

DEBATING THE POTENTIAL  
OF MACHINE LEARNING  
IN ASTRONOMICAL SURVEYS #2

November 27th - December 1st 2023  
IAP, Paris / Flatiron institute, New York

# Data-driven galaxy morphologies at $z > 3$ with contrastive learning and cosmological simulations

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<https://arxiv.org/abs/2302.07277v2>  
(Accepted in ApJ)

Collaborators

**L. Costantin, P.G. Pérez-González, R. Sarmiento, A. Pillepich**

+ JWST/CEERS collaboration + TNG collaboration + external collaborators



Universidad de Valladolid



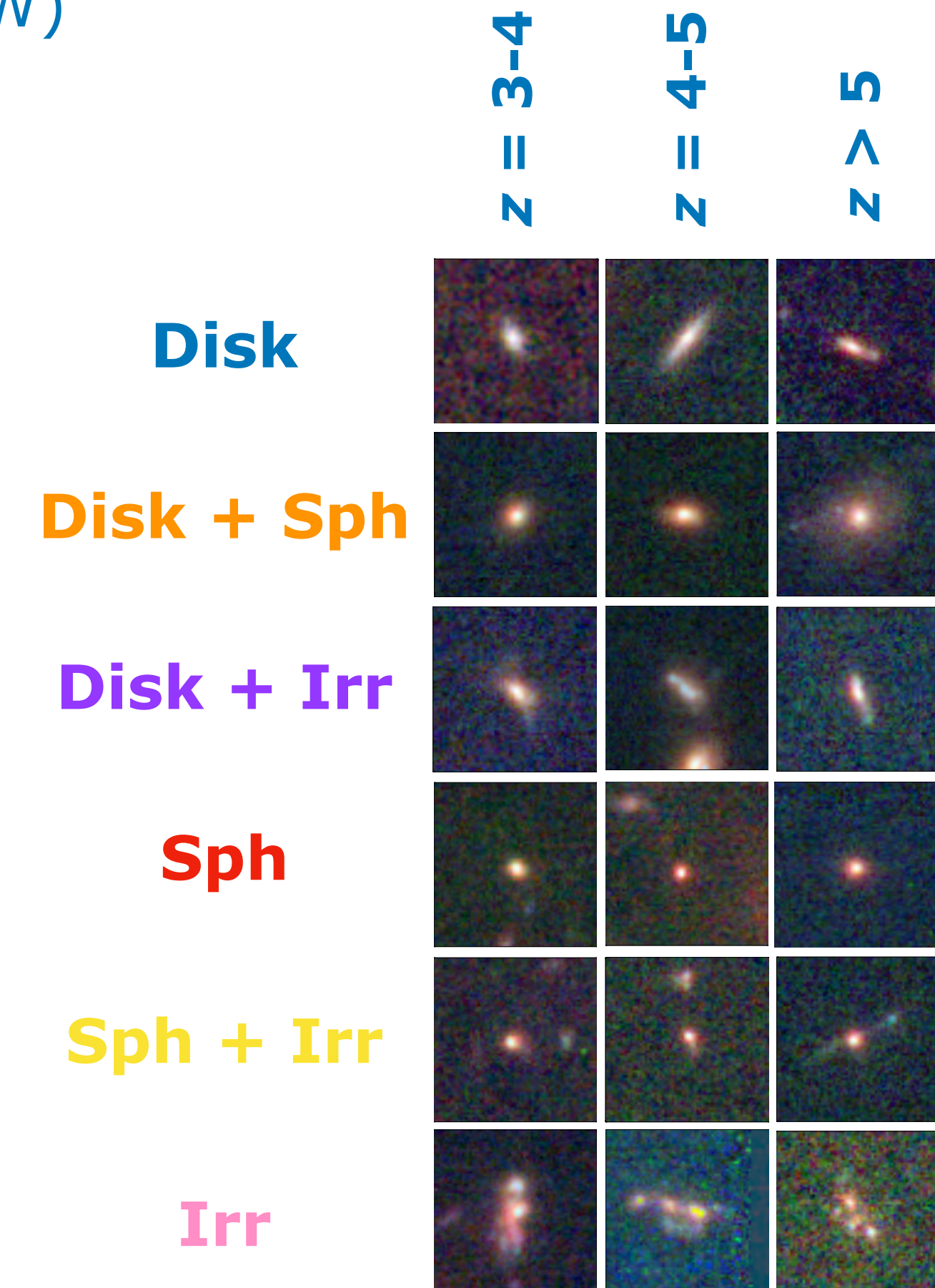
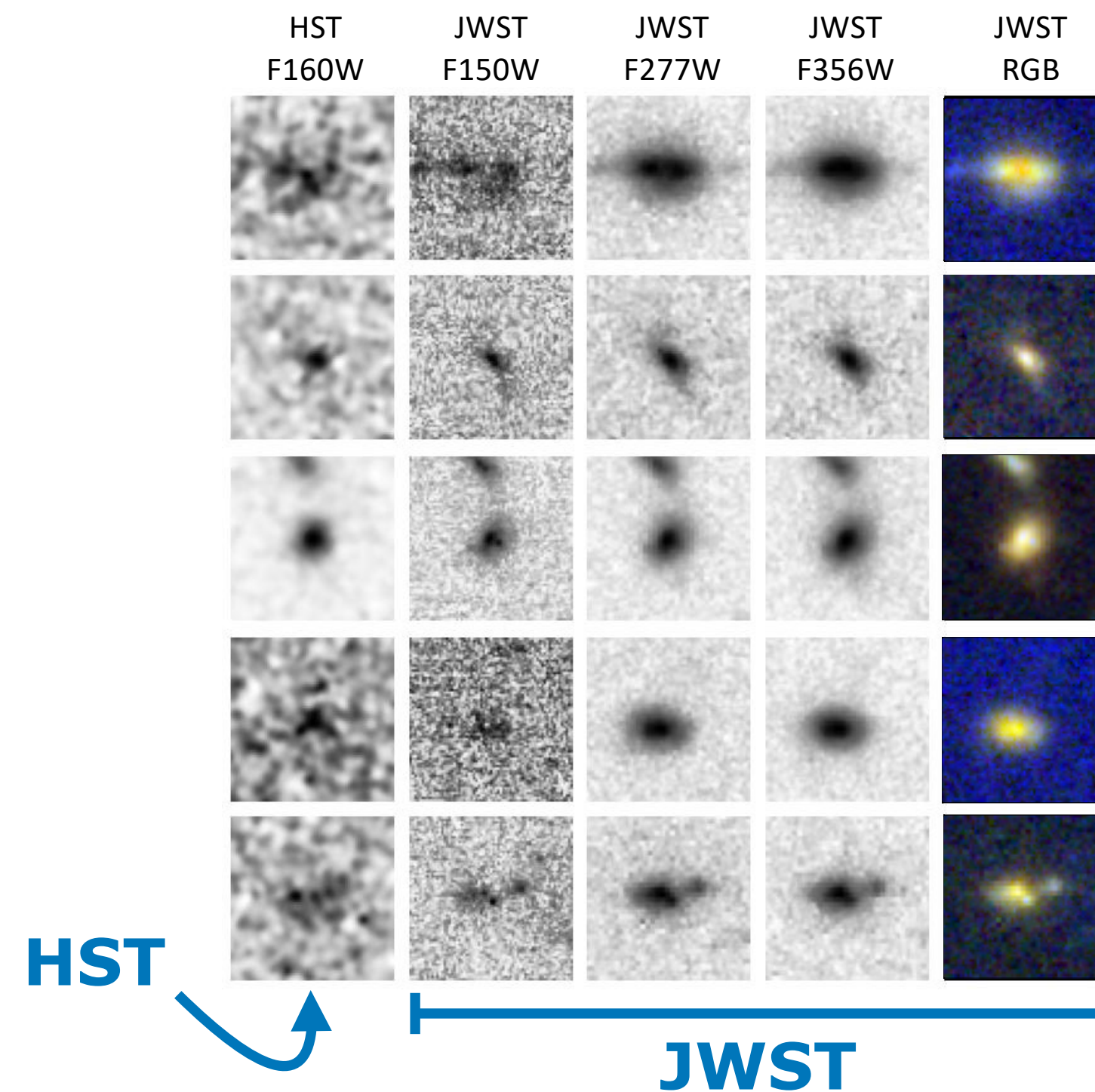
# Data-driven galaxy morphologies at $z > 3$

## JWST: a galaxy morphology machine at high- $z$

☉ **High spatial resolution:** 0.03"/px (SW) and 0.06"/px (LW)

☉  $\sim 0.5$  kpc resolution at  $z \sim 3$

☉ **Optical rest-frame at  $z > 3$**



*Credits: Kartaltepe et al. 2023*

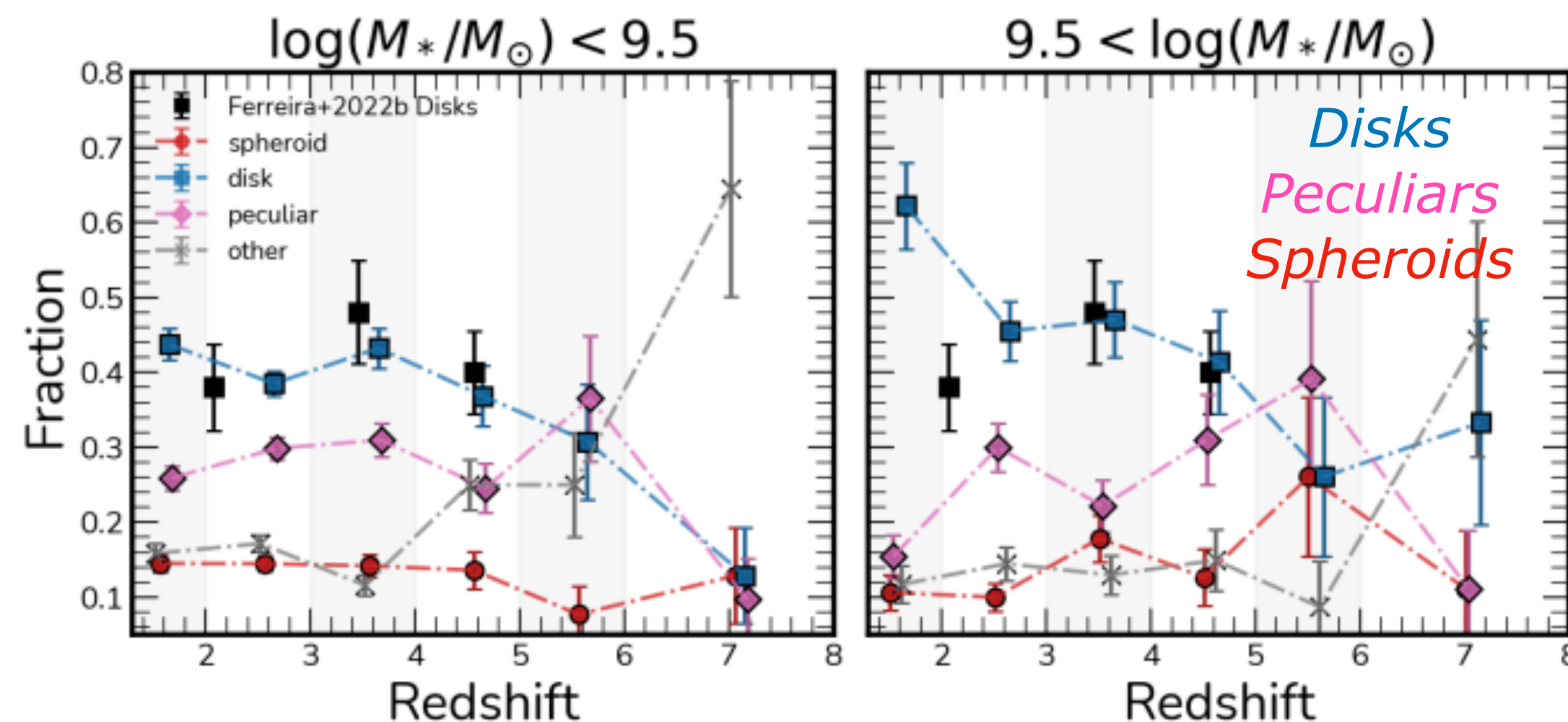
# Data-driven galaxy morphologies at $z > 3$

## JWST: a galaxy morphology machine at high- $z$

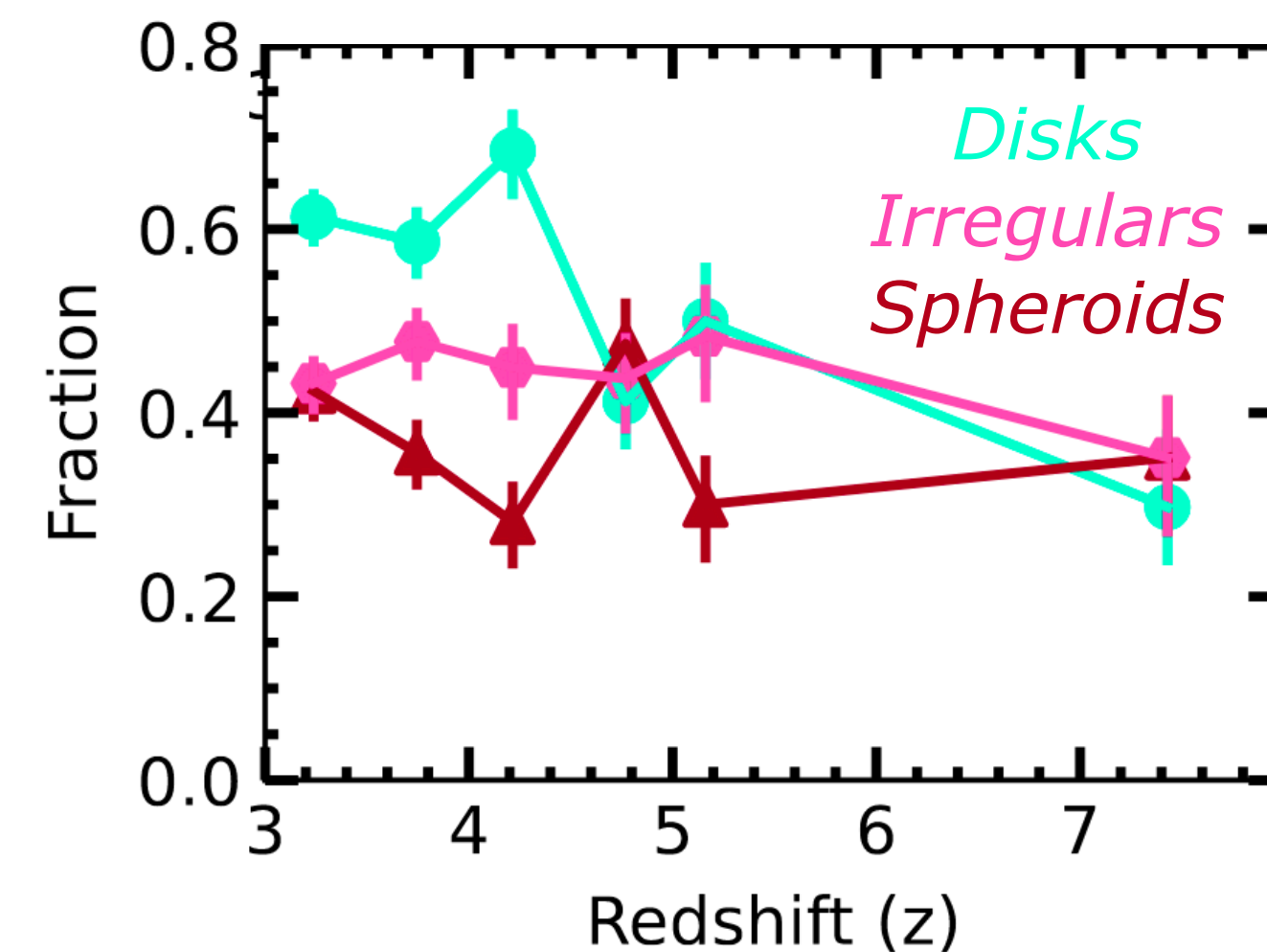
☉ First JWST observations: **disk galaxies in place very early on**

(Kartaltepe et al. 2023, Ferreira et al. 2023, Robertson et al. 2023, Huertas-Company et al. 2023b)

### Visual classifications



Ferreira et al. 2023



Kartaltepe et al. 2023

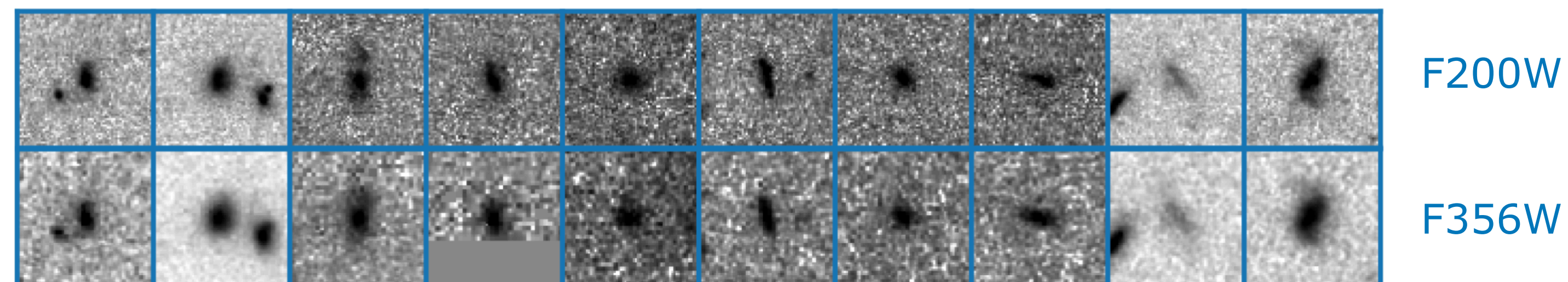
## Morphological classification of high- $z$ JWST galaxies

⌚ **Visual classifications are less reliable at high- $z$**

PSF, cosmological dimming, noise, contaminants,...

⌚ **'No secure' labelling** of high- $z$  galaxies, even at depths and resolution of JWST

### Disks at $z > 3$ (*Kartaltepe et al. 2023*)



**Can we represent galaxy images in a space that is robust to these perturbations?**

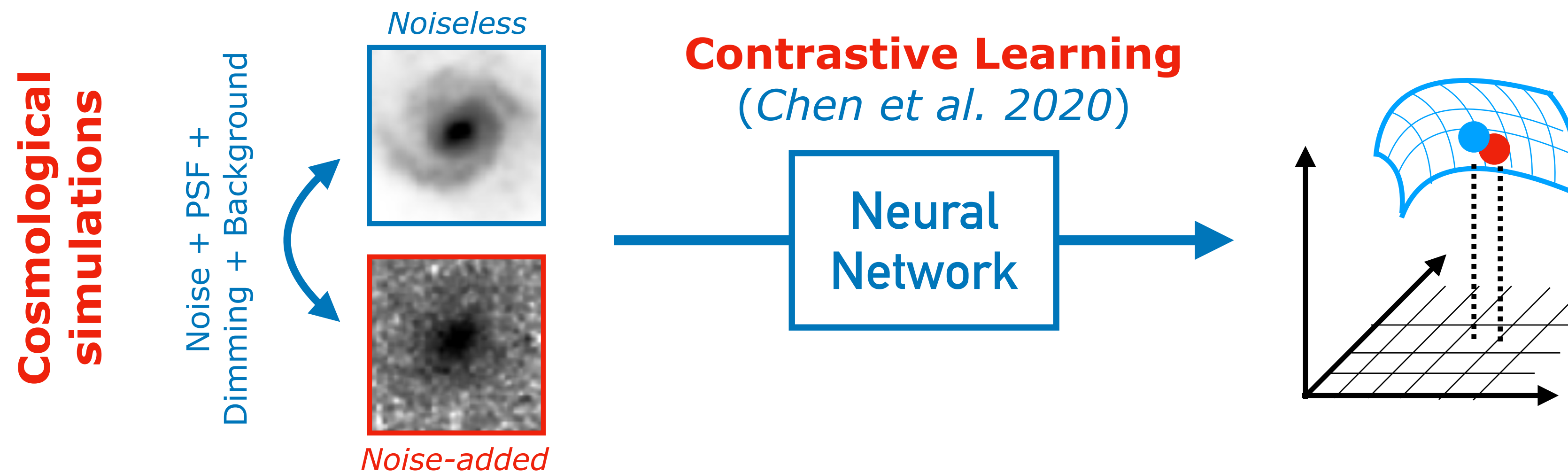
**Can we say more about the nature of these objects?**

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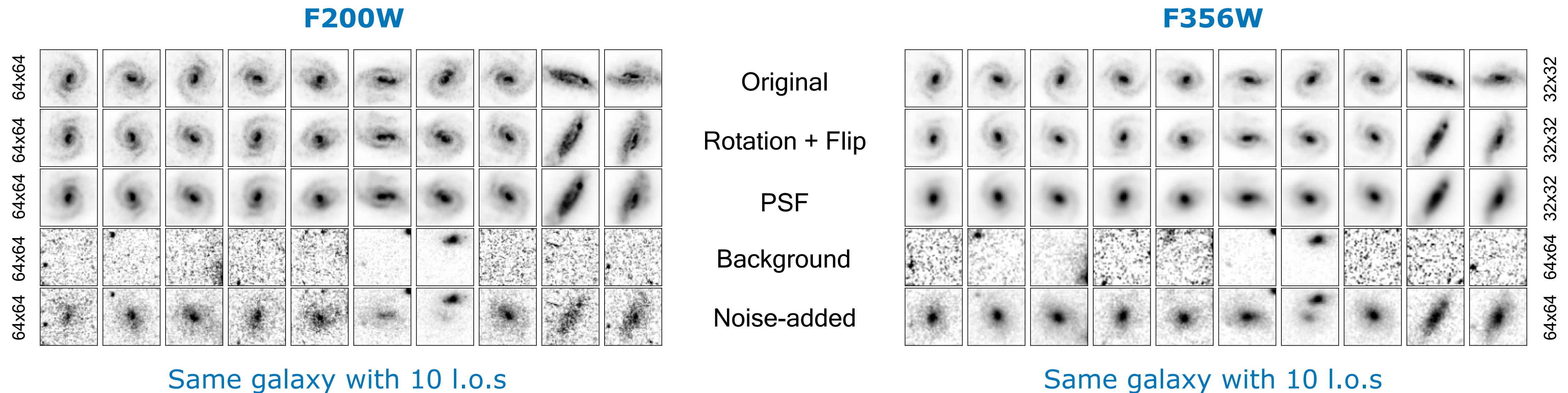
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# Data-driven galaxy morphologies at $z > 3$

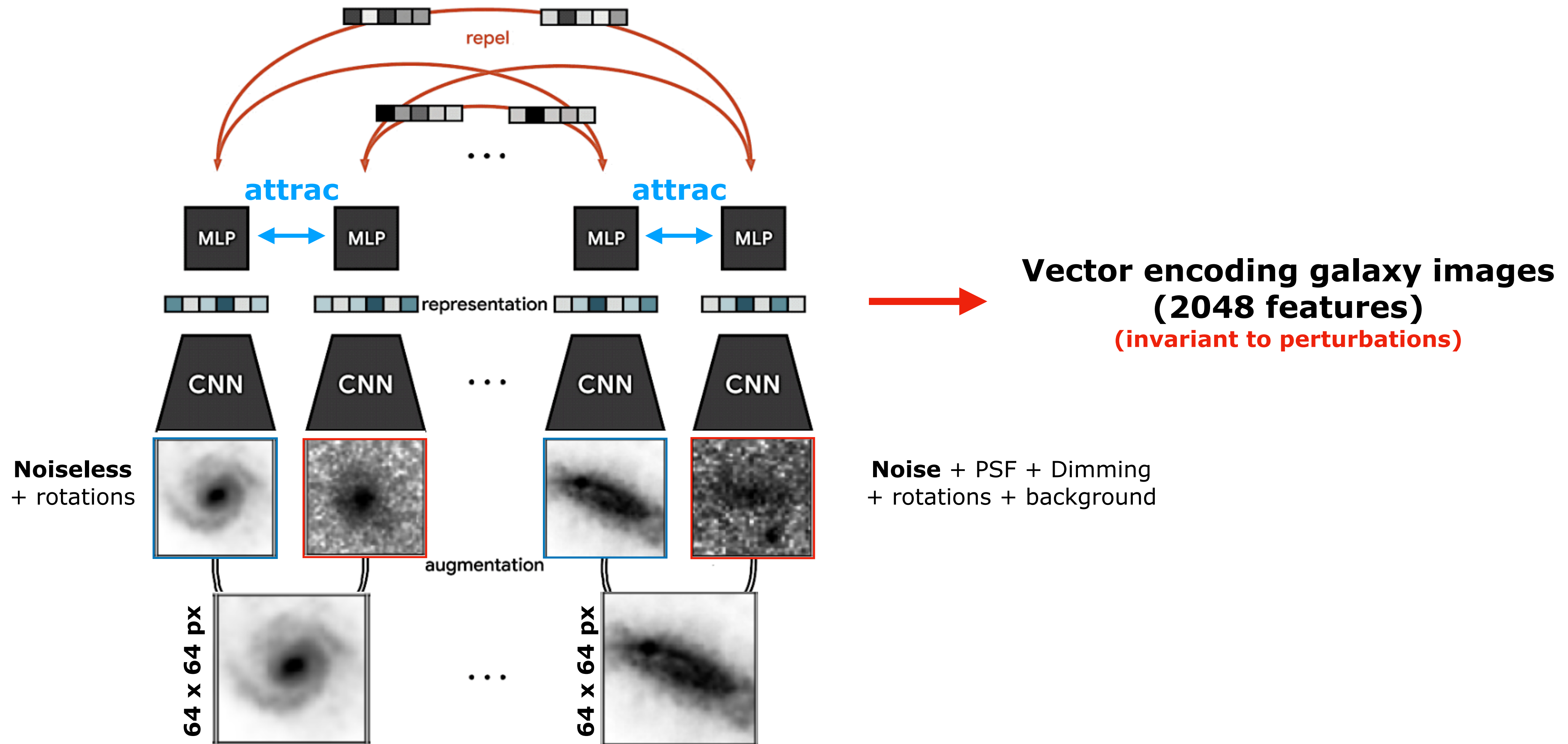
## Mock JWST images of TNG50 galaxies (*Costantin et al. 2022*)

<https://www.tng-project.org/data/docs/specifications/#sec5v>

- 1,238 TNG50 galaxies with  $M_{\star} > 10^9 M_{\odot}$
- Redshift range:  $3 < z < 6$
- 20 l.o.s. projections per galaxy
- **24,760 galaxy images**
- **F200W** (SW) and **F356W** (LW) filters
- **Image augmentations** (rotations, source noise, etc.)



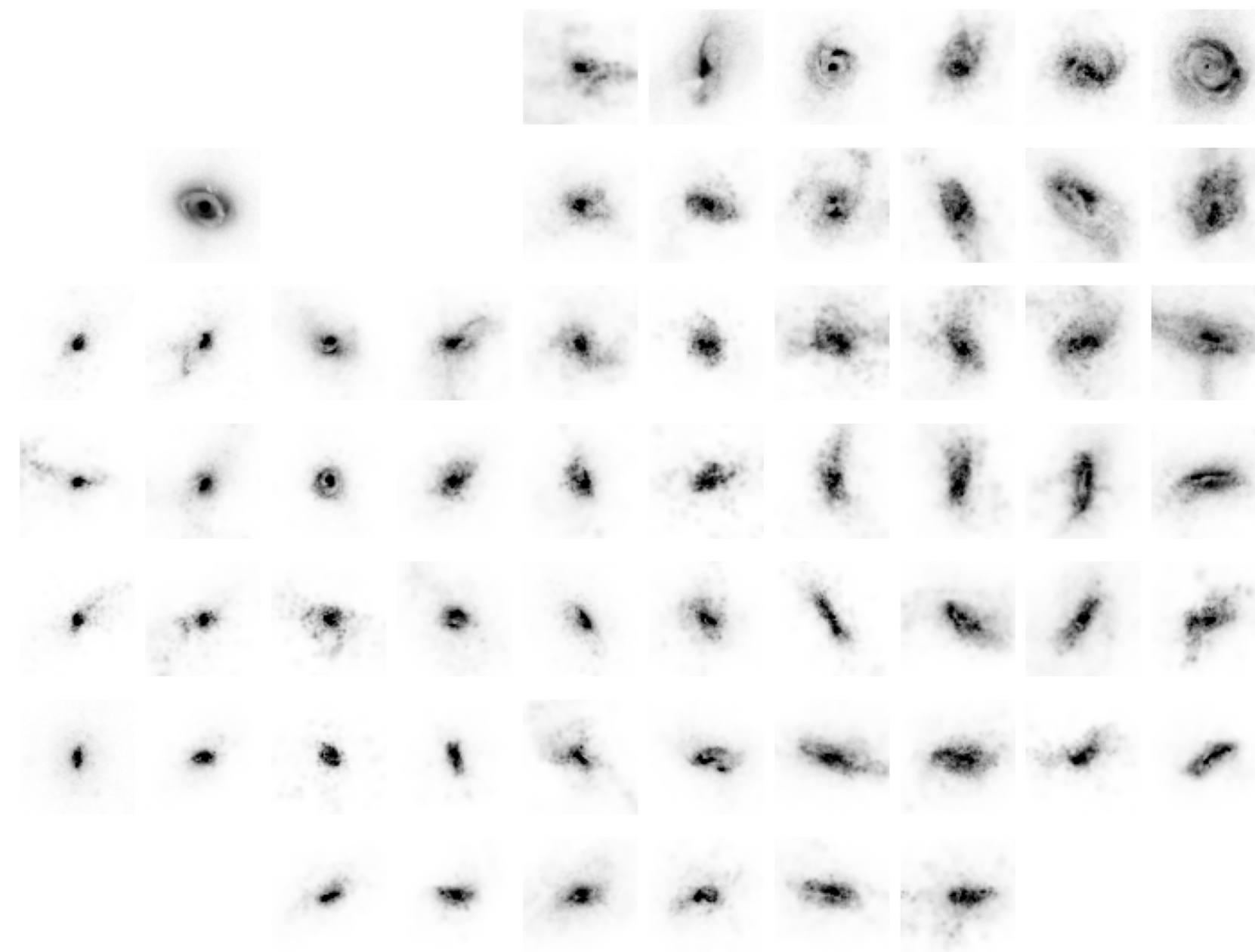
## Contrastive learning for morphological classification



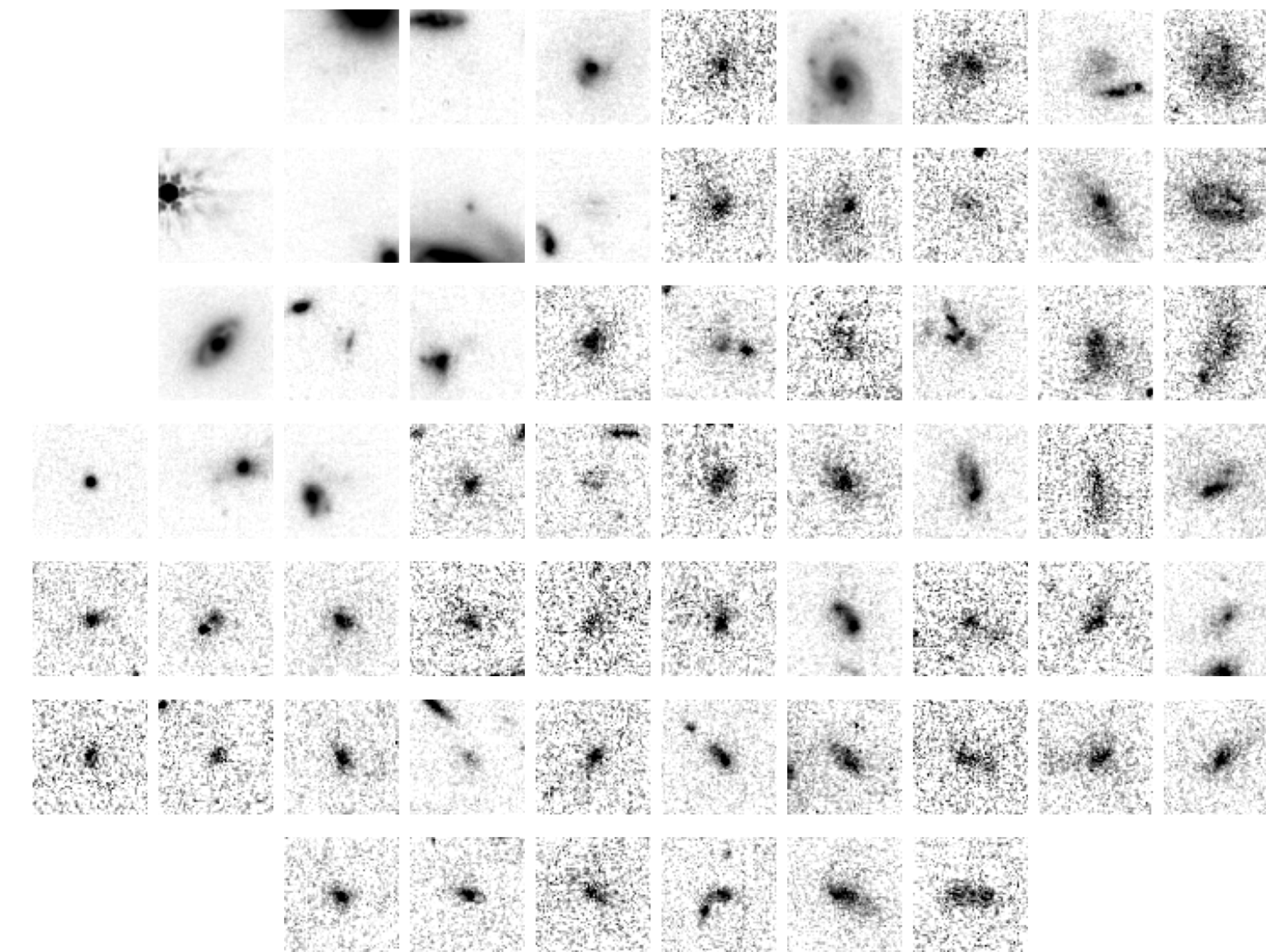
## Contrastive learning for morphological classification

UMAP 2D projection

**Noiseless (TNG50)**  
F200W



**Noise-added (TNG50)**  
F200W

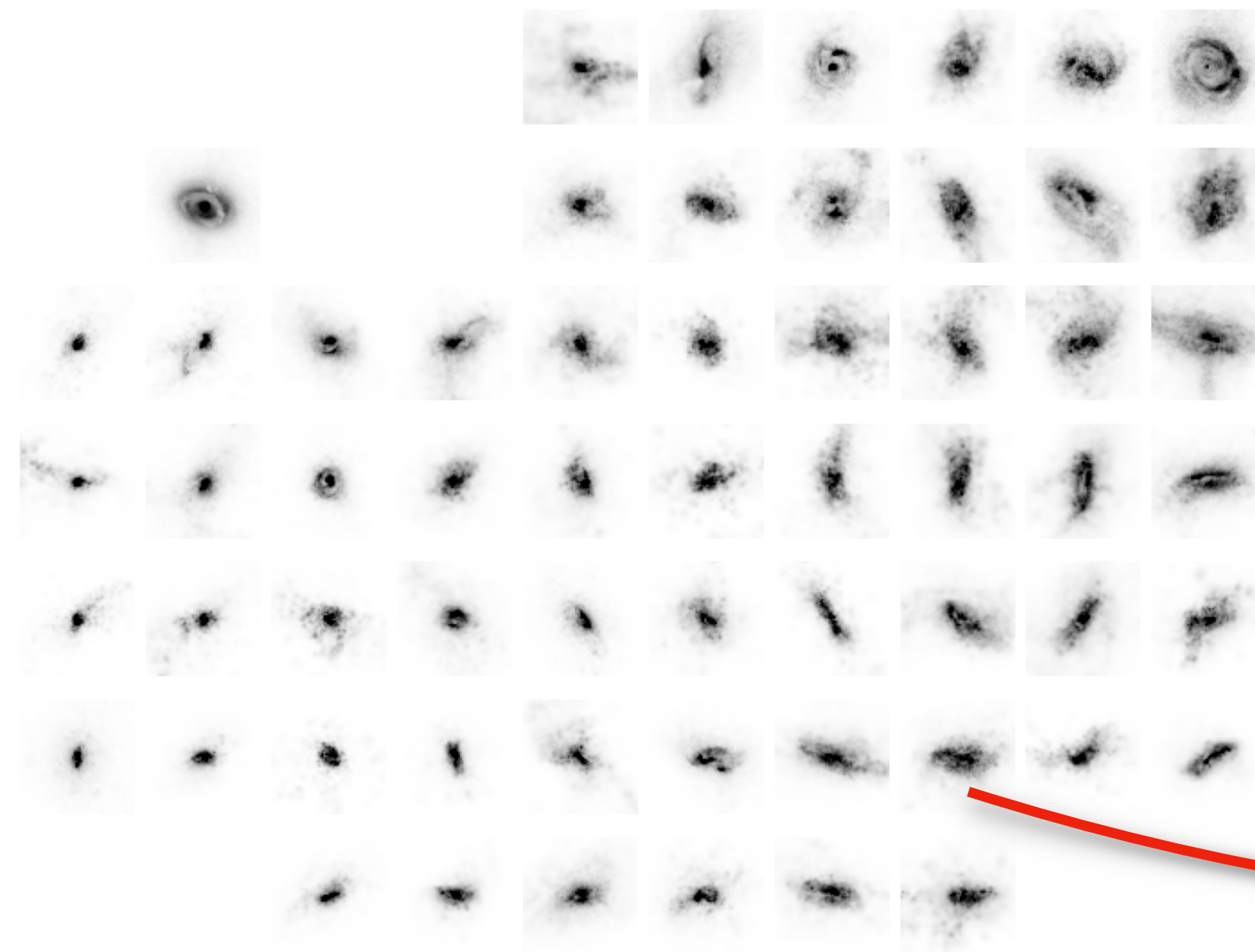




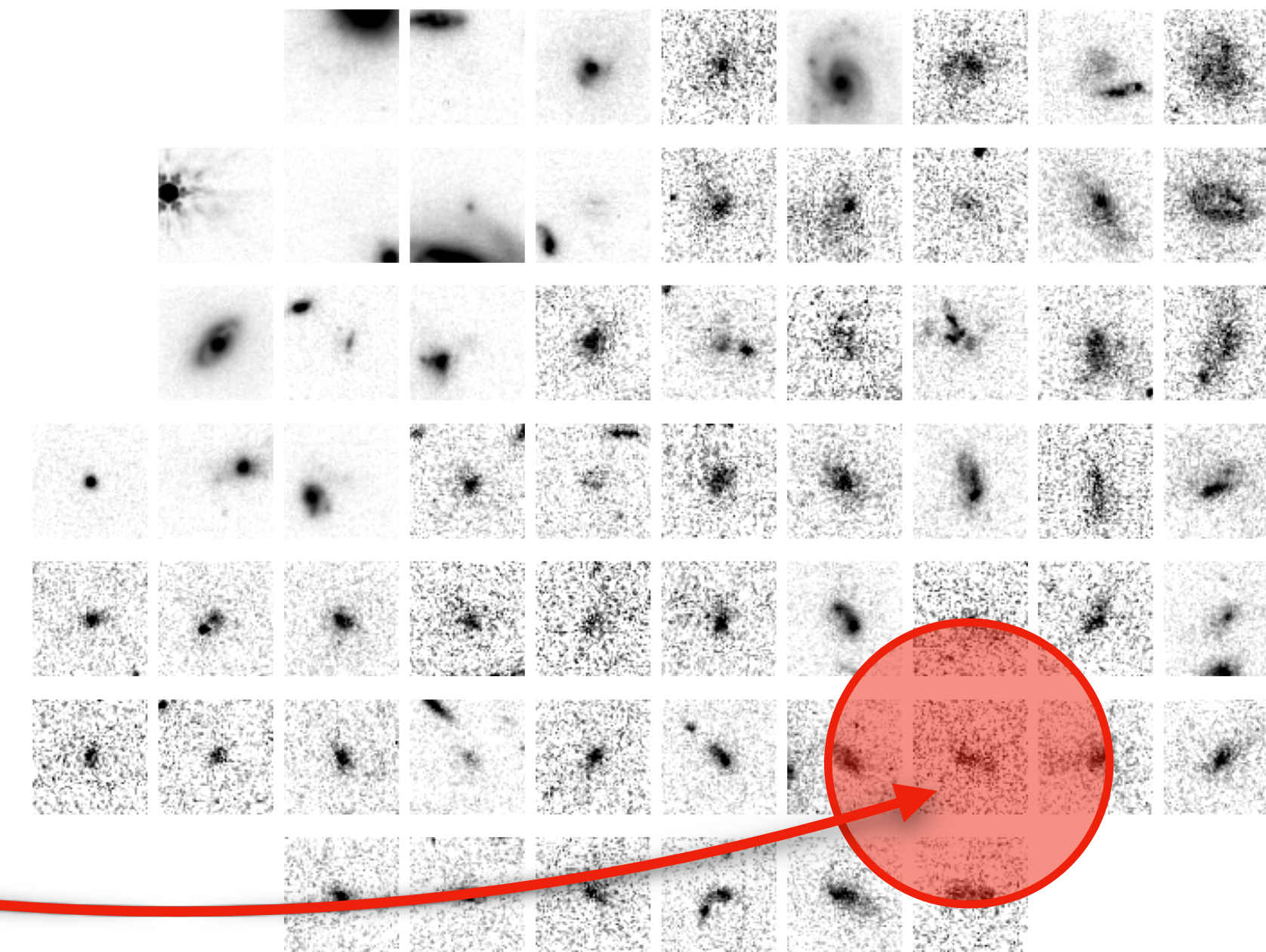
## Contrastive learning for morphological classification

A morphological representation robust to noise

**Noiseless (TNG50)**  
F200W



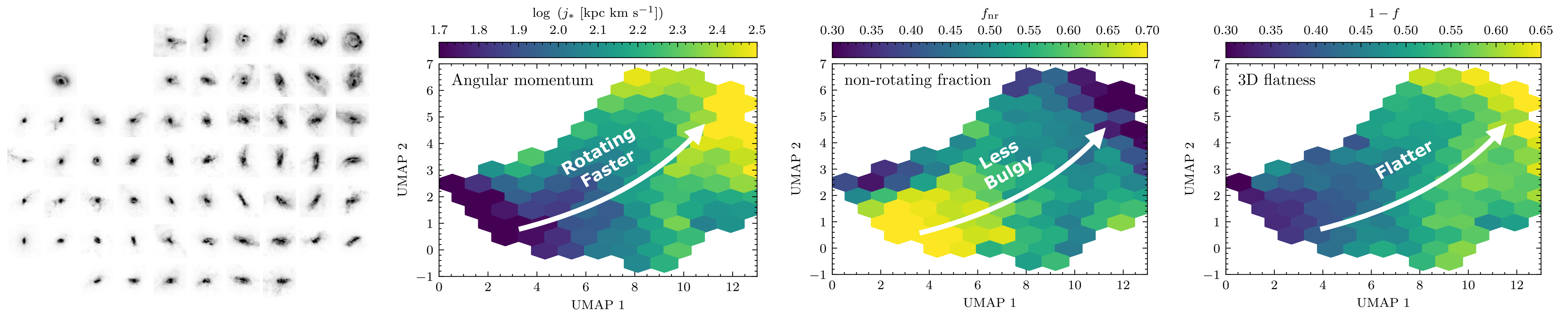
**Noise-added (TNG50)**  
F200W



# Data-driven galaxy morphologies at $z > 3$

## Contrastive learning for morphological classification

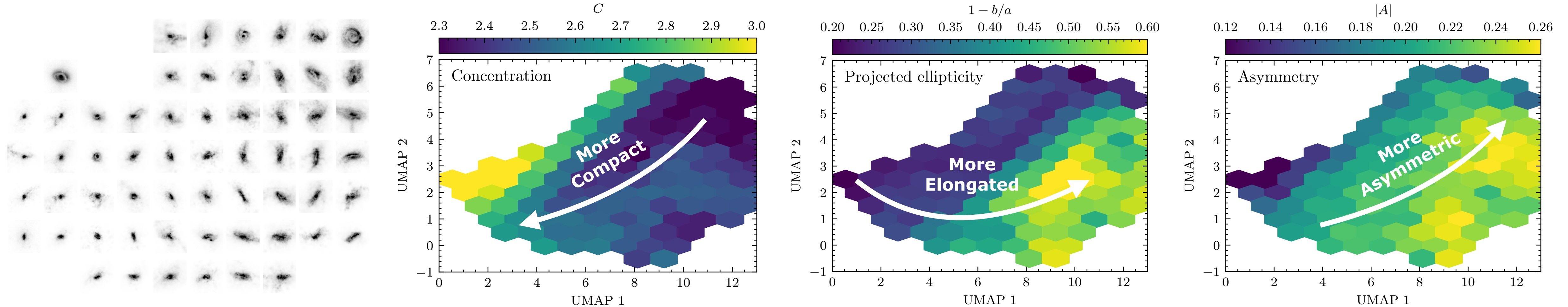
### Correlation with physical properties



# Data-driven galaxy morphologies at $z > 3$

## Contrastive learning for morphological classification

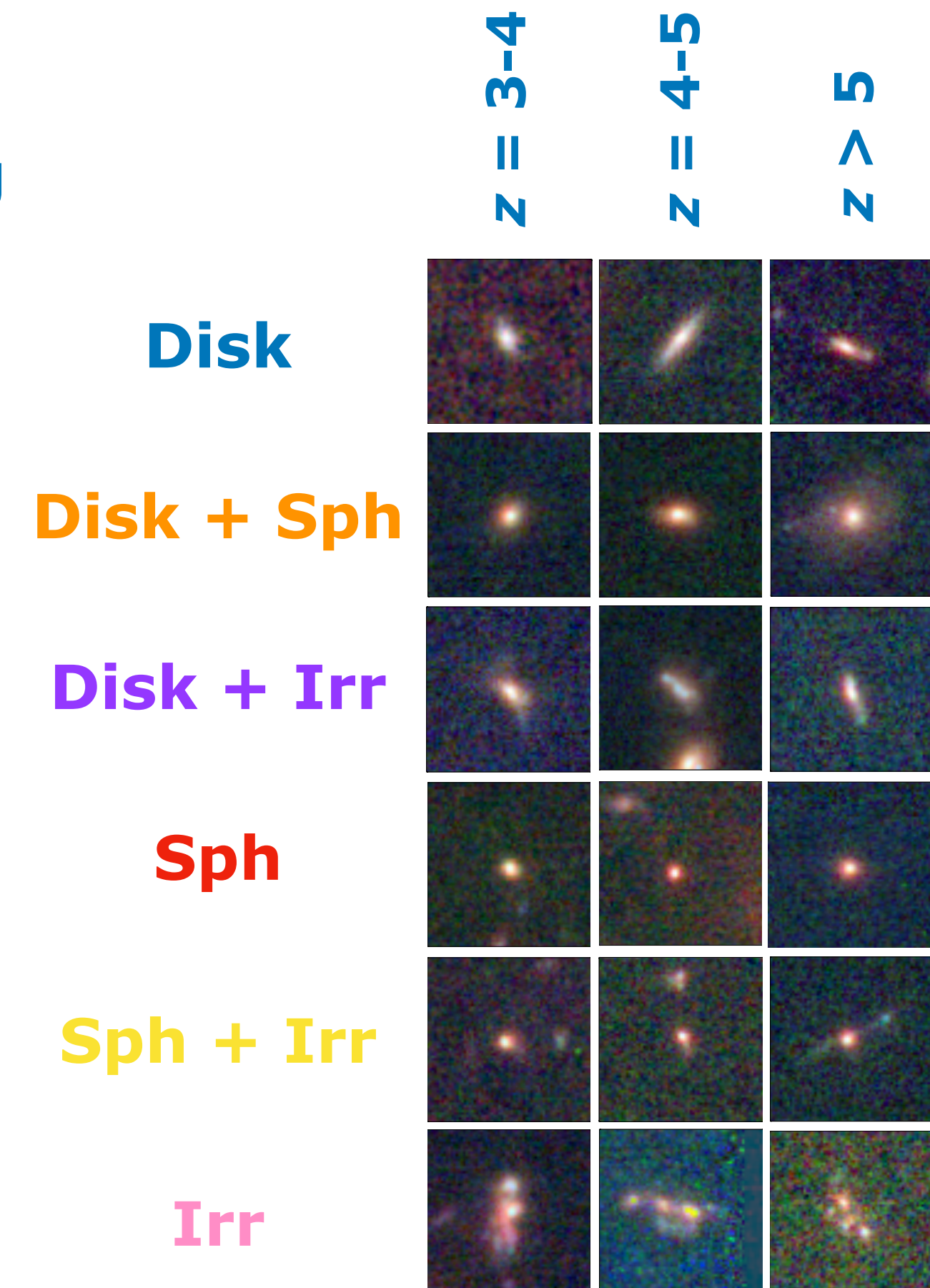
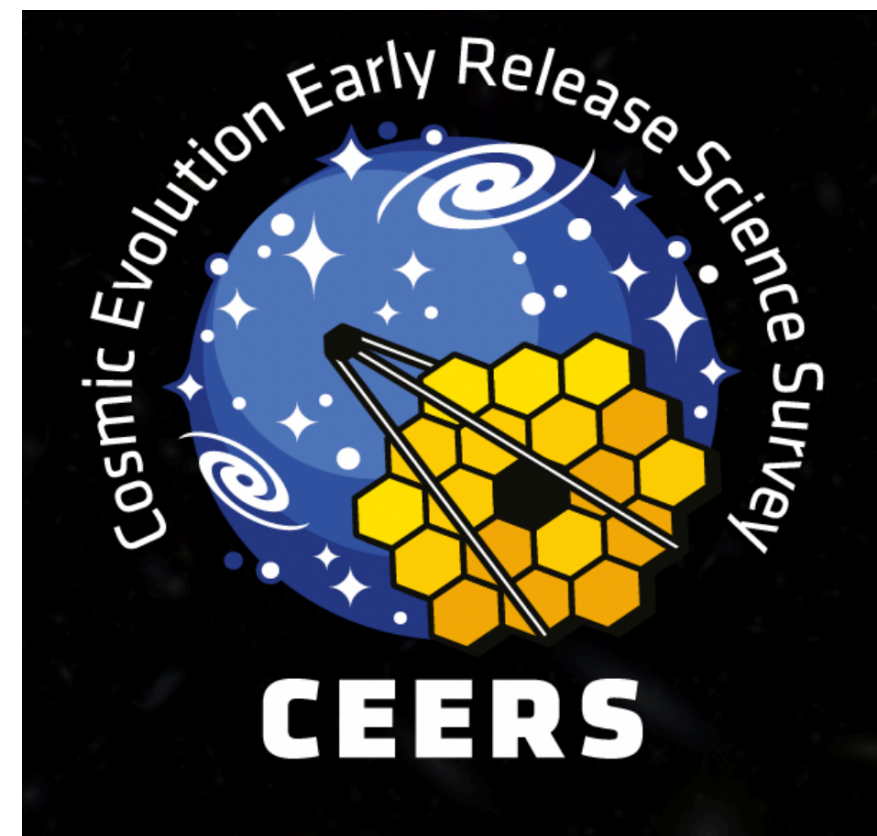
### Correlation with standard morphological parameters



# Data-driven galaxy morphologies at $z > 3$

## Contrastive learning representations of observed JWST galaxies

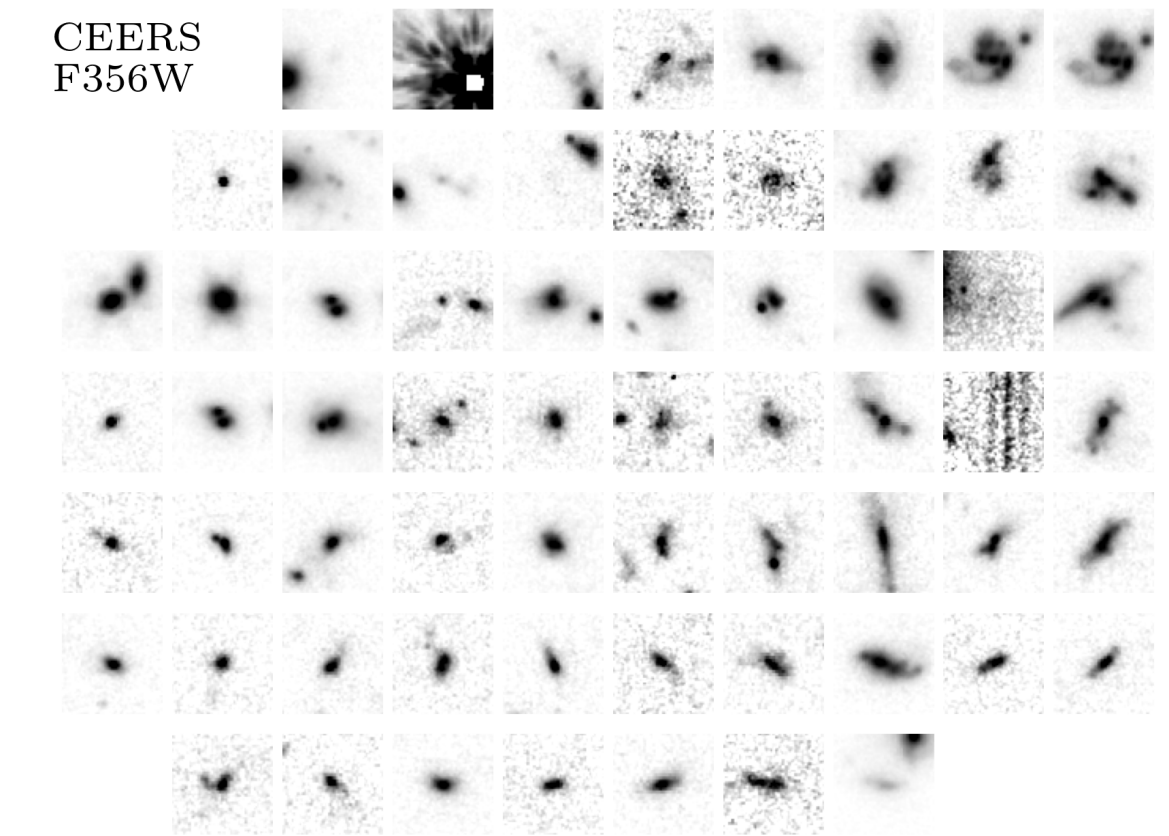
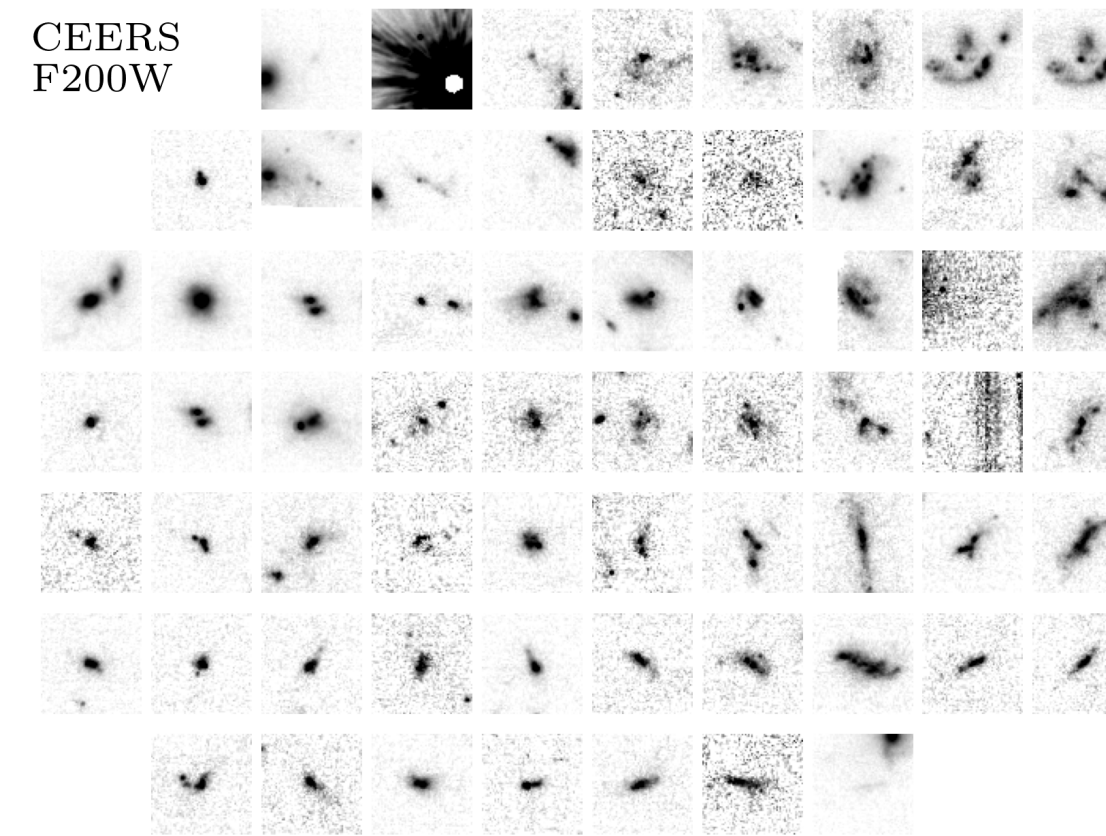
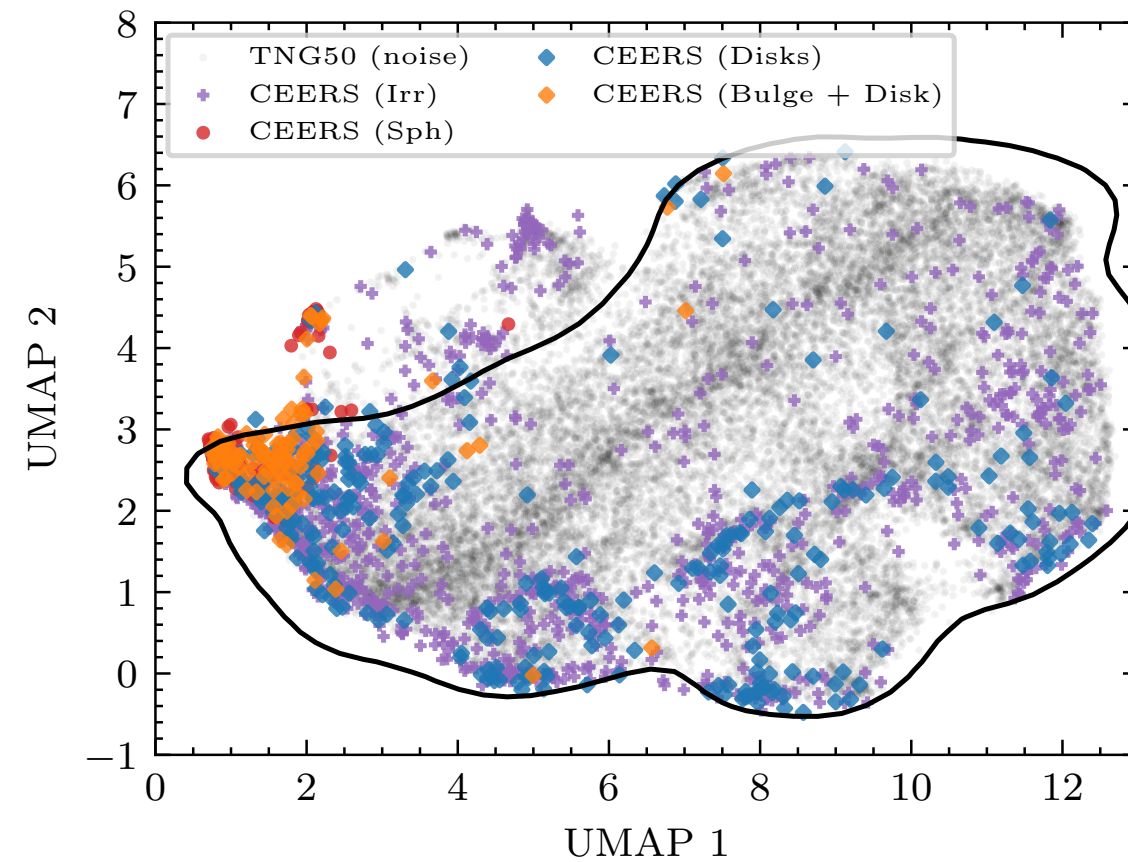
- ☉ **CEERS** (*Finkelstein et al. 17,22,23; Bagley et al. 22*)
- ☉ NIRCam imaging of 100 arcmin<sup>2</sup> with **depth 28.8-29.7 mag**
- ☉ Mass-complete ( $M_{\star} > 10^9 M_{\odot}$ )
- ☉ Redshift-selected sample ( $3 < z < 6$ )



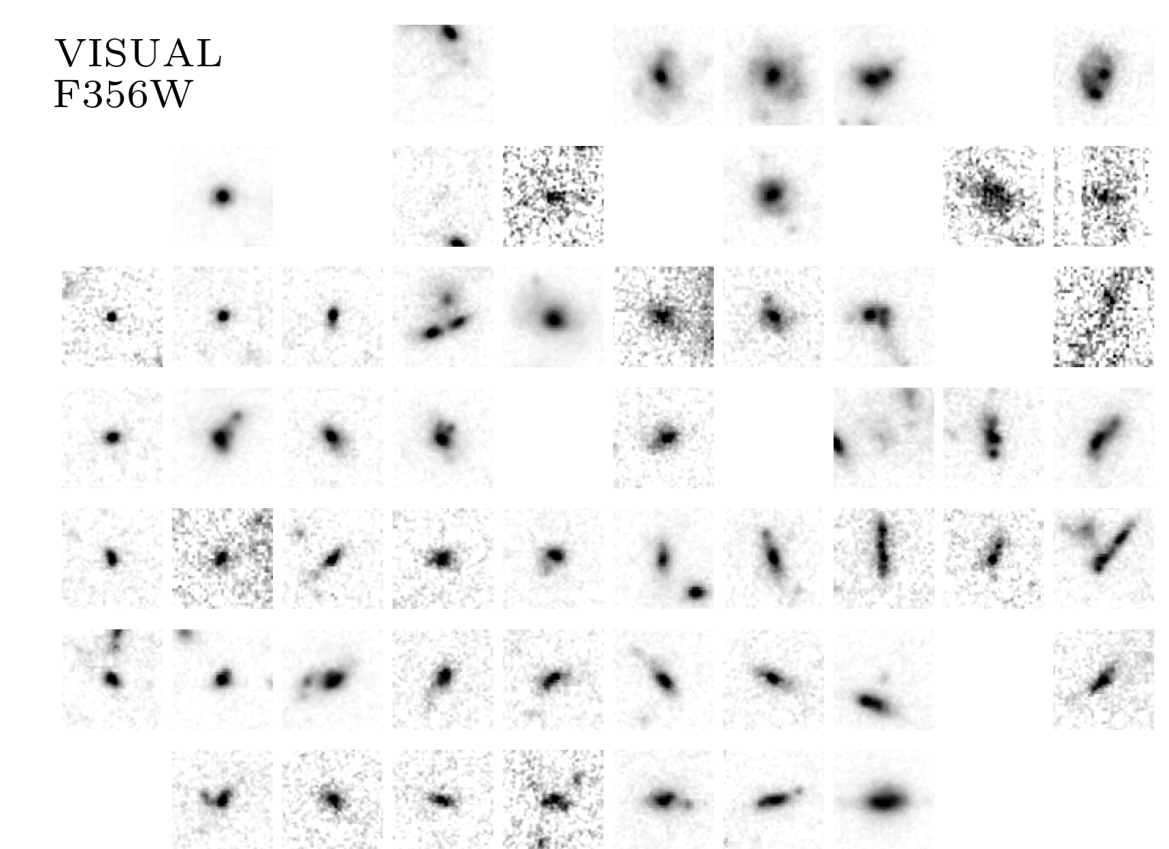
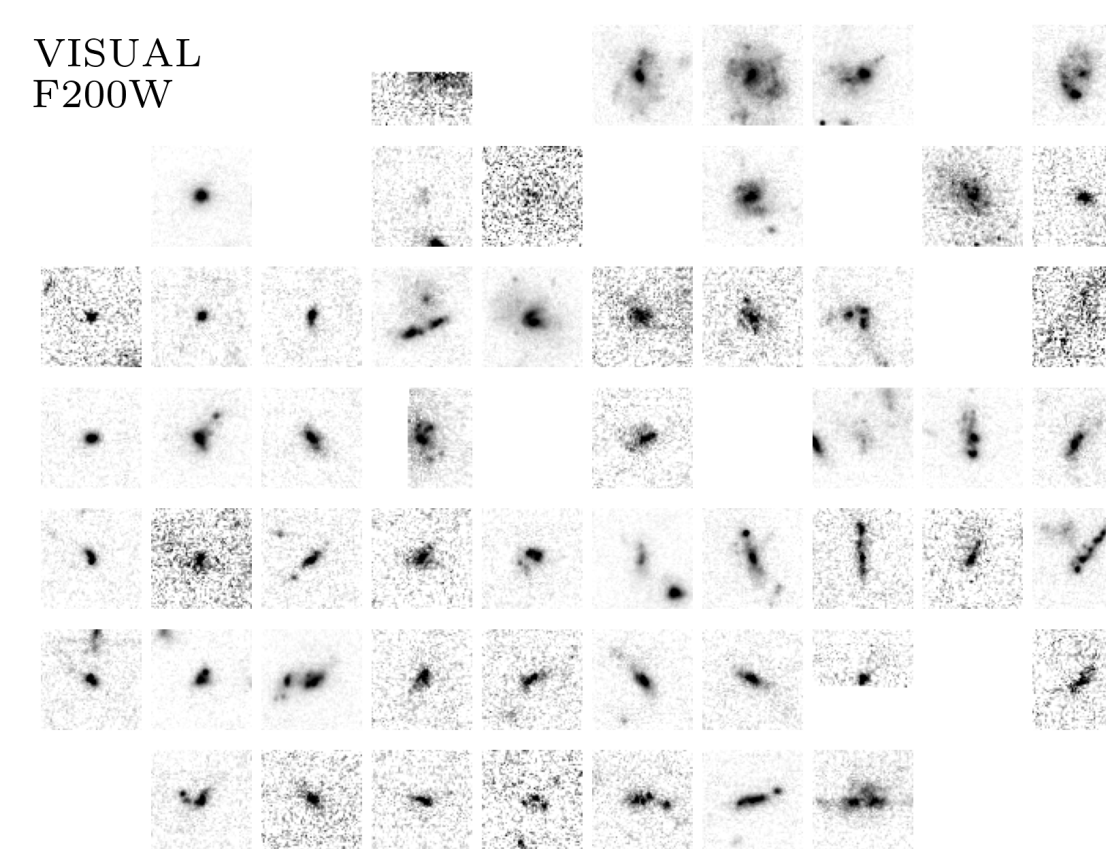
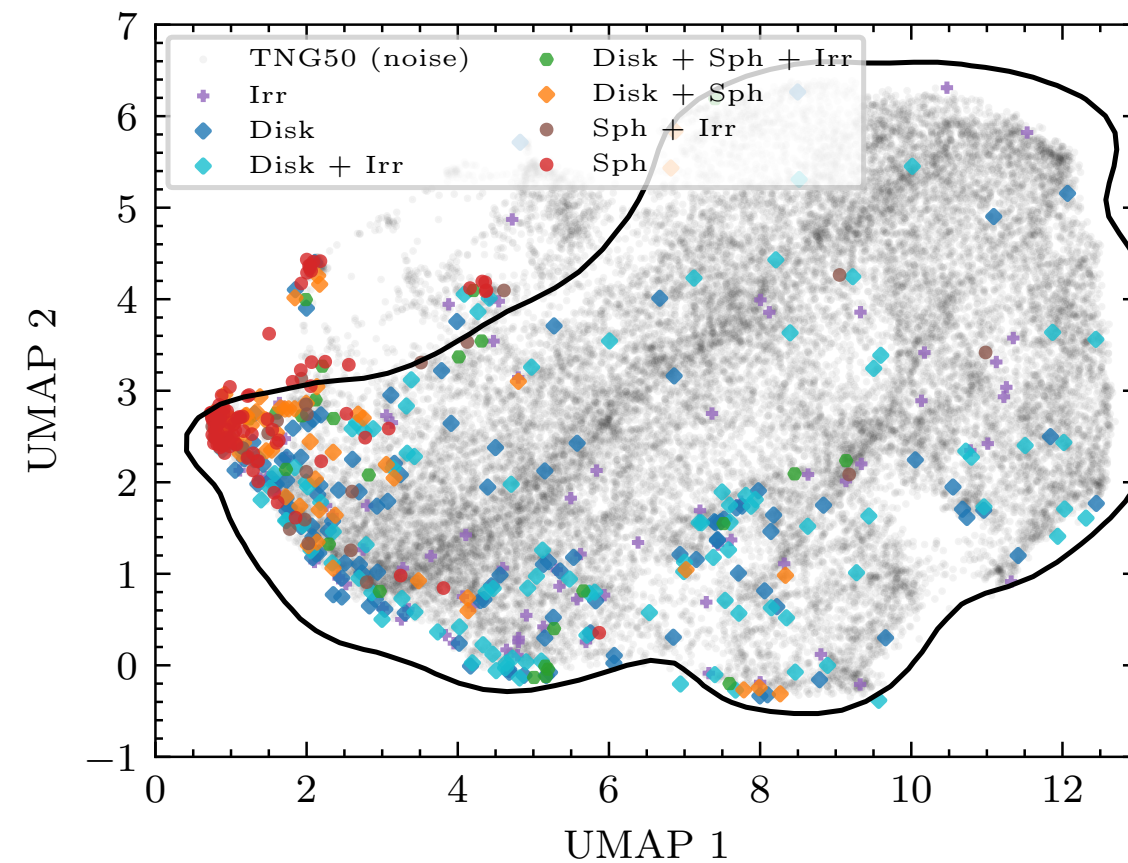
**Credits:** Kartaltepe et al. 2023

## Contrastive learning representations of observed CEERS galaxies

**MHC+2023b**  
CNN-based  
Domain adaptation  
from CANDELS labels



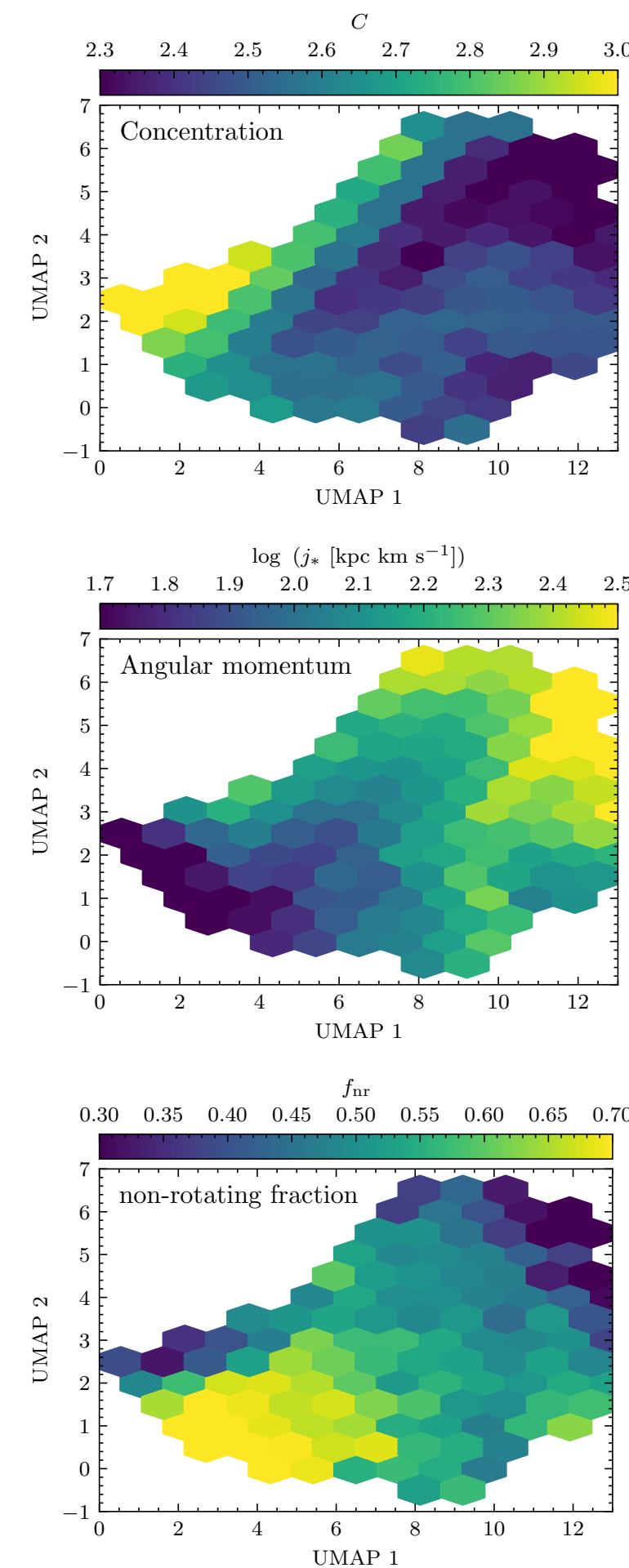
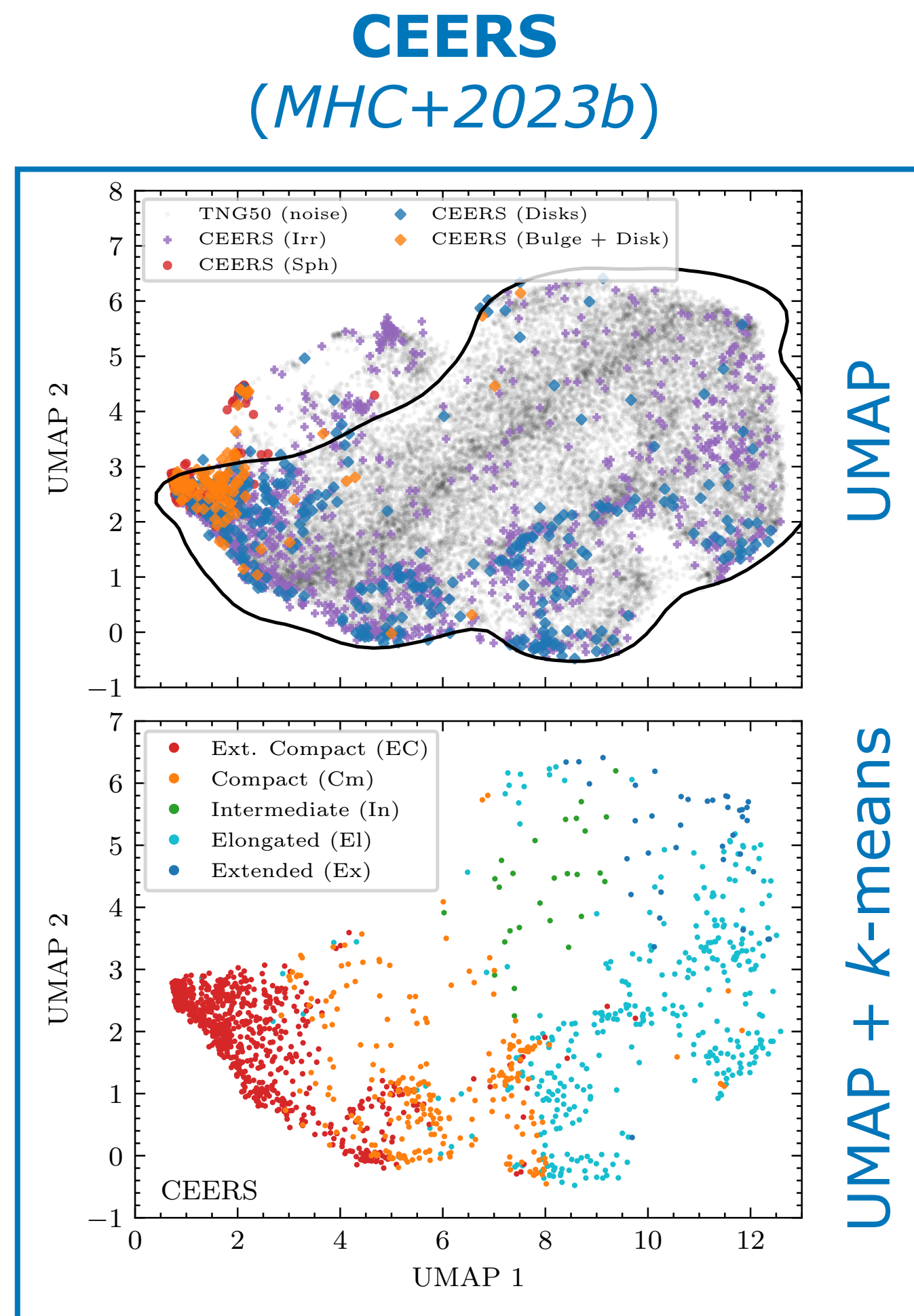
**Kartalpe+2023**  
Visual classifications



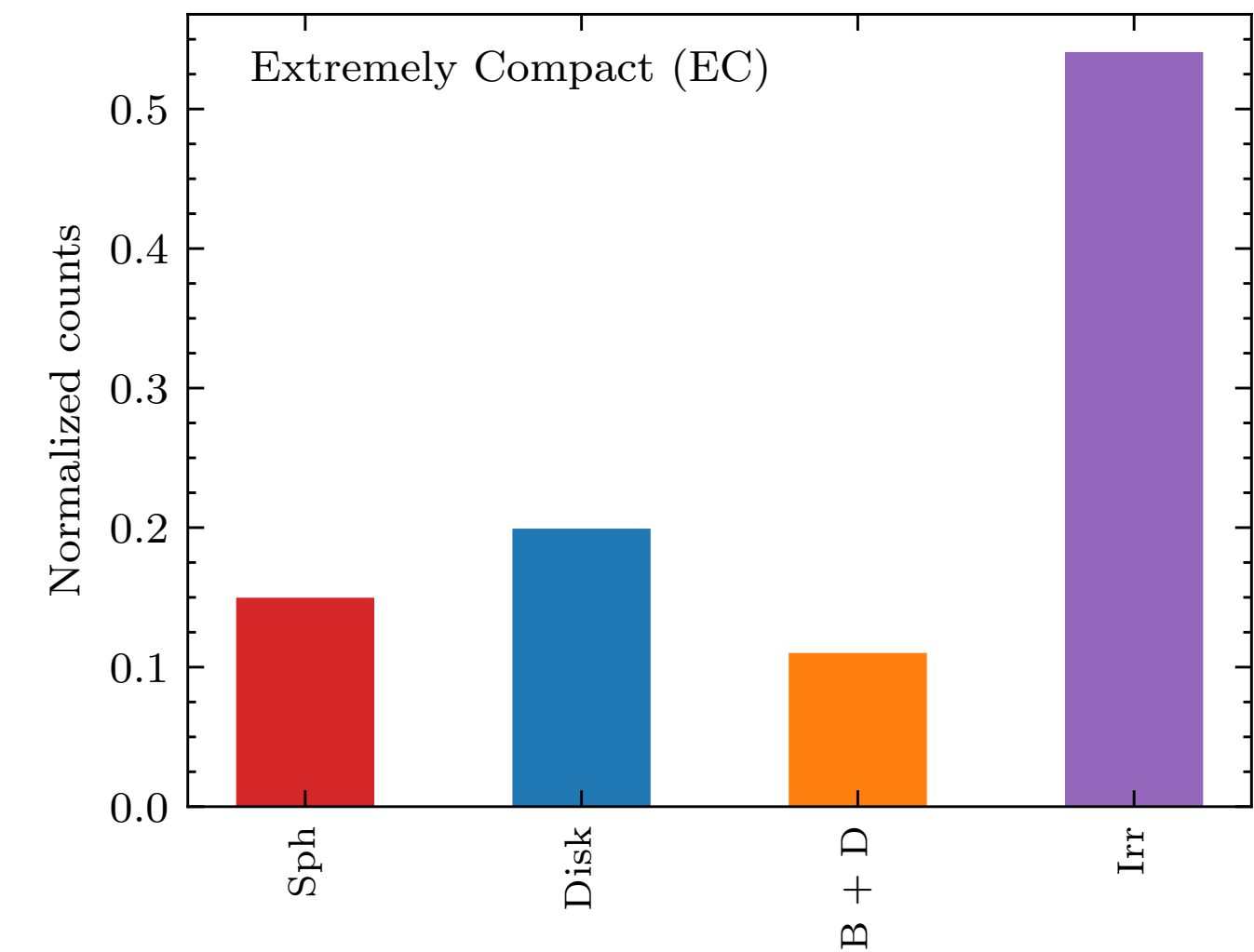
**Not a good overlap between simulations and observations**

# Data-driven galaxy morphologies at $z > 3$

## Can we say more about the nature of the observed disk galaxies?



### Extremely Compact (EC)

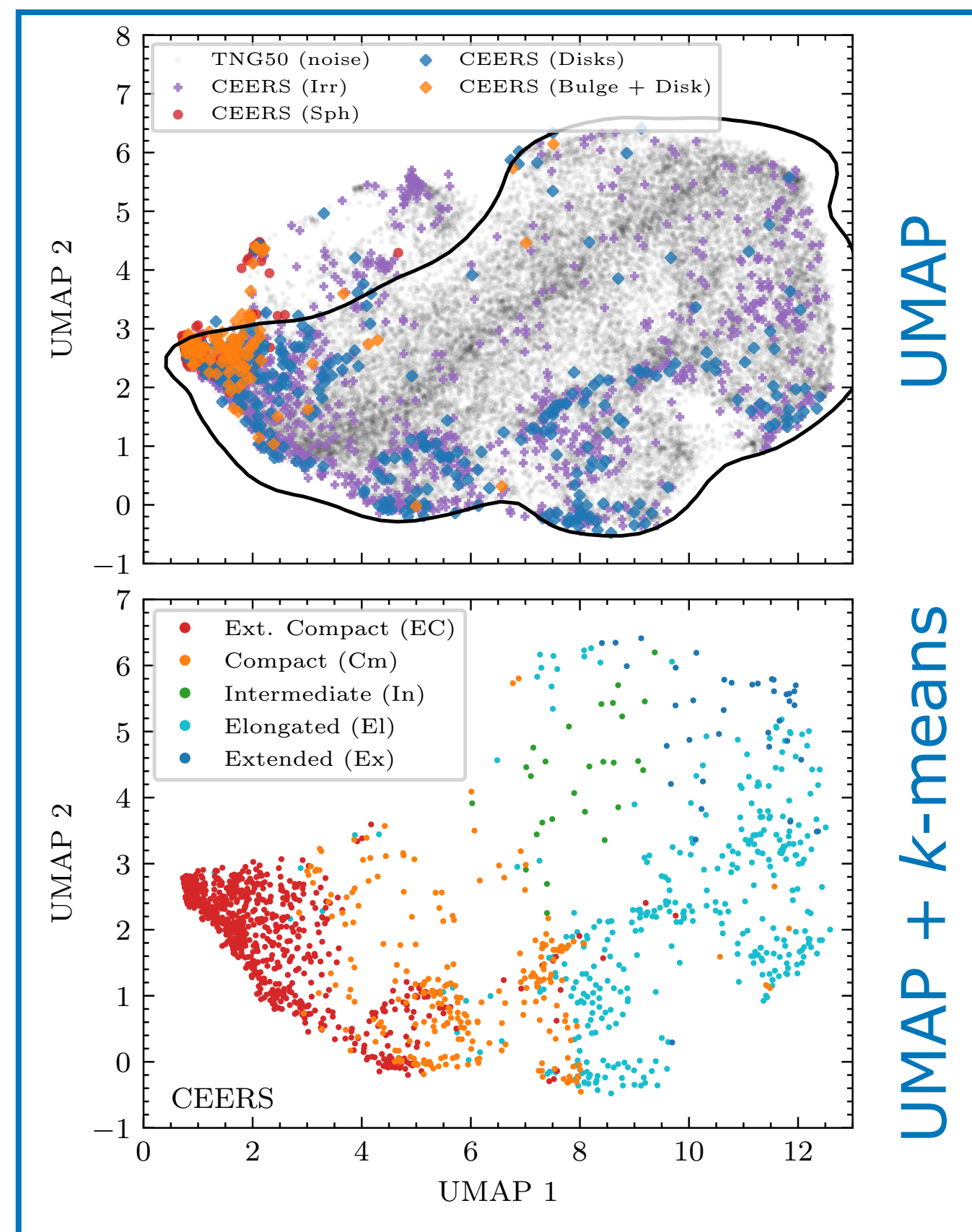


**~50% of Irr and Disk are EC, slow rotators, bulge dominated**

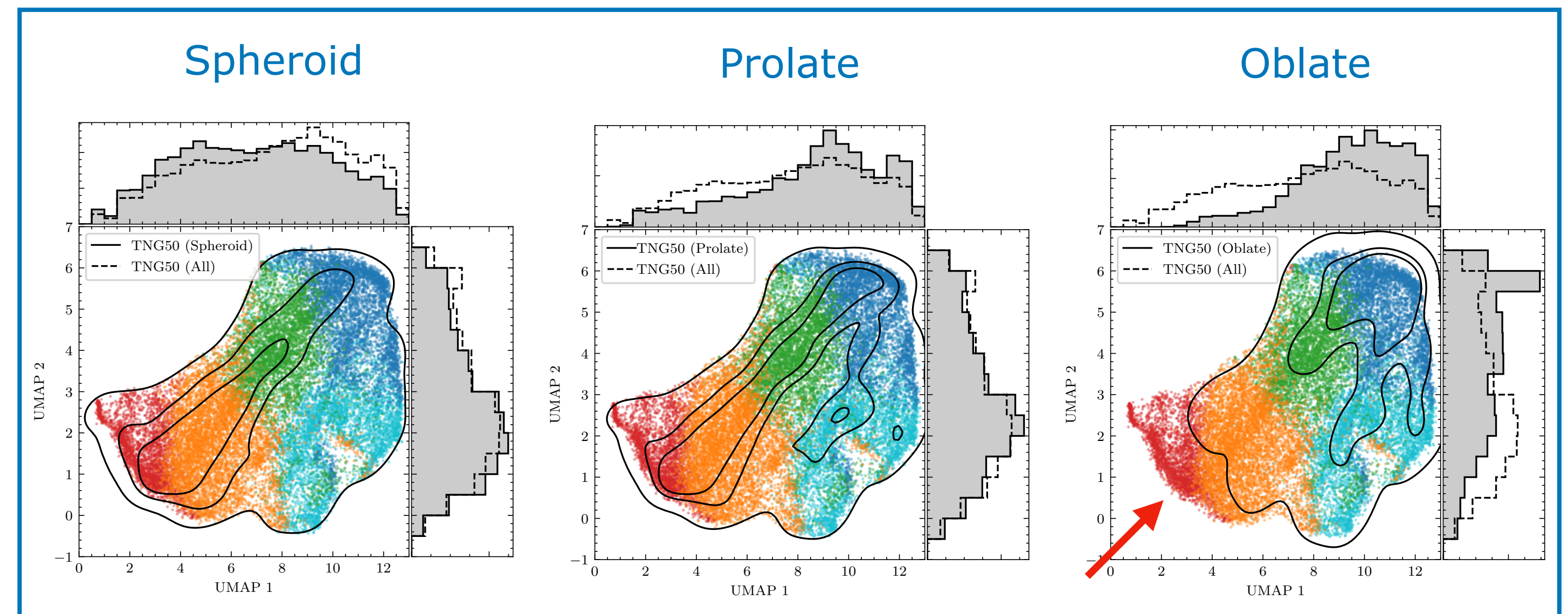
# Data-driven galaxy morphologies at $z > 3$

## Can we say more about the nature of the observed disk galaxies?

**CEERS**  
(MHC+2023b)



**TNG50**



**~50% of Disks not likely to be pure Disks,  
consistent with prolate stellar structure**

*Pandya et al. 2023*

**50%-80% prolate dwarfs at  $z = 3-8$**

## Conclusions

- ④ **Morphologies of JWST galaxies at  $z > 3$  with contrastive learning** (data-driven)
- ④ Method **calibrated on mock JWST galaxy images of TNG50 galaxies**
- ④ Representations robust to noise, color, orientation, S/N
- ④ Representations correlate with physical, photometric and structural properties
- ④ Morphological distributions of **CEERS and TNG50 galaxies are different** (observed galaxies are more compact and elongated than simulated ones)
- ④ **CNN-based and visually classified Disks** similar to **TNG50 compact galaxies** with **low angular momentum** and **non-oblate structure**
- ④ **Disk fractions at  $z > 3$  uncertain and possibly overestimated**
- ④ **Deeper imaging** and/or **spectroscopic follow-ups** (also model calibrations with **other simulations**) needed to determine the true nature of these galaxies

<https://arxiv.org/abs/2302.07277v2>