MomsData_Demo

November 21, 2018

1 Data analysis demo for H-core-M25 stellar hydro project

Last update: Nov 16, 2018.

This notebook contains a demonstration how to analyse the 3D filtered *moms* data and the 1D radial profile *rprof* data from *PPMstar* 3D hydrodynamic simulations.

1.0.1 Data for this demo

The examples are for the project H-core-M25 (this is the project identifier), the H-core convection simulations of a $25M_{\odot}$ star.

Two runs are used: * M29: 768³ grid * M35: 1536³ grid

M29, M35 are the run identifier. Keep run and project identifier attached to all derived data products.

Both runs have 1000x heating which increases their convective velocities by a factor of 10.

For each run there are two types of data to be read for this demo: * moms data is the spatially filtered data (2-byte data on reduced grid by factor four in each direction) in 3D * rprofs data are spherically averaged radial profiles

1.0.2 Location of data

The data is staged on the UVic Astrophysics Simulation Data Repository (ASDR) mounted in /data/ASDR. The repository contains the project folder H-core-M25.

1.0.3 Python asumptions

The server defaults each notebook to %pylab ipympl

```
In [28]: ## use this for final run to export with images to pdf, markdown or html %pylab inline
```

DEBUG:matplotlib.pyplot:Loaded backend module://ipykernel.pylab.backend_inline version unknown

Populating the interactive namespace from numpy and matplotlib

```
/usr/local/lib/python3.6/dist-packages/IPython/core/magics/pylab.py:160: UserWarning: pylab imgatplotlib` prevents importing * from pylab and numpy
```

[&]quot;\n`%matplotlib` prevents importing * from pylab and numpy"

```
In [29]: import numpy as np
         import matplotlib as mpl
         import matplotlib.pyplot as plt
         import matplotlib.colors as color
         import nugridpy.utils as utils
         import sys, os, time
         # if you make changes to the ppmpy module (e.g. add your analysis methods via a pull
         # request) in the https://github.com/PPMstar/PyPPM repo you may want use that
         # updated version
         #sys.path.insert(0,'/user/david/PyPPM/')
         sys.path.insert(0,'/user/PyPPM/')
         from ppmpy import ppm
         cb = utils.linestylecb # colours
In [30]: %%bash
        ls /data/ASDR/H-core-M25/
M29-768
M35-1536
In [31]: dir_repo = '/data/ASDR'
        dir_project = 'H-core-M25'
        rprof = {}; moms = {}
                                     # initialize dictionaries to hold rprof and moms instan
                    = ['M29-768','M35-1536'] # select runs
        moms_dumps = {}
        moms_dumps[runs[0]] = 650
                                       # select dump numbers for moms
        moms_dumps[runs[1]] = 375
                     = ['M29-768'] # select runs
         #moms_dumps = [ 650 ] # select dump numbers for moms
         # rprof instance holds radial profiles for all dumps
         # moms instance holds only one dump at a time
         for run in runs:
            path = os.path.join(dir_repo,dir_project,run)
             # radial profile:
             rprof[run] = ppm.RprofSet(os.path.join(path,'rprofs'))
             moms[run] = ppm.MomsDataSet(os.path.join(path,'myavsbq'),moms_dumps[run])
        print("moms and rprof dictionary created")
748 rprof files found in '/data/ASDR/H-core-M25/M29-768/rprofs/.
Dump numbers range from 0 to 747.
Reading history file '/data/ASDR/H-core-M25/M29-768/rprofs/HcoreE00768-0000.hstry'.
748 .aaa files found in '/data/ASDR/H-core-M25/M29-768/myavsbq/.
Dump numbers range from 0 to 747.
```

```
The PPMstar grid is being constructed, this can take a moment
51 rprof files found in '/data/ASDR/H-core-M25/M35-1536/rprofs/.

Dump numbers range from 375 to 425.

Reading history file '/data/ASDR/H-core-M25/M35-1536/rprofs/HcoreE01536-0000.hstry'.

51 .aaa files found in '/data/ASDR/H-core-M25/M35-1536/myavsbq/.

Dump numbers range from 375 to 425.

The PPMstar grid is being constructed, this can take a moment moms and rprof dictionary created

In [32]: # get info about moms instance
# help(moms['M29-768'])
```

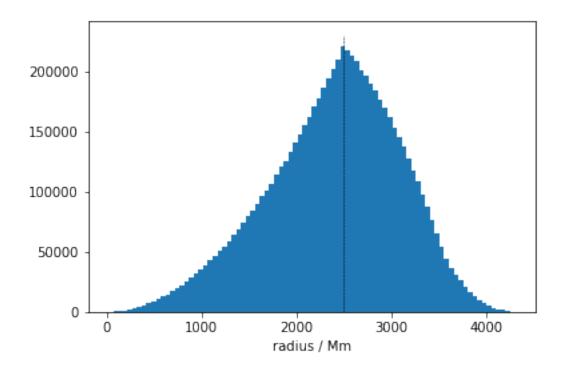
1.1 Basic grid properties

```
In [33]: x,y,z,r=moms['M29-768'].get_grid()
In [34]: print(192**3,len(r))
7077888 7077888
In [35]: print("Distance center of grid to max x value of domain: %6.4f Mm" % moms['M35-1536']
```

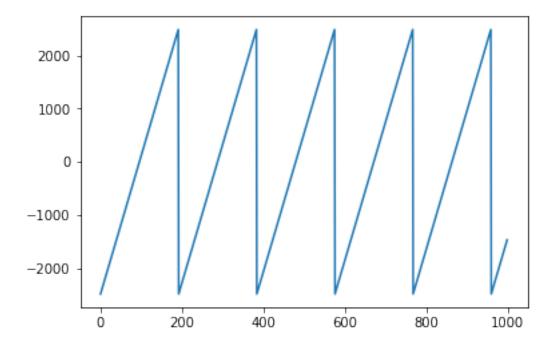
1.1.1 Histogram of radii

• increasing to 1/2 length of grid, then decreasing as only fraction of shell in box

Distance center of grid to max x value of domain: 2493.4895 Mm



1.1.2 Some more experiments with coordinates



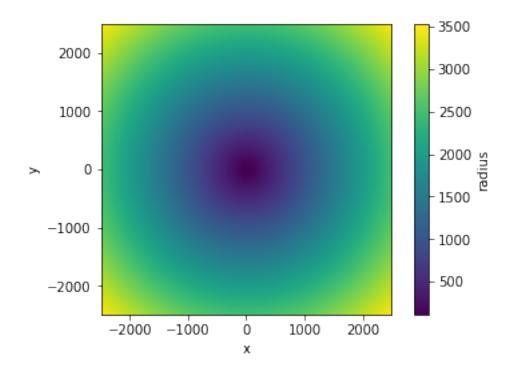
1.1.3 Planar slice image

Radius

DEBUG:matplotlib.colorbar:locator: <matplotlib.colorbar._ColorbarAutoLocator object at 0x7f3779
DEBUG:matplotlib.colorbar:Using auto colorbar locator on colorbar
DEBUG:matplotlib.colorbar:locator: <matplotlib.colorbar._ColorbarAutoLocator object at 0x7f3779
DEBUG:matplotlib.colorbar:Setting pcolormesh

```
Out[40]: Text(0, 0.5, 'radius')
```

```
DEBUG:matplotlib.axes._base:update_title_pos
```



1.1.4 Properties of the grid and time resolution

In [41]: # spatial resolution for 768 momsdata
 print('The spatial resolution of 768 momsdata is {:0.3f}'.format(np.diff(moms['M29-76]
 print('While PPMStar 768 has a spatial resolution of {:0.3f}'.format(np.diff(moms['M29-76])

```
print('')
         # spatial resolution for 1536 momsdata
        print('The spatial resolution of 1536 momsdata is {:0.3f}'.format(np.diff(moms['M35-1
        print('While PPMStar 1536 has a spatial resolution of {:0.3f}'.format(np.diff(moms['M
        print('')
         # what is the extent of the simulation?
         print('The extent of the simulation is then {:0.0f}'.format(np.diff(moms['M35-1536'].
The spatial resolution of 768 momsdata is 26.042 Mm
While PPMStar 768 has a spatial resolution of 6.510 Mm
The spatial resolution of 1536 momsdata is 13.021 Mm
While PPMStar 1536 has a spatial resolution of 3.255 Mm
The extent of the simulation is then 2500 Mm
```

```
In [42]: # for 768 momsdata
         print('The temporal resolution of 768 and 1536 momsdata is the same as the PPMStar ou
               format(np.mean(np.diff(rprof['M29-768'].get_history().get('time(mins)')))), 'min'
         print('')
```

The temporal resolution of 768 and 1536 momsdata is the same as the PPMStar output which average

print('The run-time temporal resolution of the PPMStar output averages around {:0.2f}

format(np.mean(rprof['M35-1536'].get_history().get('dt(secs)'))), 'seconds per c

The run-time temporal resolution of the PPMStar output averages around 2.73 seconds per cycle

1.2 Find Times For Dumps

As hinted at in the above section, there is a history file that gives us information about the run. This is located in the rprof files themselves

```
In [43]: # get the simulation time in seconds for dump 100 in the 768 and 1536 runs
         print('{:0.1f} seconds '.format(rprof['M29-768'].get_history().get('time(secs)')[moms]
         'have passed since the simulation started for the 768 run')
         print('{:0.1f} seconds '.format(rprof['M35-1536'].get_history().get('time(secs)')[moments
               'have passed since the simulation started for the 1536 run')
6507520.0 seconds have passed since the simulation started for the 768 run
```

7

3756320.0 seconds have passed since the simulation started for the 1536 run

1.3 What quantities have what index?

The following quantities written into the moms data file which can be called with an index:

| index | quantity |
|-------|---------------------|
| 0 | х |
| 1 | $\vec{u_x}$ |
| 2 | $ec{u_y} \ ec{u_z}$ |
| 3 | $\vec{u_z}$ |
| 4 | $ \vec{u_t} $ |
| 5 | $ \vec{u_r} $ |
| 6 | $ \vec{\omega} $ |
| 7 | P |
| 8 | rho |
| 9 | fv |
| | |

- Note that these are just 10 out of 32 quantities that can be made available in the moms data.
- fv is the fractional volume of the material initially only outside the convection zone.

```
Some Helpful Definitions \mu = \text{fv} \times 0.617 + (1 - \text{fv}) \times 0.669
```

$$T = \frac{P\mu}{\rho R_{gas}}$$

$$R_{gas} = 8.314462$$

$$\vec{\omega} = \vec{\nabla} \times \vec{u}$$

1.4 Radial profiles

Radial profiles can be taken from the *rprof* data sets. They can also be constructed from the *moms* data. This is demonstrated below.

```
In [44]: # define variables for dump number, rprof and moms

runid = 'M29-768'  # select run id for the rest of the notebook
runid = 'M35-1536'
thisdump = moms_dumps[runid]
thisrprof = rprof[runid]
thismoms = moms[runid]

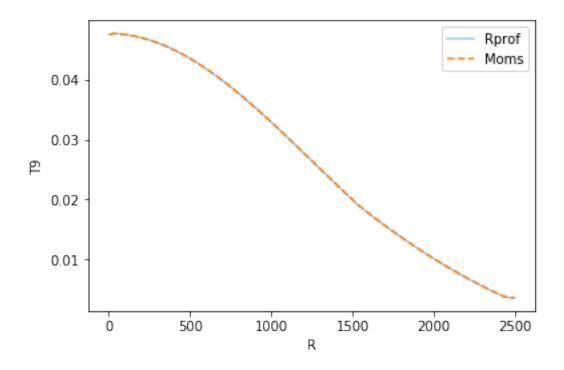
# get T9 and Ut
P_rprof = thisrprof.get('P0',fname=thisdump,resolution='h')[0::2] + thisrprof.get('P1 rho_rprof = thisrprof.get('Rho0',fname=thisdump,resolution='h')[0::2] + thisrprof.get
FV_rprof = thisrprof.get('FV',fname=thisdump,resolution='h')[0::2]

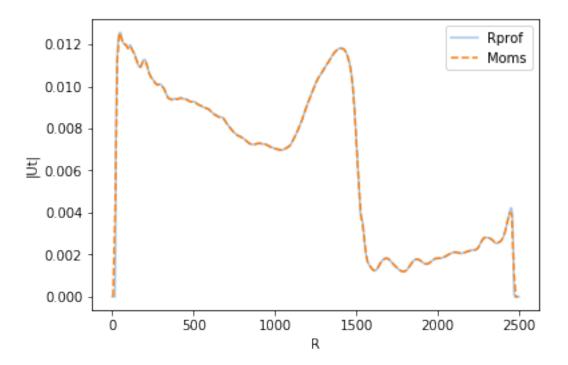
# T9 in rprof class is not correct, calculate directly
T9_rprof = P_rprof * (0.617*FV_rprof + 0.669*(1-FV_rprof)) / (8.314462 * rho_rprof)

R_rprof = thisrprof.get('R',fname=thisdump,resolution='l')
```

Ut_rprof = thisrprof.get('|Ut|',fname=thisdump)

```
In [45]: # make an rprof of temperature and ut
         ut_avg, radial_axis = thismoms.get_rprof(4,thisdump)
         # first we need to construct T from quantities
         mu = 0.617 * thismoms.get(9,fname=thisdump) + (1 - thismoms.get(9,fname=thisdump))*0.
         P = thismoms.get(7,fname=thisdump)
         rho = thismoms.get(8,fname=thisdump)
         Rgas = 8.314462
         # put it all together
         T = (mu * P) / (Rgas * rho)
         # we can give the rprof method an array to be spherically averaged
         T_avg, radial_axis = thismoms.get_rprof(T,thisdump)
/usr/local/lib/python3.6/dist-packages/scipy/stats/_binned_statistic.py:607: FutureWarning: Us
 result = result[core]
In [46]: # plot
         ifig += 1; plt.close(ifig); plt.figure(ifig)
         plt.plot(R_rprof,T9_rprof,label='Rprof',ls=cb(0)[0],color=cb(0)[2])
         plt.plot(radial_axis,T_avg,label='Moms',ls=cb(1)[0],color=cb(1)[2])
         plt.xlabel('R')
        plt.ylabel('T9')
        plt.legend()
Out[46]: <matplotlib.legend.Legend at 0x7f37795d6550>
DEBUG:matplotlib.axes._base:update_title_pos
DEBUG:matplotlib.axes._base:update_title_pos
DEBUG:matplotlib.axes._base:update_title_pos
DEBUG:matplotlib.axes._base:update_title_pos
DEBUG:matplotlib.axes._base:update_title_pos
```





1.4.1 Planar Slice Image

```
In [48]: x,y,z,r = moms[runid].get_grid()
    # they are flattened arrays, rearrange
    resolution = moms[runid].momsdata.resolution
    r_matrix = np.reshape(r,(resolution,resolution)))
    # extent x,y
    extent=[min(x),max(x),min(y),max(y)]

# slice number
    slice_num = int(resolution/2)

T9
In [49]: T_matrix = np.reshape(T,(resolution,resolution,resolution))
In [50]: ifig += 1; plt.close(ifig); plt.figure(ifig)
    plt.imshow(T_matrix[:][:][slice_num],extent=extent)
    plt.ylabel('y')
    plt.xlabel('x')
    cbar = plt.colorbar()
```

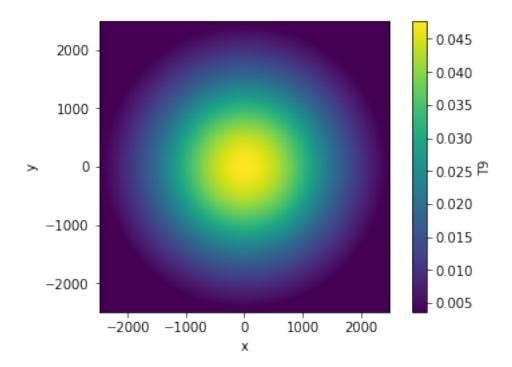
label colorbar cbar.ax.set_ylabel('T9')

DEBUG:matplotlib.colorbar:locator: <matplotlib.colorbar._ColorbarAutoLocator object at 0x7f377 DEBUG:matplotlib.colorbar:Using auto colorbar locator on colorbar DEBUG:matplotlib.colorbar:locator: <matplotlib.colorbar._ColorbarAutoLocator object at 0x7f377

DEBUG:matplotlib.colorbar:Setting pcolormesh

Out[50]: Text(0, 0.5, 'T9')

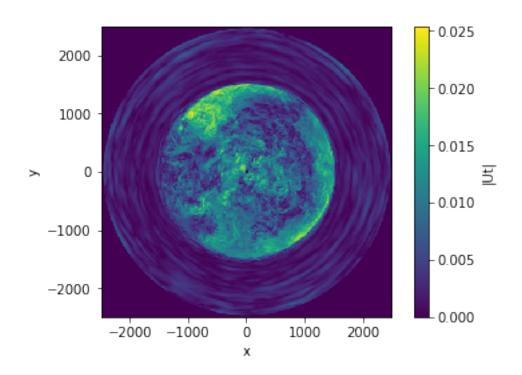
```
DEBUG:matplotlib.axes._base:update_title_pos
```



| Ut |

Out[52]: Text(0, 0.5, '|Ut|')

```
DEBUG:matplotlib.axes._base:update_title_pos
```



1.5 FV Colourmap of a Plane (x=y=0)

```
In [53]: \# x=y=0, a particular z slice. convert to an 8-bit number
         fv = np.reshape(thismoms.get(9,thisdump),(resolution,resolution,resolution))
         fv_bit = 251 + 13.35455532 * np.log(fv[:][:][96])
In [54]: FV_cmap_str = '''
         Anot: 0 0.0
         Anot: 18 0.1058824
         Anot: 56 0.2745098
         Anot: 75 0.7843137
         Anot: 123 1.0
         Anot: 158 1.0
         Anot: 184 0.5490196
         Anot: 203 0.454902
         Anot: 255 0.1254902
         Cnot: 0 0.0 0.0 0.0
         Cnot: 48 0.0 0.0 0.2509804
         Cnot: 56 0.0 0.2352941 0.627451
         Cnot: 65 0.0 0.7843137 1.0
         Cnot: 75 1.0 1.0 1.0
         Cnot: 100 1.0 1.0 0.0
         Cnot: 186 1.0 0.0 0.0
         Cnot: 244 0.5019608 0.0 0.0
         Cnot: 255 0.5019608 0.0 0.0
         1.1.1
         cmap = ppm.colourmap_from_str(FV_cmap_str, segment=(5, 251))
         # normalize to our 255 bit range
         norm = mpl.colors.Normalize(vmin=5, vmax=251)
Square Image
In [55]: my_dpi = 300
         ifig+=1; plt.close(ifig); plt.figure(ifig,figsize=(536/my_dpi, 536/my_dpi), dpi=my_dp
         x,y,z,r = thismoms.get_grid()
         plt.pcolor(np.unique(x),np.unique(y),fv_bit,cmap=cmap,norm=norm)
         plt.axis('off')
Out [55]: (-2493.489501953125, 2493.489501953125, -2493.489501953125, 2493.489501953125)
DEBUG:matplotlib.axes._base:update_title_pos
DEBUG:matplotlib.axes._base:update_title_pos
DEBUG:matplotlib.axes._base:update_title_pos
DEBUG:matplotlib.axes._base:update_title_pos
```

DEBUG:matplotlib.axes._base:update_title_pos



- In []:
- In []:
- In []: