# WALLABY/DINGO kinematic pipeline : A new Bayesian MCMC tilted-ring fitter 

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- An overview of WALLABY kinematic pipeline
- A new Bayesian MCMC 2D tilted-ring fitter
- Performance test using sample galaxies from LVHIS
- Summary \& future works

Kinematic parameter extraction for WALLABY/DINGO: ASKAP WALLABY/DINGO $(\sim 5 ; 000)+$ WSRT WNSHS $(\sim 7,000)$


See posters + talks (Thursday) by Peter, Ed and Kristine for more details

NGC 5055 (Battaglia et al. 2005)





Galaxy Radius


Free: XPOS, YPOS, VSYS, PA, INCL, VROT Fixed: VEXP

Free: PA, INCL, VROT Fixed: XPOS, YPOS, VSYS, VEXP


Free: XPOS, YPOS, VROT
Fixed: VSYS, VEXP, PA, INCL
$\downarrow$

## Free: VROT

Fixed: XPOS, YPOS, VSYS, VEXP, PA, INCL
I

## Derive final rotation curve

$$
\begin{aligned}
& V_{\mathrm{obs}}(x, y)=V_{\mathrm{sys}}(x, y)+\sin \left(i\left\{V_{t}(x, y) \cos \theta+V_{r}(x, y) \sin \theta\right\}\right. \\
& \cos \theta=\frac{-(x-\mathrm{XPOS}) \sin (\mathrm{PA})+(y-\mathrm{YPOS}) \cos (\mathrm{PA})}{r} \\
& \sin \theta=\frac{-(x-\mathrm{XPOS}) \cos (\mathrm{PA})-(y-\mathrm{YPOS}) \sin (\mathrm{PA})}{r \cos (\mathrm{INCL})}
\end{aligned}
$$



- 6 free parameters
- VROT/INCL degenerated
- sensitive to initial estimates
- non-parametric models for PA/INCL
- affected by non-circular motions
$V_{\text {MODEL }}(x, y)=V_{\text {SYS }}(x, y)+V_{\text {ROT }}(r) \times \cos \theta \sin \mathrm{I}+V \exp (r) \times \sin \theta \sin \mathrm{I}$
$\left(\cos \theta=\frac{-(x-X P O S) \times \sin P A+(y-Y P O S) \times \cos P A}{r}\right.$
$\sin \theta=\frac{-(x-X P O S) \times \cos P A-(y-Y P O S) \times \sin P A}{r \cos \mathrm{I}}$

$r=\sqrt{[-(x-X P O S) \times \sin \theta+(y-Y P O S) \times \cos P A]^{2}+\left[\frac{(x-X P O S) \times \cos P A+(y-Y P O S) \times \sin P A}{\cos I}\right]^{2}}$

1. Kinematic position angle

- Several dynamical structures in galaxies (e.g., lopsideness, bar-like potential, sprial arms, non-circular motions etc.) change kinematic PA in radial.
- Usually, well modeled by a polynomial function with a moderate order (e.g., $\mathrm{m}=5$ )

$$
P A=\sum_{i=0}^{m} p_{i} r^{i}
$$



de Blok et al. (2008)


Radius

- Kinematic INCL change is often seen in galaxies but its sudden change in the inner region (probably due to non-circulr motions or low filling factor) is unphysical except for outer regions where warps may exist.
- A modified Sersic profile is used for INCL
- Constant or linear variation of INCL in the inner region (e.g., $n=0$ or 1)

$$
\mathrm{I}=\sum_{i=0}^{n} i_{i} r^{i}+\kappa \exp \left(\left[\frac{r}{\alpha}\right]^{\beta}\right)
$$



Radius
$V_{\text {MODEL }}(x, y)=V_{S Y S}(x, y)+V_{\text {ROT }}(r) \times \cos \theta \sin \mathrm{I}+V \exp (r) \times \sin \theta \sin \mathrm{I}$ $\cos \theta=\frac{-(x-X P O S) \times \sin P A+(y-Y P O S) \times \cos P A}{r}$

$$
P A=\sum_{i=0}^{m} p_{i} r
$$

$\sin \theta=\frac{-(x-X P O S) \times \cos P A-(y-Y P O S) \times \sin P A}{r \cos \mathrm{I}}$ $\mathrm{I}=\sum_{i=0}^{n} i_{i} r^{i}+\kappa \exp \left(\left[\frac{r}{\alpha}\right]^{\beta}\right)$
$r=\sqrt{[-(x-X P O S) \times \sin \underline{P A}+(y-Y P O S) \times \cos \underline{P A}]^{2}+\left[\frac{(x-X P O S) \times \cos \underline{P A}+(y-Y P O S) \times \sin \underline{P A}}{\cos L}\right]^{2}}$
$\rightarrow r=f\left(x, y, X P O S, Y P O S, p_{0}, p_{1}, \ldots, i_{0}, i_{1,}, \ldots, \kappa, \alpha, \beta\right)$
$\rightarrow$ Solve this non-linear equation and derive the radius, $r$ in the galaxy plane for given ( $x, y$ ), XPOS, YPOS, p0, p1, $\cdots, i 0, i 1, \ldots$, к, $\alpha, \beta$ (e.g., Newton-Rapson method etc.)
$V_{\text {MODEL }}(x, y)=V_{S Y S}(x, y)+V_{\text {ROT }}(r) \times \cos \theta \sin \mathrm{I}+V \exp (r) \times \sin \theta \sin \mathrm{I}$
$r)=f\left(x, y, X P O S, Y P O S, p_{0}, p_{1}, \ldots, i_{0}, i_{1}, \cdots, \kappa, \alpha, \beta\right)$
()$_{s o}(r)=\sqrt{4 \pi G \rho_{0} r_{c}^{2}\left[1-\frac{r_{r}}{r} \arctan \left(\frac{r}{r_{r}}\right)\right]}$
$\rightarrow V_{\text {MODEL }}=F($ XPOS , YPOS , VSYS , VEXP $, \overbrace{r_{c}, \rho_{0,},}^{\text {VROT }} \overbrace{p_{0, p} p_{1, \cdots}, \ldots}^{P A}, \overbrace{i_{0}, i_{1}, \ldots, \kappa, \alpha, \beta}^{I N C L})$

$$
\log \mathscr{L}=-\frac{N}{2} \log 2 \pi-\sum_{i=0}^{N A X I} \sum_{j}^{N A X 2} \log \sigma_{i j}-\frac{1}{2} \sum_{i=0}^{N A X I} \sum_{j=0}^{N A X 2}\left[\frac{V_{\text {ose }}(i, j)-V_{\text {noon }}(i, j)}{\sigma_{i}}\right]^{2}
$$

Log likelihood

- Bayesian parameter estimation
- Markov Chain Monte Carlo (MCMC) sampling (see Mackay 2003 and refs therein)
- less sensitive to initial values and gives good error estimation
- MCMC sampling (e.g., Metropolis-Hastings algorithm and its variants, Gibbs or Hamiltonian samplings)
- CPU intensive and sampling problems in multimodal posteriors
- Bayesian model selection
- CPU expensive for the calculation of the Bayesian evidence which is used to assign relative probabilities to different models
- thermodynamic integration method (e.g., O Ruanaidh \& Fitzgerald 1996)
- inefficient sampling in multimodal posteriors
- Improves the sampling efficiency and robustness based on the clustered nested sampling in Shaw et al. (2007)
- Calculates the evidence and explores parameter space even with multimodals and curving degeneracies in high dimensions
- Refer to Feroz \& Bridges (2008) for a complete discussion on the new sampling scheme, "the improved simultaneous ellipsoidal nested sampling method"
$\rightarrow$ a fully parallelized algorithm using MPI
- Successfuly implemented in astrophysics and cosmology (e.g., CosmoMC, SuperBayeS, SUSY, gravitational lensing, exo-planet detection, ASKAP FLASH absorption line finder (Allison et al. 2012))
- Standalone C program for 2D tilted-ring fits based on Bayesian MCMC - MultiNest v2.18, CFITSIO, standard ANSI C libraries
- fully automatic: estimation for initial values, convergence check and derivation of the final rotation curve for a given 2D velocity field
- several builtin rotation curve shape functions are provided (e.g., pseudo-isothermal, Burkert, polynomial rotation curves etc.)
- the larger number of sampling, the higher quality of fits but the more cpu time
$\rightarrow$ supports MPI which enables us to do parallel computing
seheon@darkmatter rotcur. develop.mpi]\$


## Program run.

```
+ WALLABY 2D TILTED-RING FITTER
WALLABY 2D TILTED-RING FITTER
+ by SE-HEON OH (ICRAR/UWA) + WALLABY KINEMATICS WORKING GROUP
Development history
    V.1.0 14/Sept/2013
+ Usage
    mpirun -loadbalance -np [0. N-cpus= 8] ./wallaby_2D_TRfitter
```

+ A. Input 2D velocity field to fit
[1. 2D VELOCITY FIELD= vf.fits)]
[2. NAX1 $=1024$ ] [3. NAX2 $=1024$ ]
+ [4. XLOWER= 200] [5. YLOWER= 200] [6. XUPPER= 800] [7. YUPPER= 800]
+ B. Binning option
[8. grid_X_ISOFIT= 5] [9. grid_Y_ISOFIT= 5] [10. BIN_X_TRFIT= 1] [11. BIN_X_TRFIT= 1]

|  | [12. XPOS= 512] | [13. xpos0= 500] | [14. $\times$ pos $1=520]$ |
| :---: | :---: | :---: | :---: |
| + | [15. YPOS= 512] | [16. ypos0= 500] | [17. ypos $1=520]$ |
|  | [18. VSYS $=0]$ | [19. vsyso $=-10]$ | [20. vsysi= 10] |
|  | [21. $\mathrm{PA}=45$ ] | [22. $\mathrm{paO}=0.1]$ | [23. pa1= 0.9] |
|  | [24. $\mathrm{INCL}=45$ ] | [25. incl0 $=0.1$ ] | [26. incl1 $=0.9]$ |

[27. VROT $=10][28, \operatorname{vrot} 0=0.0][29, \operatorname{vrot} 1=200]$


```
| EXAMPLE
```

mpirun -loadbalance -np 8 ./wallaby_2D_TRfitter \}
test.VF1024.fits
10241024
2002008008001
$5 \quad 5111$
512490530
512490530
$\begin{array}{llll}0 & -20 & 20\end{array}$
$\begin{array}{llll}45 & 0.1 & 0.9\end{array}$
$60 \quad 0.10 .9$
$10 \quad 0.0 \quad 120$
310031
205 \}
$\begin{array}{llllll}5 & 1 & 1 & & & \\ 100 & 0.8 & 0.1 & 0 & 0 & 0\end{array}$

## Performance test

: a model galaxy (NFW halo: c. $=8.1 . \mathrm{V} 200=140 \mathrm{~km} / \mathrm{s}+$ non-circular motions


Input velocity field
model velocity field

## Performance test

## : model galaxy (NFW halo: c=8.1 V200 $=140 \mathrm{~km} / \mathrm{s}+$ non-circular motions

model galaxy (NFW halo: $\mathrm{c}=8.1 \mathrm{~V} 200=140 \mathrm{~km} / \mathrm{s}$ )



Performance test : IC 5152.( $>5$ beams )








## Performance test <br> IC 5152.(model velocity fields)



Input ${ }^{\text {nt TRfit }}$


IC 5152

TR fit


Input ${ }_{\text {nmin }}$ ISOfit

weight map


| $0^{1.0}$ |  |
| :---: | :---: |
|  | 0.9 |
| -0.18 |  |
| $-0.7$ |  |
| 0.6 |  |
| -0.5 0.4 |  |
|  |  |
| 0.3 |  |
| 0.2 |  |
|  | 0.1 |
|  | 0.0 |

ISO fit


TRfitmisofit

weight map



## : ESO 154 g23 (model velocity fields)

HIPASS-J0256_54


Input


TR $\mathrm{R}_{\mathrm{fit}}$


ISO fit


TRfit

weight map


Performance test : HIPASS-J0731-68 (~ 4 beams)


Rotation Curves



Performance test

## : HIPASS-J0731-68 (model velocity fields):

HIPASS-J0731 68


ISO fit




Input ISOfit
TRfit ISOfit




Correcting for beam smearing effect :

Swaters et al. (2009)


Distance along major axis (arcmin)

Construct a model cube using the derived ring params $(\mathrm{VROT}+\Delta \mathrm{V})$

Calculate $x^{2}$ values of residual cubes between models and the observation

Find the optimal $\Delta V(R)$
minimising the $x^{2}$ value

## - Time consuming work

- manual adjustment of $\Delta V$ (e.g., galmod)
- regularised $x^{2}$ minimisation (e.g., TiRiFiC)
$\rightarrow$ Bayesian search for $\Delta \mathrm{V}$ ?


## Log likelihood

## Observed cube

Model cube

$$
\begin{array}{r}
\log \mathscr{L}=\frac{-M}{2} \log 2 \pi-\sum_{i=0}^{\text {NAXI }} \sum_{i=0}^{\text {NAX2 }} \sum_{k=0}^{\text {NAX3 }} \log \sigma_{i, j, k}-\frac{1}{2} \sum_{i=0}^{\text {NAXI }} \sum_{i=0}^{\text {NAX2 }} \sum_{k=0}^{\text {NAX3 }}\left[\frac{\left.F_{O B S}^{\downarrow}(i, j, k)-F_{\text {MODEL }}^{\downarrow}(i, j, k)^{\downarrow}\right]_{i, j, k}^{2}}{\left.F_{\text {MODEL }}=F(V R O T), X P O S, Y P O S, V S Y S, V E X P, P A, I N C L, V D I S P, D E N S, z 0\right)}\right. \\
\text { (from 2D tilted-ring model) (from mom0 \& mom2) }
\end{array}
$$

- A new 2D tilted-ring fitting program based on Bayesian MCMC developed
- Gives similar results as Se-Heon did for moderately or well resolved galaxies

- (will be) fully automatic for estimating initial priors, deriving rotaiton curves and visualising the results of > 10,000 resolved galaxies from ASKAP WALLABY/DINGO + WSRT WNSHS (+ also useful for MeerKAT MHONGOOSE)
- Under test using sample galaxies from LVHIS(26), LITTLE THINGS(27), and THINGS(25)
$\rightarrow$ will include sub-routines for deriving mass models of baryons + DM halo
- Beam smearing correction will be added...
- Also applicable to velocity fields from IFU or CO observations (e.g., SAMI, Wifes, MANGA etc.)
- Other types of galaxy kinematic 2D or 3D models can be plugged into the platform by defining their likelihood functions (evolution to 3D tilted-ring fits?)
- Will be tuned for open MPI using a cluster machine at ICRAR, and GUI (with PyQT) will be provided
- Statistical revisit of HI rotation curves of galaxies from all available literature data (e.g., rotation curve shape, cusp/core, etc.)

