

# The Multimodal Universe:

## Enabling Large-Scale Machine Learning with 100 TB of Astronomical Scientific Data

The Multimodal Universe Collaboration

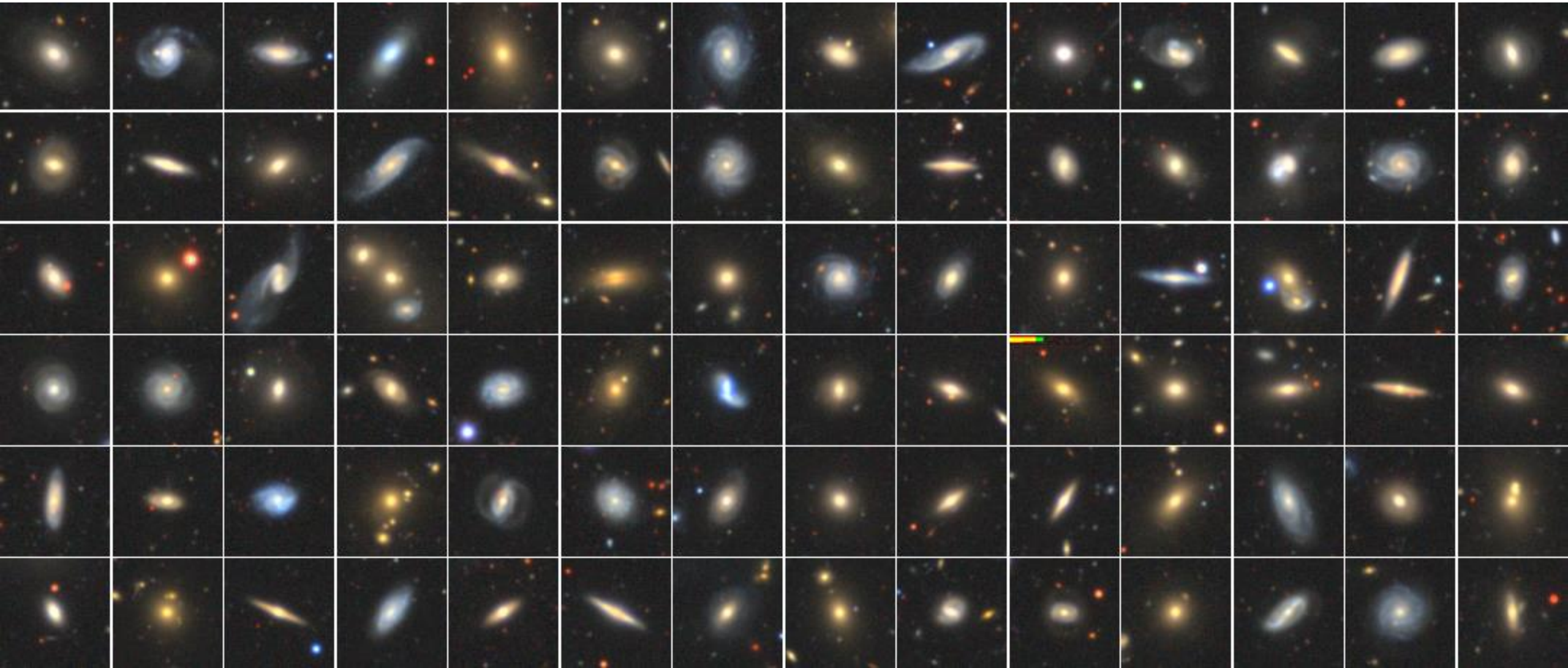


Wed, Dec 11, 16:30-19:30 PST (Poster Session 2)



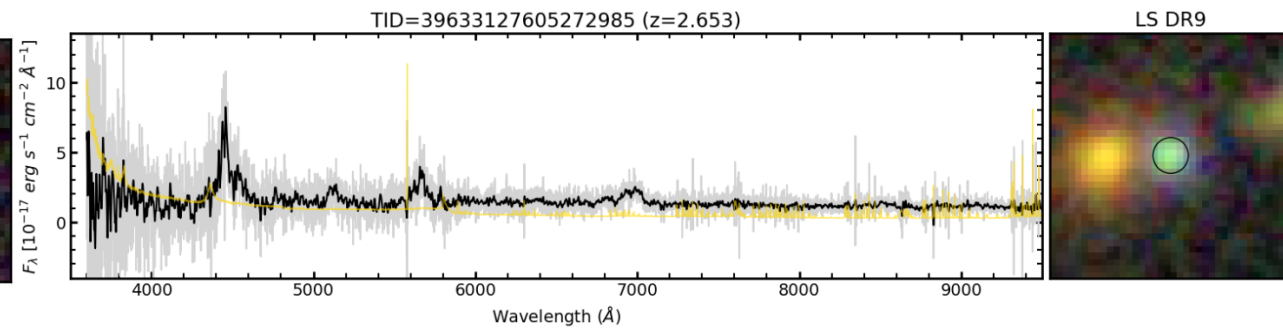
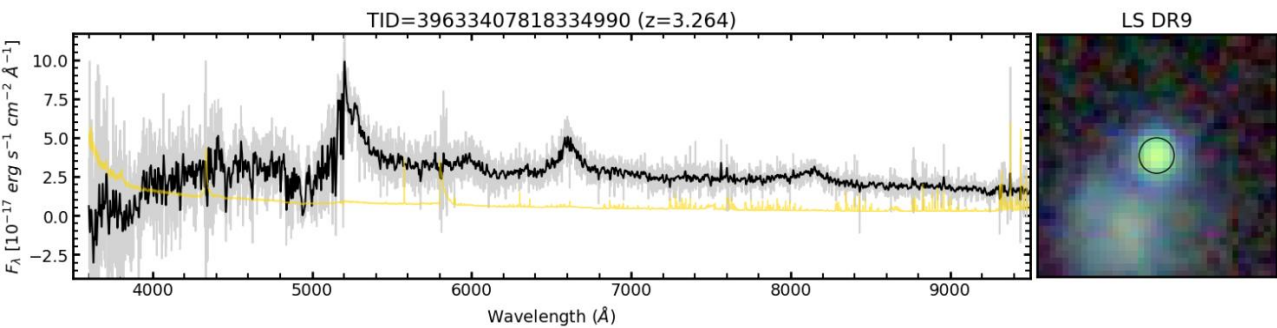
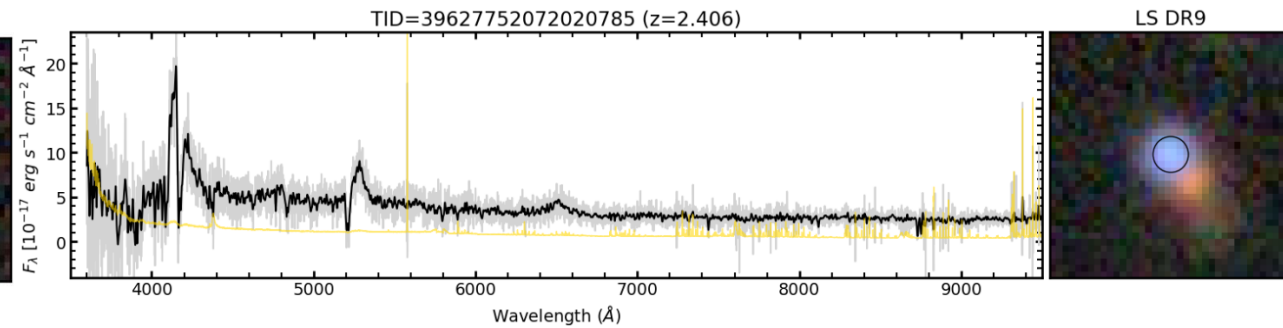
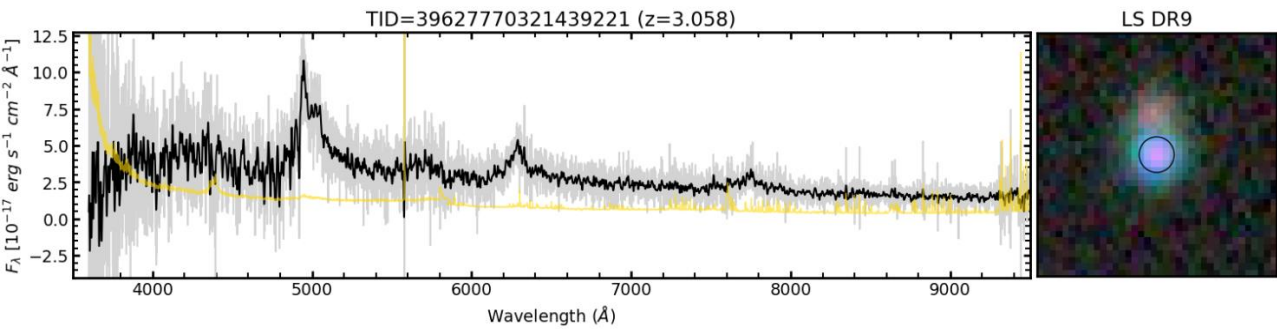
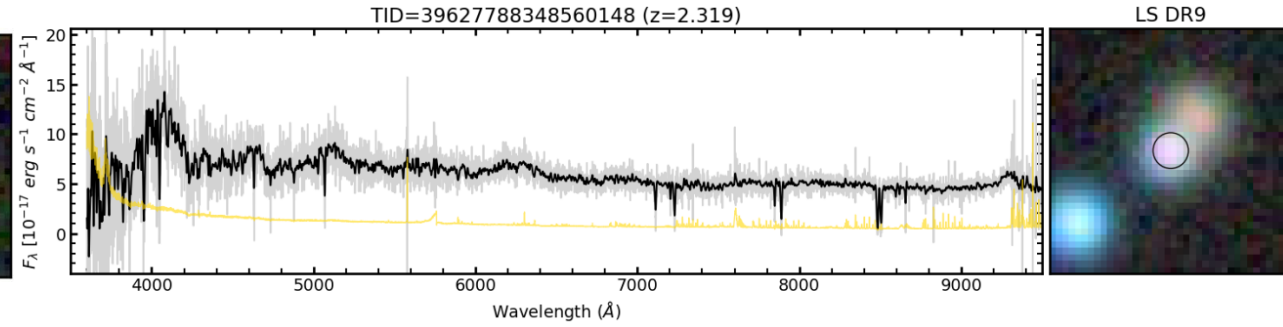
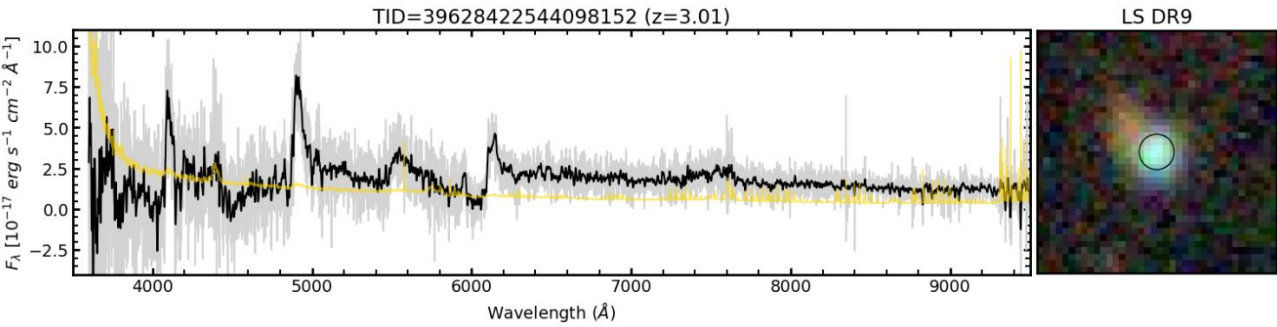
# Imaging

DESI Legacy Imaging Surveys

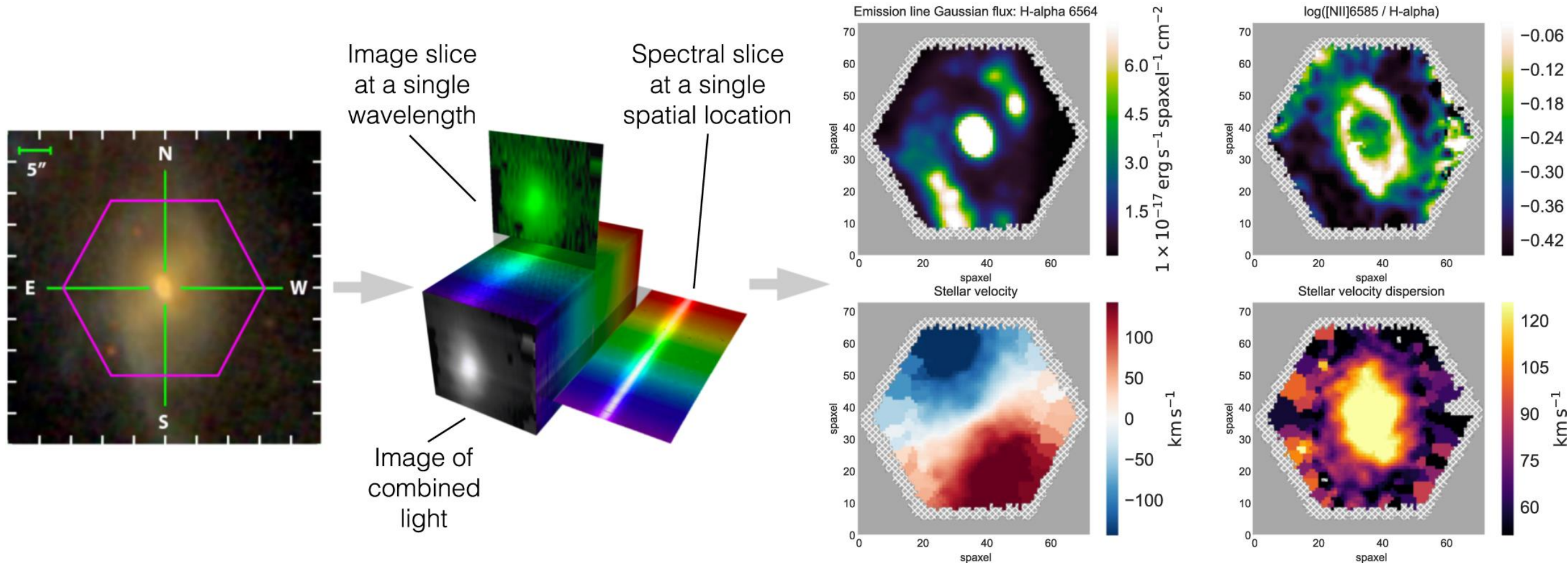


# Spectra

## DESI Spectra

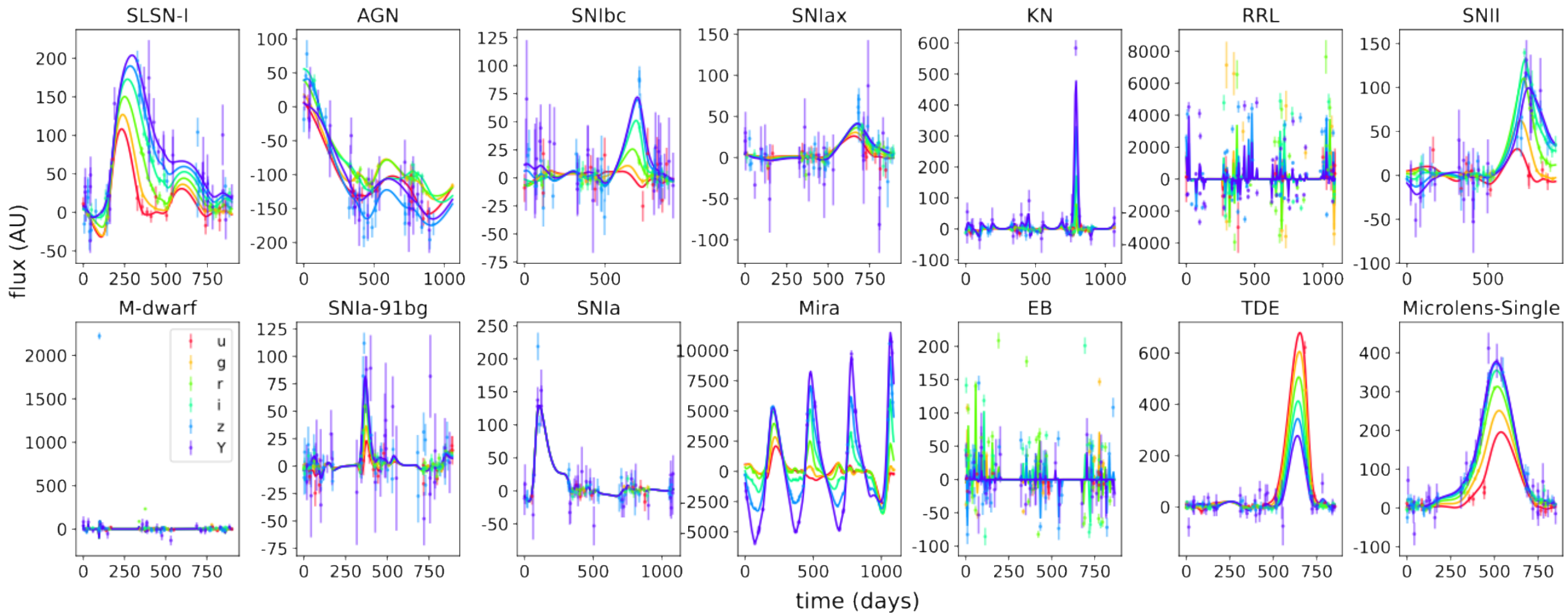


# Hyperspectral MaNGA

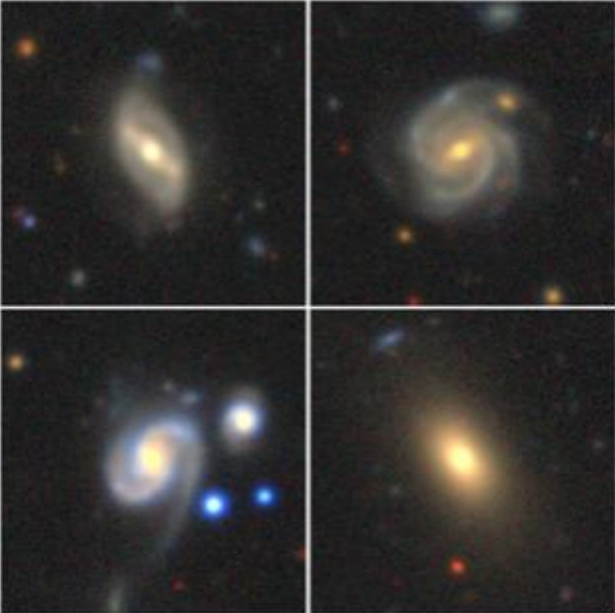
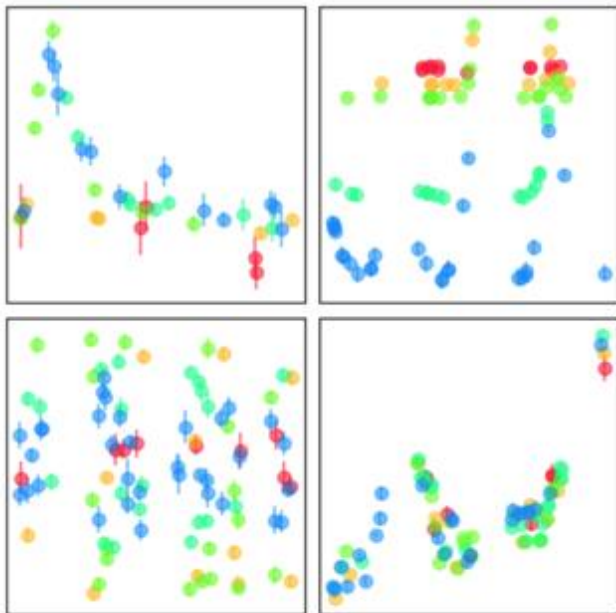
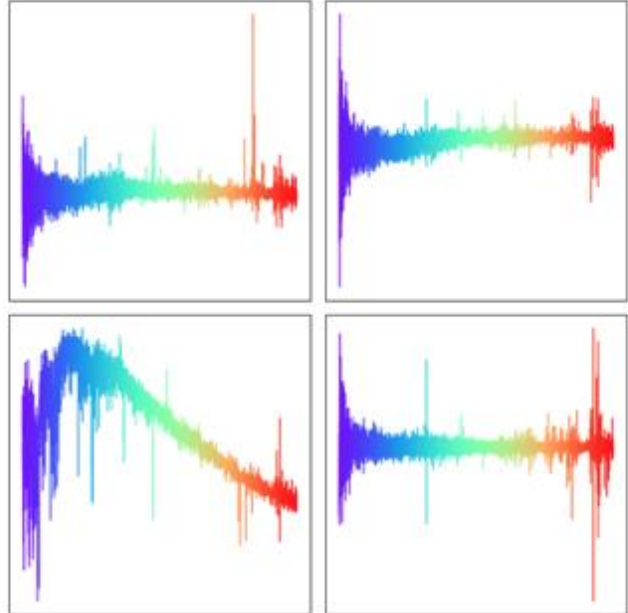


# Time Series

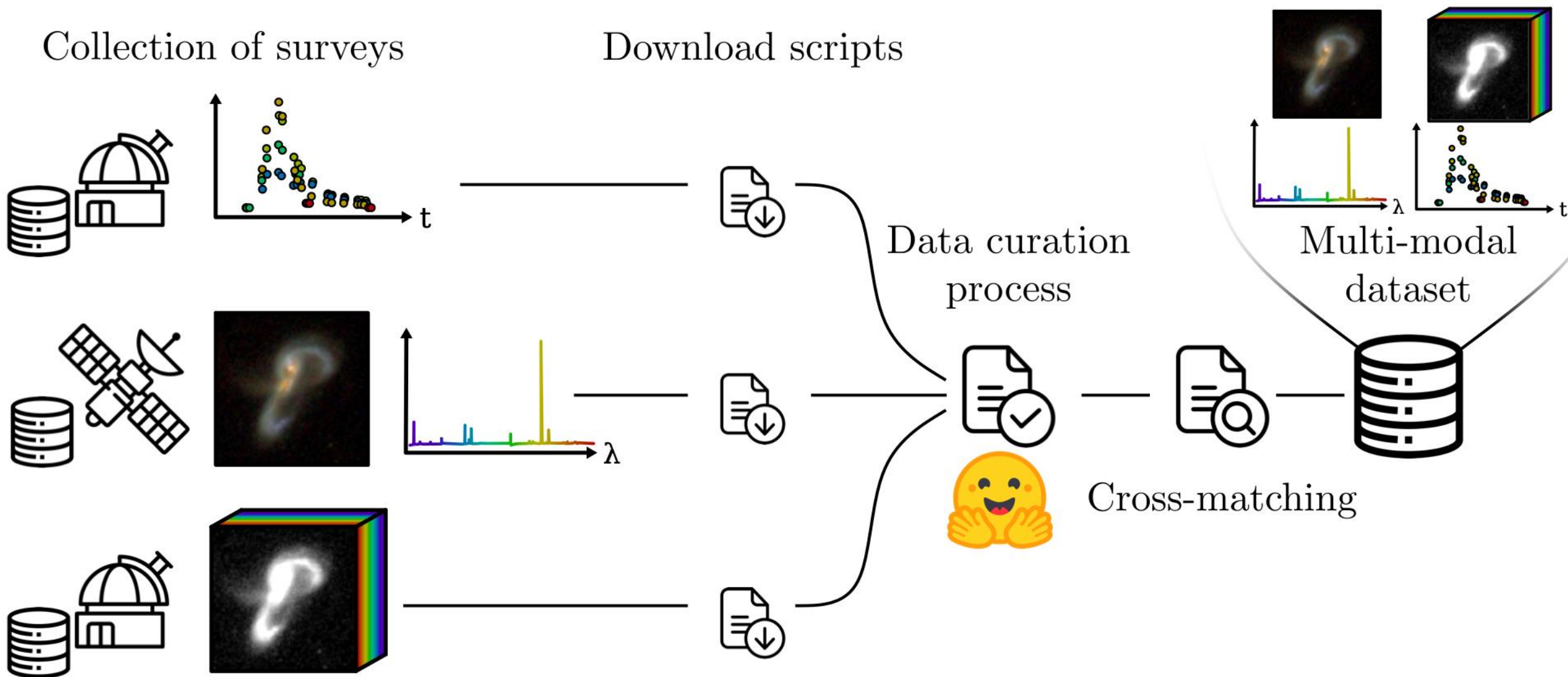
PLAsTiCC



# The Multimodal Universe

	Images	Time-Series	Spectra
# examples	140M	4.5M	225M
Description	images in a variety of wavelength ranges, including optical and infrared	multivariate time-series of flux + uncertainty in different wavelength ranges	flux as a function of wavelength
Tasks	galaxy classification, physical property estimation	time-series classification, redshift estimation	physical property estimation
Examples			

# Framework

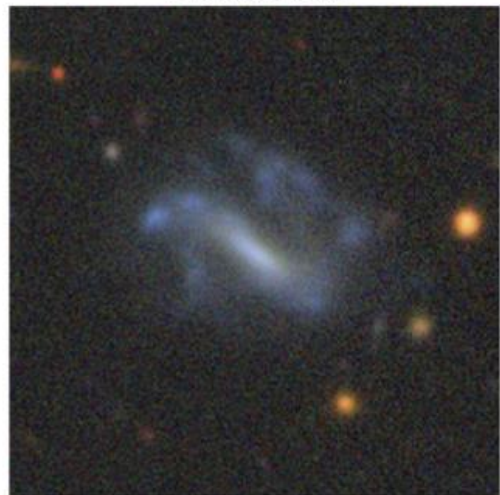


# Benchmarks

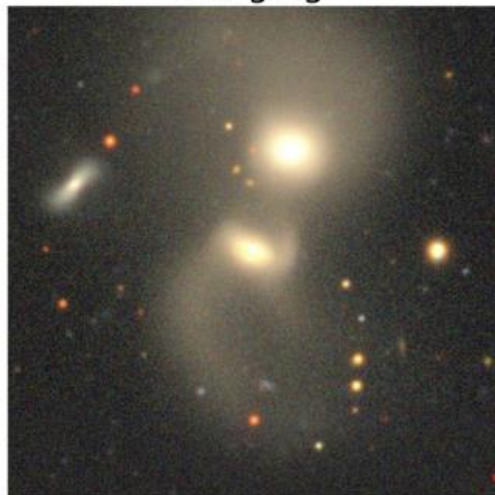


# Benchmarks

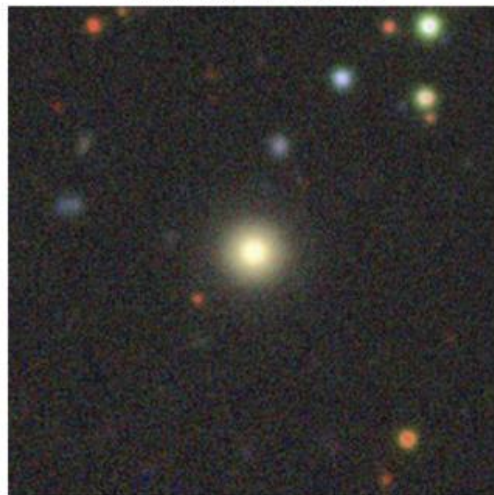
Disturbed



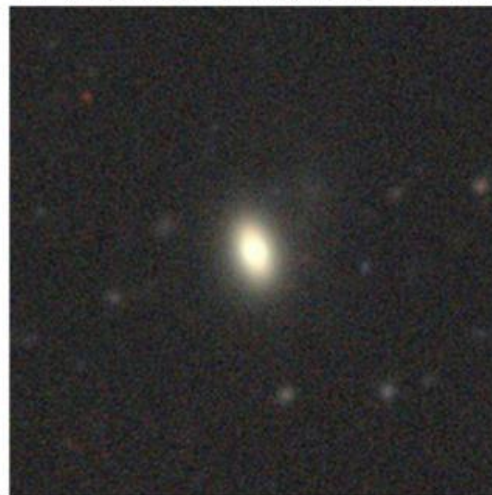
Merging



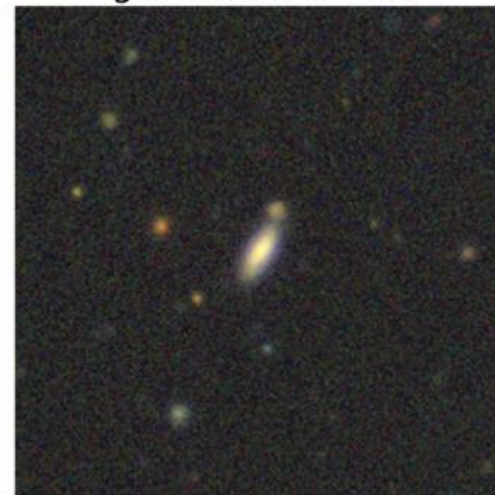
Round Smooth



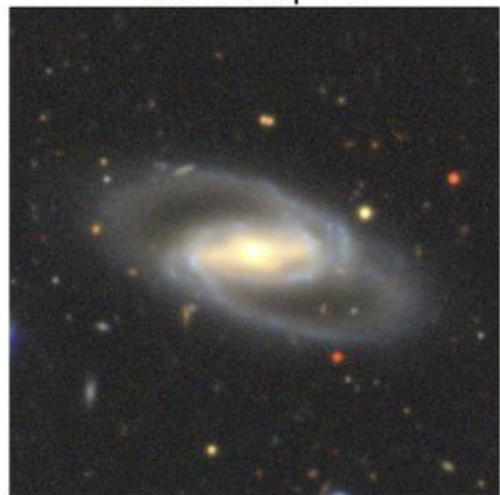
In-between Round Smooth



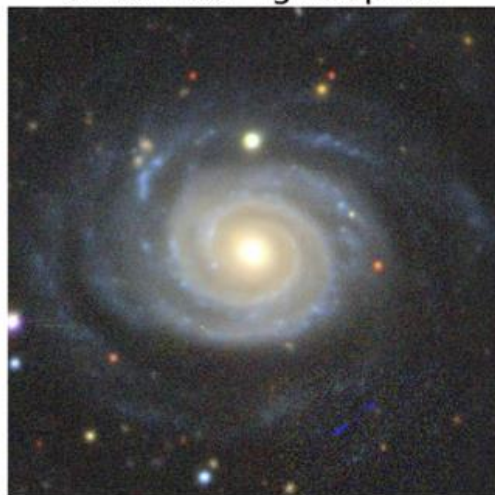
Cigar Round Smooth



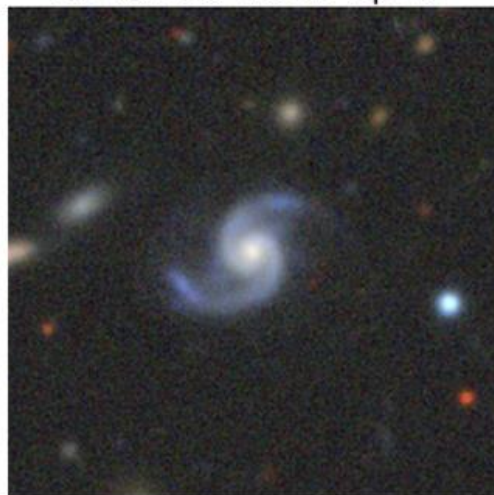
Barred Spiral



Unbarred Tight Spiral



Unbarred Loose Spiral



Edge-on without Bulge



Edge-on with Bulge



# Benchmarks

Disturbed

Merging

Round Smooth

In-between Round Smooth

Cigar Round Smooth

Table 3: Top-1 Accuracy on the Galaxy10 DECaLS morphology classification dataset.

Pretraining	Model	Top-1 Accuracy
No pretraining	EfficientNetB0	<b>80.9</b> $\pm 0.1$ %
	ConvNext-nano [144]	75.6 $\pm 1.8$ %
	ResNet18	73.9 $\pm 0.9$ %
	DenseNet121	73.5 $\pm 2.4$ %
Galaxy Zoo	ConvNext-nano [144]	<b>89.3</b> $\pm 0.1$ %
ImageNet-12k	ConvNext-nano [144]	83.9 $\pm 0.3$ %

Barred Spiral

Edge-on with Bulge

# Benchmarks

Table 4: Model  $R^2$  performance comparison for predicting galaxy properties from different observational data modalities.

Round Smooth      In-between Round Smooth      Cigar Round Smooth

Modality	Source Survey	Model	$Z_{HP}$	$\log M_*$	$Z_{MW}$	$t_{age,MW}$	$sSFR$
Image	Legacy Surveys	ResNet18	0.771	0.725	0.381	0.210	0.405
		DenseNet121	0.774	0.734	0.414	0.267	0.446
		EfficientNetB0	0.697	0.645	0.395	0.260	0.421
		Conv+Att [104]	0.982	0.871	0.659	0.488	0.679
Spectrum	DESI	MLP	0.696	0.681	0.383	0.308	0.343
		ResNet18					
Photometry	PROVABGS	DenseNet121					
		ConvNext-nano [144]					
No pretraining		Galaxy Zoo	ConvNext-nano [144]	<b>89.3 ± 0.1 %</b>			
No pretraining		ImageNet-12k	ConvNext-nano [144]	<b>83.9 ± 0.3 %</b>			



Barred Spiral

# Benchmark

Table 4: Model  $R^2$  performance compared to a baseline on observational data modalities.

Modality	Source Survey	Model	Performance
Image	LeGOS	BTSbot [119, 120]	0.985
		Random forest	0.90
		Avocado [20]	77.4
		Connect Later [117]	79.9
		Connect Later [117]	0.247

Table 5: Results for astronomical time-series tasks. The performance of each model on the associated metric for the task is shown in the Performance column. Avocado is a Random Forest architecture while Connect Later is a transformer architecture.

Source Survey	Task	Metric	Model	Performance
BTS	Transient Candidate Identification	AUC	BTSbot [119, 120]	0.985
YSE	SN Ia classification	AUC	Random forest	0.90
	14-way classification	Accuracy	Avocado [20]	77.4
	14-way classification	Accuracy	Connect Later [117]	79.9
PLAsTiCC	Redshift estimation	RMSE	Connect Later [117]	0.247
			Avocado [20]	0.581
			Connect Later [117]	0.210
			Avocado [20]	0.405
			Connect Later [117]	0.446
			Avocado [20]	0.421
			Connect Later [117]	0.679
			Avocado [20]	0.343
			Connect Later [117]	0.308

ConvNext-nano [144] **89.3 ± 0.1 %**

ConvNext-nano [144] **83.9 ± 0.3 %**

# Benchmarks

Table 6:  $R^2$  performance of zero-shot prediction of galaxy properties from image and spectrum. Avocado is a Random AstroCLIP embeddings following the strategy described in [112].

Modality	Source Survey	Model	Performance
Image	Legacy Surveys	RTSbot [119, 120]	0.985
Spectrum	DESI	random forest	0.90
Image	ZHP	Avocado [20]	77.4
Spectrum	log $M_*$	Later [117]	79.9
Image	ZMW	Later [117]	0.247

Table 5: Results for astro-forest architecture while Conn-forest associated metric for the task.

Source Survey	Task	$R^2$	$t_{age, MW}$	$sSFR$
BTS	Transient Candidate Ia	0.801	0.432	0.405
YSE	SN Ia classification	0.986	0.737	0.441
PLAsTiCC	14-way classification	0.879	0.584	0.435
	14-way classification		0.240	0.643
	Redshift estimation		0.441	
	Net121			
	ConvNext-nano [144]	<b>89.3</b> $\pm 0.3$ %		
	ConvNext-nano [144]	83.9 $\pm 0.3$ %		



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