



# DYNAMICAL MODELLING OF BARRED GALAXIES

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ACADEMY OF ATHENS, 23 APRIL 2024

# Synopsis

- ▶ Motivation
  - ▶ Barred galaxies are ubiquitous ( $\sim 50\%$  of disk galaxies)
  - ▶ Complex morphology and kinematics
  - ▶ Interplay between bars and supermassive black holes
- ▶ The problem of deprojection
  - ▶ Non-uniqueness
  - ▶ Forward modelling of 3d shapes
- ▶ Modelling approaches for barred galaxies
  - ▶ Overview of methods
  - ▶ Schwarzschild's method in brief
- ▶ First results for barred galaxies and supermassive black holes

# The problem



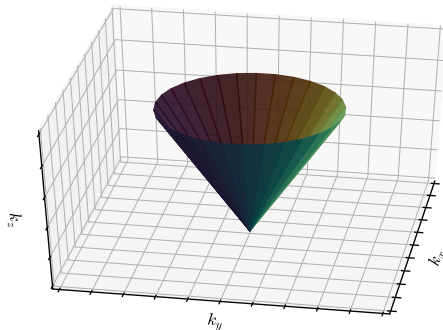
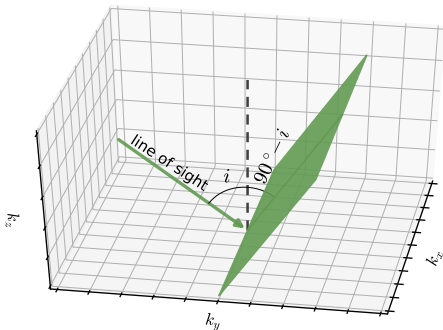
Pathway from 2d surface brightness profile to 3d density profile is non-unique

# The problem

Fourier Slice Theorem [Rybizki 1987]:

surface density  $\Sigma(X, Y) \implies$  its Fourier transform  $\hat{\Sigma}(k_X, k_Y)$   
corresponds to the Fourier transform of the 3d density  $\hat{\rho}(k_X, k_Y, k_Z = 0)$ ,  
i.e. provides **no** constraints on  $\hat{\rho}(\dots, k_Z \neq 0)$ .

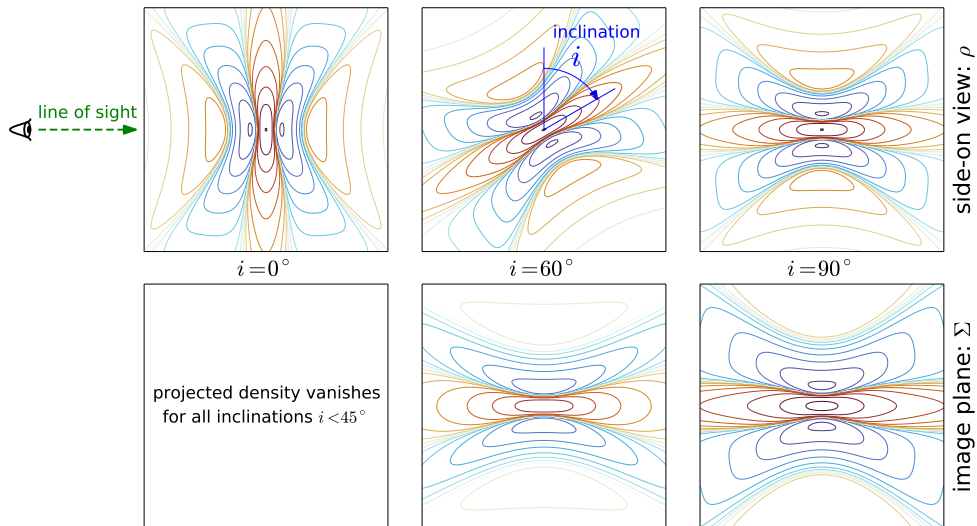
For an axisymmetric system at an inclination  $i$ , nothing is known of its  $\hat{\rho}$   
in the “cone of ignorance” with opening angle  $90^\circ - i$  around  $k_z$ .



# Illustration of non-uniqueness of axisymmetric deprojection

It turns out that there is a large family of axisymmetric “konus density” profiles that are completely invisible at any inclination  $i \leq i_{\min} < 90^\circ$

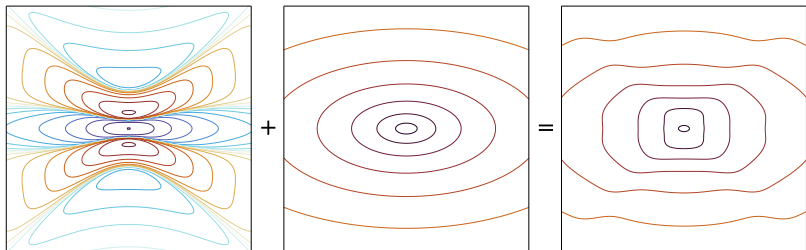
[Gerhard & Binney 1996; Kochanek & Rybizki 1996].



## Illustration of non-uniqueness of axisymmetric deprojection

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Adding it to an ordinary ellipsoidal density profile, one can make it boxy or disky, while still appearing perfectly elliptical in projection.

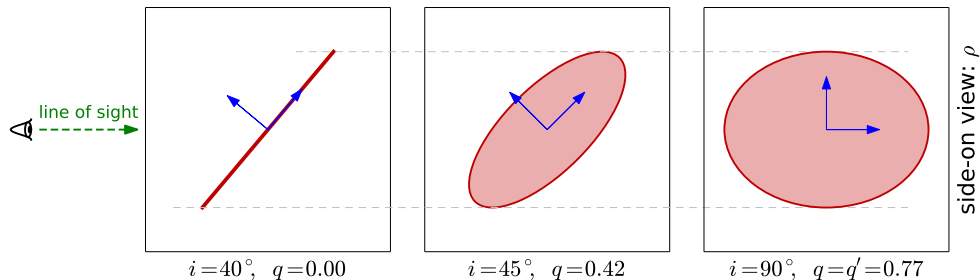


The degeneracy is **much** worse for triaxial systems.

Adding kinematic information should lift the degeneracy [Magorrian 1999].

# Approaches to deprojection

1. Ellipsoidal assumption:  $\rho(x, y, z) = \rho(m)$ ,  $m \equiv \sqrt{x^2 + (y/p)^2 + (z/q)^2}$ .



In the axisymmetric case, the projected axis ratio  $q' = \sqrt{q^2 \sin^2 i + \cos^2 i}$   
 $\implies$  deprojection is possible for inclination angles  $i > i_{\min} \equiv \arccos q'$ .

Generalization to a triaxial case: for a given projected shape and assumed orientation (viewing angles), the deprojection is either *unique* or impossible.

Widely used in practice, e.g. Multi-Gaussian Expansion [Cappellari 2002].

# Approaches to deprojection

1. Ellipsoidal assumption:  $\rho(x, y, z) = \rho(m)$ ,  $m \equiv \sqrt{x^2 + (y/p)^2 + (z/q)^2}$ .
2. Forward modelling of projected density:

parametric



- ▶ choose a suitable functional form for  $\rho(x, y, z; \mathbf{p})$
- ▶ assume some viewing angles  $\psi$  and parameters  $\mathbf{p}$
- ▶ compute the projected profile  $\Sigma(X, Y)$
- ▶ compare with the observed surface density and compute deviation  $\chi^2$
- ▶ repeat for different choices of  $\psi$  and  $\mathbf{p}$  to minimize  $\chi^2$

non-parametric

(or rather, multiparametric)

- ▶ choose a *very general / flexible* functional form with many free params (e.g., splines or a basis set expansion)
- ▶ assume some angles  $\psi$  and params  $\mathbf{p}$
- ▶ compute projected  $\Sigma$
- ▶ compare with observations; compute  $\chi^2$  and add some regularization penalty
- ▶ repeat for many choices of  $\psi$  and  $\mathbf{p}$   
[e.g., Magorrian 1999; de Nicola+ 2020]



## Photometric fitting



[reasonably] simple models  
multiple components  
sky subtraction  
foreground masking  
PSF convolution

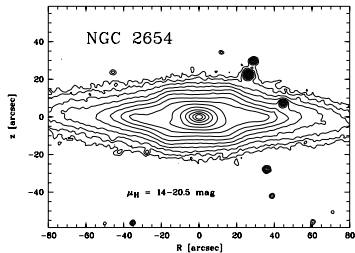
- ▶ MGEFit [Cappellari 2002]:  
“nonparametric” (multiple elliptical Gaussians)  $\Rightarrow$  ellipsoidal deprojection
- ▶ GalFit [Peng+ 2002, 2010]: many flexible 2d profiles,  
but deprojection is straightforward only for ellipsoidal models
- ▶ ImFit [Erwin 2015]: many 2d **and 3d profiles** (including user-defined),  
may project 3d model to 2d instead of deprojecting 2d to 3d

## 3D structure of bars

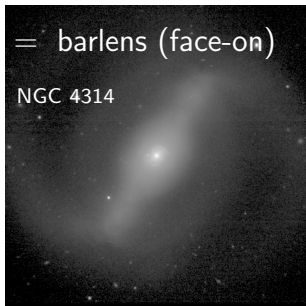
Bars often buckle vertically from the disk plane, but only in the inner part where the planar orbits are unstable;

shorter and vertically thick part is associated with boxy/peanut (B/P) bulges, and the longer and thinner component can be seen in face-on barlens galaxies [Athanasoula 2005, 2013].

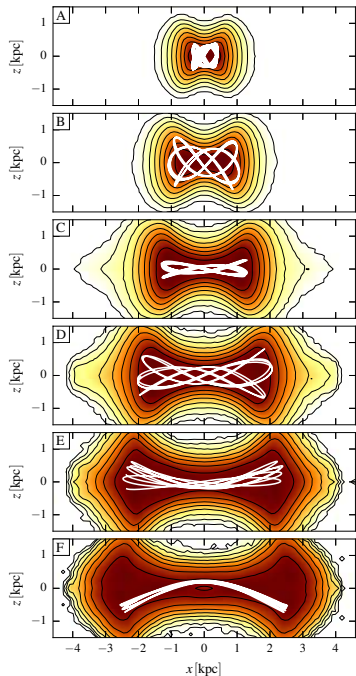
boxy/peanut bar (edge-on) = barlens (face-on)



[Lütticke+ 2000]

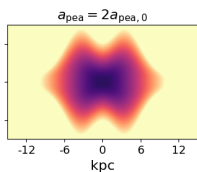
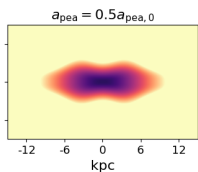
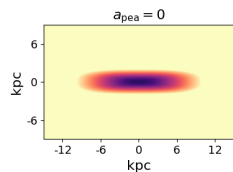
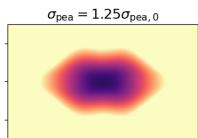
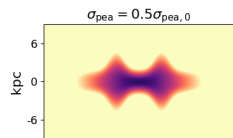
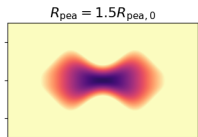
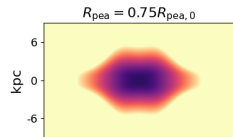
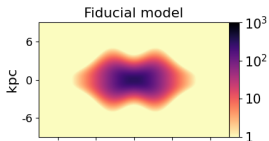


[Laurikainen+ 2011]



[Portail+ 2015]

# X-shaped bar model [Robin+ 2012; Fragkoudi+ 2015]



$$\rho \propto \text{sech}^2(-\mathcal{R})$$

$$\mathcal{R} = \left( \left[ \left( \frac{x}{X_{\text{bar}}} \right)^{c_{\perp}} + \left( \frac{y}{Y_{\text{bar}}} \right)^{c_{\perp}} \right]^{c_{\parallel}/c_{\perp}} + \left( \frac{z}{Z_{\text{bar}}} \right)^{c_{\parallel}} \right)^{1/c_{\parallel}}$$

$c_{\parallel}, c_{\perp}$ : boxiness coefficients

$$Z_{\text{bar}} = z_0 +$$

$$A_{\text{pea}} \exp \left( - \frac{(x - R_{\text{pea}})^2 + y^2}{2W_{\text{pea}}^2} \right) +$$

$$A_{\text{pea}} \exp \left( - \frac{(x + R_{\text{pea}})^2 + y^2}{2W_{\text{pea}}^2} \right)$$

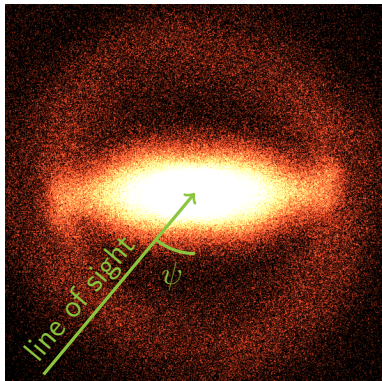
$A_{\text{pea}}$ : X/peanut amplitude

$R_{\text{pea}}$ : peanut location

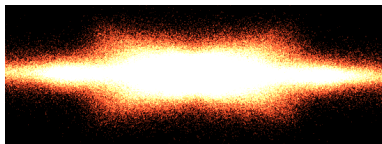
$W_{\text{pea}}$ : peanut width

# First application: edge-on projections

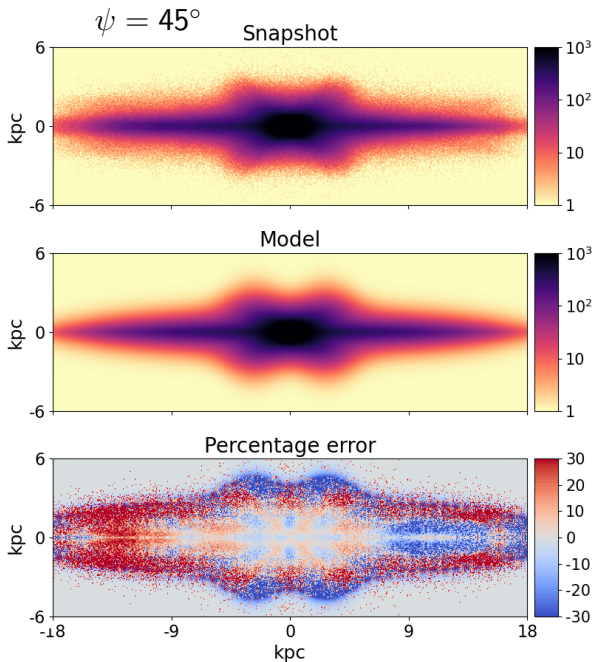
face-on view



edge-on view,  $\psi = 0$



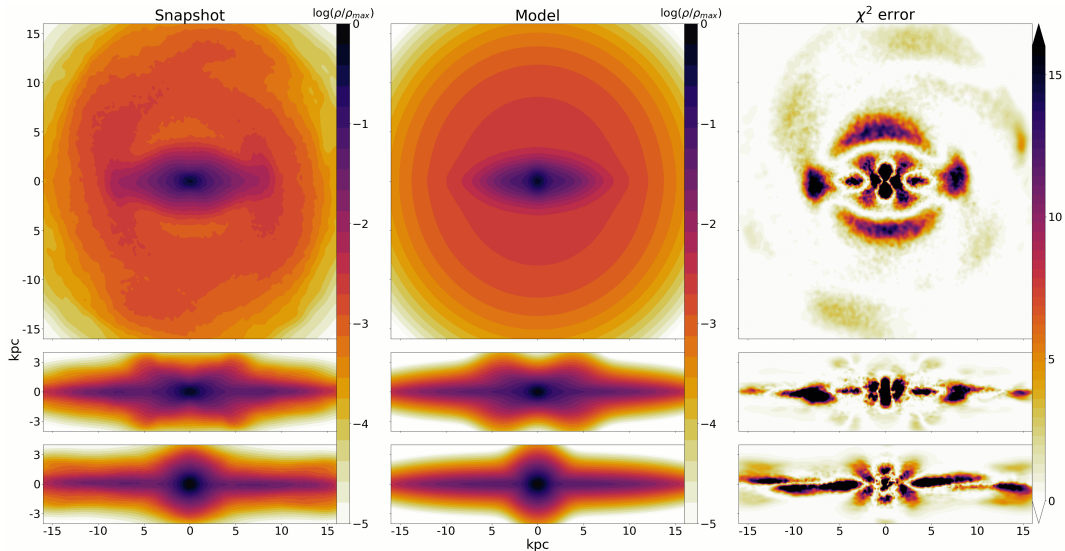
$\psi = 45^\circ$



# First application: edge-on projections

The fitted model qualitatively recovers the 3d density profile, though not without some defects

$$\psi = 45^\circ$$

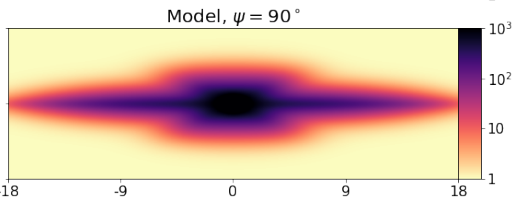
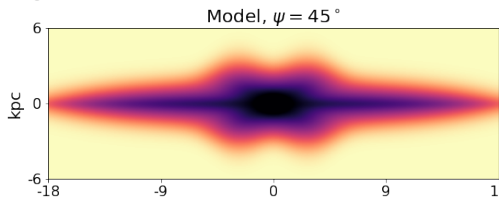
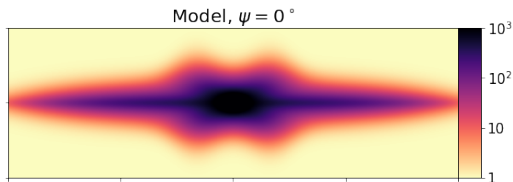
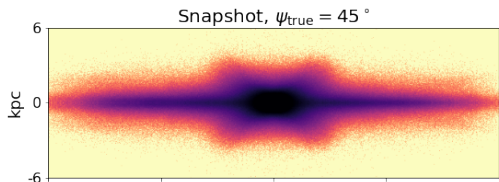
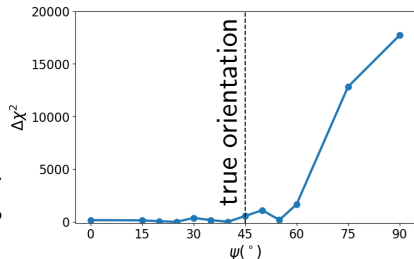


# Degeneracies in determining bar orientation

It is impossible to distinguish a rotated bar ( $0 < \psi < i_{\max} \lesssim 90^\circ$ ) from a shorter bar viewed at  $\psi = 0^\circ$  just from photometry.

(It might be easier at lower inclinations  $i < 90^\circ$ ).

Kinematics / dynamical modelling should help?



# Modelling approaches for barred galaxies

Challenges: triaxial geometry, chaotic regions in phase space

Goals:	$\Omega$	$\Phi$
Jeans modelling	–	–
Distribution functions, e.g., $f(\mathbf{J})$	?	?
Tremaine–Weinberg	$\pm$	–
Orbital response models	+	+
Guided $N$ -body simulations (made-to-measure)	+	+
Schwarzschild orbit-superposition modelling	+	+

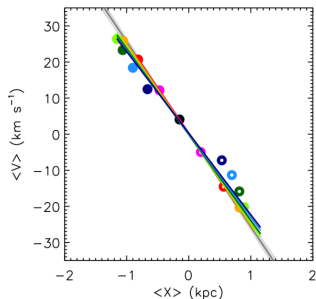
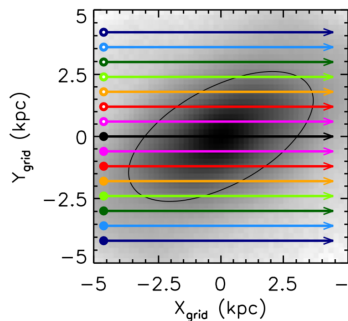
# Measurement of the pattern speed: Tremaine & Weinberg (1984)

$$\Omega_p \sin i = \frac{\langle V_{\text{los}} \rangle}{\langle X \rangle}$$

$$\langle V_{\text{los}} \rangle(Y) \equiv \frac{\int_{-\infty}^{\infty} V_{\text{los}}(X, Y) \Sigma(X, Y) dX}{\int_{-\infty}^{\infty} \Sigma(X, Y) dX}$$

$$\langle X \rangle(Y) \equiv \frac{\int_{-\infty}^{\infty} X \Sigma(X, Y) dX}{\int_{-\infty}^{\infty} \Sigma(X, Y) dX}$$

- + Simple to implement
- Works only in a limited range of inclinations  
 $15^\circ \lesssim i \lesssim 50\text{--}70^\circ$  [Zou+ 2019; Borodina+ 2023]
- Need to integrate to  $\pm\infty$   
(see Dehnen+ 2023 for improvement)



[Zou+ 2019]

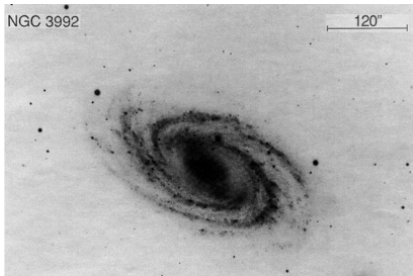


# Modelling approaches for barred galaxies: response models

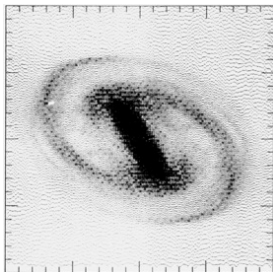
[Contopoulos & Grosbøl 1986, 1988; Patsis+ 1991; Kaufmann & Contopoulos 1996]

2d response models:

- assume parameters for potential, pattern speed, etc.
- construct the network of periodic orbits
- populate nearby orbits and compute their surface density
- compare morphological features with observations
- vary the parameters until a good match is found



observed galaxy



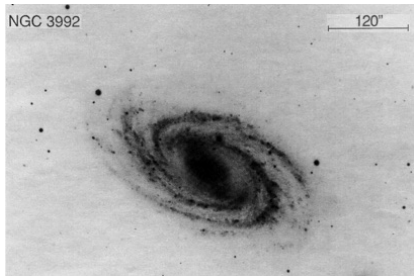
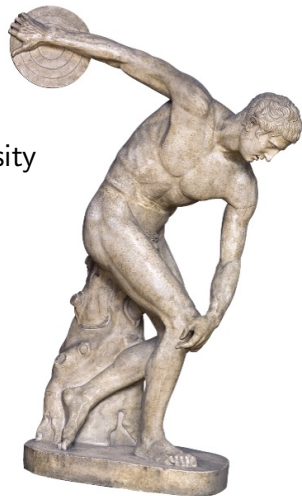
response model

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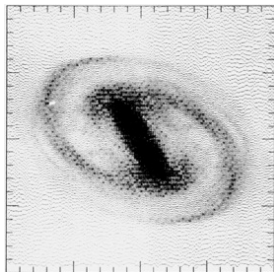
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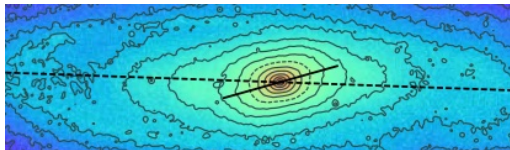
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# Modelling approaches for barred galaxies: made-to-measure

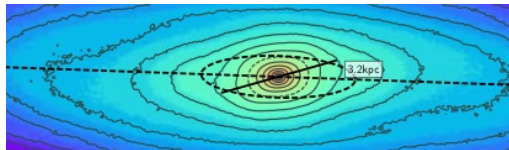
Introduced by Syer & Tremaine 1996,  
grown up and flourished in Ortwin Gerhard's group  
[Bissantz+ 2004, de Lorentzi+ 2007, Portail+ 2015; Blaña+ 2019];  
several other implementations exist [Dehnen 2009;  
Long & Mao 2012; Hunt & Kawata 2013; Malvido & Sellwood 2015].

Idea: evolve an  $N$ -body model while continuously  
adjusting particle weights to match the observables  
(density and kinematics).

Has been applied to the Milky Way bar [Portail+ 2017]  
and to a few external galaxies.



observed galaxy (M31)



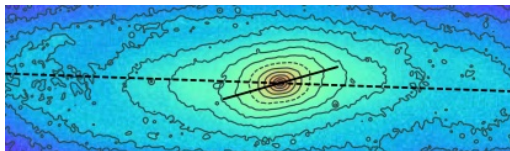
M2M model [Blaña+ 2019]

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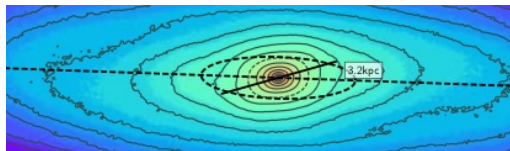
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# Modelling approaches for barred galaxies: orbit superposition

Introduced by Schwarzschild (1979) as a practical approach for constructing dynamically self-consistent triaxial models with prescribed  $\rho(\mathbf{x}) \Leftrightarrow \Phi(\mathbf{x})$ .

To invert the equation  $\rho(\mathbf{x}) = \iiint f(\mathcal{I}[\mathbf{x}, \mathbf{v} | \Phi]) d^3\mathbf{v}$ ,

integrals of motion

discretize both the density profile and the distribution function:

$\rho(\mathbf{x}) \implies$  cells of a spatial grid; mass of each cell is  $M_c = \iiint_{\mathbf{x} \in V_c} \rho(\mathbf{x}) d^3x$ ;

$f(\mathcal{I}) \implies$  collection of orbits with unknown weights [to be determined]:

$$f(\mathcal{I}) = \sum_{k=1}^{N_{\text{orb}}} w_k \delta(\mathcal{I} - \mathcal{I}_k)$$

each orbit is a delta-function in the space of integrals of motion

adjustable weight of each orbit

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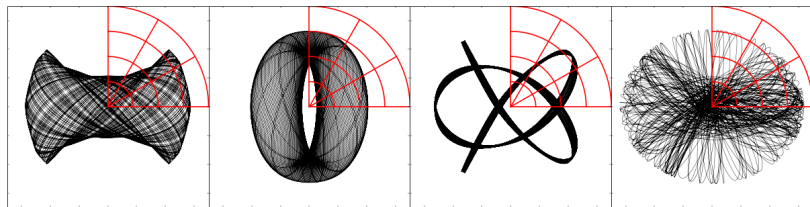
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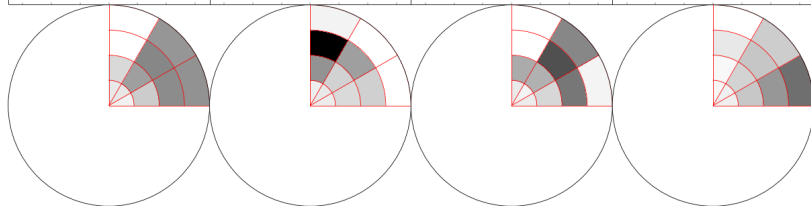
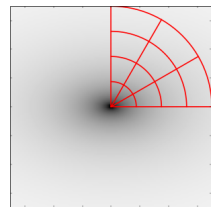


# Schwarzschild's orbit-superposition method: self-consistency

orbits in the model



target density



discretized orbit density

(fraction of time  $t_{kc}$  that  $k$ -th orbit spends in  $c$ -th cell)

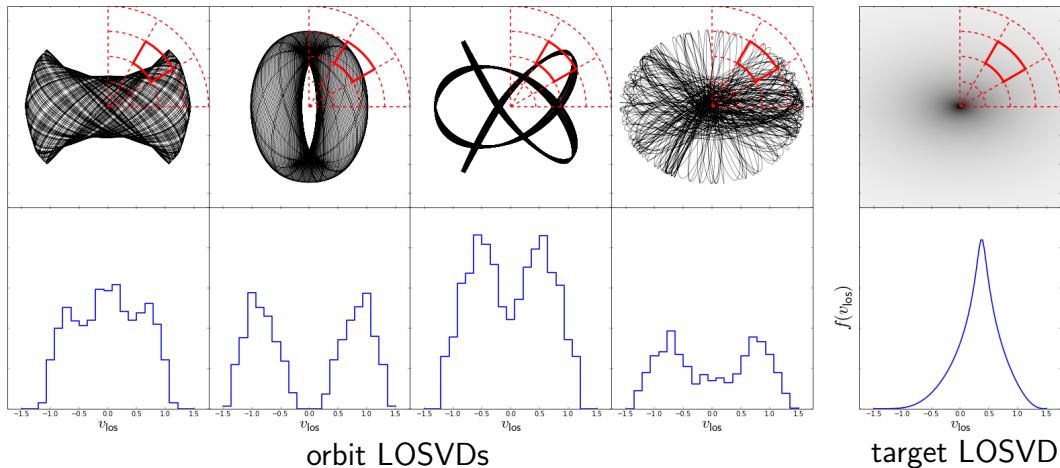
discretized density

(mass  $M_c$  in grid cells)

For each  $c$ -th cell we require  $\sum_k w_k t_{kc} = M_c$ , where  $w_k \geq 0$  is orbit weight

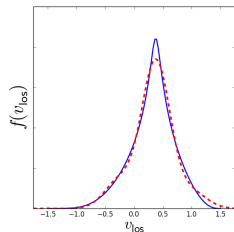
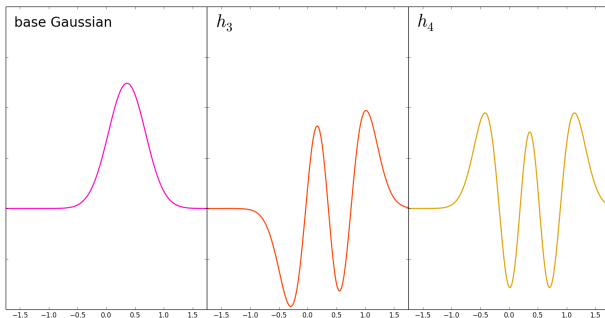
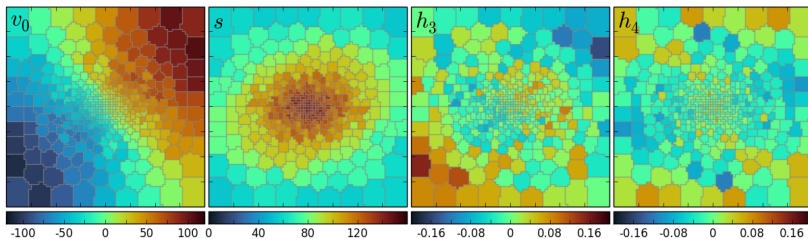
# Schwarzschild's orbit-superposition method: kinematics

orbits in the model





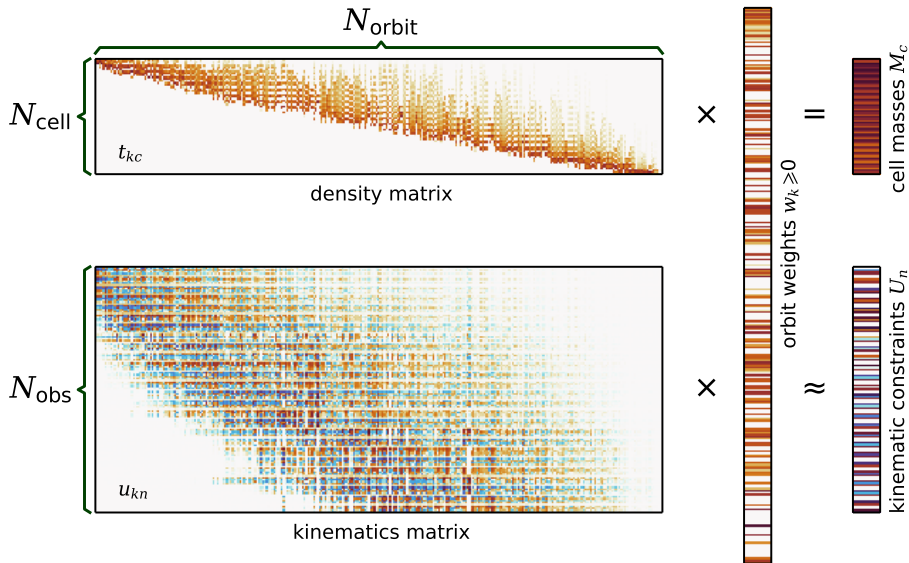
# Schwarzschild's orbit-superposition method: kinematics



Gauss-Hermite parametrization of LOSVDs [van der Marel & Franx 1993; Gerhard 1993]

# Schwarzschild's orbit-superposition method: fitting procedure

Solve the linear system with non-negativity constraints on the solution vector  $w_k \geq 0$   
(linear or non-linear optimization problem)



# Schwarzschild's orbit-superposition method: fitting procedure

- ▶ Assume some potential  $\Phi(\mathbf{x})$   
(e.g., from the deprojected luminosity profile plus parametric DM halo or SMBH)
- ▶ Construct the orbit library in this potential:  
for each  $k$ -th orbit, store its contribution to the discretized density profile  $t_{kc}$ ,  $c = 1..N_{\text{cell}}$  and to the kinematic observables  $u_{kn}$ ,  $n = 1..N_{\text{obs}}$
- ▶ Solve the constrained optimization problem to find orbit weights  $w_k$ :

$$\text{minimize } \chi^2 + \mathcal{S} \equiv \sum_{n=1}^{N_{\text{obs}}} \left( \frac{\sum_{k=1}^{N_{\text{orb}}} w_k u_{kn} - U_n}{\delta U_n} \right)^2 + \mathcal{S}(\{w_k\})$$

subject to  $w_k \geq 0$ ,  $k = 1..N_{\text{orb}}$ ,

$$\sum_{k=1}^{N_{\text{orb}}} w_k t_{kc} = M_c, \quad c = 1..N_{\text{cell}}$$

Annotations:  
-  $\delta U_n$ : their uncertainties  
-  $\mathcal{S}(\{w_k\})$ : regularization term  
-  $w_k$ : observational constraints

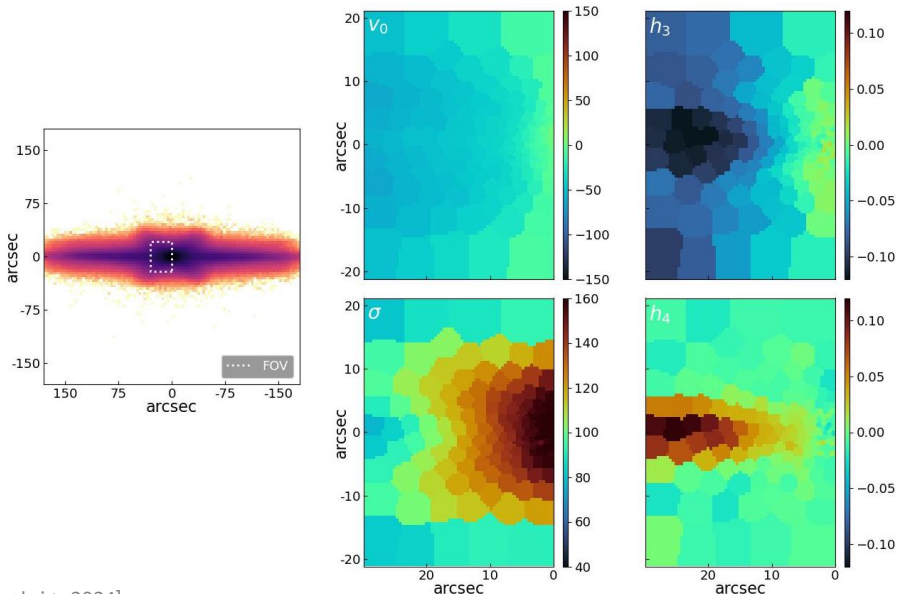
- ▶ Repeat for different choices of potential and find the one that has lowest  $\chi^2$

# Schwarzschild's orbit-superposition method: implementations

- ▶ theoretical studies in triaxial geometry: Schwarzschild 1979, 1993; Pfenniger 1984; Statler 1987; Merritt & Fridman 1996; Siopis & Kandrup 2000; Vasiliev 2013
- ▶ spherical codes: Richstone & Tremaine 1984; Rix+ 1997; Jalali & Tremaine 2010; Breddels & Helmi 2013; Kowalczyk+ 2017
- ▶ axisymmetric: “Leiden” [van der Marel, Cretton, Cappellari, ... – since 1998]
- ▶ axisymmetric: “Nukers” [Gebhardt, Richstone, Kormendy, ... – since 2000]
- ▶ axisymmetric: “MasMod” [Valluri, Merritt, Emsellem – since 2004]
- ▶ triaxial/Milky Way bar: Zhao, Wang, Mao 1996, 2012
- ▶ triaxial: van den Bosch, van de Ven ... – since 2008 ⇒ “Dynamite”
- ▶ triaxial: “SMART” [Neureiter+ 2021]
- ▶ triaxial: “Forstand” [Vasiliev & Athanassoula 2015; Vasiliev & Valluri 2020]

# Schwarzschild modelling of deprojected bars

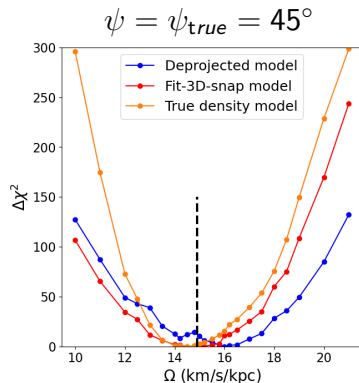
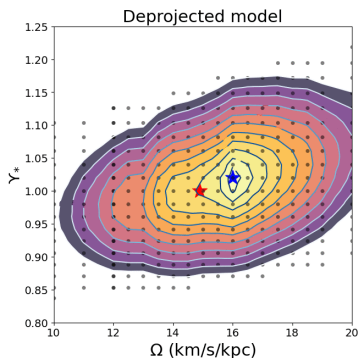
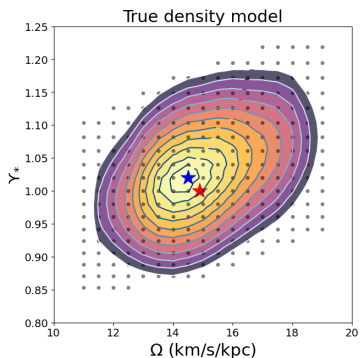
MUSE-like kinematic maps (1' FoV) of a Milky Way-like galaxy at  $D = 20$  Mpc



## Recovery of bar pattern speed

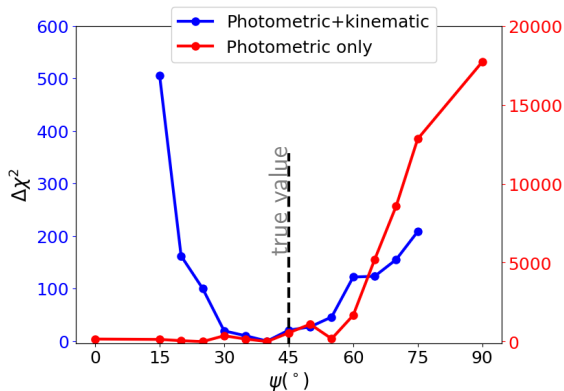
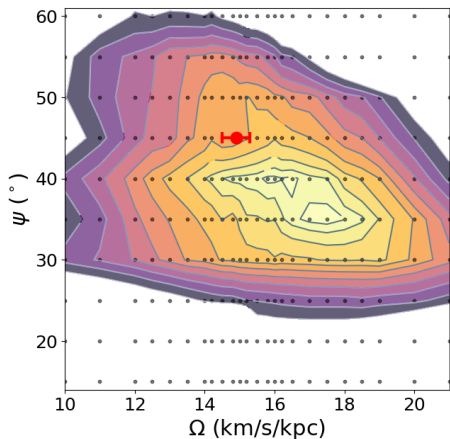
$\Omega$  is recovered almost perfectly if the true 3d density is used, or to within 10% if the deprojected density is used.

This is for the most challenging edge-on orientation, where the Tremaine–Weinberg method is not applicable!



# Recovery of bar orientation

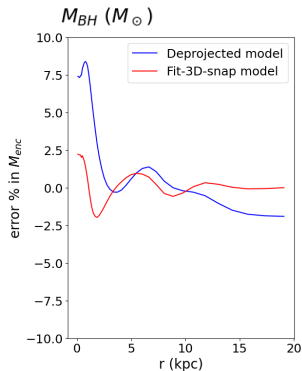
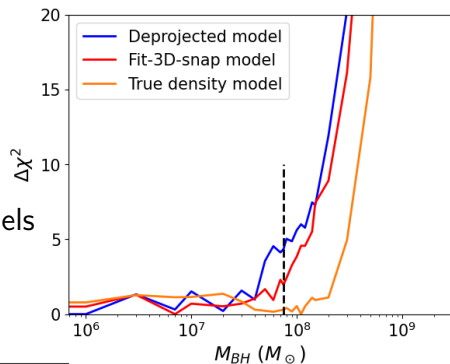
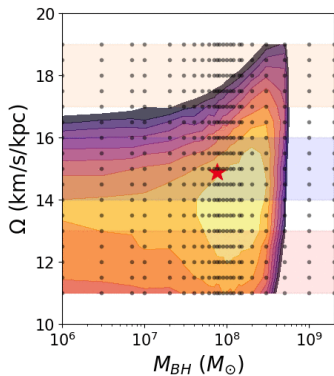
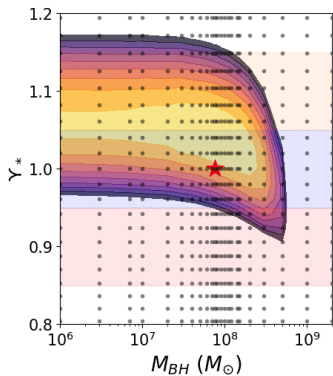
Bar orientation is also constrained much better than from pure photometry



# Recovery of black hole mass

Central supermassive black hole

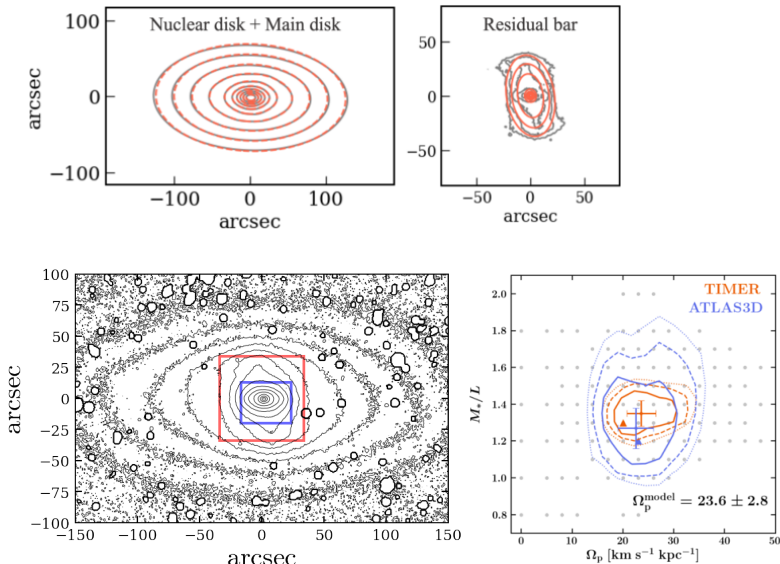
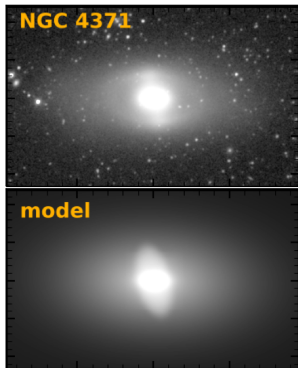
- ▶ does not destroy the bar [Wheeler+ 2023]
- ▶ has only an upper limit on  $M_{\bullet}$  in these models
- ▶ is very sensitive to the accuracy of reconstruction of enclosed stellar mass





# Another implementation of Schwarzschild's method for bars

by Tahmasebzadeh+ 2021, 2022, based on the triaxial code DYNAMITE;  
use MGE for deprojection, separately for the main disc and the bar



## Summary

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possible with IMFIT, but has some degeneracies.
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recovers pattern speed, orientation and stellar mass.

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