

Combining neural networks and galaxy light subtraction for discovering strong lenses with HSC

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What is strong lensing?

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The phenomenon that we can see the multiple images of background galaxy when a foreground galaxy and bright background galaxy are aligned.

Usage of strong lenses

- The mass structure of galaxies
- The properties of dark matter
- High-resolution studies of distant sources
- The measurement of cosmological parameters
(e.g., Refsdal 1964; Wong et al. 2020)



Tanaka et al. (2016)

Searching for Strong Lenses

- Strong lenses are very rare : source galaxy, lensing galaxy, and earth should be in line
 - deep, wide-field, multiband imaging surveys that cover a large area of the sky (e.g., Sloan Digital Sky Survey (SDSS), Kilo-Degree Survey (KiDS), Hyper Suprime-Cam Subaru Strategic Program (HSC SSP))
 - We have to search for strong lenses from millions of galaxies
 - efficient searching method is needed
 - Citizen science (e.g., Space Warps; More et al. (2023))
 - Arc-finding algorithms (e.g., YATTALENS; Sonnenfeld et al. 2018)
 - Machine learning (e.g., Canameras et al. 2021; Shu et al. 2022)

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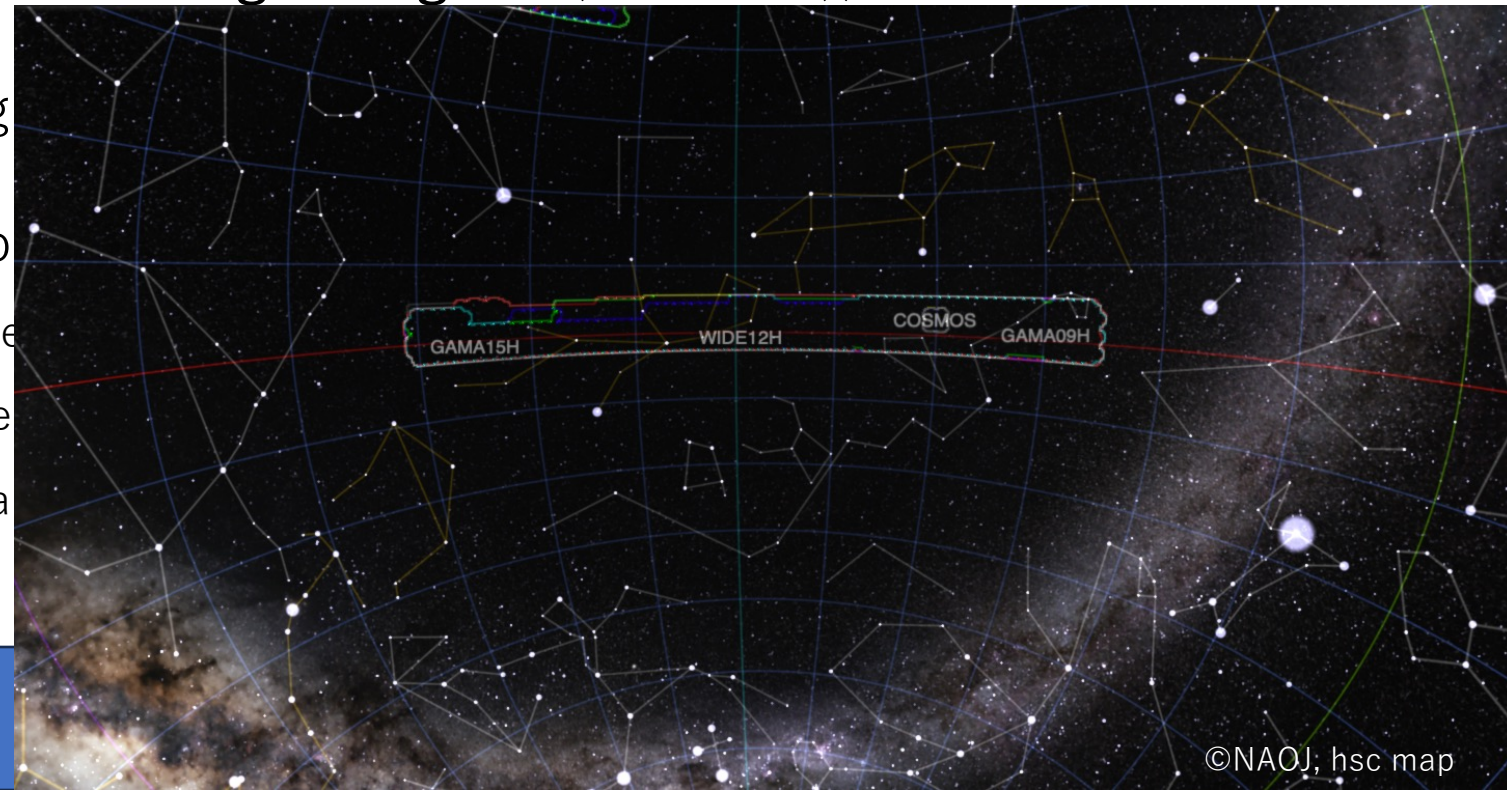
→ We have to search for strong

→ efficient searching methods

Citizen science (e.g., Spacewarps)

Arc-finding algorithms (e.g., Arclet)

Machine learning (e.g., CausticNet)



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⇒ **Combining**

Features of our Neural Networks

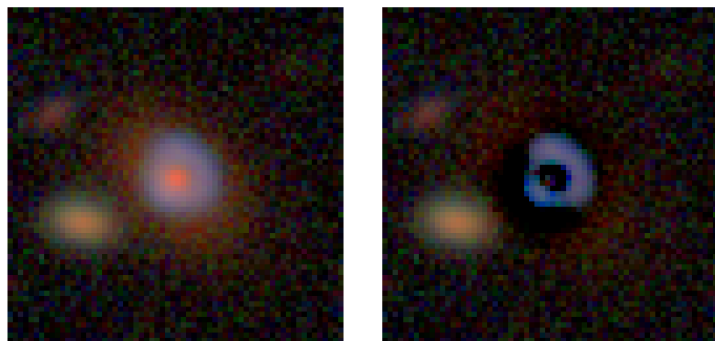
- Architectures
 - CNNs (Convolutional Neural Networks)
 - CNNs can process imaging data efficiently and identify characteristic patterns.
 - Auto tuning of hyperparameters with Keras Tuner
- Data
 - We used g, r, i-band images from HSC SSP wide (Aihara et al. 2019)
 - We used mock lenses as training dataset (Jaelani et al. 2023)
 - Subtraction of central (i.e., lens) galaxy with modified version of YATTALENS software (Sonnefeld et al. 2018)
 - highlight the characteristics of source galaxy light

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Training Dataset

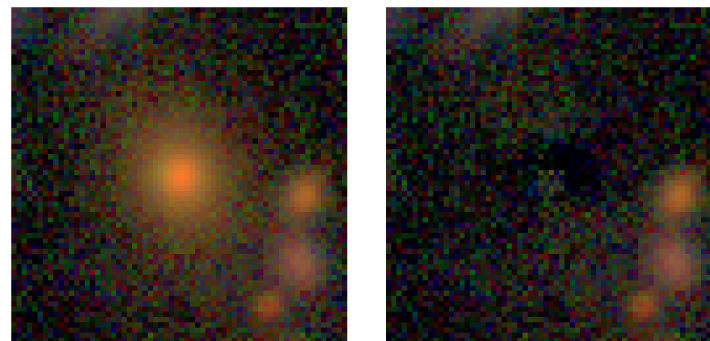
Mock Lens



Normal image

Subtracted image

Non-Lens



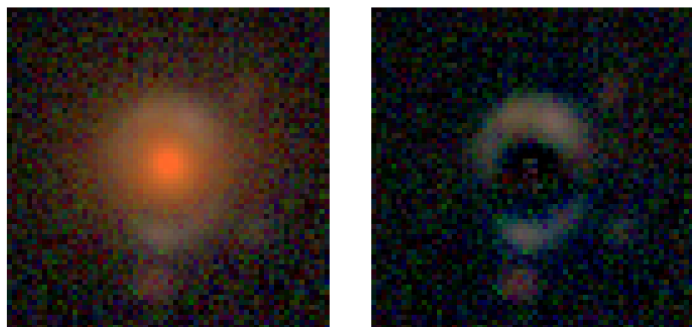
Normal image

Subtracted image

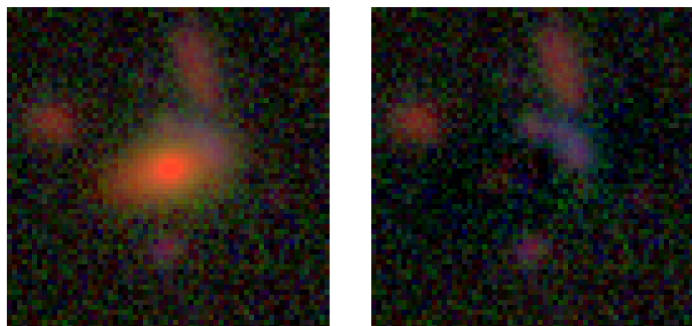
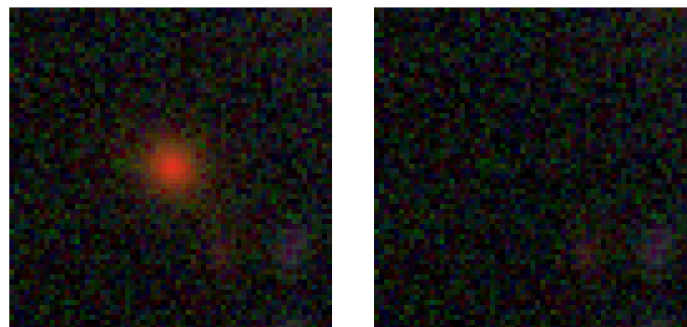
(Jaelani et al. 2023)

Test Dataset

Real Lens

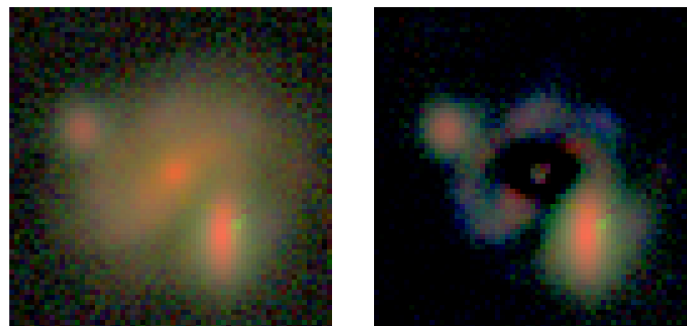


Non-Lens



Normal image

Subtracted image



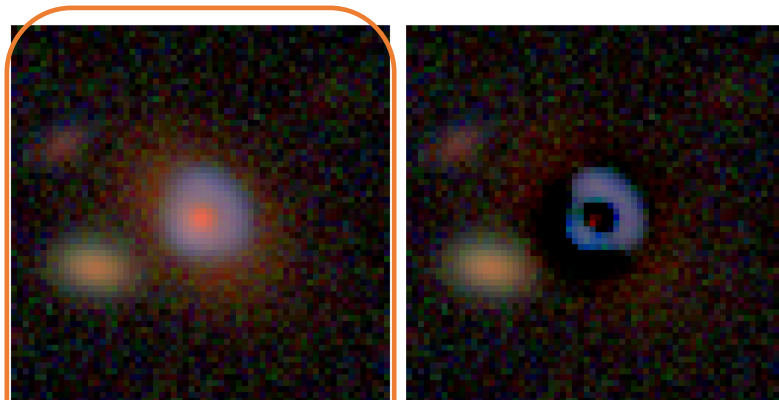
Normal image

Subtracted image

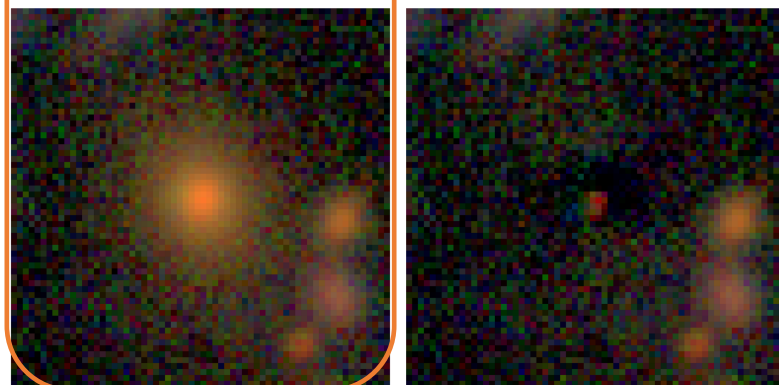
(Sonnenfeld et al. 2018, 2020; Wong et al. 2018)

Dataset

training dataset
mock lens



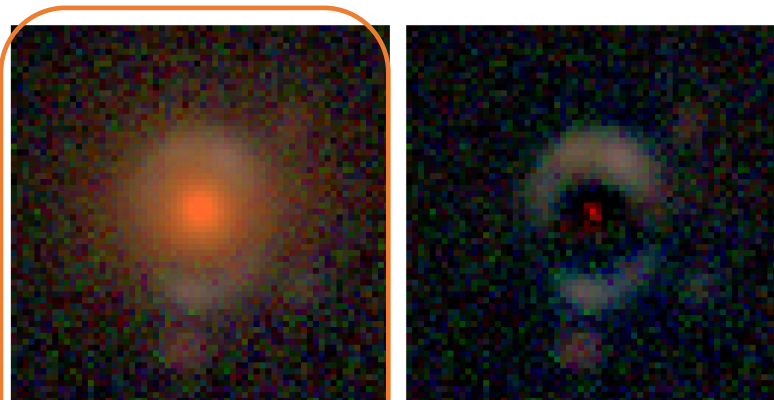
nonlens



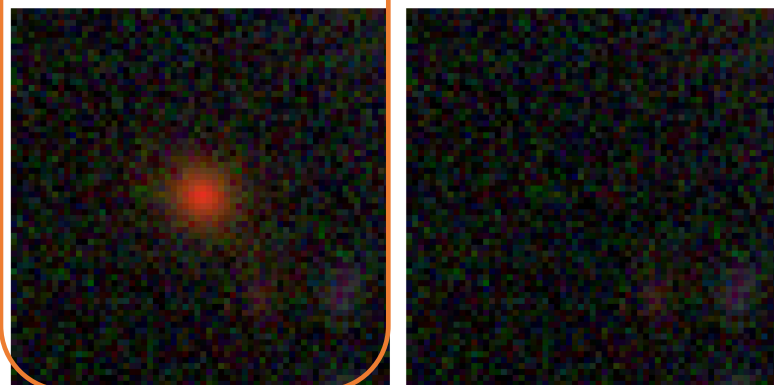
Normal

Subtracted

test dataset
real lens



nonlens



Normal

Subtracted

fid_dataset: The fiducial HSC gri-band imaging data

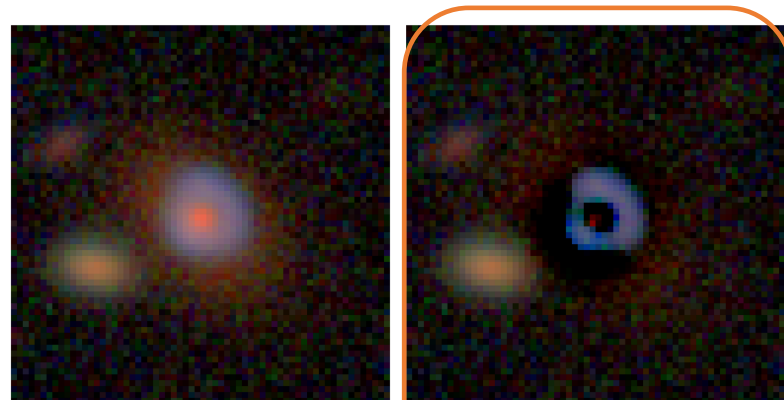
sub_dataset: The HSC imaging data with light subtraction

stack_dataset: The combination of “fid” and “sub”

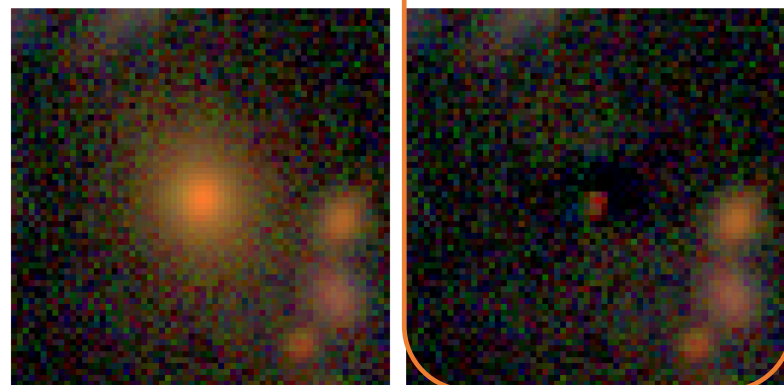
(Jaelani et al. 2023, Sonnenfeld et al. 2018, 2020; Wong et al. 2018)

Dataset

training dataset
mock lens



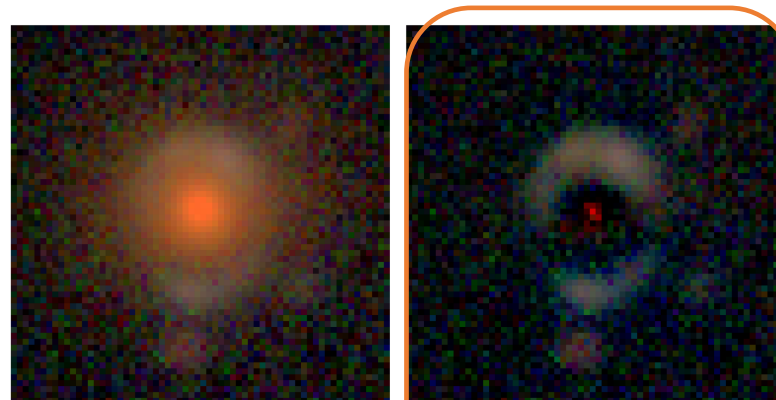
nonlens



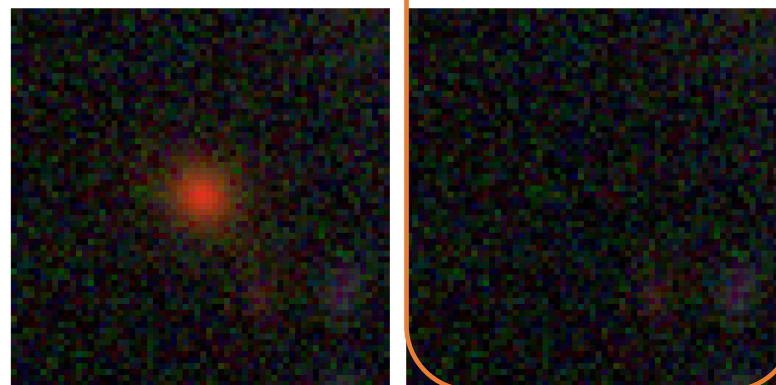
Normal

Subtracted

test dataset
real lens



nonlens



Normal

Subtracted

fid_dataset: The fiducial HSC gri-band imaging data

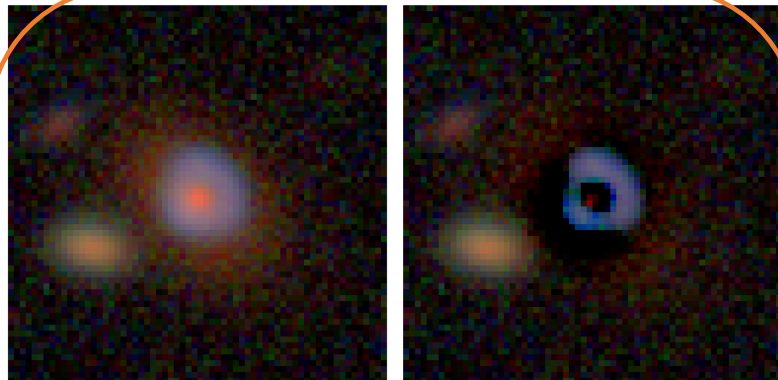
sub_dataset: The HSC imaging data with light subtraction

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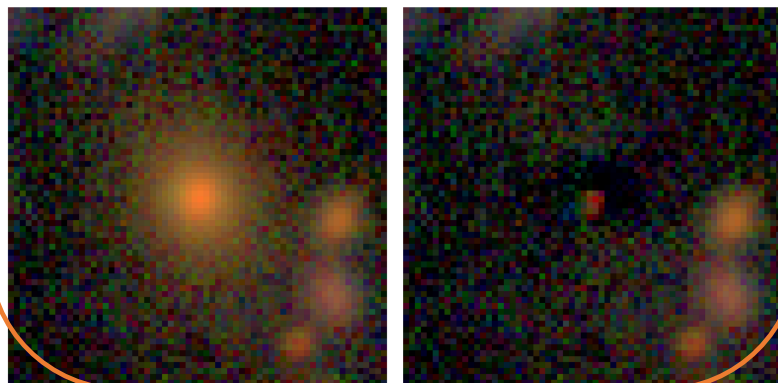
(Jaelani et al. 2023, Sonnenfeld et al. 2018, 2020; Wong et al. 2018)

Dataset

training dataset
mock lens



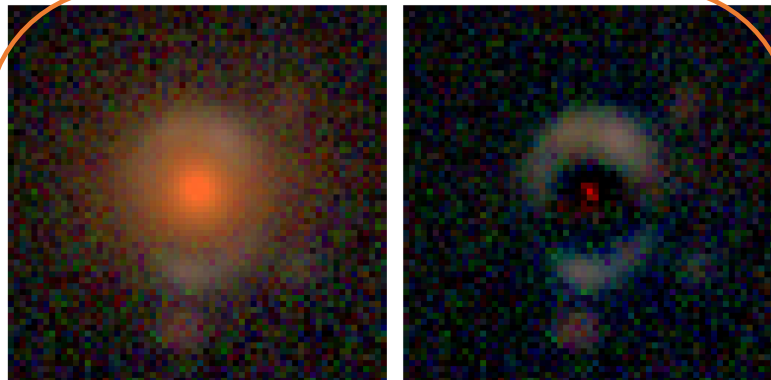
nonlens



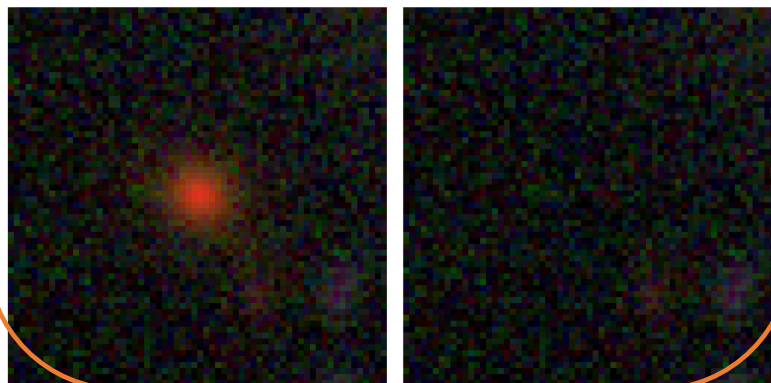
Normal

Subtracted

test dataset
real lens



nonlens



Normal

Subtracted

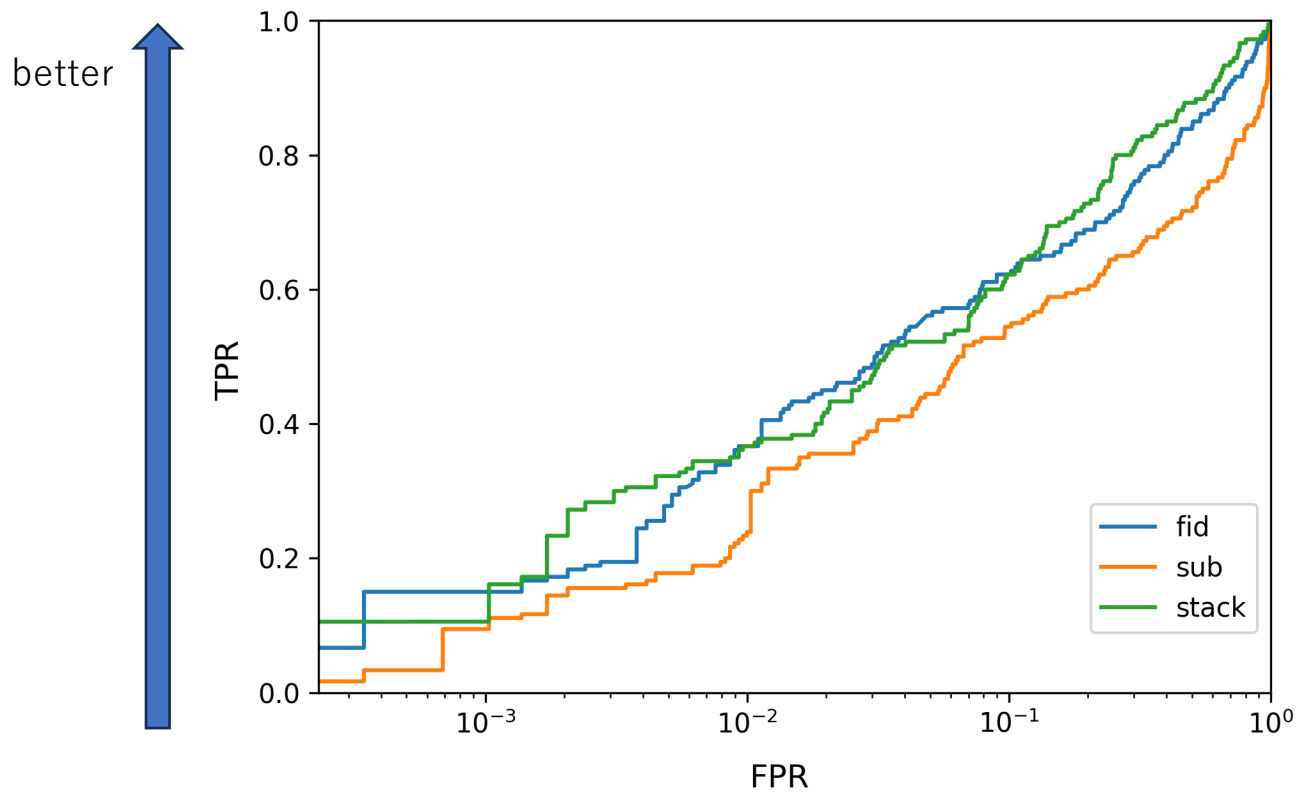
fid_dataset: The fiducial HSC gri-band imaging data

sub_dataset: The HSC imaging data with light subtraction

stack_dataset: The combination of “fid” and “sub”

(Jaelani et al. 2023, Sonnenfeld et al. 2018, 2020; Wong et al. 2018)

Comparison of neural networks optimized for each dataset



ROC curve of models optimized for each_dataset

Higher is better. Maximum is 1.



	AUROC
model_fid	0.808
model_sub	0.712
model_stack	0.837

©

Vertical axis: TPR (True Positive Rate)

$$\text{TPR} = \frac{\text{TP}}{\text{TP} + \text{FN}}$$

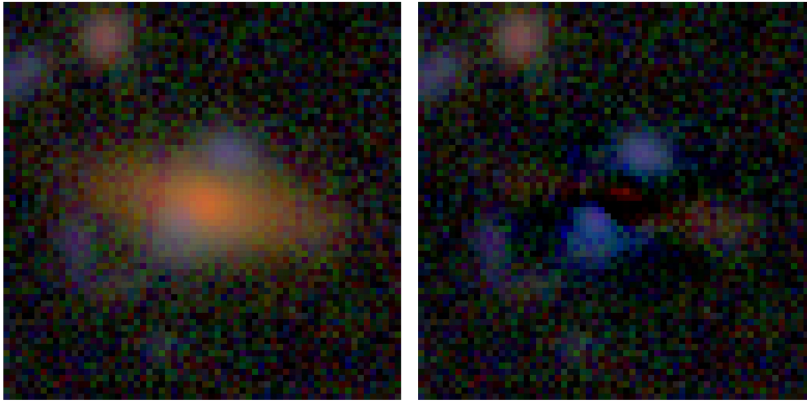
Horizontal axis: FPR (False Positive Rate)

$$\text{FPR} = \frac{\text{FP}}{\text{FP} + \text{TN}}$$

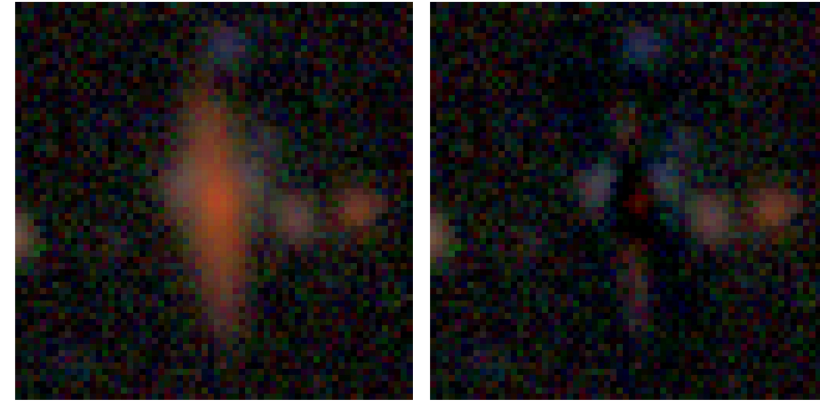
Fid vs. Sub vs. Stack (lens)

lens samples from test dataset (fid vs. sub vs. stack)

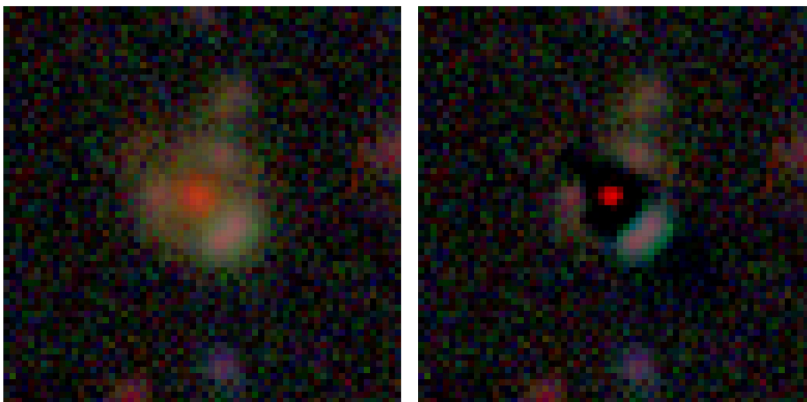
O (fid) – O (sub) – O (stack)



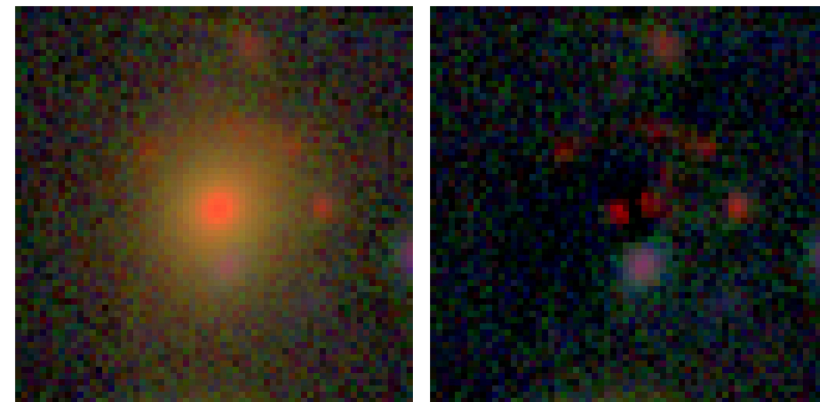
X (fid) – O (sub) – O (stack)



O (fid) – X (sub) – O (stack)



O (fid) – X (sub) – X (stack)

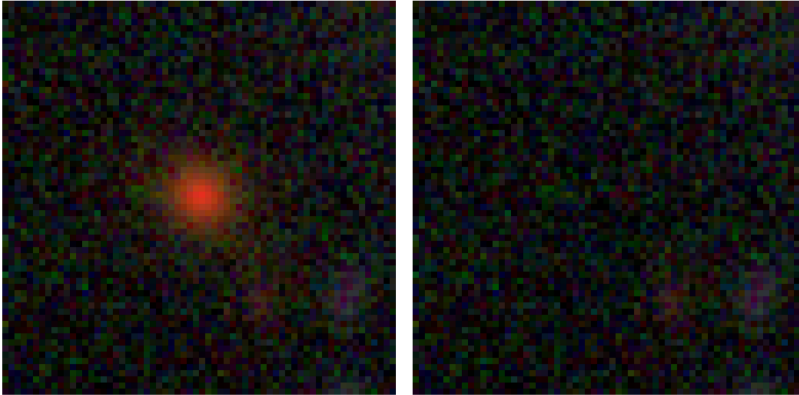


o = correct
x = incorrect

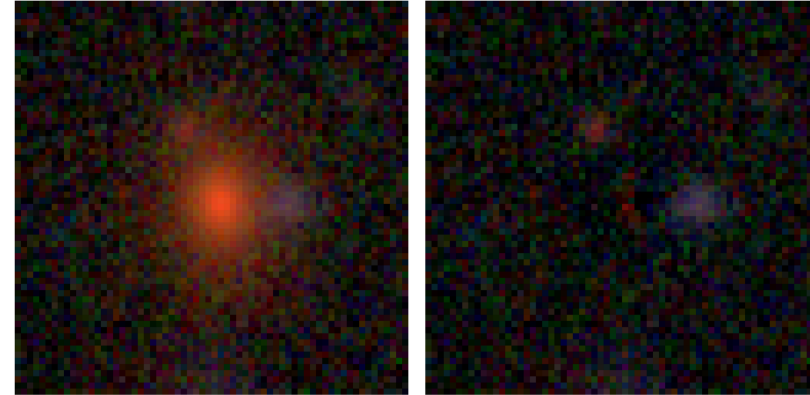
Fid vs. Sub vs. Stack (Non-lens)

Fid vs. Sub vs. Stack in nonlens

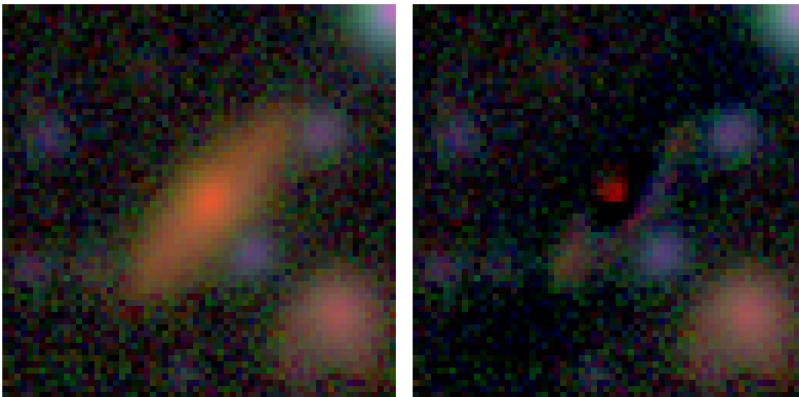
O (fid) – O (sub) – O (stack)



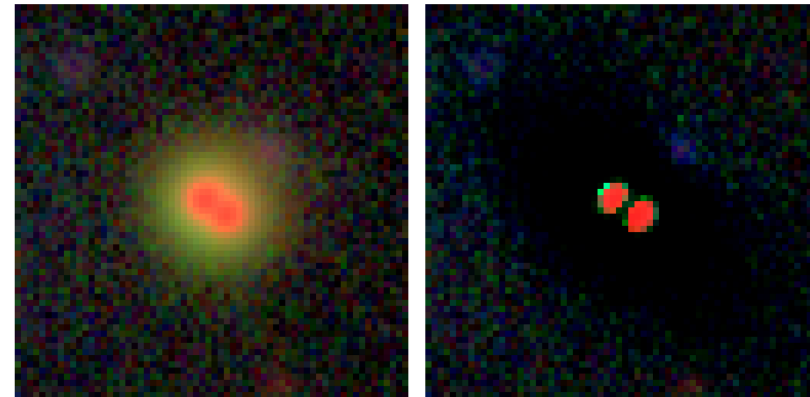
X (fid) – O (sub) – O (stack)



O (fid) – X (sub) – O (stack)



O (fid) – X (sub) – X (stack)



o = correct
x = incorrect

Collaboration Work

- Ensemble method (Holloway et al. 2023)
 - Our prediction is used for Ensemble method

Combined Classifier : Citizen Science (Sonnenfeld et al. 2020)

CNNs (Canameras et al. (2021), Shu et al. (2022), Jaelani et al. (2023), Ishida et al. (in prep))

- Comparison of different neural network with common test dataset (More et al. in prep)
 - compare our network with other CNNs (Canameras et al. (2021), Shu et al. (2022), Jaelani et al. (2023)) to evaluate performance

Summary and Future Work

- Summary

- We subtract the lens galaxy light to improve the performance of CNNs for discovering strong lenses in the HSC SSP
- combining subtracted images and fiducial images gives better performance (Ishida et al. in prep)

- Future Work

- We will reduce residuals in the center of subtracted images to improve the performance
- We will explore the specific examples more to investigate characteristics of misclassified objects.
- investigate other methods of light subtraction

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