

Statistics of flux ratios in strong lenses: probing of dark matter on small scales

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CDM:

- structure on all scales;
plethora of halos $< 10^8$
solar masses

WDM:

- no structure below a
'free streaming' scale

Self-interacting DM:

- cored density profiles

**How do flux ratios
help us tell the
difference?**

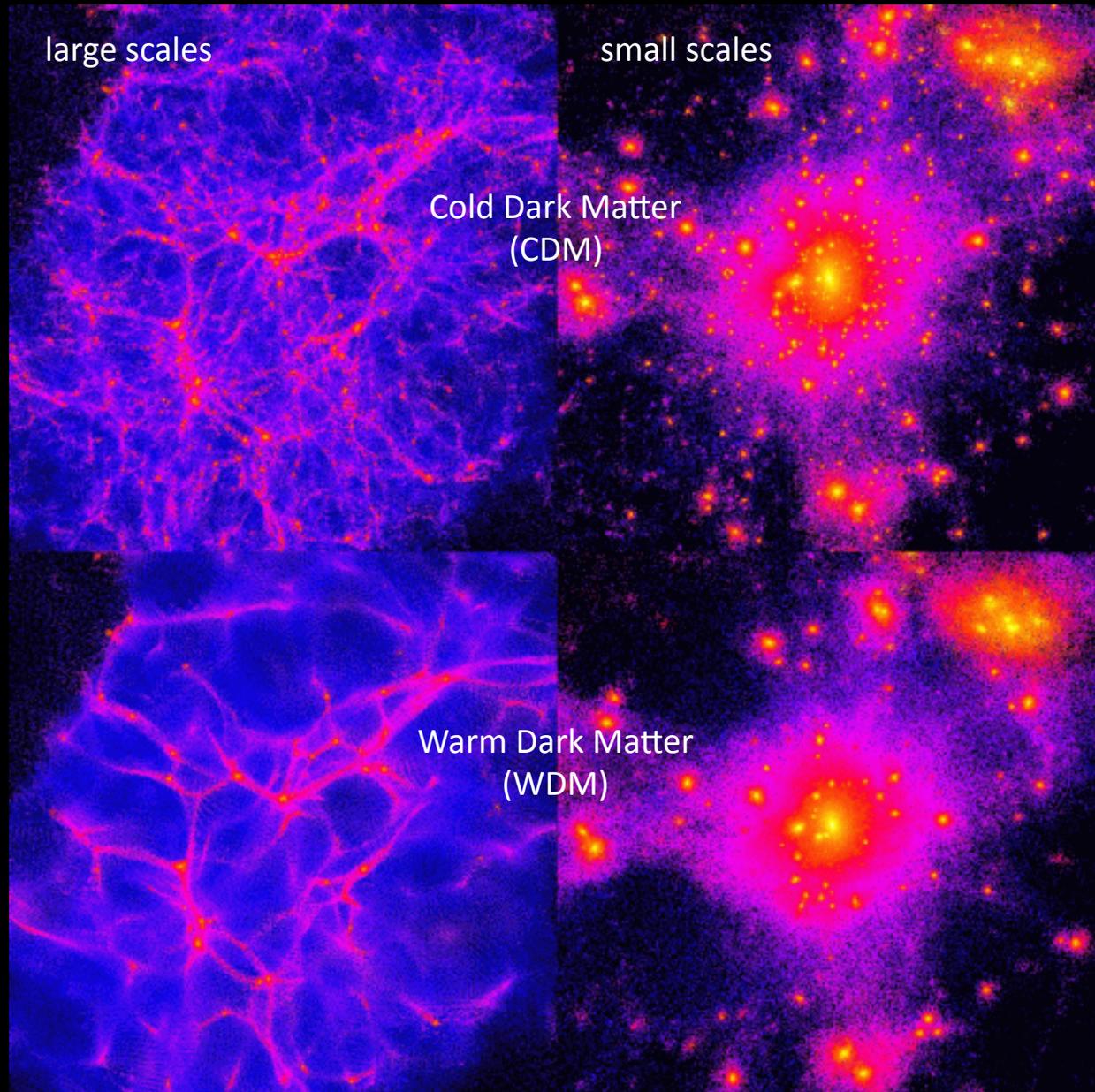
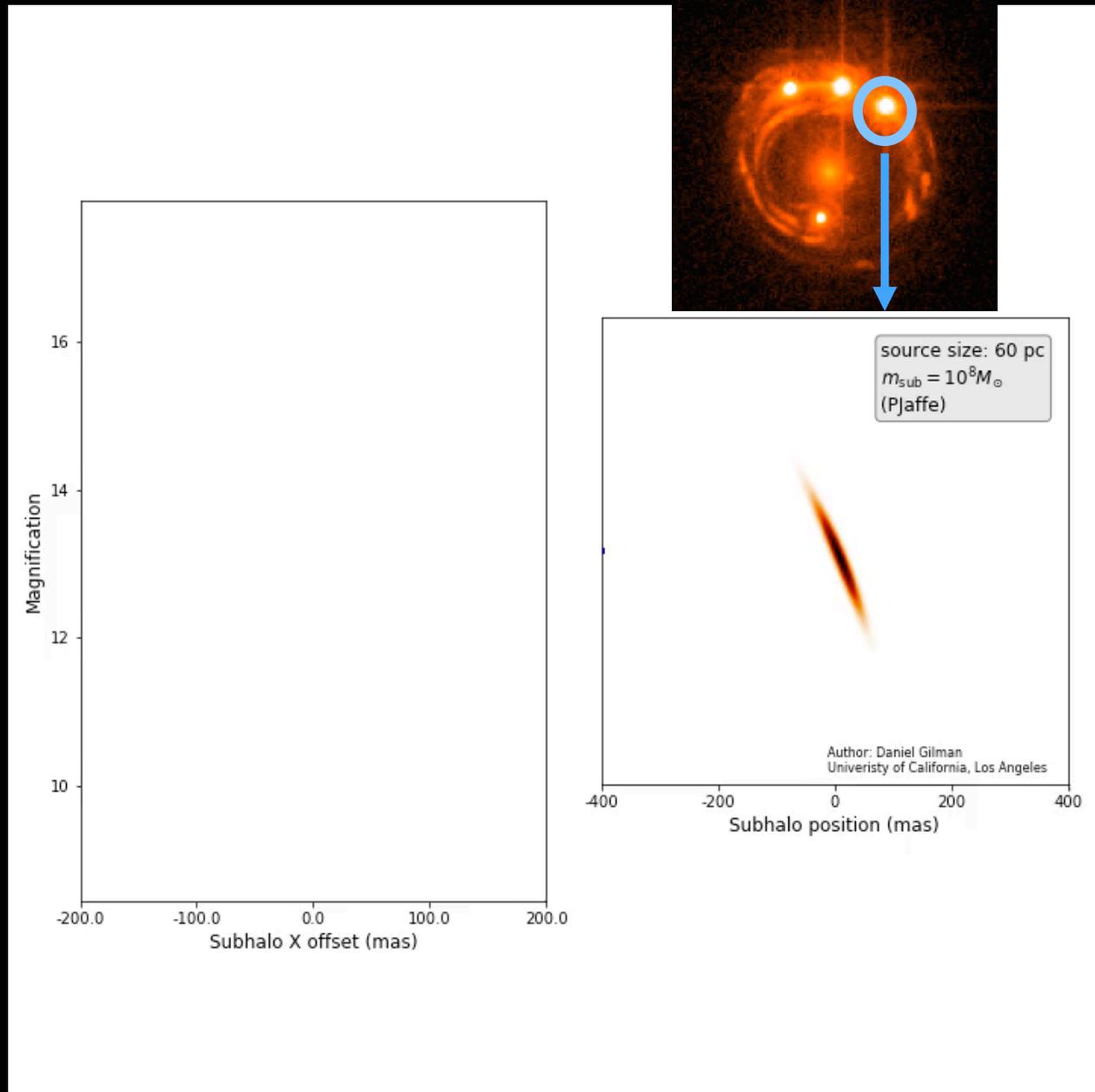


Image magnifications probe structure on small scales

Massive, dense objects
the the most efficient
lenses.

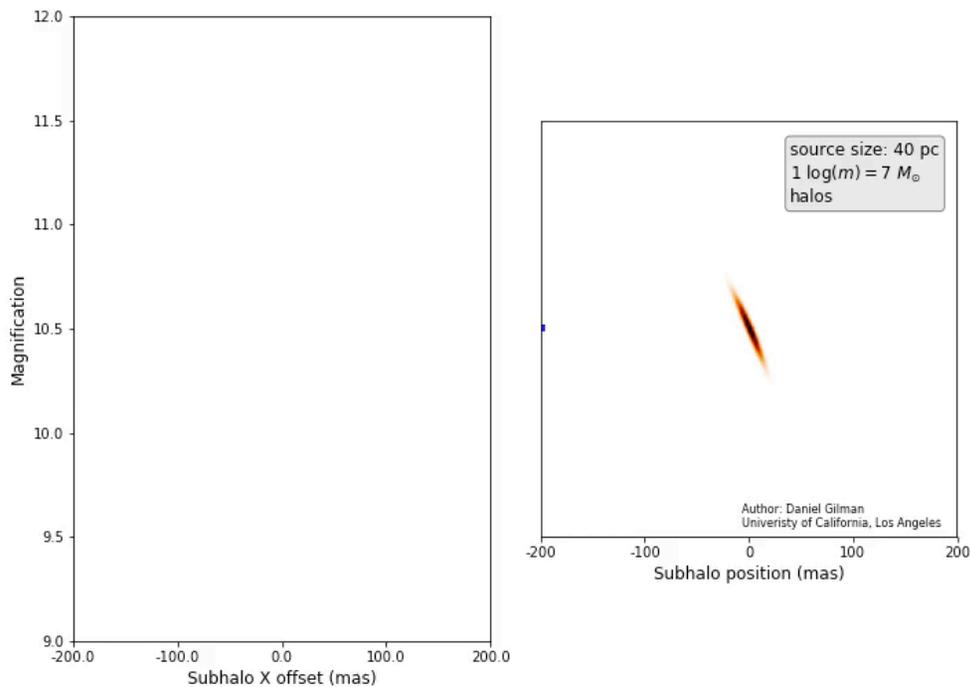
example: 10^8 solar mass
isothermal perturber



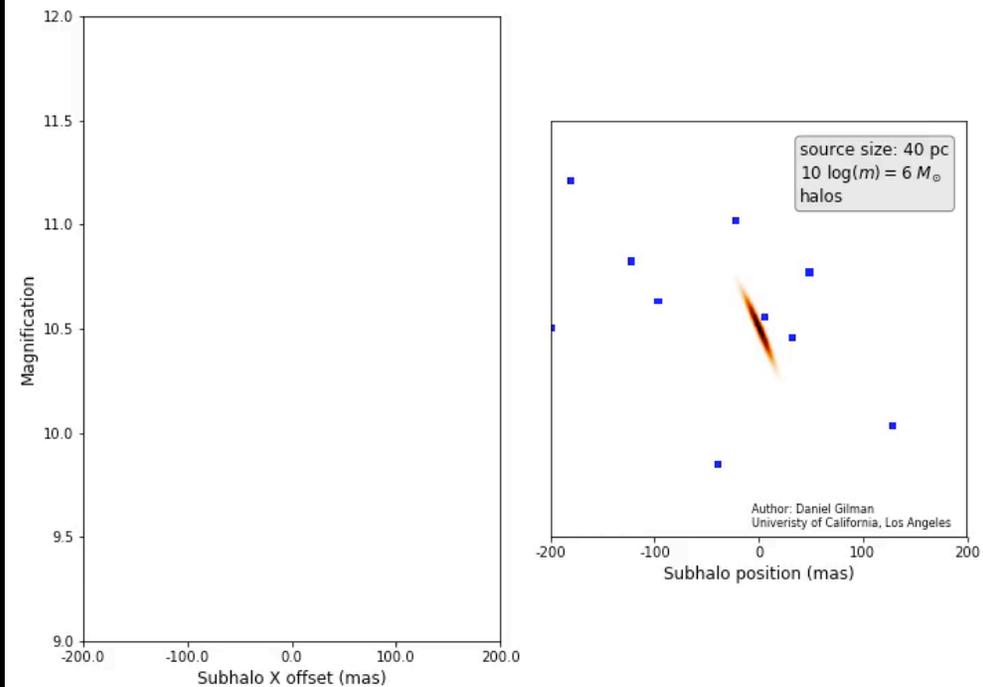
Subhalos are believed to be less dense (NFW)

Flux ratios are non-linear: multiple (smaller) subhalos can boost the signal

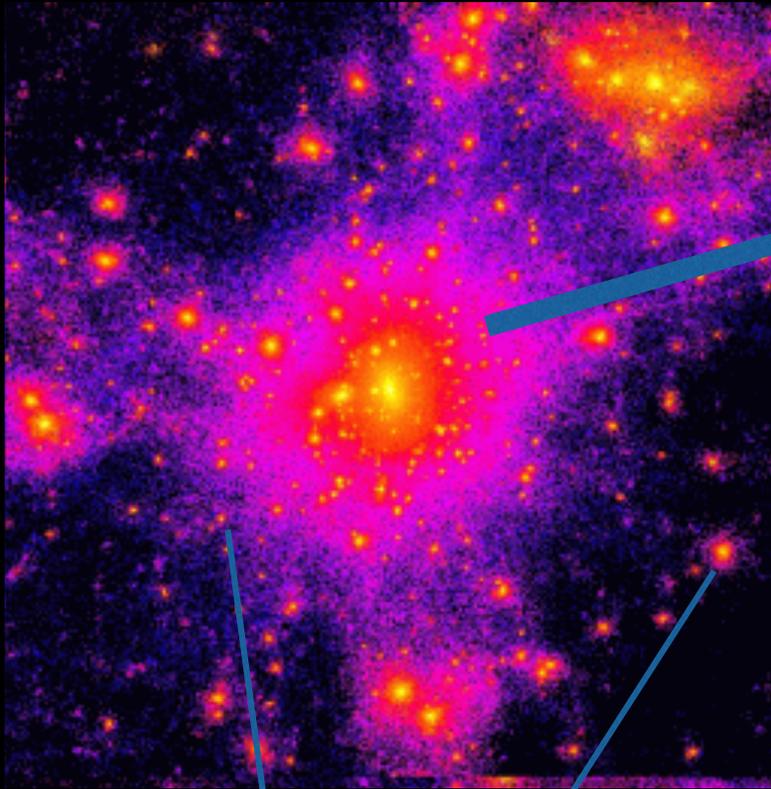
10^7 solar mass NFW halo



10×10^6 solar mass NFW halos



Modeling strong lenses

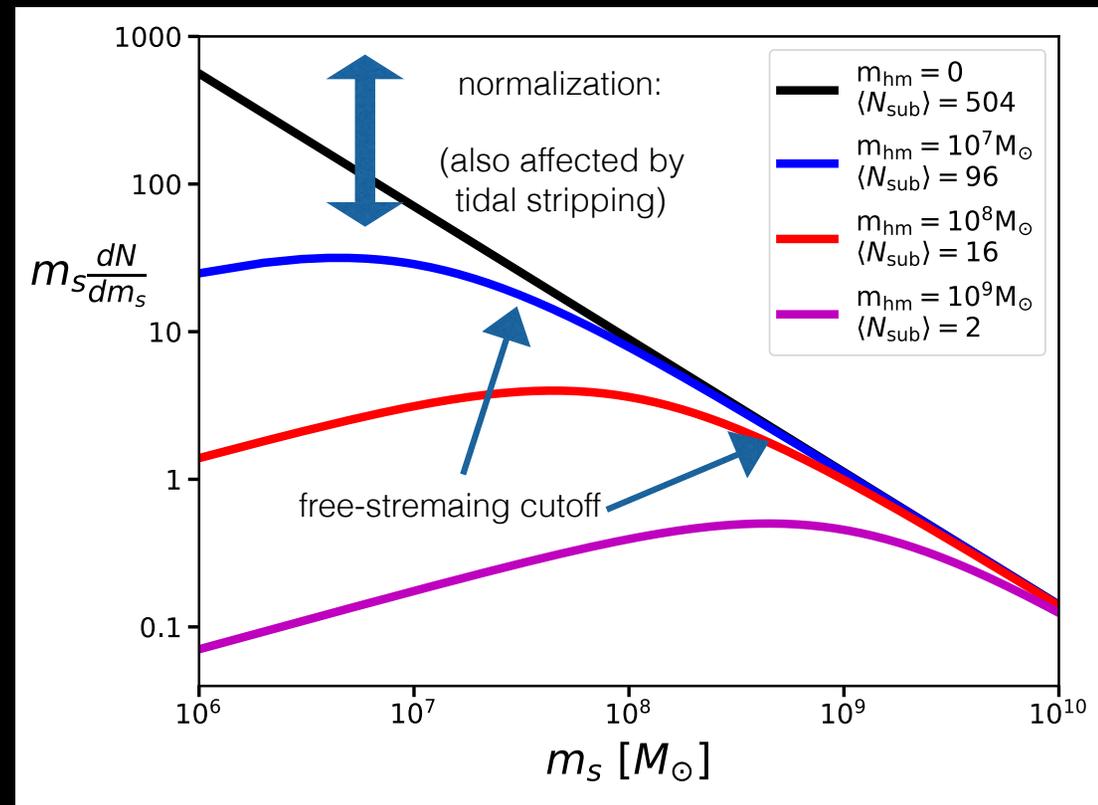


dark substructure:

- strongly perturb the fluxes
- dark matter theories predict normalization, shape of mass function

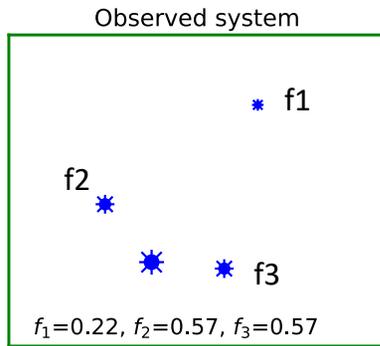
“macromodel”:

- luminous galaxy + smooth dark matter field + external shear
- 99% of the deflecting mass; image positions, time delays



Forward modeling procedure

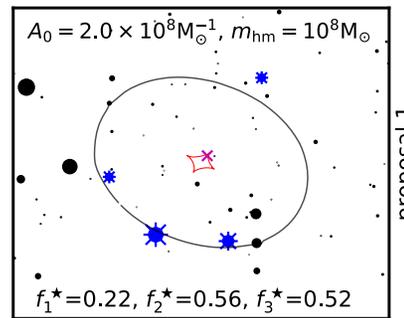
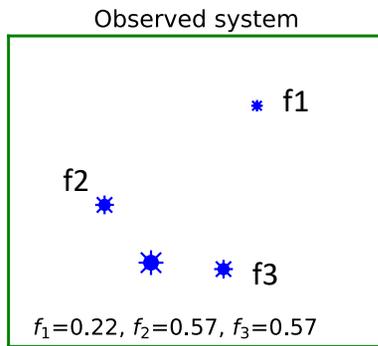
Forward modeling flux ratios step-by-step (1)



1. observe
positions, time
delays, flux ratios

Forward modeling procedure

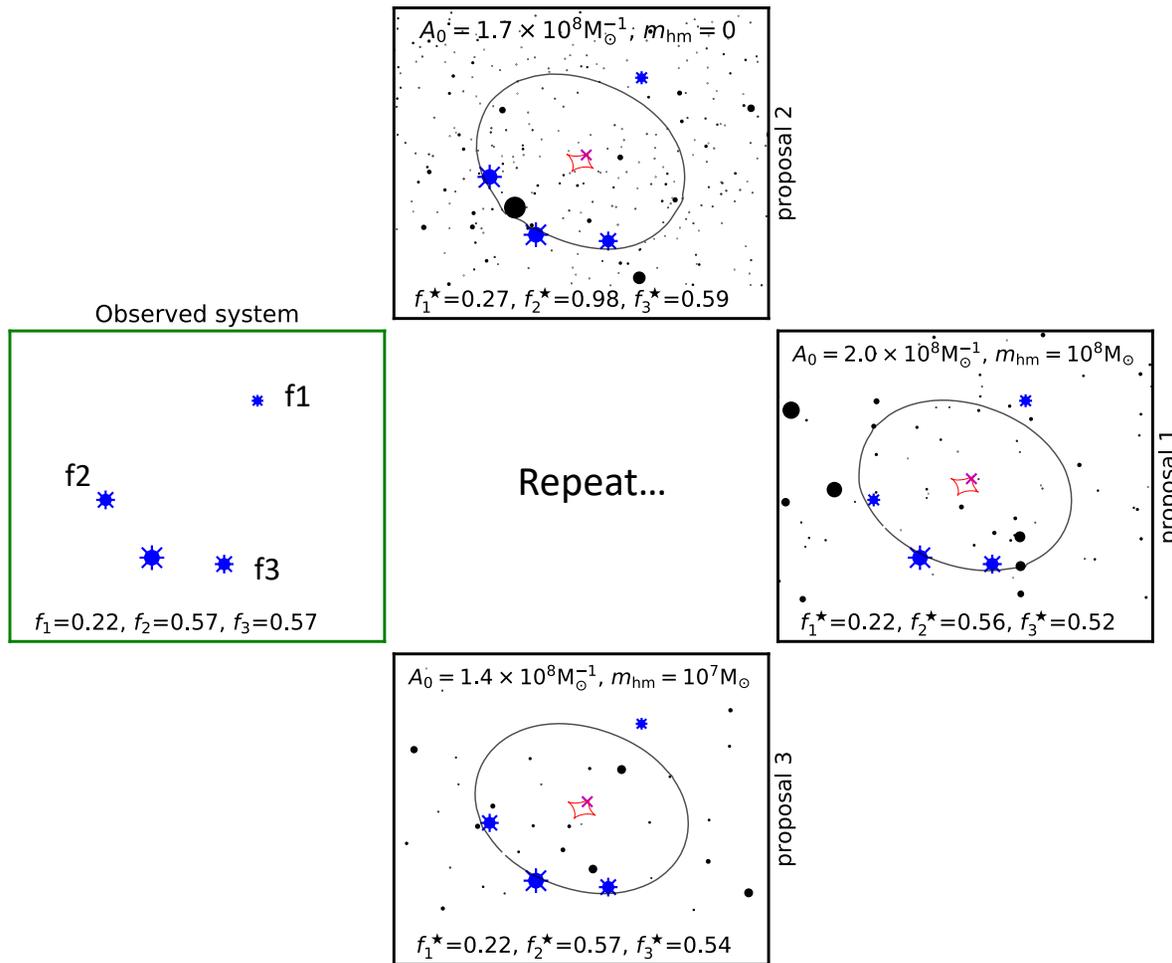
Forward modeling flux ratios step-by-step (2)



1. observe
positions, time
delays, flux ratios
2. Render
substructure
realization
 - A. re-fit
macromodel
to positions
 - B. compute
model flux
ratios

Forward modeling procedure

Forward modeling flux ratios step-by-step (2)

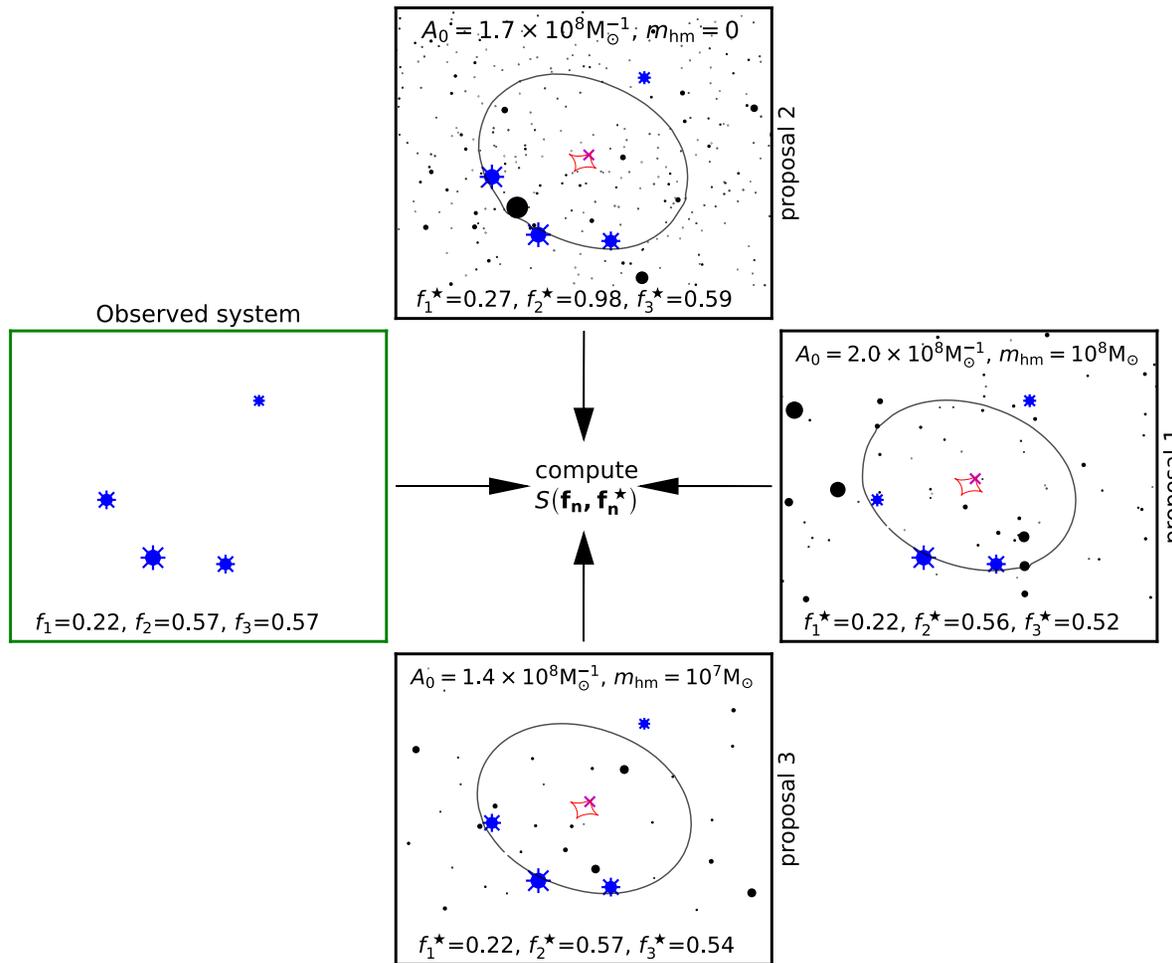


1. observe positions, time delays, flux ratios
2. Render substructure realization
 - A. re-fit macromodel to positions
 - B. compute model flux ratios
3. repeat 840,000 times per lens...

5.

Forward modeling procedure

Forward modeling flux ratios step-by-step (3)



1. observe positions, time delays, flux ratios
2. Render substructure realization
 - A. re-fit macromodel to positions
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3. repeat 840,000 times per lens...
4. compute a summary statistic:

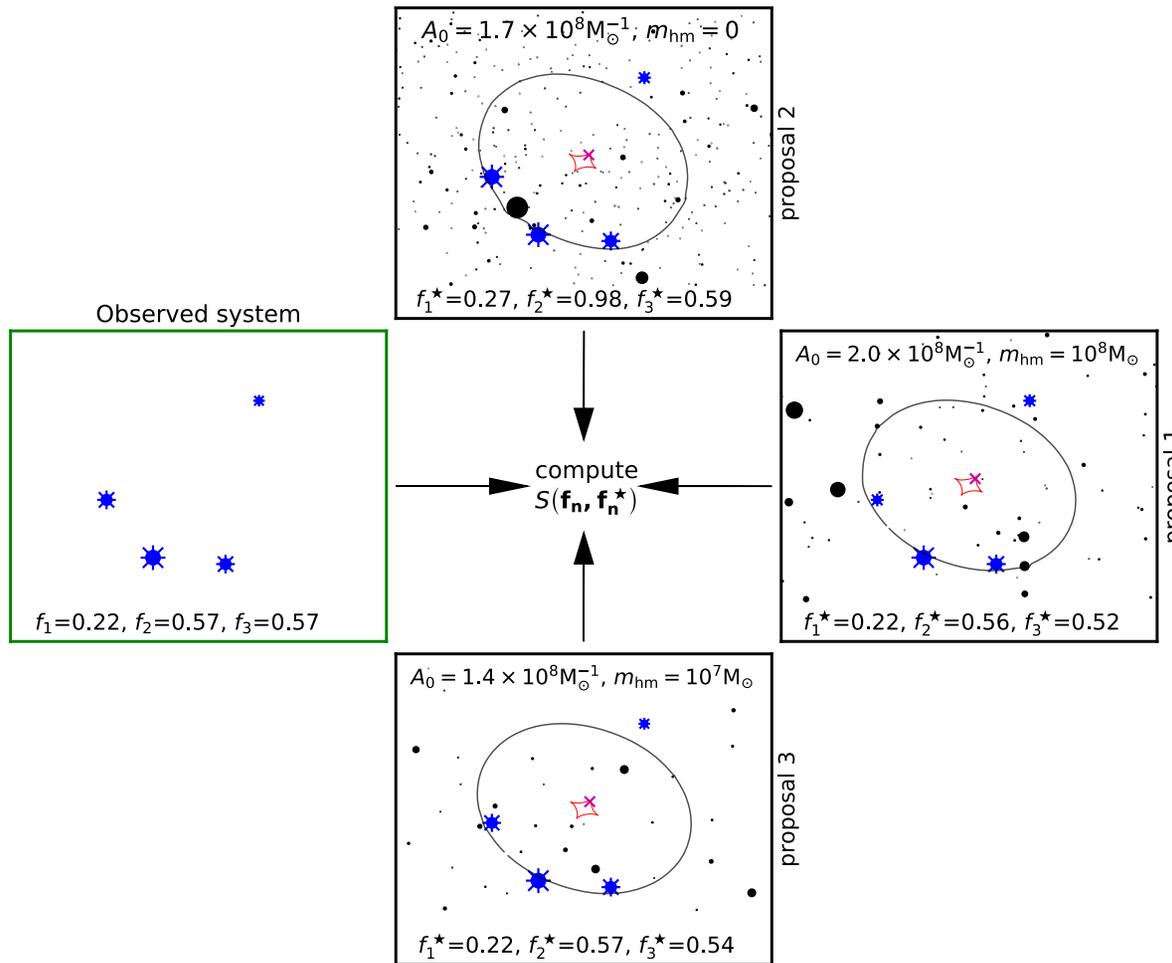
$$S(\mathbf{f}_n, \mathbf{f}_n^*) = \sqrt{\sum_{i=1}^3 (f_{n(i)} - f_{n(i)}^*)^2}$$

observed
flux ratio

model
flux ratio

Forward modeling procedure

Forward modeling flux ratios step-by-step (3)



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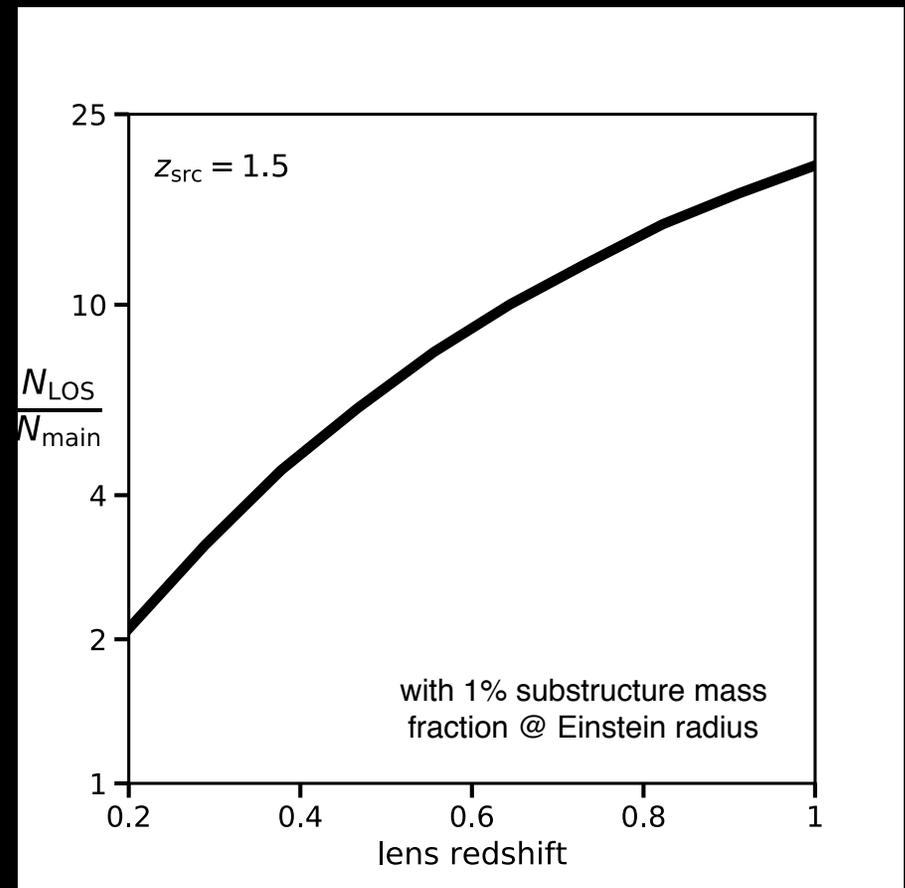
model
flux ratio

5. select based on summary statistic

Note: for now we are considering only lens plane halos

- the line of sight is important and boosts the signal!
(see e.g. Despali et al 2018)

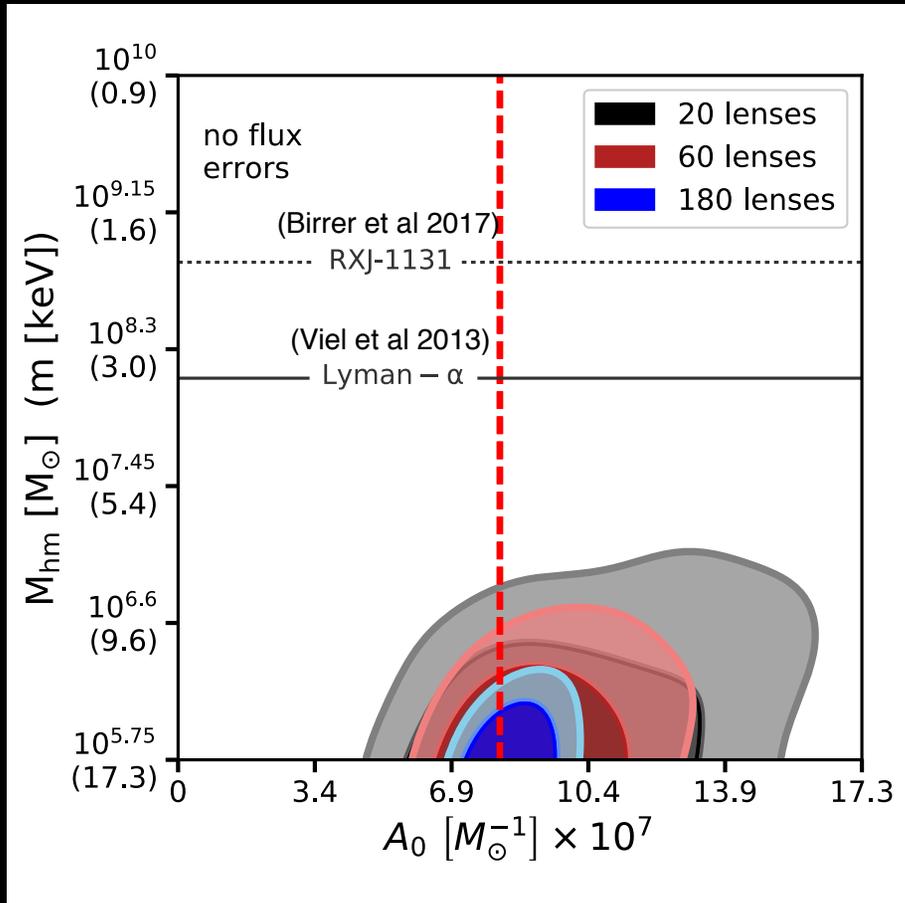
- By omitting the extra signal from the LOS our projected constraints are lower limits



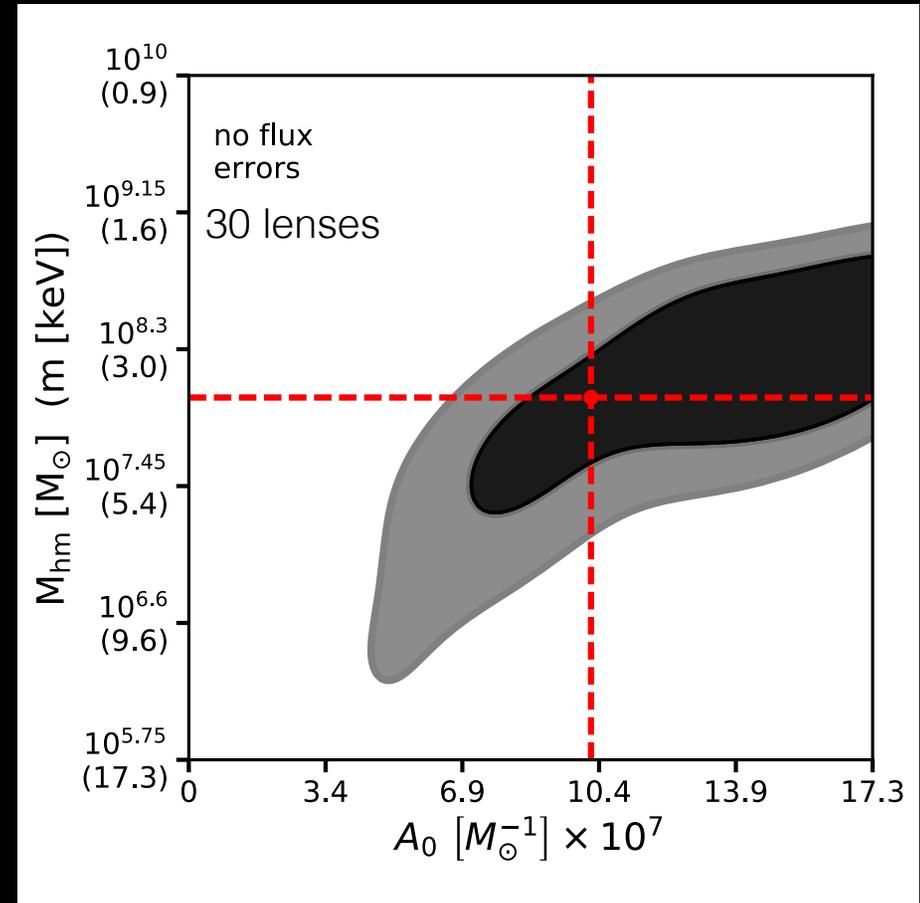
Does this method work, in principle?

Mock data: SIE+shear, plus substructure

CDM



WDM



But what about more realistic lens models?

Towards more realistic lenses

- We distinguish between dark matter mass functions through flux ratios computed w.r.t. a macromodel

An accurate macromodel (and no microlensing) is therefore crucial!

- SIE+shear can yield incorrect fluxes at the percent level (Gilman et al 2017, Hsueh et al 2017)

Towards more realistic lenses

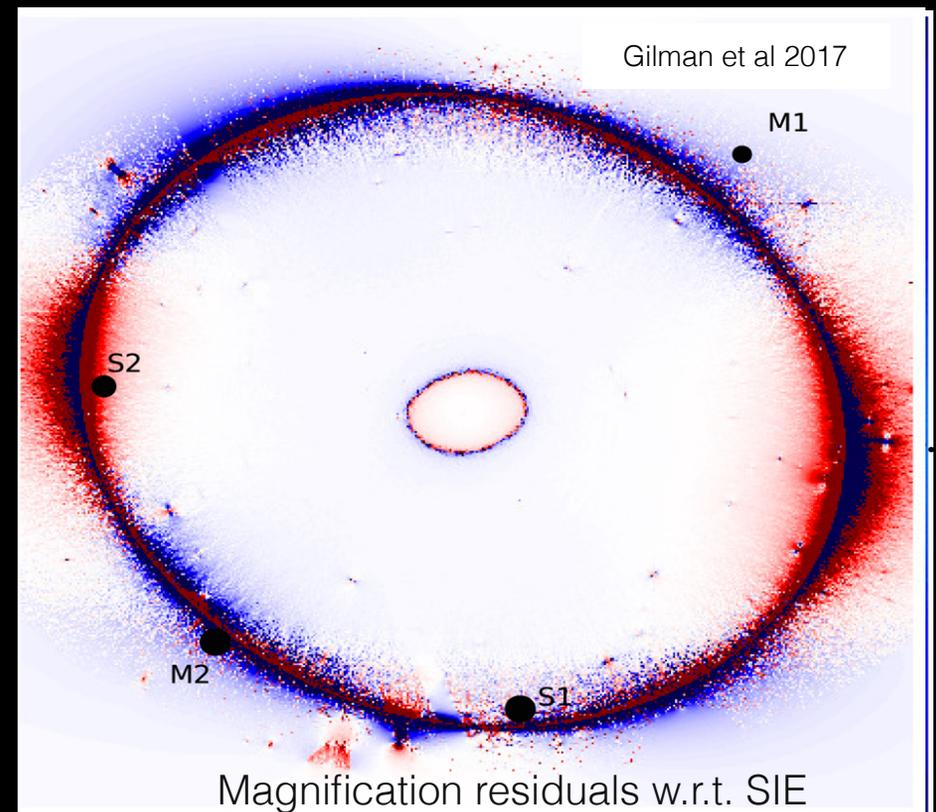
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Example (right): Surface mass density of Virgo galaxy VCC1062 plus NFW halo (turned strong lens @ $z = 0.5$)

How can we handle these uncertainties?



Towards more realistic lenses; handling uncertainties

1. **Modeling**: Include additional features (e.g. the power law slope) in the forward model
 - can also explicitly model features like disks (e.g. Hsueh et al 2016, 2017)

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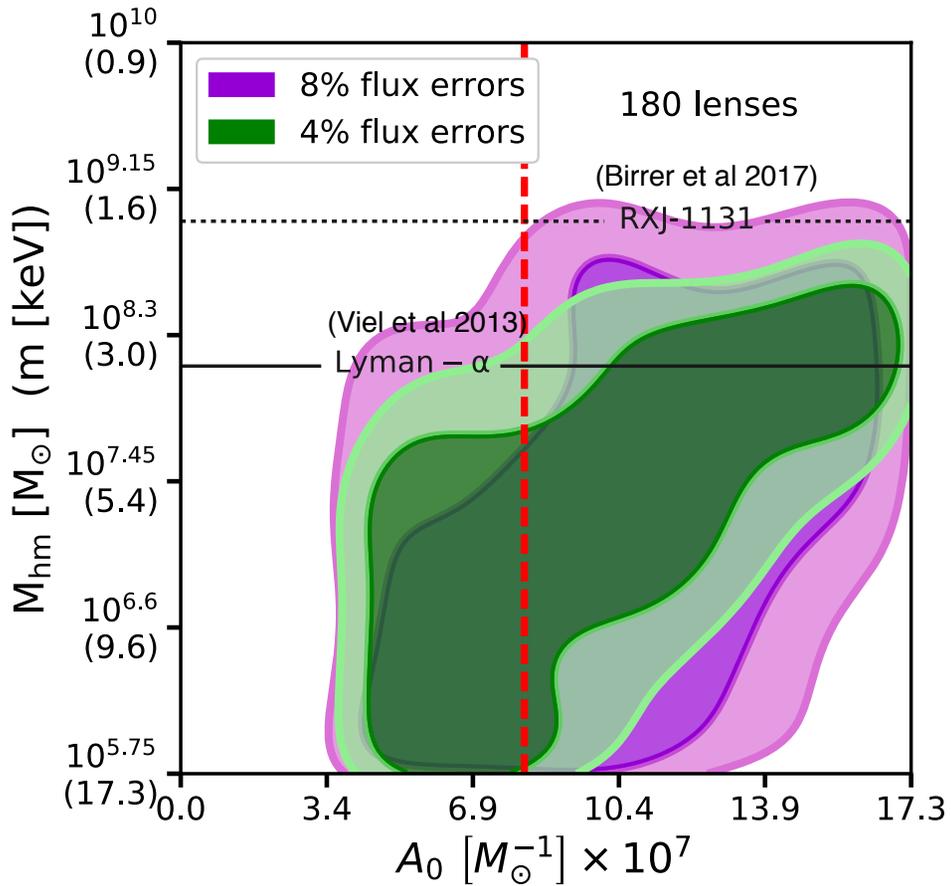
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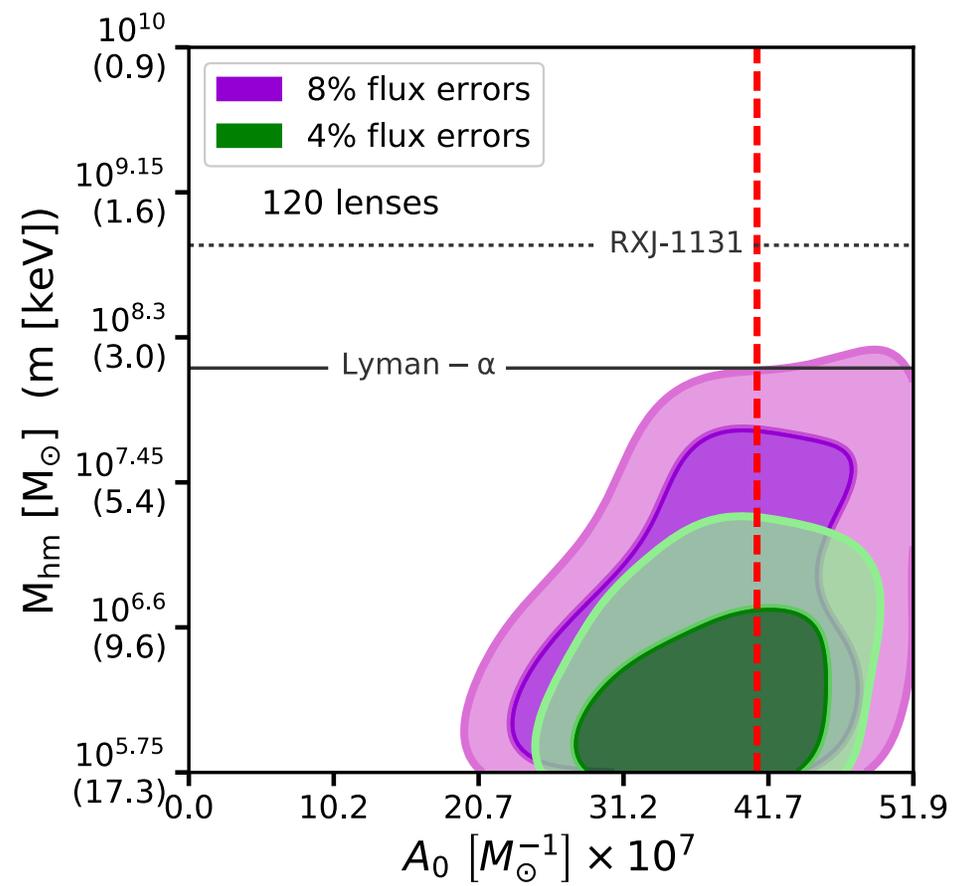
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How does the precision of the inference change after introducing uncertainties in flux ratios?

Inference with flux perturbations in mock data, forward model



Gilman et al 2018

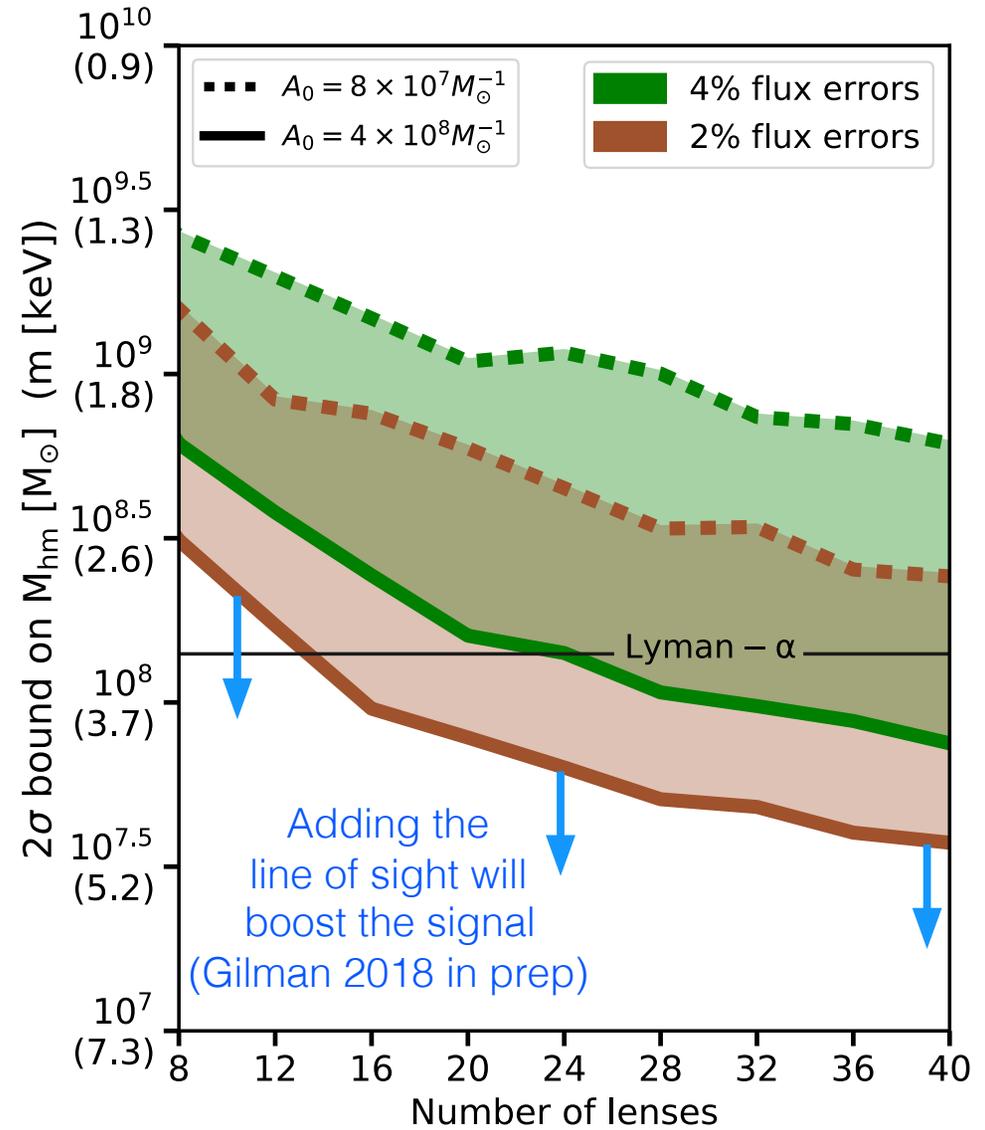
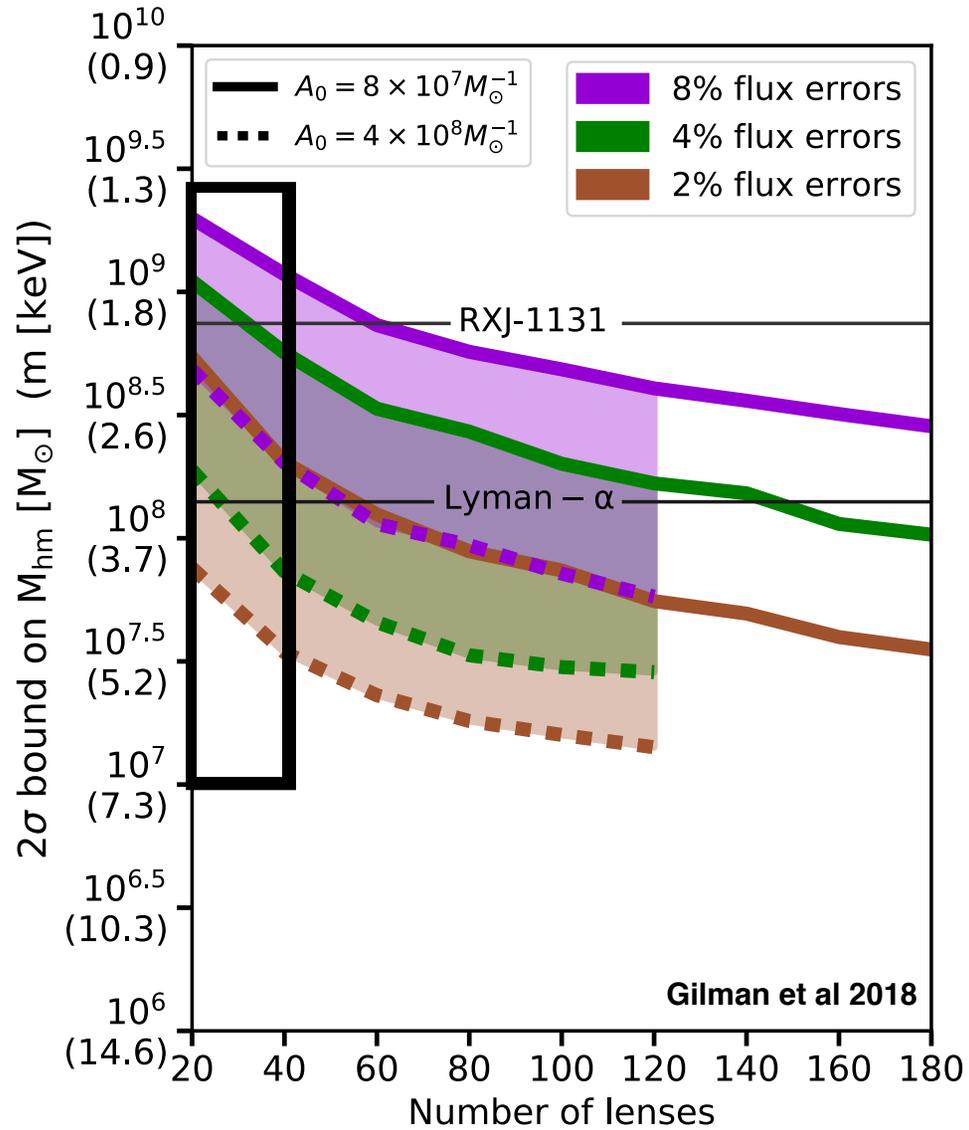


Gilman et al 2018

Normalization: 10^{13} mass parent halo; **all** subhalos within scale radius destroyed by tidal effects

Normalization: $10^{13.5}$ parent halo; **~50%** of subhalos within scale radius destroyed

The Future...



Takeaways

With ELTs affording < 5 m.a.s. precision in image positions and $< 4-8\%$ precision in image fluxes, we can measure:

1. The shape of the subhalo mass function
2. The normalization of the subhalo mass function

testing a fundamental prediction of particle dark matter theories.

Thanks!