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Measuring dark matter substructure with galaxy–galaxy flexion statistics

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ABSTRACT

It is of great interest to measure the properties of substructures in dark matter haloes at galactic and cluster scales. Here we suggest a method to constrain substructure properties using the variance of weak gravitational flexion in a galaxy–galaxy lensing context; this is a statistical method, requiring many foreground–background pairs of galaxies. We show the effectiveness of flexion variance in measuring substructures in N -body simulations of dark matter haloes, and present the expected galaxy–galaxy lensing signals. We show the insensitivity of the method to the overall galaxy halo mass, and predict the method’s signal-to-noise ratio for a space-based all-sky survey, showing that the presence of substructure down to $10^9 M_\odot$ haloes can be reliably detected.

Key words: cosmology: observations – dark matter – gravitational lensing: weak.

1 INTRODUCTION

The amount of substructure in dark matter haloes on galactic and cluster scales is a question of considerable interest. Lambda cold dark matter (Λ CDM) cosmological simulations predict thousands of dark satellites within the virial radius of the Milky Way, yet to date only ~ 20 have been observed (e.g. Kauffmann, White & Guiderdoni 1993; Klypin et al. 1999; Diemand, Kuhlen & Madau 2007). Similar results have been obtained for our nearest neighbour Andromeda (McConnachie & Irwin 2006) and in galaxy groups (D’Onghia et al. 2007). Averaging over galaxies observed in the field, there appears to be a suppression in the expected number below a baryonic mass of $\sim 10^{10} M_\odot$ (e.g. Read & Trentham 2005). This dearth of low-mass galaxies could be telling us that galaxy formation becomes increasingly inefficient below the peak of the luminosity function (e.g. Dekel & Silk 1986; Bullock, Kravtsov & Weinberg 2000; Benson et al. 2002; Kravtsov, Gnedin & Klypin 2004; Read, Pontzen & Viel 2006). Alternatively, it could be telling us something about the nature of dark matter (e.g. Avila-Reese et al. 2001; Barkana, Haiman & Ostriker 2001; Bode, Ostriker & Turok 2001; Knebe et al. 2008), or about the details of inflation (Zentner & Bullock 2003).

Gravitational lensing is a powerful tool for probing substructure within galaxies and clusters. Unlike photometric galaxy surveys, lensing directly probes the dark Universe; in principle, even if completely dark subhaloes exist, these will have some influence on the observed gravitational lensing signal and may therefore be detected. Several lensing techniques for constraining substructure already ex-

ist: one can examine the strong lensing of quasars by galaxies and clusters (e.g. Mao & Schneider 1998; Metcalf & Madau 2001; Metcalf & Zhao 2002; Keeton, Gaudi & Petters 2003; Moustakas & Metcalf 2003; Kochanek & Dalal 2004; Metcalf et al. 2004; Metcalf 2005; Aazami & Natarajan 2006; Amara et al. 2006; Miranda & Macciò 2007; Shin & Evans 2008) including the time-delay phenomenon (Keeton & Moustakas 2009). One can also constrain the substructure using weak galaxy–galaxy shear (Natarajan et al. 2004). Here we examine the usefulness for measuring substructure of another weak lensing phenomenon: flexion, building on several earlier studies (cf. Goldberg & Natarajan 2002; Goldberg & Bacon 2005; Bacon et al. 2006; Irwin & Shmakova 2006; Schneider & Er 2008; Leonard, King & Wilkins 2009).

It has been noted by these authors that flexion responds to small-scale variations in the gravitational potential. In a galaxy–galaxy lensing context, the mean flexion in annuli around a halo will fall off rapidly, as the mean halo density gradient is small away from the central region. However, if substructure is present, the flexion variance in annuli may not fall off so quickly, as the substructures will lead to potential fluctuations which will cause a flexion varying rapidly from place to place. It is this idea that this paper will examine.

The topic is of particular interest in the context of forthcoming and planned large lensing surveys. Ground-based surveys such as Pan-STARRS¹ and the Dark Energy Survey² will obtain many thousands of square degrees of lensing-quality data, allowing very precise galaxy–galaxy lensing constraints. Beyond this, space-based

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¹ <http://pan-starrs.ifa.hawaii.edu>

² <https://www.darkenergysurvey.org>

survey telescopes such as Euclid³ will obtain extraordinary accuracy for galaxy–galaxy lensing due to a high number density of galaxies and a survey area of 20 000 deg². This paper will make some initial predictions for the level of accuracy of flexion variance measurements with the latter survey.

The paper is organized as follows. In Section 2, we consider the relevant theory of weak gravitational lensing, concentrating on the flexion formalism. We proceed to define the galaxy–galaxy flexion statistics that we require, and show how they can be estimated theoretically. In Section 3 we describe N -body simulations of haloes with and without substructure, and explain the procedure for creating lensing maps from these simulations. In Section 4 we use the maps to illustrate the sensitivity of flexion to substructure, then calculate the expected galaxy–galaxy flexion variance signal together with its signal-to-noise ratio (S/N) for a Euclid-like survey. We examine the effect of varying the dominant halo mass, finding little impact, and show the much more pronounced effect of varying the amount of substructure present. This latter finding will illustrate the usefulness of the technique as a probe of dark matter temperature. We present our conclusions in Section 5.

2 THEORY

2.1 Flexion

The flexion formalism described here is more fully developed in Bacon et al. (2006). We begin by noting that lensing in the weak regime can be described by mapping the surface brightness of a galaxy in the source plane, $f_s(\beta_i)$, to the surface brightness in the image plane, $f_i(\theta_i)$:

$$f_i(\theta_i) = f_s(\beta_i) = f_s \left(A_{ij}\theta_j + \frac{1}{2}D_{ijk}\theta_j\theta_k \right). \quad (1)$$

Here we have introduced several quantities: first, the Jacobian matrix $\mathbf{A}$, which is taken to be constant across the galaxy image in the weak regime. It can be written in terms of lensing quantities

$$\mathbf{A} = \begin{pmatrix} 1 - \kappa & 0 \\ 0 & 1 - \kappa \end{pmatrix} + \begin{pmatrix} -\gamma_1 & -\gamma_2 \\ -\gamma_2 & \gamma_1 \end{pmatrix}, \quad (2)$$

where κ is the convergence; this maps a unit circle in the image plane to a circle with radius $1 - \kappa$ in the source plane. In the case of an isolated lens, the convergence is proportional to the projected density of matter in the lens (cf. Bartelmann & Schneider 2001),

$$\kappa(\boldsymbol{\theta}) = \frac{\Sigma(D_1\boldsymbol{\theta})}{\Sigma_{\text{cr}}}, \quad (3)$$

where Σ is the 2D projection of the density ρ ,

$$\Sigma(\boldsymbol{\xi}) = \int dr_3 \rho(\boldsymbol{\xi}, r_3) \quad (4)$$

where the integration is over the radial distance r_3 and Σ_{cr} is given by

$$\Sigma_{\text{cr}} = \frac{c^2}{4\pi G} \frac{D_s}{D_1 D_{\text{ls}}}, \quad (5)$$

where D_1 , D_s and D_{ls} are the angular diameter distances from the observer to the lens, from the observer to the source and from the lens to the source, respectively. If we also define the lensing potential ψ ,

which is proportional to the projection of the gravitational potential Φ ,

$$\psi(\boldsymbol{\theta}) = \frac{4\pi G}{c^2} \frac{D_1 D_s}{D_{\text{ls}}} \int dr_3 \Phi(D_1\boldsymbol{\theta}, r_3); \quad (6)$$

then we can also write the convergence as

$$\kappa = \frac{1}{2} (\partial_1^2 + \partial_2^2) \psi. \quad (7)$$

The other term in $\mathbf{A}$ is the shear γ_i ; this maps a circle in the image plane to an ellipse in the source plane. Its components can be written as

$$\gamma_1 = \frac{1}{2} (\partial_1^2 - \partial_2^2) \psi, \quad \gamma_2 = \partial_1 \partial_2 \psi. \quad (8)$$

The next term in equation (1) is the $\mathbf{D}$ -tensor; this contains the lensing information at the next order of approximation, and corresponds to the varying of convergence and shear across an object. As shown in Bacon et al. (2006), the $\mathbf{D}$ -tensor can be written in terms of the flexions:

$$\begin{aligned} -2D_{ij1} &= \begin{pmatrix} 3F_1 & F_2 \\ F_2 & F_1 \end{pmatrix} + \begin{pmatrix} G_1 & G_2 \\ G_2 & -G_1 \end{pmatrix}, \\ -2D_{ij2} &= \begin{pmatrix} F_2 & F_1 \\ F_1 & 3F_2 \end{pmatrix} + \begin{pmatrix} G_2 & -G_1 \\ -G_1 & -G_2 \end{pmatrix}, \end{aligned} \quad (9)$$

where G_i are the components of 3-flexion, describing the degree to which an object resembles a trefoil and F_i are the components of 1-flexion, describing the skewed shape of an object. We will only consider the 1-flexion for the purposes of this paper, as 3-flexion is considerably noisier. 1-flexion has the property of being the gradient of the convergence:

$$\begin{aligned} F_1 &= \frac{1}{2} (\partial_1^3 + \partial_1 \partial_2^2) \psi = \partial_1 \kappa, \\ F_2 &= \frac{1}{2} (\partial_1^2 \partial_2 + \partial_2^3) \psi = \partial_2 \kappa. \end{aligned} \quad (10)$$

It is this property that is so important for the technique of this paper; 1-flexion will respond wherever the density is varying rapidly from place to place, which is the case with substructure.

2.2 Galaxy–galaxy flexion

We can now introduce the flexion variance in annuli as a substructure probe. It is usual in galaxy–galaxy lensing (e.g. Kleinheinrich et al. 2006; Mandelbaum et al. 2006) to choose particular galaxy samples which act as lens and source sets. The sets may have some members in common, depending on the technique used for lens–source correlation, but at any rate the selections can be achieved either via photometric redshifts, spectroscopic redshifts or a combination of the two.

Let us orientate ourselves with the most familiar case of galaxy–galaxy lensing, which involves shears. The angular separation θ of foreground object f and background object b is measured, and the shear of b is decomposed into components tangential and diagonal to the line connecting f and b ,

$$\gamma_t = -\Re(\gamma e^{-2i\phi}), \quad \gamma_B = -\Im(\gamma e^{-2i\phi}), \quad (11)$$

where ϕ is the position angle of the line, $\gamma = \gamma_1 + i\gamma_2$, and $\Re$ and $\Im$ denote real and imaginary parts, respectively. For a circular foreground lens, γ_t will be activated by gravitational lensing, while γ_B will not; if it is present, it is due to systematic effects.

Similarly a background flexion can be decomposed, but whereas shear is decomposed into tangential and diagonal components, flexion is decomposed into radial and tangential components due to the

³ <http://www.ias.u-psud.fr/imEuclid>

fact that it has different rotational properties to shear:

$$F_r = -\Re(Fe^{-i\phi}), \quad F_B = -\Im(Fe^{-i\phi}), \quad (12)$$

where $F = F_1 + iF_2$. Here the r component is activated by gravity for a circular lens, while the B component is unactivated unless systematics are present.

The behaviour of flexion at an angular distance θ from a singular isothermal sphere (SIS) was considered by Bacon et al. (2006). They show that while the surface density of the SIS is proportional to θ^{-1} , the 1-flexion and 3-flexion drop as θ^{-2} . While real galaxies are not truly SIS, they are close enough to this profile (for instance, cf. Treu et al. 2009) that if we take foreground–background pairs, measure F_r and consider the mean F_r in annuli, we will see a similarly rapid drop in galaxy–galaxy flexion with annulus radius θ .

This will remain true if there are substructures at angular distance θ from the centre of foreground galaxies. Although in the extreme locality there will be a larger flexion than usual, the mean signal averaged around the galaxy will still drop rapidly with θ .

However, this will not be true for the flexion *variance* in the annulus. This will respond to any density fluctuations within the annulus. Therefore, in any annuli with non-negligible substructure, even if the mean flexion is small (as the mean gradient of density is small), the flexion variance will remain comparatively large. It is this behaviour that we will use to constrain substructure on galactic scales.

At this point one may ask whether flexion variance is the best tool for our task; wouldn't flexion correlation functions in annuli provide more information? The question could be informed by experience in cosmic shear studies, where shear correlation functions provide more finesse than shear variance in cells.

However, in this case a correlation function does not seem to be helpful. Since the correlation function in question would be in annuli around a foreground galaxy, it constitutes a form of galaxy–galaxy–galaxy lensing (Schneider & Watts 2005); it is a three-point statistic. In order to use this to measure substructures, it is necessary for two background sources to be close to the same substructure as well as to the foreground galaxy; this rarely happens, leading to low S/N. On the other hand, the flexion variance in annuli only involves two points, a foreground–background pair, with much greater S/N as we shall see.

3 SIMULATIONS

In this section we describe N -body simulations which we will use to demonstrate the utility of flexion variance as a probe of substructure.

3.1 3D density

We use the cosmological Λ CDM simulation already presented in Diemand, Madau & Moore (2005). The simulation was run using PKDGRAV (Stadel 2001), with cosmological parameters $(\Omega_m, \Omega_\Lambda, \sigma_8, h) = (0.268, 0.732, 0.7, 0.71)$, and a box of size $L_{\text{box}} = 90$ Mpc, with 300^3 particles. The initial conditions were generated with GRAFIC-2 (Bertschinger 2001). From the simulation volume, we extracted four Milky Way sized haloes at a mass resolution of $m_p = 5.7 \times 10^5 M_\odot$; their virial masses are $[2.1, 1.5, 1.2, 1.3] \times 10^{12} M_\odot$. While we are therefore very limited in our number of lenses (to three projections of each of four high-resolution galaxies), we will find that this is sufficient to give the initial indicative results required by this paper.

As in Read et al. (2008), the subhaloes inside each ‘Milky Way’ at redshift $z = 0$ were identified using the AHF algorithm (Gill,

Figure 1. N -body simulation of one of our Milky Way mass haloes, with substructure (top panel) and without substructure (bottom panel), with contours showing projected logarithmic density over the range $[0.02, 672] M_\odot \text{pc}^{-2}$.

Knebe & Gibson 2004). We use subhaloes with >500 particles in order to minimize spurious numerical disruption near the Galactic Centre. In some cases we will remove substructure; this is achieved by subtracting all particles not assigned to the main halo. An example halo, with and without substructure, is shown in Fig. 1.

3.2 2D convergence

The 3D numerical simulations discussed above represent the density field using discrete particles. We transform these into convergence maps by projecting the particles along particular spatial directions and placing the particles on to a 2D 1024×1024 grid, which we carry out using the IDL cloud-in-cell routine available as part of The IDL Astronomy User's Library.⁴ We produce three projection maps for each 3D halo (by projecting along the x -, y - or z -axis).

⁴ <http://idlastro.gsfc.nasa.gov>

Figure 2. Standard deviation of 1-flexion in annuli for one galaxy halo, in units of arcsec^{-1} . Arrows show the positions of subhaloes, with arrow length proportional to subhalo mass.

We investigated a number of techniques for filtering the mass maps. This is important because the finite number of simulation particles introduces shot noise into the 2D maps; this can compete with the substructure signal we are investigating. Here we show the results obtained using multiscale entropy filtering (MEF; Starck, Pires & Réfrégier 2006). This provides superior performance to a simple Gaussian filter, as it reduces the shot noise while preserving the density fluctuations, as we shall demonstrate. For this purpose we use the routines supplied in the public code Multi-Resolution methods for gravitational LENSing (MRLENS),⁵ decomposing the image into eight wavelet scales, and removing the first of these (which corresponds to the pixel scale). The 2D mass maps (Σ) are then rescaled using the critical density (Σ_c) to give the convergence (κ). We will use a lens–source configuration of $D_s = 1640$ Mpc, $D_l = 740$ Mpc and $D_{ls} = 1170$ Mpc corresponding to our choice of median redshifts for foreground ($z = 0.22$) and background ($z = 0.92$) sources appropriate for our projected Euclid survey.

4 ANALYSIS

Armed with this set of simulations, we are in a position to examine whether substructure can be reliably measured using galaxy–galaxy flexion variance.

4.1 Flexion variance sensitivity to substructure

Using the convergence maps for each halo, we calculate the related 1-flexion map using $F_i = \partial_i \kappa$, and smooth the resulting flexion map with a 0.4 arcsec diameter top-hat filter to remove small-scale flexion peaks inaccessible to galaxy shape measurements. We then calculate the mean and standard deviation of F in annuli with width 0.9 arcsec, centred on the mode of the galaxy’s κ distribution.

As an initial example, Fig. 2 shows the standard deviation σ_F for the radial 1-flexion in annuli, for one example halo. Also displayed are arrows showing the radial positions of subhaloes found in the simulation, with arrow size proportional to the mass of the subhalo. We can see that there is a rather close correspondence between the flexion variance and the subhalo positions and masses. This

encourages us to examine what the signal will be for an ensemble of galaxy haloes in a galaxy–galaxy lensing context.

4.2 Galaxy–galaxy flexion signal

In this section we will present an estimate of the galaxy–galaxy flexion signal; but in order to begin with a familiar quantity, we first examine the convergence for our haloes in annuli of width 0.9 arcsec, centred on the mode of the main haloes’ κ distribution. Fig. 3 shows the mean convergence $\bar{\kappa}$ and standard deviation σ_κ for an ensemble of background galaxies behind the 12 stacked haloes, in annuli with radius θ . Note that σ_κ is smaller than $\bar{\kappa}$ for $\theta < 50$ arcsec; since κ is proportional to the projected surface density, this reflects the fact that the substructure fluctuations are fairly small in amplitude in relation to the mean density. Hence the difficulty in using convergence or shear to measure substructures; the S/N on σ_κ will be smaller than that on $\bar{\kappa}$.

The upper and lower lines for $\bar{\kappa}$ and σ_κ can be compared to see the effect of including or omitting substructure; the mean of the convergence is a little higher on scales $\gtrsim 30$ arcsec if we include substructure, and the standard deviation of the convergence is several times larger. This is to be expected; the presence of substructure changes the profile a little, and significantly alters the spatial variation of the matter distribution.

We now consider the 1-flexion galaxy–galaxy signals, again calculated in annuli with width 0.9 arcsec, centred on the mode of the galaxy’s κ distribution, and smoothed with a 0.4-arcsec diameter top hat. Fig. 4 shows the mean $\bar{F}$ and standard deviation σ_F for the radial 1-flexion in annuli, for an ensemble of background galaxies behind the haloes. We show results with either MEF filtering or no filtering, and with or without substructure, as it is important to understand the impact of our filter on the results. We should emphasize that this filter is applied to the N -body simulations to increase their realism; it is not a filter applied at the measurement stage.

We immediately see that the filtering of the N -body simulation is valuable; in the cases where no filtering is applied, the flexion standard deviation is substantial even in the absence of substructure. This is due to the particle noise in the simulations that leads to spurious high-frequency convergence gradients. However, with the

Figure 3. Mean (dotted lines) and standard deviation (solid lines) of convergence κ in annuli for the 12 halo orientations combined. Upper lines in each case show results with substructure, while lower lines show the results when substructure is removed.

⁵ http://irfu.cea.fr/Phocce/Vie_des_labos/Ast/ast_visu.php?id_ast=878

Figure 4. Mean (dotted lines) and standard deviation (solid lines) of 1-flexion in annuli for the 12 halo orientations combined, in units of arcsec^{-1} . Upper lines in each case show results with substructure, while lower lines show results when substructure is removed. Top panel: MEF results; bottom panel: unfiltered results.

MEF filter in the absence of substructure, the flexion standard deviation is at least an order of magnitude smaller than the unfiltered standard deviation.

Importantly, we see that when substructure is included (with MEF filtering of the simulations), σ_F dominates over $\bar{F}$ on scales greater than about 10 arcsec, and is two orders of magnitude larger than the no-substructure signal. The former observation confirms our claim that flexion variance is of interest, as it has much larger S/N than standard galaxy–galaxy flexion. The latter observation shows that it is substructure which is activating this signal; this is important, as the overall ellipticity of the central halo might have given a flexion variance (cf. Hawken & Bridle 2009) but our result demonstrates that this is negligible for our statistic. To measure halo ellipticity with flexion, one should instead follow the methods of Hawken & Bridle (2009).

The flexion variance also dies off more slowly with θ than $\bar{F}$ does. This is presumably due to the different phenomena being probed: $\bar{F}$ is probing the mean gradient of the overall galaxy halo density; while σ_F is dominated by the gradients of substructure haloes within

Figure 5. Ratio of σ_F^2 with modified central halo amplitude to the usual σ_F^2 . Solid: no central halo; dashed: total mass standard deviation of $4 \times 10^{12} M_\odot$; dotted: total mass standard deviation of $10^{13} M_\odot$.

the annuli, which remain similar regardless of which annulus they are found in.

4.3 Insensitivity to varying dominant halo mass

An issue that we have not yet dealt with is the question of whether combining galaxies with various masses contributes a dominant unwanted signal to our galaxy–galaxy flexion signal. It is common in galaxy–galaxy lensing to divide up or scale one’s foreground lens set according to a mass proxy such as luminosity (e.g. Kleinheinrich et al. 2006) or stellar mass (e.g. Mandelbaum et al. 2006). Nevertheless, within such a bin in mass, will the flexion variance signal be dominated by the overall mass variance rather than the substructure signal?

In order to test this, we measured the mean flexion signal for each of the main haloes. We then subtracted this signal from each halo’s flexion map, leaving the flexion due to substructure alone. We then optionally re-added a random proportion of the mean flexion signal, leading to a total mass standard deviation of 4×10^{12} or $10^{13} M_\odot$, and remeasured the ensemble flexion variance signal for our 12 realizations. The effects on our signal are shown in Fig. 5.

First we note that the difference between the shear variance with and without central haloes is very small; most of the signal in our original ensemble is therefore coming from substructure. This is borne out by the case where we have a halo mass variance of $4 \times 10^{12} M_\odot$. The ratio of this signal to our original is within a few per cent of unity beyond $\theta \gtrsim 20$ arcsec. In the extreme case where the mass variance is $10^{13} M_\odot$, we obtain about 10 per cent of the flexion variance from the main haloes for $\theta > 30$ arcsec, but this case corresponds to hardly having any mass binning at all.

We conclude, then, that provided reasonable steps are taken to deal with the galaxy–galaxy lensing in mass or luminosity bins, flexion variance is totally dominated by the substructure signal.

4.4 Signal-to-noise ratio for galaxy–galaxy flexion

The question now arises whether flexion variance is measurable with sufficient accuracy on the relevant scales to constrain substructure. The difficulty is that source galaxies are measured with a substantial intrinsic flexion (cf. Goldberg & Bacon 2005); the distribution

Figure 6. Flexion variance estimator (solid line) in units of arcsec^{-2} , and expected noise level in each annulus (dashed line) for the Euclid survey described in the text.

has a strong central peak and wide wings, such that the measured distribution has a 68 per cent range $\sigma_F \simeq 0.1 \text{ arcsec}^{-1}$. Clearly this intrinsic variance dominates over the substructure variance at all scales. However, following a statistical approach from Bacon, Refregier & Ellis (2000), one can estimate the flexion variance due to substructure, σ_{sub}^2 , as an excess variance within an annulus, over and above that due to shape noise $\sigma_{\text{intrinsic}}^2$:

$$\sigma_{\text{sub}}^2 \simeq \sigma_{\text{annulus}}^2 - \sigma_{\text{intrinsic}}^2, \quad (13)$$

where $\sigma_{\text{annulus}}^2$ is the measured total variance within an annulus, while $\sigma_{\text{intrinsic}}^2$ is the variance measured for galaxies in an aperture with the largest radius available. Since this equation is only exact for Gaussian distributions, we have checked that it is approximately true for our simulation flexion distribution and realistic shape noise distribution; we find that σ_{sub}^2 is correctly estimated to within 5 per cent. As described by Bacon et al. (2000), the error on this estimator is approximately

$$\sigma[\sigma_{\text{sub}}^2] \simeq \sigma_{\text{intrinsic}}^2 / \sqrt{N}, \quad (14)$$

where N is the number of objects in the annulus.

We can now use this equation to estimate how strong a signal we expect for a Euclid-like survey. We use $\sigma_{F\text{int}} = 0.1$, and calculate N using the standard survey parameters of Amara & Réfrégier (2007). We examine the case of choosing $1.3 \text{ foreground galaxies arcmin}^{-2}$ with $z_{\text{med}} = 0.22$, $34 \text{ background galaxies arcmin}^{-2}$ with $z_{\text{med}} = 0.92$ and a survey area of $20\,000 \text{ deg}^2$. Note that this foreground number density only permits us to measure the statistic out to $\theta = 30 \text{ arcsec}$; beyond this we will typically be in the region inhabited by another foreground galaxy.

Fig. 6 shows the expected value of our estimator $\sigma_{F\text{sub}}^2$ (solid line) and the noise level on this estimator in each annulus (dashed line). Note that for a large range in θ , the S/N is considerable: $S/N \simeq 1\text{--}5$ for each of 11 annuli between 10 and 30 arcsec. This suggests that substructure can be studied in detail using this method.

4.5 Sensitivity to substructure content

To pursue this point, we provide some examples of the degree to which we can distinguish between different substructure scenarios.

First, we examine the impact of removing half of the subhaloes at random in each of our simulations. We carry out the analysis of Sections 4.2 and 4.4 for these modified simulations, and show how their flexion variance signal differs from the usual case in Fig. 7. For the figure we have used 50 bins between 0 and 90 arcsec; however, we only use bins between 10 and 30 arcsec for our discrimination analysis below.

We note that the flexion variance is itself approximately halved in this case. A χ^2 measure for the significance of this difference is the sum of $(\sigma_F^2 - \sigma_{F\text{half}}^2)^2 / \sigma[\sigma_{\text{sub}}^2]^2$ for usable annuli, and is in this case $\simeq 110$, confirming that such a substructure configuration would be strongly distinguished.

However, that scenario is not expected physically. More plausibly, if dark matter has a non-negligible temperature, this will preferentially remove substructures up to a certain mass threshold. To examine whether this type of phenomenon would be detectable by flexion variance, we remove all haloes with mass $< 5 \times 10^8 M_\odot$ or $< 10^9 M_\odot$ from our simulations. We again carry out the analysis of Sections 4.2 and 4.4 for the modified simulations. The results of this process are shown in Fig. 8.

We see that there is only a small difference between the signals for all subhaloes and for all subhaloes $> 5 \times 10^8 M_\odot$; they are only slightly distinguishable within the error expected as shown by grey shading, with $\Delta\chi^2 = 3.5$ using annuli between 10 and 30 arcsec. Care should be taken in drawing conclusions from this, as haloes with mass $5 \times 10^8 M_\odot$ are still near our lower mass threshold; we will therefore explore lower mass substructure in future work.

On the other hand, the difference between the $> 10^9 M_\odot$ case and the full substructure case is easily detected, with $\Delta\chi^2 = 22$.

This is very significant, as it is expected that in a warm dark matter (WDM) scenario, free streaming of the warm particles will suppress small haloes. As shown by Barkana et al. (2001, equations 4 and 5), a $\simeq 1 \text{ keV}$ mass scale for WDM leads to suppression of haloes of mass $< 10^9 M_\odot$. We have shown that this suppression is detectable with flexion variance, which can therefore be used as a constraint on the mass of dark matter in the WDM scenario.

Figure 7. Ratio of flexion variances with substructure and without 50 per cent of the subhaloes. Grey shading shows region within 1σ uncertainty of flexion variance with substructure.

Figure 8. Ratio between (a) flexion variances with all subhaloes, and with all subhaloes $> 5 \times 10^8 M_\odot$ (dashed line); (b) flexion variances with all subhaloes, and with all subhaloes $> 10^9 M_\odot$ (solid line). Grey shading shows region within 1σ uncertainty of flexion variance with all subhaloes.

5 CONCLUSIONS

In this paper we have explored the utility of galaxy–galaxy flexion variance for the purpose of measuring the degree of substructure on galactic scales.

We have described the relevant statistics, showing how flexion radial and systematic modes are constructed in a galaxy–galaxy lensing context, and introducing the concept of flexion variance in annuli. We have explained how this is a more suitable probe than galaxy–galaxy–galaxy flexion (or flexion correlation functions in annuli), as the latter has a very rapid drop in signal as a function of angle.

We have gone on to test the use of these statistics by adopting a set of N -body simulations of galaxy haloes, including substructure, or removing this substructure by means of halo-finding algorithms. We have then calculated the lensing convergence associated with these haloes, applying the suitable MEF filter to reduce the effect of particle shot noise.

We have shown that the flexion variance is able to detect substructures on a halo-by-halo basis. This carries through to the full galaxy–galaxy flexion variance expected with an ensemble of haloes; the signal found is substantially larger than that for the flexion mean, or indeed the flexion variance in the absence of substructures. The underlying central halo mass variance is not found to be a dominant source of noise for this signal.

We have made predictions for the level of the flexion variance S/N for a space-based survey such as Euclid, finding that substructure amplitudes will be measured with significant precision. This allows us to discriminate between substructure scenarios with different numbers and masses of haloes, and will enable a constraint on dark matter particle mass.

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