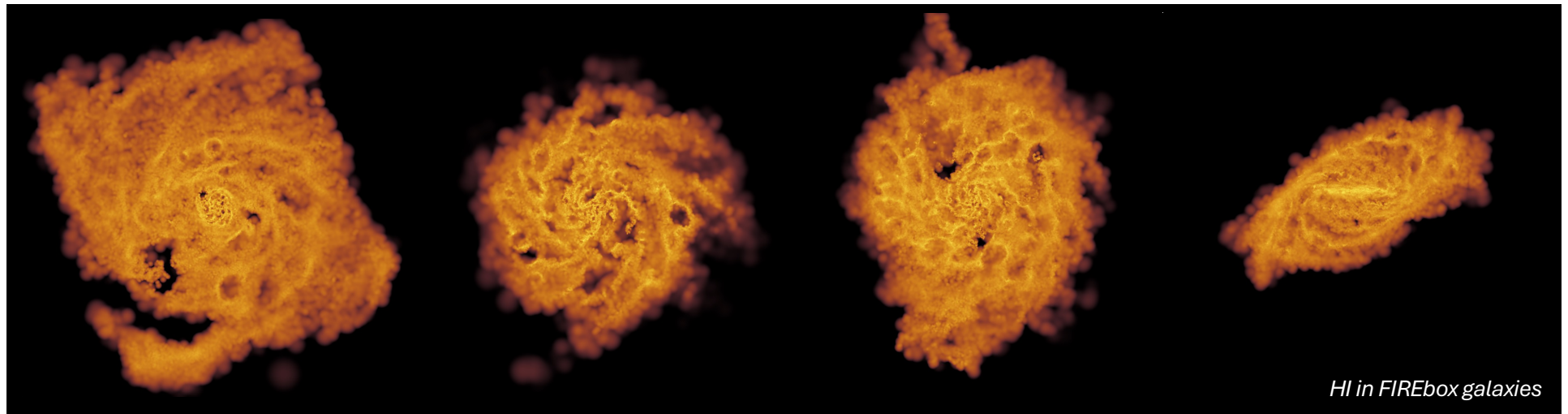
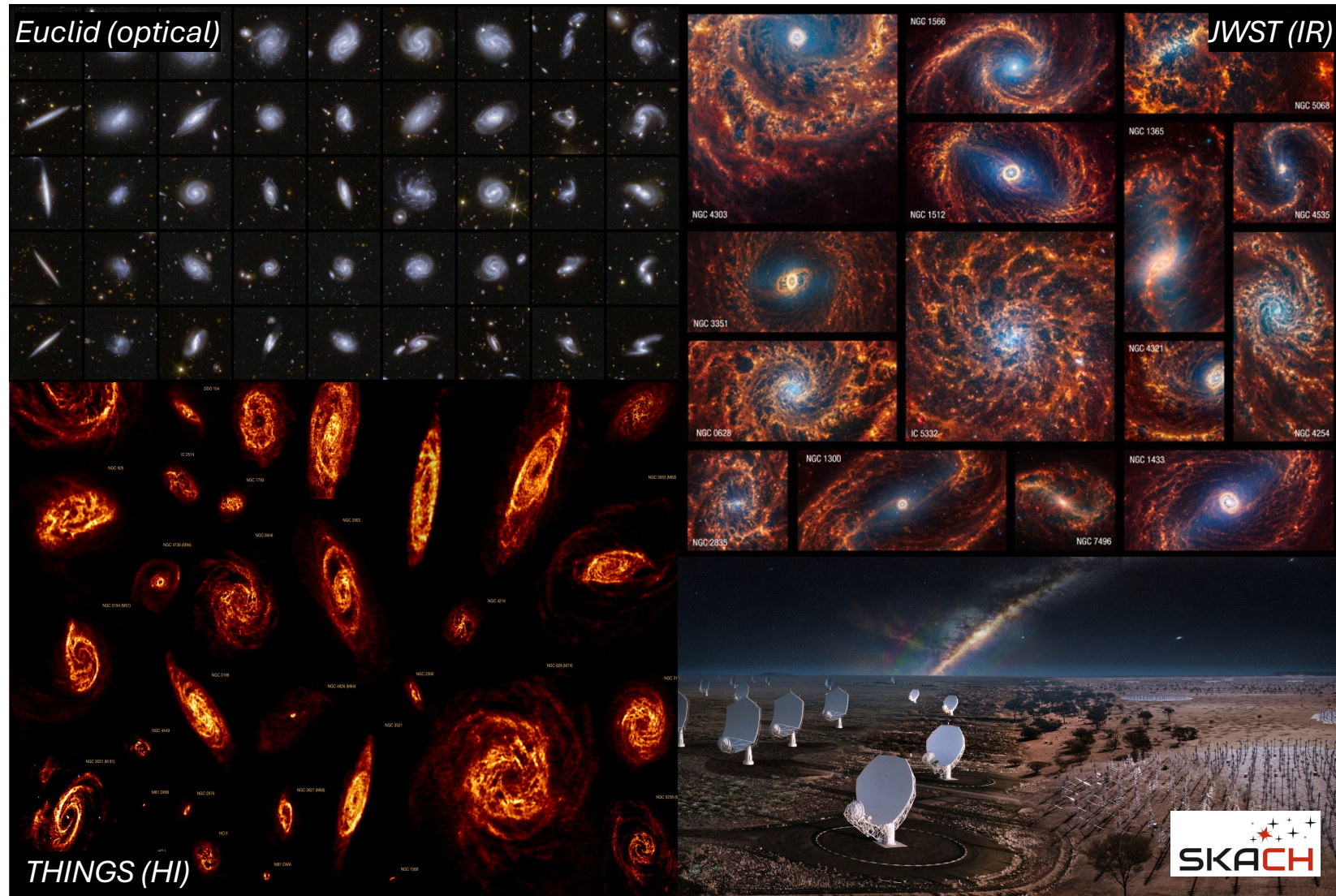


EMBER-2: towards implicit emulators for FIRE-based galaxy-formation simulations

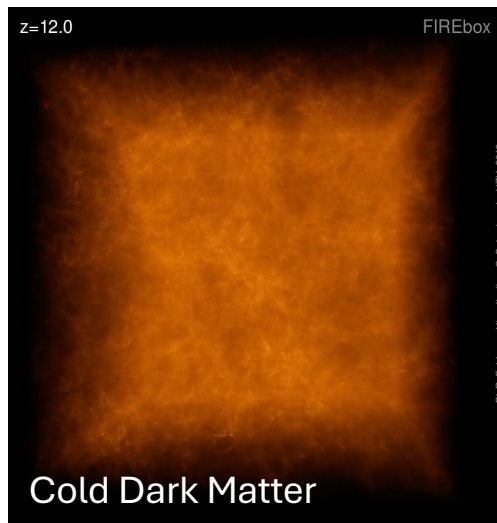


Mauro Bernardini



Current challenges of numerical modeling

Gravity



$$\frac{df}{dt} = \partial_t f + v \partial_r f - \partial_r \Phi \partial_v f = 0$$

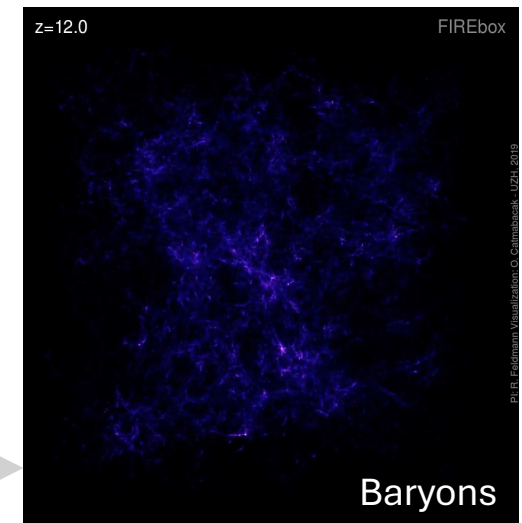
$$\Delta \Phi = 4\pi G \int f dv$$

$$\partial_t \rho + \nabla \cdot (\rho v) = 0$$

$$\partial_t (\rho v) + \nabla \cdot (\rho v \otimes v + P) = 0$$

$$\partial_t (\rho e) + \nabla \cdot (\rho e + P)v = 0$$

Hydrodynamics



Feldmann+23

„cheap“

ML

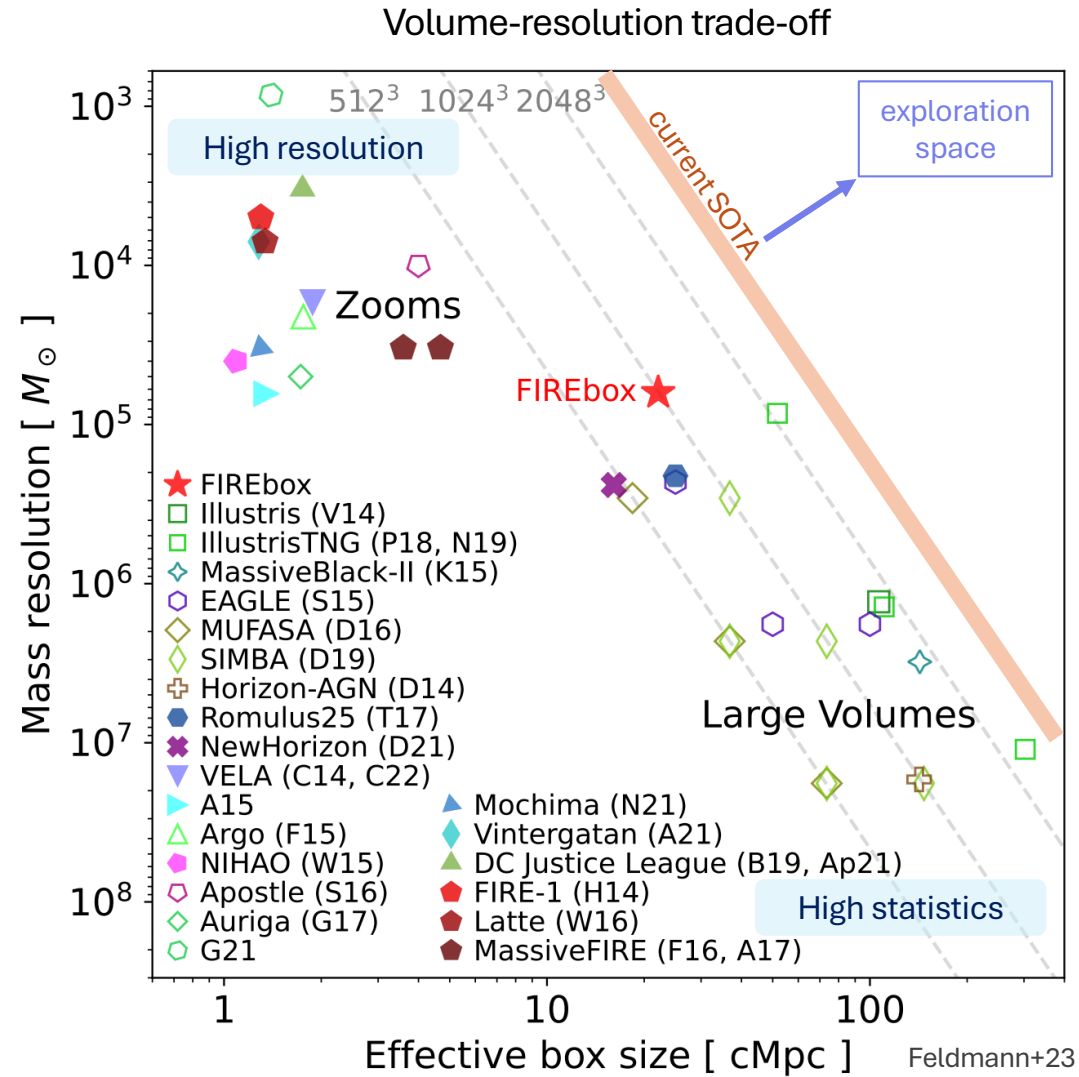
„costly“

„uncertain“

Gas cooling	Interstellar medium	Star formation	Stellar feedback	Supermassive black holes	Active galactic nuclei	Magnetic fields	Radiation fields	Cosmic rays
Atomic/ molecular/ metals/ tabulated network	Effective equation of state/ multiphase	Initial stellar mass function/ probabilistic sampling/ enrichment	Kinetic/ thermal/ variety of sources from stars, supernovae	Numerical seeding/ growth by accretion prescription/ merging	Kinetic/ thermal/ radiative/ quasar model radio mode	Ideal MHD/ cleaning schemes/ constrained transport	Ray tracing/ Monte Carlo/ moment based	Production/ heating/ anisotropic diffusion/ streaming
Stellar physics				BH + AGN		MHD		

Vogelsberger+20

Simulation costs!

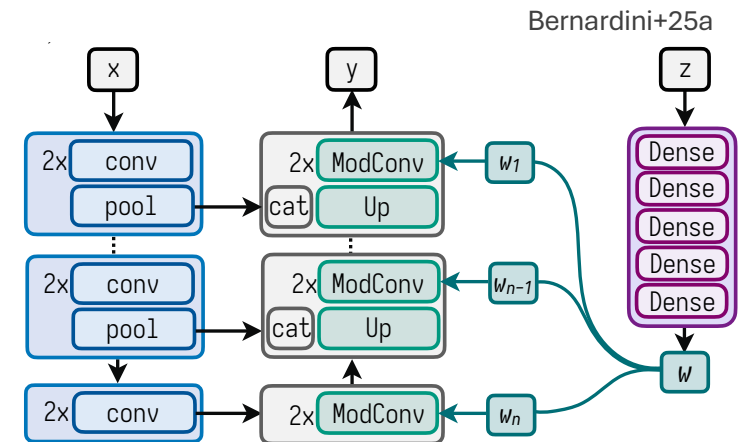


EMBER

EMulating Baryonic EnRichment

Architecture:

- StyleGAN-2 inspired NN
- Emulates gas fields from dark matter maps
- Across large redshift range (0-6)
- Fully convolutional, thus scalable



Designed for:

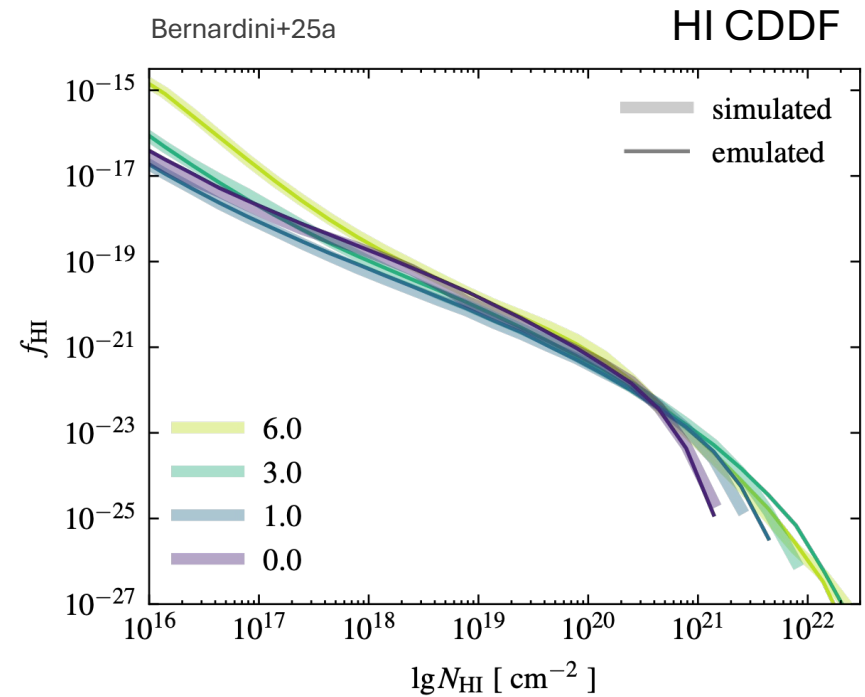
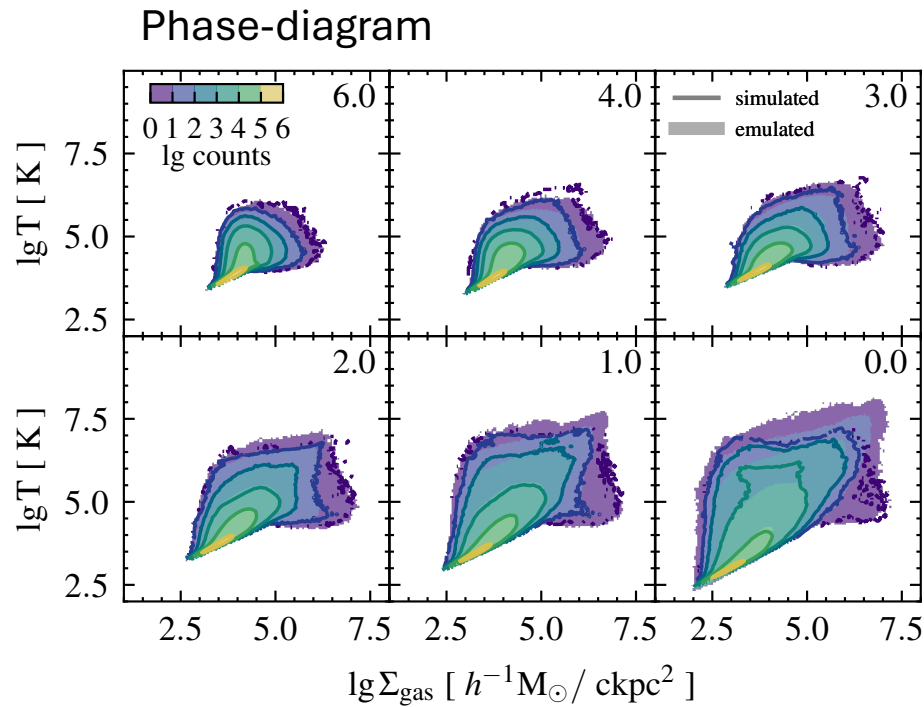
- (1) Domain translation
 - (2) Resolution / Volume upscaling
 - (3) Field-level emulation conditioned on different astrophysical models
- } *This talk!*

→ **Caveat:** No physics-constraints in the model; purely data-driven!

EMBER *dark matter* $\rightarrow$ *baryons*

Key idea: Train to emulate **baryon** from **dark matter** maps

$$f : (\Sigma_{\text{dm}}, v_{\text{dm}}) \rightarrow (\Sigma_{\text{gas}}, v_{\text{gas}}, T_{\text{gas}}, \Sigma_{\text{HI}})$$

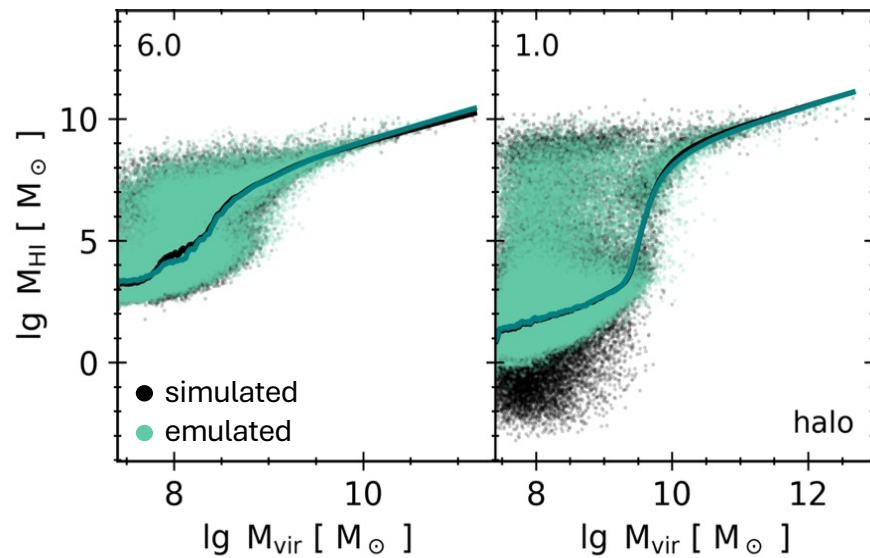


EMBER *dark matter* $\rightarrow$ *baryons*

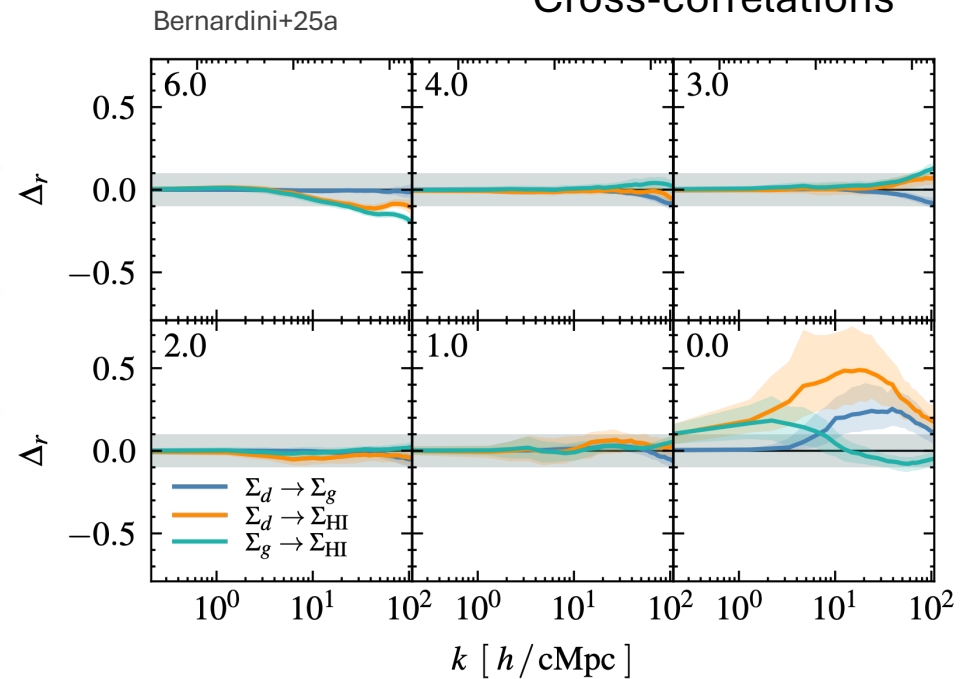
Key idea: Train to emulate **baryon** from **dark matter** maps

$$f : (\Sigma_{\text{dm}}, v_{\text{dm}}) \rightarrow (\Sigma_{\text{gas}}, v_{\text{gas}}, T_{\text{gas}}, \Sigma_{\text{HI}})$$

HM-HI relation (halo-based)



Cross-correlations



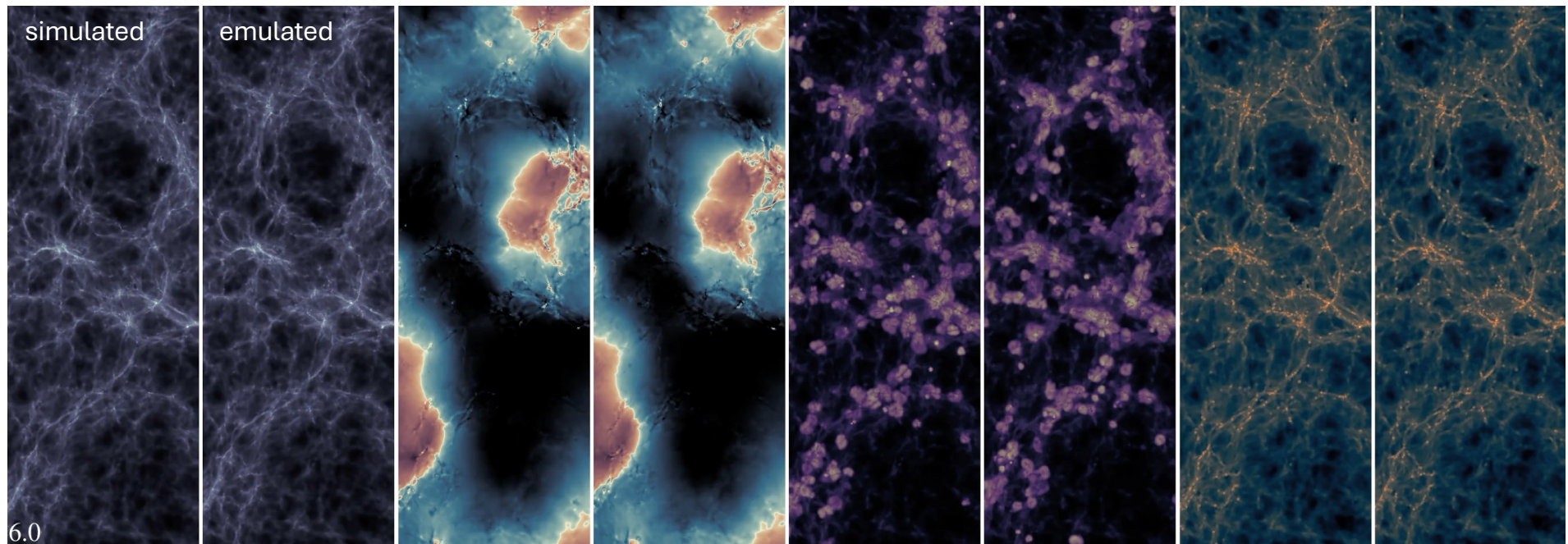
EMBER *dark matter* $\rightarrow$ *baryons*

Gas Density

Radial velocity

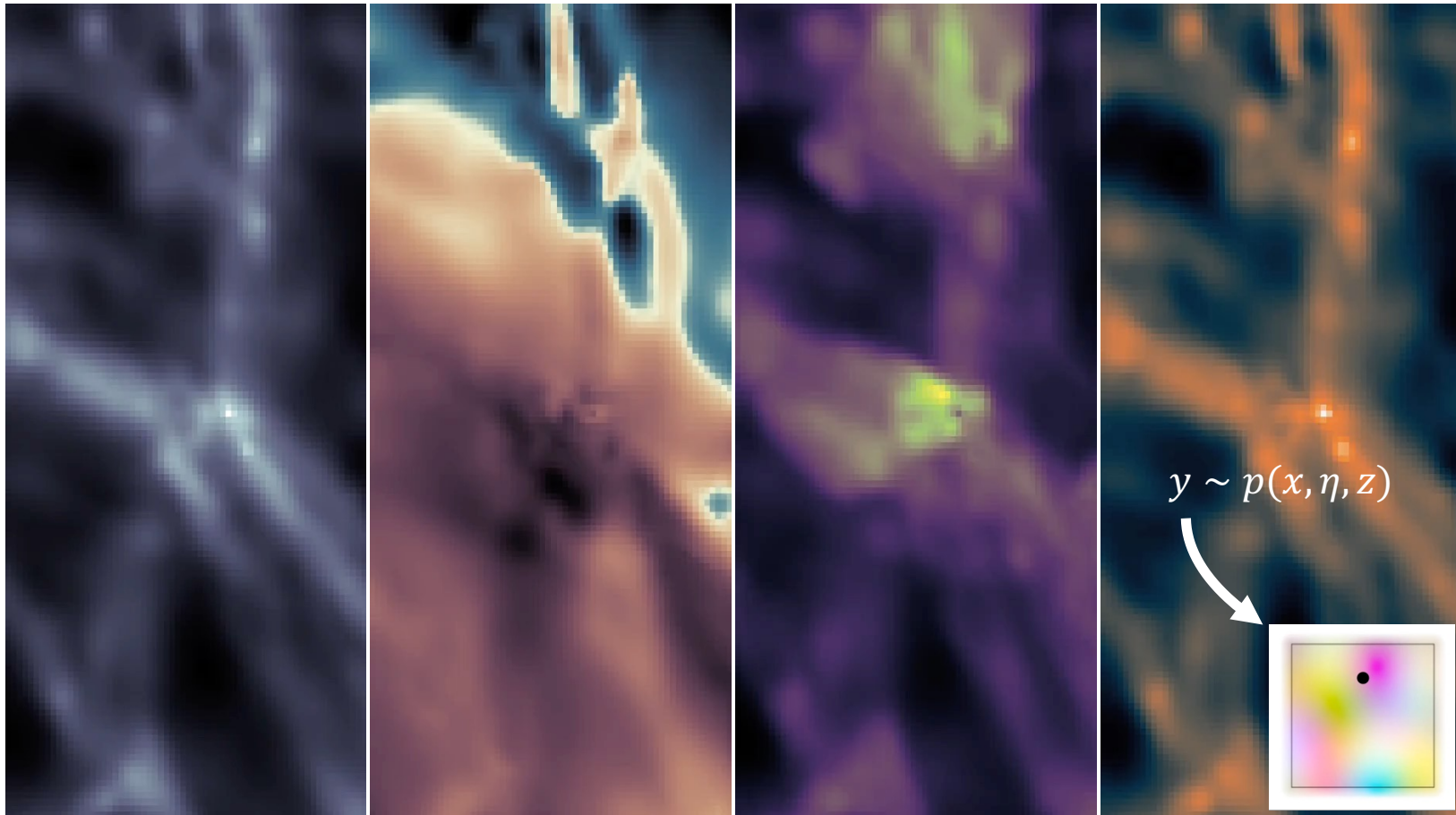
Temperature

HI



Bernardini+25a

EMBER *dark matter* $\rightarrow$ *baryons*



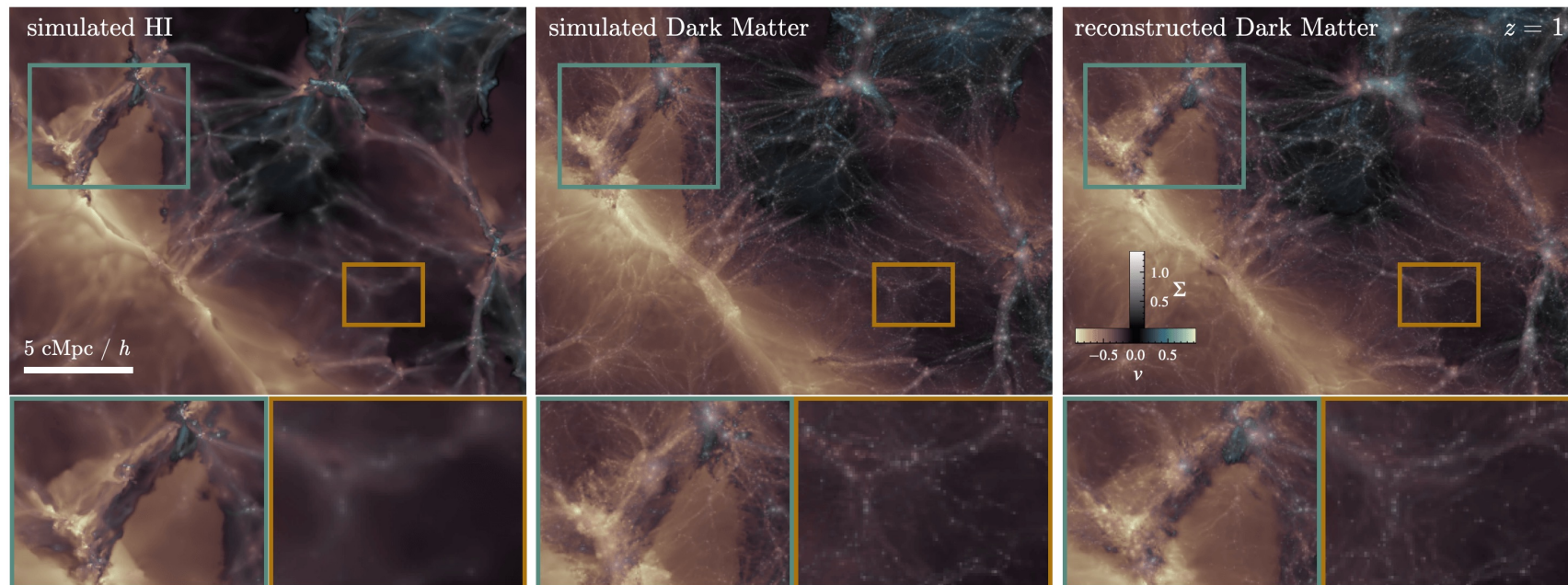
EMBER: $HI \rightarrow dark\ matter$



Key idea: Train to predict **dark matter** from **HI** maps

$$f : (\Sigma_{HI}, v_{HI}) \rightarrow (\Sigma_{dm}, v_{dm})$$

Bernardini+25b



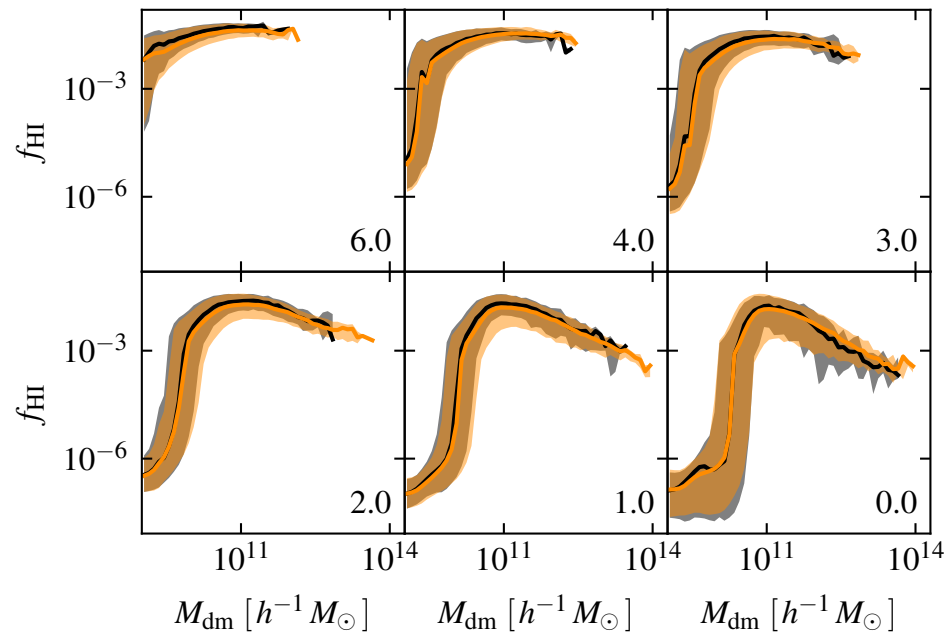
→ Can be integrated into downstream analysis pipelines of HI observations

EMBER $HI \rightarrow$ dark matter

Key idea: Train to predict **dark matter** from **HI** maps

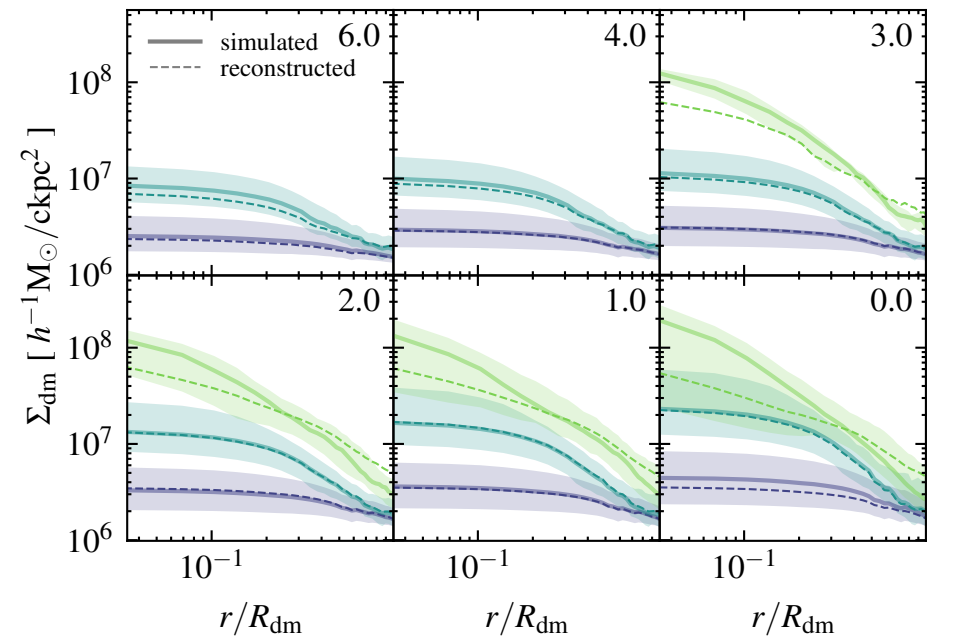
$$f : (\Sigma_{HI}, v_{HI}) \rightarrow (\Sigma_{dm}, v_{dm})$$

HI-DM fractions



Radial profiles

Bernardini+25b



— $9 \leq \lg M_{dm} < 10.5$ — $10.5 \leq \lg M_{dm} < 12.5$ — $12.5 \leq \lg M_{dm} < \text{inf}$

EMBER *upsampling simulation outputs*

Training:

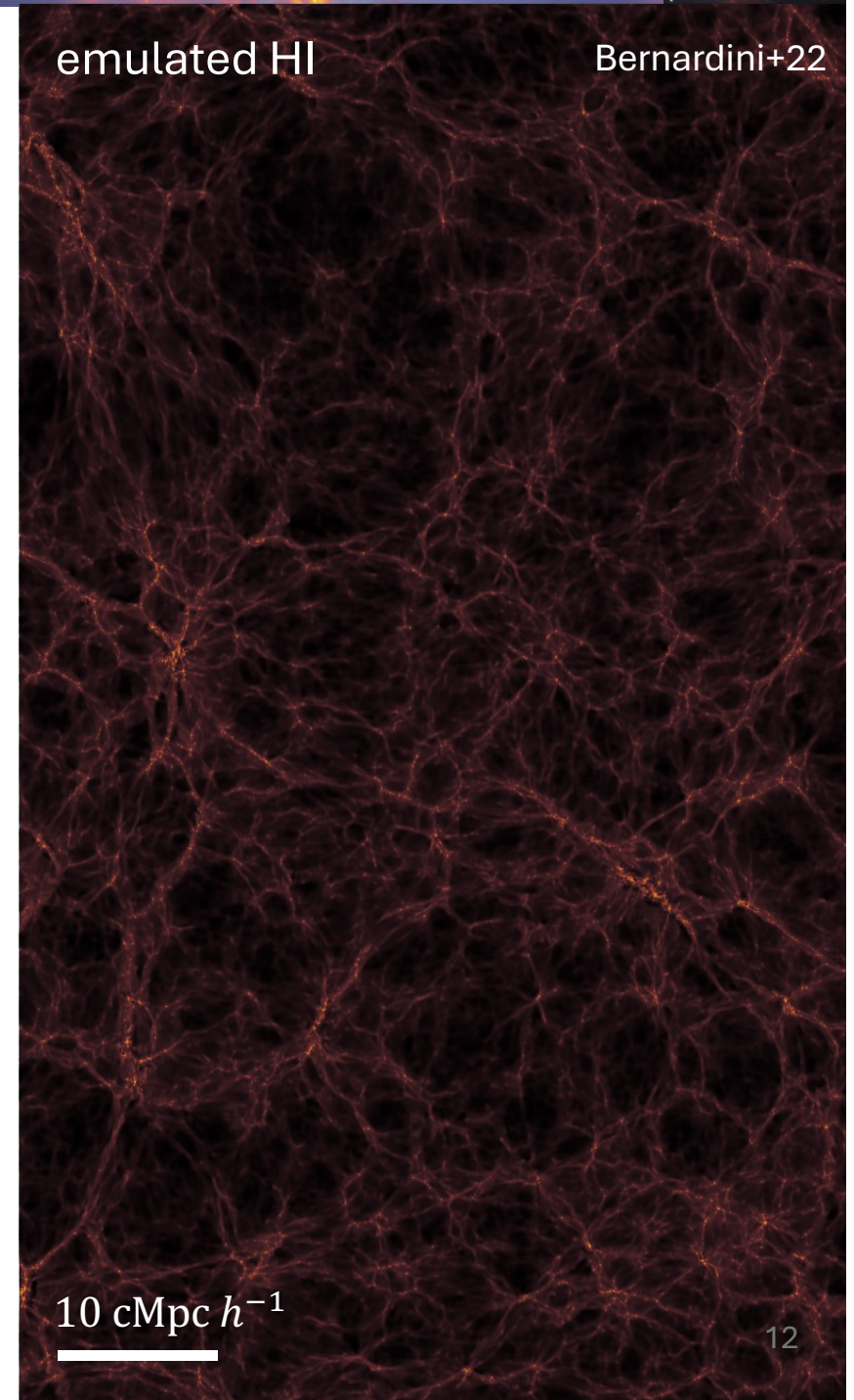
- Training on small high-resolution volumes and high-resolution zoom-in simulations
- Degrading the input dark matter resolution
- Training on paired maps of **low-res DM $\rightarrow$ high-res baryon**
- Upscalable due to fully convolutional nature

Application:

- „Emulate“ high resolution baryon fields onto much larger volumes (here 300x)

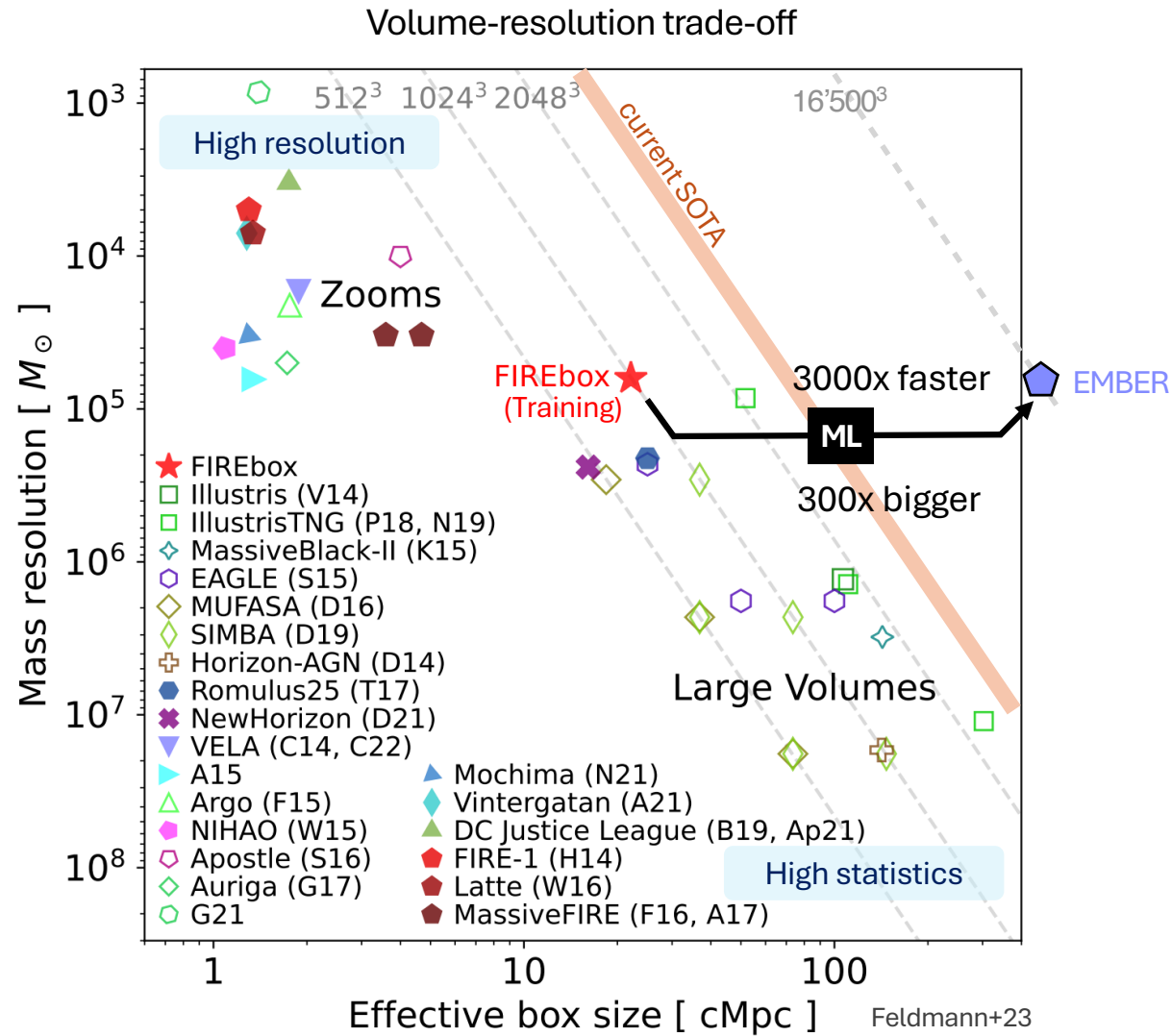
emulated HI

Bernardini+22



10 cMpc h^{-1}

Simulation costs!

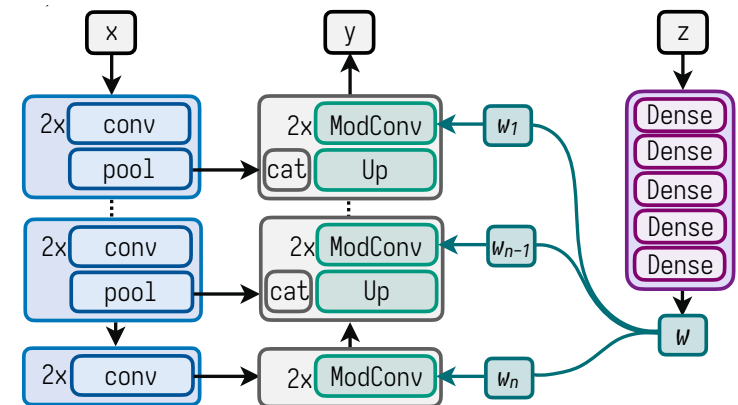


EMBER

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Designed for:

- (1) Domain translation ✓
- (2) Resolution / Volume upscaling ✓

Based on:

Bernardini+22 (2110.11970)
Bernardini+25a (2502.15875)
Bernardini+25b (to appear)

Thank you!