

# Breaking the degeneracy between anisotropy and mass: the dark halo of the E0 galaxy NGC 6703

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## ABSTRACT

We have measured line-of-sight velocity profiles (VPs) in the E0 galaxy NGC 6703 out to  $2.6R_e$ . Comparing these with the VPs predicted from spherical distribution functions (DFs), we constrain the mass distribution and the anisotropy of the stellar orbits in this galaxy.

We have developed a non-parametric technique to determine the DF  $f(E, L^2)$  directly from the kinematic data. We test this technique on Monte Carlo simulated data with the spatial extent, sampling, and error bars of the NGC 6703 data. We find that smooth underlying DFs can be recovered to an rms accuracy of 12 per cent inside three times the radius of the last kinematic data point, and the anisotropy parameter  $\beta(r)$  can be recovered to an accuracy of 0.1, in a *known* potential. These uncertainties can be reduced with improved data.

By comparing such best-estimate, regularized models in different potentials, we can derive constraints on the mass distribution and anisotropy. Tests show that, with presently available data, an asymptotically constant halo circular velocity  $v_0$  can be determined with an accuracy of  $\pm \lesssim 50 \text{ km s}^{-1}$ . This formal range often includes high- $v_0$  models with implausibly large gradients across the data boundary. However, even with extremely high quality data some uncertainty on the detailed shape of the underlying circular velocity curve remains.

In the case of NGC 6703, we thus determine the true circular velocity at  $2.6R_e$  to be  $250 \pm 40 \text{ km s}^{-1}$  at 95 per cent confidence, corresponding to a total mass in NGC 6703 inside 78 arcsec ( $13.5 h_{50}^{-1} \text{ kpc}$ , where  $h_{50} \equiv H_0/50 \text{ km s}^{-1} \text{ Mpc}^{-1}$ ) of  $1.6\text{--}2.6 \times 10^{11} h_{50}^{-1} M_\odot$ . No model without dark matter will fit the data; however, a *maximum stellar mass* model in which the luminous component provides nearly all the mass in the centre will. In such a model, the total luminous mass inside 78 arcsec is  $9 \times 10^{10} M_\odot$  and the integrated *B*-band mass-to-light ratio out to this radius is  $Y_B = 5.3\text{--}10$ , corresponding to a rise from the centre by at least a factor of 1.6.

The anisotropy of the stellar distribution function in NGC 6703 changes from near-isotropic at the centre to slightly radially anisotropic ( $\beta = 0.3\text{--}0.4$  at 30 arcsec,  $\beta = 0.2\text{--}0.4$  at 60 arcsec) and is not well-constrained at the outer edge of the data, where  $\beta = -0.5$  to  $+0.4$ , depending on variations of the potential in the allowed range.

Our results suggest that also elliptical galaxies begin to be dominated by dark matter at radii of  $\sim 10 \text{ kpc}$ .

**Key words:** line: profiles – celestial mechanics, stellar dynamics – galaxies: elliptical and lenticular, cD – galaxies: individual: NGC 6703 – galaxies: kinematics and dynamics – dark matter.

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## 1 INTRODUCTION

Current cosmological models predict that, similarly to spiral galaxies, elliptical galaxies should be surrounded by dark matter haloes. The observational evidence for dark matter in ellipticals is still weak, however. In a few cases, masses have been determined from X-ray observations (e.g. Awaki et al. 1994; Kim & Fabbiano 1995) or H I ring velocities (Franx, van Gorkom & de Zeeuw 1994). In others it has been possible to rule out constant  $M/L$  ratios from extended velocity dispersion data (Saglia et al. 1993), absorption-line profile measurements (Carollo et al. 1995; Rix et al. 1997) or globular cluster or planetary nebula velocities (e.g. Arnaboldi et al. 1994; Grillmair et al. 1994). Gravitational lensing statistics (Maoz & Rix 1993) and individual image lens separations (Kochanek & Keeton 1997) favour models with extended dark matter haloes around ellipticals. Despite this, the detailed radial mass distribution in elliptical galaxies remains largely unknown. Similarly, although we know from the tensor virial theorem that giant ellipticals are globally anisotropic (Binney 1978), their detailed anisotropy structure is only poorly known.

The origin of this uncertainty is a fundamental degeneracy – in general, it is impossible to disentangle the anisotropy in the velocity distribution and the gravitational potential from velocity dispersion and rotation measurements alone (Binney & Mamon 1982; Dejonghe & Merritt 1992). Tangential anisotropy, for example, can mimic the presence of dark matter. Recent dynamical studies have indicated, however, that the anisotropy of the stellar distribution function (DF) is reflected in the shapes of the line-of-sight velocity profiles (VPs) in a way that depends on the gravitational potential (Gerhard 1993, hereafter G93; Merritt 1993). G93 and Merritt (1993) argued that the extra constraints derived from the VP measurements may be enough to break the degeneracy and determine the mass distribution.

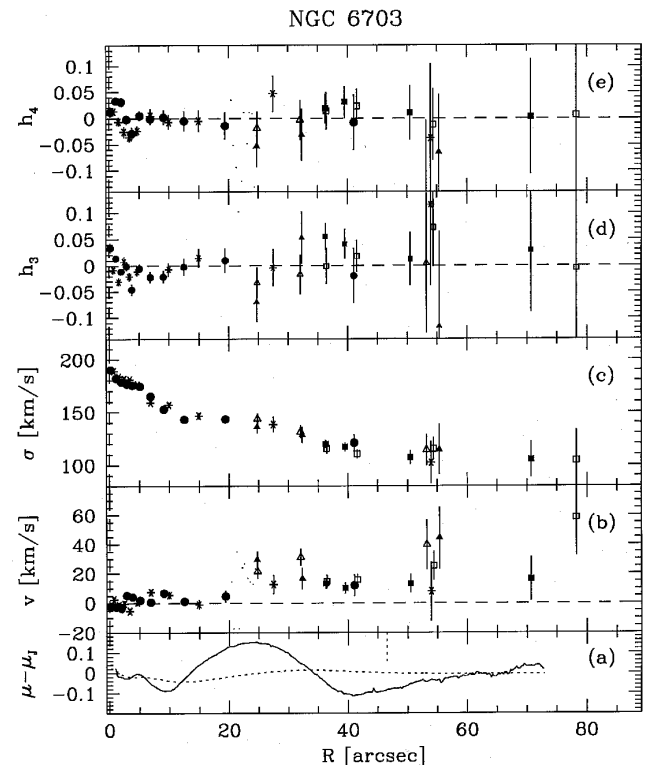
If this is correct, it provides a new method with which to investigate the properties of the dark matter haloes around elliptical galaxies at intermediate radii: VPs can now be estimated from high-quality absorption-line measurements out to  $\sim 3$  effective radii. Dynamical models are then used to disentangle the effects of orbital anisotropy and potential gradient on the VP shapes. In this paper, we implement these ideas for the analysis of real data, analysing the E0 galaxy NGC 6703. This study is part of an observational and theoretical program aimed at understanding the mass distribution and orbital structure in elliptical galaxies. Preliminary accounts of this work have been given in Jeske et al. (1996) and Saglia et al. (1997a).

We have obtained long-slit spectroscopy for NGC 6703, and have measured VPs to  $\sim 2.6R_e$  with the method of Bender (1990). The results are quantified by a Gauss–Hermite decomposition (G93, van der Marel & Franx 1993) as described by Bender, Saglia & Gerhard (1994, hereafter BSG). These observations are described in Section 2. In Section 3 we use simple dynamical models to describe the variation of the VP shapes with anisotropy and potential, generalizing the results of G93 for scale-free models. These models, taken from a systematic study of the relation between DFs and VPs in spherical potentials (Jeske 1995), are described in Appendix A. In Section 4 we develop a non-

parametric method for inferring the DF and potential from absorption-line profile measurements. Tests on Monte-Carlo-generated data are used to determine the degree of confidence with which the DF and potential can be inferred from real data. In Section 5 we analyse the kinematic data for NGC 6703. Comparison with the dynamical models from Jeske (1995) already shows that no constant  $M/L$  model will fit the data. The non-parametric method developed in Section 4 is then used to derive quantitative constraints on the mass distribution and anisotropy of this galaxy. Finally, in Section 6 we present a discussion of the results and our conclusions.

## 2 OBSERVATIONS OF NGC 6703

NGC 6703 is an E0 galaxy at a distance  $D=36$  Mpc (Faber et al. 1989) for  $H_0=50$  km s $^{-1}$  Mpc $^{-1}$ . From a  $B$ -band CCD frame taken at the prime focus of the 3.5-m telescope on Calar Alto and kindly provided by U. Hopp we have measured the inner surface brightness profile using the algorithm by Saglia et al. (1997b): this follows an  $R^{1/4}$  law with  $R_e=30$  arcsec  $=5.2 h_{50}^{-1}$  kpc, or a Jaffe model with  $r_j=46.5$  arcsec  $=8.1 h_{50}^{-1}$  kpc, with small residuals (Fig. 1). A 3 per



**Figure 1.** (a) Residuals of a Jaffe model fit to photometry for NGC 6703. Full line: in surface brightness; dotted line: in the curve of growth. The vertical dotted line marks the scaling Jaffe radius,  $r_j$ . (b) Folded mean velocity, (c) velocity dispersion, (d)  $h_3$  and (e)  $h_4$  profiles. Crosses and filled circles refer to the two sides of the galaxy and the major axis spectrum. The small dots refer to the unbinned spectrum (see text). Open and filled triangles refer to the two sides of the galaxy and the spectrum taken parallel to the minor axis and shifted 24 arcsec from the centre. Open and filled squares refer to the two sides of the galaxy and the spectrum taken parallel to the minor axis and shifted 36 arcsec from the centre.

cent increase of the sky value reduces the measured Jaffe radius to 35.5 arcsec; a 1 per cent decrease of the sky values increases it to 54 arcsec. Isophote shapes deviate little from circles ( $\varepsilon < 0.05$ ,  $|a_4/a| < 0.005$ ) and show small twisting ( $\Delta PA \simeq 10^\circ$ ). From the Jaffe profile fit we derive a fiducial (calibrated and corrected for Galactic absorption following Faber et al. 1989)  $M_B = -21.07$ , or luminosity  $L_B = 4.16 \times 10^{10} h_{50}^{-2} L_{\odot,B}$ . Note that the values of  $R_c$  and  $M_B$  derived here are slightly larger than those ( $R_c = 24$  arcsec,  $M_B = -20.79$ ) given by Faber et al. (1989).

The spectroscopic observations were carried out in 1994 October, 1995 May and 1995 August with the 3.5-m telescope on Calar Alto, Spain. In all of the runs the same setup was adopted. The Boller & Chivens long-slit twin spectrograph was used with a 1200 line  $\text{mm}^{-1}$  grating, giving  $36 \text{ \AA} \text{ mm}^{-1}$  dispersion. The detector was a Tektronix CCD with  $1024 \times 1024$  24- $\mu\text{m}$  pixels and a wavelength range 4760–5640  $\text{\AA}$ . The instrumental resolution obtained using a 3.6-arcsec-wide slit was  $85 \text{ km s}^{-1}$ . We collected 1.5 h of observations along the major axis of the galaxy, 4 h of observations perpendicular to the major axis and shifted to the north-east of the centre by 24 arcsec ( $0.8R_c$ ) and 13 h of observations perpendicular to the major axis and shifted to the north-east of the centre by 36 arcsec ( $1.2R_c$ ). Spectra taken parallel to the minor axis and shifted from the centre allow at the same time good sky subtraction and a symmetry check of the data points.

The analysis of the data was carried out following the steps described by BSG. The logarithmic wavelength calibration was performed at a smaller step ( $\Delta v = 30 \text{ km s}^{-1}$ ) than the actual pixel size ( $\approx 50 \text{ km s}^{-1}$ ) to exploit the full capabilities of the Fourier Correlation Quotient method. A sky subtraction better than 1 per cent was achieved. The heliocentric velocity difference between the 1995 May and the 1994 September–1995 August frames was taken into account before co-adding the observed spectra. The spectra were rebinned along the spatial direction to obtain a nearly constant signal-to-noise ratio larger than 50 per resolution element. The effects of the continuum fitting and instrumental resolution were extensively tested by Monte Carlo simulations. The residual systematic effects on the values of the  $h_3$  and  $h_4$  parameters are expected to be less than 0.01. The resulting fitted values for the folded velocity  $v$ , velocity dispersion  $\sigma$ ,  $h_3$  and  $h_4$  profiles are shown in Fig. 1 as a function of the distance from the centre, reaching  $\sim 2.6R_c$ . The  $v$  and  $\sigma$  profiles are folded antisymmetrically with respect to the centre for the major axis spectrum, and symmetrically with respect to the major axis for the spectra parallel to the minor axis. Template mismatching was minimized by choosing the template star that gave the minimal symmetric  $h_3$  profile derived along the major axis of the galaxy. The systematic effect resulting from residual mismatching on the derived  $h_4$  values, estimated from the remaining symmetric components, is less than 0.01.

The galaxy shows very little rotation ( $\approx 0 \text{ km s}^{-1}$  for  $R < R_c$ ,  $\approx 20$ – $30 \text{ km s}^{-1}$  for  $R > R_c$ ). The (cylindrical) rotation measured parallel to the minor axis is slightly larger ( $\approx 35 \text{ km s}^{-1}$ ) along the 24-arcsec shifted spectrum, but consistent with the peak velocity reached along the major axis at  $R \approx 22$  arcsec. This is shown by the velocities derived from the unbinned major axis spectra (dots in Fig. 1). The velocity dispersion drops from the central  $\approx 190 \text{ km s}^{-1}$  to

$\approx 140 \text{ km s}^{-1}$  at  $R_c/2$ , slowly declining to about  $110 \text{ km s}^{-1}$  in the outer parts. The  $h_3$  and  $h_4$  values are everywhere close to zero. The error bars are determined from Monte Carlo simulations. Noise is added to template stars (rebinned to the original wavelength pixel size) broadened following the observed values of  $\sigma$  and  $h_4$ , matching the power-spectrum noise to the peak ratio of the galaxy spectra. The accuracy of the estimated error bars (the rms of 30 replicas of the data points) is about 20 per cent (determined from the scatter of the estimated signal-to-noise ratios).

Data points at the largest distances for the different data sets have a lower signal-to-noise ratio than the mean and therefore have larger error bars. In addition, these data points are expected to suffer more from the systematic effects caused by the galaxy light contamination of the sky subtraction (see the discussion in Saglia et al. 1993). They are, in any case, consistent within the error bars with the more accurate values derived from the other available spectra.

The observed scatter is sometimes slightly larger than expected from the error estimates. This excess could be real and a result of the faint structures apparent in an unsharp mask image of the galaxy. In particular, this applies to the asymmetries observed in the  $h_4$  profile in the central 5 arcsec. The negative  $h_4$  values detected for the first data points of the 24-arcsec shifted spectrum are also real. They are detected in the unbinned major axis spectra at  $R \approx 22$  arcsec (dots in Fig. 1).

### 3 VELOCITY PROFILES IN SPHERICAL GALAXIES

To better understand the relation between VP shape, anisotropy and gravitational potential, we have constructed a large number of anisotropic models for spherical galaxies in which the stars follow a Jaffe (1983) profile. The gravitational potential was taken to be either that of the stars (self-consistent case) or one that had constant rotation speed everywhere (‘halo potential’). The latter case corresponds to a mass distribution with a dark halo that has equal density to the stars at  $r \simeq 0.4r_j$  and equal interior mass at  $r \simeq r_j$ , where  $r_j$  is the scale radius of the Jaffe model. Anisotropic quasi-separable DFs  $g(E)h(E, L^2)$  were calculated by the method of Gerhard (1991, hereafter G91), but contrary to G93 the circularity function (which specifies the distribution of angular momenta on energy shells) was allowed to vary with energy. These models include DFs in which the anisotropy changes radially: from tangential to radial or vice versa, from isotropic to radial to tangential, etc. (see Fig. 18, later). For comparison, we have also constructed families of Osipkov (1979)–Merritt (1985) models, which become strongly radially anisotropic beyond a certain radius. Details of this model data base and the properties of the VPs are given in Appendix A and in Jeske (1995); here we only give a brief summary relevant to the comparison with NGC 6703.

The shapes of the observable VPs are most sensitive to the anisotropy of the DF, but depend also on the potential (G93). For rapidly falling luminosity profiles, the VPs are dominated by the stars at the tangent point. Radially (tangentially) anisotropic DFs then lead to more peaked (more

flat-topped) VPs than in the isotropic case; in terms of the Gauss–Hermite parameter  $h_4$ , this corresponds to  $h_4 > (h_4)_{\text{iso}}$  and  $h_4 < (h_4)_{\text{iso}}$ , respectively (figs 8 and 9 in G93).

Fig. 2 shows that these trends are also seen in the present models in which both the luminosity density and the anisotropy change with radius. An increase in radial (tangential) anisotropy at intrinsic radius  $r$  is accompanied by an increase (decrease) of  $h_4$  at projected radius  $R \simeq r$ . The correspondence is strongest in the outer parts of the models, but is also seen to a lesser extent in the centre of a Jaffe model, where  $\rho(r) \propto r^{-2}$ , in contrast to a homogeneous core where radial orbits lead to broadened VPs (Dejonghe 1987). Quantitatively, the correspondence depends also on the anisotropy gradient. Osipkov–Merritt models show a reversal of this trend near their anisotropy radius  $r_a$  because of the large number of high-energy radial orbits turning around near  $r_a$ ; this leads to flat-topped VPs in a small radius range near  $r_a$ . However, the properties of these models are extreme and they are in general not very useful for modelling observed VPs.

Fig. 2 and figs 8 and 9 in G93 also show that as the mass of the model at large  $r$  is increased at constant anisotropy, both the projected dispersion and  $h_4$  increase. Increasing  $\beta$  at constant potential, on the other hand, lowers  $\sigma$  and increases  $h_4$ . This suggests that by modelling  $\sigma$  and  $h_4$  both mass  $M(r)$  and anisotropy  $\beta(r)$  can in principle be found.

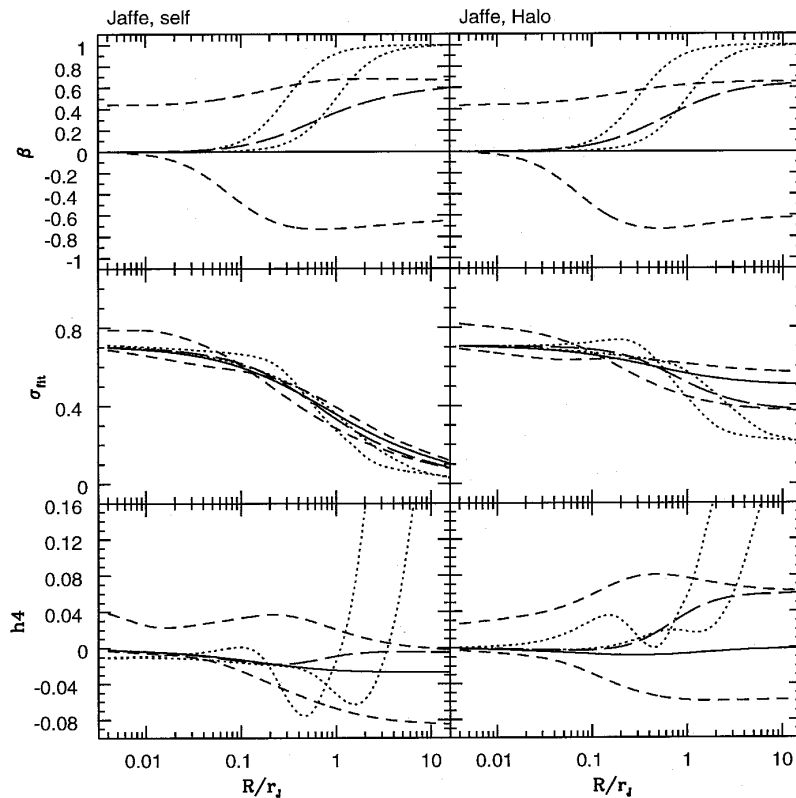
#### 4 MODELLING ABSORPTION-LINE PROFILE DATA

Having seen the effect of anisotropy and potential variations on the line-profile parameters, we now proceed to construct an algorithm by which the distribution function and potential of a spherical galaxy can be constrained from its observed  $\sigma$  and  $h_4$  profiles. Such absorption-line profile data contain a subset of the information given by the projected distribution function  $N(\mathbf{R}, v_{\parallel})$ , which in the spherical case is related to the full DF by

$$\begin{aligned} N(\mathbf{R}, v_{\parallel}) &= \int dz \iint dv_x dv_y f(E, L^2) \\ &= \int dz \iint dv_x dv_y f[\tfrac{1}{2}v_r^2 + \tfrac{1}{2}v_t^2 + \Phi(r), r^2 v_t^2]. \end{aligned} \quad (1)$$

Here the position on the sky is specified by  $\mathbf{R} = (x, y)$ ;  $R = |\mathbf{R}|$ . Velocities in the sky plane are denoted by  $(v_x, v_y)$ ,  $z$  and  $v_{\parallel}$  are the line-of-sight position and velocity, and  $v_r$  and  $v_t$  are the intrinsic radial and tangential velocities. In spherical symmetry, the DF  $f(E, L^2)$  is a function of energy and squared angular momentum only.

Notice from equation (1) that the projected DF depends linearly on  $f$ , but non-linearly on the potential  $\Phi(r)$ . Thus  $f$  will generally be easier to determine from  $N(\mathbf{R}, v_{\parallel})$  than  $\Phi$ .



**Figure 2.** Fitted projected velocity dispersion  $\sigma_{\text{fit}}$ , anisotropy parameter  $\beta$  and VP parameter  $h_4$  for representative Jaffe models in self-consistent (left) and halo potentials (right). The models shown are radially and tangentially anisotropic models constructed with the method of G91 (dashed lines), the isotropic model (solid lines) and two Osipkov–Merritt models (dotted lines). Note that while  $\beta$  is a function of three-dimensional radius  $r_h$ , whereas  $\sigma_{\text{fit}}$ ,  $h_4$  are observed quantities depending on projected radius  $R$ , there is a close correspondence between features in these profiles. See text.

Moreover, while considerations like those in the last section do suggest that, in spherical symmetry and for positive  $f(E, L^2)$ , both the DF and the potential can be determined from  $N(R, v_{\parallel})$ , there is no theoretical proof that this is in fact true. We only know that in a *fixed* spherical potential the DF is uniquely determined from  $N(R, v_{\parallel})$  (Dejonghe & Merritt 1992). For these reasons we have found it useful to split our problem into two parts (see also Merritt 1993).

(1) We fix the potential  $\Phi$ , and from the photometric and kinematic data determine the ‘best’ DF  $f$  for this potential. As in practice the surface brightness (SB) profile is much better sampled than the kinematic observations and also has smaller errors, we treat it separately and determine the stellar luminosity density  $j(r)$  at the beginning. The kinematic data are then used to determine the ‘best’  $f$  for given  $j(r)$  and  $\Phi(r)$ , by approximately solving equation (1) as a linear integral equation.

(2) We then vary  $\Phi$  to find the potential that allows the best fit overall. At present, it is not practical in step (2) to attempt to determine the potential non-parametrically. Rather, we choose a parametrized form for  $\Phi$ , and find the region in parameter space for which the ‘best’ DF as determined in step (1) reproduces the data adequately.

In view of the modelling of NGC 6703 in Section 5, we have considered the following family of potentials, including a luminous and a dark matter component. The stellar component is approximated as a Jaffe (1983) sphere, with scale radius  $r_j$  and total mass  $M_j$ , so that

$$\Phi_L(r) = \frac{GM_j}{r_j} \ln \frac{r}{r+r_j}. \quad (2)$$

The dark halo has an asymptotically flat rotation curve,

$$v_c(r) = v_0 \frac{r}{\sqrt{r^2 + r_c^2}}, \quad (3)$$

so that its potential is

$$\Phi_H(r) = \frac{1}{2}v_0^2 \ln(r^2 + r_c^2). \quad (4)$$

This is specified by the asymptotic circular velocity  $v_0$  and the core radius  $r_c$ . Both luminous and dark halo components can be modified when needed, and need not be analytic functions.

In testing our method below, we use parameters adapted to NGC 6703. This galaxy is well-fitted by a Jaffe profile (Section 2), so  $r_j$  is known. This leaves three free parameters, the mass  $M_j$  or mass-to-light ratio  $\Upsilon$  of the stellar component and the halo parameters  $r_c$  and  $v_0$ . If one assumes that the central kinematics is dominated by the luminous matter,  $\Upsilon$  can be determined; then only the two halo parameters  $r_c$  and  $v_0$  are free. The assumption of maximum stellar  $\Upsilon$  is similar to the maximum disc assumption in spiral galaxies.

In any of the potentials specified by equations (2)–(4), we determine the DF by the algorithm described in Section 4.1 below. To assess the significance of the results obtained, we test the algorithm on Monte-Carlo-generated pseudo-data in Section 4.2. For kinematic data with the spatial extent and observational errors measured for NGC 6703, the algorithm recovers a smooth spherical DF  $\sim 70$  per cent of the

time to an rms level of  $\sim 12$  per cent, taken inside three times the radius of the outermost kinematic data point. In Section 4.3, we investigate the degree to which the gravitational potential can be constrained from similar data.

#### 4.1 Recovering $f$ from $\sigma$ and $h_4$ , given $\Phi$

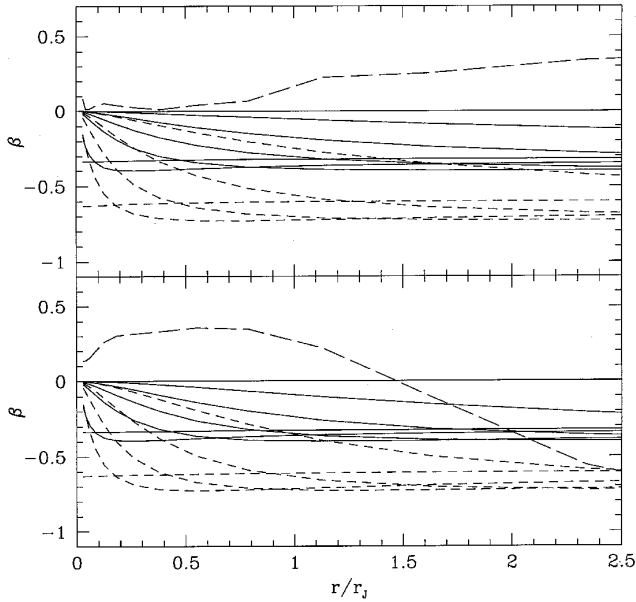
As discussed above, the projected distribution function  $N(r, v_{\parallel})$  suffices to determine DF  $f(E, L^2)$  uniquely. In practice, however, only incomplete and noisy data are available in place of  $N(r, v_{\parallel})$ , and, contrary to the two-dimensional function  $N(r, v_{\parallel})$ , the observed  $\sigma(R_i)$  and  $h_4(R_i)$  contain only one-dimensional information. This suggests that we can hope to recover only the gross features of  $f$  from such kinematic data. Indeed, the anisotropy parameter  $\beta(r)$  seems to be essentially fixed from accurate  $h_4$  measurements (e.g. figs 8 and 9 in G93). Local fluctuations in the DF will be inaccessible, but, as we will show, smooth DFs can be recovered with reasonable accuracy from presently available data.

To solve the inversion problem, we first compute a set of self-consistent models  $f_k(E, x)$  for the stellar density  $j(r)$ , in the fixed potential  $\Phi(r)$ . The  $f_k(E, x)$  are models of the kind discussed in Section 2 and Appendix A;  $E$  and  $x$  are the energy and circularity integrals of the motion. We then write the DF as a sum over these ‘basis’ functions:

$$f = \sum_{k=1, K} a_k f_k(E, x). \quad (5)$$

We do not need to use a doubly infinite, complete set of basis functions because most of the high-frequency structure represented by the higher order of basis functions in such a set will be swamped by noise in the observational data. It is sufficient to choose the number of basis functions,  $K$ , and the  $f_k(E, x)$  themselves such that the data can be fitted with a mean  $\chi^2 \simeq 1$  per data point. We have found it advantageous to use the isotropic model plus tangentially anisotropic basis models, because with these the anisotropy of the final composite DF (5) can be varied in a more local way than with radially anisotropic components. As the  $a_k$  can be negative, it is of course no problem to generate a radially anisotropic DF from the isotropic model plus a set of tangential basis models.

In Fig. 3 we illustrate the basis models that we have used, by plotting radial profiles of the anisotropy parameter  $\beta(r)$  for a subset of them. For these basis functions,  $\beta(r)$  can be regarded as a measure of the extent of the DF in circularity  $x$  at the energy  $E = \Phi(r)$ . The figure shows that the basis resolves three steps in  $x$  in the limits  $r \rightarrow 0$  and  $r \rightarrow \infty$ ; at intermediate energies it has much finer resolution. The majority of the functions are used in resolving the energy dependence, i.e. on placing the main gradient zones of the functions  $f_k$  in the  $(E, x)$  plane on a relatively dense grid in energy. The main difference from using power-law components (Fricke 1952; Dejonghe et al. 1996) is that our basis models are already reasonable in the sense that they are viable dynamical models for the density in question, and that the superposition is used only to match the kinematics. We have typically used  $K \lesssim 20$  such functions; this proved sufficient even for analysing pseudo-data of much better quality than we have for NGC 6703.



**Figure 3.** Anisotropy parameter  $\beta$  for a subset of basis functions  $f_k$  used in the self-consistent Jaffe potential (top) and in a mixed Jaffe plus halo model (bottom). The top full line in each panel shows the isotropic model. All other basis models are tangentially anisotropic. Two values of asymptotic anisotropy as  $r \rightarrow \infty$  are used (full and short-dashed lines). For illustration, the long-dashed lines show the  $\beta$  profiles of the best-fitting DFs derived with these bases from the NGC 6703 data, in both potentials.

Each of the  $f_k(E, x)$  reproduces the stellar density distribution  $j(r)$ , so the  $a_k$  satisfy

$$\sum_{k=1, K} a_k = 1. \quad (6)$$

Moreover, the coefficients  $a_k$  must take values such that

$$\sum_{k=1, K} a_k f_k(E, x) \geq 0 \quad (7)$$

everywhere in phase-space. In practice, these positivity constraints are imposed on a grid in energy and circularity ( $E = E_j$ ,  $x = x_j$ ). Subject to these constraints, the  $a_k$  are to be determined such that the kinematics predicted from the DF (5) match the observed kinematics in a minimum  $\chi^2$  sense.

The comparison between model and data is not entirely straightforward, however. As the measured ( $v$ ,  $\sigma$ ,  $h_3$ ,  $h_4$ ) are obtained by *fitting* to the line profile (the observational analogue of the projected DF), they unfortunately depend non-linearly on the underlying DF of the galaxy. We therefore cannot use the observed  $v$  and  $\sigma$  in a linear least-squares algorithm to determine the  $a_k$  from the data – they cannot be written as moments of  $f$ . In the fitting process, they have to be replaced by quantities that do depend linearly on the DF. Moreover, the error bars for these new quantities generally depend not only on the observed error bars of  $v$  and  $\sigma$ , but also on the errors and the values of the line profile parameters  $h_3$  and  $h_4$ . They must therefore be determined with some care.

We have investigated several schemes along these lines. The following scheme seemed to perform best in recovering

a known underlying spherical DF from pseudo-data. From the measured ( $\sigma$ ,  $h_4$ ), we compute an approximation to the true velocity dispersion (second moment)  $\bar{\sigma}^2(R_i)$ , by integrating over the line profile (for negative  $h_4$  only until it first becomes negative). We also evaluate a new set of even Gauss–Hermite moments  $s_n(R_i; \bar{\sigma})$  from the data, using fixed, fiducial velocity scales  $\bar{\sigma}(R_i)$  (G93; Appendix B). We have found it convenient to take for these  $\bar{\sigma}(R_i)$  the velocity dispersions  $\sigma_{\text{iso}}(R_i)$  of the isotropic model using the stellar density of the galaxy, in the current potential  $\Phi(r)$ .

The velocity-profile moments of the basis function models are transformed to the same velocity scales  $\bar{\sigma}(R_i)$ . The moments  $\bar{\sigma}^2(R_i)$  and  $s_n(R_i; \bar{\sigma})$  of the composite DF then depend linearly on the corresponding moments of the basis function models. For a regularized model they are smooth functions of  $R$ , of which the (noisy) observational moments are assumed to be a random realization within the respective errors.

To determine the best-fitting coefficient  $a_k$  we minimize the sum over data points  $i$  for all

$$\chi_{\sigma, i}^2 \equiv w_{\sigma}^2(R_i) \left[ \bar{\sigma}^2(R_i) - \sum_{k=1, K} a_k \bar{\sigma}_k^2(R_i) \right]^2 \quad (8)$$

and

$$\chi_{n, i}^2 \equiv w_n^2(R_i) \left[ s_n(R_i; \bar{\sigma}) - \sum_{k=1, K} a_k s_n^{(k)}(R_i; \bar{\sigma}) \right]^2, \quad (9)$$

for  $n=2, 4$ . These equations make use of the fact that all the  $f_k$  are self-consistent models for the same  $j(r)$ , so that all surface density factors  $\mu_k(R_i) = \mu(R_i)$  cancel.

To determine the weights  $w_{\sigma}$  and  $w_n$ , we have performed Monte Carlo simulations to study the propagation of the observational errors  $\Delta\sigma$  and  $\Delta h_4$ . Based on the results of these simulations, we have chosen

$$W_{\sigma}^{-1}(R_i) = 2.4 \bar{\sigma}(R_i) [\bar{\sigma}(R_i) / \sigma(R_i)] \Delta\sigma(R_i) \quad (10)$$

$$w_2^{-1}(R_i) = 0.7 \Delta\sigma(R_i) / \sigma(R_i), \quad (11)$$

$$w_4^{-1}(R_i) = 0.95 \sigma \Delta h_4. \quad (12)$$

The coefficients are representative in the range of values taken by the observed error bars and the measured  $h_4$ . In the presently-used fitting procedure, the additional dependence on  $h_4$  and the respective ‘other’ error bar are neglected – the tests below show that this leads to satisfactory results. As to first order the  $s_2$  moment thus measures the shift from  $\sigma$  to  $\bar{\sigma}$ , the parameter  $\alpha$  in equation (13) will be near  $\alpha \simeq 2$ . We have fixed  $\alpha$  by requiring that the distributions of  $\chi_{\sigma}^2$  and  $\chi_{h_4}^2$  have equal width (see below); this results in  $\alpha = 1.7$ .

Finally, we assume that the DF underlying the observed kinematics is smooth to a degree that is compatible with the measured data. Clearly, unless such an assumption is made, it is impossible to determine a function of two variables,  $f(E, L^2)$ , from a small number of data points with real error bars. One way to ensure that the DF is smooth is to use only a small number of terms in the expansion (5). However, this is not a good way of smoothing as it biases the recovered DF towards the functional forms of the few  $f_k$  that are used in the sum. A better way of smoothing is the method of regularization, as recently discussed by Merritt (1993) in a similar context. Regularization has a tradition in other

branches of science, and different variants exist; for references see Press et al. (1986) and Merritt's paper (1993). In the algorithm here, regularization is implemented by taking the number of basis functions that are large enough (typically 16–20) for the data to be modelled in some detail, and then constraining the second derivatives of the composite DF to be small. In other words, we also seek to minimize

$$\Lambda(f)_{ij} \equiv w_r^2(E) \times \left[ \left( D_E^2 \frac{\partial^2 f}{\partial E^2} \right)^2 + 2 \left( D_E \frac{\partial^2 f}{\partial E \partial x} \right)^2 + \left( \frac{\partial^2 f}{\partial x^2} \right)^2 \right]_{E=E_i, x=x_j} \quad (13)$$

on a grid of points  $(E_i, x_j)$  in the energy and circularity integrals. Here we normalize to the isotropic DF to ensure that fluctuations in the composite DF are penalized equally at all energies; i.e.  $w_r(E) = 1/f_{\text{iso}}(E)$ . The constant  $D_E$  is proportional to the range in potential energy in the respective model.

To find a regularized spherical DF for given kinematic data in a specified potential, we thus minimize the quantity

$$\Delta^2 \equiv \sum_{i=1}^I \left\{ \chi_{\sigma,i}^2 + \sum_{n=2,4} \chi_{h_n,i}^2 \right\} + \lambda \sum_{i,j} \Lambda(f)_{ij} \quad (14)$$

for a given regularization parameter  $\lambda$ , subject to the equality and inequality constraints (6) and (7). For the actual numerical solution we have used the NETLIB routine LSE1 by Hanson and Haskell (1981). Once the best model is found, we redetermine its quality by evaluating its deviations from the actually measured data:

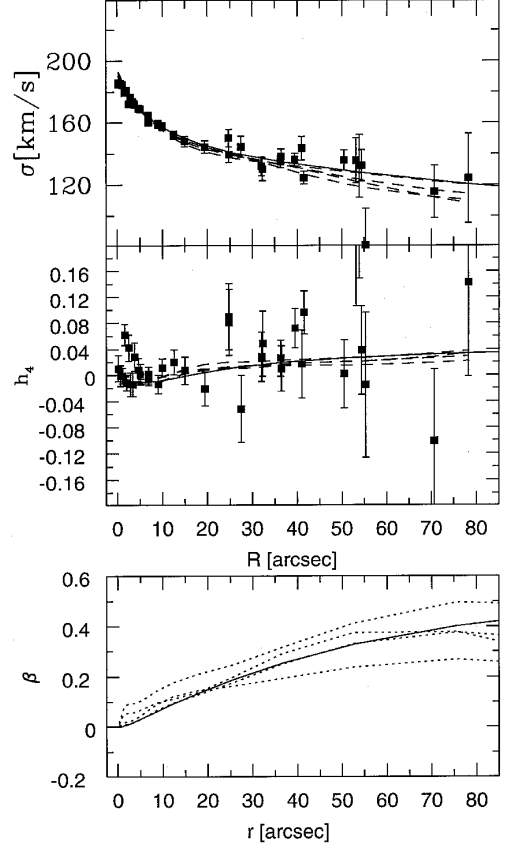
$$\chi_{\sigma}^2 = I^{-1} \sum_{i=1}^I [\sigma(R_i) - \sigma^{(f)}(R_i)]^2 / (\Delta\sigma)^2(R_i), \quad (15)$$

$$\chi_{h_4}^2 = I^{-1} \sum_{i=1}^I [h_4(R_i) - h_4^{(f)}(R_i)]^2 / (\Delta h_4)^2(R_i). \quad (16)$$

The parameters  $\sigma^{(f)}(R_i)$  and  $h_4^{(f)}(R_i)$  are determined by fitting a Gauss–Hermite series to the velocity profiles of the best-fitting model;  $I$  is the number of kinematic data points.

#### 4.2 Tests with model data

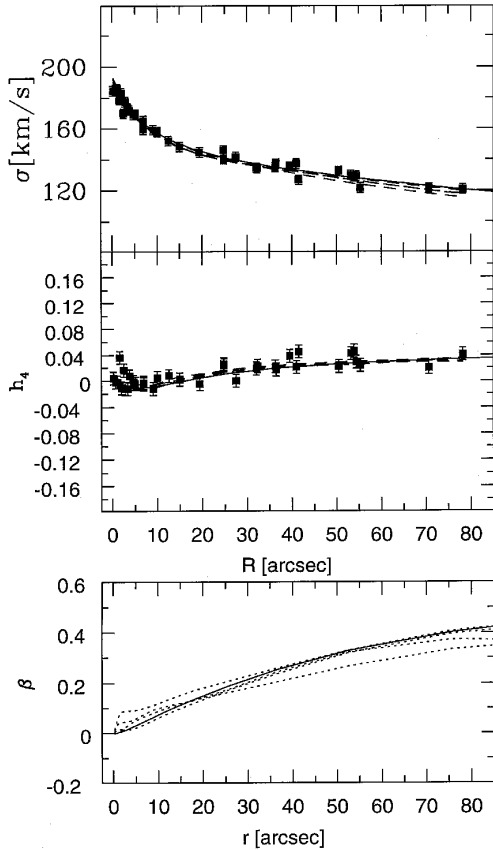
We have tested this method by applying it to kinematic data sets generated in the following Monte-Carlo-like way. First, velocity dispersions and  $h_4$  parameters are calculated from a theoretical DF of specified anisotropy in a known potential, and are interpolated to the radii  $R_i/r_j$  at which observed data points are available for NGC 6703. Error bars at these  $R_i$  are taken to be either the measured error bars for NGC 6703 (a realistic data set), or independent of radius with  $\Delta\sigma = 3 \text{ km s}^{-1}$  and  $\Delta h_4 = 0.01$  (an idealized data set). Pseudo-data are then generated from the model values of  $\sigma$  and  $h_4$  at  $R_i$  by adding Gaussian random variates with  $1\sigma$  dispersion corresponding to the respective  $\Delta\sigma$  or  $\Delta h_4$  error bar at this point. Figs 4 and 5 show data sets generated in this way from a radially anisotropic model in a potential of



**Figure 4.** Model analysis of pseudo-data generated from a radially anisotropic model DF with the sampling and the measured error bars of NGC 6703 (see text). The full curves show the true profiles of projected velocity dispersion, line-of-sight velocity distribution parameter  $h_4$  and anisotropy parameter  $\beta$  of the underlying DF. The dashed and dotted lines show the  $\sigma$ ,  $h_4$  and  $\beta$  profiles of five regularized composite DFs which were computed by the method of Section 4.1 with  $\lambda = 10^{-4}$ . One of the four curves in each panel corresponds to the data points actually shown in this figure; the other three curves derive from statistically identical kinematic data sets with randomly different values for  $\sigma$  and  $h_4$  within the same (Gaussian) errors.

a self-consistent Jaffe sphere with  $r_j = 46.5 \text{ arcsec}$ ,  $GM_j/r_j = 272 \text{ km s}^{-1}$ , and a dark halo with  $v_0 = 220 \text{ km s}^{-1}$ ,  $r_c = 56 \text{ arcsec}$ . This potential was chosen because it lies in the middle of the range of acceptable potentials for NGC 6703 (see Section 5.2), so that any systematic errors in our analysis will be similar for this model and the galaxy itself.

We have used the regularized inversion algorithm described in the last subsection to analyse several such pseudo-data sets, and have determined composite DFs as a function of the regularization parameter  $\lambda$ . The algorithm was given 16 basis function models. These included the isotropic model and a variety of tangentially anisotropic models with different anisotropy radii and circularity functions, but not, of course, the radially anisotropic model from which the data are drawn (see Fig. 3 and Appendix A). For each composite model returned by the algorithm, we have determined two diagnostic quantities. The first is the mean  $\chi^2$  per  $\sigma$  and  $h_4$  data point,  $\chi_{\sigma+h_4}^2$ , which measures the level at which this model fits the data from which it was derived. The

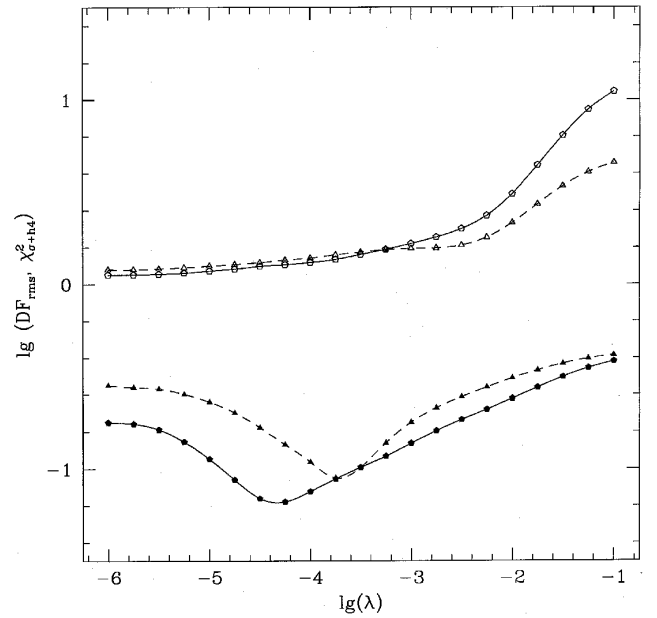


**Figure 5.** Model analysis of pseudo-data generated from the same radially anisotropic model DF as in Fig. 4, but assuming radially constant error bars  $\Delta\sigma=3 \text{ km s}^{-1}$  and  $\Delta h_4=0.01$ . The full curves again show the true  $\sigma$ ,  $h_4$  and  $\beta$  profiles of the underlying DF. The dashed lines show the projected  $\sigma$  and  $h_4$  profiles derived by the method of Section 4.1 from four sets of pseudo-data, with  $\lambda=6 \times 10^{-5}$ . The results obtained for these projected quantities from statistically identical data sets are now nearly identical. The dotted curves show the uncertainties that remain in the deprojected quantity  $\beta$  even with such small error bars.

second is the rms deviation between the returned DF and the true DF of the model from which the data were drawn, in some specified energy range.

The use of these diagnostic quantities requires some further comments. As usual, the number of degrees of freedom in such a regularized inversion problem is not well-determined. For near-zero  $\lambda$ , in the case at hand we can adjust 16 coefficients  $a_k$  and an overall mass scale. The total number of  $\sigma$  and  $h_4$  data points is 70; thus in this limit the number of degrees of freedom is 53. For large  $\lambda$ , on the other hand, the recovered DF will be linear in both  $E$  and  $x$  and the values of the  $a_k$  are essentially fixed. The number of degrees of freedom then approaches 69. In both cases, the number of degrees of freedom is of the order of the number of data points, hence the use of  $\chi^2_{\sigma+h_4}$  per  $\sigma$  and  $h_4$  data point instead of a reduced  $\chi^2$ .

The second diagnostic measuring the accuracy of the recovered DF must clearly depend on the range in energy over which it is calculated. Typically, a kinematic measurement at projected radius  $R_i$  contains information about the DF in a range of energy above the energy of the circular



**Figure 6.** Model results as a function of the regularization parameter  $\lambda$ , for the two pseudo-data sets shown in Figs 4 and 5. The two data sets are flagged by triangles (Fig. 4) and pentagons (Fig. 5). The upper two curves marked by the open symbols show the total  $\chi^2$  per  $\sigma$  and  $h_4$  data point of regularized composite models with 16 basis functions. The lower two curves marked by the filled symbols show the rms deviation between the recovered composite model DFs and the true DF from which the data sets were generated. This rms deviation was evaluated on a grid in energy and angular momentum corresponding to radii  $<3$  times the radius of the outermost data point.

orbit at radius  $R_i$ . The precise upper end of this range is model-specific; it depends on the potential, the kinematic properties of the DF itself, and, through the projection process, also on the stellar density profile. The use of the outermost kinematic data points thus contains, explicitly or implicitly, assumptions on the radial smoothness of these quantities. In a typical elliptical galaxy problem, the DF must be known fairly accurately at around  $3R_i$  for the projected kinematics at  $R_i$  to be securely predicted, and for radially anisotropic models even the values of the DF near  $10R_i$  can make some difference. The rms residuals in the recovered DF given below have therefore been calculated in a range of energies extending from  $\Phi(r=0)$  to  $\Phi(r=3R_m)$ , where the last kinematic data point in the data sets used is located at  $R_m=1.68r_j$ .

Fig. 6 shows these quantities as a function of the regularization parameter  $\lambda$  for the two pseudo-data sets shown in Figs 4 and 5. For small values of  $\lambda$ , the composite models fit the data accurately, but the recovered DF is not very accurate because it contains large spurious oscillations depending on the particular values of the data points. For large values of  $\lambda$ , the models are so heavily smoothed that they neither fit the data well nor represent a good approximation to the true DF. The optimal regime lies where the smoothing is large enough to damp out the spurious oscillations, but still permits the important structures in the underlying DF to be resolved. For values of  $\lambda$  in this regime, the fit to the data is still satisfactory, and the representation of the

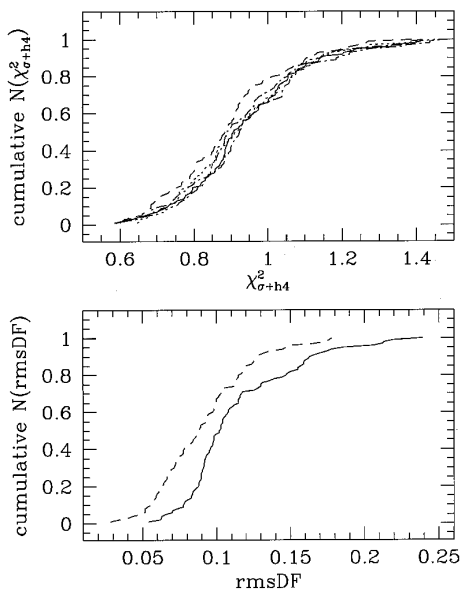


DF is optimal. Fig. 6 shows that the rms residuals of the DF go through minima at values of  $\lambda \simeq 2 \times 10^{-4}$  for the pseudo-data in Fig. 4 and  $\lambda \simeq 5 \times 10^{-5}$  for those in Fig. 5 (when  $D_E = 1$ ; see below equation 13).

The shapes of the  $\chi_{\sigma+h_4}^2(\lambda)$  curves were found to be always similar to the upper curves in Fig. 6. The shapes of the corresponding lower curves in Fig. 6 are more variable. The resulting optimal values for  $\lambda$  can vary, depending on the random realization of the data within the assumed Gaussian errors, as well as on the distribution function and potential from which the model values are drawn. We have therefore investigated 100 realizations of data generated from each of several model DFs and potentials. Based on these experiments, we have fixed optimal values of  $\lambda = 1 \times 10^{-4}$  and  $\lambda = 6 \times 10^{-5}$  for data with error bars like those in Figs 4 and 5, respectively, for all models that include a dark halo component.

We have performed similar experiments with a self-consistent model that is radially anisotropic near the centre and tangentially anisotropic in its outer parts, such as might be relevant in tests for dark matter at large radii. In these tests we have used 20 basis functions. To match the corresponding pseudo-data with similar accuracy requires smaller optimal values of  $\lambda$  ( $\sim 3 \times 10^{-9}$ ), so as to compensate for the larger derivatives in equation (13).

Fig. 7 shows the cumulative distributions of  $\chi_{\sigma+h_4}^2$  and the rms residual in the DF, as described, for the dynamical



**Figure 7.** Top: the cumulative distribution of the total normalized  $\chi_{\sigma+h_4}^2$ , for dynamical models recovered from 100 random Gaussian data sets derived from the underlying true DF and the observational errors in Figs 4 (solid line) and 5 (dashed line). Bottom: cumulative distribution of residuals between true and recovered distribution functions, evaluated on a grid extending to three times the radius of the last data point, for the same 100 data sets from both models. Also shown in the top panel are the cumulative  $\chi_{\sigma+h_4}^2$  distributions for data drawn from radially anisotropic models in the two potentials corresponding to two of the extreme solid lines in Fig. 11, with the same error bars as in Fig. 4 (dot-dashed lines), and for a self-consistent model with more complicated anisotropy structure (dotted line).

models recovered from several such sets of Monte Carlo data. From the top panel it is seen that our fitting algorithm will match kinematic data in a known potential with  $\chi_{\sigma+h_4}^2 \leq 1$  about 60–70 per cent of the time, and with  $\chi_{\sigma+h_4}^2 > 1.28$  only < 5 per cent of the time. These numbers are similar to those expected from a  $\chi^2$  distribution with 70 degrees of freedom (for Gaussian data), which should describe the statistical deviations of the Monte Carlo data points from the underlying true DF. If in modelling the data for NGC 6703 a level of  $\chi_{\sigma+h_4}^2 < 1.28$  cannot be reached, the assumed potential is not correct with 95 per cent confidence.

The bottom panel of Fig. 7 shows the distribution of residuals in the recovered DF for one model. There are three factors that limit the degree to which a given underlying DF can be recovered. The first is determined by the data, i.e. the size of the error bars, the sampling, the fact that there are no measurements at large radii, etc. The second is the level of detail that can be resolved by the modelling, given the finite number and the particular form of the basis functions. The third is the fraction of small-scale gradients in the model itself. Figs 6 and 7 show that, if the potential is known, data like those for NGC 6703 (Fig. 4) permit the recovery of reasonably smooth DFs to an rms level of  $\lesssim 12$  per cent inside  $R_m$  about 60–70 per cent of the time. Data with much smaller error bars but the same sampling (Fig. 5) would give an rms level of  $\lesssim 10$  per cent ( $\lesssim 8$  per cent for the other two halo models shown in the top panel of Fig. 7). In the radial–tangential self-consistent model the DF to  $3R_m$  is recovered with an rms accuracy of 16 and 35 per cent for data with error bars like those in Figs 5 and 4, respectively. These comparisons and the results of Fig. 7 show that the former values are dominated by the measurement uncertainties rather than the resolution of the modelling. This conclusion would be different for highly corrugated true DFs. However, we would not be able to recover such DFs from realistic data in any case.

The kinematics of the regularized composite DFs derived from pseudo-data with the chosen optimal  $\lambda$  values are shown in Figs 4 and 5. (For brevity, such model DFs obtained with near-optimal  $\lambda$  will henceforth be denoted as ‘best-estimate models’). The good match to the data points is apparent. The differences in the intrinsic anisotropy parameter  $\beta$  between the recovered models and the true model are larger. For the data set generated with the observed error bars of NGC 6703, the recovered  $\beta$  values are uncertain by  $\Delta\beta \simeq \pm 0.1$ , and a slightly larger value at the largest radii where the observational errors are large. For the pseudo-data with the small error bars in Fig. 5, this uncertainty is reduced.

### 4.3 Constraining $\Phi$

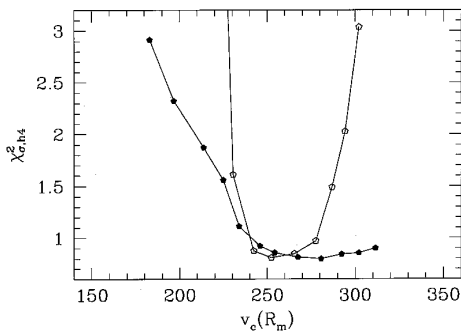
So far, we have shown the stellar DF can be recovered from VP data in a known spherical potential. In this section we return to the discussion of Section 3 and investigate the degree to which the gravitational potential itself can be constrained. Eventually, it will clearly be important to answer the theoretical question of whether, in principle, the gravitational potential is nearly or even uniquely determined from the projected DF  $N(r, v_{\parallel})$ . However, we shall not attempt to do this here; previous theoretical work sug-

gests that the range of potentials consistent with ideal data is small (G93; Merritt 1993; see Section 3). More relevant at the moment perhaps is the more practical question of how well the potential can be constrained from real observational data with realistic error bars, finite radial extent and limited sampling. What is the range of pairs  $(f, \Phi)$  that correspond to the same data? How does this range shrink as the data improve?

Here we investigate these questions with a view to the analysis of the NGC 6703 data below. We again use the model underlying Figs 4 and 5; this is a radially isotropic DF in the potential of a self-consistent Jaffe sphere and a dark halo with parameters  $GM_J/r_J = 272 \text{ km s}^{-1}$ ,  $r_c/r_J = 1.2$  and  $v_0^2/(GM_J/r_J) = 0.808$ . As before, random Gaussian data sets were generated from this model, with the positions and error bars of the data points (i) as in Fig. 4, and (ii) as in Fig. 5. The last data point is at  $R_m = 1.68r_J$ , as for the NGC 6703 data, well beyond two effective radii. However, contrary to Section 4.2, the data sets used in the following were specially selected such that  $\chi_\sigma^2 \simeq 1$  and  $\chi_{h_4}^2 \simeq 1$ . (It turns out that the data points shown in Fig. 4 have less than 5 per cent probability according to Fig. 7.)

For these pseudo-data we have determined best-estimate DFs in a number of assumed potentials, including the underlying true potential. The potentials were chosen such that (i) they correspond to approximately constant mass-to-light ratio for  $r \ll r_J$ , and (ii) they form a sequence of varying true circular velocity  $v_c(R_m)$  at the radius of the last data point. The sequence, with the slightly different velocity normalizations appropriate for NGC 6703, is shown in Section 5. For each potential, specified by the selected values of the parameters  $r_c/r_J$  and  $v_0^2/(GM_J/r_J)$ , we determined the goodness-of-fit  $\chi_{\sigma+h_4}^2$  as a function of the velocity scale  $GM_J/r_J$  from the corresponding best-estimate models. Finally, we computed the  $\chi_{\sigma+h_4}^2$  of the best-estimate DF for the optimum velocity scale. This optimal velocity scale is usually found to be slightly different for the two pseudo-data sets.

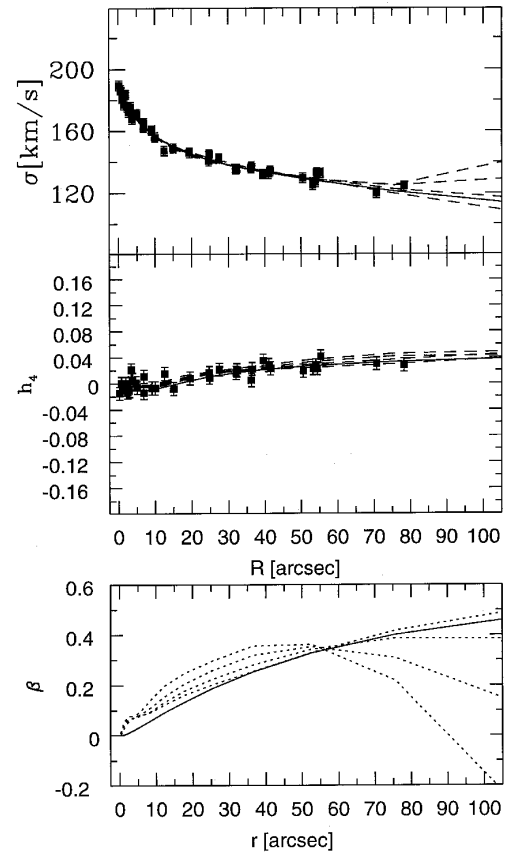
Fig. 8 shows the goodness of fit  $\chi_{\sigma+h_4}^2$  of the optimal sequence of best-estimate models as a function of  $v_c(R_m)$ , as determined for both model data sets. The underlying true potential has  $v_c(R_m) = 242 \text{ km s}^{-1}$  with the adopted param-



**Figure 8.** Goodness of fit  $\chi_{\sigma+h_4}^2$  for best-estimate DFs derived in a sequence of luminous plus dark matter potentials. The sequence of potentials, with rotation curves similar to those shown in Fig. 11, is here parametrized by the total circular velocity at the last observational radius,  $v_c(R_m)$ . The DFs were fitted to pseudo-data generated from a model with  $v_c(R_m) = 242 \text{ km s}^{-1}$ . Solid pentagons: data points with error bars as in Fig. 4. Open pentagons: data points with error bars as in Fig. 5.

eters. Fig. 8 shows that only potentials with  $v_c(R_m) < 230 \text{ km s}^{-1}$  can be ruled out (they have less than 5 per cent probability according to Fig. 7) from the data with ‘realistic’ error bars. This conclusion is not surprising, given how large these error bars are. However, the surprise is that with ‘idealized’ data there also remains a range of potentials with larger  $v_c(R_m)$  than the true value, in which these data can be fitted no worse than in the true potential: formally,  $235 \text{ km s}^{-1} < v_c(R_m) < 285 \text{ km s}^{-1}$  with 95 per cent confidence.

Fig. 9 shows the predicted kinematics and the intrinsic anisotropy of the four models in Fig. 8 that match the ‘idealized’ data set with  $\chi_{\sigma+h_4}^2 \simeq 1$ . The best-estimate DFs in the potentials with higher  $v_c(R_m)$  than that of the true potential achieve their good fit to the data by the following means. Inside  $r_J$ , they compensate for the higher circular speed of the potential by a larger radial anisotropy, which leads to slightly larger  $h_4$  values (cf. Section 3). This effect is too small to be detected even with the small error bars. Outside  $r_J$ , they compensate by a radially increasing tangential anisotropy. In this way, the velocity dispersions near the edge of the model can be lowered, because the number of high-energy orbits coming in from outside is reduced. Such orbits would contribute large line-of-sight velocities near their



**Figure 9.** Model fits to pseudo-velocity dispersion and  $h_4$  data with ‘idealized’ error bars (see text and Fig. 5). The full line shows the kinematics of the model from which these pseudo-data were generated. The dashed lines show the kinematics of best-estimate DFs in a sequence of potentials with increasing halo circular velocity, all of which fit the data perfectly with  $\chi_{\sigma+h_4}^2 \simeq 1$ . The intrinsic anisotropy of these models (dotted lines) and of the true model is shown in the bottom panel.

pericentres. The  $h_4$  values in the region concerned are also only barely affected, because the extra tangential orbits near radius  $r$  and the lack of higher energy radial orbits from beyond  $r$  nearly compensate.

Clearly, this mechanism will work less well when both the difference between the true and attempted circular velocity curves increase and the radial extent, sampling, and quality of the data points improve. With data even better and more extended than those in Fig. 5, some of the potentials consistent with the presently used model data could probably be ruled out. Thus it appears that if the asymptotic circular velocity is constant, then the value of that constant can be determined accurately with sufficiently high-quality data.

On the other hand, the potentials we have used are very simply parametrized functions. There might well exist more complicated potential functions, the circular velocity curves of which differ in only a restricted range of radii, that would be impossible to distinguish even with *extremely* good data. To test this, we have constructed a truly idealized data set of  $2 \times 70$  data points with ‘idealized’ small error bars as in Fig. 5, evenly spaced in radius and extending to  $6r_j$ . A model with potential differing only in the halo core radius (66 per cent of the true value) was found to fit even these data with  $\chi_{\sigma+h_4}^2=0.97$ , while for a model with different halo core radius *and* different asymptotic circular velocity (by  $30 \text{ km s}^{-1}$ ) a satisfactory fit could not be found.

We draw the following conclusions from these experiments.

(i) Velocity profile data with presently achievable error bars contain useful information on the gravitational potentials of elliptical galaxies. In particular, constant  $M/L$  models are relatively easy to rule out once the data extend beyond  $2R_c$ . The examples that we have studied in detail, tuned to the NGC 6703 data, certainly belong to the less favourable cases, because the dispersion profile is falling.

(ii) The detailed form of the true circular velocity curve is much harder to determine. Conspiracies in the DF are possible, which minimize the measurable changes in the line profile parameters. A good way to parametrize the results is in terms of the circular velocity at the radius of the outermost data point. With presently available data, this can be determined to a precision of about  $+50 \text{ km s}^{-1}$ .

(iii) Better results can be expected from higher quality data, of the sort one would expect from the new class of 10-m telescopes. However, even with such data some uncertainty on the detailed circular velocity curve will remain, regardless of whether or not, in theory, the potential is uniquely determined from the projected DF  $N(r, v_{||})$ . The combination of the type of analysis presented here with other information (e.g. from X-ray data) will therefore give the most powerful results.

## 5 THE ANISOTROPY AND MASS DISTRIBUTION OF NGC 6703

### 5.1 Constant $M/L$ model fits

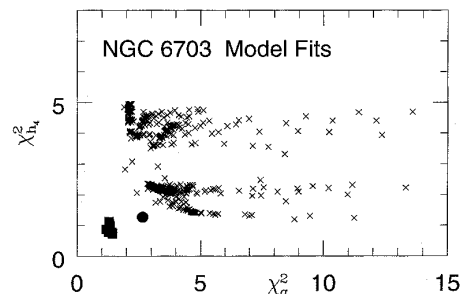
The SB profile of NGC 6703 is well fitted by a Jaffe model. The largest local residuals are  $\lesssim 15$  per cent around  $R \simeq 25$  arcsec,  $\lesssim 10$  per cent around  $R \simeq 40$  arcsec and smaller elsewhere, in particular at  $R > 60$  arcsec. A non-parametric inversion of the surface brightness profile showed that the

deviations from a Jaffe density law are not significantly larger than those quoted in SB. In the curve of growth, which measures the total luminosity inside  $R$ , the residuals are everywhere less than  $\sim 2$  per cent (Fig. 1). As the potential  $\Phi(r)$  is determined by the total mass  $M(<r)$ , we can thus to a good approximation use a Jaffe model for the gravitational potential of the stars. This enables us also to compare our kinematic data with a large set of self-consistent dynamical models from Jeske (1995).

We have fitted all models from this data base to the observational data for NGC 6703, taking the fitted Jaffe radius  $r_j=46.5$  arcsec. All data points from Fig. 1 were included and weighted equally with their individual error bars, except for two adjustments: (i) the error bars of the three  $h_4$  points near 25 arcsec have been set equal to their standard deviation, and (ii) the error bars of the innermost 11  $h_4$  points have been set to twice the measured values. These modifications prevent the total  $\chi_{h_4}^2$  from being dominated by these points, which are of no consequence for the halo of NGC 6703. Moreover, systematic effects may play a role for the innermost data points (Section 2).

The velocity scale of each model is matched to optimize the fit to the  $\sigma$  profile; that to the  $h_4$  profile is already fixed with no extra assumption. The resulting values of  $\chi_{h_4}^2, \chi_{\sigma}^2$  are normalized by the number of fitted data points. In Fig. 10 we plot a  $\chi_{h_4}^2 - \chi_{\sigma}^2$  diagram for all self-consistent models from Jeske (1995). The  $\chi^2$  values are again normalized to the number of data points. Also plotted on the figure are best-estimate models in a number of potentials, constructed with the technique described in Section 4. The optimized self-consistent model (the filled circle) lies on the bounding envelope of the self-consistent models; it has  $\chi_{\sigma}^2=2.6$ ,  $\chi_{h_4}^2=1.3$ . The squares show the normalized  $\chi^2$  values for several models with a dark halo, the rotation curves of which are shown by the full lines in Fig. 9. For most of these,  $\chi_{\sigma}^2 \simeq 1.3$ ,  $\chi_{h_4}^2 \simeq 0.8$ .

Fig. 10 shows that *no self-consistent model will fit the data*: all self-consistent models are clearly separated by a curved



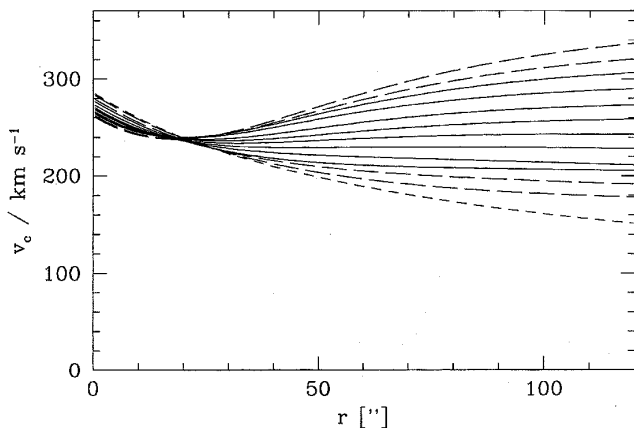
**Figure 10.**  $\chi_{h_4}^2 - \chi_{\sigma}^2$  diagram produced by fitting dynamical models to the kinematic data for NGC 6703. The crosses show fitting results for a variety of anisotropic, self-consistent models of a Jaffe sphere, the density profile of which is a good approximation for the luminous matter in NGC 6703. They fall upwards and to the right of a curved envelope that separates them clearly from a perfect fit, showing that no self-consistent model can simultaneously fit both the dispersion and line profile shape data. The filled circle is the best-estimate self-consistent model in the stars-only potential, obtained with the method of Section 4. The squares show a number of dynamical models with a dark halo; these are consistent with the data.

envelope from the lower left-hand corner of the diagram, which corresponds to a perfect fit. Either the velocity dispersion profile is matched reasonably well, but the line profiles cannot be reproduced, or when the  $h_4$  profile is fitted accurately the dispersion profile is poorly matched. The cure for the discrepancy is to raise *both*  $\sigma$  and  $h_4$  at large radii. Thus, according to Section 3 above, we require extra mass at large  $r$ . NGC 6703 must have a dark halo.

## 5.2 Dynamical models with dark halo

We will now derive constraints on the gravitational potential of NGC 6703 within the framework of the parametric mass model of equations (2)–(4). In doing this, we have in mind the following working picture. The stellar component is assigned a constant mass-to-light ratio  $\Upsilon$ , chosen maximally such that the stars contribute as much mass in the centre as is consistent with the kinematic data. The model for the halo incorporates a constant-density core, and its parameters are chosen such that the halo adds mass mainly in the outer parts of the galaxy if that is necessary. This is similar to the maximum-disc hypothesis in the analysis of disc galaxy rotation curves. Within this framework we can determine the maximum stellar mass-to-light ratio, ask whether it is reasonable, and constrain the halo parameters.

Determining the two potential parameters  $r_c$  and  $v_0$  together with the model velocity scale (equivalent to the stellar mass-to-light ratio) is a three-dimensional problem, and therefore we will proceed in steps. Fig. 11 shows circular rotation curves for the first sequence of mass models for NGC 6703 that we have investigated. In all these models the halo contribution becomes significant only outside 20 arcsec  $\lesssim R_c$ . The corresponding halo core radii mostly lie between 1.2 and 1.7 $r_j$ , i.e. 55 and 80 arcsec. These values are relatively large because of the falling dispersion curve in NGC 6703. This implies that we can determine only one of the halo parameters for this galaxy reliably.



**Figure 11.** Rotation curves for a sequence of gravitational potentials (stars plus dark halo) used in the analysis of NGC 6703. The full lines show rotation curves that are consistent with the NGC 6703 kinematic data inside the 95 per cent confidence boundary at  $\lambda = 10^{-5}$  (open symbols in Fig. 12). The other line styles show rotation curves inconsistent with the data; among these is the constant  $M/L$  model with no dark halo (short-dashed line).

We have chosen the circular velocity  $v_c(R_m)$  at the radius  $R_m$  of the last kinematic data point as this parameter. The sequence in Fig. 11 was constructed so as to vary  $v_c(R_m)$  and the rotation curve outside  $\sim R_c$ , while leaving the central rotation curve nearly invariant. However, when the model velocity scales are optimized in the determination of the best-estimate DFs, the optimal velocity scale is found to correlate inversely with  $v_c(R_m)$ . The rotation curves in Fig. 11 are plotted with their optimal scaling; the central rotation curves are then no longer identical, but become scaled versions of each other.

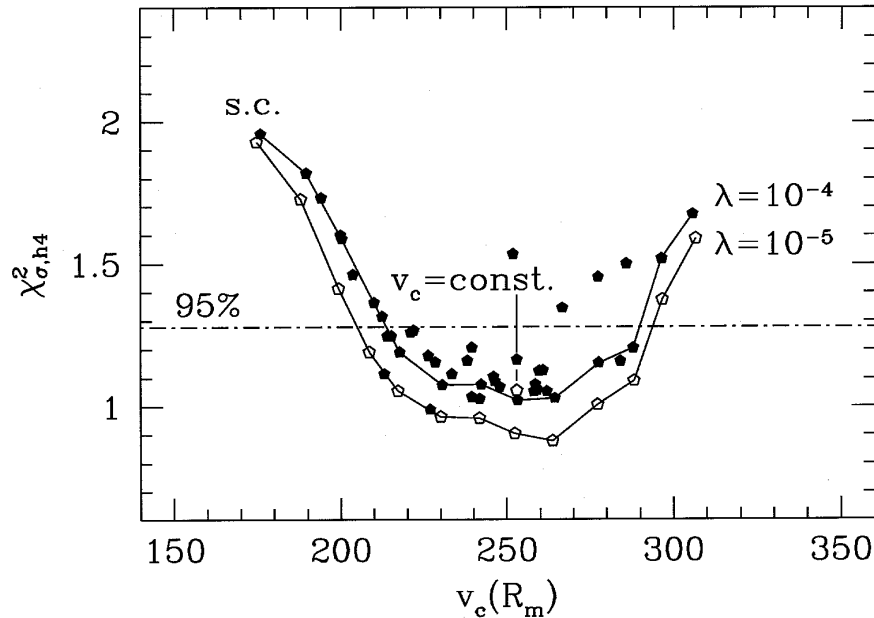
In each of the corresponding potentials, we have constructed the best-estimate DF with optimal velocity scaling, as described in Section 4. We first fit a composite DF to the velocity dispersion and line-profile shape data for a series of values of the unknown velocity scale. From this sequence of models, we determine the optimal value of the velocity scale of the model in this potential, and then recompute the best-estimate model with this velocity scale. In the following, when speaking of the best-estimate DF for a given potential we will always imply that the velocity scale has been optimized in this way. In the fitting procedure we have used a regularization parameter  $\lambda = 10^{-4}$ ; in Section 4.2 this was found to be appropriate for the error bars and sampling of the NGC 6703 data. Compared with the tests in Section 4, we have included a few extra basis functions (total  $K = 20$ ) to resolve the (possibly not real, cf. Section 2) high-frequency structure in the centre of NGC 6703.

Fig. 12 shows, as a function of  $v_c(R_m)$ , the average  $\chi^2$  per  $\sigma$  and  $h_4$  data point of the respective DFs so obtained. The connected solid symbols represent the sequence of potentials corresponding to Fig. 11. The potential with constant mass-to-light ratio appears in the upper left-hand corner in Fig. 12; it is inconsistent with the data by a large margin even for the optimum velocity scale (see Fig. 7). The best-fitting potential with a completely flat rotation curve has an optimal value of  $v_c = v_c(R_m) = 254 \text{ km s}^{-1}$  and  $\chi^2_{\sigma+h_4} = 1.17$ ; thus it does not provide the best possible fit but is consistent with the data. Of the stars plus dark halo models illustrated in Fig. 11, those models in the sequence with  $v_c(R_m) = 210$ – $285 \text{ km s}^{-1}$  have  $\chi^2_{\sigma+h_4} < 1.28$ . This is consistent with the results of Section 4.3, from which we would expect that the NGC 6703 data can be fitted by a range of gravitational potentials. Models in the sequence outside this range of  $v_c(R_m)$  are inconsistent with the kinematic data at the 95 per cent confidence level (cf. Fig. 7).

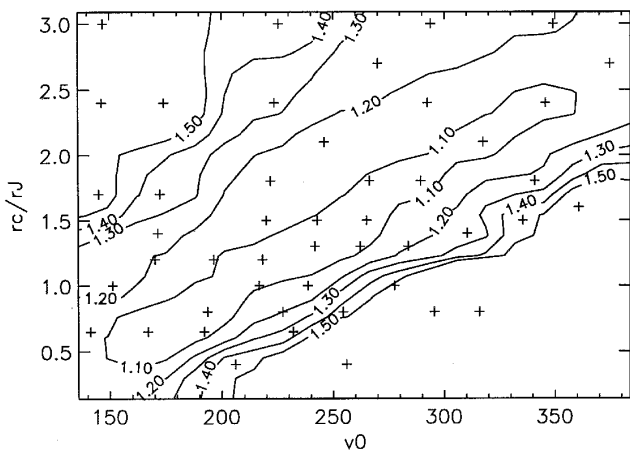
In a second step we have analysed a more complete set of potentials in a suitable part of the  $(v_0, r_c)$  plane (Fig. 13). The fitting results in these potentials are shown by the isolated filled symbols in Fig. 12. Combining these with the previous results allows us to investigate the range of acceptable potentials for NGC 6703 more fully than with the one-dimensional sequence in Fig. 11.

Fig. 13 shows contours of constant  $\chi^2_{\sigma+h_4}$  in the  $(v_0, r_c)$  plane. The most probable potentials lie in a band extending from low  $v_0$  and low  $r_c$  to high  $v_0$  and high  $r_c$ . As already discussed above, it is thus not possible to determine both halo parameters in the NGC 6703 case. However, potentials in the band of most probable  $(v_0, r_c)$  all have circular velocities  $v_c(R_m)$  in the same range,  $250 \pm 40 \text{ km s}^{-1}$ , as before.

In fact, the best-fitting velocity scales of all models in Fig. 13 are found to be such that the resulting values of  $v_c(R_m)$



**Figure 12.** Quality with which the kinematics of NGC 6703 can be fitted in different potentials. The figure shows the average  $\chi^2$  per  $\sigma$  and  $h_4$  data point, of the best-estimate distribution function fitted to the velocity dispersion and line profile data, as a function of the circular rotation velocity of the assumed potential at the observed radius of the last kinematic data point. Filled symbols show best-estimate models derived with the optimal  $\lambda=10^{-4}$  determined in Section 4.2; open symbols represent models derived with  $\lambda=10^{-5}$ . The self-consistent ( $M/L=\text{constant}$ ) and  $v_c=\text{constant}$  models are marked separately. The horizontal dashed line shows the 95 per cent confidence boundary derived from Fig. 7.



**Figure 13.** The  $(v_0, r_c)$  halo parameter plane. Values of  $r_c$  are scaled with respect to  $r_j=46.5$  arcsec. The luminous plus dark matter models investigated are shown as plus signs. The contours show lines of constant  $\chi^2_{\sigma+h_4}$  obtained by interpolating between the model values. Acceptable potentials lie in a band extending from low  $v_0$  and low  $r_c$  to high  $v_0$  and high  $r_c$ . Models at the upper left are ruled out because they do not contain enough mass at large radii. Models at the lower right are ruled out because no satisfactory fit can be found for any constant value of the stellar mass-to-light ratio.

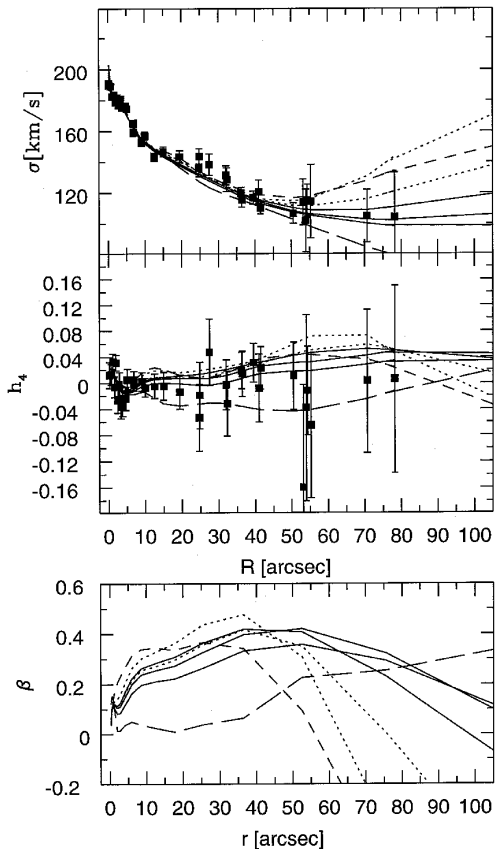
are in the range  $[189, 318]$   $\text{km s}^{-1}$ . The fitting procedure tends to move  $v_c(R_m)$  into the correct range even when no satisfactory DF can be found. For example, some models in the lower right of Fig. 13 with  $\chi^2_{\sigma+h_4} \simeq 1.5$  appear in Fig. 12 at  $v_c(R_m) = 250\text{--}300$   $\text{km s}^{-1}$ , some with  $\chi^2_{\sigma+h_4} \gtrsim 2.4$  (not shown) at  $v_c(R_m) = 280\text{--}340$   $\text{km s}^{-1}$ . All models shown in the upper left of Fig. 13 fall near the line defined by the sequence of

models discussed at the beginning of this section. The fact that  $v_c(R_m)$  varies relatively little for a wide range of luminous matter plus dark halo models suggests that our results, when expressed in terms of this parameter, are not sensitive to the choice of halo model in equations (3) and (4).

From Fig. 12 we conclude that the true circular velocity of NGC 6703 at 78 arcsec is  $v_c(R_m) = 250 \pm 40$   $\text{km s}^{-1}$  (this is the formal 95 per cent confidence interval obtained from the filled symbols, according to Fig. 7). In Section 4.3 we found, however, that most of this indeterminacy is towards large circular velocities, at least for a galaxy with a dispersion curve like that of NGC 6703. By contrast, in potentials with lower values of  $v_c(R_m)$  than the true circular velocity, it quickly becomes impossible to find a satisfactory DF. Based on these results, the lower values in the quoted range of  $v_c(R_m) = 250 \pm 40$   $\text{km s}^{-1}$  appear to be the more probable ones.

The open symbols in Fig. 12 show that the range of potentials consistent with the data is enlarged only slightly when the DF is allowed to be less smooth. These points are from best-estimate models derived in the sequence of potentials of Fig. 11 with  $\lambda=10^{-5}$  instead of the optimal  $\lambda=10^{-4}$ . The resulting curve that connects the  $\lambda=10^{-5}$  models in Fig. 12 surrounds the corresponding  $\lambda=10^{-4}$  curve. Generally, it appeared that, in the models obtained with  $\lambda=10^{-5}$ , the DF came close to zero more easily and more often. The last two facts, when taken together, suggest that, in addition to the data themselves, the positivity constraints on the DF play an important role in determining the boundary of the region in  $(f, \Phi)$  space that is consistent with given kinematic data.

Fig. 14 shows best-estimate models in some of the potentials consistent with the kinematic data (at  $\lambda=10^{-4}$ ). The three full lines are models with  $v_c(R_m) = (218, 231, 242)$  km



**Figure 14.** Dynamical models for the kinematics of NGC 6703 in several luminous plus dark matter potentials, compared with projected velocity dispersion (top panel) and VP shape parameter  $h_4$  (middle panel). The bottom panel shows the intrinsic anisotropy parameter  $\beta(r)$  of the models, with the same line styles: self-consistent model (stars only; long-dashed line),  $v_c = \text{constant}$  model (short-dashed line), three models with  $v_c(78 \text{ arcsec})$  in the lower part of the acceptable range (solid lines) and two models with  $v_c(78 \text{ arcsec})$  in the upper part of this range (dotted lines).

$\text{s}^{-1}$ , the two dotted lines are models with  $v_c(R_m) = (253, 277) \text{ km s}^{-1}$ . Also shown is the best-estimate model in the potential of only the stars with constant  $M/L$  (long-dashed line), and that in the best potential with  $v_c = \text{constant} = 254 \text{ km s}^{-1}$  (short-dashed line).

The dispersion profile of the best-estimate self-consistent model clearly falls below the data at both intermediate and large radii, as does its  $h_4$  profile. From the discussion in Section 3, this is a clear sign of extra mass at large radii. The models including dark halo contributions differ mainly in the outermost parts of the velocity dispersion profile. As expected, those with the highest velocity dispersions at large radii correspond to the potentials with the largest asymptotic circular velocities. This again suggests that with smaller error bars at large radii, and with spatially more extended data, we will be able to significantly narrow down the uncertainties in the halo parameters. The model with constant rotation speed everywhere is constrained tightly by the kinematic data in the central parts, where the  $\rho \propto r^{-2}$  profile is presumably dominated by the stars. It then has some difficulties both with the  $h_4$  values at intermediate radii and the velocity dispersions at large radii.

The lower part of Fig. 14 shows that, in order to match the observed kinematics in a potential with large circular velocity, the DF must become rapidly tangentially anisotropic at the radii of the last data points and beyond. As discussed in Section 4, this is different from the better-known effect of increasing the projected dispersions in a potential that does not have enough mass at large radii, by making the DF more tangentially anisotropic. Here, without the tangential anisotropy, our models would predict values of the velocity dispersion that are too high, because too much mass at large radii is implied. The tangential anisotropy at radii outside those reached by the observations reduces the number of orbits that come into the observed range, orbits which would contribute relatively large line-of-sight velocities near their turning points. In this way, the velocity dispersions can be reduced down to the observed values. Clearly, more spatially extended data would reduce this freedom.

From Fig. 14 we conclude that the stellar distribution function in NGC 6703 is near-isotropic at the centre and then changes to slightly radially anisotropic at intermediate radii ( $\beta = 0.3\text{--}0.4$  at 30 arcsec,  $\beta = 0.2\text{--}0.4$  at 60 arcsec). It is not well-constrained near the outer edge of the data, where formally  $\beta = -0.5\text{--}+0.4$ , depending on the correct potential in the allowed range. However, the models with large asymptotic halo circular velocities shown in Fig. 14 appear less plausible, because they are the models with the most rapidly increasing velocity dispersions outside  $R \approx 60$  arcsec. The same models also show the most rapid increase in tangential anisotropy at and beyond  $R_m = 78$  arcsec, which again appears a priori implausible, because it implies rapid changes in the DF just outside the observed range. The combined signature of both effects is strongly reminiscent of Fig. 9, where it was clearly an artefact of the limited radial range of the data. If this assumption is correct, one would again conclude that lower  $v_c(R_m)$  models in the formal range are favoured. Fig. 15 shows the recovered DF for the potential with  $v_c(R_m) = 231 \text{ km s}^{-1}$  in Fig. 14. Note, however, that all models shown in Fig. 14 except the self-consistent one are formally consistent with the presently available data.

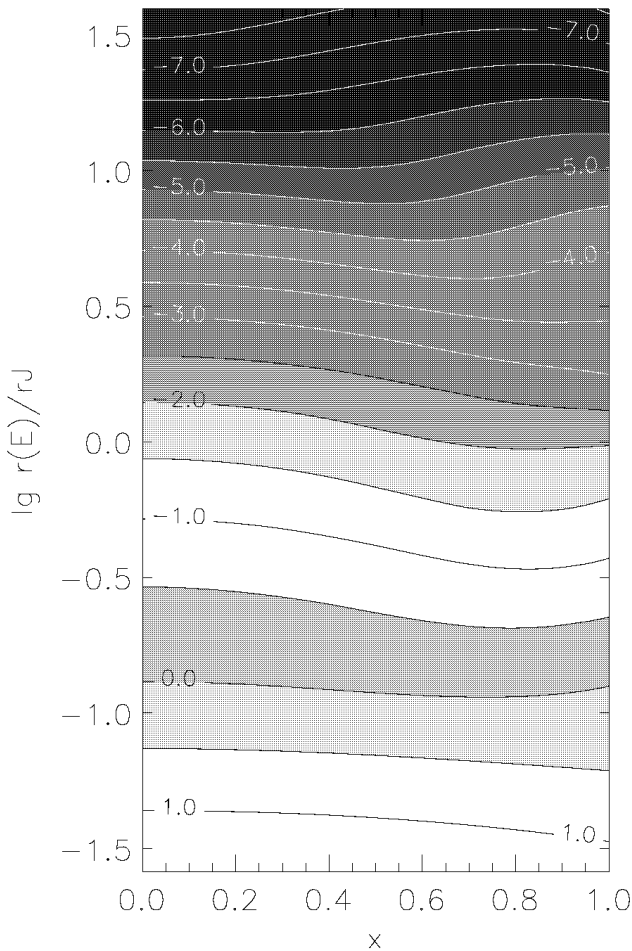
From the constraints on the circular velocity,  $v_c(R_m) = 210\text{--}290 \text{ km s}^{-1}$ , the range in mass inside  $R_m = 13.5 h_{50}^{-1} \text{ kpc}$  is

$$M(< R_m) = 1.6\text{--}2.6 \times 10^{11} h_{50}^{-1} M_\odot, \quad (17)$$

where  $h_{50} \equiv H_0/50 \text{ km s}^{-1} \text{ Mpc}^{-1}$ . The total mass in stars inside this radius is  $8\text{--}9 \times 10^{10} M_\odot$ , assuming constant mass-to-light ratio  $\Upsilon$  and a maximum stellar mass model, and taking an average value from the models that is consistent with the kinematic data. The radial run of the luminous, dark, and total mass is shown in Fig. 16 for the models that span the allowed range according to Fig. 12. After dividing by the luminosity  $L_B(r)$  for the stars, the mass-to-light ratios shown in Fig. 17 result. Between the centre and the last data point  $r = 78 \text{ arcsec} \approx 2.6 R_c$ , the mass-to-light ratio of NGC 6703 rises by a factor of 1.6–3.

### 5.3 Uncertainties

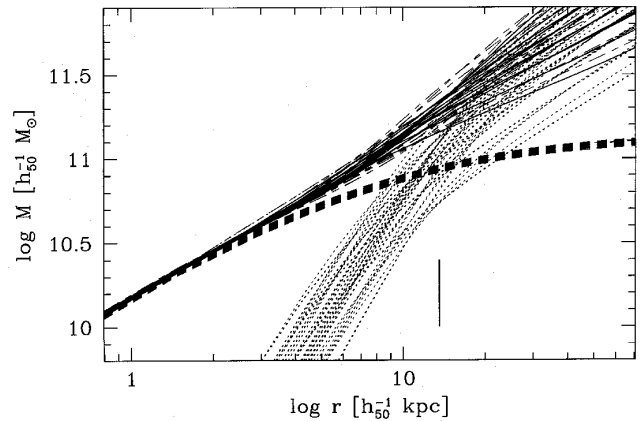
There are a number of possible sources of systematic error that would affect the mass-to-light ratio derived for NGC 6703. Most of the errors so introduced are probably small compared with the considerable uncertainty arising



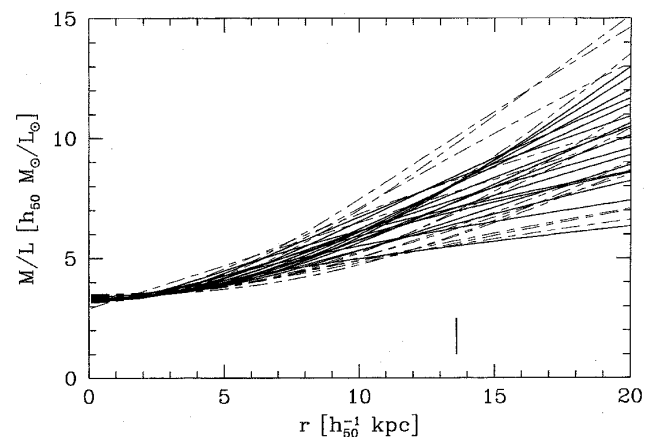
**Figure 15.** Distribution function in the energy-circularity plane derived for NGC 6703, in the luminous plus dark matter potential with  $v_c(R_m) = 231 \text{ km s}^{-1}$  shown as the middle solid line in the upper panel of Fig. 14. Energy is specified by the value of  $r(E)$ , where  $E = \phi(r)$ , in units of the fitted Jaffe radius  $r_j$ , and the DF is given in arbitrary logarithmic units. The last measured kinematic data point is located at  $\log r(E)/r_j = 0.23$ .

from the kinematic measurement errors and limited radial sampling, discussed above. One systematic error on the absolute mass-to-light ratios in NGC 6703 comes from the uncertainty in the distance, although this does not change the ratio of outer to central values. A further systematic effect on this ratio can be introduced by the sky brightness level. If this increased by 2–3 per cent, the fitted  $r_j$  decreases to 35 arcsec. To the extent that the outermost kinematic data point at 78 arcsec (which then moves to large  $r/r_j$ ) is in the flat part of the circular rotation curve, the inferred  $M/L$  changes only to second order because the luminosity inside 78 arcsec remains essentially unchanged. The same is true if the sky value is decreased by 1 per cent, in which case the fitted  $r_j$  increases to 54 arcsec.

In the previous analysis, we have ignored a possible small rotation in the outer parts of NGC 6703 (perhaps  $\sim 20\text{--}30 \text{ km s}^{-1}$  at  $R \gtrsim 50$  arcsec, but the errors are large; cf. Fig. 1). The simplest possible estimate of the effect of this rotation on the derived masses is to replace  $\sigma \simeq 105 \text{ km s}^{-1}$  in this radial range by  $(\sigma^2 + v^2)^{1/2} \simeq 110 \text{ km s}^{-1}$ . This gives a factor



**Figure 16.** Luminous, dark, and total mass as a function of radius for the range of acceptable models of NGC 6703, according to Fig. 12 (short-dashed, dotted and dash-dashed or full lines, respectively). Mass distributions in which a DF with  $\chi^2_{\sigma+h_4} \leq 1.12$  (including 87 per cent of the distribution in Fig. 7,  $1.5\sigma$ ) are coded by full lines, those with  $\chi^2_{\sigma+h_4} \leq 1.28$  (including 95 per cent,  $2.0\sigma$ ) by dash-dashed lines. The vertical line denotes the position of the last kinematic data point. At this radius, the luminous mass fails by at least a factor of 1.6.



**Figure 17.** The  $B$ -band mass-to-light ratio of NGC 6703. The solid and dash-dashed lines (coding as in Fig. 16) are derived from the dynamical models that span the range of acceptable  $v_c(R_m)$  in Fig. 12. The central mass-to-light ratio is  $\Upsilon = 3.3$ , while that at the position of the last kinematic data point at 78 arcsec (vertical bar) is in the range  $\Upsilon = 5.3\text{--}10$ .

of 1.1, neglecting changes in the model structure that would result because the central kinematics remain unchanged.

Next we consider the possibility that NGC 6703 may contain a face-on extended disc (see de Vaucouleurs, de Vaucouleurs & Corwin 1976). From an  $R^{1/4}$  law plus disc decomposition, we estimate that the contribution of such a disc in the region where we model the kinematics could be up to 10–20 per cent. In this case we expect the velocity dispersion to be decreased and the  $h_4$  coefficient to become more positive where the disc contributes significant light (Dehnen & Gerhard 1994; see NGC 4660 as an example in BSG), most likely in the outer parts.

Similarly, it is conceivable that NGC 6703 is in reality slightly triaxial, and is seen from a special direction so as to appear E0. The likelihood of this is less the more triaxial the intrinsic axial ratios are; thus slightly triaxial shapes are the most plausible ones. Again, this will imply some extra loop orbits seen nearly face-on, similarly increasing  $h_4$  and decreasing  $\sigma$ .

In both cases, we therefore expect the spherical component in NGC 6703 to have lower  $h_4$  and larger  $\sigma$  than the measured values. A similar analysis of such kinematics would, according to the discussion in Section 3, lead to a model with greater tangential anisotropy at large radii, with the mass distribution less affected. Recall that decreasing the mass at large  $r$  in a spherical model lowers *both*  $\sigma$  and  $h_4$ .

## 6 DISCUSSION AND CONCLUSIONS

This study is part of an observational and theoretical programme aimed at understanding the mass distribution and orbital structure in elliptical galaxies. In the following we first discuss general results on potential and anisotropy determination, and then proceed to the specific case of NGC 6703.

### 6.1 Velocity profiles, anisotropy and mass

The analysis of the VPs of simple dynamical models in Section 3 has broadly confirmed the conclusions of G93. At large radii, where the luminosity profile falls rapidly, the VPs are dominated by the stars at the tangent point. Radially (tangentially) anisotropic DFs can then be recognized by more peaked (more flat-topped) VPs with more positive (negative)  $h_4$  than for the isotropic case. Increasing  $\beta$  at constant potential thus lowers  $\sigma$  and increases  $h_4$ . On the other hand, increasing the mass of the system at large  $r$  at constant anisotropy increases *both* the projected dispersion and  $h_4$ . This suggests that by modelling  $\sigma$  and  $h_4$  both mass  $M(r)$  and anisotropy  $\beta(r)$  can be constrained.

In practical applications, such an analysis is complicated by a number of factors. Radial orbits at large radii may lead to increased central velocity dispersions and flat-topped central VPs (already pointed out by Dejonghe 1987). The former effect can be compensated for by a decrease in the stellar mass-to-light ratio. The latter effect is independent of this, but can be compensated for by changes to the distribution function in the inner parts of the galaxy (as in a number of cases studied in Sections 4 and 5). A more serious uncertainty is introduced by the possibility of significant gradients in the orbit population across the radii of interest. For example, a population of high-energy radial orbits with pericentres in a limited radial range may mimic tangential anisotropy there. In many cases it will be possible to exclude such a population of orbits by examining its effects on the VPs at exterior radii, i.e. by simultaneously analysing a number of observed VPs. This is least possible, however, at precisely the largest observed radii, where mass determination is most interesting. Thus this chain of argument suggests (correctly – see below) that the largest uncertainty in determining masses and anisotropies in ellipticals from VP data is the finite radial extent of these data.

To analyse realistic data we have constructed an algo-

rithm by which the distribution function and potential of a spherical galaxy can be constrained directly from its observed  $\sigma$  and  $h_4$  profiles. To assess the significance of the results obtained, we have tested the algorithm on Monte-Carlo-generated data sets tuned to the spatial extent, sampling and observational errors as measured for NGC 6703. From such data, the present version of the algorithm recovers a smooth spherical DF,  $\sim 70$  per cent of the time, to an rms level of better than  $\sim 12$  per cent inside three times the radius of the outermost kinematic data point.

We have used this algorithm to study quantitatively the degree to which the gravitational potential can be determined from such data. Our main conclusion is that velocity profile data with presently achievable error bars already constrain the gravitational potentials of elliptical galaxies significantly. In particular, constant  $M/L$  models are relatively easy to rule out once the data extend beyond  $2R_c$ . The examples that we have studied in detail, tuned to the NGC 6703 data, certainly belong to the less favourable cases, because in this galaxy the dispersion profile is falling.

A good way to parametrize the results is in terms of the true circular velocity  $v_c(R_m)$  at the radius of the outermost data point,  $R_m$ . With the presently available data,  $v_c(R_m)$  can be determined to a precision of about  $\pm \lesssim 50 \text{ km s}^{-1}$ . This will improve when high-quality data at several  $R_c$  become available, of the kind expected from the new class of 10-m telescopes. Apart from the fact that smaller error bars will decrease the formally allowed range in  $v_c(R_m)$ , tests show that this range often includes high  $v_c(R_m)$  models, which become rapidly tangentially anisotropic just outside the data boundary. These (not very plausible) models can be eliminated with data extending to larger radii.

On the other hand, the detailed form of the true circular velocity curve is much harder to determine than  $v_c(R_m)$ . Conspiracies in the DF are possible that minimize the measurable changes in the line profile parameters. Our tests showed that two potentials differing by just the value of the halo core radius could not be distinguished even with very good data out to  $6R_c$ . Some uncertainty will therefore remain in practice, regardless of whether or not in theory the potential is uniquely determined from the projected DF  $N(r, v_{\parallel})$ .

A similar picture holds for the related determination of the anisotropy of the DF. For the present error bars in the data,  $\beta(r)$  is relatively well-determined out to about half the limiting radius of the observations. Near the edge of the data, uncertainties can be large, depending on the gravitational potential [recall that in a fixed spherical potential the DF is uniquely determined by the complete projected DF  $N(r, v_{\parallel})$ ]. Again, the unknown nature of the orbits beyond the last data point has a large part in this uncertainty.

Because the largest uncertainties in determining masses and anisotropies from VPs occur near the outer radial limit of these data, the combination of the type of analysis presented here with other information (e.g. from X-ray data) will be particularly powerful.

### 6.2 The dark halo of NGC 6703

Fig. 10 shows that *no self-consistent model will fit the kinematic data for NGC 6703*. Our non-parametric best-esti-



mate self-consistent model is inconsistent with the data at the  $>99$  per cent level (Figs 7 and 12). With self-consistent models, either the velocity dispersion profile is matched reasonably well, but then the line profiles cannot be reproduced, or, when the  $h_4$  profile is fitted accurately, the dispersion profile is poorly matched. The cure for the discrepancy is to raise *both*  $\sigma$  and  $h_4$  at large radii. Thus, as discussed above, we require extra mass at large  $r$ . NGC 6703 must have a dark halo.

We have next derived constraints on the parameters of this halo as follows. The luminous component is assigned a constant mass-to-light ratio  $\Upsilon$ , chosen maximally such that the stars contribute as much mass in the centre as is consistent with the kinematic data. Our parametric model for the halo incorporates a constant-density core, and its parameters (core radius  $r_c$  and asymptotically constant circular velocity  $v_0$ ) are chosen such that the halo adds mass mainly in the outer parts of the galaxy if that is necessary. We call these models *maximum stellar mass* models (analogous to the maximum disc hypothesis in the analysis of disc galaxy rotation curves).

We find that maximum stellar mass models in which the luminous component provides nearly all the mass in the centre fit the data well. In these models, the total luminous mass inside the limiting observational radius  $R_m = 78$  arcsec  $= 13.5h_{50}^{-1}$  kpc is  $9 \times 10^{10}h_{50}^{-1} M_\odot$ , corresponding to a central  $B$ -band mass-to-light ratio  $\Upsilon = 3.3h_{50} M_\odot/L_\odot$ . According to Worthey's (1994) models, this is a rather low value for the stellar population of an elliptical galaxy and would point to a relatively low age (5 Gyrs) and/or low metallicity (less than solar). However, the galaxy has a colour  $(B - V)_0 = 0.93$  and a central line index  $Mg_2 = 0.280$  (Faber et al. 1989) which are typical for ellipticals of similar velocity dispersion.

A larger value of  $H_0$  could increase the  $M/L$  value and alleviate the demands on the stellar populations. However, the distance used here (36 Mpc) includes a correction for the larger inferred peculiar velocity of the galaxy. If we had used a distance based on the larger radial velocity in the cosmic microwave background (CMB) frame, our derived  $M/L$  would be even lower. It is also implausible that the low central value of  $M/L$  stems from the contribution of a young stellar population in a disc component, which we estimate cannot be larger than 20 per cent of the total light (see above). We therefore conclude that the dark halo in NGC 6703 is unlikely to have higher central densities than inferred from our maximum stellar mass models, because otherwise the  $M/L$  of the stellar component would be reduced to implausibly small values.

In a recent preprint, Rix et al. (1997) have analysed the velocity profiles of the E0 galaxy NGC 2434 with a linear orbit superposition method. This galaxy provides an interesting contrast to NGC 6703 because it has an essentially flat dispersion profile. Its kinematics are likewise inconsistent with a constant  $M/L$  potential, but are well-fitted by a model with  $v_c = \text{constant}$ . This can be interpreted as a maximum stellar mass model in the sense defined above, in which the luminous component with maximal  $\Upsilon$  contributes most of the mass inside  $R_c$ . The kinematics of NGC 2434 are also well-fitted by a range of specific, cosmologically motivated mass models which, if applicable, would imply lower  $\Upsilon$  and significant dark mass inside  $R_c$ . In

NGC 6703, a model with  $v_c = \text{constant}$  is formally consistent with the present data (within  $2\sigma$ ), but it is not a very plausible fit at large  $R$  and requires large anisotropy gradients, between 40 and 70 arcsec. It will be interesting to see whether future studies confirm differences between the shapes of the true circular velocity curves of elliptical galaxies.

Because of the falling dispersion curve in NGC 6703, we can determine only one of the halo parameters. (The halo parameters of the most probable potentials lie in a band extending from low  $v_0$  and low  $r_c$  to high  $v_0$  and high  $r_c$ .) The circular velocity  $v_c(R_m)$  at the data boundary is relatively well-determined for all these models, however. We therefore find (Fig. 12) that the true circular velocity of NGC 6703 at 78 arcsec is  $v_c(R_m) = 250 \pm 40$  km s $^{-1}$  (formal 95 per cent confidence interval). Tests on pseudo-data have shown that this range often includes high  $v_c(R_m)$  models, which become rapidly tangentially anisotropic just outside the data boundary. Such models may not be very plausible, so the lower values in the quoted range of  $v_c(R_m) = 250 \pm 40$  km s $^{-1}$  may be the more probable ones.

At  $R_m = 78$  arcsec  $= 13.5h_{50}^{-1}$  kpc the total mass enclosed is therefore  $M(<R_m) = 1.6\text{--}2.6 \times 10^{11}h_{50}^{-1} M_\odot$ , and the integrated mass-to-light ratio out to this radius is  $\Upsilon = 5.3\text{--}10$ , corresponding to a rise from the centre by at least a factor of 1.6. We have already noted that NGC 6703 is an unfavourable case because of its falling dispersion curve. The fact that relatively small variations in  $\Upsilon$  can nonetheless be detected shows the power of the method. Note that a scheme based on the analysis of the line-of-sight velocity dispersions alone (Binney, Davies & Illingworth 1990; van der Marel 1991) would conclude that constant mass-to-light ratio models can provide good fits.

The stellar distribution function in NGC 6703 is near-isotropic at the centre and then changes to slightly radially anisotropic at intermediate radii ( $\beta = 0.3\text{--}0.4$  at 30 arcsec,  $\beta = 0.2\text{--}0.4$  at 60 arcsec). It is not well-constrained near the outer edge of the data, where formally  $\beta = -0.5$  to  $+0.4$ , depending on the correct potential in the allowed range. Models near the lower end of this range may be consistent with the data only because of the limited radial extent of the measurements.

### 6.3 Conclusions

In summary, we have shown that the mass distribution  $M(r)$  and anisotropy structure  $\beta(r)$  for spherical galaxies can *both* be constrained from VP and velocity dispersion measurements. NGC 6703 must have a dark halo, contributing about equal mass at  $2.6R_c$  to that from stars. The circular velocity at the last kinematic data point (78 arcsec) is constrained to lie in the range  $250 \pm 40$  km s $^{-1}$  at 95 per cent confidence. The anisotropy of the stellar orbits changes from near-isotropic at the centre to slightly radially anisotropic at intermediate radii, and may be either radially or tangentially anisotropic at 78 arcsec. With more extended and more accurate data it will be possible to narrow down these uncertainties considerably.

If the results for this galaxy are typical, they suggest that also in elliptical galaxies the stellar mass dominates at small radii, and the dark matter begins to dominate at radii around 10 kpc. It is important to obtain extended kinematic

data and to perform a similar analysis for a number of elliptical galaxies. When we know the systematics and the spread in the circular velocity curves and anisotropy profiles for a sample of ellipticals, we will have an important new means for testing the currently popular formation theories.

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## APPENDIX A: LIBRARY OF ANISOTROPIC SPHERICAL DISTRIBUTION FUNCTIONS

To understand the connection between anisotropy structure and observable line profile shapes we have constructed a number of spherical distribution functions of the quasi-separable form (Gerhard 1991, hereafter G91)

$$f(E, L) = g(E)h(x), \quad (\text{A1})$$

where the variable  $x$  depends on both energy and angular momentum:

$$x = \frac{L}{L_0 + L_c(E)}. \quad (\text{A2})$$

$L_0$  is an angular momentum constant, or, equivalently, an anisotropy radius times a characteristic velocity;  $L_c(E)$  is the angular momentum of the circular orbit at energy  $E$ . DFs of the form given by equation (A1) have the following properties.

(i) The circularity function  $h(x)$  has the effect of shifting stars between orbits of different angular momenta on surfaces of constant energy, while  $g(E)$  controls the distribution of stars between energy surfaces.

(ii) For most bound stars  $L < L_c(E) \ll L_0$ ; the model therefore becomes isotropic in the centre unless  $L_0 = 0$ .

(iii) For loosely bound stars  $L \sim L_c(E) \gg L_0$ , i.e. the angular momentum distribution becomes a function of circularity  $L/L_c(E)$ , which is one-to-one related to orbital eccentricity. Outside the anisotropy radius, the DF (A1) therefore corresponds to an energy-independent orbit distribution with constant anisotropy, radial or tangential.

In these models,  $h(x)$  is an assigned function built into the model to achieve the desired anisotropy (orbit distribution). Radially biased distribution functions correspond to circularity functions  $h(x)$  decreasing with  $x$ ; for example

$$h(x) = h_z(x) \equiv (1 - x^2)^\alpha. \quad (\text{A3})$$

The family can also be used to construct tangentially anisotropic models, such that

$$h_{c,z}(x) = c + (1 - c)[1 - (1 - x^2)^\alpha]. \quad (\text{A4})$$

In these tangentially anisotropic models, one cannot choose  $h(0) = 0$  unless  $L_0 = 0$ , otherwise the density at  $r = 0$  would be zero. Of course, other forms for  $h(x)$  are possible, such as Gaussians.

Given the assigned function  $h(x)$ , the integral equation for  $\rho(r)$  in terms of  $f(E, L^2)$  is solved for the derived func-

tion  $g(E)$ ; see G91 and Jeske (1995). Fig. 2 shows line profile shape parameters for representative DFs constructed in this way. Fig. 3 shows the anisotropy profiles of two sets of tangentially anisotropic models. Here the density of stars has been taken to be that of a Jaffe sphere, and the potential in which the stars orbit is either the self-consistent potential or one of the mixed stars plus halo potentials used in Sections 4 and 5. Sequences like that in Fig. 3 are used as basis functions in the non-parametric analysis in Section 5.

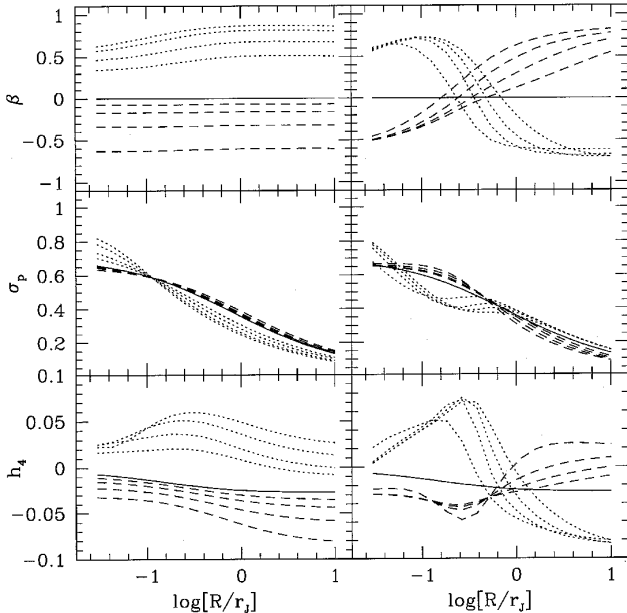
Models, the anisotropy of which changes from radial to tangential or vice versa, were constructed by linearly combining the above circularity functions with energy-dependent coefficients. In this way one obtains DFs of the more general form

$$f(E, L) = g(E)h(E, x). \quad (\text{A5})$$

For self-consistent Jaffe models we used energy-dependent coefficients  $\mu(E)$  of the following form:

$$\mu(E) = \frac{1}{2} + \frac{1}{\pi} \arctan\left[\frac{k}{2} \frac{E^2 - \bar{E}^2}{E^2}\right]. \quad (\text{A6})$$

The parameters  $\bar{E}$  and  $k$  determine the orbital energy near which the anisotropy transition occurs, and the width of the transition. A similar function  $\mu(E)$  was used for models in halo potentials. Figure A1 shows the intrinsic and projected properties of a number of DFs of this kind, constructed in the self-consistent potential of a Jaffe sphere. Notice the



**Figure A1.** Anisotropy parameter  $\beta(r)$ , projected velocity dispersion  $\sigma_p$ , and line profile shape parameter  $h_4$  for several families of anisotropic DFs. Left: scale-free radially anisotropic (dotted lines) and tangentially anisotropic (dashed lines) models in the potential of a self-consistent Jaffe sphere. For these models, circularity functions of the type (A3) and (A4) were used, with different  $\alpha$  and  $c$ , respectively. The isotropic model is shown for reference (solid line). Right: families of models with anisotropy changing from radial to tangential (dotted lines) or from tangential to radial (dashed lines). These were constructed from the same circularity functions and weight factor as in equation (A6).

wide variety of kinematical profiles that can be constructed in this way.

## APPENDIX B: TRANSFORMING TO LINEAR KINEMATIC DATA

In velocity line profile measurements, the depth of an absorption line in a spectral resolution element is assumed to be proportional to the number of stars with line-of-sight velocities corresponding to this wavelength interval. The line-of-sight velocity distribution measured from the line profiles is a discretized function linearly related to the underlying DF (cf. equation 1). This linearity is lost when line profile measurements are represented by the quantities  $v$ ,  $\sigma$ ,  $h_3$ ,  $h_4$ : these quantities are obtained by a least-squares fit of a Gauss–Hermite series to the observed line profile.

To re-express the observed kinematics in terms of quantities that depend linearly on  $f$  we proceed as follows. Consistent with the assumption of spherical symmetry, we set the mean streaming velocity  $\bar{v}$  and all odd velocity-profile moments to zero. Next, we obtain an estimate for the velocity dispersion (second moment),  $\bar{\sigma}^2$ , by integrating over the line profile specified by  $(\sigma, h_4)$ ; for negative  $h_4$  until it first becomes zero. For small  $h_4$ , the linear correction formula  $\bar{\sigma} = \sigma(1 + \sqrt{6}h_4)$  holds (van der Marel & Franx 1993); this correction results in  $\bar{\sigma} > \sigma$  for peaked profiles with  $h_4 > 0$ . The numerical correction from integrating over the velocity profile also has this property (BSG).

From the measured  $h_4(R_i)$ , we compute new even Gauss–Hermite moments  $s_n(R_i; \bar{\sigma})$  by expanding the series

$$\mathcal{L}(R_i, v_{\parallel}) = \mathcal{L}_0 \sum_{j=0,4} h_j(R_i) H_j(x) \exp(-x^2/2) \quad (\text{B1})$$

$$(h_0 = 1, h_1 = h_2 = h_3 = 0) \text{ with } x \equiv v_{\parallel}/\sigma(R_i) \text{ as}$$

$$\mathcal{L}(R_i, v_{\parallel}) = \sum_{n=0,2,4,\dots} N_n s_n H_n(\tilde{x}) \exp(-\tilde{x}^2/2), \quad (\text{B2})$$

where  $\tilde{x} \equiv v_{\parallel}/\bar{\sigma}(R_i)$ . Here  $H_n$  are Hermite polynomials, the  $N_n$  are normalization constants (G93) and  $\bar{\sigma}(R_i)$  are fiducial scaling velocities, generally different from  $\bar{\sigma}(R_i)$ . In practice, we have found it convenient to take for  $\bar{\sigma}(R_i)$  the velocity dispersions  $\sigma_{\text{iso}}(R_i)$  of the isotropic model in the given potential  $\Phi(r)$  with the same stellar density as the galaxy being analysed. The  $s_n(R_i, \bar{\sigma})$  are estimates for the Gauss–Hermite moments, related to the true velocity profile by

$$s_n(R_i; \bar{\sigma}) = (2^{n-1}n!)^{1/2} \times \int_{-\infty}^{\infty} dv_{\parallel} H_n(v_{\parallel}/\bar{\sigma}) \exp(-v_{\parallel}^2/2\bar{\sigma}^2) \mathcal{L}(R_i, v_{\parallel}). \quad (\text{B3})$$

For a theoretical model, we obtain corresponding moments by inserting the projected DF  $N(R_i, v_{\parallel})$  from equation (1) into equation (B3) instead of  $\mathcal{L}(R_i, v_{\parallel})$ , using the same  $\bar{\sigma}(R_i)$ . Clearly, the new  $s_n$  moments of the composite model are linear in  $a_k$ , i.e.

$$s_n(R_i; \bar{\sigma}) = \sum_{k=1, K} a_k s_n^{(k)}(R_i; \bar{\sigma}), \quad (\text{B4})$$

with the  $s_n^{(k)}$  corresponding to the respective  $f_k$ .