Measuring bar pattern speeds from single simulation snapshots

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ABSTRACT
We describe methods to measure simultaneously the orientation angle $\psi$ and pattern speed $\Omega$ from single snapshots of simulated barred galaxies. Unlike previous attempts, our approach is unbiased, precise, and consistent in the sense that $\psi = \int \Omega \, dt$. It can be extended to obtain the rate and axis of rotation, i.e. the vector $\Omega$. We provide computer code implementing our method.

Key words: methods: numerical – galaxies: kinematics and dynamics – galaxies: structure.

1 INTRODUCTION
About two-thirds of spiral galaxies in the local Universe host a central stellar bar (Eskridge et al. 2000; Menéndez-Delmestre et al. 2007; Sheth et al. 2008; Masters et al. 2011; Cheung et al. 2013; Erwin 2018). Such bars are thought to rotate almost rigidly with an angular frequency or pattern speed $\Omega$. In simulations, $\Omega$ is commonly found to slowly decrease with time, as angular momentum is transferred from the bar to the dark halo, which is generally accompanied by a growth in bar strength and length (Sellwood 1980; Weinberg 1985; Little & Carlberg 1991; Debatista & Sellwood 1998; Athanassoula 2003). The rate of slowdown depends on the mass and structure of the bar, and on the balance between the angular momentum absorbed by the halo and that surrendered by gas driven into the galactic centre by the bar. Recently, Chiba, Friske & Schönrich (2021) found in the stellar kinematics of the solar neighbourhood evidence for the slowdown of the Milky Way bar.

Overlaying this continuous slowdown, bar pattern speeds (along with other bar parameters, see Wu, Pfenniger & Taam (2016)) are subject to short-term oscillations, e.g. by bar–spiral interactions (Wu, Pfenniger & Taam 2018; Hilmi et al. 2020).

A dimensionless parameter for the rotation of a bar is the ratio $\mathcal{R} = R_{CR}/R_{bar}$ between the radius $R_{CR}$ at which a star on a circular orbit corotates with the bar and the bar’s actual size $R_{bar}$. The confinement of most bar-supporting orbits to $R < R_{CR}$ sets the theoretical limit $\mathcal{R} \geq 1$. Observational determinations of both $R_{CR}$ and $R_{bar}$ are plagued with difficulties and systematic uncertainties, but largely suggest that bars rotate nearly as fast as possible, i.e. $\mathcal{R} \lesssim 1.4$ (Corsini 2011; Aguerri et al. 2015; Guo et al. 2019). Simulations of galaxy formation, on the other hand, tend to predict bars to be slower (Algorry et al. 2017; Peschken & Łokas 2019) or shorter (Frankel et al. 2022, using the IllustrisTNG simulation). Fragkoudi et al. (2021) suggest that this tension lessens when increasing the resolution of the models (in other words, the models may not yet be converged on $\mathcal{R}$).

A deeper understanding of bar slowdown, pattern speed oscillations, expected distribution of $\mathcal{R}$, and other bar rotation-related topics all depends on accurate measurements of $\Omega(t)$ from simulations. In simulations with high output cadence, the most common method is to derive the bar angle $\psi$ from an $m = 2$ Fourier analysis (Sellwood & Athanassoula 1986) of consecutive snapshots and calculate the pattern speed as finite difference:

$$\Omega \approx \Omega_{FD} \equiv \frac{\Delta \psi}{\Delta t}. \quad (1)$$

Since $\psi$ is $\pi$-periodic, this simple method requires $\Omega \Delta t \ll \pi$ to unambiguously identify $\Delta \psi$, and is therefore not viable for simulations with long output intervals $\Delta t$ (or if $\Omega$ is required on the fly during a simulation). This is the typical situation for large cosmological simulations, when data volume limits the output frequency.

In this situation, Peschken & Łokas (2019) and Fragkoudi et al. (2021) applied the Tremaine & Weinberg (1984) method for determining $\Omega$ of external galaxies from line-of-sight velocities. They report an accuracy of ~10 per cent and ~5 km s$^{-1}$ kpc$^{-1}$, respectively, depending also on the adopted viewing angle. Applying the Tremaine–Weinberg method to simulations may be justified for direct comparison to observations, but is certainly not ideal. This is because it relies on the assumption of stationarity of the pattern (which is generally not satisfied as mentioned above) and utilizes only one of three Cartesian velocity components.

The orientation $\psi$ of simulated bars is well measured from the particle positions $x_i$ (and their masses) as phase of the $m = 2$ Fourier component. Since the particles move, $\psi$ is an implicit function of time $t$, which can be differentiated to obtain

$$\Omega = \frac{d\psi}{dt} = \sum_i \frac{\partial \psi}{\partial x_i} \frac{dx_i}{dt}. \quad (2)$$

Crucially, $\psi$ depends on $x_i$ not only through the azimuth $\varphi$, which enters the Fourier analysis, but also through the spatial selection, usually from an annulus. Neglecting this dependence (Wu et al. 2018) ignores the difference between the particle sets from which $\psi$ is measured at $t$ and $t + \Delta t$ and results in systematic errors of 5–25 per cent (see Fig. 5).
When using radial bins (annuli), their sharp boundaries generate divergent \( \partial \psi / \partial R \), which cannot be evaluated for particle systems. Instead, Frankel et al. (2022) estimated the resulting dependence on the particle velocities in a not reproducibly specified way and report that the accuracy for \( \Omega \) seems to be \( \sim 10 \) per cent. However, such a treatment cannot be consistent in the sense that \( \psi \) and \( \Omega \) measured from the particles satisfy \( \psi = \int \Omega \, dt \).

Therefore, for a consistent measurement of \( \Omega \), annuli with sharp boundaries must be avoided in favour of smoothly varying window functions. This is analogous to the way local properties are estimated in smoothed particle hydrodynamics (SPH).

We also show that the \( m = 2 \) Fourier method for identifying \( \psi \) is equivalent to obtaining \( \psi \) as the direction of the eigenvectors of the moment of inertia. This insight provides a way to measure the particular axis. This generalized method may be suitable to measure the particle velocities in a not reproducibly specified way and report the statistical uncertainty \( \sigma_\Omega \). This approach trivially generalizes to more general windows \( W(R, z) \).

This paper is organized as follows. Section 2 contains the derivation of the Fourier and moment-of-inertia methods, Section 3 describes tests of the method on a suite of \( N \)-body simulations, and Section 4 summarizes and discusses our findings.

2 METHODS

Before measuring a pattern speed, the centre of rotation must be known. Here, we do not discuss finding the centre, as various good methods have been published (e.g. the shrinking sphere; Power et al. 2003), but note that also the rate of change of the centre position (which may differ from the central velocity) must be known with uncertainty well below the velocity dispersion. In the remainder, \( x \) and \( \rho \) denote position and velocity relative to that centre.

We begin by assuming that rotation is around the \( z \)-axis, i.e. in azimuth \( \varphi \), and that the density \( \rho \) is stationary in a frame rotating with angular rate \( \Omega(t) \), but will later relax both assumptions. With these assumptions, \( \rho(R, z, \varphi, t) = f(R, z, \varphi - \psi) \) with some function \( f \) and the instantaneous orientation

\[
\psi(t) = \int \Omega(t) \, dt
\]

of the rotating frame, such that \( \partial \rho / \partial t = -\Omega \partial \rho / \partial \varphi \). Combining this with the continuity equation \( \partial \rho / \partial t + \nabla \cdot (\rho \vec{v}) = 0 \), we find

\[
\frac{\partial \rho}{\partial \psi} = \nabla \cdot (\rho \vec{v}),
\]

where \( \vec{v}(x) \) is the mean (streaming) velocity. We exploit equation (4) by multiplying both sides by a weight function \( w(x) \) and integrating over all space to find

\[
\Omega \int d^3 x \, \rho \frac{\partial w}{\partial \varphi} = \int d^3 x \, \rho \vec{v} \cdot \nabla w,
\]

where we have integrated each side by parts to shift the derivatives onto \( w \). The Tremaine & Weinberg (1984) method is obtained from equation (5) by weighing with the Heaviside function, \( w = \Theta(y - y_0) \), which reduces the velocity term to its \( y \) component and the integrals to a slit at \( y = y_0 \).

2.1 Fourier methods

A natural choice for the weight function is \( w(x) = W(R) e^{-i \varphi} \), where \( m \) is an azimuthal wavenumber and \( W \geq 0 \) some window function. Equation (5) then yields

\[
\Omega = \frac{\int d^3 x \, \rho \left[ W \varphi + \frac{i}{m} \bar{w} \left( \partial W / \partial R \right) \right] e^{-i \varphi}}{\int d^3 x \, \rho \ W e^{-i \varphi}}.
\]

Since this equation was derived under the assumption that \( \rho \) is stationary in the rotating frame, the right-hand side is real valued, provided this assumption is satisfied. However, since bars often evolve, we now relax this assumption, when the right-hand side of equation (6) generally includes an imaginary part. In this case, the real part remains the correct answer for \( \Omega \). To show this, we define the window-averaged surface density \( \Sigma(\varphi, t) \equiv \int_0^\infty W(R) \, dR \, dz \) and take its azimuthal Fourier transform

\[
\tilde{\Sigma}_m(t) = \frac{1}{2\pi} \int d^3 x \, W(R) \rho(x, t) e^{-i \varphi}
\]

with time derivative

\[
\frac{d \tilde{\Sigma}_m}{dt} = \frac{1}{2\pi} \int d^3 x \, \rho \left[ -im \tilde{\varphi} W + \tilde{w} \left( \partial W / \partial R \right) \right] e^{-i \varphi},
\]

where we used the continuity equation to eliminate \( \partial \rho / \partial t \). Expressing \( \tilde{\Sigma}_m \) in polar form with amplitude \( \Sigma_m \) and phase \( \psi_m \), and identifying \( \Omega = \psi_m \) gives with equation (8)

\[
\Omega + \frac{i}{m} \frac{\Sigma_m}{\tilde{\Sigma}_m} = \frac{1}{2\pi} \int d^3 x \, \rho \left[ \tilde{\varphi} W + \frac{1}{m} \tilde{w} \left( \partial W / \partial R \right) \right] e^{-i \varphi}
\]

right-hand side identical to equation (6). Thus, only the real part of these right-hand sides measures a pattern speed, namely that of the azimuthal \( m \)-wave in the window \( W \), while the imaginary part is related to the rate of change in wave amplitude \( \Sigma_m \).

For \( N \)-body models, the distribution function is a sum of \( \delta \)-peaks, resulting in the substitutions \( \int d^3 x \rho \rightarrow \sum \mu_i \), \( \rho \rightarrow \sum \mu_i \vec{v}_i \), with particle masses \( \mu_i \), in equation (9), which becomes

\[
\Omega + \frac{i}{m} \frac{\Sigma_m}{\tilde{\Sigma}_m} = \sum \frac{\mu_i \vec{v}_i \left( \partial W / \partial R \right) \bar{W}}{\sum \mu_i \bar{W} e^{-i \varphi}} e^{-i \varphi}
\]

An equivalent expression using only real-valued arithmetic is provided in Appendix A, which also specifies the estimation of the statistical uncertainty \( \sigma_\Omega \). This approach trivially generalizes to more general windows \( W(R, z) \).

2.2 Moment-of-inertia methods

We define a generalized two-dimensional moment of inertia

\[
M = \int d^3 x \, \rho(x, t) W(R) \left( \hat{x} \hat{\varphi} + \hat{y} \hat{\varphi} \right) \text{ with } \hat{\varphi} = \frac{x \hat{\varphi} + y \hat{\varphi}}{R}.
\]

The symmetric matrix \( M \) has orthonormal eigenvectors

\[
e_1 = \begin{pmatrix} \cos \psi \\ \sin \psi \end{pmatrix}, \quad e_2 = \begin{pmatrix} -\sin \psi \\ \cos \psi \end{pmatrix}
\]

that rotate with the bar as \( e_1 = \Omega e_2 \) and \( e_2 = -\Omega e_1 \) with \( \Omega = \dot{\psi} \). Moreover, with \( \lambda_i \) the eigenvalue associated with \( e_i \), \( M \) can be written

\[
M = \sum \lambda_i e_i \otimes e_i,
\]

where \( \otimes \) denotes the usual outer (or tensor) product. Differentiating with respect to time gives

\[
M = \sum \dot{\lambda}_i e_i \otimes e_i + \Omega (\lambda_1 - \lambda_2) (e_1 \otimes e_2 + e_2 \otimes e_1).
\]
Multiplying from left by \( e_1 \) and from right by \( e_2 \) and using their orthonormality, we find
\[
\Omega = \frac{e_1 \cdot M \cdot e_2}{\lambda_1 - \lambda_2}.
\] (15)

This relation for \( \Omega \) is in fact identical to equation (9) for \( m = 2 \), as one can verify by expressing \( \lambda_i \) and \( e_i \) in terms of the matrix elements \( M_{ij} \) and exploiting \( \cos 2\varphi = \hat{x}^2 - \hat{y}^2, \sin 2\varphi = 2\hat{x} \hat{y} \). For \( W = R^2, M \) is the moment of inertia of the whole system, when equation (15) agrees with equation (12) of Wu et al. (2018), who did not notice the close relation to the \( m = 2 \) Fourier method.

The relative magnitude is the rotation rate. We again have equation (15) with \( \Omega = \frac{\lambda_1 - \lambda_2}{\lambda_1 + \lambda_2} \).

\[
A_2 = \frac{\sum_i 2}{\sum_0} \left[ \int d^3x \rho W e^{-2\varphi} \right] = \frac{\lambda_1 - \lambda_2}{\lambda_1 + \lambda_2}.
\] (16)

The concept of rotation of the eigenvectors of a generalized moment of inertia easily extends to three dimensions. To this end, we supplement the matrix \( M \) with the \( z \)-coordinate and take the window to be a function of spherical radius \( r \):
\[
\mathbf{M} = \int d^3x \rho(x, t) W(r) \hat{x} \otimes \hat{x} \quad \text{with} \quad r = |x| \quad \text{and} \quad \hat{x} = x/r
\] (17)
with time derivative
\[
\dot{\mathbf{M}} = \int d^3x \rho \left[ \mathbf{v} \cdot \hat{x} \left( \frac{dW}{dr} - \frac{W}{r} \right) \hat{x} \otimes \hat{x} + \frac{W}{r} (\mathbf{v} \otimes \hat{x} + \hat{x} \otimes \mathbf{v}) \right].
\] (18)

Again, the discrete forms for \( \mathbf{M} \) and \( \dot{\mathbf{M}} \) are easily obtained via the substitutions \( \int d^3x \rho \rightarrow \sum_i \mu_i \) and \( \int d^3x \rho \mathbf{v} \rightarrow \sum_i \mu_i \mathbf{v}_i \). The three orthonormal eigenvectors of \( \dot{\mathbf{M}} \) form a triad rotating as
\[
\dot{e}_i = \Omega \times e_i,
\] (19)
where the vector \( \Omega \) points along the axis of rotation, while its magnitude is the rotation rate. We again have equation (13) with time derivative
\[
\mathbf{M} = \sum_i \lambda_i e_i \otimes e_i + \sum_0 \Omega \cdot e_i (\lambda_2 - \lambda_3) (e_2 \otimes e_3 + e_3 \otimes e_2),
\] (20)
when multiplying from left and right by \( e_i \) and \( e_j \), we find
\[
\Omega = \frac{e_2 \cdot M \cdot e_3}{\lambda_2 - \lambda_3} + \frac{e_3 \cdot M \cdot e_1}{\lambda_3 - \lambda_1} + \frac{e_1 \cdot M \cdot e_2}{\lambda_1 - \lambda_2}.
\] (21)

The component of \( \Omega \) in direction \( e_i \) is only well defined if the eigenvectors associated with the other two eigenvectors are distinct. Geometrically, this simply states that if a system is axisymmetric, the pattern speed for rotation around the symmetry axis is ill defined.

One may define a relative \( m = 2 \) Fourier amplitude with respect to each eigenvector as \( \lambda_i \). These are given in terms of the eigenvalues via (assuming the order \( \lambda_1 \geq \lambda_2 \geq \lambda_3 \))
\[
A_{2,1} = \frac{\lambda_2 - \lambda_3}{\lambda_2 + \lambda_3}, \quad A_{2,2} = \frac{\lambda_1 - \lambda_3}{\lambda_1 + \lambda_3}, \quad A_{2,3} = \frac{\lambda_1 - \lambda_2}{\lambda_1 + \lambda_2},
\] (22)
when \( A_{2,3} \) agrees with the azimuthal Fourier result (16) for \( \dot{z} = e_3 \), as expected. The time derivatives of these relative amplitudes can be obtained from the rates of change of the eigenvalues
\[
\dot{\lambda}_i = e_i \cdot M \cdot \dot{e}_i
\] (23)
and provide a measure of the non-stationarity of the pattern.

2.3 Particle selection and systematic errors

When selecting particles for measuring \( \Omega \), for example those in the bar region, one must distinguish the instantaneous selection of particles currently in the bar region from the evolving selection of particles that are in the bar region at any given time. Since the bar orientation \( \psi \) is measured from all particles inside the bar region at the time of measurement, its rate of change \( \dot{\Omega} \) must be measured from the evolving selection to be consistent.

In our method, the bar region is selected via the window function \( W(R) \) and the term involving \( \dot{\psi}(\partial W/\partial R) \) in equation (9) accounts for the difference between the instantaneous and evolving selections. We demonstrate this by considering a top-hat window, when \( W = 1 \) for \( R_0 \leq R \leq R_1 \) and zero otherwise, representing an annulus. Then \( \partial W/\partial R = \delta(R - R_0) - \delta(R - R_1) \) and equation (9) gives for \( m = 2 \)
\[
\dot{\Omega} = \frac{\int d\varphi \cos 2\varphi \int_{R_0}^{R_1} dR(R\dot{\psi}R^2 - \frac{1}{2} \sin 2\varphi [R^2 \ddot{\psi} - R\dot{\psi} R]^2/R_0)}{\int d\varphi \cos 2\varphi \int_{R_0}^{R_1} dR R^2 \ddot{\psi} R},
\] (24)
where the mean velocity is averaged over \( z \) and the coordinate system aligned with the bar. The second term in the numerator accounts for the flux of particles into and out of the window. As we show in Fig. 1, this flux is generally non-zero, owing to the motion of stars along the bar, and has a sin \( 2\varphi \) pattern. Previous authors have either omitted this term (Wu et al. 2018) and thereby implemented the wrong instantaneous selection, or have estimated it only approximately (Frankel et al. 2022), both are prone to systematic errors for \( \Omega \).

For particle systems, this term cannot be easily evaluated (since the chance to find a particle at \( \varphi \) is generally non-zero, owing to the motion of stars along the bar, and has a sin \( 2\varphi \) pattern. Previous authors have either omitted this term (Wu et al. 2018) and thereby implemented the wrong instantaneous selection, or have estimated it only approximately (Frankel et al. 2022), both are prone to systematic errors for \( \Omega \).
A similar issue occurs if the bar region changes on account of bar evolution. In this case, the window function $W = W(R, t)$ and its time derivative must also be taken into account for a strict implement of the evolving selection. However, since the pattern speed should not differ between different parts of the bar and because determining the bar region with differentiable edges $R_0(t)$ is non-trivial, we neglect these terms but do not find significant deviations from $\psi = \int \Omega \, dr$.

2.4 Measuring the bar pattern speed

The window for measuring the bar pattern speed should contain most of the bar and not much else. We identify the bar region $[R_0, R_1]$ as a continuous range of radial bins with large $\Lambda_z$ and similar $\psi_2$, see Appendix B for details. The bar orientation $\psi$ and pattern speed $\Omega$ is measured using the $m = 2$ Fourier method from all star and gas particles in the bar region using the window function (25) with $R_m$ taken to be the median radius in the bar region.

Owed to the high-velocity dispersion for motion along the bar, the flux term $\rho v_R (\partial W/\partial R)^{\text{imp}}$ in the numerator of equation (9) contributes non-negligibly to the statistical uncertainty $\sigma_\Omega$. In order to reduce this contribution, the window (25) is near-maximally smooth. For less smooth window functions (with a larger central part of $W = 1$ and steeper $\partial W/\partial R$ at the edges), we find larger $\sigma_\Omega$ and larger deviations to the finite-difference estimate for $\Omega$.

3 TESTING ON N-BODY MODELS

We test our methods for identifying the bar and measuring its pattern speed on a set of N-body models previously used by Semczuk et al. (2022). We briefly describe these models, before presenting the test results.

3.1 The N-body models

Our models are generated using the ‘growing-disc’ technique or Aumer & Schönrich (2015), by which star particles are continuously added to a running N-body model. We follow the star formation recipe of Aumer & Schönrich by placing stars on near circular orbits with velocity dispersion of 10 km s$^{-1}$. The total star formation rate is initially 16.7 $M_\odot$ yr$^{-1}$ and decays exponentially with a decay time-scale of 8 Gyr. The spatial distribution of new stars follows an exponential disc profile with scale length growing from 0.6 to 3 kpc in 10 Gyr (following equation 1 of Schönrich & McMillan 2017), to emulate an inside-out growth. Once the bar is formed, as inferred from the on-fly $m = 2$ Fourier analysis, star formation is halted at 0.05 $< R/R_{\text{CR}} < 0.7$, where $R_{\text{CR}}$ is the corotation radius.

This is done to mimic gas depletion and subsequent suppression of star formation in the bar and causes the star particles to satisfy the continuity equation (which underlies the methods of Section 2) in the bar region.

We implemented in the growing-disc technique with our code GRIFFIN that uses the fast multipole method as force solver (Dehnen 2000, 2014). The growing disc is embedded in a dark matter whose initial distribution follows a spherical Dehnen & McLaughlin (2005) profile with scale radius $r_s = 31.25$ kpc, smoothly truncated at 10 kpc and with circular velocity at $r_s$ of to 126.6 km s$^{-1}$. Dark matter particles are seeded from an ergodic distribution function.

We used and tested our method for measuring $\Omega$ on many such growing-disc N-body models with different values for the various parameters of dark halo and star formation. However, here we present only two illustrative typical models, which we call ‘fiducial’ and ‘hot’.

Fig. 2 shows snapshots of surface density for the fiducial model, which is identical to the fiducial model (Fig. 2) at $t \leq 4$ Gyr, but has no star formation thereafter.

Fig. 3 shows snapshots of the stellar surface density from the hot growing-disc model, which is identical to the fiducial model (Fig. 2) at $t \leq 4$ Gyr, but has no star formation thereafter.

Figure 2. Snapshots of the stellar surface density for our fiducial growing-disc model. Note the growth of disc and bar as well as the relative strength of spiral arms.

Figure 3. Snapshots of the stellar surface density from the hot growing-disc model, which is identical to the fiducial model (Fig. 2) at $t \leq 4$ Gyr, but has no star formation thereafter.

For less smooth window functions (with a larger central part of $W = 1$ and steeper $\partial W/\partial R$ at the edges), we find larger $\sigma_\Omega$ and larger deviations to the finite-difference estimate for $\Omega$.
Fig. 4 shows the time evolution of the relative $m = 2$ Fourier amplitude measured from all particles inside 5 kpc (top) and of the bar length and the corotation radius for the fiducial and hot N-body models.

Our estimate for $\Omega$ (red) fluctuates around $\Omega_{FD}$ (on top of its variations), but shows no significant systematic bias – the mean deviation of 0.13 per cent is insignificant. The amplitude of the fluctuations of 1.34 per cent is about twice the mean relative uncertainty $\langle \sigma_2 / \Omega_{FD} \rangle = 0.75$ per cent due to particle shot noise ($\sigma_2$ is shown as error bars in the top panel of Fig. 5). This suggests that the fluctuations of the instantaneously measured $\Omega$ on time-scales $< \Delta t = 40$ Myr are only partly due to shot noise, but mostly reflect true variations of $\Omega$ on these time-scales. On time-scales much shorter than the shot-noise correlation time, the instantaneously measured $\Omega$ and $\Omega_{FD}$ do indeed agree (if the bar region is kept fixed), as we verify using a simulation with output interval $10^4$ times shorter (not plotted).

If we decrease the number of particles used to estimate $\Omega$ by a factor of 6, the standard deviation of the relative deviations to $\Omega_{FD}$ is 1.94 per cent, not quite twice the value reported in Fig. 5, and the measured statistical uncertainty $\sigma_2$ rises to 1.75 per cent (from 0.75 per cent) close to the expected rise by $\sqrt{6} \approx 2.45$.

In Fig. 6, we show pattern speeds measured for the hot model (for $t > 4$ Gyr, when it differs from the fiducial simulation). Our method is unbiased, while the naive approach incurs a $\sim 14$ per cent bias, corroborating our findings for the fiducial model. We also clearly see...
that the pattern-speed oscillations at \( t = 7\)–8 Gyr have substantially larger amplitude for \( \Omega \) measured instantaneously than its sliding average, as provided by the finite difference.

In Fig. 7, we use the naive method with a bar region without central hole, i.e. \( R_0 = 0 \). This has the advantage that the bar region no longer has an inner boundary such that the neglected flux is limited to that through the outer boundary. We find that the bias of the method has changed sign and is smaller compared to the situation with inner boundary (in Fig. 5). However, we also see that the method has become very noisy. Upon inspection of equation (10), this is not surprising, since \( \dot{\psi} = v_\psi / R \) can become arbitrarily large for particles near \( R = 0 \). We conclude from this exercise that for the purpose of calculating \( \Omega \), the bar region must exclude the origin and therefore necessarily have an inner boundary, though our smooth-window approach suffers much less from this problem, as it weights particles at \( R \sim R_0 \) only very little.

4 DISCUSSION

Rotating galactic bars are always dominated by their azimuthal \( m = 2 \) Fourier mode. Consequently, their orientation \( \psi \) is well estimated by the phase of the \( m = 2 \) Fourier transform measured in the bar region. The bar pattern speed therefore is naturally defined as the time derivative of that phase.

We show in Section 2.3 and demonstrate in Section 3.2 that in order to compute \( \Omega = d\psi / dt \) without bias, one must account for the net particle flux into the bar region. Previous implementations of this approach have either overlooked this flux term (Wu et al. 2018), resulting in large systematic errors, or have estimated it only approximately (Frankel et al. 2022), when the measured \( \dot{\psi} \) and \( \dot{\Omega} \) are inconsistent, i.e. do not in general satisfy \( \dot{\psi} = \int \Omega \, dt \). This flux term, which never vanishes, is most naturally determined in a consistent way by weighing the simulation particles with a function \( W(R) \) that smoothly drops to \( W = 0 \) outside the bar region.

When attempting to use the Tremaine–Weinberg method to separately measure pattern speeds for the inner and outer parts of an external galaxy, a similar flux term, which cannot be measured, occurs at their boundary. Such applications therefore are erroneous.

In Section 2.2, we show that the Fourier method is equivalent, modulo a radial weight function, to measuring \( \Omega \) as the rate at which the eigenvectors of the planar moment of inertia rotate. We also show that this method can be generalized to obtain the vector \( \vec{\Omega} \), the rate and axis of rotation, as that by which the triad of the principal axes of a 3D moment-of-inertia-like tensor rotates. We have not tested this generalization on 3-body models of galactic bars (for which the direction of rotation is unambiguous), but suggest it for measuring the tumbling rate and axis of spheroidal components, such as elliptical galaxies and dark matter haloes.

While defining the bar orientation \( \psi \) as phase of the \( m = 2 \) azimuthal Fourier mode appears natural and works well, alternative ideas are worth considering. One is to trace the phase of the azimuthal maximum, i.e. the ridge of the bar, which is more like what humans do when eye-ball the bar orientation. To find this maximum, some azimuthal smoothing is necessary using a smoothing kernel \( w(\Delta \phi) \), when only particles near the maximum contribute to the estimate. Such an approach is therefore likely to be more noisy, unless the azimuthal smoothing is maximally wide, like \( w = 1/2 [1 + \cos 2 \Delta \phi] \), when it reverts to the \( m = 2 \) Fourier method.

Can one determine \( \Omega = \dot{\psi} \) without any concept of the bar orientation \( \psi \)? This seems impossible, but Wu et al. (2018) also proposed a method that does, by estimating \( \Omega \) as the value for which the Jacobi integral \( J = E - \Omega L_z \) is conserved (giving \( \Omega = E / L_z \) with least-squares solution \( \Omega = \sum_i E_i L_{z,i} / \sum_i L_{z,i}^2 \)). Unfortunately, this clever approach has two serious problems. First, \( J \) is conserved only if the gravitational potential is stationary (not just the pattern as required by the Tremaine–Weinberg method) and \( \Omega \) is constant in time. Neither of these conditions is likely satisfied for realistic simulations and systematic errors are unavoidable. Secondly, this method requires knowledge of the time derivative of the gravitational potential for all particles, which is not usually computed by 3-body force solvers.

All these considerations strongly favour the \( m = 2 \) Fourier method, including its formulation as the moment-of-inertia method, over other contemporary approaches for measuring bar pattern speeds.

The task of determining \( \Omega \) for spiral structure is harder than that for bars, since spirals are weaker, evolve faster, and can contain patterns rotating at different rates. However, our method should in theory be adequate for measuring their pattern speeds, one for each azimuthal wavenumber \( m \) and (sufficiently resolved) radial range, as well as the time derivatives of the wave amplitudes. While we have not attempted or tested this so far, this is a promising idea warranting further investigation.

5 CONCLUSIONS

We provide the first unbiased and precise method for measuring the bar pattern speed \( \Omega \) from single simulation snapshots. This is valuable because time intervals between snapshots are typically too long for determining \( \Omega \) by following the bar rotation.

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DATA AVAILABILITY

Computer code in PYTHON for finding the bar region and estimating bar pattern speed from N-body data is publicly available at https://github.com/WalterDehnen/patternSpeed. The simulation data can be shared on reasonable request.

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APPENDIX A: REAL-VALUED FOURIER METHOD

We begin by noting that the azimuthal harmonics are best recursively computed from \( \cos \varphi = x/R, \sin \varphi = y/R \),

\[
\cos (m+1) \varphi = \cos m \varphi \cos \varphi - \sin m \varphi \sin \varphi, \tag{A1a}
\]

\[
\sin (m+1) \varphi = \cos m \varphi \sin \varphi + \sin m \varphi \cos \varphi, \tag{A1b}
\]

[in particular \( \cos 2 \varphi = (x^2 - y^2)/R^2 \) and \( \sin 2 \varphi = 2xy/R^2 \)], which is computationally much faster than calls to trigonometric functions.

Given the sums

\[
C_m = \sum_i \mu_i W(x_i) \cos m \varphi_i, \quad S_m = \sum_i \mu_i W(x_i) \sin m \varphi_i, \tag{A2}
\]

with \( \partial W/\partial \varphi = 0 \) and their time derivatives

\[
\dot{C}_m = \sum_i \mu_i [v_i \cdot \nabla W(x_i) \cos m \varphi_i - m \varphi_i W(x_i) \sin m \varphi_i], \tag{3a}
\]

\[
\dot{S}_m = \sum_i \mu_i [v_i \cdot \nabla W(x_i) \sin m \varphi_i + m \varphi_i W(x_i) \cos m \varphi_i], \tag{3b}
\]

the Fourier amplitude, phase, and their time derivatives are

\[
\Sigma_m = \sqrt{\dot{C}_m^2 + \dot{S}_m^2}, \quad \frac{\dot{C}_m}{\Sigma_m} = \frac{C_m + \dot{S}_m S_m}{C_m^2 + \dot{S}_m^2}, \tag{4a}
\]

\[
\psi_m = \frac{1}{m} \tan^{-1} \frac{S_m}{C_m}, \quad \dot{\psi}_m = \frac{m C_m S_m - \dot{S}_m \dot{C}_m}{m(C_m^2 + \dot{S}_m^2)}. \tag{4b}
\]

For the correct \( \Sigma_m \), the window function must be normalized to \( N_W = 2\pi \int W \, dR = 1 \) [or \( \Sigma_m \) from equation (A4a) divided by \( N_W \)].

Each of the terms in equations (A2) and (A3) is of the form \( \sum p_i = N(p) \), i.e. a sample mean. Hence, its variance can be estimated as

\[
\sigma^2 \approx \frac{N}{N-1} \left( \langle p_i \rangle - \langle p \rangle \right)^2, \tag{A5}
\]

and equivalently the co-variances between any two such quantities. From these, the co-variance matrix for the derived quantities in equations (A4) can be estimated via linear error propagation.

APPENDIX B: IDENTIFYING THE BAR REGION

We again assume that all positions and velocities are relative to the centre (determined before) and rotation is around the \( z \)-axis.

Our implementation first sorts particles in cylindrical radius \( R \) and assigns a number \( N_i \) of radial bins in \( R \) (annuli). The innermost bin starts at \( R_{\text{min}} \) and extends to \( R_{\text{max}} \) such that \( N_{\text{min}} \) particles are contained. Each subsequent bin starts at \( R_{\text{min}} \) equal to the next particle just outside the particle itself, and contains at least \( N_{\text{min}} \) particles, but more if \( R_{\text{max}} / R_{\text{min}} < 10^2 \) with parameter \( \Delta = 0.15 \) by default, though not exceeding a certain maximum \( N_{\text{max}} \). Next, we add \( N_1 \) - 1 intermittent bins that cover the radii between the medians of two adjacent primary bins. This gives \( N_2 = 2N_1 - 1 \) overlapping cylindrical bins. In each of these, we perform the \( m = 2 \) Fourier analysis of Appendix A to determine \( A_2 = \Sigma_m - 2/\Sigma_m = 0 \) and the phase \( \psi \). For this purpose, we employ the top-hat window, which is more efficient, slightly more accurate (the effective particle number is higher than with non-uniform weighting), and sufficient for unbiased estimates of \( \Sigma_m \) and \( \psi_m \).

If the maximum \( A_2 \) across all bins is below a threshold (0.2 by default), we do not attempt to identify a bar. Otherwise, we start by setting the bar region to the radial bin with maximum \( A_2 \) and extend it as follows. We consider the next inner and outer bins for extension, if their \( A_2 \) exceeds half the maximum. If both qualify, we take that which keeps the range \( \Delta \psi \) of phases covered by the bar region smallest. In this way, the bar region is extended until \( A_2 \) of the candidate bins is too small or \( \Delta \psi \) would exceed a certain width (typically 10°). The inner and outer edges of the bar region are then identified as \( R_0 \) and \( R_1 \) of, respectively, the inner- and outermost bin in the bar region.

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