# Inferring the assembly and merger histories of galaxies

#### with the IllustrisTNG simulations and machine learning

Based on Eisert+ in prep.

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Or





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MERGE or



## We can only observe that one!

### What can this observation tell us about the assembly history?





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# We can only observe that one!





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#### "EASY":

Observable morphology and/or properties of the bright main galaxy body

#### e.g. SDSS





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Observable faint stellar halo surrounding them

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The problem...

### We can only observe galaxies at one specific point in time...



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How to learn something about their histories?

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### We can only observe galaxies at one specific point in time...



### A solution...

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How to learn something about their histories?

- Study galaxies via (full) cosmological simulations!
- Get observables and the history for each simulated galaxy and train a Neural Network to it!

### The IllustrisTNG Simulations (2016-2019)

- Annalisa Pillepich (Co-PI: TNG50) Dylan Nelson (Co-PI: TNG50)
- Federico Marinacci
- Jill Naiman
- Lars Hernquist
- Mark Vogelsberger
- Ruediger Pakmor
- Paul Torrey
- Shy Genel

CFA III T

Max-Planck-Institut

für Astrophysik

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• Volker Springel (PI)

ING5

TNG10

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S R High-Performance Computing Center | Stuttga

GCS

MAX PLANCK COMPUTING & DATA FACILITY

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IM GARCHING DER MAX-PLANCK-GESELLSCHAFT

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00 Mpc

2D matter distribution



esp. merger histories i.e.

- Average mass ratios of mergers
- Average time of mergers
- Time of latest major merger
- Fraction of ex-situ stars

### **Cosmological Simulations** (e.g. IllustrisTNG)

### Potential loss of information



# Galaxy properties (observables) c

Integral properties:

- Stellar mass
- Redshift
- Diskyness
- Galaxy size
- Average stellar age
- Integrated Color
- Stellar Metallicity











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### Approach #1: Scalars to scalars

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### Approach #2: Maps to scalars

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#### Galaxy stellar maps:



esp. merger histories i.e.

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### Approach #1: Scalars to scalars

Look for the posterior distribution p(x|c)to account for ambiguities connected to such inverse Problems!

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### **Conditional Invertible Neural Network (CINN)**





### Store information into z which would be otherwise lost!

### **Conditional Invertible Neural** Network (cINN)



### Training with Maximum Likelihood Loss

$$\mathcal{L} = \mathbb{E}_{i} \left[ \frac{||f(\mathbf{x}_{i}; \mathbf{c}_{i}, \theta)||_{2}^{2}}{2} - \log |J_{i}| \right]$$



### **Conditional Invertible Neural Network (CINN)**





### **Conditional Invertible Neural Network (CINN) Posteriors**

Mean Merger Lookbacktime [Gyr]

Prior Posterior **Ground Truth** Max. a posteriori estimate (MAP)





















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### 1 row = 1 example ga

#### Lookback Time of Last Major Merger [Gyr]

al	axy



TNG100 Redshift <= 1 Log Total Stellar Mass > 10.0 ~ 150.000 Training galaxies ~ 18.000 Test galaxies Logarithmic colorbar

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#### NMM = No Major Merger

Lukas Eisert, ML-IAP, 2021/10/21



10<sup>0</sup>



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# Summary & Outlook

- We are able to construct well behaving posteriors for the average merger lookback time and stellar exsitu fractions from a simple set of scalar observables!
- In the full paper:
  - Convergence Analysis
  - Closer look at the mass ratios and the lookback time of the latest major merger
  - Sensitivity Analysis
- Approach #2: Use Light Maps as input
- Get representations via contrastive learning
- Infer more parameters; maybe even the merger tree or maps of unobservable physical quantities
- Apply the models to real observational data!











# Supplementary

Average Merger Lookback Time

Average Merger Mass Ratio

Lookback Time Of Latest Major Merger

Total Stellar Exsitu Fraction

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Absolute MAP Error





#### Embedding of the galaxies into a 2D representation with UMAP

