```
Version 0.0
- The initial version of the old TSM v11 that is now being tracked in Git.
from __future__ import division
import numpy
import scipy.integrate
import pickle
import profiles
import time
import sys
import cosmo
# Constants and conversions
G = 4.3*10**(-9) #newton's constant in units of Mpc*(km/s)**2 / M sun
c = 3e5 \# speed of light km/s
sinGyr = 31556926.*10**9 # s in a Giga-year
kginMsun = 1.98892*10**30 # kg in a solar mass
kminMpc = 3.08568025*10**19 # km in a Megaparsec
minMpc = 3.08568025*10**22 # m in a Megaparsec
r2d = 180/numpy.pi # radians to degrees conversion factor
def randdraw(A,index=None):
    This function randomly draws a variable from either and input gaussian
    distribution or an array of random draws from a parent distribution function
    (e.g. bootstrap sampling).
    Input:
    A = [(float, float) or (1D array of floats)] If (float, float) is input then
        it is interpreted as (mu, sigma) of a Gaussian distribution. If (1D
        array of floats with length > 2) then it is interpreted as a random
        sample from a parent distribution.
    index = [int] rather than drawing a random value from the array the value
    Output:
    a = [float] a random draw from either the Gaussian or sample distribution.
    N_A = numpy.size(A)
    if N_A == 2: #then assumed (mu, sigma) format
        a = A[0]+A[1]*numpy.random.randn()
    elif N_A > 2: #Then assume distribution array format
        if index == None:
           index = numpy.random.randint(N_A)
        a = A[index]
    elif N_A <2:
        print 'MCMAC.randdraw: Error, parameter input array is not a valid size, exiting'
        sys.exit()
    return a, index
def vrad(z1,z2):
    Given the redshifts of the two subclusters this function calculates their
    relative line-of-sight velocity (km/s).
    Input:
    z1 = [float] redshift of subcluster 1
    z2 = [float] redshift of subcluster 2
    v los = [float; units:km/s] line-of-sight relative velocity of the two
        subclusters
    v1 = ((1+z1)**2-1)/((1+z1)**2+1)*c
    v2 = ((1+z2)**2-1)/((1+z2)**2+1)*c
    return numpy.abs(v1-v2)/(1-v1*v2/c**2)
def f(x,a,b):
    This is the functional form of the time since merger integral for two point
    masses.
```

```
18/03/2023, 19:08
                  https://s3.us-west-2.amazonaws.com/secure.notion-static.com/11baa7cd-0737-411a-905e-f69ba073e7a9/MCMAC.py?X-A...
     return 1/numpy.sqrt(a+b/x)
 def TSMptpt(m_1,m_2,r_200_1,r_200_2,d_end,E):
     This function calculates the time it takes for the system to go from a
     separation of r_200_1+r_200_2 to d_end. It is based on the equations of
     motion of two point masses, which is valid in the regime where the
     subclusters no longer overlap.
     Input:
     m_1 = [float; units: M_sun] mass of subcluster 1
     m_2 = [float; units: M_sun] mass of subcluster 2
     r_200_1 = [float; units: Mpc] NFW r_200 radius of subcluster 1
     r_200_2 = [float; units: Mpc] NFW r_200 radius of subcluster 2
     d_end = [float; units: Mpc] the final separation of interest
     E = [float; units: (km/s)^2*M_sun] the total energy (PE+KE) of the two subcluster system
     Output:
     t = [float; units: Gyr] the time it takes for the system to go from a
         separation of r 200 1+r 200 2 to d end
     d start = r 200 1 + r 200 2
     C = G*m 1*m 2
     mu = m 1*m 2/(m 1+m 2)
     if E < 0:
         integral = scipy.integrate.quad(lambda x: f(x,E,C),d start,d end)[0]
         t = numpy.sqrt(mu/2)*integral/sinGyr*kminMpc
         print 'TSMptpt: error total energy should not be > 0, exiting'
         sys.exit()
     if t < 0:
         print 'TSM < 0 encountered'</pre>
     return t
 def PEnfwnfw(d,m_1,rho_s_1,r_s_1,r_200_1,m_2,rho_s_2,r_s_2,r_200_2,N=100):
     This function calculates the potential energy of two truncated NFW halos.
     Input:
     d = [float; units:Mpc] center to center 3D separation of the subclusters
     m 1 = [float; units:M sun] mass of subcluster 1 out to r 200
     rho_s_1 = [float; units:M_sun/Mpc^3] NFW scale density of subcluster 1
     r_s_1 = [float; units:Mpc] NFW scale radius of subcluster 1
     r_200_1 = [float; units:Mpc] r_200 of subcluster 1
     m_2 = [float; units:M_sun] mass of subcluster 2 out to r_200
     rho_s_2 = [float; units:M_sun/Mpc^3] NFW scale density of subcluster 2
     r_s_2 = [float; units:Mpc] NFW scale radius of subcluster 2
     r_200_2 = [float; units:Mpc] r_200 of subcluster 2
     N = [int] number of mass elements along one coordinate axis for numerical
         integration approximation
     Output:
     V_total = [float; units:(km/s)^2*M_sun] total potential energy of the two
         subcluster system
     if d >= r 200 1+r 200 2:
         V_{total} = -G*m_1*m_2/d
     else:
         # mass element sizes
         dr = r 200 2/N
         dt = numpy.pi/N
         i,j = numpy.meshgrid(numpy.arange(N),numpy.arange(N))
         # distance of 2nd NFW halo mass element from center of 1st NFW halo
         r = numpy.sqrt(((2*i+1)*dr/2*numpy.sin((2*j+1)*dt/2))**2+
                         (d+(2*i+1)*dr/2*numpy.cos((2*j+1)*dt/2))**2)
         #mass elements of 2nd NFW halo
         m = 2*numpy.pi*rho_s_2*r_s_2**3*(numpy.cos(j*dt)-numpy.cos((j+1)*dt))*(1/(1+1)*dt)
 (i+1)*dr/r_s_2)-1/(1+i*dr/r_s_2)+numpy.log(((i+1)*dr+r_s_2)/(i*dr+r_s_2)))
         #determine if 2nd halo mass element is inside or outside of 1st halo
         mask_gt = r >= r_200_1
         mask_lt = r < r_200_1
         # potential energy of each mass element (km/s)^2 * M_sun
         # NFW PE
```

```
V_nfw = numpy.sum(-4*numpy.pi*G*rho_s_1*r_s_1**3/(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)*(numpy.log(1+r_s_1)*(r[mask_lt]+dr)
(r[mask_lt]+dr)/r_s_1)-(r[mask_lt]+dr)/(r_s_1+r_200_1))*m[mask_lt])
            # Point mass PE
            V_pt = numpy.sum(-G*m_1*m[mask_gt]/r[mask_gt])
            V_total = V_nfw+V_pt
      return V_total
def NFWprop(M_200,z,c):
      Determines the NFW halo related properties. Added this for the case of user
      specified concentration.
      Input:
      M_200 = [float; units:M_sun] mass of the halo. Assumes M_200 with
            respect to the critical density at the halo redshift.
      z = [float; unitless] redshift of the halo.
      c = [float; unitless] concentration of the NFW halo.
      # CONSTANTS
      rho cr = cosmo.rhoCrit(z)/kginMsun*minMpc**3
      #calculate the r 200 radius
      r 200 = (M 200*3/(4*numpy.pi*200*rho cr))**(1/3.)
      del c = 200/3.*c**3/(numpy.log(1+c)-c/(1+c))
      r s = r 200/c
      rho s = del c*rho cr
      return del_c, r_s, r_200, c, rho_s
def MCengine(N mc,M1,M2,Z1,Z2,D proj,prefix,C1=None,C2=None,del mesh=100,TSM mesh=200):
      This is the Monte Carlo engine that draws random parameters from the
      measured distributions and then calculates the kenematic parameters of the
      Input:
      N mc = [int] number of successful Monte Carlo samplings to perform
      M1 = [(float, float) or (1D array of floats); units: M sun] M 200 of
            subcluster 1. If (float, float) is input then it is interpreted as
             (mu, sigma) of a Gaussian distribution. If (1D array of floats with
            length > 2) then it is interpreted as a random sample from a parent
            distribution.
      M2 = [(float, float) or (1D array of floats); units: M sun] M 200 of
            subcluster 2. Must have same format as M1.
      Z1 = [(float, float) or (1D array of floats)] redshift of subcluster 1.
            Similar input format to M1.
      Z2 = [(float, float) or (1D array of floats)] redshift of subcluster 2.
            Similar input format to M1.
      D_proj = [(float, float, float) or (1D array of floats); units:Mpc]
            projected separation of the two subclusters. If (float, float, float)
            is input then it is interpreted as (d_proj, c1_sigma, c2_sigma), where
            d proj is the projected distance of the two halos and c1 sigma and
            c2 sigma are the centroid errors of each halo. If (1D array of floats
            with length > 3) then it is interpreted as a random sample from the
            parent distribution of d proj.
      prefix = [string] prefix to attach to all output
      C1 = [(float, float) or (1D array of floats); unitless] NFW concentration of
            subcluster 1. If (float, float) is input then it is interpreted as
             (mu, sigma) of a Gaussian distribution. If (1D array of floats with
            length > 2) then it is interpreted as a random sample from a parent
            distribution. If None then Duffey et al. (2008) scaling relation is used
            to determine the concentration based on M1 value. Must have same format
      C2 = [(float, float) or (1D array of floats); unitless] NFW concentration of
            subcluster 2. Must have same format as M2.
      del_mesh = [int] number of mass elements along one coordinate axis for
            numerical integration approximation of potential energy
      TSM_mesh = [int] number of elements along the separation axis when
            performing the numerical integration appoximation of the TSM
      Output:
      In general the 1D arrays of floats are output for each of the following
      parameters. The Nth element of each array corresponds to the system
      properties of the Nth viable solution.
```

```
m_1_out = [units: M_sun] M_200 of halo 1
m_2_out = [units: M_sun] M_200 of halo 1
z_1_out = redshift of halo 1
z_2_out = redshift of halo 2
alpha_out = [units: degrees] merger axis angle with respect to the plane of
    the sky. 0 corresponds to in the plane of the sky, 90 corresponds to
    along the line-of-sight.
d_proj_out = [units: Mpc] projected center to center separation of the
    halos in observed state
d_3d_out = [units: Mpc] 3D center to center separation of the halos in
    observed state
d_max_out = [units: Mpc] 3D maximum center to center separation of the two
    halo at the apoapsis
v_rad_obs_out = [units: km/s] relative line-of-sight velocity of the two
    halos at the observed time
v 3d obs out = [units: km/s] relative 3D velocity of the two halos at the
    observed time
v 3d col out = [units: km/s] relative 3D velocity of the two halos at the
    collision time
TSM 0 out = [units: Gyr] time it took the system to reach the observed state
    from the collision state. Assuming the system has not reached its
    apoapsis since colliding.
TSM 1 out = [units: Gyr] time it took the system to reach the observed state
    from the collision state. Assuming the system has passed throught the
    apoapsis once since colliding.
T out = [units: Gyr] period of the system
prob out = This is the probability of observing the system with the randomly
   drawn paramers. This accounts for the fact that it is more likely to
   observe the system near the apoapsis rather than with zero separation,
   due to the system spending more of its time at large separation. Simply
  an effect of the velocity of the system.
# Verify user input
# Check to make sure mass inputs for each halo are the same
N M1 = numpy.size(M1)
N M2 = numpy.size(M2)
if N M1 != N M2:
    print 'MCMAC.MCengine: Error, the mass inputs for the two halos must \
        be the same type and size. For example, you cannot mix input \
        format (mu, sigma) for halo1 with (1D array of floats) format \
        for halo 2. Nor can the size of the (1D array of floats) input \
        be different. This is to facilitate correct covariance handeling. \
        Exiting.'
    sys.exit()
if C1 != None or C2 != None:
    N C1 = numpy.size(C1)
    N C2 = numpy.size(C2)
    if N C1 != N M1 or N C2 != N M2:
        print 'MCMAC.MCengine: Error, the concentration inputs for the two \
        halos must be the same type and size as the mass inputs. For \
        example, you cannot mix input format (mu, sigma) for M1 with \
        (1D array of floats) format for C1, or vise versa. Nor can the size\
        of the (1D array of floats) input be different. This is to \
        facilitate correct covariance handeling. Exiting.'
        sys.exit()
    elif N M1 > 2:
        print 'MCMAC.MCengine: Assuming that the order and values of the \
        C1 and C2 (1D array of floats) are correlated with the order and \
        values of the M1 and M2 (1D array of floats). This is meant to \
        maintain proper covariance.'
N_D_proj = numpy.size(D_proj)
if N_D_proj == N_M1:
    print 'MCMAC.MCengine: Assuming that the order and values of the \
    D_proj (1D array of floats) are correlated with the order and values of\
    the M1 and M2 (1D array of floats). This is meant to maintain proper\
    covariance.'
```

```
N_Z1 = numpy.size(Z1)
    if N_Z1 == N_D_proj:
        print 'MCMAC.MCengine: Warning. It is currently assumed that the halo\
        redshifts and projected separation estimates are uncorrelated. \
        Covariance will not be handled correctly.'
    if N_Z1 > 2 and N_Z1 == N_M1:
        print 'MCMAC,MCengine: Warning. It is currently assumed that the halo ∖
        redshifts and mass estimates are uncorrelated. Covariance will not be \setminus
        handled correctly.'
    i = 0
   # Create the output arrays
   m_1_out = numpy.zeros(N_mc)
   m_2_out = numpy.zeros(N_mc)
    z_1_out = numpy.zeros(N_mc)
    z 2 out = numpy.zeros(N mc)
    d proj out = numpy.zeros(N mc)
    v rad obs out = numpy.zeros(N mc)
    alpha out = numpy.zeros(N mc)
    v 3d obs out = numpy.zeros(N mc)
    d 3d out = numpy.zeros(N mc)
    v 3d col out = numpy.zeros(N mc)
    d max out = numpy.zeros(N mc)
    TSM 0 out = numpy.zeros(N mc)
    TSM 1 out = numpy.zeros(N mc)
    T out = numpy.zeros(N mc)
    prob out = numpy.zeros(N mc)
    N d = numpy.size(D proj)
    t start = time.time()
    while i < N mc:
        # Draw random mass and redshift parameters
        m_1, index = randdraw(M1)
        m 2, index = randdraw(M2,index)
        if m 1 <= 0 or m 2 <=0:
            # Discard these unphysical draws, this can happen if a Gaussian
            # distribution is speficied since at least part of the tail will be
            # in the negative mass range.
            continue
        z_1, temp = randdraw(Z1)
        z 2, temp = randdraw(Z2)
        # Draw random projected separation
        if N_d == 3: #then assumed (D_proj, c1_sigma, c2_sigma)
            #draw del_r1, del_r2, phi1 and phi2, where phi is the
            del_r_1 = D_proj[1] * numpy.random.randn()
            del r 2 = D proj[2] * numpy.random.randn()
            phi 1 = numpy.random.rand()*2*numpy.pi
            phi 2 = numpy.random.rand()*2*numpy.pi
            #calculate the projected separation based on drawn centers
            d_proj = numpy.sqrt((D_proj[0]-del_r_1*numpy.cos(phi_1)+del r 2*numpy.cos(phi 2))**2+
(del r 1*numpy.sin(phi 1)+del r 2*numpy.sin(phi 2))**2)
        elif N d > 3: #then assume distribution array format
            if N D proj == N M1:
                # assume D proj array values and order correlated with M1 and M2
                # array values and order.
                d proj, index = randdraw(D proj,index)
            else:
                # assume D proj is uncorrelated with M1 and M2
                d proj = D proj[numpy.random.randint(N d)]
        elif N d < 3:
            print 'MCMAC.MCengine: Error, D proj parameter input array is not a valid size,
exiting'
            sys.exit()
        d_proj = abs(d_proj)
        # Define NFW halo properties
        if C1 == None:
            # Then use scaling relation to determine concentration
```

```
del_c_1, r_s_1, r_200_1, c_1, rho_s_1 = profiles.nfwparam_extended(m_1/1e14,z_1)
        else:
            # Then user specified concentration
            # Draw random concentration parameter
            c_1, index = randdraw(C1,index)
            del_c_1, r_s_1, r_200_1, c_1, rho_s_1 = NFWprop(m_1, z_1, c_1)
        if C2 == None:
            del_c_2, r_s_2, r_200_2, c_2, rho_s_2 = profiles.nfwparam_extended(m_2/1e14,z_2)
        else:
            c_2, index = randdraw(C2,index)
            del_c_2, r_s_2, r_200_2, c_2, rho_s_2 = NFWprop(m_2, z_2, c_2)
        # Reduced mass
        mu = m_1*m_2/(m_1+m_2)
        # Calculate potential energy at time of collision
        PE col = PEnfwnfw(0,m 1,rho s 1,r s 1,r 200 1,m 2,rho s 2,r s 2,r 200 2,N=del mesh)
        # Calculate maximum 3D free-fall velocity
        # Note that this assumes that the clusters begin with infinite
        # separation and their observed mass. It also guarntees that all
        # solutions will be for a bound system.
        v 3d max = numpy.sqrt(-2/mu*PE col)
        # Calculate observed radial velocity
        v rad obs = vrad(z 1, z 2)
        # Draw a random alpha, merger axis angles with respect to the sky
        ALPHA = numpy.random.rand()*numpy.pi/2
        #Since not all alpha are equally likely.
        alpha = (1-numpy.cos(ALPHA))*numpy.pi/2
        # Calculate the 3D velocity at observed time
        v_3d_obs = v_rad_obs/numpy.sin(alpha)
        if v 3d obs > v 3d max:
            # then the randomly chosen alpha is not valid
            continue
        # Calculate the 3D separation
        d_3d = d_proj/numpy.cos(alpha)
        # Calculate the potential energy at observed time
        PE_obs = PEnfwnfw(d_3d,m_1,rho_s_1,r_s_1,r_200_1,m_2,rho_s_2,r_s_2,r_200_2,N=del_mesh)
        # Calculate the 3D velocity at collision time
        v_3d_col = numpy.sqrt(v_3d_obs**2+2/mu*(PE_obs-PE col))
        if v 3d col > v 3d max:
            # then the randomly chosen alpha is not valid
            continue
        # Total Energy
        E = PE obs+mu/2*v 3d obs**2
        # Calculate PE from d = 0 to r_200_1+r_200_2
        del TSM mesh = (r 200 1+r 200 2)/(TSM mesh-1)
        d = numpy.arange(0.00001,(r 200 1+r 200 2)+del TSM mesh,del TSM mesh)
        PE_array = numpy.zeros(TSM_mesh)
        for j in numpy.arange(TSM_mesh):
            PE array[j] =
PEnfwnfw(d[j],m_1,rho_s_1,r_s_1,r_200_1,m_2,rho_s_2,r_s_2,r_200_2,N=del_mesh)
        # Calculate d max
        if E >= -G*m_1*m_2/(r_200_1+r_200_2):
            # then d_max > r_200_1+r_200_2
            d_{max} = -G*m_1*m_2/E
        else:
            # d_max <= r_200_1+r_200_2
```

```
## find closet d value less than d_max
    #idx=(numpy.abs(PE_array-E)).argmin()
    \#d_max = d[idx]
    # determine d_max, ensuring that E is always > PE
    mask_tmp = E >= PE_array
    d_{max} = d[mask_{tmp}][-1]
# Calculate TSM_0
if d_3d >= r_200_1+r_200_2:
    # then halos no longer overlap
    # calculate the time it takes to go from d=0 to r_200_1+r_200_2
    TSM_0a = numpy.sum(del_TSM_mesh/numpy.sqrt(2/mu*(E-PE_array))*kminMpc/sinGyr)
    # calculate th time it takes to go from d = r_200_1+r_200_2 tp d_3d
    TSM_0b = TSMptpt(m_1, m_2, r_200_1, r_200_2, d_3d, E)
    TSM 0 = TSM 0a+TSM 0b
else:
    # then d_3d < r_200_1+r_200_2, halos always overlap
    # calculate the time it takes to go from d=0 to d 3d
    mask = d <= d 3d
    TSM 0 = numpy.sum(del TSM mesh/numpy.sqrt(2/mu*(E-PE array[mask]))*kminMpc/sinGyr)
# Check that TSM 0 < Age of Universe at (z 1+z 2)/2
age = cosmo.age((z_1+z_2)/2)
if TSM 0 > age:
    # unlikely that this system could occur
    continue
# Calculate period
if E >= -G*m \ 1*m \ 2/(r \ 200 \ 1+r \ 200 \ 2):
    # then d max > r 200 1+r 200 2
    # calculate th time it takes to go from d = r 200 1+r 200 2 to d max
    if d 3d < r 200 1+r 200 2:
        #then TSM 0a has not been previously defined
        TSM 0a = numpy.sum(del TSM mesh/numpy.sqrt(2/mu*(E-PE array))*kminMpc/sinGyr)
    TSM_0b = TSMptpt(m_1, m_2, r_200_1, r_200_2, d_max, E)
    T = 2*(TSM 0a+TSM 0b)
    # then d max < r 200 1+r 200 2
    # calculate the time it takes to go from d=0 to d_max
    mask = d \le d max
    T = 2*numpy.sum(del_TSM_mesh/numpy.sqrt(2/mu*(E-PE_array[mask]))*kminMpc/sinGyr)
# Calculate probability of d 3d
prob = TSM_0/(T/2)
# Calculate TSM 1
TSM 1 = T-TSM 0
if TSM 0 < 0:
    print 'TSM < 0 encountered'</pre>
# Write calculated merger parameters
m_1_out[i] = m_1
m = 2 \text{ out}[i] = m = 2
z 1 out[i] = z 1
z_2_{out[i]} = z_2
d_proj_out[i] = d_proj
v_rad_obs_out[i] = v_rad_obs
alpha_out[i] = alpha*180/numpy.pi
v_3d_obs_out[i] = v_3d_obs
d_3d_out[i] = d_3d
v_3d_col_out[i] = v_3d_col
d_max_out[i] = d_max
TSM_0_out[i] = TSM_0
TSM_1_out[i] = TSM_1
T_{out}[i] = T
prob_out[i] = prob
i+=1
```

```
# Estimate calculation time
    if i== 10 or i == 100 or i%1000 == 0:
        del_t = time.time()-t_start
        t_total = N_mc*del_t/i
        eta = (t_total-del_t)/60
        print 'Completed Monte Carlo iteration {0} of {1}.'.format(i,N_mc)
        print '~{0:0.0f} minutes remaining'.format(eta)
# Pickle the results of the MC analysis
filename = prefix+'_m_1.pickle'
F = open(filename,'w')
pickle.dump(m_1_out,F)
F.close()
filename = prefix+'_m_2.pickle'
F = open(filename,'w')
pickle.dump(m 2 out,F)
F.close()
filename = prefix+' z 1.pickle'
F = open(filename,'w')
pickle.dump(z 1 out,F)
F.close()
filename = prefix+' z 2.pickle'
F = open(filename,'w')
pickle.dump(z 2 out,F)
F.close()
filename = prefix+' d proj.pickle'
F = open(filename,'w')
pickle.dump(d proj out,F)
F.close()
filename = prefix+' v rad obs.pickle'
F = open(filename,'w')
pickle.dump(v_rad_obs_out,F)
F.close()
filename = prefix+' alpha.pickle'
F = open(filename,'w')
pickle.dump(alpha_out,F)
F.close()
filename = prefix+'_v_3d_obs.pickle'
F = open(filename,'w')
pickle.dump(v_3d_obs_out,F)
F.close()
filename = prefix+'_d_3d.pickle'
F = open(filename,'w')
pickle.dump(d_3d_out,F)
F.close()
filename = prefix+' v 3d col.pickle'
F = open(filename,'w')
pickle.dump(v_3d_col_out,F)
F.close()
filename = prefix+' d max.pickle'
F = open(filename,'w')
pickle.dump(d_max_out,F)
F.close()
filename = prefix+' TSM 0.pickle'
F = open(filename,'w')
pickle.dump(TSM_0_out,F)
F.close()
filename = prefix+' TSM 1.pickle'
F = open(filename,'w')
pickle.dump(TSM_1_out,F)
F.close()
filename = prefix+'_T.pickle'
F = open(filename,'w')
pickle.dump(T_out,F)
F.close()
filename = prefix+'_prob.pickle'
F = open(filename,'w')
pickle.dump(prob_out,F)
```

F.close()

return

m_1_out,m_2_out,z_1_out,z_2_out,d_proj_out,v_rad_obs_out,alpha_out,v_3d_obs_out,d_3d_out,v_3d_col_
out,d_max_out,TSM_0_out,TSM_1_out,T_out,prob_out