### Modelling the dwarf Spheroidals: Cores, Cusps and orbital structure

Nicola Cristiano Amorisco

with Prof. Wyn Evans

Institute of Astronomy University of Cambridge

7th June 2012 Workshop CIAS Meudon 2012 Dark Haloes in dSphs

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## dSphs and Near-Field Cosmology



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## dSphs and GCs



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### dSphs and GCs



mass does not follow light:

dSphs are the most dark matter dominated systems known a unique testing ground for cosmology

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# **Dynamics and Near-Field Cosmology**

### dark halo

- mass profile?
- core or cusp?
- concentration?
- phase-space structure?

### stellar populations

- orbital structure?
- evolutionary history?
- accreted/expelled gas?
- baryonic feedback?

Because of degeneracies, it has not proved easy to address these questions. Dark Haloes in dSphs

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### Mass-Anisotropy degeneracy



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### luminous density $\rho_*$

• gravitational potential  $\Phi$ 

Solve the Jeans' eq. and calculate the LOS kinematic profil

$$\sigma_{LOS}^2(R) = \frac{2}{\Sigma_*(R)} \int_R^\infty r dr \left(1 - \beta \frac{R^2}{r^2}\right) \frac{\rho_* \sigma_r^2}{\sqrt{r^2 - R^2}}$$

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 $\begin{array}{ll} \text{luminous density} & \rho_* \\ \text{halo potential} & \Phi \\ \text{orbital structure} & \beta \end{array}$ 

are NOT independent quantities

### Not all combinations guarantee $f \ge 0!$

Plummer or King any cusped halo isotropic

is NOT a physical mode (Ciotti & Pellegrini 1992, An & Evans 2009)

this introduces an artificial degeneracy
 no prediction for Σ<sub>\*</sub>

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### # 2 - Multiple stellar populations



Sculptor: Tolstoy et al. 2004

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# # 3 - Unused data: Line Profiles



LOSVDs are the imprint of orbital structure.

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# Cores from multiple stellar pops



Jeans' modelling in Sculptor for MP and MR stellar populations.

- Marginal preference for a core
- NFW fit still statistically acceptable

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### **Cores from mass estimators**



### Mass estimator

$$M(R_h) \approx K \frac{R_h \sigma_{los}^2}{G}$$

Walker et al. 2009, Wolf et al. 2010, Amorisco & Evans 2011 Dark Haloes in dSphs

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### **Cores from mass estimators**



- significant improvement: metallicity distribution
- not a dynamical model: detailed information on the halo? orbital structure?

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### Cores from phase-space modelling

We adopt the observables as in Battaglia et al. 2008



MP: [Fe/H] < -1.7 MR: [Fe/H] > -1.5

The phase-space approach gives a prediction for both kinematics and photometry.

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What level of complexity do we need?





The tidal cut is incompatible with the MR stellar population.

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The energy-cut in phase-space of the MR population is not tidal in origin. Possibly depending on the original gas distribution.

Hence, we allow for different truncations.



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Also, isotropic models are not enough, and a mild radial anisotropy is needed:



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Michie-King DFs:

$$f_* = f_{\rm MK}(E,L) \propto \exp\left(\frac{-L^2}{2r_{\rm a}^2\sigma^2}\right) \left\{ \exp\left[\frac{\Phi(r_{\rm t}) - E}{\sigma^2}\right] - 1 \right\}$$

These can cover a wide range of dependences on Energy

- from the isothermal limit:  $f_* \sim \exp(-E)$
- up to strongly truncated systems:  $f_* \sim (\Phi(r_t) E)$ This is equivalent to an ordering in the ratio  $r_t/R_h$ .

Isotropic in the center with an adjustable degree of radial anisotropy at larger radii.

Each stellar population has 3 free parameters:

$$(R_{\rm h}, r_t, \bar{\beta} \equiv \beta(R_{\rm h}))$$

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### Halo density profile

the usual NFW profile

$$\rho(r) = \frac{\rho_0}{\left(\frac{r}{r_0}\right) \left(1 + \frac{r}{r_0}\right)^2}$$

an intermediate cusp

$$\rho(r) = \frac{\rho_0}{\left(\frac{r}{r_0}\right)^{1/2} \left(1 + \frac{r}{r_0}\right)^{3/2}}$$

a cored halo

$$\rho(r) = \frac{\rho_0}{\left(1 + \left(\frac{r}{r_0}\right)^2\right)^{3/2}}$$

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### **Maximum-likelihood Analysis**

For each parametrization of the dark halo we have 8 free parameters:

 $(r_0, 
ho_0; \hat{R}_{\mathrm{h}}, \hat{r}_{\mathrm{t}}, ar{eta}; \hat{R}_{\mathrm{h}}, \hat{r}_{\mathrm{t}}, ar{eta})$ 

We consider the likelihood

$$L_{\text{tot}} = L_{\text{MP}}(r_0, \rho_0; \hat{R}_{\text{h}}, \hat{r}_{\text{t}}, \bar{\beta}) \cdot L_{\text{MR}}(r_0, \rho_0; \hat{R}_{\text{h}}, \hat{r}_{\text{t}}, \bar{\beta})$$
$$\chi_{\text{tot}} = \chi_{\text{MP}}^2 + \chi_{\text{MR}}^2$$

For each population

$$\chi^2 = \chi_{\Sigma}^2 + \chi_{\sigma}^2$$

$$\chi^2_{\rm MP} = \chi^2_{\Sigma} + \chi^2_{\sigma} + \chi^2_{r_{\rm t}}$$

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### **Cored Halo**



### Amorisco & Evans 2012



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## **Cusped Halo**



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### **Mass Profile**

Both NFW halo and cored halo have approximately the same mass profile in the range

 $200 \text{pc} \le r \le 1.2 \text{kpc}$ 



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### **Cores from the Virial Theorem**

 $\frac{\text{MP } 2K_{los} + W_{los} = 0}{\text{MR } 2K_{los} + W_{los} = 0}$ 

both populations satisfy the Virial Theorem

$$K_{los} \propto \int_{0}^{\infty} R dR \ \Sigma_{*} \ \sigma_{los}^{2}$$
 $W_{los} \propto \int_{0}^{\infty} R dR \ \Sigma_{*} \ \int_{0}^{R} r^{2} dr \ \frac{\rho_{dm}}{\sqrt{R^{2} - r^{2}}}$ 

A fundamental constrain based on measured quantities only: no dependence on the orbital structure  $\beta$ !

Agnello & Evans 2012

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### **Cores from the Virial Theorem**

$$\rho_{dm} = \rho_0 \ (\epsilon^2 + r^2/r_s^2)^{-1/2} (1 + r^2/r_s^2)$$



- Deviations from spherical symmetry
- Self-gravity contributions

Any core smaller than  $\approx 120$  pc is not compatible (less than  $2\sigma$ ) with the VT.

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## **Line Profiles**

- Line Profiles constrain the orbital structure;
- break degeneracies: mass profile at the center and at large radii;
- constrain feasible formation scenarios.

LOSVDs are usually assumed to be Gaussians. Deviations are measured by using Gauss-Hermite expansions.



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- Gauss-Hermite series are best suited for continuous data;
- non-uniform observational uncertainties;
- limited sampling limits the accuracy.



Accuracy Limits: Standard Deviation for  $h_3$  and  $h_4$  at sample size N.

For N smaller than 200, noise may be larger than expected signal.

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A tracer  $v_i \pm \delta_i$  is associated with the velocity distribution  $\mathscr{L} * \mathscr{G}(\delta_i)$ , rather than with the intrinsic  $\mathscr{L}$ .



Attenuation by observational uncertainties.

On the contrary, a Bayesian implementation directly measures the intrinsic distribution  $\mathscr{L}$ .

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# A Bayesian Framework

Using all available information

- the velocities  $v_i$ ;
- the uncertainties  $\delta_i$ ;
- the probabilities of membership  $p_i$ .

$$L(\vec{\Theta}) = \prod_{i=1}^{N} p_i \left[ \mathscr{L}(\vec{\Theta}) * \mathscr{G}(\delta_i) \right] (v_i)$$
$$\vec{\Theta} = \{\mu, \sigma\} \cup \vec{\Theta}_{\rm sh} = \{\mu, \sigma, s, a\}$$

no binning in velocity space;

- reliable uncertainties for any parameter;
- lacksquare intrinsic distribution  ${\mathscr L}$  recovered.

Amorisco & Evans 2012

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### Symmetric deviations: s



The symmetric distributions:  $\mathscr{L}(s; v)$ .

Constructed by using the simple model

$$f(v_r, |\vec{v}_t|) \propto |\vec{v}_t|^{-2s} \exp\left[-\frac{v_r^2 + |\vec{v}_t|^2}{2\sigma_r^2}\right]$$

with anisotropy  $\beta = s$  and los direction  $\varphi(s)$ .

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### Asymmetric deviations: a



Asymmetry is driven by a suitable tranformation of the symmetric family:

$$\begin{array}{l} \mathscr{L}(s,a;v) \equiv \\ \mathscr{L}(s;X(s,a;v)) \end{array} \\$$

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The asymmetric distributions:  $\mathscr{L}(s, a; v)$ .

### Performance

Does it work any better?



Comparing Accuracy: Standard Deviation for  $h_3$  and  $h_4$  at a given sample size N.

The relative gain in accuracy is significant even with no observational uncertainties or probabilities of membership.

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# **Assessing Statistical Significance**



Comparing the maximum likelihood

$$\bar{L} = \prod_{i=1}^{N} p_i \left[ \mathscr{L}(\vec{\Theta}) * \mathscr{G}(\delta_i) \right] (v_i)$$

with the *average* likelihood for the same parameters

$$\langle \prod_{i=1}^{N} p_i \, \mathscr{L} \ast \mathscr{G} \rangle = \prod_{i=1}^{N} p_i \, \int \left[ \mathscr{L} \ast \mathscr{G}(\delta_i) \right]^2$$

and the natural scatter induced by sample size

$$\chi = \left(\bar{L} - \langle L \rangle\right) / \text{StD}\left[\langle L \rangle\right]$$

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What if this family is not general enough?

### Carina dSph

758 giants with  $p_i \geq 0.9$ ;  $\langle \delta \rangle / \sigma \approx 0.53$ 



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### Sextans dSph

424 giants with  $p_i \geq 0.9$ ;  $\langle \delta \rangle / \sigma \approx 0.42$ 



Profiles in circular annuli for the Sextans dSph;  $R_{core} \approx 16.6$ arcmin. Data from Walker et al. 2009.

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## Sculptor dSph

1355 giants with  $p_i \ge 0.9$ ;  $\langle \delta \rangle / \sigma \approx 0.33$ 



Profiles in circular annuli for the Sculptor dSph;  $R_h \approx 11.3$ arcmin. Data from Starkenburg et al. 2010 in green.

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### Fornax dSph

2409 giants with  $p_i \geq 0.9$ ;  $\langle \delta \rangle / \sigma \approx 0.22$ 



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### Conclusions

- Detailed modelling can tackle the dynamical structure of dSphs
- **Multiple-pop.** systems allow to probe the presence of cusps
- Phase-Space modelling of both photometric and kinematic data in Sculptor exclude an NFW halo
- This result is confirmed by different analyses, by using different data
- A Bayesian framework for measuring line profiles allows to double accuracy
- Sextans, Carina and Sculptor have LOSVD more peaked than Gaussian, suggesting some radial anisotropy

- Fornax is the only system with flat-topped LOSVDs
- Modelling that use all observables is required

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### **Mass Estimator**



$$R_{\rm h}/r_0 = 0.3$$

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$$M\left[(1.67 \pm 0.04)R_{\rm h}\right] = (5.85 \pm 0.2)\frac{R_{\rm h}\sigma_{\rm los}(R_{\rm h})^2}{G}$$

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### **Independent Populations - Core**



• MP: quasi isothermal,  $R_{\rm h,1} \approx r_0$ ,  $\chi^2 = 40$ 

■ MR: signs of truncation, radially anisotropic,  $R_{\rm h,2}$  undetermined,  $\chi^2 = 2.1$ 

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### **Independent Populations - NFW**



**MP**: mildly radial ,  $R_{\rm h,1} < 0.25 r_0$ ,  $\chi^2 = 49$ 

MR: strongly truncated, radially anisotropic, R<sub>h,2</sub> undetermined, χ<sup>2</sup> = 3.4

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$$\left(\frac{R_{\rm h,2}}{R_{\rm h,1}}\right)^{\delta} \le \left(\frac{\sigma_{\rm los,2}(R_{\rm h,2})}{\sigma_{\rm los,1}(R_{\rm h,1})}\right)^{2}$$
$$\Gamma = \frac{\ln\left[M(\lambda R_{\rm h,2})/M(\lambda R_{\rm h,1})\right]}{\ln\left(R_{\rm h,2}/R_{\rm h,1}\right)}$$

If  $k_R \geq k_\sigma$ , then

$$\Gamma = 1 + 2\frac{\ln k_{\sigma}}{\ln k_R} \ge \delta + 1$$

which is incompatible with  $\rho \sim r^{\delta-2}$ 

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	$\chi^2_{\Sigma}$	$\chi^2_{\Sigma} + \chi^2_{\sigma}$	$\chi_{\Sigma}^2 + \chi_{\sigma}^2 + \chi_{r_{\rm t}}^2$
NFW	39.3, 41.5, 45.7	48.2, 66.9, 68.3	49.0, 67.8, 69.7
cored	32.7, 36.3, 39.7	39.0, 54.3, 59.6	40.3, 55.7, 60.5
NFW	2.3, 5.6, 9.1	3.4, 11.1, 13.8	-
cored	1.1, 3.3, 4.9	2.1, 6.8, 9.9	-

**Table:** Results of the independent analysis of the metal-poor (upper) and metal-rich (lower) stellar component. The table gives the values of the  $\chi^2$ -quantities referring, in order, to the best fit models, to the 68% and to the 95% confidence regions.

# **Disentangling Populations**



Figure: Metallicity distribution in the Fornax dSph.

$$L = \prod_{i=1}^{N} \left[ \sum_{j} f_{j} \ p_{R,j}(R_{i}) \ p_{\Sigma,j}(\Sigma_{i}) \right]$$

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- Plummer density profiles
- Gaussian metallicity distributions

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### MR - Intermediate - MP





- MR:  $R_h \approx 10.5 \text{arcmin}$ ,  $\langle \Sigma'_{Mg} \rangle \approx 0.55 \text{\AA}, \ f \approx .1$
- Int:  $R_h \approx 15.3 \mathrm{arcmin}$ ,  $\langle \Sigma'_{Mg} \rangle \approx 0.45 \mathrm{\AA}$ ,  $f \approx .6$
- $\label{eq:main_states} \blacksquare \begin{array}{l} \mathsf{MP:} \ R_h \approx 23 \mathsf{arcmin}, \\ \langle \Sigma'_{Mg} \rangle \approx 0.26 \mathring{A}, \ f \approx .3 \end{array}$

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### **Kinematics**



Figure: Kinematics of the disentangled populations.

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### ...and Counter-Rotation



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dSphs



Figure: The effect of apparent rotation on circular annuli.

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**Figure:** 'Unstable' kinematics for the 2-pop division in the Fornax dSph.



Figure: Metallicity distribution in the Sculptor dSphs.

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