Revealing systematics in strong lens substructure detection with CNNs

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Background: Strong lensing



CNNs and systematics in substructure detection

Background: Estimating sensitivity with ML



CNNs and systematics in substructure detection

Method: Training data

Galaxy isophotes and simulations show that <1% perturbations are very common.

Order 4 is more common than 3, as it fits boxy/discy isophotes.

This informs our model choice: a reasonable prior of <1%, and a more extreme version with <3%

See Naab & Burket (2003), Hao et al, (2006), Chaware et al (2014) Mitsuda et al (2017)

+ MP

PL

+ MP

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	Lensing components	Observation	Subhaloes
	Hubble deep field sources	S/N from 10 ¹ to 10 ³	$v_{\max} - r_{\max}$ to set concentration
Fγ _{ext}	Elliptical power- law mass	HST pixel scale and PSF	$M_{ m max}$ from 107.6 to 1011
	External shear	Poisson limited lens subtraction	Zero or 1-4, randomly placed
(<1%)	Order 1, 3, 4 MPs <1%		
(<3%)	Order 1, 3, 4 MPs <3%		

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Results: HST mock images





Results: Sensitivity mapping with multipoles



Including multipoles in the macro-model removes sensitivity away from the arc

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Sensitivity in and close to the arc remains similar, especially when going from modest to extreme multipoles

Results: Expected detections with multipoles



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With a smaller sensitive area, the number of expected detections drops by a factor of 4 for the 1% case.

But with only a small change in depth, the number of detections deviate from CDM at similar HM masses.

Conclusions

We tested the effect of angular structure on the sensitivity maps of 100 mock HST lenses. For multipoles up to 1% amplitude we find...



61% Loss in sensitivity area

Substructure detection efforts must allow for angular structure in the lens to remain reliable.

But, the constraining power on the underlying DM model should not drastically change.

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Orders of magnitude loss in depth

0.28

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Extra slides

Background: Sensitivity function

Find the best smooth model 1 Yields subhalo 0 Find corrections detections to the potential -1Calculate the Places limits on $^{-1}$ sensitivity of the non-detections observation*

*expensive!

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Despali et al (2021)



Method: Modelling angular structure

Order 1



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PL + 10% MP perturbation

PL unperturbed



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Results: Substructure false positives



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We add multipoles to our mock HST data and run the images through our PL-only model.

Substructure is detected very often for even modest multipole strengths.

Order 3 multipoles have the strongest degeneracy, followed closely by order 4.

Models with 1% and 3% MPs have zero false positives in their prior ranges

Results: Sensitivity mapping with multipoles



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Minimum sensitivity (M_{\odot})

Order 3



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Order 4

Sensitivity map uncertainty



Fixed Poisson

Varying Poisson noise in lensed

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Some of the noise in the image changes between making the sensitivity map and running the subhalo detections

This introduces uncertainty - what was detectable at 5σ in the SM realisation, might not be when we run the detections, and vice versa

Accounting for uncertainty will always boost the number of expected detections because the mass function is steep.

Sensitivity map uncertainty

Here I define the detectability of a subhalo: $\log \left(M_{\text{max}} / M_{\text{map}} \right)$



Running many realisations of the same single subhalo through the detector gives a distribution of detection significances, as a function of how detectable the subhalo is.

Repeating for many lenses and masses we can estimate the uncertainty as a function of mass

At all masses, $\sigma_{M_{\rm map}}/M_{\rm map} < 10^{-1}$



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Sensitivity map uncertainty is too small to fully explain the boost.



Multiple object detections



We draw pairs of substructures and record:

- Individual isolated detection significances s_0 and s_1
- Joint detection significance s_{01}

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S01

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We already compute s_0 and s_1 for all positions and masses when we produce the sensitivity maps.

We can map the substructures in the realisations to those in the map data and compute s_{01} for the two largest subhaloes in each realisation.

Accounting for the situation with two large but not quite detectable subhaloes in this way, we gain more expected detections...

Interactions between substructures



 d_{01} is the separation between the two most detectable substructures in each realisation

Interactions between substructures



In realisations where we did not expect a detection, the largest substructures were on average closer than in the rest of the population

Results: Detections from subhalo populations



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In warmer DM models, the excess of detections seen in CDM disappears.



n = 3.0

n = 4.0

n = 5.0























