TSC_CM_different_prior

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This program calculates the CM frame relic position contraints

```
In [469]: from __future__ import division
          %autosave 60
         %load_ext nbtoc
         %nbtoc
         from astropy import wcs
         import astropy.coordinates as coord
         import astropy.units as unit
         from astropy.coordinates import Angle
         import astropy.io.fits
         from astropy.cosmology import FlatLambdaCDM
         from astropy.coordinates import ICRS
         import numpy as np
         import time
         import numpy.random as rand
         import pdb
         import pandas as pd
         import sys
tstart = time.time()
         #sys.path.append('/afs/sapphire.physics.ucdavis.edu/home/karenyng/Documents/Research_c
```

Autosaving every 60 seconds
The nbtoc extension is already loaded. To reload it, use:
%reload_ext nbtoc

```
In [470]: from wcs_ICRS import wcs_ICRS as WICRS
```

load homebrew modules

```
In [471]: from plotmod_dict import *
   import cosmo
```

Part I

Initialization !!!!!!

```
In [472]: #---initializations----
         data_path = "/Users/karenyng/Documents/Research/code"+\
                      "/TSM/mercury_elGo/Feb_data/"
         prefix = data_path + "ElGordo_"
         oprefix = data_path + "polar_"
          index = [ str(i) for i in range(20) ]
         NW_relic = ICRS('01h02m46s - 49d14m43s',
                          unit=(unit.degree, unit.degree))
         SE\_relic = ICRS('01h03m01s - 49d17m14s',
                          unit=(unit.degree, unit.degree))
          #Histogram bins
         N_bins_2d = 130
         N_bins_1d = 200
         N_bins_TSM = 45
         N_bins_alpha = 90
          #specify the number of iterations
         Iter = 500000
          #Iter = 475000
         'd_proj',
              'v_rad_obs',
               'alpha',
              'v_3d_obs',
'd_3d',
               'v_3d_col',
              'd_max',
'TSM_0',
               'TSM_1',
               'T',
               'prob']
          fitspath = '/Users/karenyng/Documents/Research/code/'+\
                      'ElGordo-Dynamics-Paper/Analysis/HSTlensing/'
          fits = fitspath + 'header.fits'
         getwcs = WICRS('01h03m01s -49d17m14s', fitsname=fits)
         w = getwcs.wcs
```

load pickles into suitable formats

1 Calculate CM coordinates in pixel then convert back to WCS coordinates

```
In [492]: # these values r from El Gordo WL paper latest arxiv version 2
           NW_pix = np.array
           NW = ICRS('01h02m50.60s -49d15m04.5s')
           SE = ICRS('01h02m56.31s - 49d16m23.2s')
           [NW_pix, SE_pix] = w.wcs_world2pix(np.array([[NW.ra.deg, NW.dec.deg],
                                                            [SE.ra.deg, SE.dec.deg]]), 1)
In [494]: ### the following values are from James via private communication
           ### and are what we use for our MCMAC
           \#NW\_pix = np.array([2966.6336, 5431.9614])
           \#SE\_pix = np.array([2954.5809, 3152.1517])
           #[NW_wcs, SE_wcs] = w.wcs_pix2world(np.array([NW_pix, SE_pix]), 1)
In [495]: \#NW = ICRS('01h02m51.68s -49d15m04.40s')
           \#SE = ICRS('01h02m38.38s - 49d16m37.64s')
           #[NW_pix, SE_pix] = w.wcs_world2pix(np.array([[NW.ra.deg, NW.dec.deg],
                                                           [SE.ra.deg, SE.dec.deg]]), 1)
 Just use the mean of the mass blobs as the center of mass location
In [524]: cm_pix_x = (NW_pix[0] * NW_mean_mass + SE_pix[0] * SE_mean_mass) / 
           (NW mean mass + SE mean mass)
           cm_pix_x
Out [524]:
          3133.7272010974093
In [526]: cm_pix_y = (NW_pix[1] * NW_mean_mass + SE_pix[1] * SE_mean_mass) / <math>\setminus
           (NW_mean_mass + SE_mean_mass)
           cm_pix_y
Out [526]:
          4725.4838968134009
In [529]: NW_pix
Out [529]:
          array([ 3142.04012987, 5551.36265773])
In [512]: NW_mean_mass = data['m_1'].mean()
SE_mean_mass = data['m_2'].mean()
           cm_pix = (NW_pix * NW_mean_mass + SE_pix * SE_mean_mass) /\
                   (NW_mean_mass + SE_mean_mass)
           [cm_wcs, junk] = w.wcs_pix2world(np.array([cm_pix, cm_pix]), 1)
cm_wcs = ICRS(ra=cm_wcs[0], dec=cm_wcs[1],
                         unit=(unit.degree, unit.degree))
           SE_relic_CM_sep = cm_wcs.separation(SE_relic)
           NW_relic_CM_sep = cm_wcs.separation(NW_relic)
           print cm wcs
```

```
<ICRS RA=15.72101 deg, Dec=-49.26060 deg>
```

2 Compare all the relevant coordinates side by side

Something went wrong with the coordinates, let's try to visualize

```
In [664]: dec_axis = np.arange(-49.30, -49.19, 0.02)
          ra_axis = np.arange(15.68, 15.76, 0.01)
          print dec_axis
          print ra_axis
          def ra_deg_to_wcs(deg):
              deg = deg / 15.
              [h, m] = str(deg).split(".")
              [m, s] = str(float("." + m) * 60).split(".")
              s = '{0:.1f}'.format(float("." + s) * 60.)

return h + 'h' + m + 'm' + s + 's'
          def dec_deg_to_wcs(deg):
              [d, m] = str(deg).split(".")
              [m, s] = str(float("." + m) * 60).split(".")
              s = '\{0:.1f\}'.format(float("." + s) * 60.)
              return d + 'h' + m + 'm' + s + 's'
          raAXIS = [ra_deg_to_wcs(ra) for ra in ra_axis]
          decAXIS = [dec_deg_to_wcs(dec) for dec in dec_axis]
          print raAXIS
          print decAXIS
         [-49.3 -49.28 -49.26 -49.24 -49.22 -49.2]
```

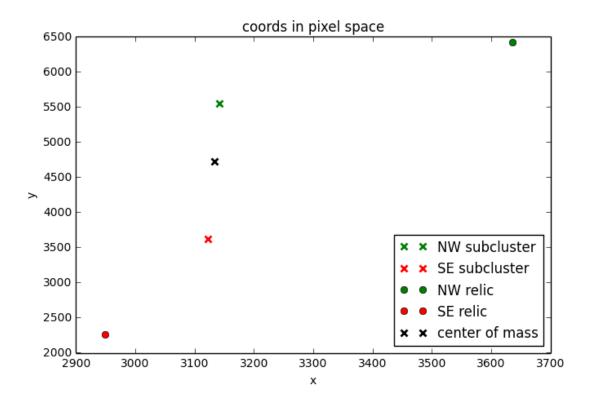
```
[-49.3 -49.28 -49.26 -49.24 -49.22 -49.2]
[ 15.68 15.69 15.7 15.71 15.72 15.73 15.74 15.75 15.76]
['1h2m43.2s', '1h2m45.6s', '1h2m48.0s', '1h2m50.4s', '1h2m52.8s',
'1h2m55.2s', '1h2m57.6s', '1h3m0.0s', '1h3m2.4s']
['-49h18m0.0s', '-49h16m48.0s', '-49h15m36.0s', '-49h14m24.0s',
'-49h13m12.0s', '-49h12m0.0s']
```

```
#plt.ylim(-49.3, -49.2)
#plt.xlim(15.69, 15.76)

plt.xlabel('x')
plt.ylabel('y')
plt.title('coords in pixel space')
```

Out [665]:

<matplotlib.text.Text at 0x115210850>



```
In [656]: print 'NW_pix = {0} '.format(NW_pix)
            print 'SE_pix = {0} '.format(SE_pix)
            print 'NW_relic_pix = {0}'.format(NW_relic_pix)
print 'SE_relic_pix = {0}'.format(SE_relic_pix)
print 'cm_pix = {0}'.format(cm_pix)
            NW_pix = [3142.04012987 5551.36265773]
            SE_pix = [3122.60864656 3620.86983129]
            NW_relic_pix = [ 3636.50083636 6418.49706285]
            SE_relic_pix = [2949.2945903]
                                                          2262.66993921]
            cm_pix = [3133.7272011]
                                                 4725.48389681]
In [666]: plt.plot(NW_wcs[0], NW_wcs[1], "gx", label="NW subcluster",
                         markeredgewidth=2)
            plt.plot(SE_wcs[0], SE_wcs[1], "rx", label="SE subcluster",
                         markeredgewidth=2)
            plt.plot(NW_relic.ra.deg, NW_relic.dec.deg, "go", label="NW relic")
plt.plot(SE_relic.ra.deg, SE_relic.dec.deg, "ro", label="SE relic")
plt.plot(cm_wcs.ra.deg, cm_wcs.dec.deg, "kx", label="center of mass",
                         markeredgewidth=2)
```

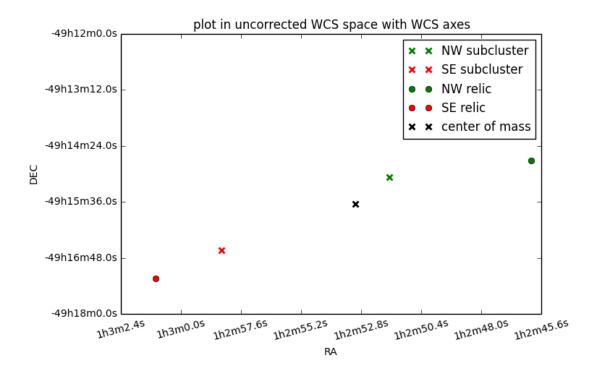
```
a = plt.gca()
a.get_xaxis().get_major_formatter().set_useOffset(False)
a.get_yaxis().get_major_formatter().set_useOffset(False)

plt.ylim(-49.3, -49.2)
xlim_min, xlim_max = plt.xlim()
plt.xlim(xlim_max, xlim_min)
plt.ylim(-49.3, -49.2)
plt.xlim(15.76, 15.69)
plt.title('plot in uncorrected WCS space with WCS axes')

a = plt.gca()
a.set_xticklabels(raAXIS, rotation = 15)
a.set_yticklabels(decAXIS)
plt.xlabel("RA")
plt.ylabel("DEC")
plt.legend(loc = 'best')
```

Out [666]:

<matplotlib.legend.Legend at 0x1152068d0>

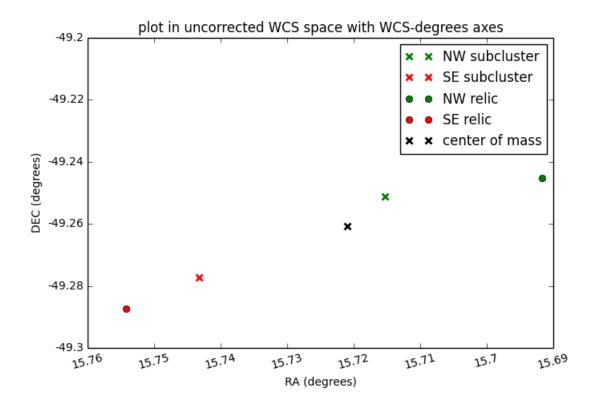


```
plt.ylim(-49.3, -49.2)
plt.xlim(15.76, 15.69)
plt.title('plot in uncorrected WCS space with WCS-degrees axes')
plt.xlabel("RA (degrees)")
plt.ylabel("DEC (degrees)")
plt.legend(loc = 'best')

a = plt.gca()
a.set_xticklabels(ra_axis, rotation = 15)
a.set_yticklabels(dec_axis)
```

Out [667]:

```
[<matplotlib.text.Text at 0x1162b0050>, <matplotlib.text.Text at 0x1162aae10>, <matplotlib.text.Text at 0x11794a690>, <matplotlib.text.Text at 0x11794ab90>, <matplotlib.text.Text at 0x117947310>, <matplotlib.text.Text at 0x117947a50>]
```



```
In []: NW = ICRS(NW_wcs[0], NW_wcs[1], unit = (unit.deg, unit.deg))
SE = ICRS(SE_wcs[0], SE_wcs[1], unit = (unit.deg, unit.deg))
```

use ΛCDM

```
print "NW relic separation from CM is {0}".format(r_proj_NW)
```

Now repeat the calculation for the separation of the NW subcluster to the center of mass

Which is consistent with the picture given by Lindner et al. 's paper Fig 1.

Part II

How to do the relic calculation

3 Info from Lindner et al. 2013

"shock speed can be interpreted as an upper limit on the collision speed"

implies that this a relative speed between the two subclusters?

This is different than saying that the shock speed is in the CM frame.

3.1 the position of the center of mass and thus the separations of the relics to the CM

- using the full realizations takes up a lot of time for doing conversion between coordinates and calculating the
 physical separation since we have half a million realizations and the code is not vectorized python loops are
 slow
- using the mean location of the center of mass which is a lot faster but will not fold the uncertainties in maybe I should rewrite this part using Cython

3.2 the velocities of the relics

we do not know the time evolution of the speed but we can assume an average velocity to be either:

- the collisional velocity of the NW subcluster
- the collision velocity of the NW subcluster times a slow-down factor

In the center of mass frame, the speed of m_1 and m_2 are related by:

$$v_{2,CM} = -\frac{m_1}{m_2} v_{1,CM}$$

-(1)

but in our simulation, we only know about the relative velocities of the two subclusters,

 $v_{rel} = v_1 - v_2$ – (2) which is true in any frame

Plug (1) in (2) to get

$$v_{1,CM} = \left(\frac{m_2}{m_1 + m_2}\right) v_{rel}$$
$$v_{2,CM} = -\left(\frac{m_1}{m_1 + m_2}\right) v_{rel}$$

3.3 Relic calculation - v.1 - just getting a feeling of how the distances compare

- using a mean location for the center of mass
- using NW $v_{relic} = 4300$ km/s ± 800 km/s assuming this is in CM frame

Is this shock speed in 3D or 2D??? it makes a difference - should be in 3D...Calculation of the simulated separation for relic from the CM:

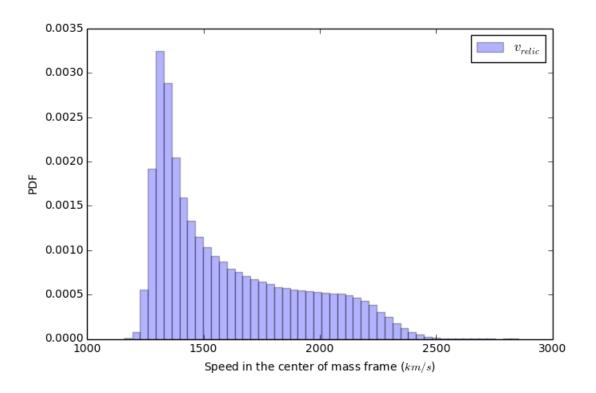
$$s_{relic} = v_{relic_i} t_{tsc_i} \cos(\alpha)$$

There will be 2 outputs from the calculations just from simulation since there are two TSCs. Unit conversion is done to convert units of (km/s * Gyr) to Mpc

```
Mpc / km = 60 \times 60 \times 365 \times 24 \times 1e9/(1e13 \times 1e6 \times 3.086)
```

3.4 Initialize the velocities

3.5 See if the speed distribution looks alright



3.6 Initialize the lower bound of the predicted location

$$x_{NW} = \frac{m_{SE}}{m_{NW}} x_{SE}$$

$$x_{rel} = x_{NW} - x_{SE}$$

$$x_{NW} = \frac{m_{SE}}{m_{NW} + m_{SE}} x_{rel}$$

3.7 Initialize the upper bound of the predicted location

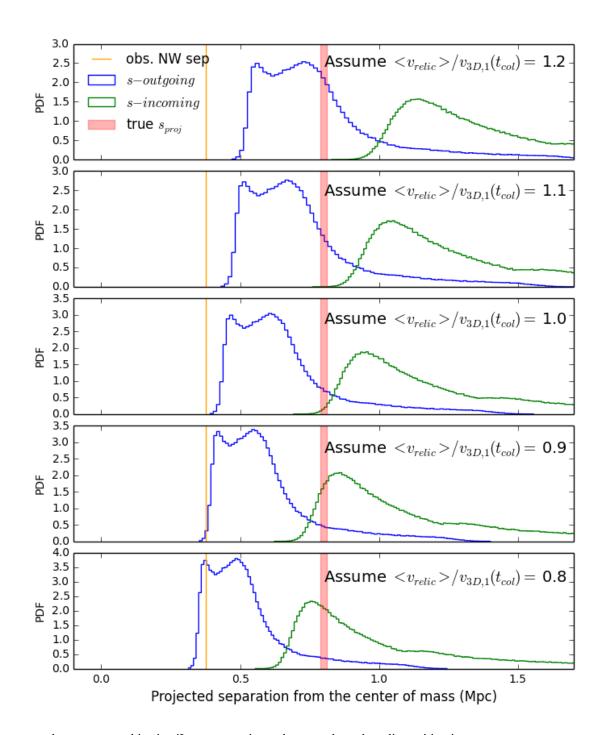
```
data = data[~np.isnan(data['r_proj0_max'])]
data = data[~np.isnan(data['r_proj1_max'])]
```

3.8 Include the uncertainty of the width of the relic

```
In [535]: rupper = r_proj_NW.value - 23 / 2 / 1000
        rlower = r_proj_NW.value + 23 / 2 / 1000
        print "range is {0}, {1}".format(rupper, rlower)
        range is 0.788629498264, 0.811629498264
In [617]: def make_hist(this_ax, data1, data2, data1_mean, data2_mean, beta, i):
            j1, j2, j3 = this_ax.hist(data1,
                            histtype='step', bins=100, label=r'$s-outgoing$',
                            normed=True)
            j1, j2, j3 = this_ax.hist(data2),
                            histtype='step', label=r'$s-incoming$',
                            bins = 1000, normed = True)
            this_ax.axvspan(rlower, rupper, color='red', label=r'true $s_{proj}$', alpha=0.3)
            this_ax.axvline(NW_subclstr_sep.value, color='orange', label=r'obs. NW sep')
            this_ax.set_xlim(-0.1, 1.7)
            this_ax.set_ylabel('PDF')
```

3.9 Create plot with varying beta ratio

```
In [623]: beta = np.arange(1.2, 0.7, -0.1)
         f, ax = plt.subplots(beta.size, sharex=True)
         for i in range(beta.size):
             # initialize the different ratio of velocities
            unitconversion * SE_mean_mass /
                    (SE_mean_mass + NW_mean_mass)
            data1 = temp1[temp1 < 10]
             data1_mean = np.mean(data1)
             temp2 = data['v_3d_col'] * data['TSM_1'] * \
                    np.cos(data['alpha'] / 180 * np.pi) * beta[i] * \
                    unitconversion * SE_mean_mass /
                    (SE_mean_mass + NW_mean_mass)
            data2 = temp2[temp2 < 10]
             data2_mean = np.mean(data2)
            make_hist(ax[i], data1, data2, data1_mean, data2_mean, beta[i], i)
         f.subplots_adjust(hspace=0.1)
         f.set_size_inches(8, beta.size * 2)
         plt.xlabel('Projected separation from the center of mass (Mpc)', size = 12)
         #ax[0].legend(bbox_to_anchor=(1.05, 0), loc='lower left', borderaxespad=0.)
         ax[0].legend(bbox_to_anchor=(0.005, 0.1), loc='lower left', frameon=False)
         # text explanation
         for i in range(beta.size):
            plt.text(.8, 3.5 + 4.45 * i, r"Assume < v_{relic} > / v_{3D,1} (t_{col}) = " + 
                         " {0}".format(beta[beta.size - i - 1]), va='top', size = 14)
         plt.savefig("default_prior_bounds.pdf", bbox_inches="tight")
```

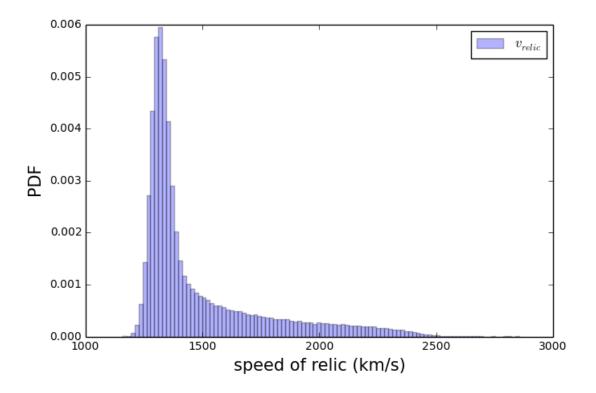


turns out we need to use proper bin size !Let us zoom in on the part where the relic position is

Part III Apply radio relic polarization prior

3.10 Initialize the speeds for the data with polarization data

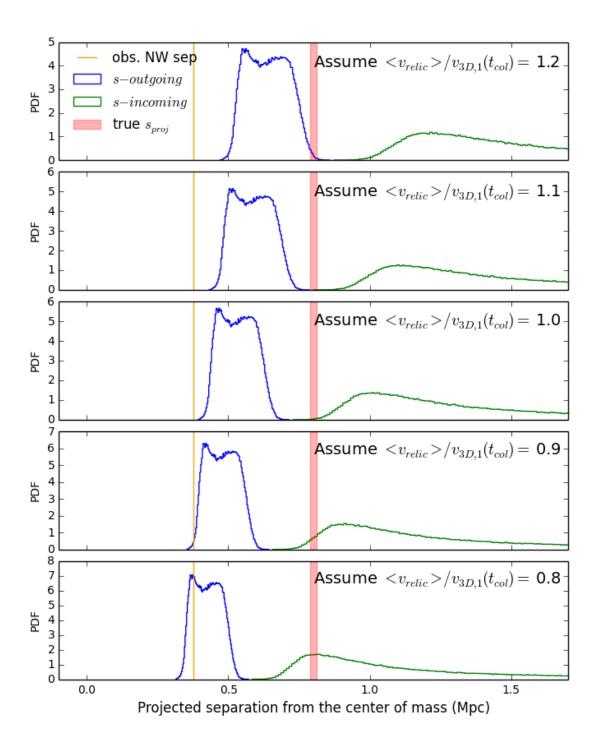
3.11 See if the speed distribution looks alright



3.12 Calculate the predicted separation

```
In [541]: r_data.columns
Out [541]:
          Index([u'm_1', u'm_2', u'z_1', u'z_2', u'd_proj', u'v_rad_obs',
          u'alpha', u'v_3d_obs', u'd_3d', u'v_3d_col', u'd_max', u'TSM_0',
         u'TSM_1', u'T', u'prob', u'v_relic', u'r_proj0_min', u'r_proj1_min',
         u'r_proj0_max', u'r_proj1_max'], dtype='object')
In [542]: r_data['TSM_0'][r_data['TSM_0']<6].mean()</pre>
Out [542]:
          0.40473912377000321
In [543]: r_data['TSM_1'][r_data['TSM_1']<6].mean()</pre>
Out [543]:
         1.3572542421114591
In [621]: beta = np.arange(1.2, 0.7, -0.1)
          f, ax = plt.subplots(beta.size, sharex=True)
          for i in range(beta.size):
              # initialize the different ratio of velocities
              temp1 = r_data['v_3d_col'] * r_data['TSM_0'] * \
                      np.cos(r_data['alpha'] / 180 * np.pi) * beta[i] *\
```

```
unitconversion * SE_mean_mass /\
            (SE_mean_mass + NW_mean_mass)
   r_{data1} = temp1[temp1 < 10]
   r_data1_mean = np.mean(r_data1)
   unitconversion * SE_mean_mass /\
           (SE_mean_mass + NW_mean_mass)
   r data2 = temp2[temp2 < 10]
   r_data2_mean = np.mean(r_data2)
   f.subplots_adjust(hspace=0.1)
f.set_size_inches(8, beta.size * 2)
plt.xlabel('Projected separation from the center of mass (Mpc)', size = 12)
#ax[0].legend(bbox_to_anchor=(1.05, 0), loc='lower left', borderaxespad=0.)
ax[0].legend(bbox_to_anchor=(0.005, 0.1),loc='lower left', frameon=False)
for i in range(beta.size):
   plt.text(.8, 7.5 + 8.75 * i,
                r"Assume $<v_{relic}> / v_{3D, 1}(t_{col}) = + \
                " {0}".format(beta[beta.size - i - 1]), va='top', size = 14)
plt.savefig("polar_prior_bounds.pdf", bbox_inches="tight")
```



Part IV

Sort the different arrays to get the extreme values as bounds

4 First from the data with default prior

```
In [545]: data.columns
Out [545]:
         Index([u'm_1', u'm_2', u'z_1', u'z_2', u'd_proj', u'v_rad_obs',
         u'alpha', u'v_3d_obs', u'd_3d', u'v_3d_col', u'd_max', u'TSM_0',
         u'TSM_1', u'T', u'prob', u'v_relic', u'r_proj0_min', u'r_proj1_min',
         u'r proj0 max', u'r proj1 max'], dtype='object')
In [546]: print "true location of relic from CM is\n" + \
           {0:.2f} Mpc".format(r_proj_NW.value)
         true location of relic from CM is
          0.80 Mpc
In [547]: print "the bounds with default priors for outgoing scenario is\n" + \ "\{0:.2f\} Mpc < r_relic < \{1:.2f\} Mpc".format(data['r_proj0_min'].min(),
                                                data['r_proj0_max'].max())
         the bounds with default priors for outgoing scenario is
         0.37 Mpc < r relic < 1.56 Mpc
In [548]: print "the bounds with default priors for incoming scenario is\n" + \n
          data['r_proj1_max'].max())
         the bounds with default priors for incoming scenario is
         0.37 Mpc < r relic < 43616847.30 Mpc
In [549]: print "the bounds with polarization priors for outgoing scenario is\n" + \n
          r_data['r_proj0_max'].max())
         the bounds with polarization priors for outgoing scenario is
         0.37 \text{ Mpc} < r_{relic} < 0.72 \text{ Mpc}
In [550]: print "the bounds with polarization priors for incoming scenario is\n" + \n
          .
'{0:.2f} Mpc < r_relic < {1:.2f} Mpc".format(r_data['r_proj1_min'].min(),
                                                 r_data['r_proj1_max'].max())
```

the bounds with polarization priors for incoming scenario is 0.37 Mpc < r_relic < 43616847.30 Mpc