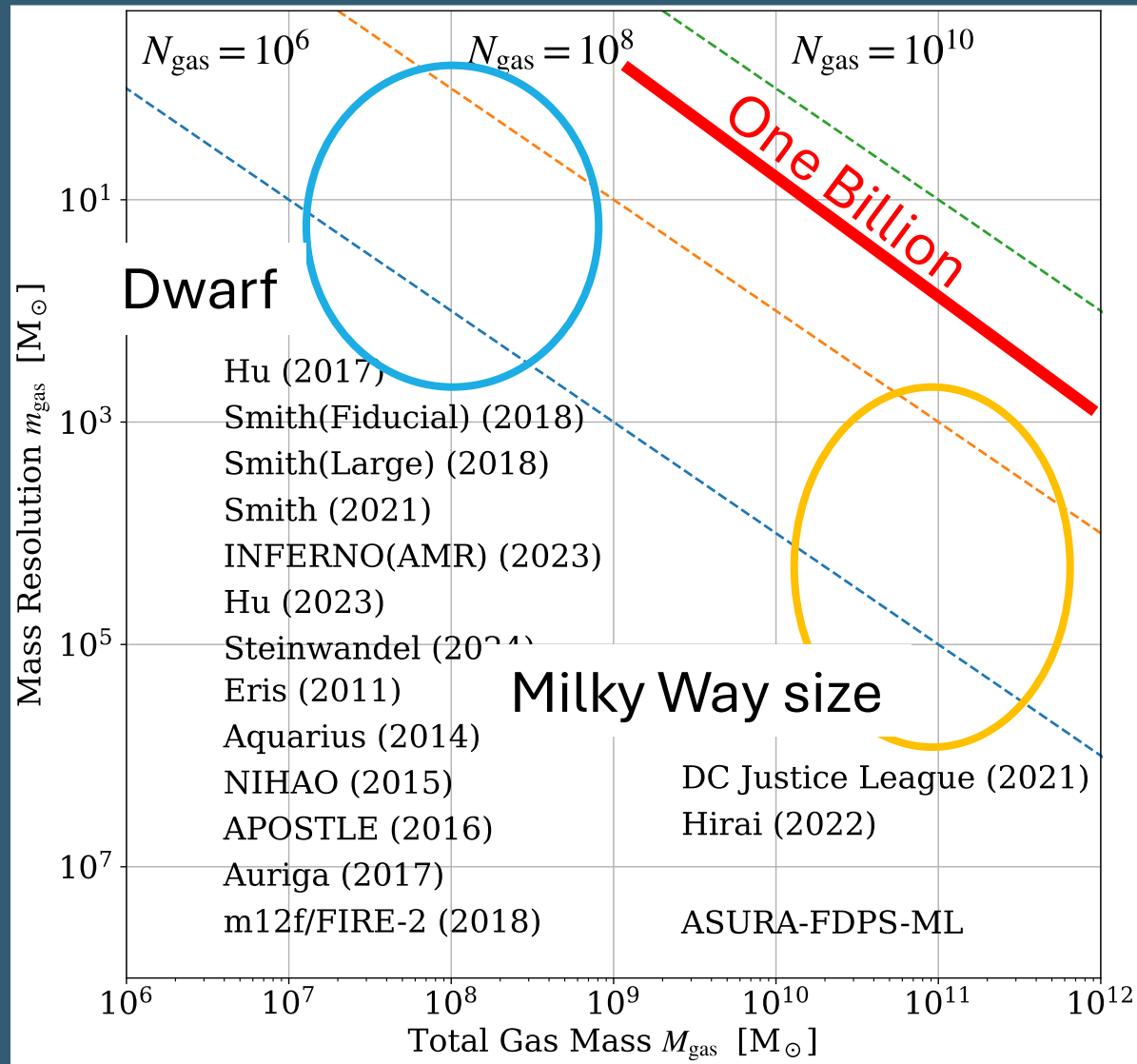
- Three components
  - Gas: hydro + gravity (4 Msun)
  - Dark matter (DM): gravity (4k Msun)
  - Stars: gravity (down to 4 Msun)



DM halo is 20—100 times more massive than the disk



# Progress in Resolution of Galactic Disk Simulations



State-of-the-art simulations

1. Star-by-star simulations of dwarf galaxies.
2. Low-resolution Milky Way-sized galaxy simulations.

- Higher resolution further increases the cost of resolving small-scale physics.

# Why do we need to pursuit high-resolutions?

Observation of M74

1975

2003~2005

2022



$\theta > 1'' \sim 47\text{pc}$

Optical

Star-forming regions  
Spiral arms

KPNO 4-meter Mayall telescope, 1975  
Credit: NOIRLab/NSF/AURA



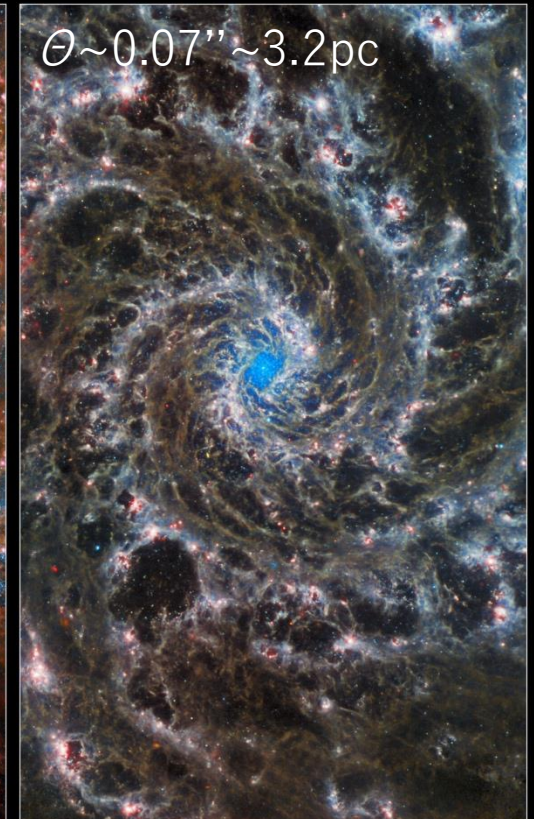
$\theta \sim 0.05'' \sim 2.3\text{pc}$

Hubble / Optical

HII regions  
Star clusters



Hubble & Webb



$\theta \sim 0.07'' \sim 3.2\text{pc}$

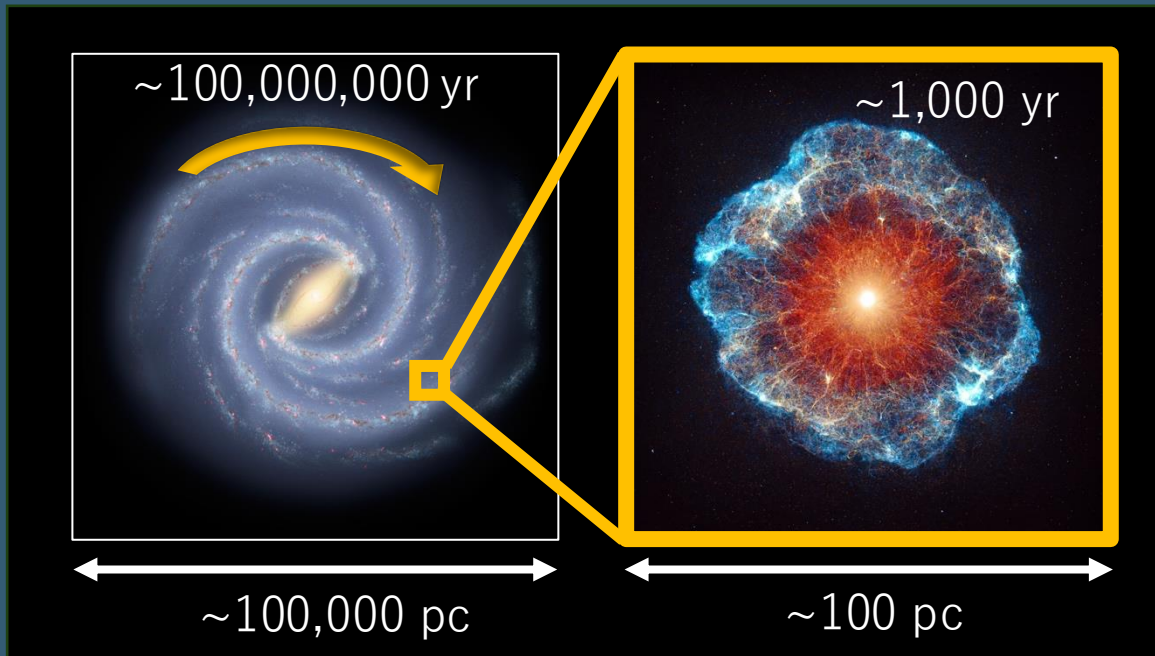
Webb / Infrared

Filaments  
Shock waves

ESA/Webb, NASA & CSA, J. Lee and the PHANGS-JWST Team; ESA/Hubble & NASA, R. Chandar Acknowledgement: J. Schmidt

# Higher Resolution Demands Greater Costs

- High-resolution simulation domains demand short timesteps due to high density and temperature (e.g., SNe)



NASA/JPL-Caltech/ESO/R. Hurt

- Estimating a typical timestep
  - Size of SPH particles

$$\lambda = \left( \frac{3 m_{\text{SPH}}}{4\pi \rho} \right)^{1/3}$$

Mass resolutions:  $m_{\text{SPH}}$   
Density:  $\rho$

- Internal energy per particle

$$U_{\text{SN}} := \frac{E_{\text{SN}}}{N_{\text{NB}} m_{\text{SPH}}} \simeq \frac{5.0 \times 10^{17}}{N_{\text{NB}}} \left( \frac{1 M_{\odot}}{m_{\text{SPH}}} \right) [\text{erg g}^{-1}]$$

- Sound speed

$$c_{\text{SN}} := \sqrt{\gamma(\gamma - 1)U_{\text{SN}}} \simeq \frac{7.5 \times 10^3}{N_{\text{NB}}^{1/2}} \left( \frac{1 M_{\odot}}{m_{\text{SPH}}} \right)^{1/2} [\text{km s}^{-1}].$$

Neighbor number  $N_{\text{NB}} \sim 100$

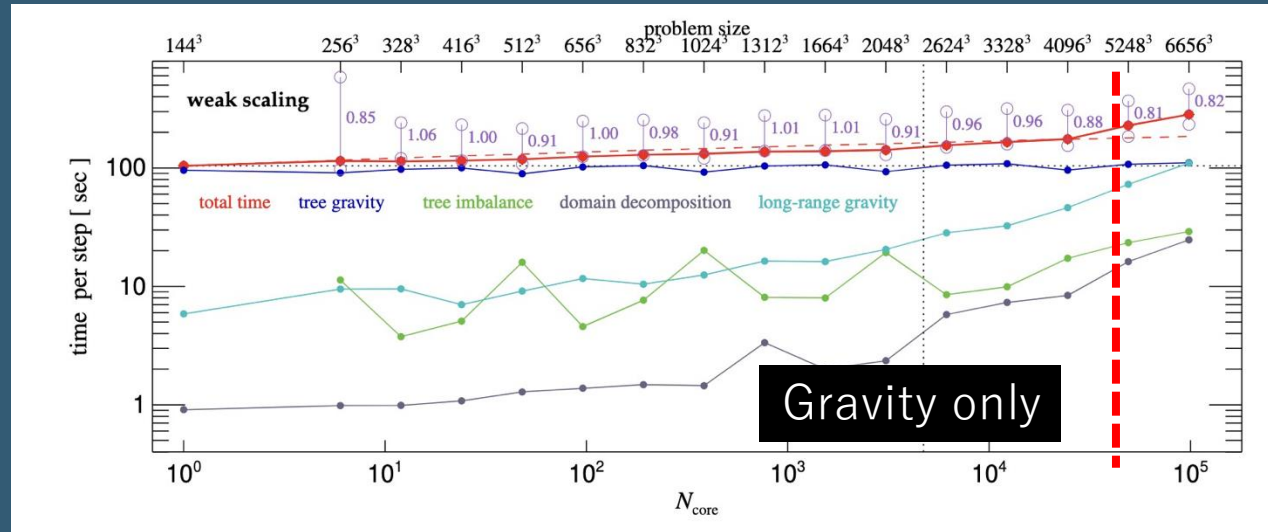
- Timescale of SNe at certain resolution and density

$$t_{\text{SN}} \simeq 1.3 \times 10^3 \left( \frac{m_{\text{SPH}}}{1 M_{\odot}} \right)^{5/6} \left( \frac{100 [\text{cm}^{-3}]}{n_{\text{H}}} \right)^{1/3} [\text{yr}].$$

- Typical timestep:  $0.1 t_{\text{SN}} \sim 10^2 \text{ yr}$

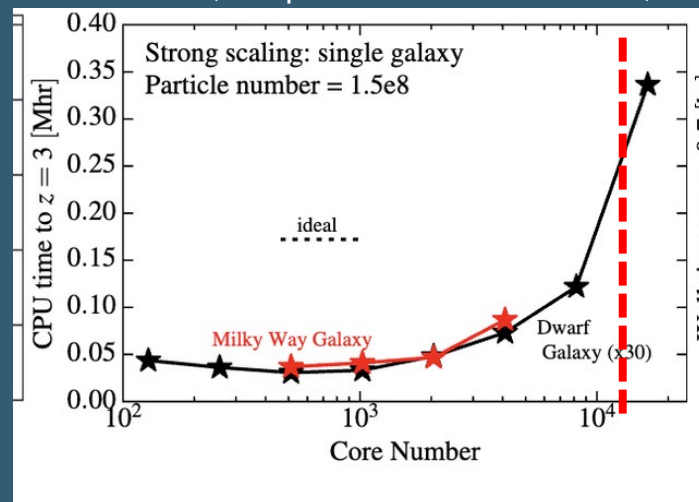
# Parallelization Efficiency in Galaxy Simulations

Gadget4 (Springel et al. 2021)



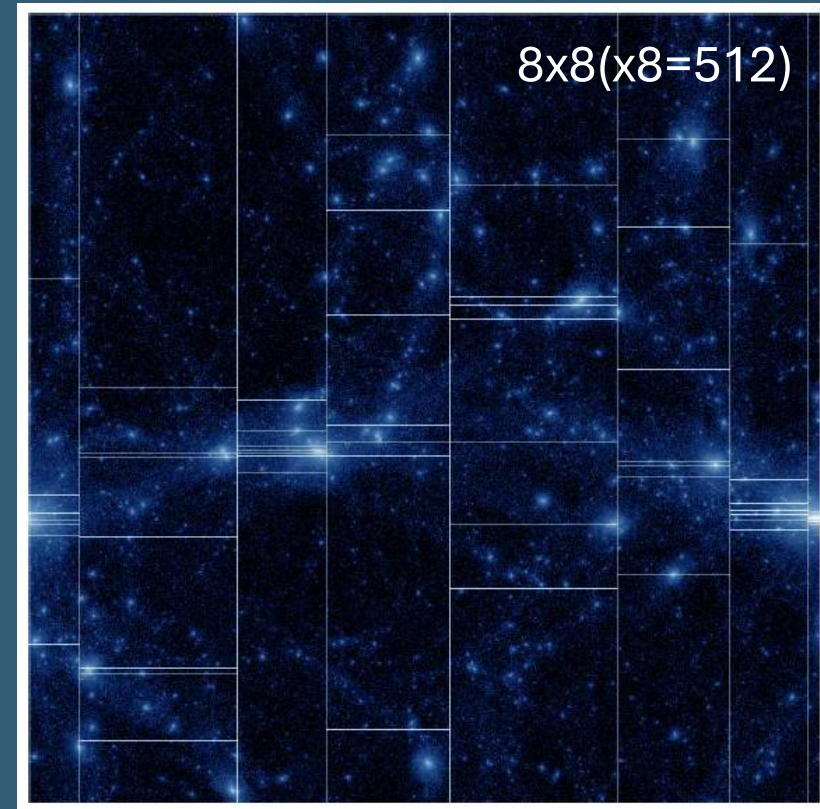
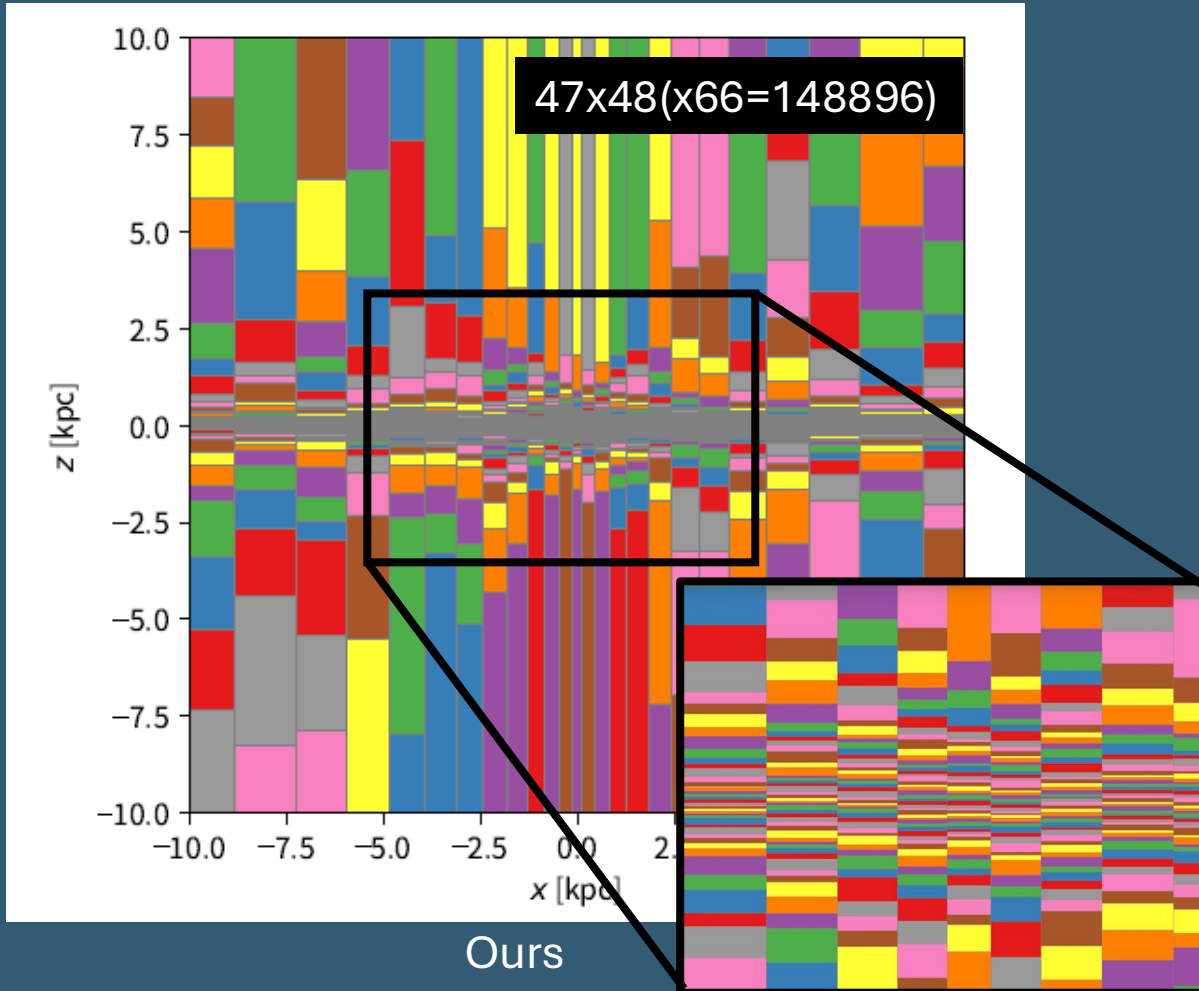
- The communication overhead increases at  $\sim 1\text{k}$  nodes.
- Given a certain mass, the number of required steps is proportional to the cubic root of #particles,  $N^{1/3}$ .
- Small-scale physics cause load-imbalance.

GIZMO (Hopkins et al. 2018)



# Challenges in Load Balance: Domain Decomposition

In a galaxy, the density increases toward the center, following a power-law profile, which could lead to severe load imbalance across nodes after domain decomposition.

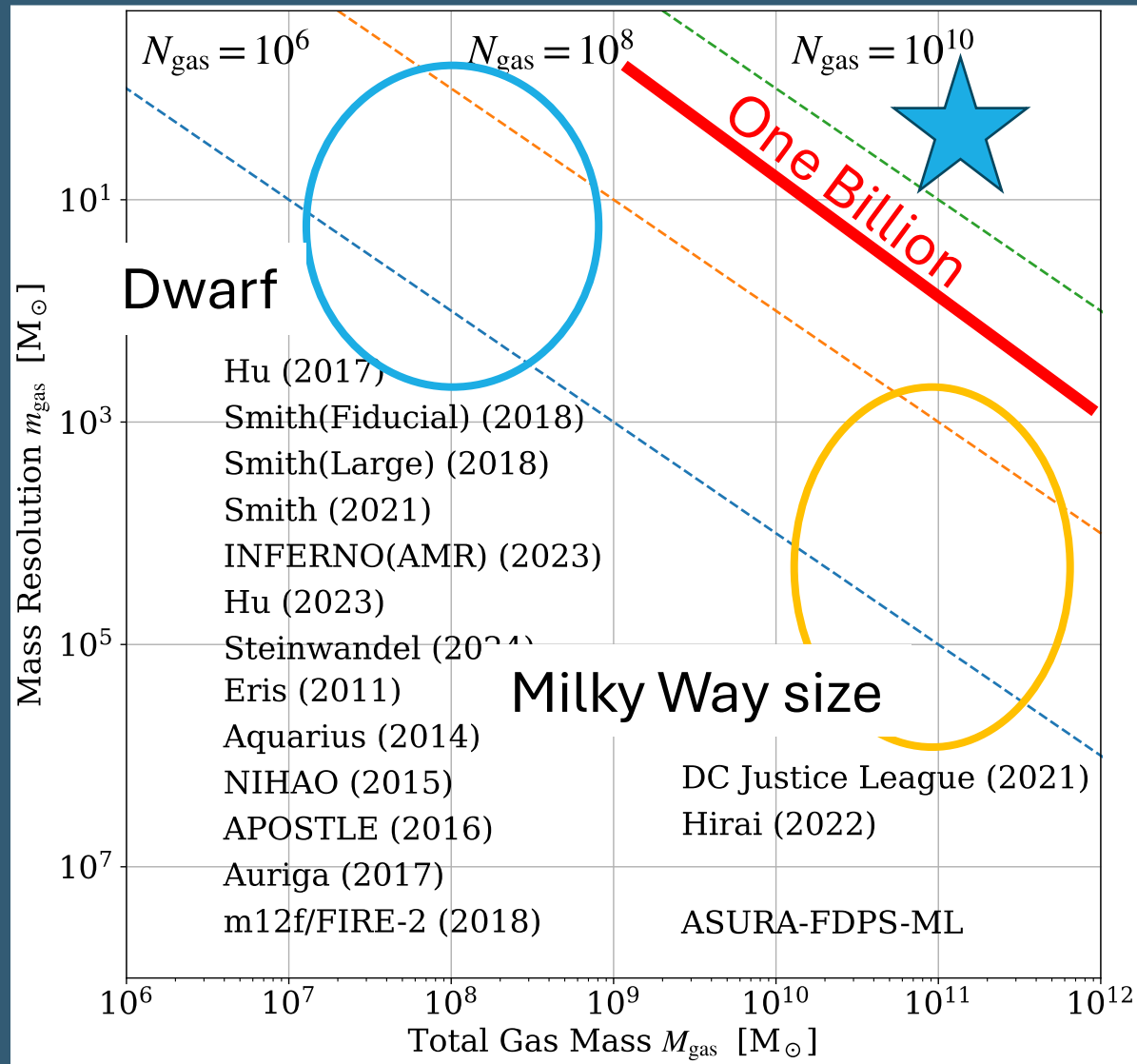


Ishiyama et al. (2012), Gordon-Bell prize

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# Current Status of Galactic Disk Simulations



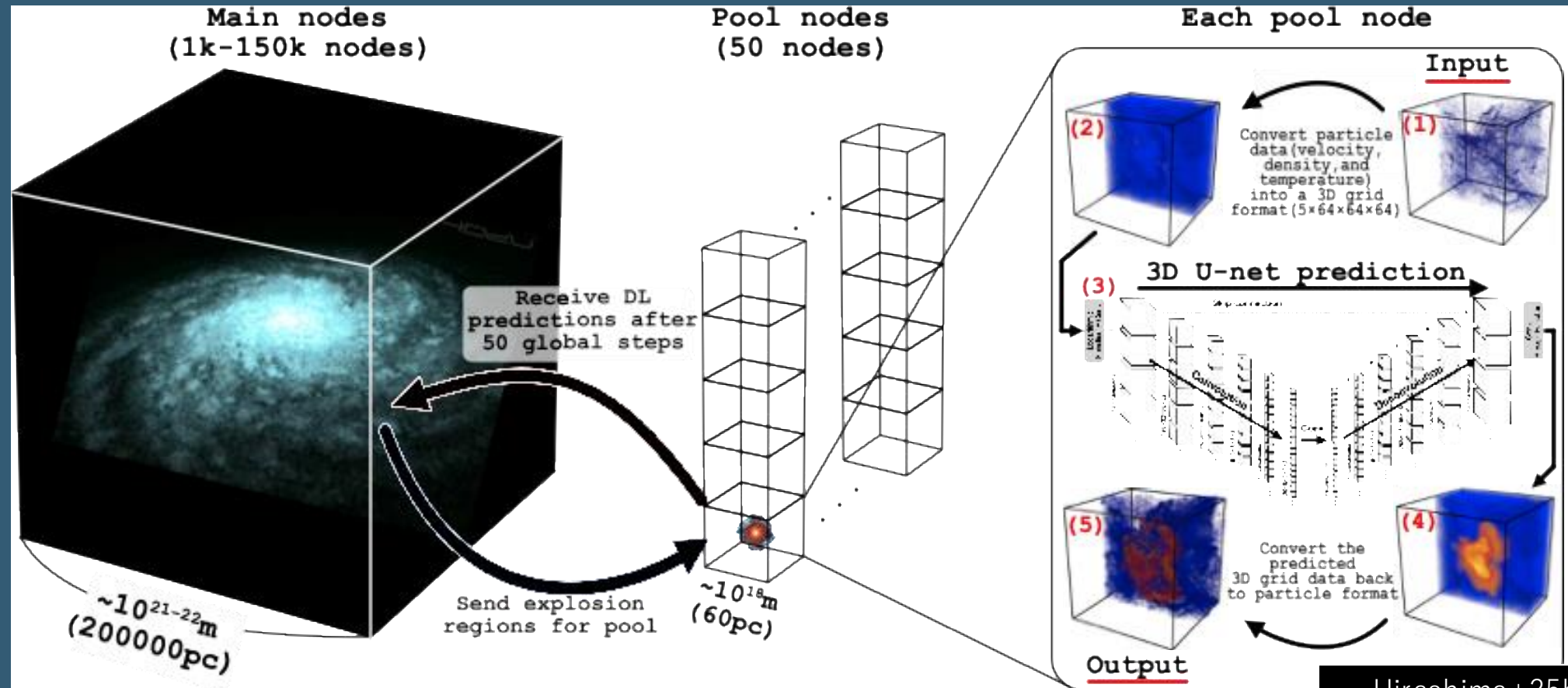
- State-of-the-art simulations
1. Star-by-star simulations of dwarf galaxies.
  2. Low-resolution Milky Way-sized galaxy simulations.

**Aim  $10^{11}$  particles for star-by-star Milky Way-sized simulations**

- Higher resolution further increases the cost of resolving small-scale physics.

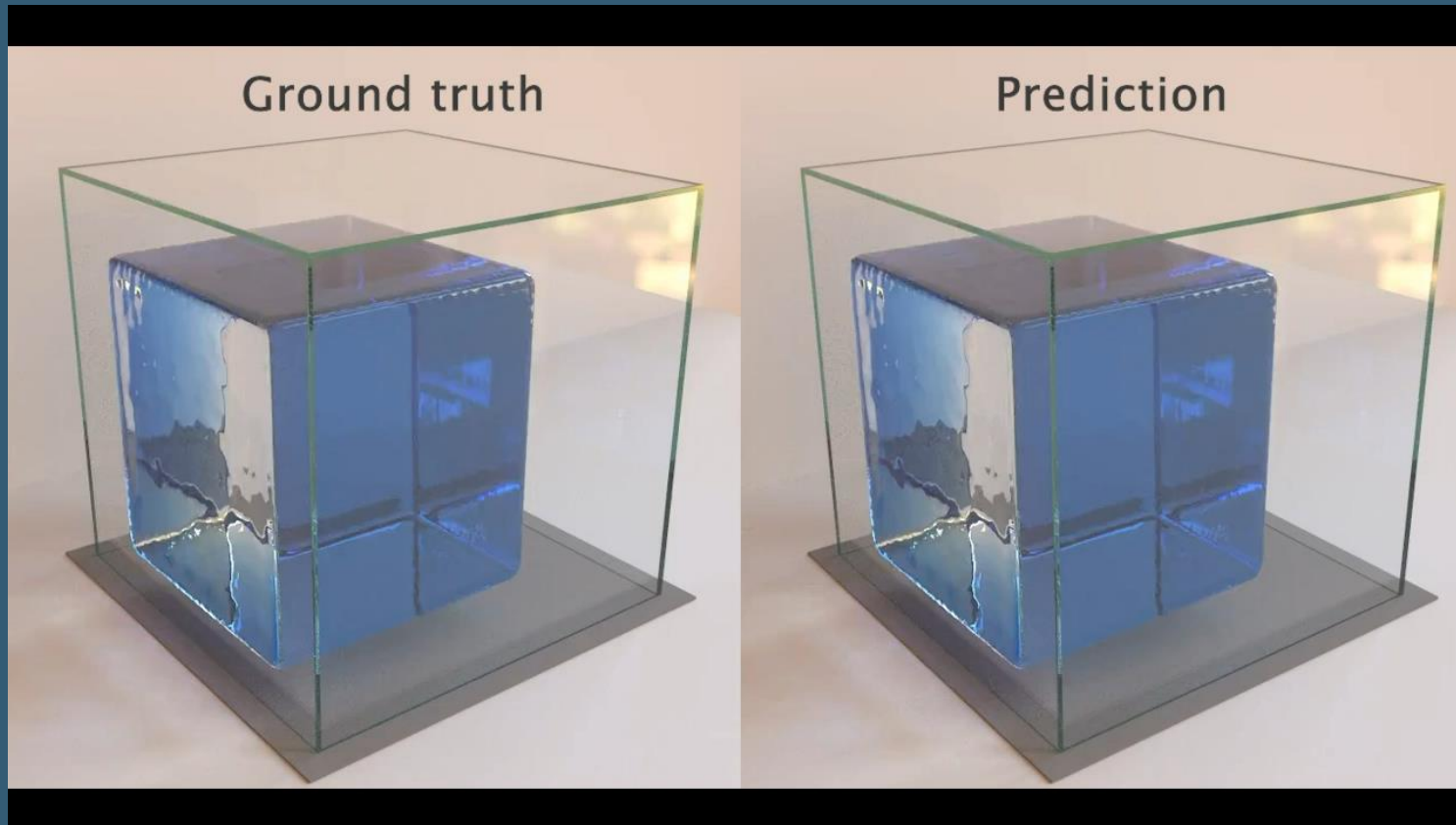
# Our Framework with AI: ASURA-FDPS-ML

- ASURA (Physical model for galaxy simulations)
- FDPS (Framework for Developing Particle Simulator in parallel computing)
- PIKG (an automatic Particle-particle Interaction Kernel Generator)
- Surrogate Modeling (Deep learning for fast computation of supernovae)



# Surrogate Model

- Surrogating (replacing) expensive simulations with mathematical models (incl. machine learning)
- surrogate model, emulator, forward modeling are the same in the context of “AI for Science”.



Sanchez-Gonzalez+2020

SPH interaction estimation  
using Graph Neural  
Network

# Training Data (3D Cartesian Grids)

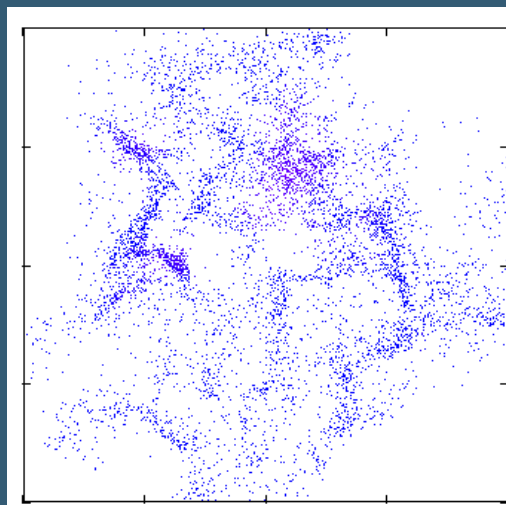
Training dataset: 300 simulations of a SN in turbulent clouds.

Temperature	10 [K]
Mean ambient density	4 ~ 1000 [cm <sup>-3</sup> ]
Input energy	10 <sup>51</sup> [erg]
Total mass	10 <sup>6</sup> [M <sub>⊙</sub> ]
Mass of a gas particle	1 [M <sub>⊙</sub> ]
Softening parameter	0.5 [pc]

The dataset is publicly available at **The Well** ([https://github.com/PolymathicAI/the\\_well](https://github.com/PolymathicAI/the_well)).



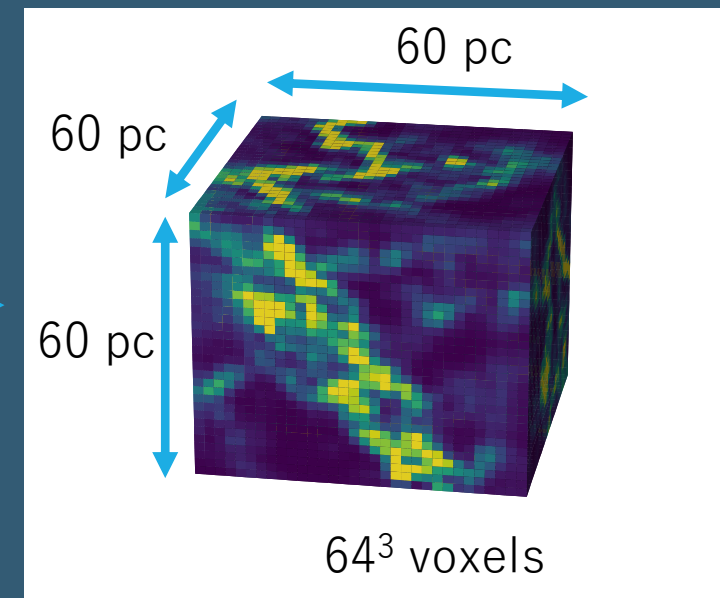
Ohana+(incl. KH),  
2024, NeurIPS



Interpolation using

SPH summation

$$\rho_i = \sum_j m_j W(r_{ij}, h_j)$$



# Convolutional Neural Network

Input

0	0	0	0
0	0	1	2
0	3	4	5
0	6	7	8

Kernel  
(Learnable)

0	1
3	4

\*

=

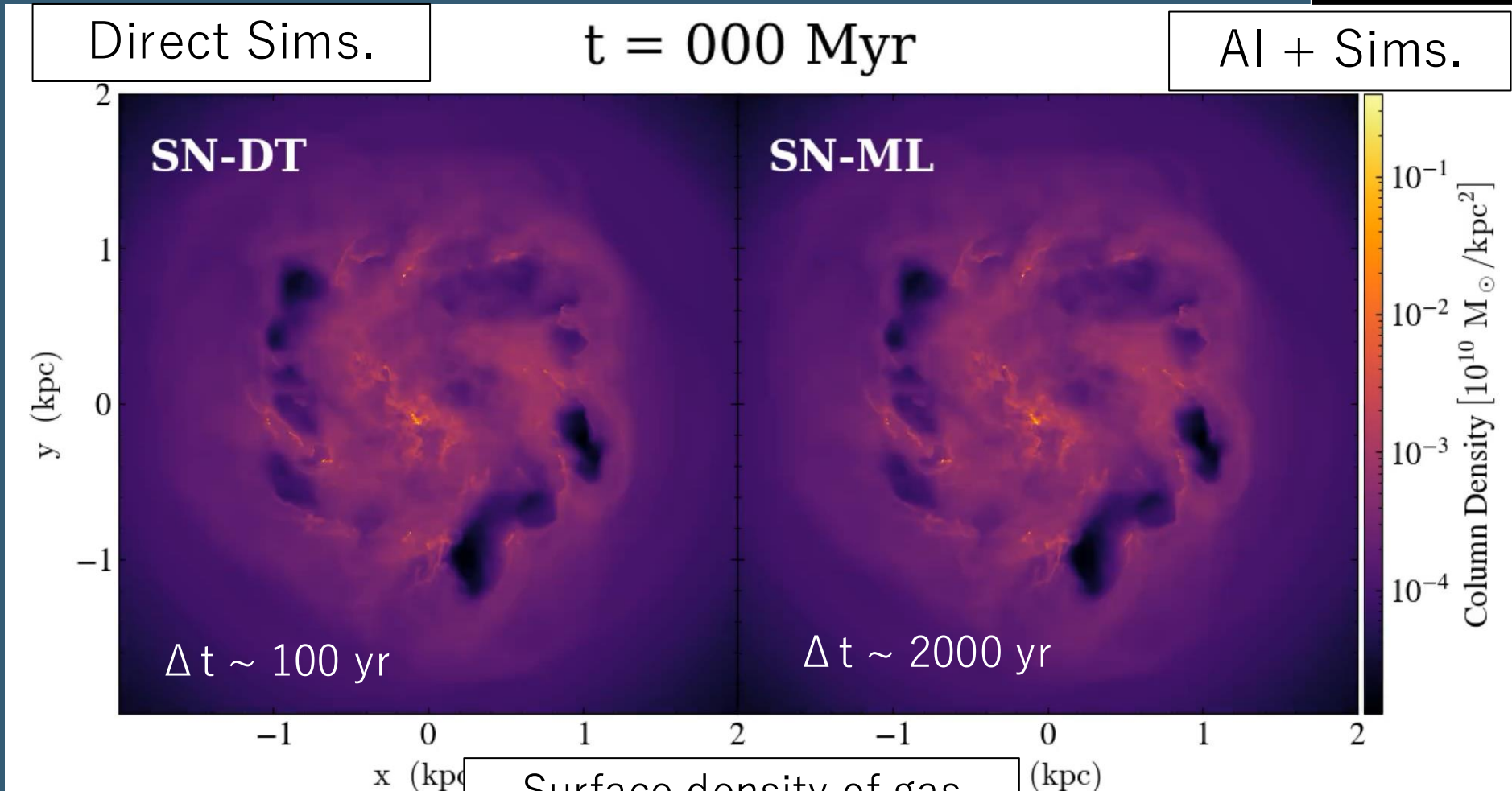
Output

0	4	11
12	26	34
27	50	58

# Galaxy Simulations with Surrogate Model

- Our surrogate model is used for SNe at  $> 1 \text{ cm}^{-3}$
- Tested on a dwarf galaxy ( $M_{\text{vir}} \sim 10^{10} M_{\odot}$  &  $m_{\text{baryon}} \sim 4 M_{\odot}$ )
- **x4 faster** and 6 months can be saved!

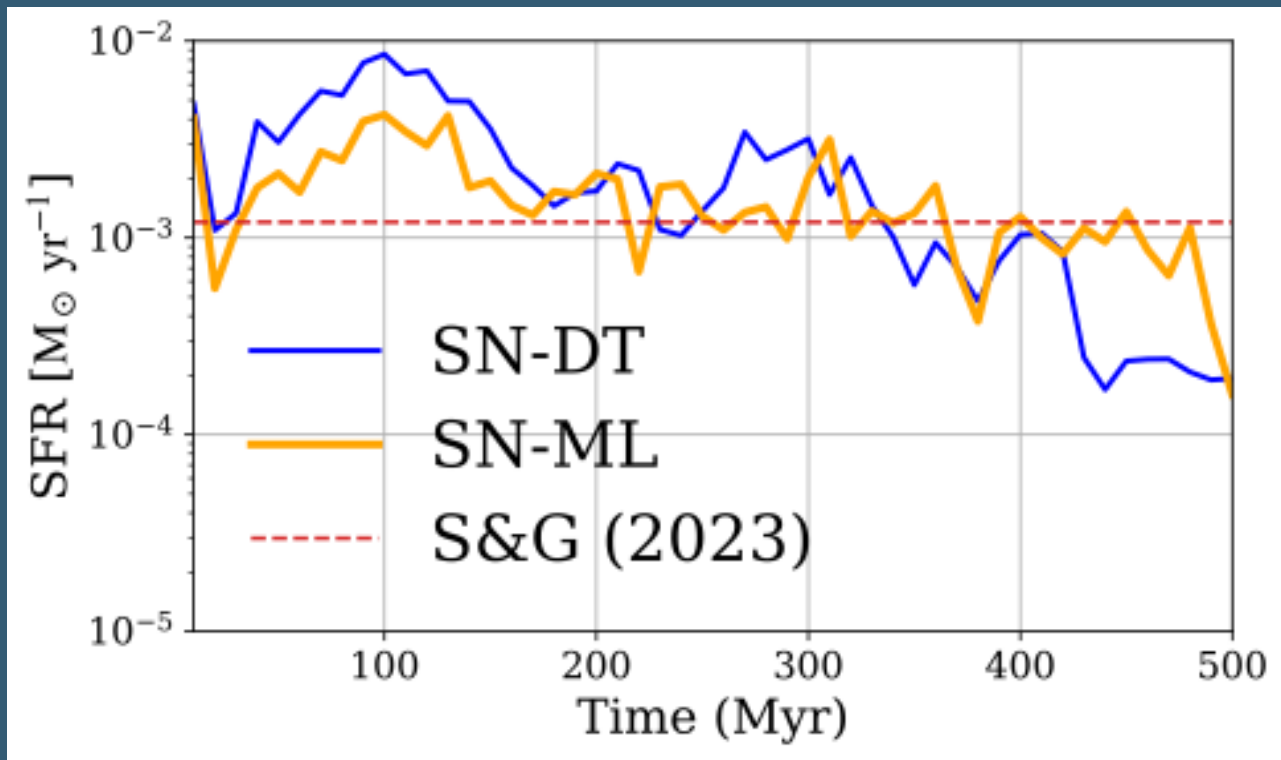
Hirashima+25a, ApJ



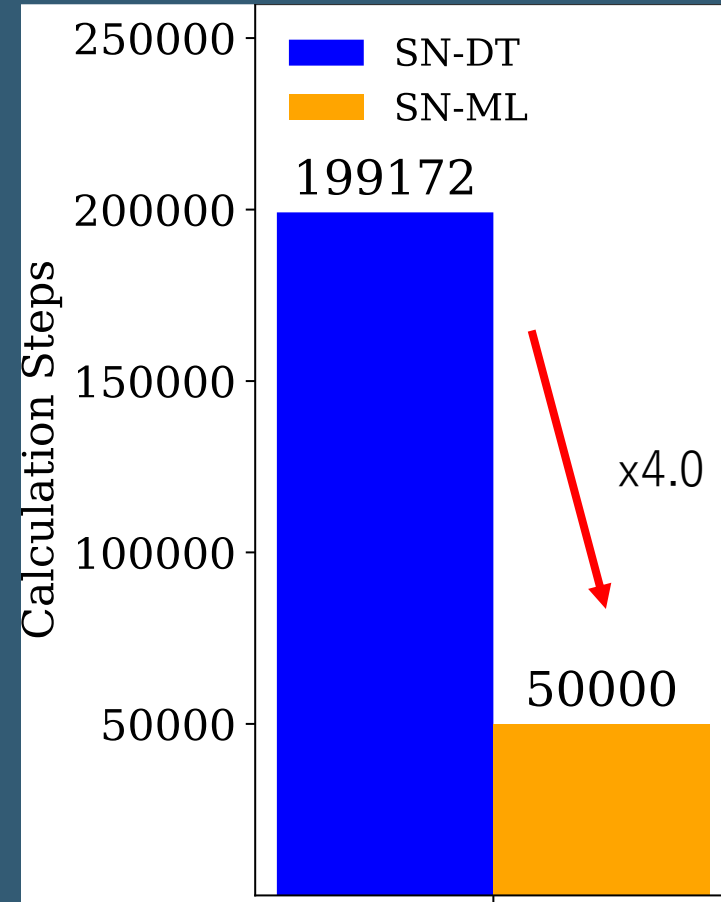
Video's Here!

# Validation

## Star Formation Efficiency



## X4 speed-up



8 months to 2 months

# Outflows

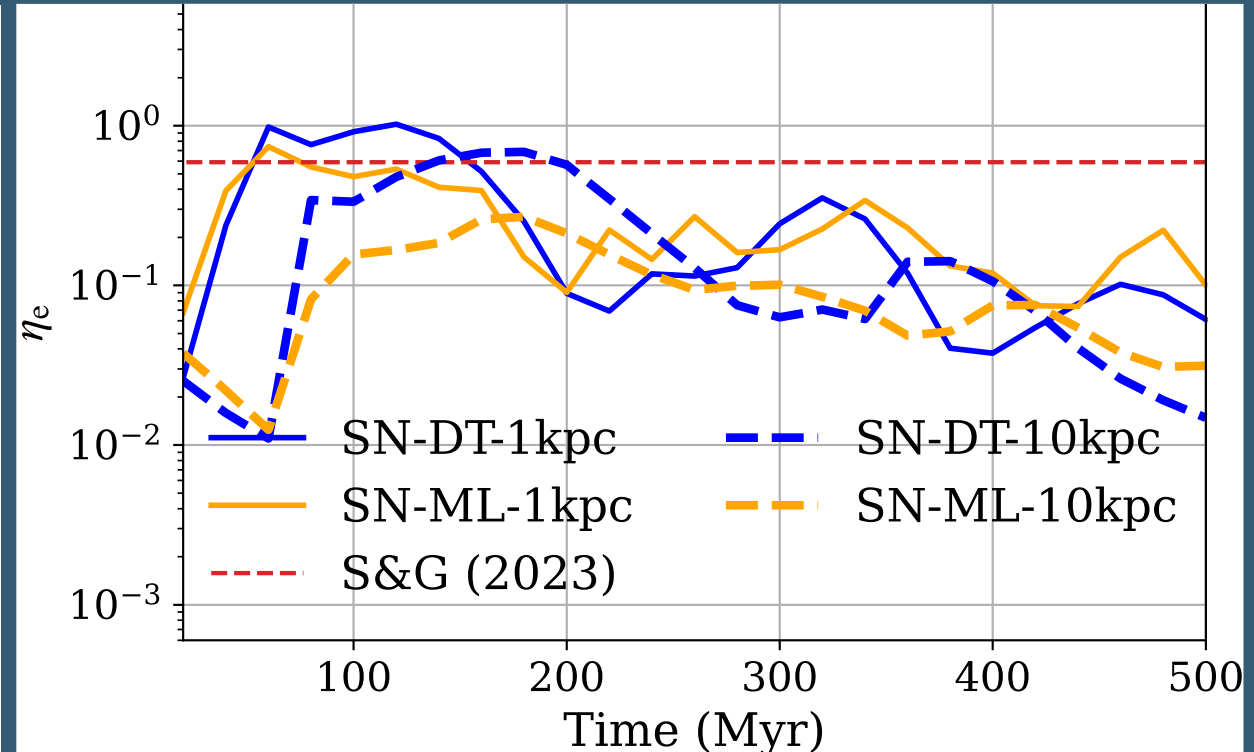
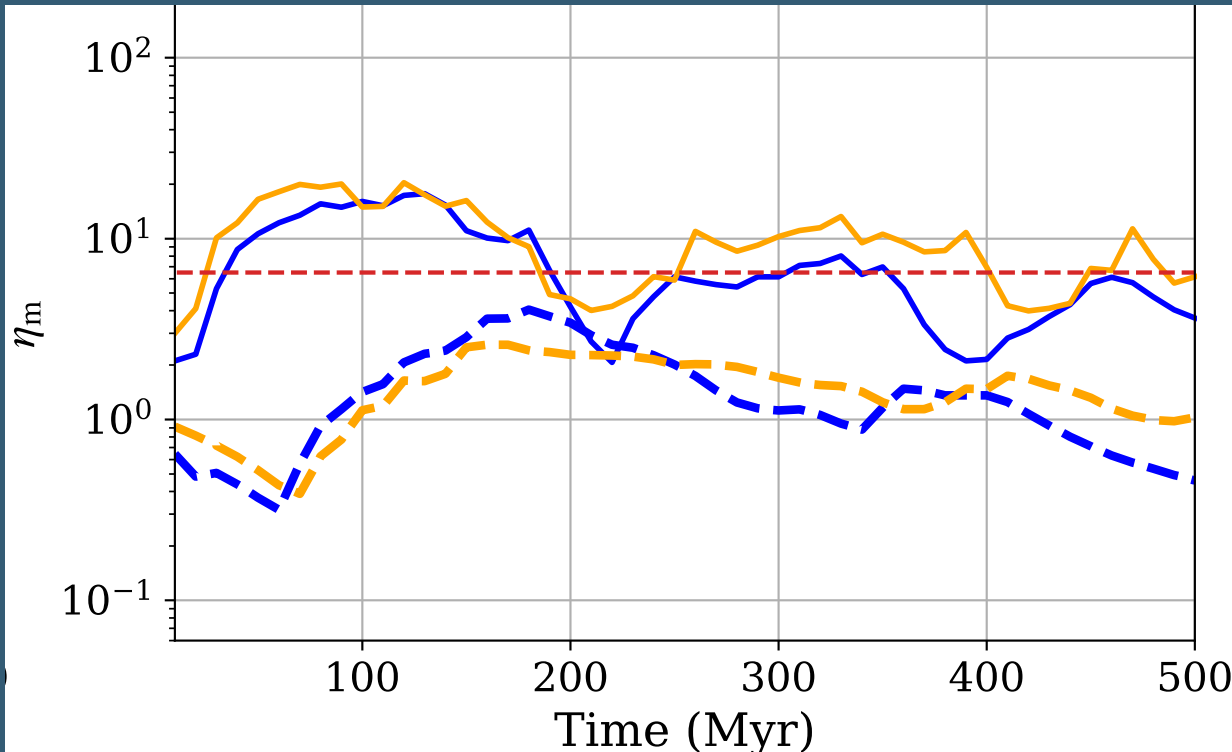
- Similar results between conventional and novel framework with AI.
- Only hot supersonic winds spout of the disk (Kim+20, Steinwandel+24)。

Mass loading factor

$$\eta_m^{\text{out}} = \dot{M}_{\text{out}} / \overline{\text{SFR}}$$

Energy loading factor

$$\eta_e^{\text{out}} = \dot{E}_{\text{out}} / (E_{\text{SN}} \overline{R_{\text{SN}}})$$



2025年11月21日

理化学研究所

神戸大学

筑波大学

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## 3,000億粒子の天の川銀河シミュレーションをAI×富岳で実現

一星一つ一つを再現する高解像度モデルで銀河進化に迫る

[English Page](#)

### The First Star-by-star $N$ -body/Hydrodynamics Simulation of Our Galaxy Coupling with a Surrogate Model

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- 7M CPU cores (Fugaku)
- High performance across Arm & x86 & CUDA
  - **Efficient on most commercial architectures and NVIDIA GPUs**
- Resolving the Milky Way down to 1 Msun.
- 36 yr- $\rightarrow$  3 months for 1 Gyr

Hirashima+25b, SC25

# Scaling Performance across multi-architectures

	# Nodes	Processor	Architecture	PFLOPS	Efficiency	#
Fugaku	148,900	A64FX	Armv8.2-A SVE	90.2	9.9%	$10^{11}$
Rusty	193	Genoa	AVX-512	0.863	35.5%	$10^{11}$
Miyabi	1024	GH200	CUDA	5.60	8.1%	$10^{10}$

- The full simulation could finish within half a year if each step takes less than 10 seconds.
- When using ~20k nodes, the communication overhead is significant.
- As the number of particles per process increases, the relative impact of communication decreases.
- Gravity kernel is heavy but scalable.

