

## ✓ Practicas 'Temas de Astronomia moderna - 2024A'

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A continuacion se desarrollara el codigo y las preguntas correspondiente a las practicas propuestas por el profesor y discutidas en clases con los demas companeros. Se obviara la instalacion de los paquetes y tutorial (En mi caso, como usuario de MacOS fue necesario actualizar las librerias de anaconda y descargar la version de desarrollador de GALA, <http://gala.adrian.pw/en/latest/>)

*Referencias:*

- <https://doi.org/10.5281/zenodo.4159870>
- <https://doi.org/10.21105%2Fjoss.00388>

## ✓ Practica I

Objetivo: Explorar el comportamiento de orbitas estelares en distintos potenciales galacticos

1. Que componentes principales debemos elegir para una galaxia tipo  $L^*$  tardio?

Hacemos un **potencial compuesto** usando la funcion de gala `gp.CCompositePotential`. El potencial de dicha galaxia se puede dividir en:

- Disk: Miyamoto-Nagai
- Bulge: Hernquist
- Dark Matter halo: Navarro Frenk and White (en primera instancia usaremos el esferico, no triaxial)

```

1 #We import the libraries and the different tools of
2 #python to make figures and units from astropy:
3 #python version 3.12.2
4 import gala.potential as gp
5 import astropy.units as u
6 import numpy as np
7 import gala.dynamics as gd
8 from gala.units import galactic
9 import matplotlib.pyplot as plt

1 total_potential = gp.CCompositePotential()
2 total_potential['disk'] = gp.MiyamotoNagaiPotential(m = 1E11 , a=3, b=0.15, un
3 total_potential['bulge'] = gp.HernquistPotential(m = 3E9 , c = 0.67, units=gala
4 total_potential['dm_halo'] = gp.NFWPotential.from_circular_velocity(v_c=200*u.kpc
5                                     r_s=10.*u.kpc,
6                                     units=galactic)

```

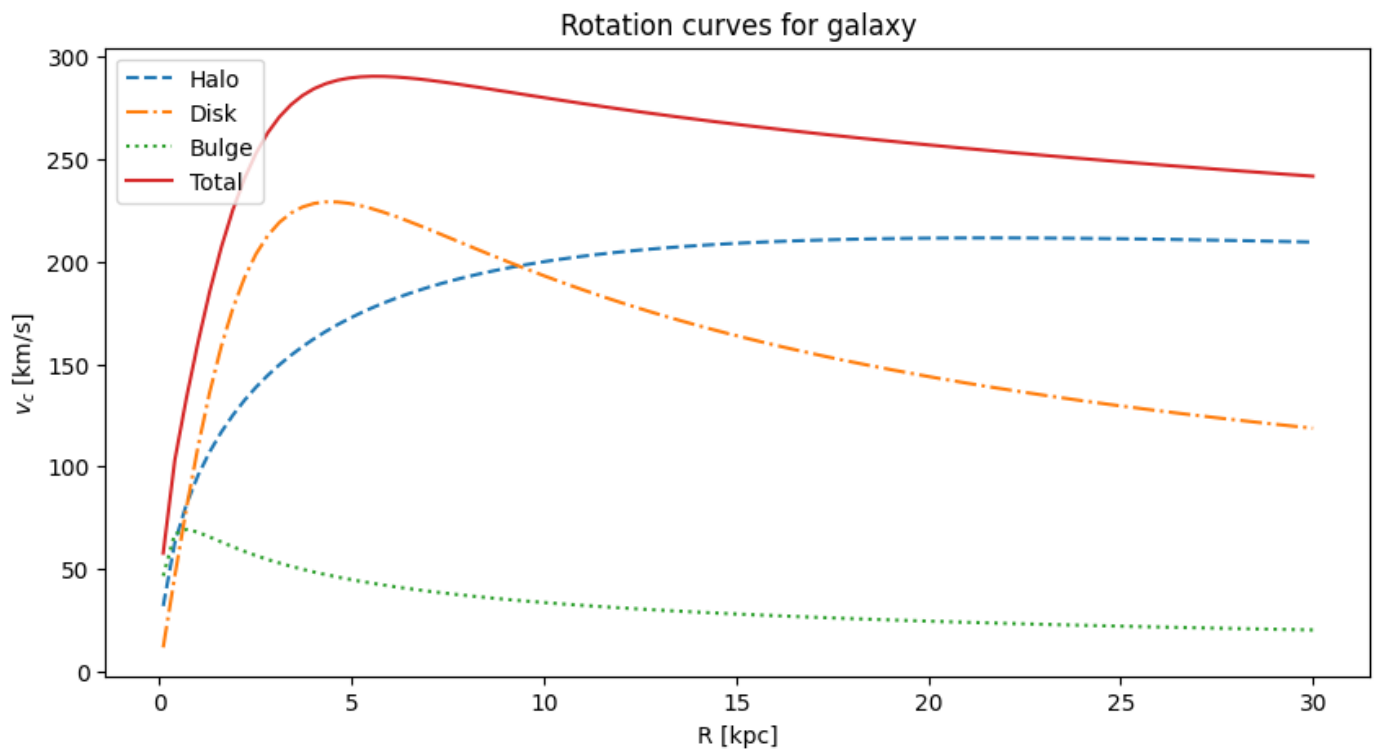
Análogo al tutorial que es para una galaxia tipo MW, (<https://gala.adrian.pw/en/latest/tutorials.html>), lo completamos para el potencial de nuestra galaxia L\* tardío (total\_potential).

2. Calcular las curvas de velocidades para:
  - a. Cada componente
  - b. Sistema total

```

1 R_grid = np.linspace(0.1, 30, 100) * u.kpc
2 xyz = np.zeros((3,) + R_grid.shape) * total_potential['dm_halo'].units["length"]
3 xyz[0] = R_grid
4
5 vcirc_halo = total_potential['dm_halo'].circular_velocity(xyz)
6 vcirc_disk = total_potential['disk'].circular_velocity(xyz)
7 vcirc_bulge = total_potential['bulge'].circular_velocity(xyz)
8 vcirc_gal = total_potential.circular_velocity(xyz)
9
10 fig = plt.figure(figsize=(10, 5))
11 plt.plot(R_grid, vcirc_halo, label='Halo',linestyle='--')
12 plt.plot(R_grid, vcirc_disk, label='Disk',linestyle='-.')
13 plt.plot(R_grid, vcirc_bulge, label='Bulge',linestyle=':')
14 plt.plot(R_grid, vcirc_gal, label='Total')
15 plt.xlabel('R [kpc]')
16 plt.ylabel(f'$v_c$ [km/s]')
17 plt.title('Rotation curves for galaxy')
18 plt.legend()
19 plt.show()

```



- b. **Orbita hiperbolica** ( $E > 0$  [km/\$s^2\$])

Calcularemos la órbita de las siguientes estrellas de prueba (a y b), usando solo el potencial del DM halo (`total_potential['dm_halo']`).

```
1 a_star_p = [6,0,0]
2 a_star_v = [20,80,0]
3
4 b_star_p = [-100,0,0]
5 b_star_v = [500,10,0]
6
7 a_particle = gd.PhaseSpacePosition(pos = a_star_p*u.kpc,
8                                     vel = a_star_v*u.km/u.s)
9 b_particle = gd.PhaseSpacePosition(pos = b_star_p*u.kpc,
10                                    vel= b_star_v*u.km/u.s)
```

```

1 # Usando el hamiltoniano calculamos la energia de la orbita,
2 # dadas las condiciones iniciales que definimos en la celda anterior
3 # esto para corroborar que tipo de orbita es.
4 a_orbit_DM = gp.Hamiltonian(total_potential['dm_halo']).integrate_orbit(a_part
5                                     dt=0.5:
6                                     t1=0,
7 b_orbit_DM = gp.Hamiltonian(total_potential['dm_halo']).integrate_orbit(b_part
8                                     dt =0.:
9                                     t1=0,
10

```

```

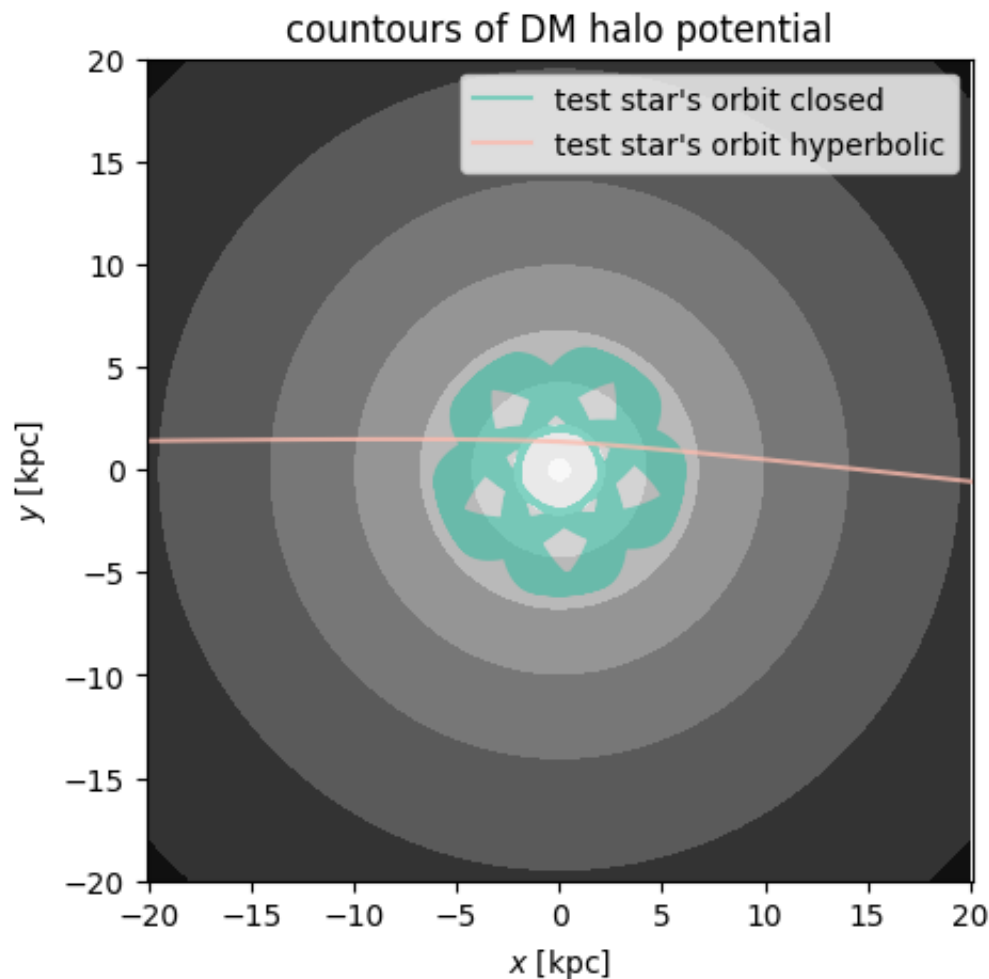
1 grid = np.linspace(-20,20,128)
2 fig, ax = plt.subplots(1,1, figsize=(5,5))
3 fig = total_potential['dm_halo'].plot_contours(grid=(grid, grid, 0), cmap='Gre
4 fig = a_orbit_DM.plot(['x','y'], color='#1ABC9C',
5                       alpha=0.5, axes=[ax],
6                       auto_aspect=True,
7                       label="test star's orbit closed")
8 fig2 = b_orbit_DM.plot(['x','y'], color='#FCBBAE',
9                        alpha=0.8, axes=[ax],
10                       auto_aspect=True,
11                       label="test star's orbit hyperbolic")
12 ax.set_xlim(-20,20)
13 ax.set_ylim(-20,20)
14
15 ax.legend()
16 ax.set_title('countours of DM halo potential')
17

```

```

➡ Text(0.5, 1.0, 'countours of DM halo potential')

```



#### 4. Integrar orbitas anteriores e ir agregando componentes.

- DM halo (`total_potential['dm_halo']`)
- DM halo + disco (`total_potential['dm_halo'] + total_potential['disk']`)
- DM halo + disco + bulge (`total_potential['dm_halo'] + total_potential['disk'] + total_potential['bulge']`)

```
1 a_orbit = gp.Hamiltonian(total_potential).integrate_orbit(a_particle, dt=0.5*u
2                                     t1=0, t2=1)
3 b_orbit = gp.Hamiltonian(total_potential).integrate_orbit(b_particle, dt =0.5*u
4                                     t1=0, t2=1)
5
```

```

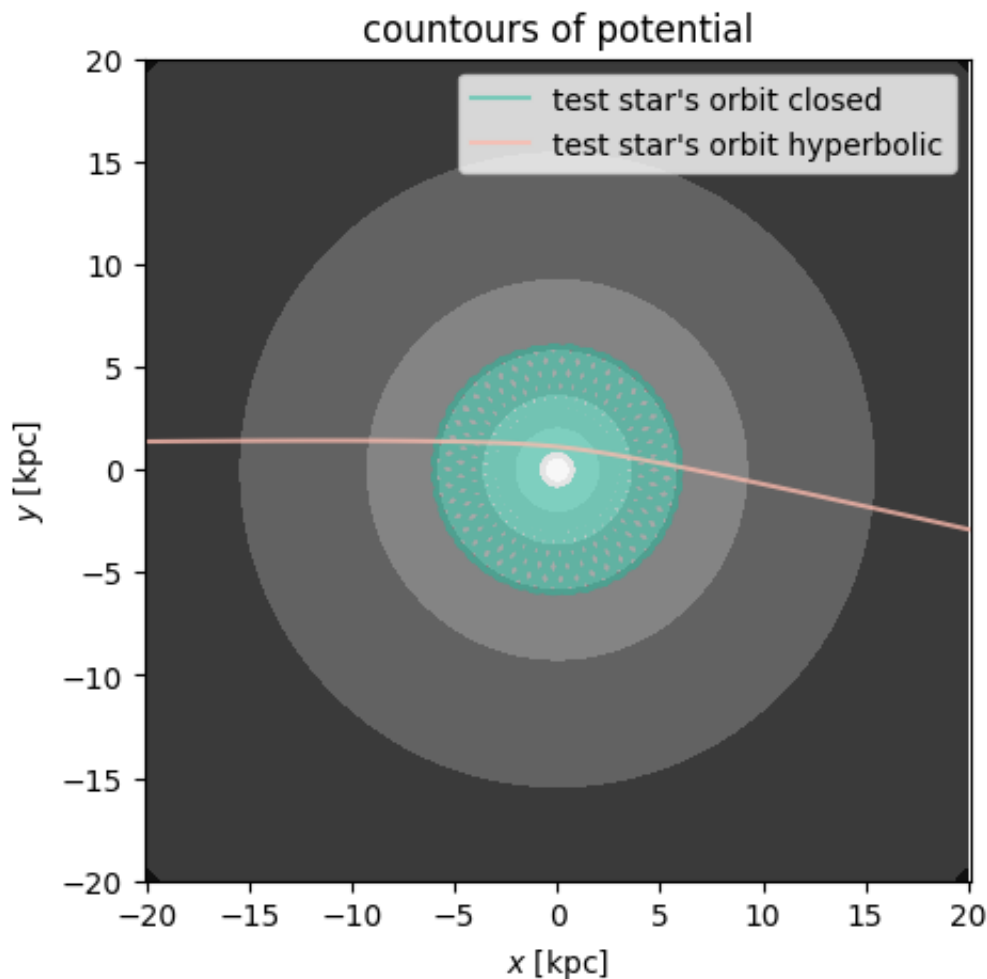
1 grid = np.linspace(-20,20,128)
2 fig, ax = plt.subplots(1,1, figsize=(5,5))
3 fig = total_potential.plot_contours(grid=(grid, grid, 0),
4                                     cmap='Greys', ax=ax)
5 fig = a_orbit.plot(['x','y'], color='#1ABC9C',
6                   alpha=0.5, axes=[ax],
7                   auto_aspect=True,
8                   label="test star's orbit closed")
9 fig2 = b_orbit.plot(['x','y'], color='#FCBBAE',
10                    alpha=0.8, axes=[ax],
11                    auto_aspect=True,
12                    label="test star's orbit hyperbolic")
13 ax.set_xlim(-20,20)
14 ax.set_ylim(-20,20)
15
16 ax.legend()
17 ax.set_title('countours of potential')

```

```

➡ Text(0.5, 1.0, 'countours of potential')

```



Podemos notar que el espacio se hace denso mucho mas rapido que cuando solo trabajamos sobre el halo esferico de materia oscura. Por otro lado la orbita hiperbolica, sigue siendo hiperbolica.

Tambien graficaremos la energia de la particula en el tiempo, y comprobaremos si efectivamente son cerrada e hiperbolica respectivamente.

```
1 b_totalenergy = b_orbit.energy().to(u.kpc**2/u.Myr**2)
2
3 a_totalenergy = a_orbit.energy().to(u.kpc**2/u.Myr**2)
```

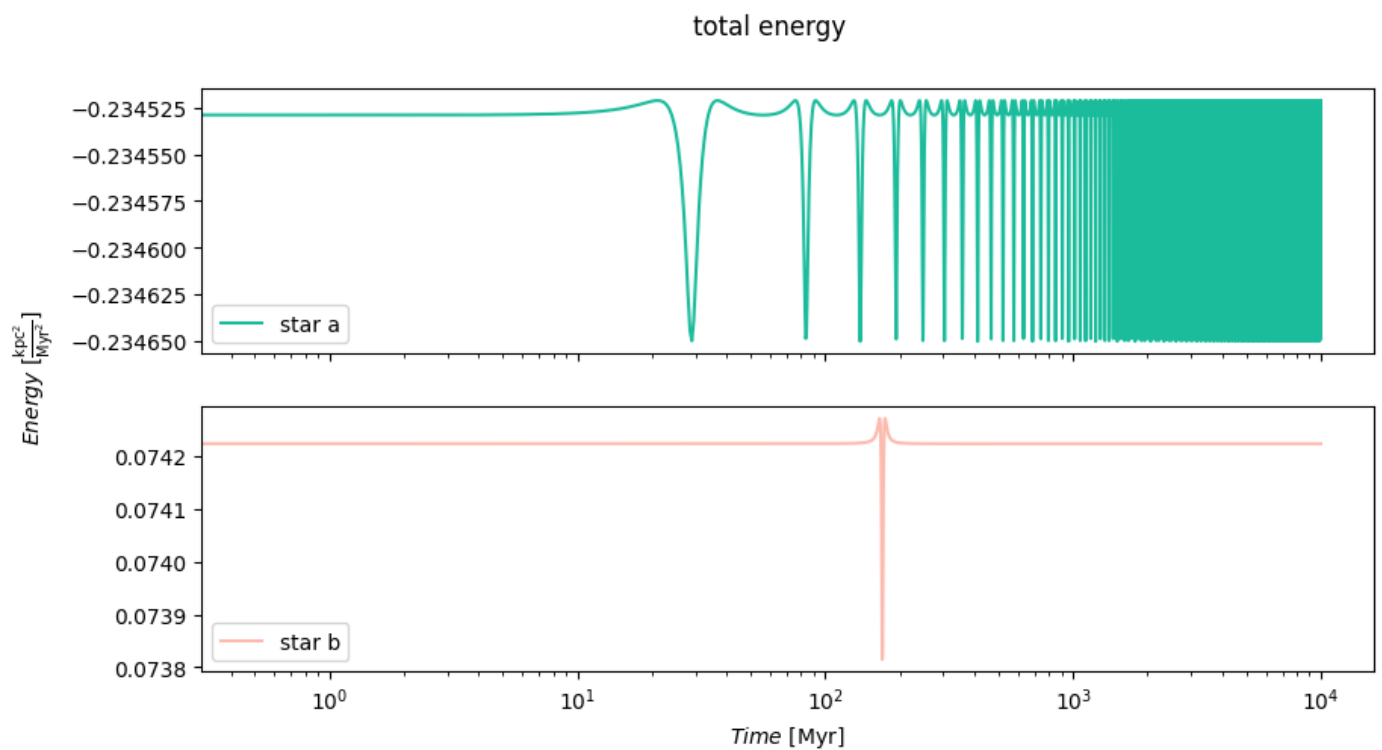


```

1 fig, ax = plt.subplots(2,1, figsize=(10,5), sharex=True)
2
3 ax[0].plot(a_orbit.t, a_totalenergy,
4           color='#1ABC9C', label='star a')
5 ax[1].plot(b_orbit.t, b_totalenergy,
6           color='#FCBBAE', label='star b')
7
8 ax[1].set_xlabel("$Time$ [{}]").format(a_orbit.t.unit.to_string(format="latex"))
9 ax[0].set_xscale('log')
10 ax[1].set_xscale('log')
11
12 fig.text(0.03, 0.5, "$Energy$ [{}]").format(a_totalenergy.unit.to_string(format=
13       horizontalalignment='right',
14       verticalalignment='center',
15       rotation='vertical')
16 plt.suptitle('total energy')
17 ax[0].legend(loc='best')
18 ax[1].legend(loc='best')

```

 <matplotlib.legend.Legend at 0x1485df4d0>



Como podemos notar, la energía total es negativa en el caso de la estrella a, es decir, esta cerrada, mientras que para la estrella b (cabe destacar que su posición es mucho más lejana a la galaxia que la posición inicial de a), la energía total se mantiene siempre positiva. Podemos observar también que la evolución de la energía en el tiempo nos indica que tan lejos o cerca se encuentra el cuerpo del área de mayor potencial de la galaxia. Si consideramos que la energía total es la suma de la energía cinética y potencial, notamos que estas se mantienen en equilibrio para la órbita cerrada, subiendo y bajando de manera complementaria, mientras que para la órbita hiperbólica, la energía potencial baja cuando se acerca a la galaxia, pero no es suficiente para que la energía total sea negativa, es decir la energía cinética no disminuye y por ende la energía total es mayor a 0.

```

1 a_eccentricity = a_orbit.eccentricity()
2 b_eccentricity = b_orbit.eccentricity()
3
4 a_apocenter = a_orbit.apocenter()
5 b_apocenter = b_orbit.apocenter()
6
7 a_pericenter = a_orbit.pericenter()
8 b_pericenter = b_orbit.pericenter()

➡ /opt/anaconda3/envs/astro/lib/python3.12/site-packages/numpy/core/fromnumeric.
    return _methods._mean(a, axis=axis, dtype=dtype,
/opt/anaconda3/envs/astro/lib/python3.12/site-packages/numpy/core/_methods.py:
    ret = ret.dtype.type(ret / rcount)

```

Gala utiliza la ecuación:  $e = \frac{r_{apo}-r_{per}}{r_{apo}+r_{per}}$  Y con ello podemos obtener los valores correspondientes a la eccentricidad. Mientras que el apocentro y pericentro promedio provienen de la integración numérica de la órbita

```

1
2 print(a_eccentricity, a_apocenter, a_pericenter)
3 print(b_eccentricity,b_apocenter, b_pericenter )

➡ 0.7095922426112573 6.01542239391078 kpc 1.021837420421007 kpc
   nan nan kpc 1.1152944507957727 kpc

```

star	eccentricidad	apocentro	pericentro
a star	0.7095922426112573	6.015 kpc	1.022 kpc
b star	nan	nan	1.12 kpc

## 6. Comparar y discutir:

Podemos notar que la eccentricidad de la orbita de a es menor a 1, que por definicion significa que es una orbita eliptica. Luego tenemos que su apocentro medio es 6 kpc y su pericentro ~~~1 kpc~~. Considerando que la estrella b parte por fuera de la galaxia y no queda ligada, notamos que ~~no hay un valor para la eccentricidad y el apocentro, que deben ser  $e_b \gg 1$  y apocentro  $\rightarrow \infty$ , pero que su pericentro es similar al de a (1 kpc)~~. Esto podria deberse a que ambos se acercan a la galaxia en el mismo plano (es decir con posicion inicia en z iguales)

---

## ✓ Practica II

Objetivo: Seguimiento a la practica anterior para explorar mas propiedades aprendidas en clases.

### 1. Considerar orbitas cerradas solamente. De forma secuencial graficar:

- Energia vs tiempo
- momento angular total vs tiempo
- componentes de momento angular vs tiempo
- eccentricidad vs tiempo

```
1 def calculate_eccentricity(apocenter, pericenter):
2     eccentricity = (apocenter - pericenter) / (apocenter + pericenter)
3     return eccentricity

1 # Respecto a la practica anterior utilizaremos la estrella a
2 # que es cerrada, y calcularemos su momento angular
3
4 #para el dm halo
5 aDM_totalenergy = a_orbit_DM.energy()
6
7 aDM_angularmoment = a_orbit_DM.angular_momentum
8 aDM_Lx = aDM_angularmoment()[0]
9 aDM_Ly = aDM_angularmoment()[1]
10 aDM_Lz = aDM_angularmoment()[2]
11 aDM_totalangularmoment = np.sqrt(aDM_Lx**2 +
12                                     aDM_Ly**2 +
13                                     aDM_Lz**2)
14 aDM_pericenter = (a_orbit_DM.pericenter(return_times=True, func=None)[0]).to_val
15 aDM_apocenter = (a_orbit_DM.apocenter(return_times=True, func=None)[0]).to_val
```

```

16 aDM_apocenter = np.delete(aDM_apocenter, -1)
17 timeDMecc = a_orbit_DM.pericenter(return_times=True, func=None)[1]
18 aDM_ecc = calculate_eccentricity(aDM_apocenter, aDM_pericenter)
19
20 #para el disco
21 a_orbit_disk = gp.Hamiltonian(total_potential['disk']).integrate_orbit(a_parti
22                                     dt=0.5*u.Myr
23                                     t1=0, t2=10
24 a_disk_totalenergy = a_orbit_disk.energy()
25
26 a_disk_angularmoment = a_orbit_disk.angular_momentum
27 a_disk_Lx = a_disk_angularmoment()[0]
28 a_disk_Ly = a_disk_angularmoment()[1]
29 a_disk_Lz = a_disk_angularmoment()[2]
30 a_disk_totalangularmoment = np.sqrt(a_disk_Lx**2 +
31                                     a_disk_Ly**2 +
32                                     a_disk_Lz**2)
33 a_disk_pericenter = (a_orbit_disk.pericenter(return_times=True, func=None)[0])
34 a_disk_apocenter = (a_orbit_disk.apocenter(return_times=True, func=None)[0]).t
35 a_disk_apocenter = np.delete(a_disk_apocenter, -1)
36 timediskecc = a_orbit_disk.pericenter(return_times=True, func=None)[1]
37 a_disk_ecc = calculate_eccentricity(a_disk_apocenter, a_disk_pericenter)
38
39
40
41 #para el bulbo
42 a_orbit_bulge = gp.Hamiltonian(total_potential['bulge']).integrate_orbit(a_par
43                                     dt=0.5*u.Myr
44                                     t1=0, t2=10
45 a_bulge_totalenergy = a_orbit_bulge.energy()
46
47 a_bulge_angularmoment = a_orbit_bulge.angular_momentum
48 a_bulge_Lx = a_bulge_angularmoment()[0]
49 a_bulge_Ly = a_bulge_angularmoment()[1]
50 a_bulge_Lz = a_bulge_angularmoment()[2]
51 a_bulge_totalangularmoment = np.sqrt(a_bulge_Lx**2 +
52                                     a_bulge_Ly**2 +
53                                     a_bulge_Lz**2)
54
55 a_bulge_pericenter = (a_orbit_bulge.pericenter(return_times=True, func=None)[0]
56 a_bulge_apocenter = (a_orbit_bulge.apocenter(return_times=True, func=None)[0])
57 #a_bulge_apocenter = np.delete(a_bulge_apocenter, -1)
58 timebulgeecc = a_orbit_bulge.pericenter(return_times=True, func=None)[1]
59 a_bulge_ecc = calculate_eccentricity(a_bulge_apocenter, a_bulge_pericenter)
60
61
62 #para todos los potenciales
63 a_angularmoment = a_orbit.angular_momentum

```

```

64 a_Lx = a_angularmoment()[0]
65 a_Ly = a_angularmoment()[1]
66 a_Lz = a_angularmoment()[2]
67 a_totalangularmoment = np.sqrt(a_Lx**2 +
68                                 a_Ly**2 +
69                                 a_Lz**2)
70
71 timeecc = a_orbit.pericenter(return_times=True, func=None)[1]
72 a_apocenter = (a_orbit.apocenter(return_times=True, func=None)[0]).to_value()
73 a_pericenter = (a_orbit.pericenter(return_times=True, func=None)[0]).to_value()
74 a_apocenter = np.delete(a_apocenter, -1)
75 a_ecc = calculate_eccentricity(a_apocenter, a_pericenter)
76
77

1 from mpl_toolkits.axes_grid1.inset_locator import zoomed_inset_axes
2 from mpl_toolkits.axes_grid1.inset_locator import mark_inset

1 t = a_orbit.t
2
3
4 fig, ax = plt.subplots(4,4, figsize=(20,20), constrained_layout = True, sharex:
5 ax[0,0].set_xscale('log')
6 ax[0,0].plot(t, aDM_totalenergy, color='pink', label='total energy in DM halo'
7 ax[0,0].legend(loc='best')
8 ax[0,0].set_ylim(-0.3,0.1)
9
10 axins = ax[0,0].inset_axes([0.5, 0.5, 0.47, 0.47])
11 # axins = zoomed_inset_axes(ax[0,0], 3, loc=4) # zoom = 6
12 # sub region of the original image
13
14 axins.set_xscale('log')
15 axins.plot(t, aDM_totalenergy, color='pink', label='total energy in DM halo')
16
17 # draw a bbox of the region of the inset axes in the parent axes and
18 # connecting lines between the bbox and the inset axes area
19 ax[0,0].indicate_inset_zoom(axins, edgecolor="black")
20 mark_inset(ax[0,0], axins, loc1=2, loc2=4, fc='none',ec='0.1')
21
22
23
24
25 ax[0,1].plot(t, aDM_totalangularmoment, color='purple', label='total angular m
26 ax[0,1].legend(loc='best')
27 ax[0,1].set_yscale('log')
28

```

```

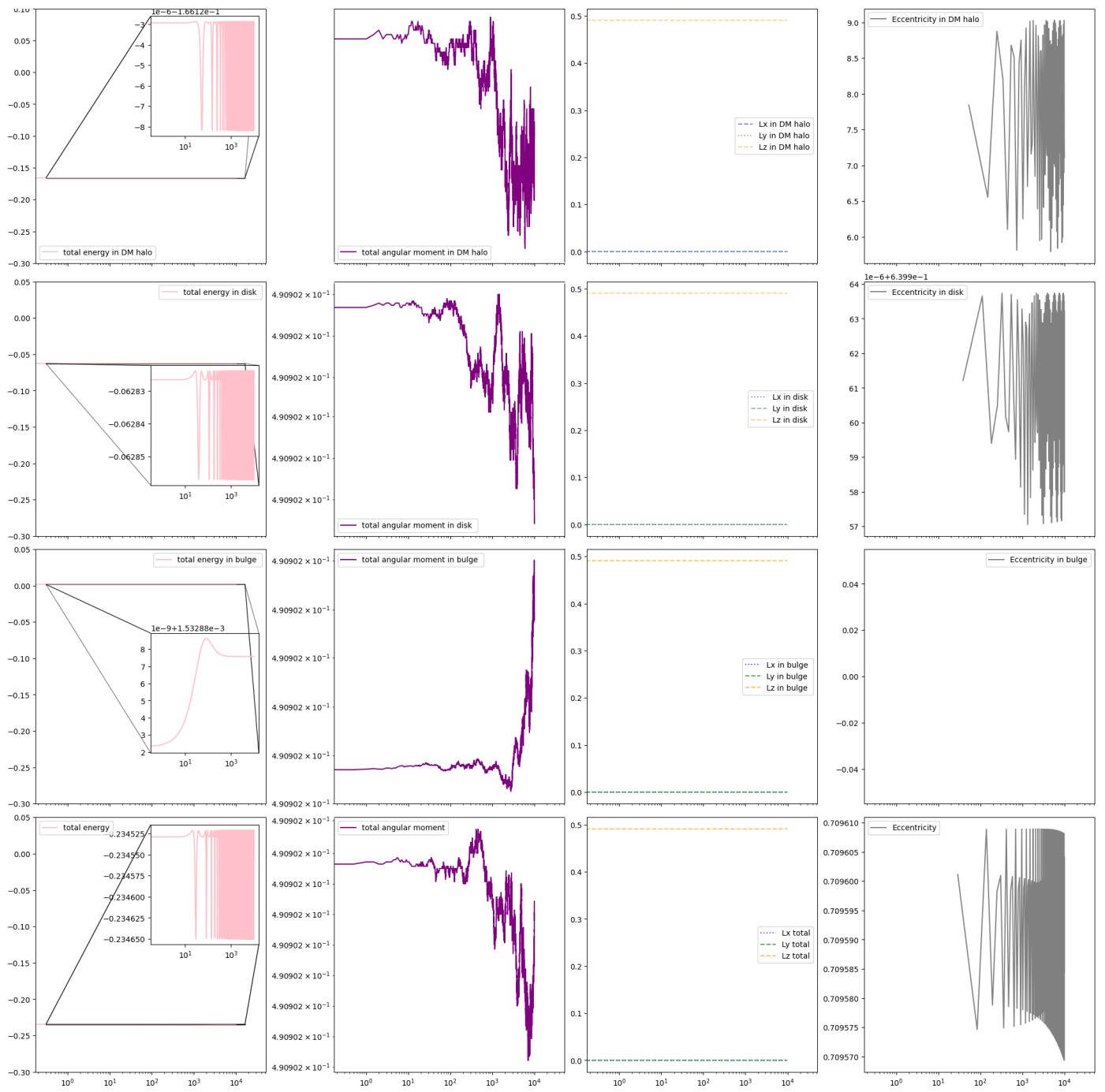
29
30 ax[0,2].plot(t, aDM_Lx, color='blue', alpha=0.5,label='Lx in DM halo', linestyle
31 ax[0,2].plot(t, aDM_Ly, color='green', alpha=0.5, label='Ly in DM halo', lines
32 ax[0,2].plot(t, aDM_Lz, color='orange', alpha=0.5, label='Lz in DM halo', line
33 ax[0,2].legend(loc='best')
34
35 ax[0,3].plot(timedMecc, aDM_ecc, color='gray', label='Eccentricity in DM halo'
36 ax[0,3].legend(loc='best')
37
38
39 ax[1,0].set_xscale('log')
40 ax[1,0].plot(t, a_disk_totalenergy, color='pink', label='total energy in disk
41 ax[1,0].legend(loc='best')
42 ax[1,0].set_ylim(-0.3,0.05)
43
44 axins1 = ax[1,0].inset_axes([0.5, 0.2, 0.47, 0.47])
45 # axins = zoomed_inset_axes(ax[0,0], 3, loc=4) # zoom = 6
46 # sub region of the original image
47
48 axins1.set_xscale('log')
49 axins1.plot(t, a_disk_totalenergy, color='pink')
50
51 # draw a bbox of the region of the inset axes in the parent axes and
52 # connecting lines between the bbox and the inset axes area
53 ax[1,0].indicate_inset_zoom(axins1, edgecolor="black")
54 mark_inset(ax[1,0], axins1, loc1=2, loc2=4, fc='none',ec='0.1')
55
56 ax[1,1].plot(t, a_disk_totalangularmoment, color='purple', label='total angula
57 ax[1,1].legend(loc='best')
58 ax[1,1].set_yscale('log')
59
60
61 ax[1,2].plot(t, a_disk_Lx, color='blue', alpha=0.5,label='Lx in disk ',linesty
62 ax[1,2].plot(t, a_disk_Ly, color='green', alpha=0.5, label='Ly in disk ',lines
63 ax[1,2].plot(t, a_disk_Lz, color='orange', alpha=0.5, label='Lz in disk ',line
64 ax[1,2].legend(loc='best')
65
66 ax[1,3].plot(timediskecc, a_disk_ecc, color='gray', label='Eccentricity in dis
67 ax[1,3].legend(loc='best')
68
69
70
71 ax[2,0].set_xscale('log')
72 ax[2,0].plot(t, a_bulge_totalenergy, color='pink', label='total energy in bulg
73 ax[2,0].legend(loc='best')
74 ax[2,0].set_ylim(-0.3,0.05)
75
76 axins2 = ax[2,0].inset_axes([0.5, 0.2, 0.47, 0.47])

```

```

77 axins2.set_xscale('log')
78 axins2.plot(t, a_bulge_totalenergy, color='pink')
79
80 # draw a bbox of the region of the inset axes in the parent axes and
81 # connecting lines between the bbox and the inset axes area
82 ax[2,0].indicate_inset_zoom(axins2, edgecolor="black")
83 mark_inset(ax[2,0], axins2, loc1=2, loc2=4, fc='none',ec='0.1')
84
85 ax[2,1].plot(t, a_bulge_totalangularmoment, color='purple', label='total angul
86 ax[2,1].legend(loc='best')
87 ax[2,1].set_yscale('log')
88
89 ax[2,2].plot(t, a_bulge_Lx, color='blue', alpha=0.7,label='Lx in bulge ',lines
90 ax[2,2].plot(t, a_bulge_Ly, color='green', alpha=0.7, label='Ly in bulge', lin
91 ax[2,2].plot(t, a_bulge_Lz, color='orange', alpha=0.7, label='Lz in bulge ',li
92 ax[2,2].legend(loc='best')
93
94 ax[2,3].plot(timebulgeecc, a_bulge_ecc, color='gray', label='Eccentricity in b
95 ax[2,3].legend(loc='best')
96
97
98
99 ax[3,0].set_xscale('log')
100 ax[3,0].plot(t, a_totalenergy, color='pink', label='total energy')
101 ax[3,0].legend(loc='best')
102 ax[3,0].set_ylim(-0.3,0.05)
103 axins3 = ax[3,0].inset_axes([0.5, 0.5, 0.47, 0.47])
104 axins3.set_xscale('log')
105 axins3.plot(t, a_totalenergy, color='pink')
106 # draw a bbox of the region of the inset axes in the parent axes and
107 # connecting lines between the bbox and the inset axes area
108 ax[3,0].indicate_inset_zoom(axins3, edgecolor="black")
109 mark_inset(ax[3,0], axins3, loc1=2, loc2=4, fc='none',ec='0.1')
110
111 ax[3,1].plot(t, a_totalangularmoment, color='purple', label='total angular mom
112 ax[3,1].legend(loc='best')
113 ax[3,1].set_yscale('log')
114
115 ax[3,2].plot(t, a_Lx, color='blue', alpha=0.7,label='Lx total', linestyle=':')
116 ax[3,2].plot(t, a_Ly, color='green', alpha=0.7, label='Ly total', linestyle='-
117 ax[3,2].plot(t, a_Lz, color='orange', alpha=0.7, label='Lz total', linestyle='
118 ax[3,2].legend(loc='best')
119
120 ax[3,3].plot(timeecc, a_ecc, color='gray', label='Eccentricity')
121 ax[3,3].legend(loc='best')
122

```





Notemos que el bulbo no muestra excentricidad para la estrella a. Esto es porque en la definicion del potencial del bulbo, su masa es demasiado baja para atraer gravitatoriamente a la estrella a y dejarla ligada. Se puede corroborar observando que la orbita que toma la estrella cuando solo entra al bulbo tiene  $E > 0$ . Y dentro de la segunda columna, podemos ver que en realidad el ruido que se nota sobre la curva de momento angular es infimo, por ende este es constante en general, teniendo claras variaciones para el caso del bulbo, que no representa necesariamente error debido al metodo numerico de aproximacion (leap-frog, default integrator [\[https://gala.adrian.pw/en/latest/tutorials/pyia-gala-orbit.html\]](https://gala.adrian.pw/en/latest/tutorials/pyia-gala-orbit.html)) y tiene relacion con lo mencionado previamente.

```

1 position_x_a = a_orbit.x.value
2 velocity_x_a = a_orbit.vel.d_x.value
3 position_y_a = a_orbit.y.value
4 velocity_y_a = a_orbit.vel.d_y.value
5 position_z_a = a_orbit.z.value
6 velocity_z_a = a_orbit.vel.d_z.value
7 delta = 0.2
8
9 poincare_index_a = np.where((position_y_a >= -delta) & (position_y_a <= delta)
10 #poincare_index_c = np.where((position_y_c == 0) & (velocity_y_c > 0))[0]
11
12 poincare_position_x_a = position_x_a[poincare_index_a]
13 poincare_velocity_x_a = velocity_x_a[poincare_index_a]
14 poincare_velocity_y_a = velocity_y_a[poincare_index_a]
15 poincare_position_y_a = position_y_a[poincare_index_a]
16 poincare_position_z_a = position_z_a[poincare_index_a]
17 poincare_velocity_z_a = velocity_z_a[poincare_index_a]


1 position_x_b = b_orbit.x.value
2 velocity_x_b = b_orbit.vel.d_x.value
3 position_y_b = b_orbit.y.value
4 velocity_y_b = b_orbit.vel.d_y.value
5 position_z_b = b_orbit.z.value
6 velocity_z_b = b_orbit.vel.d_z.value
7 delta = 0.2
8
9 poincare_index_b = np.where((position_y_b >= -delta) & (position_y_b <= delta)
10 #poincare_index_b = np.where((position_y_b == 0) & (velocity_y_b > 0))[0]
11
12 poincare_position_x_b = position_x_b[poincare_index_b]
13 poincare_velocity_x_b = velocity_x_b[poincare_index_b]
14 poincare_velocity_y_b = velocity_y_b[poincare_index_b]
15 poincare_position_y_b = position_y_b[poincare_index_b]
16 poincare_position_z_b = position_z_b[poincare_index_b]
17 poincare_velocity_z_b = velocity_z_b[poincare_index_b]


1 fig, ax = plt.subplots()
2
3 ax.scatter(poincare_position_x_a, poincare_velocity_x_a,color='red', alpha=0.5
4 ax.scatter(poincare_position_x_b, poincare_velocity_x_b, color='cyan', alpha=0
5
6 axins = ax.inset_axes([0.15, 0.15, 0.47, 0.47])
7 axins.set_yticklabels([])
8 axins.set_xticklabels([])
9 axins.scatter(poincare_position_x_b, poincare_velocity_x_b, color='cyan', s=1)
10

```

```

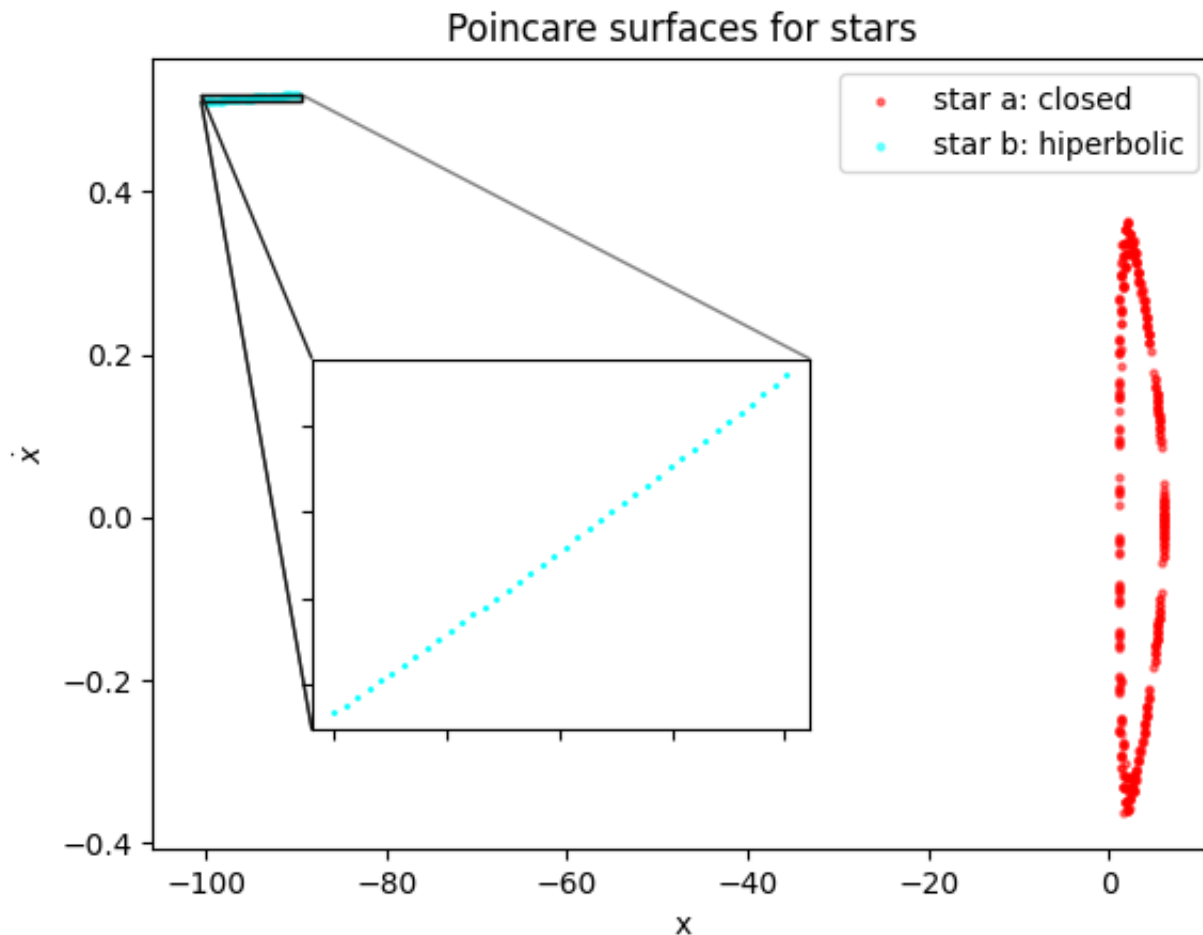
11
12 ax.indicate_inset_zoom(axins, edgecolor="black")
13 mark_inset(ax, axins, loc1=2, loc2=3, fc='none',ec='0.1')
14
15 plt.title('Poincare surfaces for stars')
16 plt.legend()
17 plt.xlabel('x')
18 plt.ylabel('$\dot{x}$')
19 plt.show()
20

```

```

⇨ <>:18: SyntaxWarning: invalid escape sequence '\d'
<>:18: SyntaxWarning: invalid escape sequence '\d'
/var/folders/_3/b44zg9zx3mjcpbfyw4_4d15c0000gn/T/ipykernel_2837/3267452816.py:
plt.ylabel('$\dot{x}$')

```



```

1 # Disk component
2 position_x_disk = a_orbit_disk.x.value
3 velocity_x_disk = a_orbit_disk.vel.d_x.value
4 position_y_disk = a_orbit_disk.y.value
5 velocity_y_disk = a_orbit_disk.vel.d_y.value
6 position_z_disk = a_orbit_disk.z.value
7 velocity_z_disk = a_orbit_disk.vel.d_z.value
8

```

```

9 # Bulge component
10 position_x_bulge = a_orbit_bulge.x.value
11 velocity_x_bulge = a_orbit_bulge.vel.d_x.value
12 position_y_bulge = a_orbit_bulge.y.value
13 velocity_y_bulge = a_orbit_bulge.vel.d_y.value
14 position_z_bulge = a_orbit_bulge.z.value
15 velocity_z_bulge = a_orbit_bulge.vel.d_z.value
16
17 # DM component
18 position_x_dm = a_orbit_DM.x.value
19 velocity_x_dm = a_orbit_DM.vel.d_x.value
20 position_y_dm = a_orbit_DM.y.value
21 velocity_y_dm = a_orbit_DM.vel.d_y.value
22 position_z_dm = a_orbit_DM.z.value
23 velocity_z_dm = a_orbit_DM.vel.d_z.value
24
25 delta = 0.2
26
27 # Selecting particles
28 poincare_index_disk = np.where((position_y_disk >= -delta) & (position_y_disk <= delta))
29 poincare_index_bulge = np.where((position_y_bulge >= -delta) & (position_y_bulge <= delta))
30 poincare_index_dm = np.where((position_y_dm >= -delta) & (position_y_dm <= delta))
31
32 # Selecting poincare section data for each component
33 poincare_position_x_disk = position_x_disk[poincare_index_disk]
34 poincare_velocity_x_disk = velocity_x_disk[poincare_index_disk]
35 poincare_velocity_y_disk = velocity_y_disk[poincare_index_disk]
36 poincare_position_y_disk = position_y_disk[poincare_index_disk]
37 poincare_position_z_disk = position_z_disk[poincare_index_disk]
38 poincare_velocity_z_disk = velocity_z_disk[poincare_index_disk]
39
40 poincare_position_x_bulge = position_x_bulge[poincare_index_bulge]
41 poincare_velocity_x_bulge = velocity_x_bulge[poincare_index_bulge]
42 poincare_velocity_y_bulge = velocity_y_bulge[poincare_index_bulge]
43 poincare_position_y_bulge = position_y_bulge[poincare_index_bulge]
44 poincare_position_z_bulge = position_z_bulge[poincare_index_bulge]
45 poincare_velocity_z_bulge = velocity_z_bulge[poincare_index_bulge]
46
47 poincare_position_x_dm = position_x_dm[poincare_index_dm]
48 poincare_velocity_x_dm = velocity_x_dm[poincare_index_dm]
49 poincare_velocity_y_dm = velocity_y_dm[poincare_index_dm]
50 poincare_position_y_dm = position_y_dm[poincare_index_dm]
51 poincare_position_z_dm = position_z_dm[poincare_index_dm]
52 poincare_velocity_z_dm = velocity_z_dm[poincare_index_dm]
53

```

```

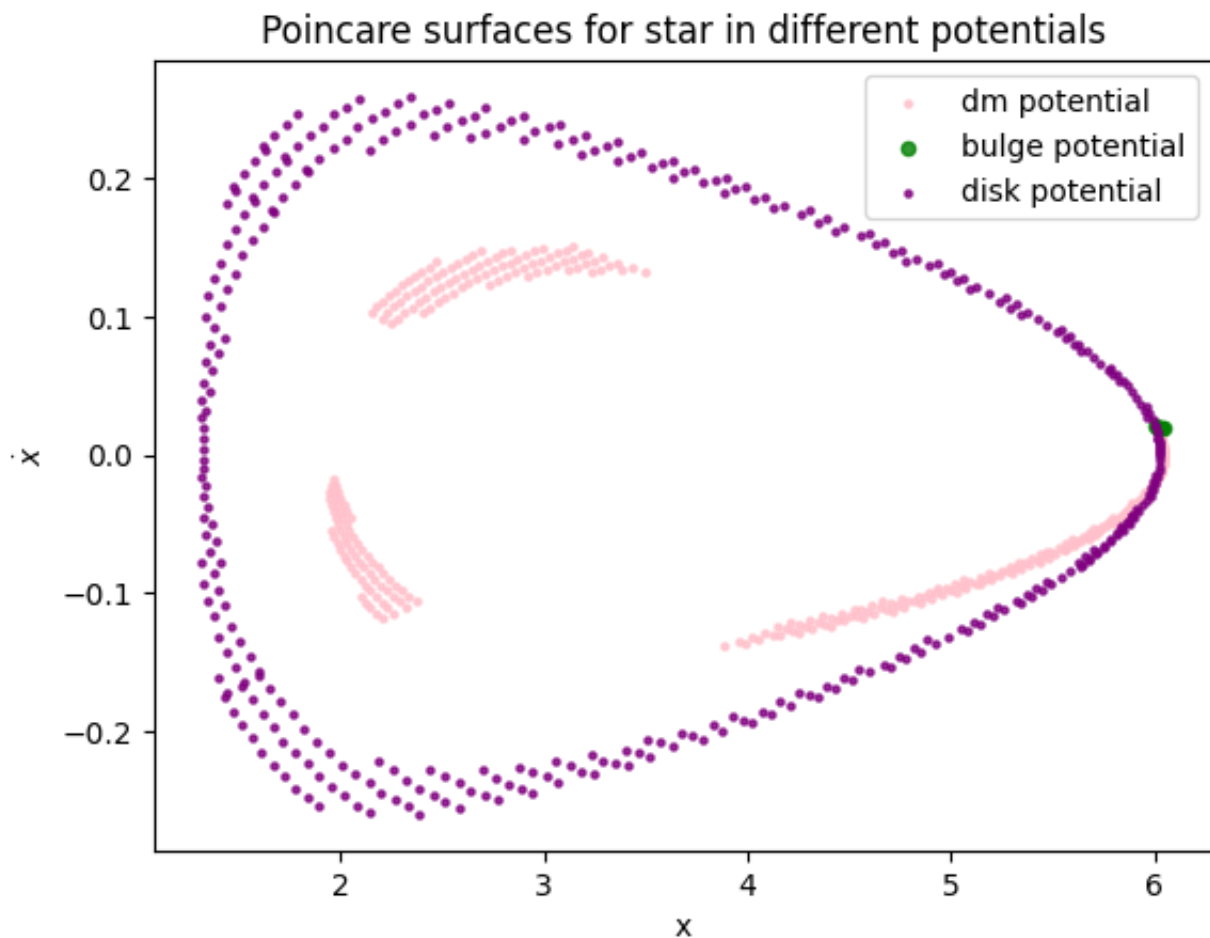
1 fig, ax = plt.subplots()
2
3 ax.scatter(poincare_position_x_dm, poincare_velocity_x_dm,color='pink', alpha=0.5)
4 ax.scatter(poincare_position_x_bulge, poincare_velocity_x_bulge, color='green'
5 ax.scatter(poincare_position_x_disk, poincare_velocity_x_disk, color='purple',
6
7
8 plt.title('Poincare surfaces for star in different potentials')
9 plt.legend()
10 plt.xlabel('x')
11 plt.ylabel('$\dot{x}$')

```

```

➡ <>:11: SyntaxWarning: invalid escape sequence '\d'
<>:11: SyntaxWarning: invalid escape sequence '\d'
/var/folders/_3/b44zg9zx3mjcpbfyw4_4d15c0000gn/T/ipykernel_2837/3715340710.py:
  plt.ylabel('$\dot{x}$')
Text(0, 0.5, '$\dot{x}$')

```



3. Utilizar orbita circular en un potencial que contenga solo DM halo y disco.

$$x_i = (x, 0, 0)$$

$$v_i = (0, v_{circ}, 0)$$

```

1 import gala.dynamics as gd

1 xi = (10,0,0)
2 vi = (0,150,0)
3 ics = gd.PhaseSpacePosition(pos = xi*u.kpc,
4                               vel = vi*u.km/u.s)
5 orbiti = gp.Hamiltonian(total_potential['dm_halo']+total_potential['disk']).in
6
7 position_x_i = orbiti.x.value
8 velocity_x_i = orbiti.vel.d_x.value
9 position_y_i = orbiti.y.value
10 velocity_y_i = orbiti.vel.d_y.value
11 position_z_i = orbiti.z.value
12 velocity_z_i = orbiti.vel.d_z.value
13 delta = 0.2
14
15 poincare_index_i = np.where((position_y_i >= -delta) & (position_y_i <= delta)
16 #poincare_index_c = np.where((position_y_c == 0) & (velocity_y_c > 0))[0]
17
18 poincare_position_x_i = position_x_i[poincare_index_i]
19 poincare_velocity_x_i = velocity_x_i[poincare_index_i]
20 poincare_velocity_y_i = velocity_y_i[poincare_index_i]
21 poincare_position_y_i = position_y_i[poincare_index_i]
22 poincare_position_z_i = position_z_i[poincare_index_i]
23 poincare_velocity_z_i = velocity_z_i[poincare_index_i]

```

4. Perturbar levemente la orbita con un  $\Delta v_x \ll v_y$

```

1 xii = (10,0,0)
2 vii = (-2,150,0)
3 ics = gd.PhaseSpacePosition(pos = xii*u.kpc,
4                               vel = vii*u.km/u.s)
5 orbitii = gp.Hamiltonian(total_potential['dm_halo']+total_potential['disk']).i
6
7 position_x_ii = orbitii.x.value
8 velocity_x_ii = orbitii.vel.d_x.value
9 position_y_ii = orbitii.y.value
10 velocity_y_ii = orbitii.vel.d_y.value
11 position_z_ii = orbitii.z.value
12 velocity_z_ii = orbitii.vel.d_z.value
13 delta = 0.2
14
15 poincare_iindex_ii = np.where((position_y_ii >= -delta) & (position_y_ii <= de
16 #poincare_iindex_c = np.where((position_y_c == 0) & (velocity_y_c > 0))[0]
17
18 poincare_position_x_ii = position_x_ii[poincare_iindex_ii]
19 poincare_velocity_x_ii = velocity_x_ii[poincare_iindex_ii]
20 poincare_velocity_y_ii = velocity_y_ii[poincare_iindex_ii]
21 poincare_position_y_ii = position_y_ii[poincare_iindex_ii]
22 poincare_position_z_ii = position_z_ii[poincare_iindex_ii]
23 poincare_velocity_z_ii = velocity_z_ii[poincare_iindex_ii]

```

```

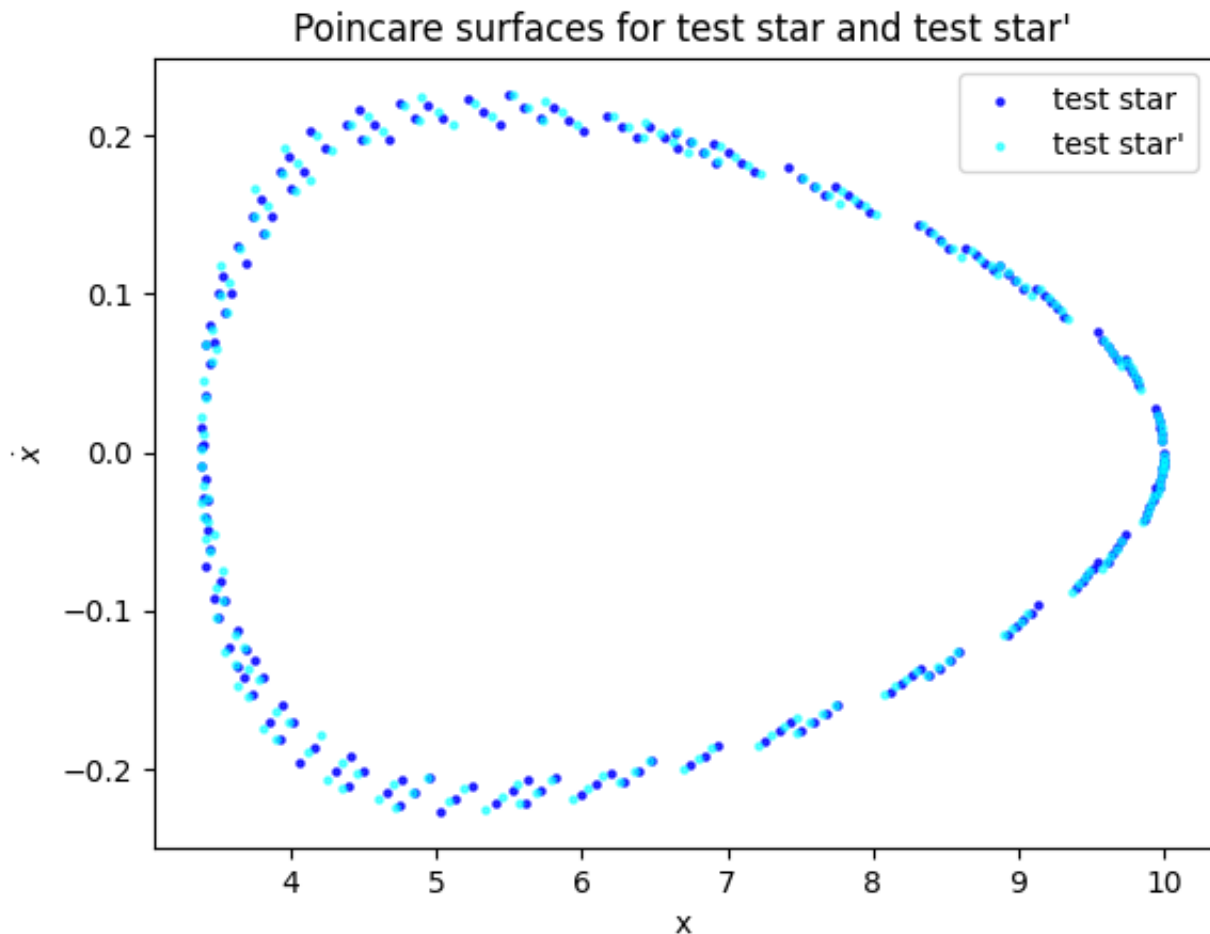
1 fig, ax = plt.subplots()
2
3 ax.scatter(poincare_position_x_i, poincare_velocity_x_i,color='blue', alpha=0.1)
4 ax.scatter(poincare_position_x_ii, poincare_velocity_x_ii,color='cyan', alpha=0.1)
5
6
7 plt.title('Poincare surfaces for test star and test star\')
8 plt.legend()
9 plt.xlabel('x')
10 plt.ylabel('$\dot{x}$')

```

```

↳ <>:10: SyntaxWarning: invalid escape sequence '\d'
<>:10: SyntaxWarning: invalid escape sequence '\d'
/var/folders/_3/b44zg9zx3mjcpbfyw4_4d15c0000gn/T/ipykernel_2837/4103845929.py:
  plt.ylabel('$\dot{x}$')
  Text(0, 0.5, '$\dot{x}$')

```



- Discutir los resultados

Se nota que la superficie de poincare que representa la estrella levemente perturbada se mueve respecto a la posicion inicial, esto significa que el tipo de orbita es igual (cerrada) pero, las perturbaciones deforman un poco las superficies de poincare.



---

## ✓ Practica III

Objetivo: Comprender la densidad local de un enjambre de orbitas y su potencial

1. Utilizando un par de orbitas sobre un potencial esferico de NFW tipo MW, calcular y graficar su evolucion temporal en  $\sim 10$  Gyrs

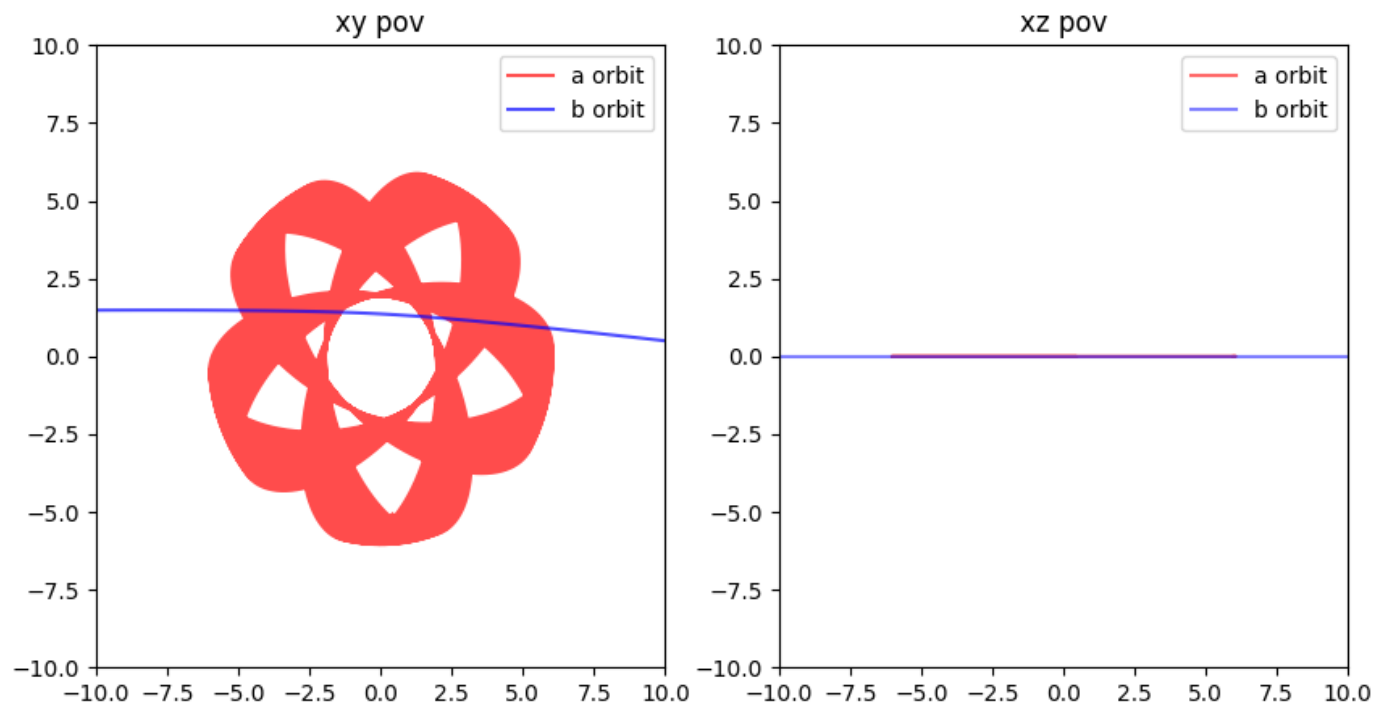
Seguiremos utilizando el potencial que tenemos definido que es tipo MW para el DM halo, y las estrellas a (orbita cerrada) y b (orbita hiperbolica)

```

1 fig, ax = plt.subplots(1, 2, figsize=(10,5))
2 ax[0].plot(a_orbit_DM.pos.x ,a_orbit_DM.pos.y,
3           color='red', alpha=0.7, label='a orbit')
4 ax[0].plot(b_orbit_DM.pos.x, b_orbit_DM.pos.y,
5           color='blue', alpha=0.7, label='b orbit')
6 ax[0].set_xlim(-10,10)
7 ax[0].set_ylim(-10,10)
8
9
10 ax[1].plot(a_orbit_DM.pos.x, a_orbit_DM.pos.z,
11           color='red', alpha=0.6, label='a orbit')
12 ax[1].plot(b_orbit_DM.pos.x, b_orbit_DM.pos.z,
13           color='blue', alpha=0.5, label='b orbit')
14 ax[1].set_xlim(-10,10)
15 ax[1].set_ylim(-10, 10)
16
17 ax[0].set_title('xy pov')
18 ax[1].set_title('xz pov')
19
20 ax[0].legend()
21 ax[1].legend()

```

 <matplotlib.legend.Legend at 0x14f53d250>



## 2. Definir un volumen alrededor con las siguientes características:

- Gaussiane en 3D para posiciones y velocidades
- Condiciones iniciales como la media de posición y velocidad
- $\sigma_r$  y  $\sigma_v$  con orden de  $10^{-5}$  kpc y 1 km/s, respectivamente.
- Tener en total 1000 condiciones iniciales un poco distintas
- Integrar todo y observar el enjambre en  $\sim 10$  Gyrs

```
1 n_orbits = 1000
2
3 ics_a = gd.PhaseSpacePosition(pos = a_star_p*u.kpc,
4                               vel = a_star_v*u.km/u.s)
5 ics_b = gd.PhaseSpacePosition(pos = b_star_p*u.kpc,
6                               vel = b_star_v*u.km/u.s)
7
8
9 new_pos_a = np.random.normal(ics_a.pos.xyz.to(u.pc).value, 0.01,
10                             size=(n_orbits,3)).T*u.pc
11 new_vel_a = np.random.normal(ics_a.vel.d_xyz.to(u.km/u.s).value, 1,
12                             size=(n_orbits,3)).T * u.km/u.s
13
14 new_ics_a = gd.PhaseSpacePosition(pos=new_pos_a, vel=new_vel_a)
15 orbit_a = gp.Hamiltonian(total_potential['dm_halo']).integrate_orbit(ics_a, d
16
17 orbits_total_a = gp.Hamiltonian(total_potential['dm_halo']).integrate_orbit(ne
18
19
20 new_pos_b = np.random.normal(ics_b.pos.xyz.to(u.pc).value, 0.01,
21                             size=(n_orbits,3)).T*u.pc
22 new_vel_b = np.random.normal(ics_b.vel.d_xyz.to(u.km/u.s).value, 1.,
23                             size=(n_orbits,3)).T * u.km/u.s
24
25 new_ics_b = gd.PhaseSpacePosition(pos=new_pos_b, vel=new_vel_b)
26 orbit_b = gp.Hamiltonian(total_potential['dm_halo']).integrate_orbit(ics_b, d
27
28 orbits_total_b = gp.Hamiltonian(total_potential['dm_halo']).integrate_orbit(ne
29
```

```

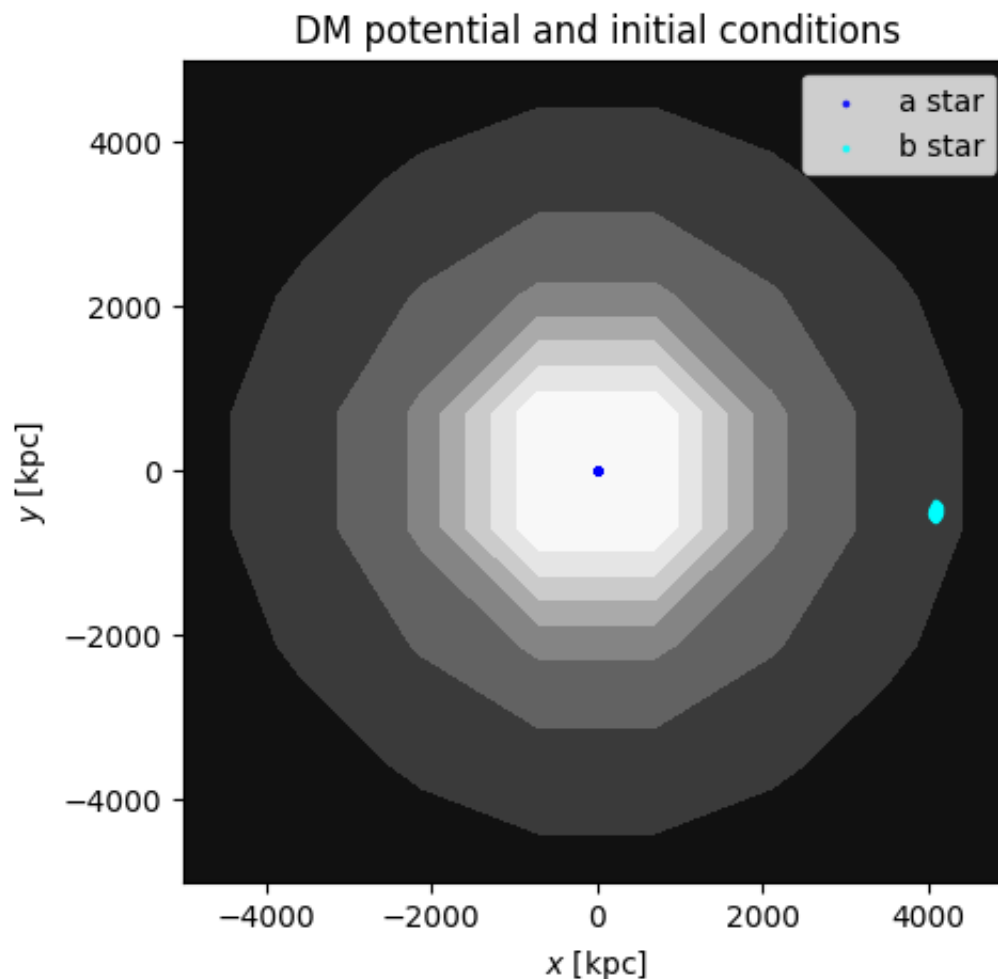
1 grid = np.linspace(-5000,5000,8)
2 fig, ax = plt.subplots(1, 1, figsize=(5,5))
3 fig = total_potential['dm_halo'].plot_contours(grid=(grid, grid, 0), cmap='Gre
4 fig = orbits_total_a[-1].plot(['x','y'], color='blue', s=10,
5                               alpha=0.8, axes=[ax], auto_aspect=False,
6                               label='a star')
7 fig = orbits_total_b[-1].plot(['x','y'], color='cyan', s=10,
8                               alpha=0.8, axes=[ax], auto_aspect=False,
9                               label='b star')
10 ax.set_xlim(-5000,5000)
11 ax.set_ylim(-5000,5000)
12 ax.legend()
13 ax.set_title('DM potential and initial conditions')

```

```

➡ Text(0.5, 1.0, 'DM potential and initial conditions')

```

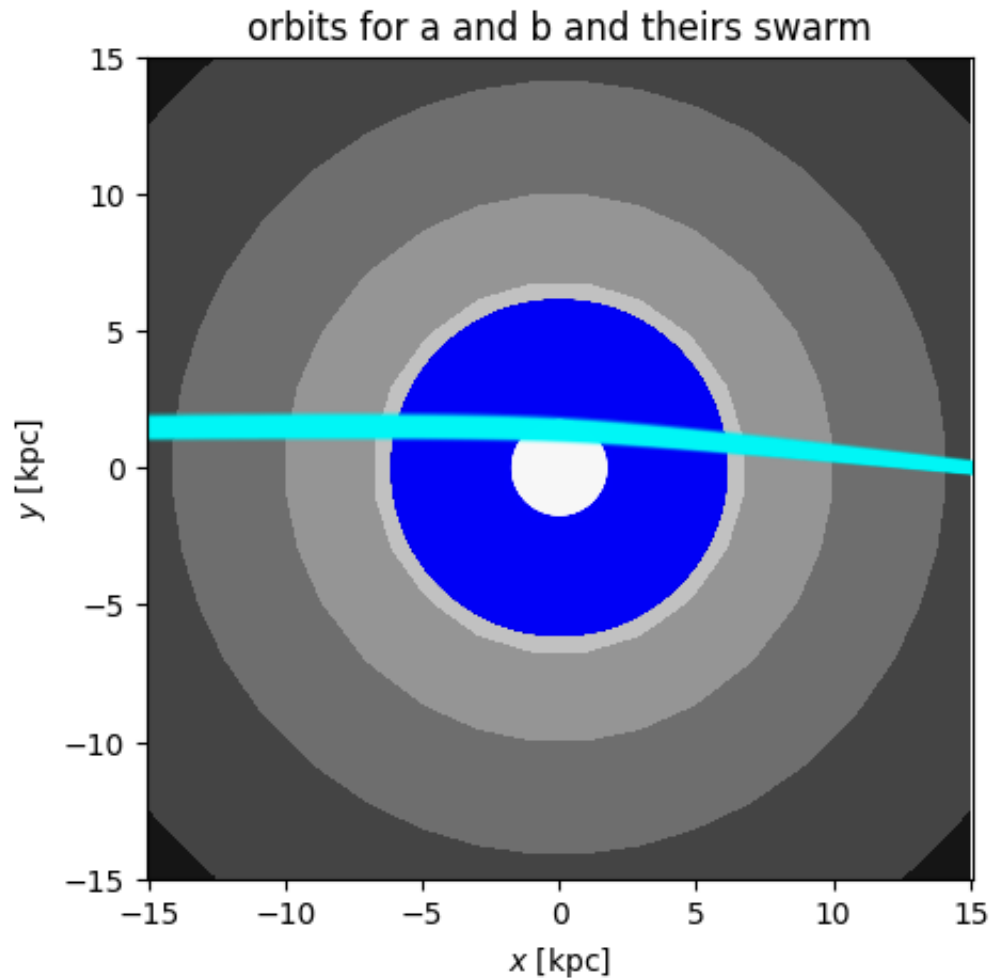


```

1 grid = np.linspace(-15,15,16)
2 fig, ax = plt.subplots(1, 1, figsize=(5,5))
3 fig = total_potential['dm_halo'].plot_contours(grid=(grid, grid, 0), cmap='Gre
4 fig = orbits_total_a.plot(['x','y'], color='blue', axes=[ax], alpha=0.1, label:
5 fig = orbits_total_b.plot(['x','y'], color='cyan', axes=[ax], alpha=0.1, label:
6 ax.set_title('orbits for a and b and theirs swarm')
7 ax.set_xlim(-15, 15)
8 ax.set_ylim(-15, 15)
9

```

→ (-15.0, 15.0)



3. Graficar y observar evolucion de  $\sigma_r$  y  $\sigma_{\dot{r}}$ .

- sigma total
- sigma por componente ( $[x,y,z]$ ,  $[\dot{x}, \dot{y}, \dot{z}]$ )

Nuestros parametros iniciales son:

- $a_{\text{star}_p} = [6,0,0]$
- $a_{\text{star}_v} = [20,80,0]$
- $b_{\text{star}_p} = [-100,0,0]$
- $b_{\text{star}_v} = [500,10,0]$

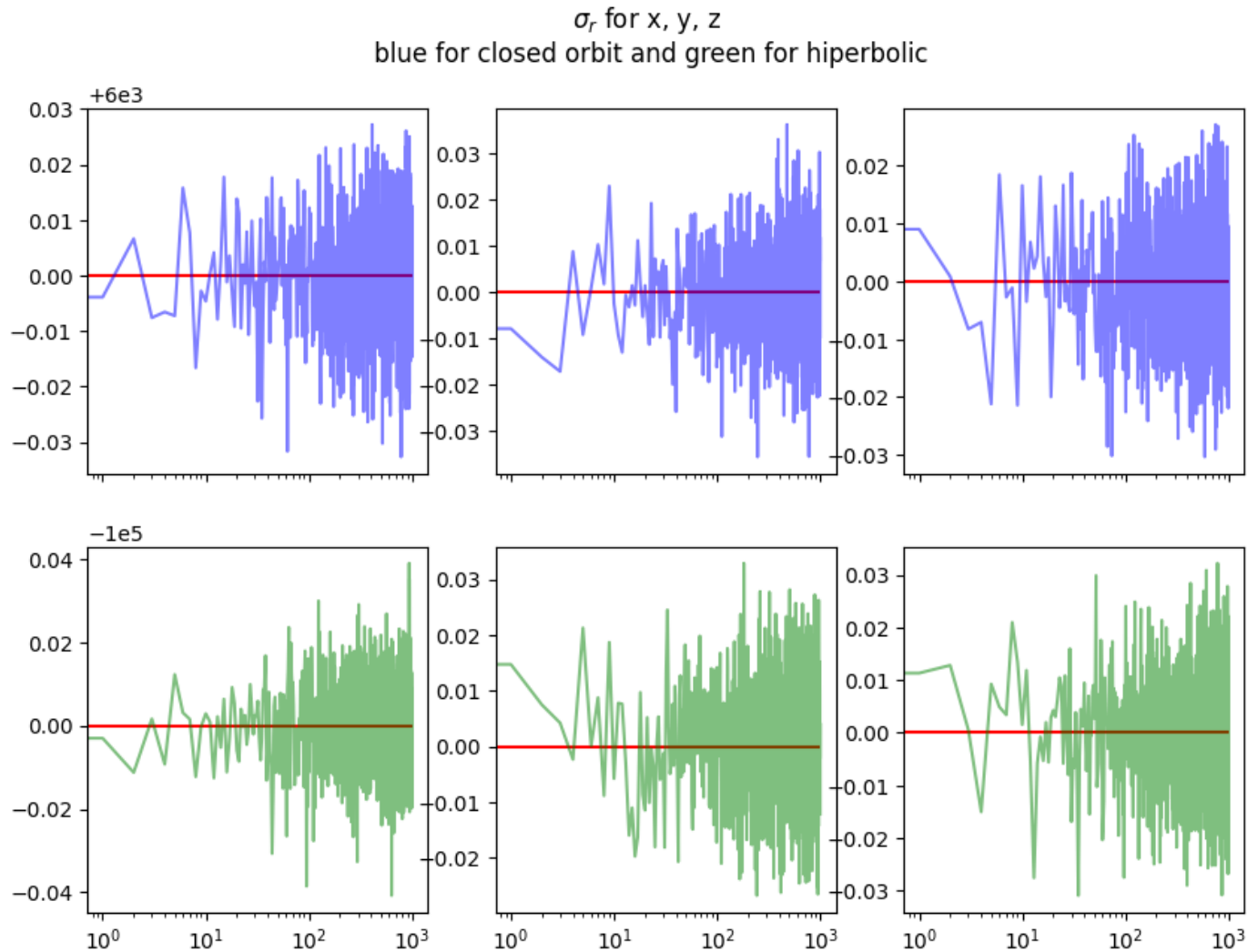
Las cuales estan en kiloparsec. Los  $\sigma$  seran graficados en parsec. Por ende los errores son muy pequenitos.

```
1 fig, ax = plt.subplots(2,3, sharex=True, figsize=(10,7))
2 xdata= np.arange(0,1000)
3
4 #x
5 ax[0,0].hlines(y=6000, xmin=0, xmax=1000, color='red')
6 ax[0,0].plot(xdata, new_pos_a[0], color='blue', alpha=0.5)
7
8 ax[0,0].set_xscale('log')
9 ax[1,0].hlines(y=-100000, xmin=0, xmax=1000, color='red')
10 ax[1,0].plot(xdata, new_pos_b[0], color='green', alpha=0.5)
11
12 #y
13 ax[0,1].hlines(y=0, xmin=0, xmax=1000, color='red')
14 ax[0,1].plot(xdata, new_pos_a[1], color='blue', alpha=0.5)
15
16 ax[1,1].hlines(y=0, xmin=0, xmax=1000, color='red')
17 ax[1,1].plot(xdata, new_pos_b[1], color='green', alpha=0.5)
18
19
20 #z
21 ax[0,2].hlines(y=0, xmin=0, xmax=1000, color='red')
22 ax[0,2].plot(xdata, new_pos_a[2], color='blue', alpha=0.5)
23
24 ax[1,2].hlines(y=0, xmin=0, xmax=1000, color='red')
25 ax[1,2].plot(xdata, new_pos_b[2], color='green', alpha=0.5)
26
27 fig.suptitle('$\sigma_{\{r\}}$ for x, y, z \n blue for closed orbit and green for |
28
29
30
31
```

```

➡ <>:27: SyntaxWarning: invalid escape sequence '\s'
<>:27: SyntaxWarning: invalid escape sequence '\s'
/var/folders/_3/b44zg9zx3mjcpbfyw4_4d15c0000gn/T/ipykernel_2837/521037230.py:2
    fig.suptitle('$\sigma_{r}$ for x, y, z \n blue for closed orbit and green fc
Text(0.5, 0.98, '$\sigma_{r}$ for x, y, z \n blue for closed orbit and green
for hiperbolic')

```



```

1 fig, ax = plt.subplots(2,3, sharex=True, figsize=(10,7))
2 xdata= np.arange(0,1000)
3
4 ax[0,0].set_xscale('log')
5
6 ax[0,0].plot(xdata, new_vel_a[0], color='purple', alpha=0.5)
7 ax[0,0].hlines(y=20, xmin=0, xmax=1000, color='red')

```

```

8
9 ax[1,0].plot(xdata, new_vel_b[0], color='black', alpha=0.5)
10 ax[1,0].hlines(y=500,xmin=0, xmax=1000, color='red')
11
12 ax[0,1].plot(xdata, new_vel_a[1], color='purple', alpha=0.5)
13 ax[0,1].hlines(y=80, xmin=0, xmax=1000, color='red')
14
15 ax[1,1].plot(xdata, new_vel_b[1], color='black', alpha=0.5)
16 ax[1,1].hlines(y=10, xmin=0, xmax=1000, color='red')
17
18 ax[0,2].plot(xdata, new_vel_a[2], color='purple', alpha=0.5)
19 ax[0,2].hlines(y=0, xmin=0, xmax=1000, color='red')
20
21 #0
22 ax[1,2].plot(xdata, new_vel_b[2], color='black', alpha=0.5)
23 ax[1,2].hlines(y=0, xmin=0, xmax=1000, color='red')
24
25 fig.suptitle('$\sigma_{\dot{r}}$ for x, y, z \n purple for closed orbit and gr
26
27

```

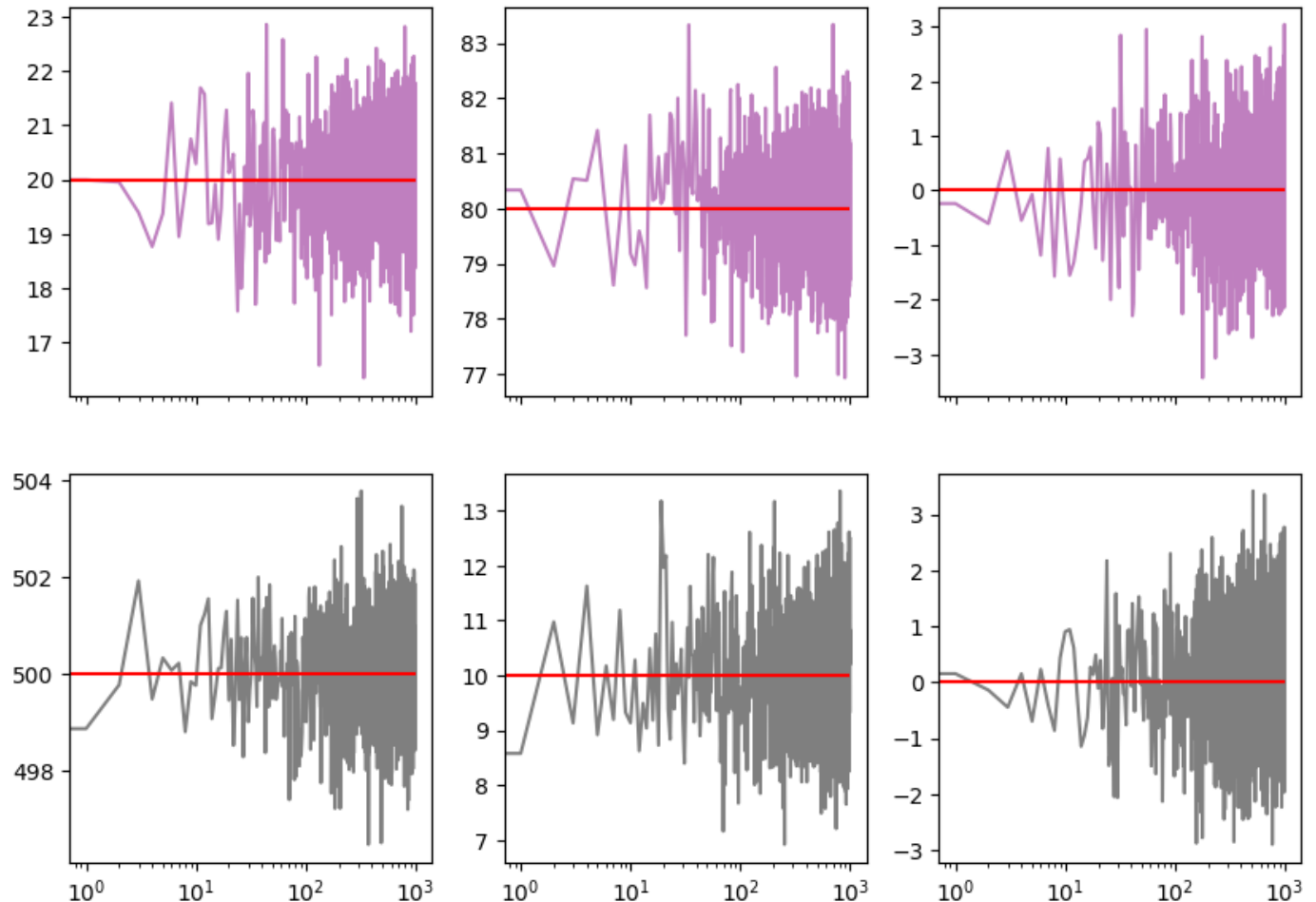


```

<>:25: SyntaxWarning: invalid escape sequence '\s'
<>:25: SyntaxWarning: invalid escape sequence '\s'
/var/folders/_3/b44zg9zx3mjcpbfyw4_4d15c0000gn/T/ipykernel_2837/1194764760.py:
    fig.suptitle('$\sigma_{\dot{r}}$ for x, y, z \n purple for closed orbit and
Text(0.5, 0.98, '$\sigma_{\dot{r}}$ for x, y, z \n purple for closed orbit
and gray for hiperbolic')

```

$\sigma_r$  for x, y, z  
purple for closed orbit and gray for hiperbolic



Si observamos y superponemos los graficos de posicion y velocidad para la misma componente donde ambos parten en cero, se puede notar lo visto en clases del paper de Gomez+2013, es decir, el comportamiento de  $x$  y  $\dot{x}$  como sin y cos (ya que  $\dot{x}$  es la derivada de  $x$ )

4. Tomar esferas de 1 y 2,5 kpc de radio, centradas a todo tiempo  $t$ , en la órbita central del enjambre.

En cada paso de integración, calcular la densidad de partículas dentro de estas esferas.

```
1 fig, anim = orbits_total_a[:1000].cylindrical.animate(components=['rho', 'z'],
2                                                         stride=10)
3 anim.save('closedswarmorbit.gif')
4 plt.close()
```

➡ MovieWriter ffmpeg unavailable; using Pillow instead.

```
1 fig, anim = orbits_total_b[:1000].cylindrical.animate(components=['rho', 'z'],
2                                                         stride=10)
3 anim.save('hiperbolicswarmorbit.gif')
4 plt.close()
```

➡ MovieWriter ffmpeg unavailable; using Pillow instead.

Con las animaciones creadas en la celda anterior, (closedswamorbit.gif y hiperbolicswarmorbit.gif), podemos observar que la densidad del enjambre aumenta a medida que se acerca mas a la parte mas masiva del potencial, y disminuye a medida que se aleja.

```
1 density25 = []
2 density10 = []
3
4 for i in range(len(orbital_total_a.t)):
5     difference = orbital_total_a.pos[i,:] - orbit_a.pos[i]
6     distance_25 = difference.norm()/(2.5*u.kpc)
7     distance_10 = difference.norm()/(1.0*u.kpc)
8
9     pos_25 = np.where(distance_25<=1)[0]
10    pos_10 = np.where(distance_10<=1)[0]
11
12    densitypoints_25 = distance_25[pos_25]
13    densitypoints_10 = distance_10[pos_10]
14
15    vol25 = (4/3) * np.pi * (2.5**3)
16    densitytotal_25 = len(densitypoints_25) / vol25
17    density25.append(densitytotal_25)
18
19    vol10 = (4/3) * np.pi * (1.0**3)
20    densitytotal_10 = len(densitypoints_10) / vol10
21    density10.append(densitytotal_10)
22
23
24
```

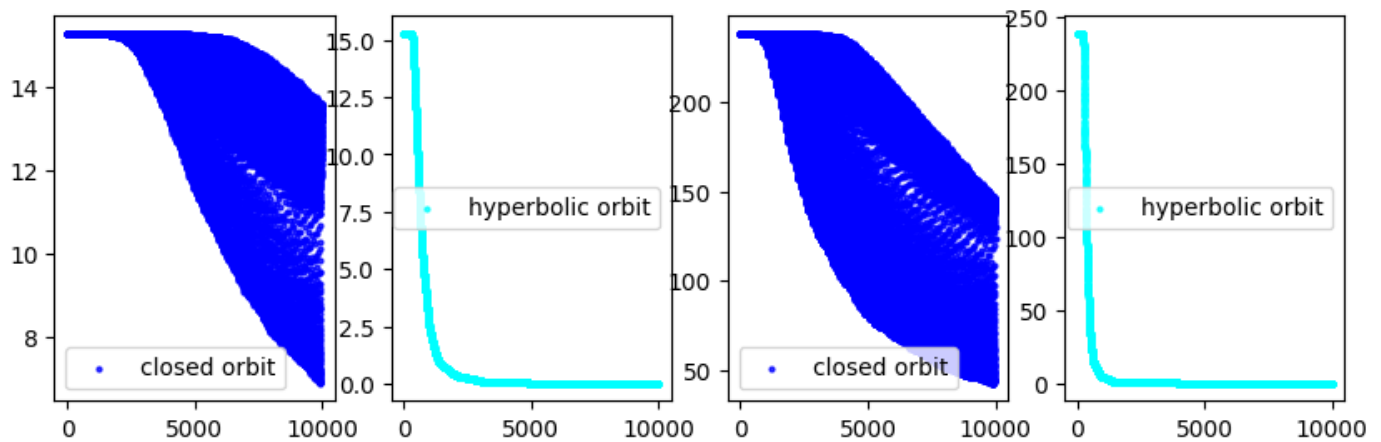
```
1 density25b = []
2 density10b = []
3
4 for i in range(len(orbital_total_b.t)):
5     difference = orbital_total_b.pos[i,:] - orbit_b.pos[i]
6     distance_25b = difference.norm()/(2.5*u.kpc)
7     distance_10b = difference.norm()/(1.0*u.kpc)
8
9     pos_25b = np.where(distance_25b<=1)[0]
10    pos_10b = np.where(distance_10b<=1)[0]
11
12    densitypoints_25b = distance_25b[pos_25b]
13    densitypoints_10b = distance_10b[pos_10b]
14
15    vol25b = (4/3) * np.pi * (2.5**3)
16    densitytotal_25b = len(densitypoints_25b) / vol25b
17    density25b.append(densitytotal_25b)
18
19    vol10b = (4/3) * np.pi * (1.0**3)
20    densitytotal_10b = len(densitypoints_10b) / vol10b
21    density10b.append(densitytotal_10b)
```

```

1 fig, ax = plt.subplots(1,4, figsize=(10,3))
2 ax[0].scatter(orbita_total_a.t, density25, s=4,
3              alpha=0.8, color='blue', label='closed orbit')
4 ax[1].scatter(orbita_total_b.t, density25b, s=4,
5              alpha=0.8, color='cyan', label='hyperbolic orbit' )
6
7 ax[2].scatter(orbita_total_a.t, density10,
8              s=4, alpha=0.8, color='blue', label='closed orbit')
9 ax[3].scatter(orbita_total_b.t, density10b,
10             s=4, alpha=0.8, color='cyan', label='hyperbolic orbit')
11 ax[0].legend()
12 ax[1].legend()
13 ax[2].legend()
14 ax[3].legend()

```

 <matplotlib.legend.Legend at 0x30309d6a0>



5. Repetir el experimento, ahora considerando:

- Un potencial NFW triaxial.
- Un potencial MW-like (compuesto completo)

```

1 total_potential_triaxial = gp.CCompositePotential()
2 total_potential_triaxial['disk'] = gp.MiyamotoNagaiPotential(m = 1E11 , a=3, b:
3 total_potential_triaxial['bulge'] = gp.HernquistPotential(m = 3E9 , c = 0.67,
4 total_potential_triaxial = gp.NFWPotential(m=5E12,r_s=80, a=2, b=3, c=7, units:

```

```

1 n_orbits = 1000
2
3
4 ics_a = a_particle
5 ics_b = b_particle
6
7
8 new_pos_a = np.random.normal(ics_a.pos.xyz.to(u.pc).value, 0.01,
9                               size=(n_orbits,3)).T*u.pc
10 new_vel_a = np.random.normal(ics_a.vel.d_xyz.to(u.km/u.s).value, 1,
11                               size=(n_orbits,3)).T * u.km/u.s
12
13 new_ics_a = gd.PhaseSpacePosition(pos=new_pos_a, vel=new_vel_a)
14 orbit_a = gp.Hamiltonian(total_potential_triaxial).integrate_orbit(ics_a, dt=
15
16 orbits_total_a = gp.Hamiltonian(total_potential_triaxial).integrate_orbit(new_
17
18
19 new_pos_b = np.random.normal(ics_b.pos.xyz.to(u.pc).value, 0.01,
20                               size=(n_orbits,3)).T*u.pc
21 new_vel_b = np.random.normal(ics_b.vel.d_xyz.to(u.km/u.s).value, 1.,
22                               size=(n_orbits,3)).T * u.km/u.s
23
24 new_ics_b = gd.PhaseSpacePosition(pos=new_pos_b, vel=new_vel_b)
25 orbit_b = gp.Hamiltonian(total_potential_triaxial).integrate_orbit(ics_b, dt=
26
27 orbits_total_b = gp.Hamiltonian(total_potential_triaxial).integrate_orbit(new_
28

```

```

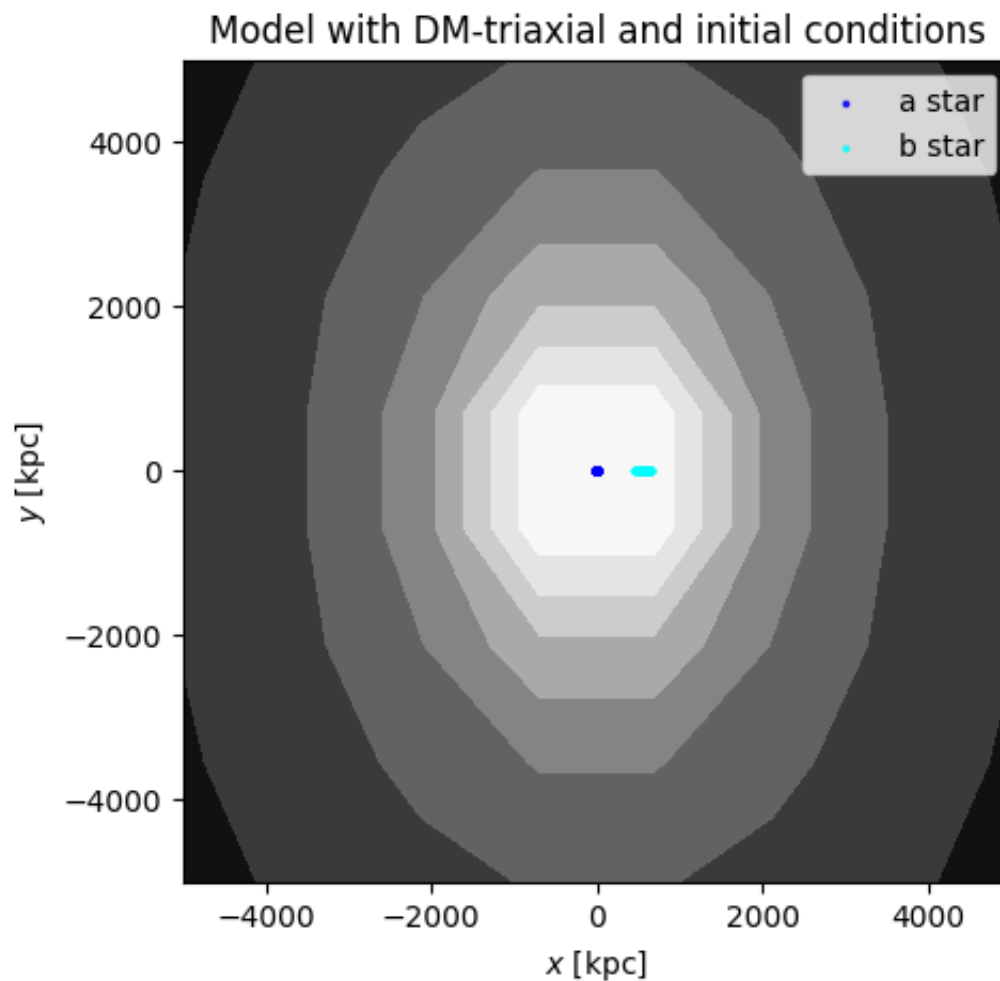
1 grid = np.linspace(-5000,5000,8)
2 fig, ax = plt.subplots(1, 1, figsize=(5,5))
3 fig = total_potential_triaxial.plot_contours(grid=(grid, grid, 0), cmap='Greys
4 fig = orbits_total_a[-1].plot(['x','y'], color='blue', s=10,
5                               alpha=0.8, axes=[ax], auto_aspect=False,
6                               label='a star')
7 fig = orbits_total_b[-1].plot(['x','y'], color='cyan', s=10,
8                               alpha=0.8, axes=[ax], auto_aspect=False,
9                               label='b star')
10 ax.set_xlim(-5000,5000)
11 ax.set_ylim(-5000,5000)
12 ax.legend()
13 ax.set_title('Model with DM-triaxial and initial conditions')

```

```

➡ Text(0.5, 1.0, 'Model with DM-triaxial and initial conditions')

```

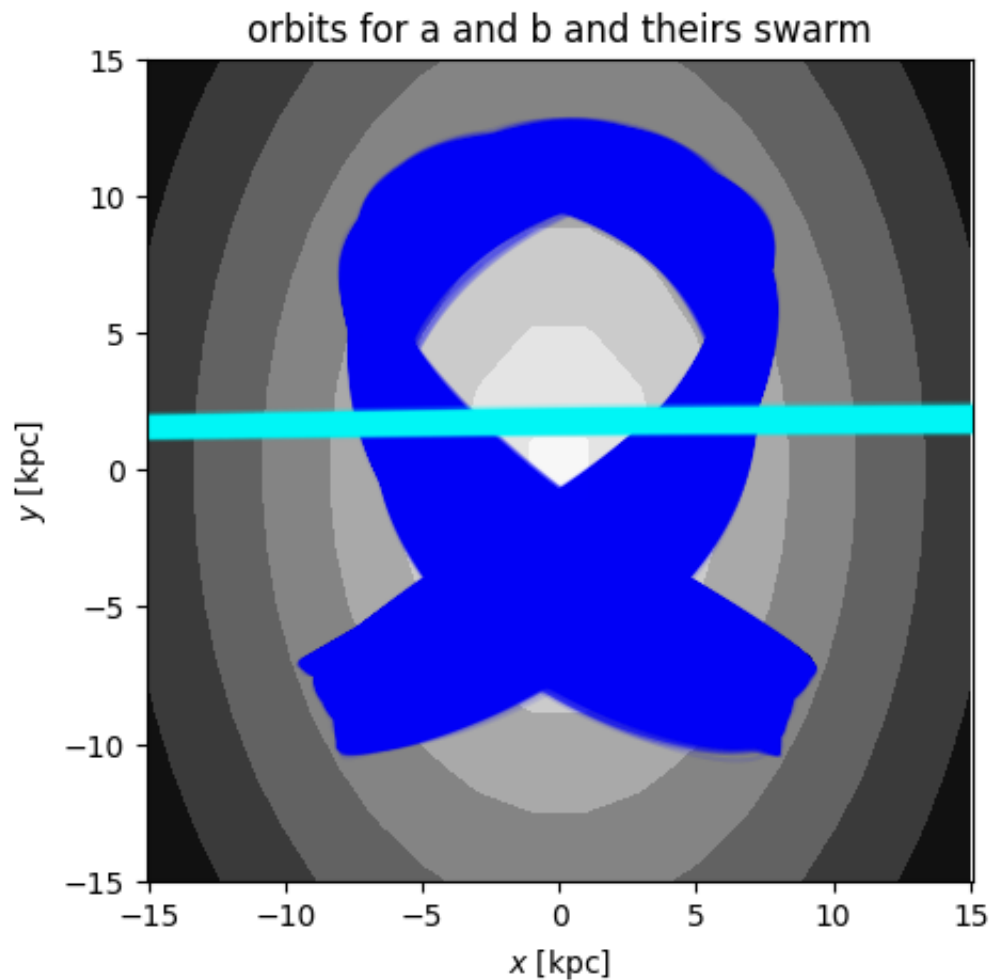


```

1 grid = np.linspace(-15,15,16)
2 fig, ax = plt.subplots(1, 1, figsize=(5,5))
3 fig = total_potential_triaxial.plot_contours(grid=(grid, grid, 0), cmap='Greys
4 fig = orbits_total_a.plot(['x','y'], color='blue', axes=[ax], alpha=0.1, label:
5 fig = orbits_total_b.plot(['x','y'], color='cyan', axes=[ax], alpha=0.1, label:
6 ax.set_title('orbits for a and b and theirs swarm')
7 ax.set_xlim(-15, 15)
8 ax.set_ylim(-15, 15)
9

```

→ (-15.0, 15.0)



3. Graficar y observar evolucion de  $\sigma_r$  y  $\sigma_{\dot{r}}$ .

- sigma total
- sigma por componente ( $[x,y,z]$ ,  $[\dot{x}, \dot{y}, \dot{z}]$ )



Nuestros parametros iniciales son:

- $a_{\text{star}_p} = [6,0,0]$
- $a_{\text{star}_v} = [20,80,0]$
- $b_{\text{star}_p} = [-100,0,0]$
- $b_{\text{star}_v} = [500,10,0]$

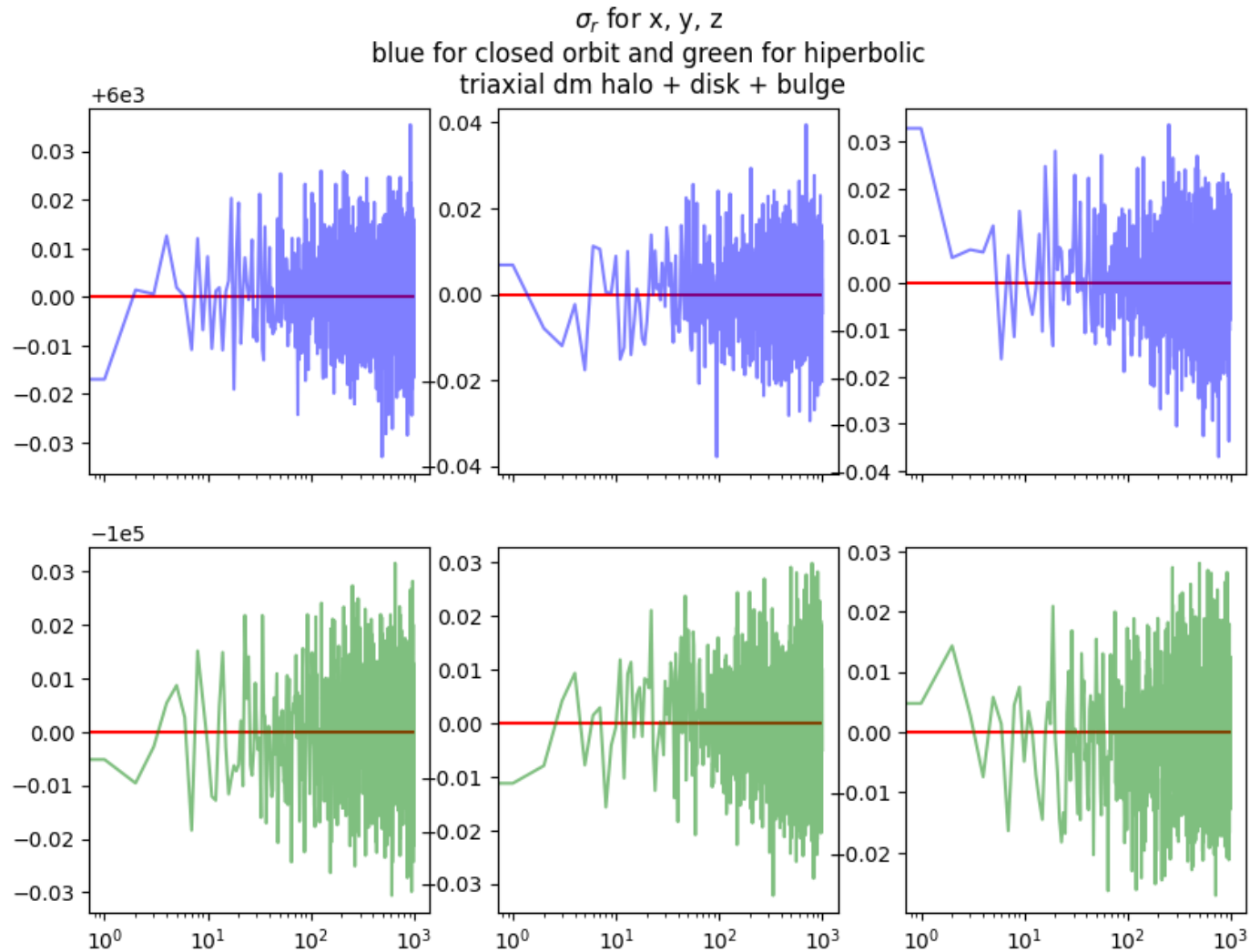
Las cuales estan en kiloparsec. Los  $\sigma$  seran graficados en parsec. Por ende los errores son muy pequenitos.

```
1 fig, ax = plt.subplots(2,3, sharex=True, figsize=(10,7))
2 xdata= np.arange(0,1000)
3
4 #x
5 ax[0,0].hlines(y=6000, xmin=0, xmax=1000, color='red')
6 ax[0,0].plot(xdata, new_pos_a[0], color='blue', alpha=0.5)
7
8 ax[0,0].set_xscale('log')
9 ax[1,0].hlines(y=-100000, xmin=0, xmax=1000, color='red')
10 ax[1,0].plot(xdata, new_pos_b[0], color='green', alpha=0.5)
11
12 #y
13 ax[0,1].hlines(y=0, xmin=0, xmax=1000, color='red')
14 ax[0,1].plot(xdata, new_pos_a[1], color='blue', alpha=0.5)
15
16 ax[1,1].hlines(y=0, xmin=0, xmax=1000, color='red')
17 ax[1,1].plot(xdata, new_pos_b[1], color='green', alpha=0.5)
18
19
20 #z
21 ax[0,2].hlines(y=0, xmin=0, xmax=1000, color='red')
22 ax[0,2].plot(xdata, new_pos_a[2], color='blue', alpha=0.5)
23
24 ax[1,2].hlines(y=0, xmin=0, xmax=1000, color='red')
25 ax[1,2].plot(xdata, new_pos_b[2], color='green', alpha=0.5)
26
27 fig.suptitle('$\sigma_{\{r\}}$ for x, y, z \n blue for closed orbit and green for |
28
29
30
31
```

```

↳ <>:27: SyntaxWarning: invalid escape sequence '\s'
<>:27: SyntaxWarning: invalid escape sequence '\s'
/var/folders/_3/b44zg9zx3mjcpbfyw4_4d15c0000gn/T/ipykernel_2837/4119965136.py:
    fig.suptitle('$\sigma_{r}$ for x, y, z \n blue for closed orbit and green for fc
Text(0.5, 0.98, '$\sigma_{r}$ for x, y, z \n blue for closed orbit and green
for hiperbolic \n triaxial dm halo + disk + bulge')

```



```

1 fig, ax = plt.subplots(2,3, sharex=True, figsize=(10,7))
2 xdata= np.arange(0,1000)
3
4 ax[0,0].set_xscale('log')
5
6 ax[0,0].plot(xdata, new_vel_a[0], color='purple', alpha=0.5)
7 ax[0,0].hlines(y=20, xmin=0, xmax=1000, color='red')

```

```

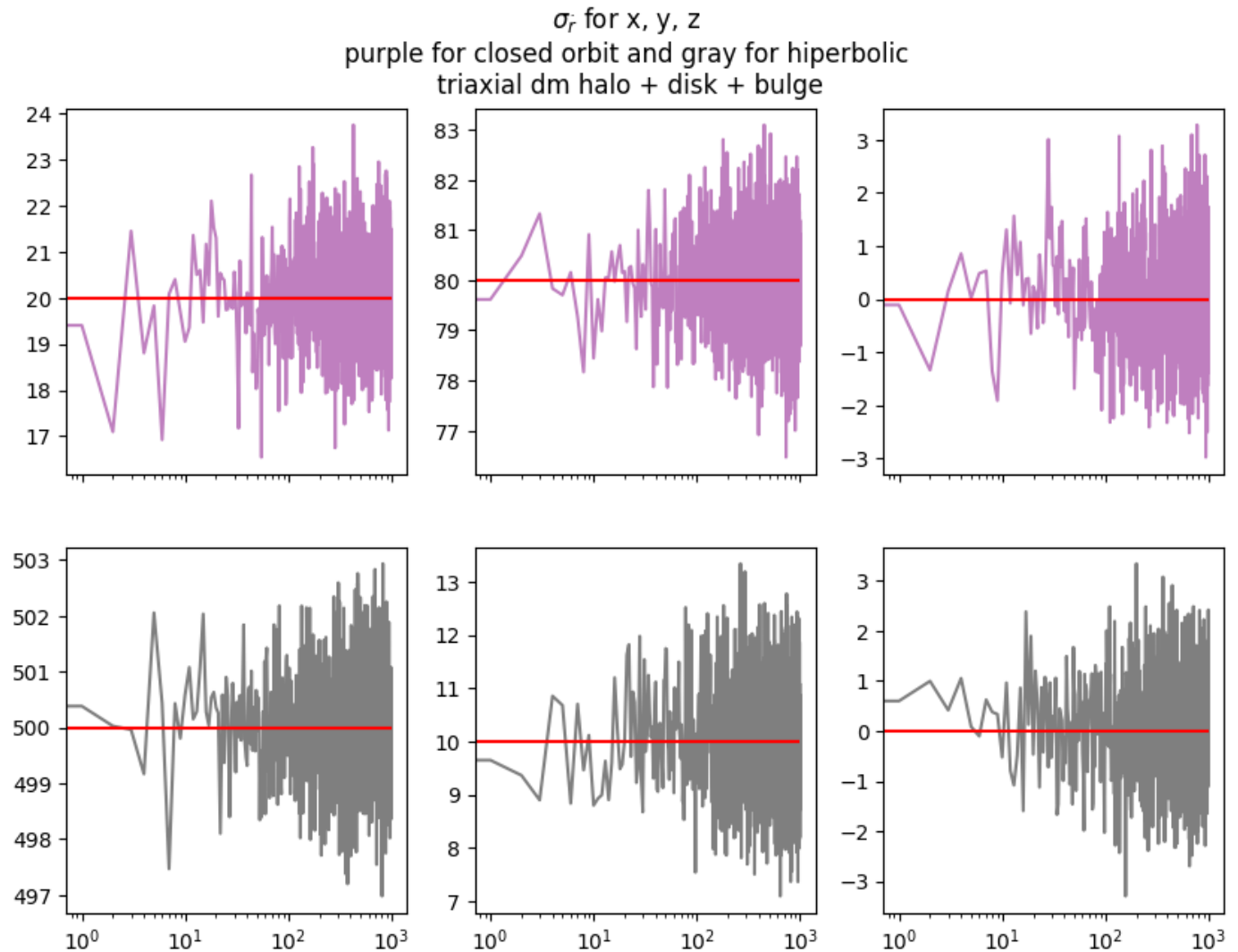
8
9 ax[1,0].plot(xdata, new_vel_b[0], color='black', alpha=0.5)
10 ax[1,0].hlines(y=500,xmin=0, xmax=1000, color='red')
11
12 ax[0,1].plot(xdata, new_vel_a[1], color='purple', alpha=0.5)
13 ax[0,1].hlines(y=80, xmin=0, xmax=1000, color='red')
14
15 ax[1,1].plot(xdata, new_vel_b[1], color='black', alpha=0.5)
16 ax[1,1].hlines(y=10, xmin=0, xmax=1000, color='red')
17
18 ax[0,2].plot(xdata, new_vel_a[2], color='purple', alpha=0.5)
19 ax[0,2].hlines(y=0, xmin=0, xmax=1000, color='red')
20
21 #0
22 ax[1,2].plot(xdata, new_vel_b[2], color='black', alpha=0.5)
23 ax[1,2].hlines(y=0, xmin=0, xmax=1000, color='red')
24
25 fig.suptitle('$\sigma_{\dot{r}}$ for x, y, z \n purple for closed orbit and gr
26
27

```

```

↳ <>:25: SyntaxWarning: invalid escape sequence '\s'
<>:25: SyntaxWarning: invalid escape sequence '\s'
/var/folders/_3/b44zg9zx3mjcpbfyw4_4d15c0000gn/T/ipykernel_2837/2812287091.py:
    fig.suptitle('$\sigma_{\dot{r}}$ for x, y, z \n purple for closed orbit and
Text(0.5, 0.98, '$\sigma_{\dot{r}}$ for x, y, z \n purple for closed orbit
and gray for hiperbolic \n triaxial dm halo + disk + bulge')

```



Si observamos y superponemos los graficos de posicion y velocidad para la misma componente donde ambos parten en cero, se puede notar lo visto en clases del paper de Gomez+2013, es decir, el comportamiento de  $x$  y  $\dot{x}$  como sin y cos (ya que  $\dot{x}$  es la derivada de  $x$ )

4. Tomar esferas de 1 y 2,5 kpc de radio, centradas a todo tiempo t, en la órbita central del enjambre.

En cada paso de integración, calcular la densidad de partículas dentro de estas esferas.

```
1 fig, anim = orbits_total_a[:1000].cylindrical.animate(components=['rho', 'z'],
2                                                         stride=10)
3 anim.save('closedswarmorbit_triaxial.gif')
4 plt.close()
```

➡ MovieWriter ffmpeg unavailable; using Pillow instead.

```
1 fig, anim = orbits_total_b[:1000].cylindrical.animate(components=['rho', 'z'],
2                                                         stride=10)
3 anim.save('hiperbolicswarmorbit_triaxial.gif')
4 plt.close()
```

➡ MovieWriter ffmpeg unavailable; using Pillow instead.

```
1 density25 = []
2 density10 = []
3
4 for i in range(len(orbits_total_a.t)):
5     difference = orbits_total_a.pos[i,:] - orbit_a.pos[i]
6     distance_25 = difference.norm()/(2.5*u.kpc)
7     distance_10 = difference.norm()/(1.0*u.kpc)
8
9     pos_25 = np.where(distance_25<=1)[0]
10    pos_10 = np.where(distance_10<=1)[0]
11
12    densitypoints_25 = distance_25[pos_25]
13    densitypoints_10 = distance_10[pos_10]
14
15    vol25 = (4/3) * np.pi * (2.5**3)
16    densitytotal_25 = len(densitypoints_25) / vol25
17    density25.append(densitytotal_25)
18
19    vol10 = (4/3) * np.pi * (1.0**3)
20    densitytotal_10 = len(densitypoints_10) / vol10
21    density10.append(densitytotal_10)
22
23
24
```

```

1 density25b = []
2 density10b = []
3
4 for i in range(len(orbital_total_b.t)):
5     difference = orbital_total_b.pos[i,:] - orbit_b.pos[i]
6     distance_25b = difference.norm()/(2.5*u.kpc)
7     distance_10b = difference.norm()/(1.0*u.kpc)
8
9     pos_25b = np.where(distance_25b<=1)[0]
10    pos_10b = np.where(distance_10b<=1)[0]
11
12    densitypoints_25b = distance_25b[pos_25b]
13    densitypoints_10b = distance_10b[pos_10b]
14
15    vol25b = (4/3) * np.pi * (2.5**3)
16    densitytotal_25b = len(densitypoints_25b) / vol25b
17    density25b.append(densitytotal_25b)
18
19    vol10b = (4/3) * np.pi * (1.0**3)
20    densitytotal_10b = len(densitypoints_10b) / vol10b
21    density10b.append(densitytotal_10b)

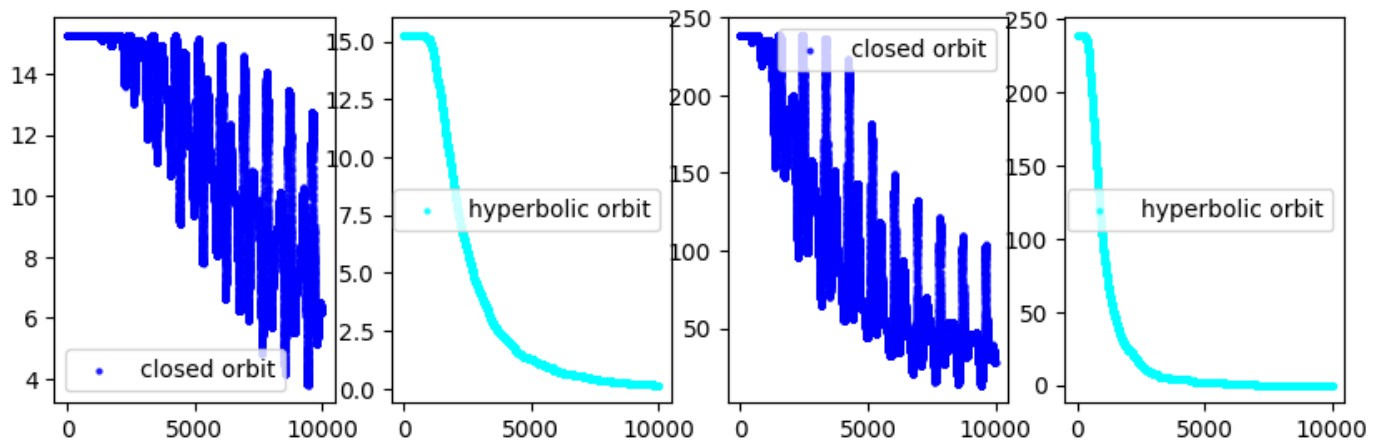
```

```

1 fig, ax = plt.subplots(1,4, figsize=(10,3))
2 ax[0].scatter(orbits_total_a.t, density25, s=4,
3             alpha=0.8, color='blue', label='closed orbit')
4 ax[1].scatter(orbits_total_b.t, density25b, s=4,
5             alpha=0.8, color='cyan', label='hyperbolic orbit' )
6
7 ax[2].scatter(orbits_total_a.t, density10,
8             s=4, alpha=0.8, color='blue', label='closed orbit')
9 ax[3].scatter(orbits_total_b.t, density10b,
10            s=4, alpha=0.8, color='cyan', label='hyperbolic orbit')
11 ax[0].legend()
12 ax[1].legend()
13 ax[2].legend()
14 ax[3].legend()

```

 <matplotlib.legend.Legend at 0x16444fef0>



Finalmente podemos observar que en el caso de que el potencial cambie, las orbitas van a verse afectadas al punto de volverse abiertas como es el caso.

Double-click (or enter) to edit

