

NUR Hand-in exercise 1

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Abstract

The source code and the outputs of Hand-in exercise 1 are shown in this report.

1 Source code

```
1 #!/usr/bin/env python
2 # coding: utf-8
3
4 # In [1]:
5
6
7 import numpy as np
8 import matplotlib.pyplot as plt
9 import sys
10 sys.stdout = open('outputs.txt', 'w')
11
12 '''
13 Question 1
14 (a)
15 Poisson probability distribution function
16 P_lambda(k)=(lambda^k)*(e^-lambda)/(k!)
17 '''
18 print('Question 1')
19 print('(a) Poisson function')
20 def Poisson(p_lam, p_k):
21     '''
22     Possion Function (lambda^k)*(e^-lambda)/(k!)
23     p_lam ——parameter lambda
24     p_k ——parameter k
25     '''
26     factorial = np.float64(1)
27     for i in range(1, p_k+1): #calculate the factorial of k
28         factorial = factorial * np.float(i)
29     num1=np.float(p_lam**p_k) #the 1st item of numerator
30     num2=np.float(np.exp(-p_lam)) #the 2nd item of numerator
31     den=factorial #denominator
32     P=np.float64(num1*num2/den) #calculatae P
33     return P
34 for i in ((1,0),(5,10),(3,21),(2.6,40)):
35     print('Poisson{} = '.format(i),Poisson(i[0],i[1]))
36
37 '''
38 (b)
39 random number generator
40 '''
```

```

41 print('\n(b) Random number generator' )
42 def generator(n,seed):
43     '''
44     combined number generator:
45     LCG(XOR-shift) ^ MMC
46     period = period of xorshift * period of MMC
47     the parameters will be given later
48     n = the amount of numbers
49     seed = initial seed
50     '''
51     #parameters of each generator
52     ##XOR-shift 64-bit
53     XOR_a1=21
54     XOR_a2=35
55     XOR_a3=4
56     bit64=2**64-1
57     ##LCG
58     LCG_a=3935559000370003845
59     LCG_c=2691343689449507681
60     mod = 2**64
61     ##MMC
62     MMC_a=4294957665
63     #seed
64     x=seed
65     l=seed
66     m=seed
67     number = np.zeros(n)
68     for i in range(n):
69         #XORshift
70         '''
71         XOR-shift : keep the number in 64-bit and do logical left or
                       right bit shift , which means some the bits that are moved out
                       of memory.
72         the 'new bits' will be filled by 0
73         '''
74         x = x ^ (x >> XOR_a1)
75         x = x ^ (x << XOR_a2) & bit64 # do a logical 'and' to cut
76         the number to 64 bits.
77         x = x ^ (x >> XOR_a3)
78         # LCG part
79         '''
80         use the output of XOR-shift to get a new number, the new
81         period = the period of XOR which is 2^64-1
82         l will not fed back into the XOR-shift , which works
83         unperturbed
84         LCG : calculate the remainder and put it back into iteration.
85         Hence, generally the period is a factor of the divisor.
86         '''
87         l = (LCG_a * x + LCG_c) % mod
88         ##MMC
89         '''
90         only the 32 bits of the old number will be manipulated and
91         propagated to the next one
92         '''
93         m = (MMC_a*(m & (2**32-1))+(m >>32)) & bit64 # constrain
94         the bits of MMC
95         #combine them
96         number[i] = (l ^ m)
97         #normalise in (0,1) /maxnumber of 64.
98         #Note that 'period' shows the repeating information (how long
99         the sequence is), not the range of radom number
100        number=np.array(number)/(2**64-1)

```

```

94     return number
95 #set the seed
96 seed = 777
97 print('seed = ', seed)
98 #first 1000 numbers
99 n1000 = generator(1000,seed)
100 fig1 = plt.figure(1)
101 ax1.1 = fig1.add_subplot(1,2,1)
102 ax1.1.scatter(n1000[0:-2], n1000[1:-1])
103 ax1.1.set_xlabel("$X_i$")
104 ax1.1.set_ylabel("$X_{i+1}$")
105 ax1.1.set_title('Sequential 1000 numbers')
106 # 1 million numbers
107 nlm=generator(10**6,seed)
108 ax1.2=fig1.add_subplot(1,2,2)
109 ax1.2.hist(nlm, bins=np.linspace(0.0, 1.0, 21))
110 ax1.2.set_title('Histogram of 1 million numbers')
111 ax1.2.set_xlabel('bins')
112 ax1.2.set_ylabel('quantity of numbers')
113 fig1.tight_layout()
114 fig1.savefig("1.png")
115 fig1.show()
116 print('figure of random number generator please see fig.1')
117 '''
118 Question 2
119 (a)
120 Statellite galaxies
121 number density profile
122 solve normalisation coefficient A
123 '''
124 print('\n\nQuestion 2')
125 print('(a) solve A ')
126 #pick up generated numbers and set them to the required range.
127 a = (2.5-1.1)*n1000[7]+1.1
128 b = (2-0.5)*n1000[77]+0.5
129 c = (4-1.5)*n1000[777]+1.5
130 print(' a = ', a)
131 print(' b = ', b)
132 print(' c = ', c)
133 '''
134 BTW, I don't think 2(a) is slovable without viral radius, so I will
    set it to 1. Then,  $x = r$ .
135  $n(x) = A \cdot 100 \cdot (x/b)^{(a-3)} \cdot \exp(-(x/b)^c)$ 
136  $3rd\_integral \ n(x) \ d(4 \cdot \pi \cdot x^3)/3 = 100 \implies 3rd\_integral \ n(x) \cdot 4 \cdot \pi \cdot$ 
 $x^2 \ dx = 100 \implies 3rd\_integral \ A \cdot (x/b)^{(a-3)} \cdot \exp(-(x/b)^c) \cdot 4 \cdot$ 
 $\pi \cdot x^2 \ dx = 1$ 
137 Hence,  $A = 1 / 3rd\_integral \ (x/b)^{(a-3)} \cdot \exp(-(x/b)^c) \cdot 4 \cdot \pi \cdot x^2 \ dx$ 
138 '''
139 N_sat=100.
140 # Now do the integral, using trapeziodal rules
141 def integrator(function, lower, upper, intervals):
142     '''
143     trapeziodal rule:
144      $s = h/2(f(x_0) + 2 \sum(f(x_{1 \text{ to } n-1})) + f(x_n))$ 
145     function =
146     lower = lower limit of the integral
147     upper = upper limit
148     interbvals = n of intervals
149     '''
150     h = (upper - lower) / intervals
151     S = 0.5*(function(lower) + function(upper))
152     for i in range(1,intervals):

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153         S += function(lower + i*h)
154         integral = h * S
155         return integral
156 # def the function
157 function_x = lambda x: (x/b)**(a-3.)*np.exp(-(x/b)**c)*4.*np.pi*x
158         **2.
159 # calculate A
160 since when x = 0, n(x) is improper
161 I set the lower limit to a small value
162 '''
163 A = 1. / integrator(function_x, 10**(-32), 5., 10000)
164 print(' A = ',A)
165
166
167
168 '''
169 (b)
170 make a log-log plot , plot 4 points and do interpolations.
171 I tried cubic, quadratic and linear spline interpolation.
172 None of the results is satisfied and linear spline is the 'best'
173         among them.
174 '''
175 print('\n(b) log-log plot. See fig.2.')
176 #def n(x)
177 n_x = lambda x : A*N_sat*(x/b)**(a-3)*np.exp(-(x/b)**c)
178 #true data x = 10^-4, 10^-2, 10^-1, 1, 5
179 data_ture_x=np.array([10**-4,10**-2,10**-1,1,5])
180 data_ture_n=n_x(data_ture_x)
181 fig2 = plt.figure(2)
182 fig2.suptitle('log-log plot of n(x)')
183 ax2_1 = fig2.add_subplot(1,1,1)
184 ax2_1.set_xscale('log')
185 ax2_1.set_yscale('log')
186 ax2_1.set_xlabel('$x = r/r_{vir}$ (log)')
187 ax2_1.set_ylabel('number density (log)')
188 ax2_1.plot(data_ture_x, data_ture_n, 'o', label='true data')
189 #Neville's method
190 def neville(x_ture, y_ture, x_new):
191     '''
192     Neville's Algorithm:
193     p_i, i(x) = y_i
194     p_i, j(x) = [(x-j-x)*p_i, j-1(x)+(x-x_i)*p_i+1, j(x)] / x_j-x_i
195     '''
196     n = len(x_ture)
197     y = y_ture.copy()
198     for m in range(1,n):
199         for i in range(n-m):
200             y[i] = ((x_new - x_ture[i+m])*y[i]+(x_ture[i]-x_new)*y[
201                 i+1])/(x_ture[i]-x_ture[i+m]))
202     return y[0]
203 inter_x=np.linspace(10**-4,5,1000)
204 inter_n_Neville=np.zeros(1000)
205 for i in range(len(inter_n_Neville)):
206     inter_n_Neville[i] = neville(data_ture_x, data_ture_n, inter_x[i]
207 ])
208 ax2_1.plot(inter_x, inter_n_Neville, 'b', label='Neville')
209 #linear interpolation
210 def linear(inter_x):
211     '''
212     linear interpolation is quite easy. Take the true points and

```

```

        calculate the slope and constant for every two adjacent points.
    '''
    #calculate k & b in each range
    # $10^{-4}, 10^{-2}$ 
    k0 = (data_ture_n[1]-data_ture_n[0])/(data_ture_x[1]-
    data_ture_x[0])
    b0 = data_ture_n[1]-k0*data_ture_x[1]
    # $10^{-2}, 10^{-1}$ 
    k1 = (data_ture_n[2]-data_ture_n[1])/(data_ture_x[2]-
    data_ture_x[1])
    b1 = data_ture_n[2]-k1*data_ture_x[2]
    # $10^{-1}, 1$ 
    k2 = (data_ture_n[3]-data_ture_n[2])/(data_ture_x[3]-
    data_ture_x[2])
    b2 = data_ture_n[3]-k2*data_ture_x[3]
    # $1, 5$ 
    k3 = (data_ture_n[4]-data_ture_n[3])/(data_ture_x[4]-
    data_ture_x[3])
    b3 = data_ture_n[4]-k3*data_ture_x[4]
    # do interpolation
    inter_n = np.zeros(len(inter_x))
    for i in range(len(inter_x)):
        if inter_x[i] < data_ture_x[1]:
            inter_n[i] = k0*inter_x[i] + b0
        if inter_x[i] >= data_ture_x[1] and inter_x[i] <
        data_ture_x[2]:
            inter_n[i] = k1*inter_x[i] + b1
        if inter_x[i] >= data_ture_x[2] and inter_x[i] <
        data_ture_x[3]:
            inter_n[i] = k2*inter_x[i] + b2
        if inter_x[i] >= data_ture_x[3]:
            inter_n[i] = k3*inter_x[i] + b3
    return inter_n
inter_n_linear = linear(inter_x)
ax2.1.plot(inter_x, inter_n_linear, 'r', label='linear spline')
ax2.1.legend(loc='lower left')
fig2.savefig('2b.png')
fig2.show()

'''
(c) Derivative
Analytical Derivative is  $(b^3 * e^{-(x/b)^c} * (x/b)^a * (-3 + a - c * (x/b)^c)) / x^4$ 
'''
print('\n(c) Derivative')
dn_x = lambda x: A*N_sat*(b**3*np.exp(-(x/b)**c)*(x/b)**a*(-3+a-c*(
x/b)**c))/x**4
print('Analytical dn(x)/dx (x=5) = ', dn_x(b))
# Use central difference to calculate the derivative.
def central_difference(function, x, h):
    '''
    derivative =  $\lim_{h \rightarrow 0} [f(x+h) - f(x-h)] / 2 * h$ 
    function =
    x =
    h = step size
    '''
    derivative = (function(x + h) - function(x - h)) / (2 * h)
    return derivative
print('Numerical dn(x)/dx (x=5) = ', central_difference(dn_x, b
, 10**-8))

'''
'''

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263 (d)
264 probability distrubution sampling
265 '''
266 print('\n(d) probability distribution and the positions of
      satellites')
267 # Choose sampling method to be rejection.
268 n2m=generator(2*10**6,seed)
269 def rejection_sample(function, x_min, x_max, y_min, y_max, n):
270     '''
271     I use rejection sampling here is because it is easy to apply
272     with my random number generator
273     according to the slides. If the  $y_i < p(x)$  then the combination
274     of  $x_i$  &  $y_i$  will be accepted.
275
276     function =
277     x_min, x_max =
278     y_min, y_max =
279     n = amount of points wanted to be tested < 2 million
280
281     '''
282     accepted_x=[]
283     accepted_y=[]
284     for i in range(n):
285         x = n2m[i] * (x_max - x_min) + x_min
286         y = n2m[-i] * (y_max - y_min) + y_min # I don't want the
287         original random number are the same.
288         if y < function(x):
289             accepted_x.append(x)
290             accepted_y.append(y)
291
292     return accepted_x, accepted_y
293
294 # def probability function
295 p_x = lambda x: A*(x/b)**(a-3)*np.exp(-(x/b)**c)*4*np.pi*x**2
296
297 sample_x, sample_y = rejection_sample(p_x, 0, 3., 0, 4., 2*10**6)
298 fig3 = plt.figure(3)
299 fig3.suptitle("rejection sampling")
300 ax3_1 = fig3.add_subplot(1,1,1)
301 ax3_1.set_xlabel('x')
302 ax3_1.set_ylabel('Probability distribution')
303 ax3_1.scatter(sample_x, sample_y,s=2.7)
304 ax3_1.plot(np.arange(10**-8, 3, 0.01), [p_x(i) for i in np.arange
305 (10**-8, 3, 0.01)], 'r-',label='p(x)dx')
306 ax3_1.legend(loc='best')
307 fig3.savefig('2d.png')
308 fig3.show()
309 # position of each galaxy
310 def halo(n):
311     '''
312     r = x * viral radius
313     as befor, set viral radius = 1
314     sample_x is the possible positions, drag out some of them (
315     literally 100)
316     n = number_of_satellites
317
318     '''
319     satel_x = sample_x[0:n] # which indicate r
320     satel_phi = 2*np.pi*n2m[0:n]
321     satel_theta = np.pi*n2m[-n:-1]
322     return satel_x, satel_phi, satel_theta
323
324 halo_100 = halo(100)

```

```

319 print('Sampling distribution. See fig.3.')
320 print('Positions of 100 satellites :')
321 print('r = ')
322 print(halo_100[0])
323 print('phi = ')
324 print(halo_100[1])
325 print('theta = ')
326 print(halo_100[2])
327
328 '''
329 (e)
330 1000 halos, each contains 100 satellites
331 '''
332 print('\n(e) 1000 haloes and histogram')
333 #zeros 1000 haloes (one halo per row) and each element represents
    the 'x = r/ r_vir' of the satellite
334 haloes=np.zeros((1000,100))
335 for i in range(1000):
336     haloes[i]=(sample_x[100*i:100*(i+1)])
337 radii = np.reshape(haloes,10**5)
338 #def N_x
339 N_x = lambda x: A*N_sat*(x/b)**(a-3)*np.exp(-(x/b)**c)*4*np.pi*x**2
340 # Plot log-log of N(x).
341 fig4 = plt.figure(4)
342 fig4.suptitle('Log-log Plot of N(x)')
343 ax4_1 = fig4.add_subplot(1,1,1)
344 ax4_1.set_xlabel('log(x)')
345 ax4_1.set_ylabel('N(x) = n(x)*4*pi*x^2')
346 ax4_1.set_xscale('log')
347 ax4_1.set_yscale('log')
348 ax4_1.plot(np.arange(10**-8, 5, 0.01), N_x(np.arange(10**-8, 5,
    0.01)), label='N(x)')
349 amount,bin_edge,_,=ax4_1.hist(radii, bins=np.logspace(np.log10
    (10**-4),np.log10(5.0), 21))
350 fig4.savefig('2e.png')
351 fig4.show()
352 print('Histogram for 1000 haloes. See fig.4.')
353 '''
354 (f)
355 rooting finding
356 Newton-Raphson method needs derivative of the original function,
    which is really difficult to calculate analytically.
357 Secant method may diverge sometimes.
358 so i chose the most basic one : bisection N_x1*N_x2
359 '''
360 print('\n(f) rooting finding')
361 def bisection_root(function, x1, x2, epsilon ,iteration):
362     '''
363     test the sign of f(x1)*f(x2), if it is negative then the root
    is in this bracket.
364     (x1,x2) = range of the root
365     epsilo = percision
366     iteration = iteration times (in case over-shooting)
367     '''
368     N_x1 = function(x1)
369     N_x2 = function(x2)
370     if N_x1*N_x2 > 0:
371         print('no root or multiple roots in this range, please
    reset initial bracket')
372         return False
373     for i in range(iteration):
374         mid = (x1 + x2) / 2.

```

```

375     N_mid = function(mid)
376     N_x1 = function(x1)
377     if N_x1 * N_mid < 0: # mid point is the new right point x2
378         x2 = mid
379         if abs(N_mid) < epsilon: # acheive the percision
380             return mid
381     else: # mid point is the new left point x1
382         x1 = mid
383         if abs(N_mid) < epsilon: # acheive the percision
384             return mid
385     print("Not converge after {} iterations".format(iteration))
386 def maximum(a,b,function):
387     '''
388     use bracket method to find the maximum
389     split the bracket into 3 pieces with 2 point and compare them
390     if the right one is larger, then use the new left point to
391     replace the original left point
392     '''
393     while b-a > 10**-8 :
394         x=a+(b-a)/3.
395         y=a+2*(b-a)/3.
396         if function(x) < function(y):
397             a=x
398         else:
399             b=y
400     return function(a)
401 y=maximum(10**-4,5,N_x)
402 N_x_root = lambda x: A*N_sat*(x/b)**(a-3)*np.exp(-(x/b)**c)*4*np.pi
403     *x**2-y/2
404 root_1 = bisection_root(N_x_root, 10**-4,1,10**-4,100)
405 root_2 = bisection_root(N_x_root, 1,5,10**-4,100)
406 print('Root_1 = ', root_1)
407 print('Root_2 = ', root_2)
408 '''
409 (g) histogram and Poisson distribution
410 '''
411 print('\n(g)Histogram and Poisson distribution')
412 radial_bin_l = bin_edge[amount.argmax(axis=0)] # located the
413     largest amount from the histogram, lower limit
414 radial_bin_r = bin_edge[amount.argmax(axis=0)+1] # upper limit
415 radial_bin=[] # the radii in that largest bin
416 #test those satellites one by one , if their radii are in the range
417     , save it.
418 for i in radii:
419     if i >= radial_bin_l and i <= radial_bin_r:
420         radial_bin.append(i)
421 def quick_sort(array,i,j):
422     '''
423     pick an element as pivot and partition the given array around
424     the picked pivot.
425     '''
426     if i < j:
427         pivot = quick_sort_process(array,i,j)
428         quick_sort(array,i,pivot)
429         quick_sort(array,pivot+1,j) # do several times
430     return array
431 def quick_sort_process(array,i,j):
432     pivot = array[i]
433     while i < j:
434         while i < j and array[j] >= pivot:

```



```

432         j -= 1
433         while i < j and array[j] < pivot:
434             array[i] = array[j]
435             i += 1
436             array[j] = array[i]
437         array[i]=pivot
438     return i
439 sorted_radii = sorted(radial_bin) # my sorting function takes too
    long
440
441 median = sorted_radii[int(len(sorted_radii) / 2)]
442 percent16 = sorted_radii[int(0.16 * len(sorted_radii))]
443 percent84 = sorted_radii[int(0.84 * len(sorted_radii))]
444 print('median = ', median)
445 print('16th percentile = ', percent16)
446 print('84th percentile = ', percent84)
447 # histogram : for each halo , test how many satellites ' radii are
    in that range and count it.
448 hist_each_halo = np.zeros(1000)
449 for i in range(1000):
450     count=0
451     for j in haloes[i,:]:
452         if j >= radial_bin_l and j <= radial_bin_r:
453             count += 1
454     hist_each_halo[i] = count
455 fig5=plt.figure(5)
456 ax5_1=fig5.add_subplot(1,1,1)
457 amount_in_bin,_,_=ax5_1.hist(hist_each_halo, bins=np.arange
    (0.5,101,1),density='true')
458 poisson_value = np.zeros(100)
459 for i in range(100):
460     poisson_value[i] = Poisson(len(radial_bin)/1000.,i)
461 ax5_1.plot(np.arange(0,100,1),poisson_value,label='Poisson
    distribution')
462 ax5_1.legend(loc='upper right')
463 fig5.savefig('2g.png')
464 fig5.show()
465 print('Histogram and distribution see fig.5')
466
467 '''
468 (h) A(a,b,c)
469 '''
470 print('\n(h) A(a,b,c)')
471 a_h = np.arange(1.1, 2.51, 0.1)
472 b_h = np.arange(0.5, 2.01, 0.1)
473 c_h = np.arange(1.5, 4.01, 0.1)
474 A_h = np.zeros((len(a_h), len(b_h), len(c_h)))
475 for i in range(len(a_h)):
476     for j in range(len(b_h)):
477         for k in range(len(c_h)):
478             a=a_h[i]
479             b=b_h[j]
480             c=c_h[k]
481             A_h[i,j,k] = 1. / integrator(function_x, 10**(-16), 5.,
    1000)
482 print(len(A_h)*len(A_h[0])*len(A_h[0][0]),' values in A table')
483 print('e.g. A(a=1.1,b=0.5,c=1.5) = ', A_h[0,0,0])
484
485 '''
486 3d interpolations
487 Do interpolation for every single dimensions
488 a—>b—>c

```

```

489 '''
490 def interpolator_3d(a_ture,b_ture,c_ture,A_ture,step_size):
491     '''
492     do the interpolation on dimensions one by one.
493     interpolate a point between two original points
494     step_size = interpolation intervals, I used 0.05 > 0.01
495     this function will generate the more points, which means make
496     the array larger and has a narrower interval.
497     '''
498     a_inter = np.arange(1.1,2.51,step_size)
499     b_inter = np.arange(0.5,2.01,step_size)
500     c_inter = np.arange(1.5,4.01,step_size)
501     A_inter_a = np.zeros((len(a_inter),len(b_ture),len(c_ture))) #
502     means A after interpolation on a direction
503     for i in range(len(b_ture)):
504         for j in range(len(c_ture)):
505             for k in range(len(a_inter)):
506                 A_inter_a[k,i,j] = neville(a_ture,A_ture[:,i,j],
507                 a_inter[k])
508     # A_inter_a is the new 'true data' for b and c
509     A_inter_ab = np.zeros((len(a_inter),len(b_inter),len(c_ture)))
510     for l in range(len(a_inter)):
511         for m in range(len(c_ture)):
512             for n in range(len(b_inter)):
513                 A_inter_ab[l,n,m] = neville(b_ture,A_inter_a[l,:,m]
514                 ,b_inter[n])
515     # A_inter_ab is the new 'ture data' for c
516     A_inter_abc = np.zeros((len(a_inter),len(b_inter),len(c_inter)))
517     for o in range(len(a_inter)):
518         for p in range(len(b_inter)):
519             for q in range(len(c_inter)):
520                 A_inter_abc[o,p,q] = neville(c_ture,A_inter_ab[o,p
521                 ,:],c_inter[q])
522     return A_inter_abc
523 #test 3d interpolation ( this is not in the question so I will not
524 show the results)
525 #A_inter_3d=interpolator_3d(a_h,b_h,c_h,A_h,0.05)
526 def A_abc(a_ture,b_ture,c_ture,A_ture,a_new,b_new,c_new):
527     '''
528     define a function that can use interpolation to output an A
529     value based on any new combination of (a,b,c)
530     As same as 3d-interpolator, do one dimension by one dimension.
531     '''
532     A_inter_a = np.zeros((len(b_ture),len(c_ture))) # after
533     interpolation on 'a' direction
534     for i in range(len(b_ture)):
535         for j in range(len(c_ture)):
536             A_inter_a[i,j] = neville(a_ture,A_ture[:,i,j],a_new)
537     # A_inter_a is the new 'true data'
538     A_inter_ab = np.zeros(len(c_ture))
539     for k in range(len(c_ture)):
540         A_inter_ab[k] = neville(b_ture,A_inter_a[:,k],b_new)
541     # A_inter_ab is the new 'ture data'
542     A_inter_abc = neville(c_ture,A_inter_ab,c_new)
543     return A_inter_abc
544 #test the function build on interpolator
545 print('test the interpolator function')
546 print('e.g. A(a=1.27,b=1.27,c=2.27) = ', A_abc(a_h,b_h,c_h,A_h
547 ,1.27,1.27,2.27))
548
549
550

```

```

541
542 """
543 '''
544 Question 3
545 (a) find a,b,c
546 '''
547 data_m15=np.genfromtxt('satgals_m15.txt',skip_header=5)
548 #data_m14=np.genfromtxt('satgals_m14.txt',skip_header=5)
549 #data_m13=np.genfromtxt('satgals_m13.txt',skip_header=5)
550 #data_m12=np.genfromtxt('satgals_m12.txt',skip_header=5)
551 #data_m11=np.genfromtxt('satgals_m11.txt',skip_header=5)
552 #the first column is for x = r/vir
553 def likelihood(data,a_range,b_range,c_range):
554     '''
555     Sorry, I can't understand the question clearly so I just wrote
556     something which I think might be interesting.
557     For every single satellite, number density 'n' is a function of
558     a,b,c; fixed x; set N_sat to be 100.
559     Find the maximum n and the location(a,b,c) for every single
560     satellite and then apply it to all satellites.
561     '''
562     x_3a = data[0]
563     abc = []
564     for l in range(len(x_3a)): # for a single satellite
565         n = np.zeros((len(a_range),len(b_range),len(c_range)))
566         for i in range(len(a_range)):
567             for j in range(len(b_range)):
568                 for k in range(len(c_range)):
569                     a = a_range[i]
570                     b = b_range[j]
571                     c = c_range[k]
572                     n[i,j,k] = n_x(x_3a[l])
573                 abc.append(np.argmax(n, axis=1))
574             '''
575             I want to indicate the position(i,j,k) of the max value of
576             n
577             Then ,I will know the (a,b,c) that maximize the n for every
578             satellite.
579             but the 'argmax' is a little complicated in 3 dimensions
580             array.
581             It doesn't work well
582             '''
583         return abc
584     '''
585     I was looking forward to getting the poistion of peak and
586     revert the index to (a,b,c) value.
587     '''
588
589 a_3a = np.arange(1.1,2.5,0.1)
590 b_3a = np.arange(0.5,2,0.1)
591 c_3a = np.arange(1.5,4,0.1)
592 #likelihood(data_m15,a_3a,b_3a,c_3a)
593 '''
594 Since it didn't work ,I will not make it running now.
595 '''
596
597 (b) Interpolator for a,b,c as function of halo mass, or fitting
598 method
599 Once I get the a,b,c for each mass bin
600 I can use following fitting method to fit
601 '''
602 def least_square(ls_x,ls_y):

```

```

595     '''
596     I have written this routine on question 1 of tutorial 7.
597     '''
598     ls_x_mean=sum(ls_x)/len(ls_x)
599     ls_y_mean=sum(ls_y)/len(ls_y)
600     numer=0
601     denom=0
602     for i in range(len(ls_x)):
603         numer += (ls_x[i] - ls_x_mean) * (ls_y[i] - ls_y_mean)
604         denom += (ls_x[i] - ls_x_mean) ** 2
605     w = numer / denom
606     b = ls_y_mean - (w * ls_x_mean)
607     return w,b
608 """
609 print('\nQuestion 3. \n The code did not run as I had expected it
        to but I wrote the least-square fitting algorithm. Please find
        it in the source code section.')

```

2 Outputs

```

1 Question 1
2 (a) Poisson function
3 Poisson(1, 0) = 0.36787944117144233
4 Poisson(5, 10) = 0.01813278870782187
5 Poisson(3, 21) = 1.0193398241110193e-11
6 Poisson(2.6, 40) = 3.615123994937689e-33
7
8 (b) Random number generator
9 seed = 777
10 figure of random number generator please see fig.1
11
12
13 Question 2
14 (a) solve A
15 a = 1.6629645031839944
16 b = 1.0274077403563826
17 c = 2.878139347989354
18 A = 0.13691978822510378
19
20 (b) log-log plot. See fig.2.
21
22 (c) Derivative
23 Analytical dn(x)/dx (x=5) = -20.665432423965154
24 Numerical dn(x)/dx (x=5) = -20.66543238754548
25
26 (d) probability distribution and the positions of satellites
27 Sampling distribution. See fig.3.
28 Positions of 100 satellites :
29 r =
30 [0.6946151470329265, 0.7310219267314602, 0.3040205414539694,
    0.25839441810422975, 1.027807936888672, 0.3030914676498381,
    1.1296044727104195, 0.1418091649404141, 0.5447139430869825,
    0.3594026879362262, 1.1172511297291998, 1.0791478707211926,
    0.7510739926994239, 0.17998057128504774, 1.3480897465759845,
    0.7272258440480012, 0.8081205175516876, 0.9496684840229044,
    0.6726386186291384, 0.6465613732093036, 1.778542642017886,
    1.0436059702924707, 0.2808786642008271, 1.0130994165806233,
    1.0801263491575153, 0.7635290007524946, 0.5919230567149196,
    0.43851557084167303, 0.28791892990364143, 1.0760334538853653,
    1.6454329057482096, 0.24594356054573324, 0.597127764080485,
    0.7060998865833868, 0.32809294305515946, 0.8641292598042791,
    0.5092276928025999, 0.9105267804243506, 0.9618478534298651,

```

```

0.6014645062654661, 1.4384635637646896, 0.2020068565018859,
0.42768909776691477, 1.2085595963048987, 0.4984193031408748,
0.6739956343270603, 0.6487096460301438, 1.623763295702238,
0.5347977552981961, 0.8659995818910158, 0.6538612037521938,
0.4144757132227146, 0.7553696866825521, 0.029187682490338013,
0.7506770093837914, 0.7638440221286465, 0.25415217392483763,
1.2304781456264542, 0.5056984695669674, 0.1655755809133595,
0.2088337850355816, 0.5507056711852045, 0.9886746874412782,
0.9082370879945934, 0.6752797823330347, 0.8229865457861786,
0.44685779279572513, 0.4900311743589316, 1.328554155744372,
0.24189085749769304, 0.33602242322477655, 0.16421531122654875,
0.38555691887346866, 1.1892608064204722, 0.37137481220979973,
0.7608554080057888, 0.17120241301041844, 0.23560776734213823,
0.22168045529867036, 1.0933155277764832, 1.3342786616653826,
1.272076192236514, 1.1047358637821365, 0.5829820949222099,
0.87676707631191733, 1.2223694826673444, 0.16535888752460376,
0.8010616800301091, 0.8906475979674384, 0.845788408640473,
0.07204278618255322, 0.7646080402389883, 0.47182057683858614,
0.2902623359212106, 0.8814734800710415, 0.9613805864733036,
0.23720991057638469, 0.541979052175256, 0.7844243523236174,
1.039293217475281]
31 phi =
32 [2.8461199 4.86275465 0.89879912 2.67828094 4.80432042 3.12891195
33 0.85379011 2.52657878 3.59534553 1.45479856 0.8386021 4.04904598
34 3.30001782 5.38044842 2.25022834 0.78996347 1.02857486 1.99422214
35 1.46929334 2.18146099 3.80306254 4.96048485 0.31809128 5.41717719
36 1.7207461 3.99708778 5.48434005 5.67413449 1.31722548 3.94216344
37 5.11182916 1.53104874 0.63673913 2.24014011 4.66422776 0.54118
38 2.70926995 2.15263591 3.36180745 5.83480251 3.16749966 5.92929208
39 3.44494492 0.49036601 4.15344513 0.13408845 4.40695424 3.44493658
40 6.18718287 3.93208215 2.73902472 5.74537343 0.28827685 3.52727101
41 2.88778487 5.73621852 2.87117533 4.72401578 5.3700832 5.34997662
42 1.38215181 3.64238664 0.44541052 5.61772423 2.74818297 5.56893001
43 5.55194144 1.14449953 6.08900042 0.68504765 2.3664095 1.44951018
44 0.20915526 2.86916328 0.05459971 4.07910868 2.42636114 2.20920038
45 5.29246299 0.63479329 1.23576997 0.57377959 5.88685435 2.81264825
46 5.53228425 2.60901237 3.5365992 2.36583808 4.06495589 2.87708493
47 1.80626858 4.03653864 0.29700442 0.92615063 1.58280142 1.85299593
48 4.63955936 2.71185457 4.97860458 1.14084621]
49 theta =
50 [2.46786272 0.29774471 1.87432966 1.24008588 2.75065107 3.02774209
51 1.94174344 2.31963649 0.29308535 2.59941275 2.91211977 1.63164754
52 1.16817586 0.04673397 2.12340846 2.96914783 1.64782274 1.0867605
53 0.84264676 1.598079 2.15326593 0.05491177 1.5447134 0.59505749
54 1.54631497 1.78498448 1.2762653 2.24139269 2.95418573 1.66764626
55 1.57949114 2.1579593 0.60161071 2.79614538 0.0388627 2.84953686
56 0.58134298 2.99508388 2.93018345 2.52870156 2.90360464 0.55186322
57 1.74383421 1.05584654 1.9974019 3.07352889 0.3624227 0.80993618
58 2.00608724 2.77597198 0.62793029 2.12536689 2.34123979 1.91473246
59 2.45125298 0.6128037 2.99320932 1.51300194 2.65007158 2.03684182
60 3.09714012 1.71116983 0.40763429 0.34074092 1.01865185 0.25836407
61 2.89326949 2.37725183 0.33753698 0.28277509 0.76286683 0.3494638
62 1.87392527 0.28267057 2.0942809 1.93551429 0.92519246 1.82554632
63 1.2100293 2.51243244 1.78229649 0.98187632 1.9018328 0.85187363
64 1.43117794 1.94088237 2.78843993 0.99906496 0.39967171 1.66934138
65 1.11630936 0.18233779 1.39821687 1.71619137 1.27651394 0.44921759
66 0.09591433 2.12243238 2.48963038]
67
68 (e) 1000 haloes and histogram
69 Histogram for 1000 haloes. See fig.4.
70
71 (f) rooting finding

```

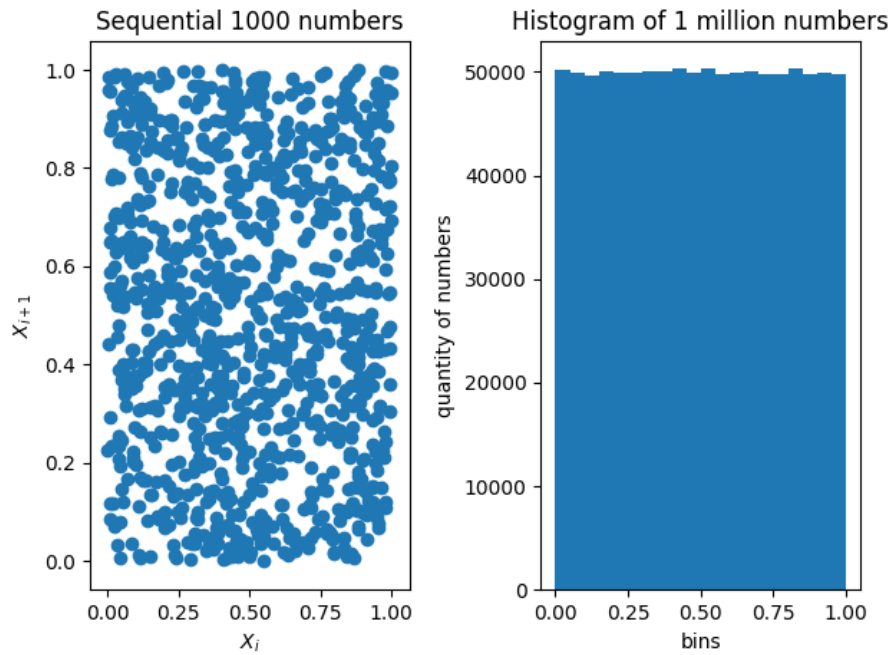


Figure 1: random number generator

```

72 Root_1 = 0.15418119373321532
73 Root_2 = 1.1334781646728516
74
75 (g) Histogram and Poisson distribution
76 median = 0.764393189628881
77 16th percentile = 0.6345954502983239
78 84th percentile = 0.9084267360075602
79 Histogram and distribution see fig.5
80
81 (h) A(a,b,c)
82 6240 values in A table
83 e.g. A(a=1.1,b=0.5,c=1.5) = 0.7673230946253291
84 test the interpolator function
85 e.g. A(a=1.27,b=1.27,c=2.27) = 0.05548288251663483
86
87 Question 3.
88 The code did not run as I had expected it to but I wrote the least
    -square fitting algorithm. Please find it in the source code
    section.

```

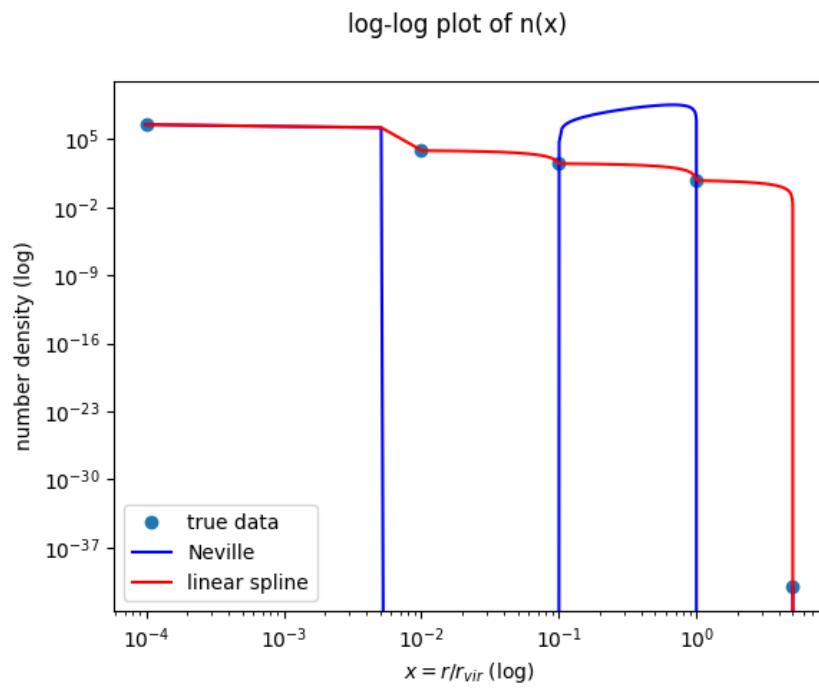


Figure 2: number density and interpolations

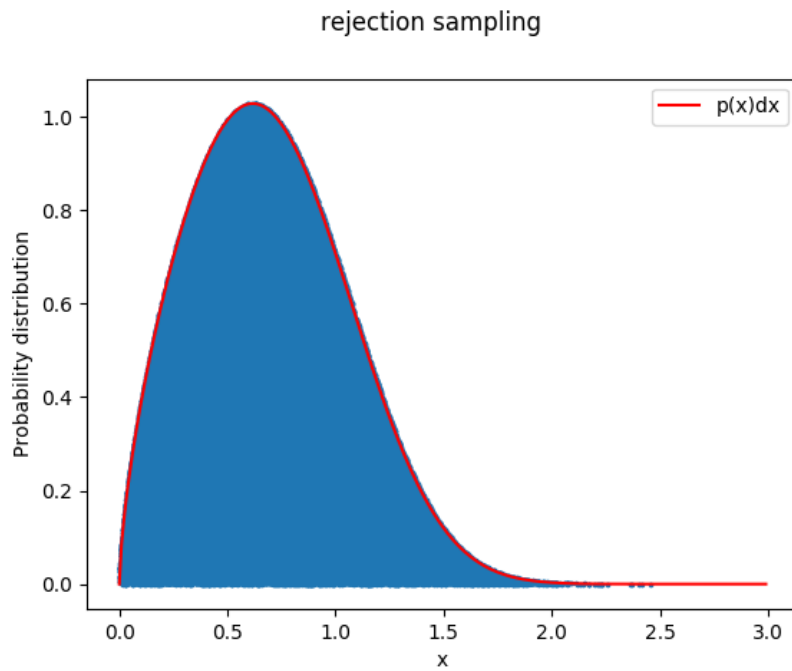


Figure 3: probability distribution

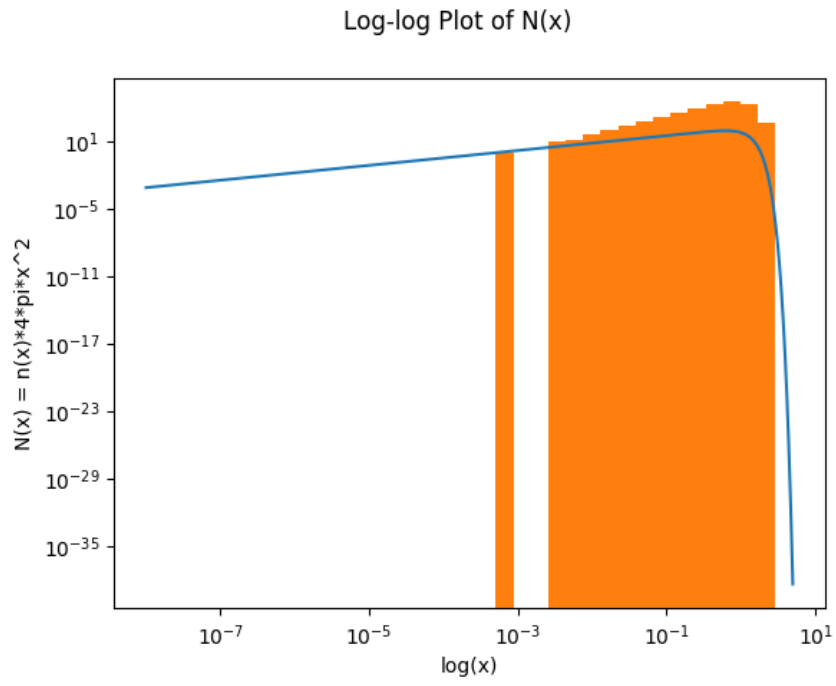


Figure 4: 1000 haloes and number of satellites in each bin

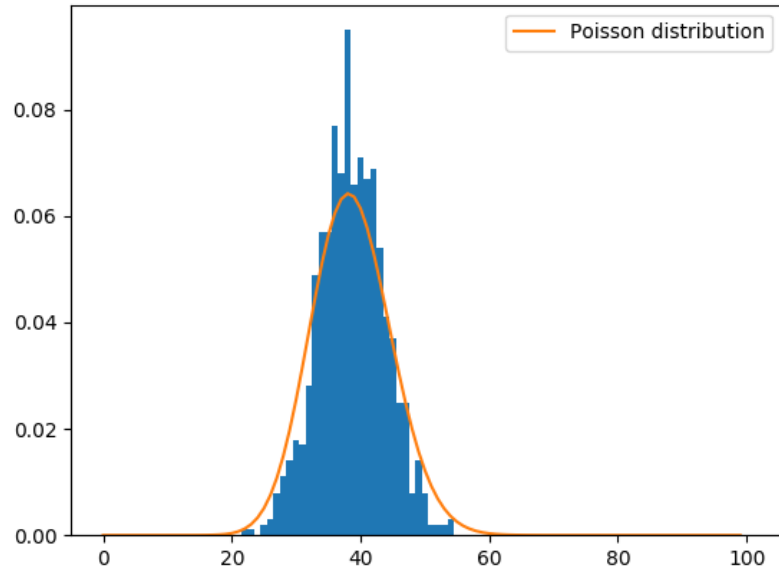


Figure 5: Poisson distribution