Schwarzschild modelling of barred s0 galaxy NGC 4371

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ABSTRACT

We apply the barred Schwarzschild method developed by Tahmasebzadeh et al. (2022) to a barred S0 galaxy, NGC 4371, observed by IFU instruments from the TIMER and ATLAS3D projects. We construct the gravitational potential by combining a fixed black hole mass, a spherical dark matter halo, and stellar mass distribution deprojected from 3.6 µm S⁴G image considering an axisymmetric disc and a triaxial bar. We independently modelled kinematic data from TIMER and ATLAS3D. Both models fit the data remarkably well. We find a consistent bar pattern speed from the two sets of models with $\Omega_p = 23.6 \pm 2.8 \text{ km s}^{-1} \text{ kpc}^{-1}$ and $\Omega_p = 22.4 \pm 3.5 \text{ km s}^{-1} \text{ kpc}^{-1}$, respectively. The dimensionless bar rotation parameter is determined to be $\mathcal{R} \equiv R_{cor}/R_{bar} = 1.88 \pm 0.37$, indicating a likely slow bar in NGC 4371. Additionally, our model predicts a high amount of dark matter within the bar region ($M_{DM}/M_{total} \sim 0.51 \pm 0.06$), which, aligned with the predictions of cosmological simulations, indicates that fast bars are generally found in baryon-dominated discs. Based on the best-fitting model, we further decompose the galaxy into multiple 3D orbital structures, including a BP/X bar, a classical bulge, a nuclear disc, and a main disc. The BP/X bar is not perfectly included in the input 3D density model, but BP/X-supporting orbits are picked through the fitting to the kinematic data. This is the first time a real barred galaxy has been modelled utilizing the Schwarzschild method including a 3D bar.

Key words: galaxies: bar - galaxies: bulges - galaxies: kinematics and dynamics - galaxies: structure.

1 INTRODUCTION

About two-thirds of disc galaxies in the local universe host bars (Eskridge et al. 2000; Erwin 2018). These bars are identified either by non-axisymmetric features in their surface density or through kinematic signatures, such as a positive correlation between the mean velocity and the third Gauss–Hermite moment h_3 (Bureau & Athanassoula 2005; Li et al. 2018).

Bars can play a significant role in the formation of galaxies by redistributing the energy and the angular momentum of the disc materials (Debattista & Sellwood 1998; Athanassoula 2003; Kormendy & Kennicutt 2004; Gadotti 2011). Numerous observations and simulations have demonstrated that bars are often associated with a boxy/peanut or X-shaped structure (hereafter, BP/X) when viewed edge-on (Combes & Sanders 1981; Raha et al. 1991; Lütticke, Dettmar & Pohlen 2000).

The key parameters characterizing a bar include its radius, strength, and pattern speed (e.g. Aguerri et al. 2015). The radius and strength of a bar are typically derived from optical or near-infrared images (Aguerri, Beckman & Prieto 1998; Buta & Block 2001). In contrast, the bar pattern speed is a dynamical parameter that is more challenging to measure, requiring kinematic data. Tremaine & Weinberg (1984) introduced a simple and model-independent method (TW method) for measuring the bar pattern speed Ω_p , which is widely used. The TW method uses the profiles of surface brightness $\Sigma(x)$ and line-of-sight (LOS) velocity V_{los} measured along the slits crossing the bar and parallel to the disc major axis, with the coordinate of x integrated from $-\infty$ to ∞ along a slit.

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In recent decades, integral field unit (IFU) surveys such as SAURON (Bacon et al. 2010), CALIFA (Sánchez et al. 2012), SAMI (Croom et al. 2012), and MaNGA (Bundy et al. 2015), have provided kinematic maps of thousands of nearby galaxies. IFU data improve the accuracy of the pattern speed measurement. The TW method has been applied to sub-samples of barred galaxies from CALIFA (Aguerri et al. 2015; Cuomo et al. 2019b), MaNGA (Guo et al. 2019; Garma-Oehmichen et al. 2022; Géron et al. 2023), and MUSE observed galaxies (Cuomo et al. 2019a, 2022; Buttitta et al. 2022). The accuracy of the TW method depends on accurately determining the disc position angle. Inaccuracies of a few degrees in the disc position angle can lead to errors of 10 per cent (and up to 100 per cent) for $\Omega_{\rm p}$ (Debattista 2003; Zou et al. 2019). There is still large uncertainty in the pattern speed measurements obtained from MaNGA-like data due to their low-spatial resolution. MUSE instrument (Bacon et al. 2010) offers IFU data with higher spatial resolution and signal-to-noise ratio (S/N). However, accurately measuring Ω_p requires kinematic data covering both the bar and the disc's outer regions. Such observations with MUSE are expensive (Cuomo et al. 2019a, 2022; Buttitta et al. 2022).

Recently, the TIMER project (Gadotti et al. 2020) observed a sample of 21 nearby barred galaxies using the MUSE instrument. These observations uncovered various structures co-existing with bars in the galaxy centres, such as classical bulges, nuclear discs, and ring-like structures. To thoroughly understand the formation of these structures, it is crucial to decompose them both morphologically and kinematically to quantify their contributions. Additionally, the bar pattern speed for the TIMER galaxies cannot be determined using the TW method, as the kinematic data are limited only to the bar region.

Dynamical modelling is a powerful method that can constrain the bar pattern speed using full kinematic information and facilitate the dynamical decomposition of the bar, classical bulge, and nuclear disc structures. The bar pattern speed of the Milky Way was strongly constrained by a few dynamical models, including the Schwarzschild (1979) orbit-superposition method (Zhao 1996; Wang et al. 2012, 2013) and Made-to-Measure method (Long et al. 2013; Portail et al. 2017). The orbit-superposition method, in particular, the van den Bosch et al. (2008) triaxial code (hereafter, VdB08) has been widely used in exploring stellar orbit distribution in a large sample of galaxies from surveys like CALIFA (Zhu et al. 2018a, b), MaNGA (Jin et al. 2020), and SAMI (Santucci et al. 2022). It has been further developed to include the stellar age and metallicity (Zhu et al. 2020; Poci et al. 2019), enabling chemodynamical decomposition of galaxy structures (Zhu et al. 2022; Ding et al. 2023; Jin et al. 2024). A new version of the VdB08 code, named DYNAMITE, has been publicly released, featuring ongoing enhancements (Jethwa et al. 2020; Thater et al. 2022).

However, the dynamical modelling of external barred galaxies is challenging due to their complicated morphological and kinematic properties. N-body simulations are used as input of the 3D density distribution in the previous barred models, like the triaxial bulge/bar/disc M2M model for M31 (Blaña Díaz et al. 2018), and Schwarzschild FORSTAND code (Vasiliev & Valluri 2020). Recently, Dattathri et al. (2024) introduced a new method that employs a parametric 3D density distribution to deproject edge-on barred galaxies with BP/X-shaped structures. This approach was validated using dynamical modelling against mock data with the FORSTAND code.

In Tahmasebzadeh et al. (2021), we presented a deprojection method to estimate the 3D density distribution of barred galaxies across various observational orientations, incorporating both an axisymmetric disc and a triaxial (predominantly prolate) bar.

Subsequently, we utilized these 3D density distributions as input and modified the VdB08 code to explicitly include the bar (Tahmasebzadeh et al. 2022). Testing our methodology with a set of mock data in various orientations has demonstrated its proficiency in accurately recovering key properties of barred galaxies, particularly the bar pattern speed and the BP/X structure.

In this study, we apply our bar modelling approach to NGC 4371, a particularly intriguing barred galaxy observed by the TIMER and ATLAS3D projects. This marks the first application of the Schwarzschild method to a real barred galaxy, with the bar explicitly included in the model. NGC 4371 is notable for its complex inner structures, including a nuclear disc and a bar. There is ongoing debate regarding the presence of a classical bulge in this galaxy (Erwin et al. 2015; Gadotti et al. 2015). The stellar population across the galaxy appears to be very old, and the bar pattern speed remains undetermined due to limited data coverage. By developing an orbit-superposition model, we aim to first constrain key parameters such as the bar pattern speed, and then quantitatively determine the contributions of the classical bulge, bar, and nuclear disc.

The paper is organized as follows. In Section 2, we introduce the photometry and spectroscopy data used for modelling. In Section 3, we describe the bar modelling steps and the technical details. In Section 4, we present our results and highlight the key properties of NGC 4371 that have been measured. Section 5 provides a summary and conclusion. The appendix discusses how our new approach improves upon previous axisymmetric models applied to a large sample of spiral galaxies without including a bar.

2 DATA

2.1 General properties of NGC 4371

NGC 4371, a massive early-type galaxy with a stellar mass of $M_{\star} \sim 10^{10.5} \,\mathrm{M_{\odot}}$ (Muñoz-Mateos et al. 2015), is located near the centre of the Virgo cluster at a distance of approximately 16.9 Mpc (Blakeslee et al. 2009). The inner region of this galaxy is composed of a few different structures as seen from photometric images and kinematic maps. It is commonly accepted that NGC 4371 has a bar (Erwin & Sparke 1999; Buta et al. 2015). Morphological studies using photometry suggest that it might have a pseudobulge or a composite bulge, including a small classical merger-built bulge alongside a pseudobulge (Fisher & Drory 2010; Erwin et al. 2015). While 2D stellar kinematics do not provide direct evidence for a classical bulge, they instead strongly demonstrate the existence of a relatively large, rapidly rotating nuclear stellar disc extending to $\sim 12^{''}$, which corresponds to the barlens observed in the photometric images (Gadotti et al. 2015, 2019). It is difficult to clearly identify the presence or absence of a BP/X bulge in NGC 4371 from morphological isophotos (Erwin & Debattista 2013), although with a high probability of presenting one at its stellar mass (Erwin, Debattista & Anderson 2023). Both the photometric images and stellar kinematic maps are with information blended along the line-ofsight. Considering the complicated composition of a few structures in this galaxy, uncovering its 3D structure combining photometric and kinematic maps might be a key to further understanding its bulge properties.

2.2 Photometry

We use the 3.6 µm image of NGC 4371 taken by the Infrared Array Camera (IRAC) Channel 1 from the Spitzer Survey of Stellar Structures in Galaxies (S^4G) (Sheth et al. 2010). The pixel size



Figure 1. Integrated MUSE spectra (black line) and PPXF fits of NGC 4371 (red lines) are shown for two Voronoi bins, one from the central region (top) and one from the outer region (bottom). The spectra are plotted over the wavelength range of 4800–7300 Å, fitted to the MILES stellar templates. The residuals between the observed spectrum and the best-fitting model are displayed as a green line. Grey-shaded areas indicate regions that were masked during the fit, often due to emission lines or insufficient sky subtraction.

of the S⁴G image is 0.75" and the point spread function (PSF) FWHM is ≈ 1.8 " (Kim et al. 2014). Due to the reduced effects of dust extinction and emission at these wavelengths, the S⁴G image effectively represents the galaxy stellar structures.

2.3 Spectroscopy

MUSE has observed NGC 4371 as part of the TIMER project (Gadotti et al. 2019). MUSE covers an almost square 1 arcmin \times 1 arcmin field of view (FOV) with contiguous sampling of 0.2 arcsec \times 0.2 arcsec and the spectral coverage of 4750–9350 Å. The spectral sampling is 1.25 Å per pixel, and the total integration time is 3840 s. The stellar kinematics maps of NGC 4371 had been studied in Gadotti et al. (2015, 2020).

To achieve our desired number of stellar kinematic constraints, we re-extract the 2D maps of v, σ , h_3 , and h_4 from the MUSE data cubes using the Penalized Pixel-Fitting (PPXF) software (Cappellari & Emsellem 2004) through the GIST pipeline (Bittner et al. 2019), We spatially binned the spectra to achieve a minimum signal-to-noise (S/N) ratio of approximately 140 per spectral pixel. This was done using the Voronoi binning technique presented by Cappellari & Copin (2003). We set a minimum signal-to-noise (S/N) threshold of ~ 3 for each pixel adopted for the binning, thereby reducing contamination from outer regions with very low S/N.

The stellar templates are taken from the Medium-resolution Isaac Newton Telescope Library of Empirical Spectra (MILES) stellar library (Sánchez-Blázquez et al. 2006; Falcón-Barroso et al. 2011). We used the full sample consisting of 980 stars that span the wavelength range of 4760–7400 Å. For our analysis, we specifically fitted the galaxy spectrum within the range of 4800–7300 Å to ensure compatibility with the coverage provided by the MILES spectral library. Additionally, in the outer region bins, the noise level significantly increases beyond 7300 Å due to weaker signals combined with the higher sky background. We also employed a nonconstant Line Spread Function (LSF) to account for the wavelength dependence of the instrumental spectral resolution. We adopted the LSF derived by Bacon et al. (2017), where the FWHM varies from 2.98 Å at a wavelength of 4800 Å to 2.54 Å at a wavelength of 7300 Å. We used the instrumental dispersion of the MILES template library, reported as 2.51 Å by Beifiori et al. (2011) and Falcón-Barroso et al. (2011). Thater et al. (2019) demonstrated that the effect of the spectral resolution variation, changing from 2.5 to 2.9 Å on the extracted velocity dispersion is on average only 5 km s⁻¹, which is within the range of kinematic errors. A χ^2 minimization was used with PPXF to fit the stellar template to the spectra from each Voronoi bin. We adopted an 8th-order multiplicative polynomial for the fitting process and a 4th-order additive Legendre polynomial to account for the underlying continuum. Emission lines and regions with poor sky subtraction were masked during the fit. We then compared the fitted spectrum with the original spectrum for each bin. The standard deviation of the residuals is shown as a green line in Fig. 1 for two Voronoi bins from the central and outer regions. The extracted kinematic maps are overall consistent with the results from Gadotti et al. (2020).

NGC 4371 is also observed with SAURON (Bacon et al. 2001) as part of the volume-limited ATLAS3D project that examined stellar and gas kinematics and photometric imaging of 260 early-type galaxies (Cappellari et al. 2011). The SAURON 33 arcsec × 41 arcsec FOV was sampled by 0.94 arcsec × 0.94 arcsec square lenslets in the low resolution mode. Therefore, SAURON covers a smaller inner region of NGC 4371 than the MUSE data cube. We use the stellar kinematic data cube provided on the ATLAS3D website¹, which is extracted with the spectral coverage of 4800–5400 Å.

¹http://www-astro.physics.ox.ac.uk/atlas3d/

3 DYNAMICAL MODEL

3.1 Gravitational potential

We assume the gravitational potential includes contributions from stellar mass, dark matter (DM), and a fixed central black hole (BH) mass. As the TIMER-MUSE kinematic data do not resolve the BH sphere of influence, we cannot constrain the BH mass in our modelling. Therefore, we fixed the BH mass with $M_{\rm BH} = 10^{6.8} \, {\rm M}_{\odot}$ for NGC 4371 constrained by the SINFONI data with higher spatial resolution (Saglia et al. 2016). Including or excluding the BH with this mass does not affect our modelling outcome. However, its presence provides additional stability in orbit integration in the very central region and helps achieve a more realistic representation of the orbital structure in the very inner region. Despite this, its overall effect on the dynamical properties of the modelled galaxy is negligible. The central black hole can influence stellar orbits beyond its sphere of influence, such as reducing the population of bar-supporting resonances with the smallest pericentre by up to \sim 15 per cent depending on the BH mass. However, this effect is still confined to a very small region, possibly a few times the sphere of influence, and will not impact the overall projected kinematics in the centre as discussed in Wheeler et al. (2023).

3.1.1 Dark matter mass

For the DM distribution, we consider a spherical Navarro-Frenk-White (NFW) (Navarro, Frenk & White 1996) halo for simplicity and to reduce the number of free parameters using the concentrationmass relation. Our kinematic data have limited spatial coverage, and what the model constrained is only the enclosed mass profile within the data coverage; the halo profile beyond the data coverage does not affect the fitting to the kinematic data. With the current data, an NFW halo with a fixed mass-concentration relation has enough freedom to represent the enclosed DM profile within the data coverage. We have tested different dark-matter profiles against the mock model presented in Tahmasebzadeh et al. (2022) and found no significant differences in the modelling results. This is a widely adopted and accepted assumption in stellar dynamical modelling of external galaxies with IFU data covering the inner regions (e.g. Cappellari et al. 2013; Zhu et al. 2018a; Jin et al. 2020; Santucci et al. 2022, among others).

The enclosed mass profile in the NFW halo can be expressed as:

$$M(< r) = M_{200}g(c) \left[\ln\left(1 + cr/r_{200}\right) - \frac{cr/r_{200}}{1 + cr/r_{200}} \right],$$
(1)

where $g(c) = [\ln(1+c) - c/(1+c)]^{-1}$ and *c* represents the concentration of the DM halo. The virial mass M_{200} , is defined as $\frac{4}{3}\pi 200\rho_c r_{200}^3$ representing the mass within the virial radius r_{200} . The adopted critical density is $\rho_c = 1.37 \times 10^{-7} \,\mathrm{M_{\odot} pc^{-3}}$, so that the two remaining free parameters are the concentration *c* and the virial mass M_{200} .

Since the data do not extend to a sufficiently large radius, we cannot constrain c and M_{200} at the same time. Therefore, we fix c based on the relation from (Dutton & Macciò 2014):

$$\log_{10} c = 0.905 - 0.101 \log_{10} \left(M_{200} / \left[10^{12} h^{-1} \,\mathrm{M_{\odot}} \right] \right), \tag{2}$$

which is inferred from galaxy simulations with h = 0.671 (Planck Collaboration XVI 2014). Our dynamical model can robustly constrain the contribution of DM within the radius of the outermost kinematic aperture. This constraint is directly related to M_{200} , assuming the validity of equation 2.

Central concentrated component (Sérsic)				
Normalized flux	Σ_0	15.24		
Effective radius	R_e	1		
Sérsic index	n	1.10		
Ellipticity	ϵ	0.70		
Position angle	PA	90.0		
Nuclear disc (exponential)				
Normalized flux	Σ_0	13.47		
Scale length	R_s	5.21		
Ellipticity	ϵ	0.45		
Position angle	PA	89.0		
Bar (Sérsic)				
Normalized flux	Σ_0	13.93		
Effective radius	R_e	27.39		
Sérsic index	n	0.20		
Ellipticity	ϵ	0.49		
Position angle	PA	14		
Main disc (exponential)				
Normalized flux	Σ_0	15.24		
Scale length	R_s	43.33		
Ellipticity	ϵ	0.47		
Position angle	PA	88.0		

3.1.2 Stellar mass

We construct the contribution of stellar mass to the gravitational potential by multiplying the galaxy intrinsic 3D density distribution with a stellar mass-to-light ratio. Obtaining the intrinsic 3D luminosity density of a barred galaxy from its 2D photometry image is not straightforward. We achieve this in three steps using the method described in Tahmasebzadeh et al. (2021): (1) We first decompose the galaxy 2D image into a disc and a bar, (2) we then apply a multi-Gaussian expansion (MGE) (Cappellari 2002) to fit the 2D surface densities of the disc and the bar separately, and (3) we deproject the disc and the bar separately, allowing for different assumptions about their internal 3D shapes. Finally, by combining the 3D densities of the disc and the bar, we derive the intrinsic 3D luminosity distribution of the entire galaxy. Further details for each step are provided in the following.

3.1.2.1 The photometric image decomposition We employ GALFIT (Peng et al. 2010) to decompose the 2D surface brightness of NGC 4371 using a four-component model, which provides a good fit to 3.6 µm image. This model includes a central compact Sérsic component, an exponential nuclear disc, a bar (Sérsic profile), and an exponential main disc. The goal of the GALFIT fitting is to obtain a parametrized model that fully matches the global surface brightness of the galaxy. The central compact component (point source) in the model is introduced solely to enhance the goodness of the GALFIT fit in the central pixels of the image, it is not necessary to be a physically defined component and is not considered as a classical bulge. The uncertainty in the image, derived from Poisson noise, is used to weight the data points during the fitting process. We combine the nuclear and main discs as the disc component and then subtract the disc component from the original image to obtain a residual bar. Note that, in the following analysis, we only use the residual bar derived from the original image and do not use the GALFIT compact S'ersic component or the bar's S'ersic model.



Figure 2. The photometric analysis of the S⁴G image and the GALFIT best-fitting model. The left column from top to bottom shows: 1D surface brightness profile, ellipticity, and position angle along the major axis of the S⁴G image (red) and GALFIT model (black). The last panel on the left column represents each component's 1D surface brightness profile in the GALFIT best-fitting model (central compact source, nuclear disc, bar, and main disc). Right column from top to bottom: 2D surface brightness distribution of the NGC 4371 S⁴G image, GALFIT best-fitting model, the residual, and the residual bar extracted by subtracting the nuclear and main discs from the original image. The black dashed line indicates the projected bar radius $R_{\text{bar}}' \sim 35$ arcsec. The image *x*-axis is flipped to align with the orientation of the kinematic map. The pink shaded area indicates the extent of the GALFIT nuclear disc component, from a half-light radius of ~ 8.8 to 20 arcsec.

Table 1 presents the structural parameters derived from the fit. The right column of Fig. 2 shows (from top to bottom) the S⁴G image, the GALFIT model, the residual, and the residual bar obtained by subtracting the c component (comprising both the nuclear and the main discs) from the S⁴G image. The left column of Fig. 2 shows (from top to bottom) the radial surface brightness profile derived from the S⁴G image and the GALFIT model with ellipse fits to the isophotes, ellipticity, position angle, and the radial surface brightness profile of the subcomponents. From the GALFIT model, we determine the projected bar semi-major axis to be $R_{\rm bar}' \sim 35$ arcsec, which aligns with the measurements reported by Gadotti et al. (2015) using image decomposition.

In Fig. 2 (top left), we mark the locations of the nuclear disc and the bar on the surface brightness profile derived from the S⁴G image. The exponential decrease in brightness between 10–20 arcsec (1 arcsec $\sim 82 \text{ pc}$) suggests the presence of a nuclear disc. The characteristic bump at a radius of $\sim 35 \text{ arcsec}$ corresponds to the projected bar radius we measured.

As discussed in Gadotti et al. (2015), the three local maxima in the ellipticity profile correspond to distinct structures of the nuclear disc, the bar, and the main disc. The nuclear disc and the main disc exhibit higher ellipticity than the bar. The two minima in the ellipticity profile signify the transition points: the first from the nuclear disc to the bar and the second from the bar to the main disc. In the position



Figure 3. Left panel: The surface density contours of disc component (grey lines), overplotted with contours of the best-fitting MGE model (orange lines). Right panel: similarly for the bar, where the twist is allowed in the MGE model.

angle (PA) profile, the nuclear disc exhibits a similar PA to that of the main disc, as also evidenced by their kinematics (Gadotti et al. 2015). The sharp drop in PA at the bar radius is attributed to the bar in NGC 4371 being nearly perpendicular to the nuclear and main discs.

The match between the ellipticity and PA profiles of our GALFIT model and the image indicates that we have effectively captured the main features of the image. The only purpose of this photometric decomposition is to separate the disc and the residual bar component, ensuring that their combination faithfully represents the image of the entire galaxy. This will be employed to construct the 3D stellar mass density and compute the stellar gravitational potential, as detailed below. We will not rely on this photometric decomposition to study the intrinsic properties of each component; instead, we will perform a structural decomposition based on the 3D dynamical models superposed by stellar orbits at the end.

3.1.2.2 MGE fitting We combine the nuclear disc and the main disc as an axisymmetric disc component and consider the residual bar as a triaxial component. We fit MGEs to each component separately as shown in Fig. 3. Note that, we use the residual barred bugle, which allows us to capture the triaxiality better than the fitted elliptical Sérsic bar.

We obtain parameters $(L_j, q'_j, \sigma'_j, \Delta \psi'_j)$ of the 2D Gaussians from the fitting, where L_j is the total luminosity, q'_j is the projected flattening, and σ'_j is the scale length along the projected major axis of each Gaussian component $j = 1 \dots N$. $\Delta \psi'_j$ is the isophotal twist of each Gaussian. The MGE fitting parameters of the barred bugle (Gaussians with $\Delta \psi'_j \neq 0$) and the disc (Gaussians with $\Delta \psi'_j = 0$) are presented in Table 2.

3.1.2.3 Deprojection The orientation of a projected system is defined by three viewing angles $(\theta, \varphi, \psi), \theta$, and φ indicate the orientation of the line-of-sight with respect to the principal axes of the object. For example, projections along the intrinsic major, intermediate, and minor axes correspond to $(\theta = 90^{\circ}, \varphi = 0^{\circ}), (\theta = 90^{\circ}, \varphi = 90^{\circ})$ and $(\theta = 0^{\circ}, \varphi \text{ irrelevant})$, respectively. ψ is the position angle, which indicates the rotation of the object around the line of sight in the sky plane (see fig. 2 in de Zeeuw & Franx 1989). The intrinsic parameters describing a 3D Gaussian component (σ_j, p_j, q_j) can be derived analytically using a set of viewing angles (θ, φ, ψ) and parameters measured for the 2D Gaussians (see equations 7–9 in VdB08). For a rigid body comprised of multiple Gaussian components, all Gaussians are fixed to have the same viewing angles; the allowed orientations are thus the intersection of allowed viewing angles (θ, φ, ψ) of all the Gaussians.

We consider the disc and the bar as two rigid body components, and we deproject them separately. Thus, we have three viewing

Table 2. The MGE fitting parameters of the bar (Gaussians with $\Delta \psi'_j \neq 0$) and the disc (Gaussians with $\Delta \psi'_j = 0$).

j	$L_j(L_\odot pc^{-2})$	$\sigma'_j(arcsec)$	q_j'	$\Delta \psi_j'(^\circ)$
1	4519.857	0.294	0.99	-59.0
2	10253.941	1.069	0.99	-59.0
3	3263.905	2.119	0.981	-66.0
4	535.953	18.296	0.58	-63.831
5	53.586	45.0	0.614	-59.611
6	3842.074	2.343	0.575	0.0
7	4078.952	5.112	0.58	0.0
8	2506.333	8.667	0.57	0.0
9	179.587	11.512	0.965	0.0
10	614.981	14.211	0.57	0.0
11	217.091	31.804	0.57	0.0
12	151.360	60.354	0.57	0.0
13	31.390	106.066	0.57	0.0

angles $(\theta_{\text{disc}}, \varphi_{\text{disc}}, \psi_{\text{disc}})$ for the disc and three viewing angles $(\theta_{\text{bar}}, \varphi_{\text{bar}}, \psi_{\text{bar}})$ for the bar.

The disc is considered as an axisymmetric oblate system with the major axis aligned with the x' axis of the image so that $\psi_{\text{disc}} =$ 90°, and all Gaussians have $\Delta \psi'_j = 0$, while φ_{disc} is irrelevant. The inclination angle of the disc θ_{disc} is left as a free parameter, with its lower limit constrained by $\cos(\theta_{\text{disc}})^2 < q''_{\text{min}}$ where q'_{min} indicate the flattest Gaussian of MGEs fitted to the disc.

The bar is triaxial, so its Gaussians can have different isophotal twists $\Delta \psi'_j$. The twists of bar Gaussians are measured with respect to the major axis of the disc in the observational plane. We therefore have $\Delta \psi'_j = \psi'_j - \psi_{\text{disk}}$, and $\psi_{\text{disk}} = 90^\circ$. The intrinsic position angle ψ of an isolated triaxial system is in principle unknown (van den Bosch et al. 2008). However, we have a reference disc. We fix the reference bar position angle to $\psi_{\text{bar}} = \psi_{\text{disc}} = 90^\circ$ and use $\Delta \psi'_j$ to include the real information of the bar position angle.

Deprojection of the model is inherently not unique; there is still a wide range of possible viewing angles. We impose certain constraints to further reduce the degree of this degeneracy: (1) We assume the bar major axis aligns with the disc plane, implying the bar inclination angle θ_{bar} , matches the disc θ_{disc} . This assumption narrows down the allowed inclination angle to be $\theta_{\text{bar}} = \theta_{\text{disc}}$, and the angle φ_{bar} is left free. (2) We further refine the inclination angle constraint using the observationally derived inclination of $\sim 60^{\circ}$ from Gadotti et al. (2015), stating that $|\theta - 60^{\circ}| \le 10^{\circ}$. These constraints significantly mitigate the degeneracy of viewing angles and narrow the range of permissible viewing angles (see fig. 4 in Tahmasebzadeh et al. 2021); they also lead to reasonable intrinsic shapes for barred galaxies according to our test with mock data (Tahmasebzadeh et al. 2021). We acknowledge that the deprojection process does not achieve absolute uniqueness. The remaining free viewing angles, although currently undetermined, will be constrained by the following dynamical model that fits the kinematic data. This approach underscores the fact that achieving complete uniqueness in the deprojection may not be necessary for our analysis, given that the fitting to kinematic data is designed to pinpoint the most plausible viewing angles for the galaxy.

We therefore have two viewing angles as free parameters: θ_{disk} and φ_{bar} , which will be just denoted as θ and φ in what follows.

Once we infer the 3D luminosity density distribution with a set of viewing angles, we multiply it by a constant stellar mass-to-light ratio M_*/L to obtain the 3D stellar mass distribution, which is another free parameter in the mass model. The assumption of a constant

 M_*/L across the entire galaxy is fairly adequate for this case study. Gadotti et al. (2015) showed that the stellar population is uniformly old across the MUSE–TIMER field. Furthermore, we utilize the S⁴G image at 3.6 um, which traces the old stellar populations that dominate the mass budget of galaxies; the M_*/L at 3.6 um should exhibit minimal variation across different populations (Meidt et al. 2012, 2014; Querejeta et al. 2015).

We assumed a stationary gravitational potential in the rotating frame for a barred galaxy with a triaxial bar model. Thus, the bar pattern speed Ω_p is left as another free parameter.

In summary, we have five so-called free hyperparameters in the model of gravitational potential: DM virial mass M_{200} , inclination angle θ , bar azimuthal angle φ , stellar mass-to-light ratio M_*/L , and the bar pattern speed Ω_p .

3.2 Generating the orbit library

We sample the initial conditions of orbits in the x-z plane, using the properties of separable models as described in VdB08. In such models, tube orbits (excluding shell orbits where the outer and inner radial turning points coincide) intersect the x-z plane perpendicularly twice above z > 0. Therefore, it is not necessary to sample the entire x-z plane. We determine the orbital energy E in a stationary frame using a logarithmic grid in radius; each energy corresponds to a grid radius r_i calculated by evaluating the potential at (x, y, z) = $(r_i, 0, 0)$. For each energy, the starting point (x, z) is chosen from a linear open polar grid of (R, ϕ) , between the location of the shell orbits and the equipotential surface with zero velocity for that energy, where $R = \sqrt{x^2 + z^2}$ and $\phi = \arctan(x/z)$ (as shown by the grey area in fig. 2 of VdB08). The initial starting points are sampled in the inertial frame with $v_y = \sqrt{2[E - \Phi(x, 0, z)]}$, and converted to velocities in the rotating frame for the orbit integration as discussed in Tahmasebzadeh et al. (2022). We sample two sets of tube orbit libraries in x-z plane, one with $v_y > 0$ and the other with $v_y < 0$ as these will represent different orbits when integrated with the presence of a rotating bar. A large number of starting points are sampled across the three integrals with $(n_E \times n_R \times n_{\phi}) = (30 \times 15 \times 13)$, and we adopt the dithering number to be 3 to impose the smoothness of orbit-superposition models. Hence, each orbital bundle contains 27 orbits with close starting points.

3.3 Weighing the orbits

The constraints for the model include 1- kinematic maps, which typically encompass the observational velocity V_o^l (*o* stands for observation) and dispersion σ_o^l for each aperture *l*, along with the Gaussian–Hermite (GH) coefficients $h_{3,o}^l$ and $h_{4,o}^l$, 2- the surface brightness in the 2D observational plane and the 3D luminosity distribution, which is deprojected from the 2D image. Note that, V_o^l and σ_o^l are the parameters of the GH function obtained by the full spectrum fitting; they are not the mean velocity and its dispersion unless all higher-order moments are zero.

The model comprises a superposition of thousands of orbit bundles, where each orbit bundle, denoted as k, is weighted by w_k . We minimize the χ^2 between the data and model to get the solution of the orbit weights. The χ^2 is contributed by two parts, the fit to the luminosity distribution and the fit to kinematic maps.

$$\chi^2_{\rm NNLS} = \chi^2_{\rm lum} + \chi^2_{\rm kin}.$$
(3)

We allow a relative error margin of 1 per cent for both 2D and 3D luminosity distribution fittings. The 2D luminosity distribution, represented as S_l , is stored within the observational apertures on the

$$\chi_{\text{lum}}^{2} = \chi_{\text{S}}^{2} + \chi_{\rho}^{2}$$

= $\sum_{l=1}^{N_{\text{kin}}} \left[\frac{\sum_{k} w_{k} S_{k}^{l} - S_{l}}{0.01 S_{l}} \right]^{2} + \sum_{n=1}^{360} \left[\frac{\sum_{k} w_{k} \rho_{k}^{n} - \rho_{n}}{0.01 \rho_{n}} \right]^{2},$ (4)

where N_{kin} represents the total number of apertures in a single kinematic map, while w_k denotes the weight assigned to orbit k.

From observations, we describe the LOSVD profile f_l in each aperture l as a GH distribution (Gerhard 1993; van der Marel & Franx 1993) with parameters $(V_o^l, \sigma_o^l, h_{3,o}^l, h_{4,o}^l)$ and corresponding errors $(\Delta V_o^l, \Delta \sigma_o^l, \Delta h_{3,o}^l, \Delta h_{4,o}^l)$. When V_o^l and σ_o^l are chosen as the centre and the width of the best-fitting Gaussian approximating the original LOSVD, this resulted in $h_{1,o}^l = h_{2,o}^l = 0$.

We denote the LOVSD contributions of orbit bundle k at aperture l as f_k^l . If we expand f_k^l in a GH series also with the central velocity and dispersion fixed at the observed V_o^l and σ_o^l , then the resulting GH coefficients $h_{n,k}^l$ with n = 1, 2, 3, and 4 will contribute linearly to the observations, so that

$$\chi_{\rm kin}^2 = \sum_{l=1}^{N_{\rm kin}} \sum_{n=1}^{n_{\rm GH}} \left[\frac{\sum_k w_k S_k^l h_{n,k}^l - S_l h_{n,o}^l}{S_l \Delta h_{n,o}^l} \right]^2, \tag{5}$$

where the model predictions are luminosity weighted in the same manner as the observations, and n_{GH} , which represents the number of kinematic moments used for the fitting, is set to 4 here. The errors of $(\Delta V_o^l, \Delta \sigma_o^l, \Delta h_{3,o}^l, \Delta h_{4,o}^l)$ are usually provided directly from observations, while we derive $\Delta h_{1,o}^l, \Delta h_{2,o}^l$ following Rix et al. (1997). The luminosity density is usually easy to fit, so that χ^2_{kin} is the dominant term contributing to goodness of fit χ^2_{NNLS} (e.g. Zhu et al. 2018b). We use the non-negative least squares (NNLS) implementation to find the solution of orbit weights by minimizing the χ^2_{NNLS} between data and model following VdB08.

3.4 Exploring the parameter space

We have five free hyperparameters in the model: 1- the stellar massto-light ratio M_*/L at 3.6 µm band, 2- the inclination angle θ , 3- the bar azimuthal angle φ , 4- the bar pattern speed Ω_p , and 5- the DM virial mass M_{200}/M_* .

In our search for the best-fitting model, we employ an optimized grid iterative process. We start with initial guesses of the hyperparameters. We then walk two steps in each direction of the parameter grid by taking relatively large intervals of 0.2, 6, 2, 2, and 0.4 for M_*/L , Ω_p , θ , φ , and $\log_{10}(M_{200}/M_*)$, respectively.

After completing the initial models, we select those with a $\chi^2 - \chi^2_{min} < 100 \times \sqrt{2n_{GH}N_{kin}}$ from the existing models. The factor of 100 is an empirical choice. Then, we run new models around the selected models. This iterative process continues until we identify the model with the lowest χ^2 , ensuring that all models in its vicinity are also evaluated. Then, we halve the parameter step sizes to better sample the grids around the best-fitting models.

Finally, once again, we repeat the iterative search process, but this time with an increased threshold of $\chi^2 - \chi^2_{min} < 500 \times \sqrt{2n_{GH}N_{kin}}$. This approach is adopted to prevent the process from getting trapped in a local minimum and to guarantee that all models within a 3σ confidence level are thoroughly calculated.

In classical statistic analysis for analytic models fitting to data, the 1 σ confidence level is determined by $\Delta \chi^2 = 1$ for one degree of freedom. However, it is unsuitable for our case where the model numerical noise dominates the χ^2 (Lipka & Thomas 2021). In Zhu et al. (2018b), $\Delta \chi^2 \equiv \sqrt{2n_{\rm GH}N_{\rm kin}}$ is adopted as 1 σ confidence level, which is consistent with the χ^2 fluctuation caused by numerical noise of their models for CALIFA galaxies.

Here, we adopt a similar approach to calculate the 1σ confidence level, utilizing a bootstrapping process. This involves randomly perturbing the kinematic data within its error margins and generating 100 new kinematic maps. Subsequently, we re-fit only our bestfitting model-which has a fixed potential and orbit library-to these 100 perturbed data sets. The standard deviation of the χ^2 values from these fittings is then used to define the χ^2 fluctuation attributable to the model's numerical noise, which is related to various factors such as the non-uniqueness of the orbit weight distribution, degree of freedom, observational errors, etc. Through this method, we determined that the 1σ confidence level for models constrained by TIMER data is approximately $\sim 0.5 \times \sqrt{2n_{\rm GH}N_{\rm kin}}$, and for those constrained by ATLAS3D data, it is about $\sim 2 \times \sqrt{2n_{\rm GH}N_{\rm kin}}$.

4 RESULTS

By exploring the parameter space, we obtain the best-fitting models that match both the target density distribution and all kinematic features. We anticipate achieving robust constraints on the free parameters in the gravitational potential (Section 4.1). This will enable us to determine the bar pattern speed (Section 4.2), the enclosed mass profiles (Section 4.3), and the viewing angles, thus the 3D intrinsic structure of the galaxy. We will facilitate a dynamical structure decomposition based on the 3D model, allowing us to explore the intrinsic properties of each structural component (Section 4.4).

4.1 Best-fitting models

We consider all models that fall within the 1σ confidence level and compute the mean and standard deviation of each parameter for these models. These calculations are then used to determine the best-fitting parameter and its corresponding 1σ error.

From the models constrained by TIMER, we obtained the bestfitting parameters as $\theta = 60 \pm 2^{\circ}$, $\varphi = -12 \pm 1^{\circ}$, $\Omega_{\rm p} = 23.6 \pm 2.8 \,\rm km \, s^{-1} \, \rm kpc^{-1}$, $M_*/L_{3.6 \,\mu\rm m} = 1.05 \pm 0.05 \,\rm M_{\odot}/L_{\odot,3.6 \,\mu\rm m}$, and DM virial mass $\log_{10}(M_{200}/M_*) = 2.4 \pm 0.5$.

From the models constrained by ATLAS3D, we obtain $\theta = 60 \pm 2^{\circ}$, $\varphi = -12 \pm 1^{\circ}$, $\Omega_{\rm p} = 22.4 \pm 3.5 \,\rm km \, s^{-1} \, kpc^{-1}$, $M_*/L_{3.6 \,\mu\rm m} = 0.99 \pm 0.11 \,\rm M_{\odot}/L_{\odot,3.6 \,\mu\rm m}$, and DM virial mass $\log_{10}(M_{200}/M_*) = 3.0 \pm 0.6$. The parameters obtained from the two sets of models are generally consistent with each other. There is a larger uncertainty in the model constrained by ATLAS3D data, due to the smaller spatial coverage of ATLAS3D compared to the MUSE data. The parameters grid of all models is included in Appendix Figs A4 and A5.

In 3.6 µm band, by assuming a Chabrier initial mass function (IMF), the stellar population synthesis gives an average of $\Upsilon_{3.6 \,\mu\text{m},*}^{\text{Chab}} \sim 0.6$, with uncertainty of 0.1 dex (Meidt et al. 2014), which is lower than the dynamical stellar mass-to-light ratio $\Upsilon_{3.6 \,\mu\text{m},*}^{\text{dyn}} \sim 1.05$ we obtained. If we scale it to a Salpeter IMF multiplied by a factor of 1.8, then $\Upsilon_{3.6 \,\mu\text{m},*}^{\text{Salp}} \sim 1.08$ is consistent with our dynamical results within 1σ uncertainty.

We present the best-fitting models of NGC 4371 in Fig. 4. Columns from left to right displaying the 2D surface density, LOS velocity, velocity dispersion, h_3 , and h_4 . The first three rows depict the



Figure 4. The best-fitting models of NGC 4371 to the data from TIMER (top panel) and ATLAS3D (the bottom), respectively. In each panel, columns from left to right represent the 2D surface density, velocity, velocity dispersion, h_3 , and h_4 , rows from top to bottom are the observational data, the best-fitting Schwarzschild barred model, and the residuals. The overplotted grey contours are the surface brightness of the NGC 4371. The TIMER data covers the regions with $r \lesssim 35$ arcsec, while ATLAS3d covers only the inner regions with $r \lesssim 25$ arcsec. The models are well matched with both sets of data.

TIMER data covering \sim 35 arcsec in radius, followed by the bestfitting Schwarzschild bar model and the residuals (calculated as the difference between the TIMER data and the model, normalized by the uncertainties in each TIMER data bin). The fourth to sixth rows illustrate the ATLAS3D data within ~ 20 arcsec in radius, the corresponding best-fitting model, and the residuals. All panels include grey contours indicating the surface brightness of NGC 4371.



Figure 5. The bar pattern speed obtained from the two sets of models constrained by TIMER and ATLAS3D separately. Top panel: the spatial coverage of TIMER (red) and ATLAS3D (blue) data comparing to the whole galaxy, with contours showing the 3.6 µm S⁴G image. Bottom panel: the model parameter grid bar pattern speed Ω_p versus the stellar mass-to-light ratio $M_*/L_{3.6um}$. The contours indicate the 1σ (solid line), 2σ (dashed line), and 3σ (dotted line) uncertainties for the best-fitting models using TIMER (red) and ATLAS3D (blue). The small dot with error bars indicates the mean and 1σ uncertainty obtained from the models within the 1σ confidence level. The triangles indicate the best-fitting model for each data set. The pattern speed obtained from the two sets of models are well consistent with each other, with smaller uncertainty from the model constrained by TIMER.

Our models successfully reproduce key kinematic features, ranging from the nuclear disc to the outer bar regions. They demonstrate improved fitting to the kinematic maps compared to nearly axisymmetric models that do not explicitly include the bar. In appendix A, we discuss the limitations of an axisymmetric model, illustrating how it still can fit the data while significantly biasing the internal properties of the best-fitting model (see Fig. A2).

4.2 The bar pattern speed and rotation parameter

4.2.1 The bar pattern speed

As depicted in the upper panel of Fig. 5, neither the TIMER nor the ATLAS3D data encompass the disc-dominated regions of NGC 4371, which complicates the determination of the bar pattern speed using the TW method. However, the bar pattern speed can be well constrained by the Schwarzschild model, even with limited data coverage. This is because the orbit-superposition method utilises full kinematic information, including higher velocity moments, which strongly depends on the internal orbital structure within the bar regions. Even when data cover only the inner region of the bar, it still contains crucial kinematic information that reflects the overall internal kinematics and orbital structure of the bar. This method has been verified through tests on a set of mock data representing various orientations and covering only the inner region of the bar (Tahmasebzadeh et al. 2022) (see also other independent tests by Vasiliev & Valluri (2020) and Dattathri et al. 2024). These tests, which all used data that covered only the inner region of the bar, consistently report the recovery of pattern speed with a high degree of precision, with an uncertainty of less than 10 per cent.

In the bottom panel of Fig. 5, we illustrate the parameter space of Ω_p versus M_*/L constrained by the TIMER (red) and ATLAS3D (blue) data. Solid, dashed, and dotted lines indicate the confidence level of 1σ , 2σ , and 3σ regions, respectively. We find the bar pattern speed of $\Omega_p = 23.6 \pm 2.8 \,\mathrm{km \, s^{-1} \, kpc^{-1}}$ and $\Omega_p = 22.4 \pm 3.5 \,\mathrm{km \, s^{-1} \, kpc^{-1}}$ from the models constrained by the TIMER and ATLAS3D data, respectively. The ATLAS3D data yield similar bar pattern speed but with larger uncertainties due to limited data coverage, in contrast to the TIMER data that fully encompass the bar regions. However, the ATLAS3D coverage still contains important information. It fully covers the peak velocity dispersion in the lower part, the higher values of h_4 in the central region, and the rotation in the nuclear disc region reflected in both velocity and h_3 , and they are consistent with the MUSE data at the same regions.

To demonstrate the impact of pattern speed on various moments in kinematic maps, Fig. 6 compares the maps generated by models with different pattern speeds. The first column displays the velocity (top) and velocity dispersion (bottom) derived from the TIMER data. The models, with pattern speeds of $\Omega_p = 0, 11, 20, 32, 41 \text{kms}^{-1} \text{kpc}^{-1}$, are presented by columns from left to right. Among them, $\Omega_p =$ 20kms⁻¹kpc⁻¹ represents the best-fitting model. All other parameters are kept the same as the best-fitting model. The second and fourth rows display the residuals, calculated as the difference between the TIMER data and the model, divided by the uncertainties in the TIMER data for each bin. The residual maps clearly demonstrate that as the pattern speed deviates from $\Omega_p = 20 \text{kms}^{-1} \text{kpc}^{-1}$, the value of χ^2 increases significantly. When Ω_p is smaller than the optimal value, the model predicts insufficient regular rotation in the outer bar regions, and the dispersion in the two high-dispersion lobes becomes too large. In contrast, a larger Ω_p leads to predictions of an overly strong regular rotation in the outer bar regions, causing the high-dispersion lobes to vanish. Both the velocity and dispersion maps are crucial for accurately constraining the pattern speed within our Schwarzschild model.

In addition to requiring extended data coverage, the TW method also demands an optimal orientation of the disc and bar for accurate pattern speed measurements. As discussed in Zou et al. (2019), a disc nearly face-on/edge-on or a bar nearly parallel/perpendicular to the disc major axis could lead to a wrong measurement of the pattern speed. The bar in NGC4371 is nearly perpendicular to the main axis of the disc. We attempted to apply the TW method to NGC 4371 and yielded a pattern speed of approximately $\sim 220 \,\mathrm{km \, s^{-1} \, kpc^{-1}}$, which is very likely to be erroneous. This is much higher than any pattern speed ever seen in observation or simulations.



Figure 6. The kinematic maps predicted by models with different pattern speeds. The first column shows velocity (top) and velocity dispersion (bottom) from TIMER data, the following columns from left to right are the model predictions with pattern speed of $\Omega_p = 0, 11, 20, 32, 41 \text{kms}^{-1} \text{kpc}^{-1}$ in which $\Omega_p = 20 \text{kms}^{-1} \text{kpc}^{-1}$ is the best-fitting model. All other parameters were kept the same as the best-fitting model. The second and fourth rows represent the corresponding residuals. The residuals of both the velocity and velocity dispersion maps become significantly larger with the pattern speed deviating from the best-fitting model ($\Omega_p = 20 \text{kms}^{-1} \text{kpc}^{-1}$).

4.2.2 The dimensionless bar rotation parameter \mathcal{R}

For each model within the 1σ confidence level, we compute the local angular frequency $\Omega(R)$ using the model potential Φ , following Binney & Tremaine (2008):

$$\Omega(R) = \sqrt{\frac{1}{R} \frac{\mathrm{d}\Phi}{\mathrm{d}R}} \tag{6}$$

To obtain a more accurate estimation of $\Omega(R)$ in our triaxial potential, we use quasi-axisymmetric approximation and calculate the average frequencies along the major and minor axes of the bar in the galaxy plane. However, since our model includes an axisymmetric disc, the $\Omega(R)$ along both axes are similar outside the bar. In Fig. 7, we present the angular frequency (black curve) as a function of radius. Assuming that the bar rotates as a solid body with angular velocity Ω_p , we determine the corotation radius $R_{\rm cor}$ using $\Omega(R_{\rm cor}) = \Omega_p^{\rm model}$, where $\Omega_p^{\rm model}$ is the pattern speed obtained by the corresponding model. At $R_{\rm cor}$, the materials within the disc rotate at the same angular velocity as the bar. For each of our models, we infer the corotation radius by analysing the angular frequency profile and the bar pattern speed, resulting in a single value for $R_{\rm cor}$ without associated uncertainty. Thus, we compute the co-rotation radius for all models within the 1σ region, which leads to a mean and standard deviation of $R_{\rm cor} = 131 \pm 24$ arcsec (~ 10.7kpc).

For the bar length of NGC 4371, we obtained the projected bar radius of $R_{\text{bar}}' = 35$ arcsec from the photometric decomposition. With the inclination angle of $\theta = 60^{\circ}$ and the bar angle of $\varphi = -12^{\circ}$ from our best-fitting Schwarzschild mode, we obtain the intrinsic bar radius of $R_{\text{bar}} \sim 69$ arcsec (~ 5.7 kpc) using $R_{\text{bar}} = R_{\text{bar}}' \sqrt{\sin^2 \varphi} + \cos^2 \varphi / \cos^2(60^{\circ})$ (Gadotti et al. 2007). Note that, φ here differs from that in Gadotti et al. (2007), so the formula is modified accordingly. We considered the range of 64 to 75 arcsec as the upper and lower limits of bar length following Erwin, Pohlen & Beckman (2008), who employed the IRAF ellipse package to estimate the possible maximum and minimum bar lengths.

To estimate the dimensionless bar rotation parameter $\mathcal{R} \equiv R_{\rm cor}/R_{\rm bar}$ and its uncertainties for NGC 4371, we first compute the corotation radius for each model, which yields a single value without error. We then divide the co-rotation radius by the bar length, as well as the upper and lower limits of the bar length. This results in three possible values for the bar rotation parameter for each model, including its upper and lower limits. For example, in the best-fitting



Figure 7. Analysis of co-rotation resonance for models within 1σ confidence level constrained by TIMER. The black dashed curve and shaded regions indicate the local angular frequency profile $\Omega(R)$ and its 1σ uncertainty. The horizontal blue line with shaded regions represents the measured bar pattern speed and its 1σ uncertainty. The vertical green line indicates the bar radius $R_{\text{bar}} \sim 69$ arcsec measured by photometric decomposition, with the shaded regions indicating the lower/upper limits of the bar radius. The vertical red line with shaded regions indicates the location of co-rotation resonance (R_{cor}) at ~ 132 arcsec, along with its 1σ uncertainty.

Schwarzschild model, the bar rotation parameter is $\mathcal{R} = 2.5 \pm 0.2$. However, we report the bar rotation parameter for the galaxy by calculating the mean and standard deviation of bar rotation parameter values across all models within the 1σ region. This results in a bar rotation parameter of $\mathcal{R} = 1.88 \pm 0.37$. A bar in a galaxy can be classified as fast (if $1 < \mathcal{R} < 1.4$) or slow (if $\mathcal{R} > 1.4$) (Debattista & Sellwood 2000). Therefore, we conclude that NGC 4371 has a slow bar, even considering the lower limit of $\mathcal{R} = 1.51$, although it is near the borderline.

4.3 Mass profile

We show the enclosed mass profiles of NGC 4371 in Fig. 8. The black, red, and blue curves represent the total enclosed mass, stellar mass, and DM mass profiles, respectively. The shaded regions represent the 1σ uncertainty from the models with TIMER. The total mass profiles are well constrained with 1σ uncertainties of 10 per cent within data coverage of 35 arcsec. The mass profiles from the models constrained by ATLAS3D are consistent with those from TIMER; however, they exhibit relatively larger uncertainties in the outer regions due to the smaller data coverage of ATLAS3D. Our model predicts that NGC 4371 has a high dark matter fraction of $M_{\rm DM}/M_{\rm total} \sim 0.51 \pm 0.06$ within the bar region. It is similar to the dark matter fraction ($\sim 0.53 \pm 0.02$) of NGC 4277 inside the bar regions, which is also a slow bar (Buttitta et al. 2023). These results are aligned with cosmological simulations, suggesting that fast bars are typically found in baryon-dominated discs (Fragkoudi et al. 2021). To confirm the correlation between the bar rotation parameter and the dark matter content in real galaxies, it is necessary to apply these measurements to a substantial number of barred galaxies.



Figure 8. The enclosed mass profiles of the best-fitting bar models constrained by TIMER (solid line) and ATLAS3D (dashed line). The red, blue, and black curves represent the stellar mass, dark matter mass, and total mass, respectively. The shaded regions indicate the 1σ uncertainty from the set of models constrained by TIMER. The vertical grey line indicates the bar length $R_{\rm bar} \sim 70$ arcsec.

4.4 Dynamical structure decomposition and their internal properties

4.4.1 Orbital decomposition

After finding the best-fitting Schwarzschild model and constraining the model parameters linked to the overall galaxy dynamics, we obtain the orbits and their corresponding weights that describe the galaxy kinematic and photometric data. We now aim to dynamically decompose this galaxy based on the orbital properties in the bestfitting models. The dynamic decomposition is fundamentally different from the photometric decomposition described earlier. In photometric decomposition, different assumed components are combined to fit the photometric image. In contrast, dynamical decomposition does not involve any further fitting process, since the best-fitting Schwarzschild model is already established. Instead, we utilize the orbits from the best-fitting model and group orbits with similar kinematic and morphological features based on our arbitrary (but meaningful) criteria to construct different components. To achieve this, we first characterize each orbit using various parameters. Then, we classify the orbits into different groups based on their properties and rebuild the density and kinematic maps with each group of orbits. This process allows us to better understand the contributions of different orbital families to the overall structure and dynamics of the galaxy.

We first characterize the stellar orbits by two parameters following Zhu et al. (2018b): the circularity λ_z defined as the angular momentum L_z (recorded in an inertial frame) normalized by the maximum angular momentum allowed by a circular orbit L_c with the same binding energy. The radius *r* is taken as the average of particles along an orbit stored with equal time steps in the Schwartzchild model. We show the stellar orbit distribution of the best-fitting model constrained by TIMER in the space of circularity λ_z versus radius *r* in the left panel of Fig. 9. The vertical dotted line indicates the bar radius, which is within the kinematic coverage of TIMER data. Prograde circular



Figure 9. The stellar orbit distribution of the best-fitting model constrained by TIMER data. *Left panel*: the probability density distribution of orbits in the space of circularity λ_z versus time-averaged radius *r*. Darker colours indicate a higher probability density, as the colour bar indicates. The vertical dotted lines indicate the bar radius and kinematics data extent of TIMER. *Right panel*: the location of orbits making up different structures: BP/X bulge (red), classical bulge (black), nuclear disc (green), and main disc (blue). The larger symbol size corresponds to orbits with higher weights from the minimum value of 10^{-7} to a maximum of 10^{-1} (the total weight of all orbits is = 1).

orbits have $\lambda_z \sim 1$, box orbits with no net rotation have $\lambda_z \sim 0$, and circular retrograde orbits have $\lambda_z \sim -1$. The internal stellar orbit distribution should be well constrained by the Schwarzschild model, as demonstrated in previous tests with many mock galaxies (Zhu et al. 2018b, a, 2022; Jin et al. 2020).

To further identify the orbits that make up the bar, we employ the frequency analysis of the orbits in the best-fitting model using the 'Numerical Analysis of Fundamental Frequencies (NAFF)' software.² We compute orbital frequencies in both Cartesian and cylindrical coordinates and classify the orbits into various types following the criteria established by Valluri & Merritt (1998) and Valluri et al. (2016). This orbit classification is crucial for identifying different structures within the galaxy, particularly those associated with bars.

We begin by segregating all orbits within the bar, considering those with apocentre radii smaller than the bar radius ($R_{apo} < R_{bar}$). These orbits are then classified into three groups: BP/X bar, Classical bulge, and Nuclear disc. The remaining orbits with $R_{apo} > R_{bar}$ are considered to form the main disc. Detailed descriptions of the orbits in each category are as follows:

(1) **BP/X bar**: including x_1 , banana (1 : 2 resonance), periodic and non-periodic *z*-tube orbits which are within the bar and elongated along the bar. These orbits exhibit prograde motion and display an X-shaped structure in edge-on projected surface density, consistent with previous studies (Portail, Wegg & Gerhard 2015; Abbott et al. 2017; Fragkoudi et al. 2017; Parul, Smirnov & Sotnikova 2020) and our analysis of the orbital structures in a few simulations with BP/X bulges (Tahmasebzadeh et al. 2024); (2) **Classical bulge**: we assume that the classical bulge composed of non-periodic box orbits with $\lambda_z < 0.4$, which contribute to a hot dispersion-dominated, and round structure in both face-on and edge-on views. It is widely accepted that classical bulges are slowly rotating and dominated by high-velocity dispersion (Babusiaux et al. 2010; Rojas-Arriagada et al. 2014; Erwin

²https://bitbucket.org/cjantonelli/naffrepo/src/master/

et al. 2021; Fragkoudi et al. 2020) (3) Nuclear disc: a rotationdominated structure perpendicular to the bar constructed by highly circular orbits with $\lambda_z > 0.4$ in the inner regions; (4) Main disc: all orbits with apocentre radii larger than the bar radius ($R_{apo} > R_{bar}$). It is dominated by highly circular orbits. Note that, the orbits in the main disc extend well beyond the range of the available kinematic data, so they are constrained solely by the luminosity distribution of the galaxy.

We show the location of the four structures in the phase space of λ_z versus *r* in the right panel of Fig. 9. Highly circular orbits distributed in the inner and outer regions make up the nuclear and main discs. The non-circular orbits could construct a classical bulge, and could also be part of the bar. The bar orbits overlap with the orbits of the classical bulge in the phase-space of λ_z versus *r*. Overall, the structures are distinctly separated in the phase space distribution; however, the bar orbits in the inner regions exhibit partial overlap with orbits constituting the classical bulge in λ_z versus *r*, illustrating the importance of detailed orbital classification using frequency analysis for separating the bar.

4.4.2 Morphology and kinematics of the four structures

4.4.2.1 Morphology of each structure We reconstruct the 3D density distribution and kinematics by summing the particles sampled from the orbits, the entire galaxy from all orbits in the model and each structure from the orbits within each group. In Fig. 10, we display the surface densities of different components projected with the same orientation as the observed galaxy, in face-on and edge-on views, respectively. The columns from left to right represent the surface densities of the entire galaxy, BP/X bar, classical bulge, nuclear disc, and main disc orbits, respectively. The white square indicates the spatial coverage of the TIMER data. The surface density of the entire galaxy built by all orbits matches the global surface density from observations well. The variation of the PA derived from the entire galaxy rebuilt by the orbits, as obtained from the



Figure 10. The surface density of different orbital structures decomposed from our best-fitting model constrained by TIMER data. The rows from top to bottom are projections with the observed galaxy orientation (top), face-on (middle), and edge-on (bottom) views. Columns from left to right show the reconstructed surface densities of the whole galaxy, the BP/X-shaped bar, the classical bulge, the nuclear disc, and the extended main disc, respectively. The white dashed squares in the first row indicate the data coverage. The luminosity fraction of each structure within the data coverage is tagged in the top row (in per cent). The scale radius of the nuclear disc component is $R_s = 5.6$ arcsec.

ellipse fitting, is presented in Fig. A7, and it is consistent with the PA variation for the GALFIT model and the S4G image shown in Fig. 2.

As shown in Fig. 10, the bar exhibits BP/X shaped structure when viewing edge-on, the classic bulge is mostly round and might still be mixed with some flat discy orbits, the nuclear disc is elongated perpendicular to the bar when seen face-on and mostly flat when seen edge-on, the main disc is flat. Our results confirm that NGC 4371 has a BP/X bar that co-exists with a classical bulge, contributing nearly similar fractions to the total surface density. The BP/X-shaped structure is not directly discernible in the 2D image of this galaxy, and note that, our projected 3D density model constructed in Section 3.1 lacks a BP/X-shaped structure where the bar is nearly prolate. Although a BP/X-shaped structure is not explicitly included in the gravitational potential, our model can still support orbits constructing the BP/X-shaped structure, and the kinematic constraints pick up the orbits contributed to the BP/X structure (see Figs 6 and 11). This capability is a key aspect of our model, as we discussed in detail in our previous works Tahmasebzadeh et al. (2021, 2022).

We then quantify the luminosity fraction of different structural components in the best-fitting model. The contributions of the four components to the total luminosity of the galaxy are 24.32 per cent BP/X bar, 15.36 per cent classical bulge, 17.29 per cent nuclear disc, and 43.02 per cent main disc. When considering only the region within the data coverage, these fractions change to 34.92 per cent BP/X bar, 25.69 per cent classical bulge, 30.23 per cent nuclear disc, and 9.16 per cent main disc, respectively. Note that, we have flexibility in separating the classical bulge and inner disc. If we choose a difference of 0.1 in the λ_z cut-off, the classical bulge and nuclear disc fractions will change by 2 per cent to 5 per cent.

In our model, the radially end-to-end separation of the X-shaped structure is approximately half of the bar length. This proportion is consistent with what is measured for the Galactic bar (Li & Shen 2012; Portail et al. 2017). A barlens structure was reported in NGC 4371 (Buta et al. 2015); however, we show that it is actually a nuclear disc in the central regions in agreement with Gadotti et al. (2015), and it does not contribute to the BP/X structure. Erwin et al. (2015) employed 1D photometric decomposition to estimate the stellar mass ratio of the classical bulge to the total galaxy, finding it to be less than 10 per cent for NGC 4371, which is lower than the fraction of the classical bulge we found dynamically.

4.4.2.2 Kinematics of each structure We investigate the kinematic properties associated with these structures to validate the accuracy of our orbital decomposition. We show the LOS velocity (the first and second rows) and velocity dispersion (the third and fourth rows) of each component in Fig. 11, by projecting them along two viewing angles: as the galaxy was observed and edge-on. In the maps projected as observed, the inner region of the BP/X bar shows the strongest regular rotation, which is caused by stellar motion through an X-like (or a ∞ -like) path along the bar and moving inward/outward from the disc plane. The total velocity dispersion map displays two high-dispersion lobes in the vertical direction. This enhancement is caused by a combination of the BP/X structure and the classical bulge. In the BX/P bar, velocity dispersion increases in two wings of the X-shaped structure. In the map projected edgeon, the BP/X bar has moderate rotation, creating a relatively high dispersion in an X-shaped area. The classical bulge, characterized by its dynamically hot nature, contributes to an enhanced velocity dispersion at the centre. A weak counter-rotating motion in the



Figure 11. The kinematic maps of different structures decomposed from our best-fitting model. The first and the second rows show the velocity maps of the galaxy projected as observed and edge-on views. The third and fourth rows show the velocity dispersion maps of the galaxy projected as observed and edge-on views. Columns from left to right show the reconstructed kinematic maps of the whole galaxy, the BP/X-shaped bar, the classical bulge, the nuclear disc, and the extended main disc, respectively. The white dashed squares in the first column indicate the data coverage.

central region of the classical bulge is due to orbits with $\lambda_z < 0$. In contrast, the nuclear and main discs demonstrate strong regular rotation coupled with low dispersion.

4.4.2.3 The 1D surface brightness profiles We extract the surface brightness profile along the major axis of the galaxy on the projected 2D observational plane for each component and compare the results of our dynamical decomposition and that of the photometric decomposition based on the S⁴G image in Fig. 12 (note that, they are not necessary to be comparable, we do not impose any of these photometric decomposed components in our dynamical model rather than the global parametrized surface brightness of the galaxy). The surface brightness profiles of our four dynamically decomposed components, BP/X bar, classic bulge, nuclear disc, and main disc, can be roughly compared with the four photometrically decomposed components, the Sérsic bar, central compact component, nuclear disc, and main disc. The major difference is that the dynamical main disc exhibits a luminosity in the inner regions much lower than the inward extrapolation of an exponential profile from the outer regions. Thus, in the inner regions, the dynamical main disc contributes less luminosity than the photometric exponential main disc; instead, the dynamical BP/X bar and the classic bulge contribute more luminosity than the corresponding Sérsic bar and central compact component from photometric decomposition. The main discs defined more physically from stellar dynamics or stellar populations are not necessarily exponential but have lower luminosity in the inner regions; this has been noted in both simulations and observations (Zhu et al. 2018b; Breda, Papaderos & Gomes 2020; Ding et al. 2023), and is consistent with our results here. At the same time, the dynamical nuclear disc closely approximates an exponential distribution consistent with the photometric decomposition results. The scale radius of the nuclear disc, determined by dynamical decomposition, is $R_s = 5.6$ arcsec, which is similar to 5.2 arcsec obtained from photometric decomposition.

5 CONCLUSIONS

We apply the triaxial Schwarzschild barred model presented in Tahmasebzadeh et al. (2022) to an S0 barred galaxy NGC 4371. The gravitational potential is adopted as the combination of a 3D stellar luminosity density multiplied by a constant stellar mass-to-light ratio, a spherical dark matter halo, and a fixed black hole. We use the 3D stellar luminosity density deprojected from a 2D photometric image, combining an axisymmetric disc, and a triaxial bar. We have five free parameters in the model: the stellar mass-to-light ratio M_*/L , dark matter halo mass log M_{200}/M_* , the inclination angle of the disc θ , the bar angle with respect to the intrinsic major axis of the disc φ , and the pattern speed Ω_p .



Figure 12. Comparison of dynamical structure decomposition with photometric decomposition. Solid lines are the 1D surface brightness profiles along the major axis for different orbital structures decomposed from our best-fitting model constrained by TIMER data. The dotted lines indicate those of the structures from the GALFIT decomposition of S^4G image. The shaded gray area represents regions beyond the kinematic data coverage.

We independently created two sets of models constrained by stellar kinematic data from the TIMER and ATLAS3D surveys. For both sets, our models match the observational data well, capturing the major properties of the bar as seen in the observed 2D images and the stellar kinematic maps. The main results are as follows:

(1) For the model using TIMER, we obtained $\theta = 60 \pm 2^{\circ}$, $\varphi = -12 \pm 1^{\circ}$, $M_*/L_{3.6um} = 1.05 \pm 0.05$, and DM virial mass $\log_{10}(M_{200}/M_*) = 2.4 \pm 0.5$. For the model based on ATLAS3D data, the best-fitting parameters are $\theta = 60 \pm 1^{\circ}$, $\varphi = -12 \pm 1^{\circ}$, $M_*/L_{3.6um} = 0.99 \pm 0.11$, and DM virial mass $\log_{10}(M_{200}/M_*) = 3.0 \pm 0.6$. The best-fitting parameters obtained from the two sets of models are generally consistent with each other, with larger uncertainty for ATLAS3D data with smaller data coverage than TIMER.

(2) We constrain the pattern speed of NGC 4371 to be $\Omega_p = 23.6 \pm 2.8 \text{ km s}^{-1} \text{ kpc}^{-1}$ using TIMER data, and $\Omega_p = 22.4 \pm 3.5 \text{ km s}^{-1} \text{ kpc}^{-1}$ using ATLAS3D data. The bar pattern speed can be effectively constrained by the Schwarzschild model using data that cover only the bar region, consistent with previous results from validation exercises with mock data (Tahmasebzadeh et al. 2022; Vasiliev & Valluri 2020; Dattathri et al. 2024). While it is necessary to have an observed velocity profile that covers both the bar and the disc outer region to accurately measure Ω_p with the traditional TW method. The TW method relies solely on information about the LOS velocity. In contrast, the orbit superposition method utilizes four linearly independent moments of the LOSVD, which tightly constrain the allowed orbits in the modelled region, reflecting the overall kinematics of the bar and thereby enabling more accurate estimates of the bar pattern speed.

(3) We determined the location of the co-rotation resonance, along with the bar pattern speed for each model. Using all our models within the 1σ confidence level, constrained by the TIMER data, we found that the mean bar co-rotation radius for NGC 4371 could be around 131 arcsec, with a standard deviation of 24 arcsec.

(4) We compute the dimensionless bar rotation parameter for each model considering the upper/lower limit of bar length of $R_{\text{bar}} = 69^{+6}_{-5}$ arcsec. In the best-fitting Schwarzschild model, the bar rotation parameter is $\mathcal{R} = R_{\text{cor}}/R_{\text{bar}} = 2.5 \pm 0.2$. However, we report the bar rotation parameter for the galaxy by calculating the mean and standard deviation of all possible bar rotation parameter values across all models within the one-sigma region that results in $\mathcal{R} = R_{\text{cor}}/R_{\text{bar}} = 1.88 \pm 0.37$. This indicates that NGC 4371 likely has a slow bar with a lower limit close to the borderline.

(5) We found a large amount of DM mass within the bar region. The fraction of dark matter to the total enclosed mass within the bar region (M_{DM}/M_{total}) is $\sim 0.51 \pm 0.06$. Our results align with cosmological simulations that suggest that fast bars are typically found in baryon-dominated discs. However, to confirm the correlation between the bar rotation parameter and the amount of dark matter in the disc region for real galaxies, it is necessary to apply our modelling to a large sample of barred galaxies.

(6) Using orbit classification, we dynamically decompose the galaxy into four components: BP/X bar, classical bulge, nuclear disc, and main disc, which contribute 34.92 per cent, 25.69 per cent, 30.23 per cent, and 9.16 per cent of the luminosity, respectively, within the MUSE data coverage of NGC 4371. We confirm that NGC 4371 has a BP/X bar and a classical bulge, consistent with kinematical features in velocity dispersion and h_4 maps (Gadotti et al. 2015). The previously reported barlens structure (Buta et al. 2015) may actually be a nuclear disc component built by rotation-dominated orbits, which does not contribute to the BP/X bulge or the classical bulge.

We illustrate that our barred Schwarzschild model can reproduce the kinematic properties of real barred galaxies. It is a powerful new tool for uncovering key properties of the barred galaxies, including the bar pattern speed and the internal BP/X-shaped orbital structure. This framework will be included as a module in the publicly available DYNAMITE page (Jethwa et al. 2020; Thater et al. 2022), which is a new implementation of the code by van den Bosch et al. (2008). This methodology will be applied to large samples of barred galaxies from different mass ranges and environments. It will help us to understand the formation and evolution of barred galaxies by investigating their pattern speed, BP/X structures, black holes, and dark halo masses.

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DATA AVAILABILITY

The raw and reduced MUSE data used in this work are publicly available at the ESO Science Archive Facility.

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APPENDIX: AXISYMMETRIC MODEL

We also considered a nearly axisymmetric model by fitting MGEs without twists ($\Delta \psi'_j = 0$) to the whole galaxy as shown in Fig. A1. Details of fitting are shown in Table A1. For the nearly axisymmetric model, we explore intrinsic shape parameters σ_j , p_j , and q_j that are more efficient than searching over the viewing angles. p_j and q_j are the intermediate-to-long and short-to-long axis rations of the Gaussian component *j*. The deprojection of an axisymmetric system cannot constrain the φ as it is irrelevant but has a finite axis-ratio between *y* and *x*. We allow for some degree of triaxiality of the galaxy by setting a non-unity *u*. The stationary axisymmetric model also has five free hyperparameters: M_*/L , q, p, u, and M_{200}/M_* . We use an interval of 0.01 for q, p, and u in the searching process.

We sample a set of tube orbits in x - z plane and box orbits from equipotential surfaces with zero velocity in the energy E, which are discretized by spherical angles of Θ and Φ . The number of start points of the box orbit library (E, Θ, Φ) is also $30 \times 15 \times 13$. To reduce the Poisson noise of the model, we consider the dithering number to be 3, so each orbital bundle contains 27 orbits with close starting points. **Table A1.** Details of MGE fit for the whole galaxy in axisymmetric limit. *j* is the number of each individual Gaussian for which, L_j is the central flux in the unit of $(L_{\odot}pc^{-2})$, σ'_j presents the size in the unit of (arcsec), q'_j indicates the flattening. $\Delta \psi'_i$ is twisted angle of Gaussian that is 0 in this case.

j	$L_j(L_{\odot}pc^{-2})$	$\sigma'_j(\operatorname{arcsec})$	q_j'	$\Delta \psi_j'(^\circ)$
1	50 033.382	1.35	0.74	0.0
2	5211.781	7.874	0.53	0.0
3	1104.101	11.242	0.95	0.0
4	616.274	19.265	0.95	0.0
5	209.024	54.744	0.53	0.0
6	38.302	109.426	0.53	0.0



Figure A1. The contours of the NGC 4371 S^4G image (grey), overplotted with contours of the best-fitting MGE (red) without twist between different Gaussians applied to the whole galaxy.

The best-fitting axisymmetric model and its comparison versus the best-fitting bar model is shown in Fig. A2. Fig. A3 shows the main differences in recovered properties between the axisymmetric model versus the bar model, including variation of the axial ratios (left panel), enclosed mass profiles (middle panel), and the stellar orbit distribution in the space of circularity λ_z versus time-averaged radius r (right panel). Although the axisymmetric model apparently matches the observation well, there are significant differences in the internal orbital structure and mass profile of the best-fitting axisymmetric model compared to the bar model. The improvement of the bar model is evident in the velocity residual map (third row, second column) when compared to the axisymmetric model (last row, second column). This improvement is attributed to the model's ability to capture the higher stellar velocities along the bar. While producing a higher stellar motion, the axisymmetric model prefers the bestfitting model with a larger inclination ($\theta = 75^{\circ}$), which leads to a significantly thicker disc as is shown in the left panel Fig. A3, q in the disc region is ~ 0.6 for axisymmetric model (green solid line) and ~ 0.4 for the bar model (green dashed line). Previous studies measured the inclination angle for this galaxy to be ~ 60 (Erwin et al. 2008; Gadotti et al. 2015) similar to our bar model. The parameter grid searched for this axisymmetric model is shown in Fig. A6.



Figure A2. Comparison of the best-fitting bar model obtained with $\theta = 59^{\circ}$ (second row) versus the best-fitting axisymmetric model (fourth row) obtained with $\theta = 75^{\circ}$ using the TIMER data (first row). The third and fifth rows show the residual of each model. The surface density and its residuals are presented in the first column, for TIMER data, we use the flux of the IFU data cube and its contours, while for the bar and axisymmetric models, we used the MGEs surface density and its counters of the best-fitting models.



Figure A3. Left panel: variation of the axial ratios q = c/a (green), p = b/a (blue), and triaxial parameter $T = (1 - p^2)/(1 - q^2)$ (red) for the best-fitting axisymmetric model (solid lines) compared with the bar model (dashed lines) using TIMER. Middle panel: mass profiles of the best-fitting axisymmetric model (solid lines) compared with the bar model (dashed lines). The red, blue, and black curves represent the stellar mass, dark matter mass, and total mass, respectively. The shaded regions indicate the 1 σ uncertainty from our axisymmetric models. The right panel represents the stellar orbit distribution in the space of circularity λ_z versus time-averaged radius *r* for the best-fitting axisymmetric model with TIMER. The vertical dashed lines indicate the kinematics data extent.





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Figure A4. The parameter grid we explored for the bar model fitting using TIMER data. The five hyperparameters are stellar mass-to-light ratio $M_*/L_{3.6um}$ in unit of M_{\odot}/L_{\odot} , dark matter halo mass $\log M_{200}/M_*$, the inclination angle of the disc θ , the bar angle with respect to the major axis of the disc φ , and the pattern speed Ω_p in units of km s⁻¹ kpc⁻¹. Each point is one model colour-coded according to their χ^2 values shown in the colour bar. The points with $(\chi^2 - \chi^2_{min})/\sqrt{2n_{GH}N_{kin}} < 1$ indicate models within 1σ confidence level. The black crosses indicate the best-fitting model.



Figure A5. Same as Fig. A4 but for the bar model fitting using ATLAS3D data.



Figure A6. Same as Fig. A4 but for the axisymmetric model fitting using TIMER data. For the axisymmetric modeling, we use intrinsic parameters of q, p, and u instead of viewing angles of θ , φ , and ψ .



Figure A7. The variation of PA in the best-fitting Schwarzschild model was extracted using ellipse fitting, similar to the method applied in Fig. 2 for the GALFIT model and the S4G image, and overall is consistent with them.

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