# Strongest gravitational lensing limit on the dark matter free streaming length with JWST

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# **DM Figure of Merit**

- 2 classes of physical observables from astro DM probes
- Is DM cold? Is DM collisionless?
- Free streaming? Interactions?
- Abundance of DM halos, profiles of DM halos
- self-interaction cross-section and particle mass



Drlica-Wagner+ 1902.01055



#### Substructure Lensing

• Gravitational potential:  

$$\psi(\vec{\theta}) = \frac{D_{DS}}{D_D D_s} \frac{2}{c^2} \int dz \Phi(D_D, \vec{\theta}, z)$$

• Deflection:  $\overrightarrow{\alpha}(\overrightarrow{\theta}) = \overrightarrow{\nabla}\psi$ 

Magnification: 
$$M^{-1} = \delta_{i,j} - \frac{d\psi}{d\theta_i d\theta_j}$$

• Image positions probe large scale (smooth) fluxes probe small scale (clumpy)



pyHalo: Gilman+ (1908.06983) MNRAS



# Flux Ratio Anomalies

QUASAR

• Four distinct images Lensed quasars • DM from (anomalous) fluxes different than predictions of main deflector

> FOREGROUND GALAXY WITH FOUR QUASAR IMAGES

#### JWST



#### Strong Lensing of Quasars

- Different regions -> different physical sizes and emit in different wavelengths -> independent datasets
- Smaller source sizes are more sensitive to lower mass DM halos
- Observe narrow-line region with HST/Keck and warm dust with JWST



#### Urry et al ASP 1995

#### **Source Sizes Matter**

- Want flux ratios from cold torus / warm dust region (~5 pc)
- More sensitive to low mass DM halos than narrow-line (~80 pc)
- JWST observations are sensitive to completely dark DM halos



Nierenberg+ (2309.10101) MNRAS



# **Statistical Signals**

- Wide range of substructure realizations can reproduce observed flux ratios
- Need to evaluate the frequency that a given set of parameters will fit the data well







Gilman+ (1908.06983) MNRAS



### Anomalous Fluxes

- Ratios of fluxes from quadruply lensed images
- Ratio to divide out unknown flux of source



- Model predicts *distribution* of fluxes, not a single value
- Need machine learning, simulation-based (likelihood-free) statistical inference





# Warm Dark Matter

- Cold DM predicts an abundance of DM halos at small masses
- Warm DM predicts few DM halos: structure suppressed at small scales
- Constrained by observing smallest DM halos
- Masses ~1-20 keV
- e.g. sterile neutrinos



### Halo Mass Function

- Half-mode mass ( $M_{\rm hm}$ ): scale where mass function suppressed by half
- Determined by the DM mass

• 
$$M_{\rm hm} = 5.5 \times 10^{10} \left(\frac{m_{\rm WDM}}{1 \,{\rm keV}}\right)^{-3.33}$$



Keeley+ (2301.07265) MNRAS



### Modeling halos

$$\frac{d^2 N}{dMdV} = \delta_{\text{LOS}} \left( 1 + \xi_{2\text{halo}} \left( M_{\text{host},z} \right) \right) \frac{d^2 N}{dMdV} \bigg|_{\text{ST}} \text{ (line-of-sight halos)}$$

$$\frac{d^2 N_{\text{WDM}}}{dMdV} = \frac{d^2 N_{\text{CDM}}}{dMdV} f_{\text{WDM}} \left( M, M_{\text{hm}} \right)$$

$$f_{\text{WDM}} \left( M, M_{\text{hm}} \right) = \left( 1 + \left( \alpha \frac{M_{\text{hm}}}{M} \right)^{\beta} \right)^{\gamma} \alpha = 2.3, \ \beta = 0.8, \ \gamma = -1.0$$

$$\frac{d^2 N_{\text{sub}}}{dMdA} = \frac{\Sigma_{\text{sub}}}{10^8 M_{\odot}} \left( \frac{M}{10^8 M_{\odot}} \right)^{\alpha} \mathcal{F}(M_{\text{host}}, z) f_{\text{WDM}} \left( M, M_{\text{hm}} \right), \text{ (subhalos)}$$

#### Modeling halos **Ingredients for pyHalo**

- Concentration mass relation  $\frac{c_{\rm WDM}(M,z)}{c_{\rm CDM}(M,z)} = (1+z)^{\beta(z)} \left(1+60\frac{M_{\rm hm}}{M}\right)^{-0.17}$
- Field halos: NFW
- Subhalos: Truncated NFW (r\_t~4-10r\_s)



#### Gilman+ (1908.06983) MNRAS



# **ABC** method



# **WDM Constraint**

- 9 lenses: 0405, 0607, 0608, 0659, 1042, 1537, 1606, 2026, 2038
- $M_{\rm hm} < 10^{7.6} M_{\odot} (m_{\rm WDM} > 6.1 \text{ keV})$ (Posterior odds 10:1)
- 3.0 keV for GUT-scale produced sterile neutrino DM,
- 15 keV for miracle keV sterile neutrino DM,
- 50 keV for Dodelson-Widrow sterile neutrino DM.



#### Summary

- $M_{\rm hm} < 10^{7.6} M_{\odot}$
- $m_{\rm WDM} > 6.1 \, \rm keV$
- With substructure lensing we are directly observing DM and characterizing the population of completely dark DM halos, with no corresponding galaxies
- Expect even stronger probes from the full sample
- Test additional DM models using this formalism (SIDM, FDM)
- Joint constraints using additional tracers (arcs, satellites, streams, Lya)



### **Ex JWST data: J1537-3010**

Observe quad lenses in 4 JWST MIRI filters f560w, f1280w, f1800w, f2550w

Keeley et al 2024 1



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# Ex Fitting Results: J1537-3010

Model data with point source, lensed light from QSO host (arc), deflector light



Model Point Sources











#### Keeley et al 2024

# **Ex SED Results**

Model the fluxes of the point sources with an SED model: -continuum -hot dust

-warm dust

Plot: Samples from posterior; temperature and flux of each component are varied



# Fitting JWST data

Observe quad lenses in 4 JWST MIRI filters f560w, f1280w, f1800w, f2550w

Use warm dust fluxes in each image for DM inference



Model the fluxes of the quasar with an SED model (continuum, hot dust, warm dust)

Cores 



Kaplinghat, Keeley et al 2024



- Cores
- Core-collapse



Gad-Nasr et al 2023



- Cores
- Core-collapse
- Core-collapse triggered by stripping



Gad-Nasr et al 2023



- Cores
- Core-collapse
- Core-collapse triggered by stripping
- Anti-correlation between concentration and pericenter



 $r_p$  (percicenter) [kpc] Kaplinghat et al 2019



- Cores
- Core-collapse
- Core-collapse triggered by stripping
- Anti-correlation between concentration and pericenter
- Suppression
- Dark acoustic oscillations



Munoz et al 2020

#### MixDM Forecast constraints

- Mock data generated from a MixDM model
- Posterior of parameters  $\Sigma_{\rm sub}, M_{\rm hm}, f$
- $f = 0, M_{\rm hm} = 10^{10}, \Sigma_{\rm sub} = 0$ corner with least structure (warmest)
- Easier to differentiate between MixDM and CDM than MixDM and WDM

