Computing for Physicists (PHYS2962) Test 1

Fill in this jupyter-notebook and upload <u>both</u> the jupyter-notebook and a pdf of it to the LMS at the end of time. (3:30)

1) The command line

For this problem, retrieve a file named "datatest.txt" from the remote computer with ip address 74.69.18.77. On your local machine, preprocess the file at the command line and plot the data.

(Note: the username is "class" and the password is "phys2962")

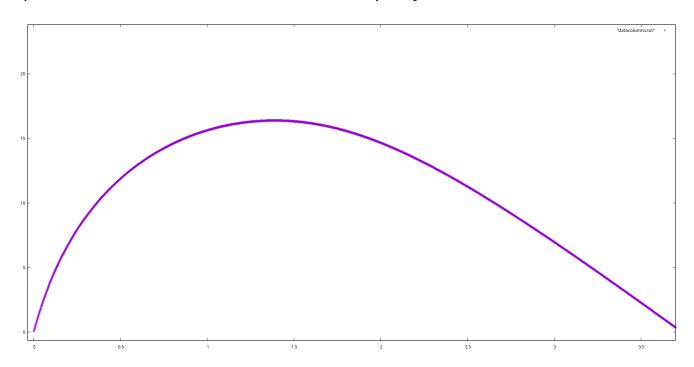
a) What command did you issue from the command prompt in order to copy the file to your local machine?

"" scp class@74.69.18.77:./a54/sdkfjhsdf/mary/a/little/lamb/arewethereyet/yes/datatest.txt data.txt

b) At the command line create a new file which contains two columns of numerical data (the y and t values). What command/s did you use?

``` cat data.txt | grep y= | gawk '{print 2,4}' >> datacolumns.txt

c) read in the data from the new file and plot y vs. t



## 2) numPy arrays

a) write code below to generate the following 3-dimensional array:

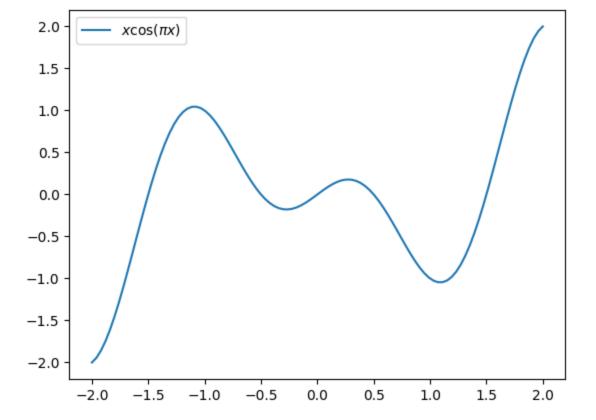
```
array([[[0, 4, 8],
 [12, 16, 20],
 [24, 28, 32],
 [36, 40, 44],
 [48, 52, 56],
 [60, 64, 68],
 [72, 76, 80],
 [84, 88,
 92],
 [96, 100, 104]],
 [[108, 112, 116],
 [120, 124, 128],
 [132, 136, 140],
 [144, 148, 152],
 [156, 160, 164],
 [168, 172, 176],
 [180, 184, 188],
 [192, 196, 200],
 [204, 208, 212]],
 [[216, 220, 224],
 [228, 232, 236],
 [240, 244, 248],
 [252, 256, 260],
 [264, 268, 272],
 [276, 280, 284],
 [288, 292, 296],
 [300, 304, 308],
 [312, 316, 320]]])
```

```
import numpy as np
arr = np.arange(0,321,4)
arr.reshape(3,9,3)
```

```
array([[[
 Θ,
 4,
 8],
Out[106]:
 16,
 [12,
 20],
 [24,
 28,
 32],
 [36,
 40,
 44],
 52,
 [48,
 56],
 [60,
 64,
 68],
 [72,
 76,
 80],
 [84,
 88,
 92],
 [96, 100, 104]],
 [[108, 112, 116],
 [120, 124, 128],
 [132, 136, 140],
 [144, 148, 152],
 [156, 160, 164],
 [168, 172, 176],
 [180, 184, 188],
 [192, 196, 200],
 [204, 208, 212]],
 [[216, 220, 224],
 [228, 232, 236],
 [240, 244, 248],
 [252, 256, 260],
 [264, 268, 272],
 [276, 280, 284],
 [288, 292, 296],
 [300, 304, 308],
 [312, 316, 320]]])
```

b) Plot  $xcos(\pi x)$  in the range of x=[-2,2]

```
import matplotlib.pyplot as plt
#Numpy imported from previous section
x = np.linspace(-2,2,100)
plt.plot(x, x*np.cos(np.pi*x), label = "$x \\cos(\\pi x)$")
plt.legend()
plt.show()
```



c) Generate an array of uniformly distributed random numbers in the range of [0,2) (generate enough so that you can do decent statistics  $\sim$  10,000). Calculate the mean and standard deviation of the dataset.

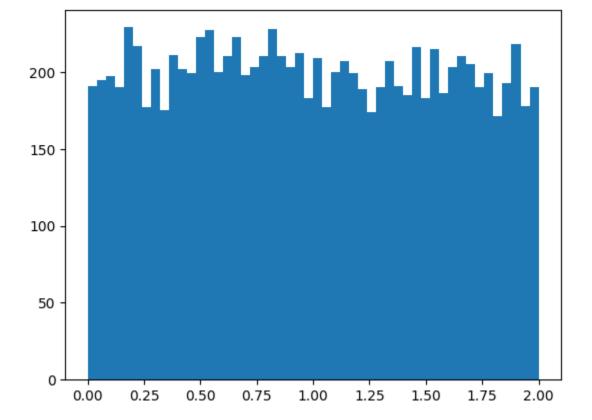
```
import random as rand
statarr = np.random.rand(10000) * 2
mean = np.mean(statarr)
print("Mean is: ", mean)
stddev = np.std(statarr)
print("Standard Deviation is:, ", stddev)

Mean is: 0.9906038963948719
```

Mean 1s: 0.9906038963948719 Standard Deviation is:, 0.5738970435849345

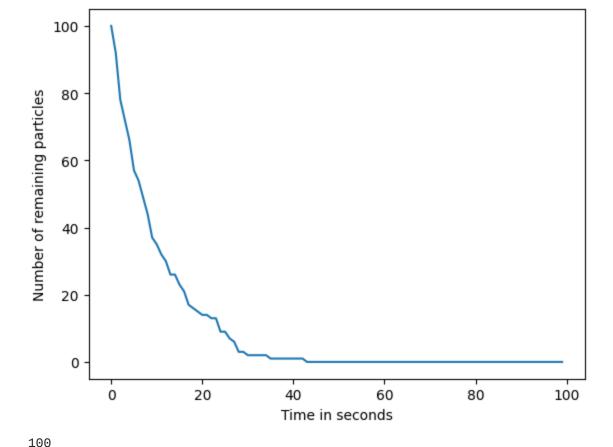
d) Display a histogram of the numbers generated in part (c)

```
In [109... plt.hist(statarr, bins=50)
 plt.show()
```



- 3) A Radioactive isotope has a 10% chance of decaying every second. Given a sample containing 100 atoms of such an isotope --
- a) Perform a Monte Carlo simulation to determine the Number of atoms (N) remaining as a function of time and plot it:

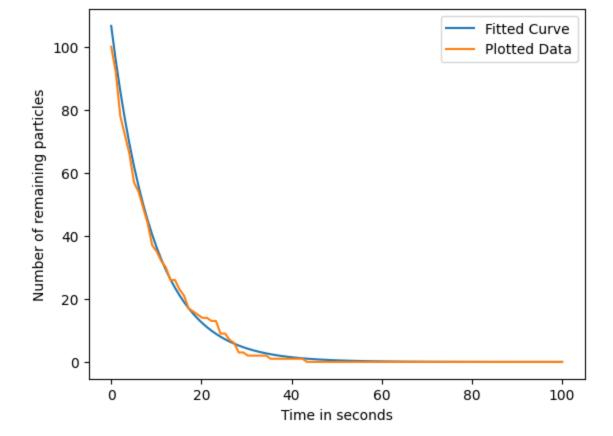
```
In [118... N = 100
Ns=[]
for i in range(100):
 Ns.append(N)
 for j in range(N):
 if(np.random.random() < 0.1): #This would be a very radioactive sample
 N = N-1
plt.ylabel("Number of remaining particles")
plt.xlabel("Time in seconds")
plt.plot(Ns)
plt.show()
print(len(Ns))</pre>
```



b) Fit your results from part a) to an exponential,  $N=N_0e^{-t/\tau}$ . What value of  $\tau$  do you obtain in your fit? Plot the result from (a) together with your exponential fit.

```
from scipy.odr import Model, RealData, ODR
In [119...
 def exponentialdecay(params,t):
 #params[0] is N_0, params[1] is tau
 return params[0] * np.exp(-t/params[1])
 def fit_exp_decay_odr(t_data, y_data):
 # Initial guess
 initial_guess = [100, 10]
 model = Model(exponentialdecay)
 data = RealData(t_data, y_data)
 odr = ODR(data, model, beta0=initial_guess)
 output = odr.run()
 params = output.beta
 y_fit = exponentialdecay(params, t_data)
 return params, y_fit
 t_{data} = np.linspace(0, 100, 100)
 Ns = np.array(Ns)
 params, yfit = fit_exp_decay_odr(t_data, Ns)
 print("Obtained value of tau is: ", params[1])
 plt.plot(t_data, exponentialdecay(params, t_data), label = "Fitted Curve")
 plt.ylabel("Number of remaining particles")
 plt.xlabel("Time in seconds")
 plt.plot(t_data,Ns, label="Plotted Data")
 plt.legend()
 plt.show()
```

Obtained value of tau is: 9.361994521114157



c) Perform a large number of realizations (100~1000) of the simulation in (a). What is the average time required for all atoms to decay? What is the standard deviation of the time required for all atoms to decay?

```
In [120...
 def monteCarloTime():
 N = 100
 Ns=[]
 for i in range(1000):
 Ns.append(N)
 for j in range(N):
 if(np.random.random() < 0.1): #This would be a very radioactive sample
 return Ns.index(0)
 fulldecaytime = []
 for i in range(1000):
 time = monteCarloTime()
 fulldecaytime.append(time)
 avgdecaytime = np.mean(fulldecaytime)
 stddecaytime = np.std(fulldecaytime)
 print("Average time = ", avgdecaytime, " Standard Devation of time: " , stddecaytime)
 Average time = 49.884 Standard Devation of time: 12.144568497892381
```

## 4) $\chi^2$ vs. least squares fit

a) Perform a least squares fit of the following data to the function  $y(x)=a+bx^2.$  Indicate the fitted values of a and b.

```
0.50505051,
 0.60606061,
 0.70707071,
 0.80808081,
 0.90909091
 1.01010101,
 1.11111111,
 1.21212121, 1.31313131,
 1.41414141,
 1.51515152, 1.61616162, 1.71717172, 1.81818182, 1.91919192,
 2.02020202,
 2.12121212,
 2.2222222, 2.32323232, 2.42424242,
 2.92929293,
 2.52525253,
 2.62626263,
 2.72727273, 2.82828283,
 3.13131313, 3.23232323, 3.33333333, 3.43434343,
 3.03030303,
 3.53535354,
 3.63636364,
 3.73737374, 3.83838384, 3.93939394,
 4.14141414,
 4.24242424, 4.34343434, 4.44444444,
 4.04040404,
 4.54545455,
 4.64646465, 4.74747475, 4.84848485,
 4.94949495,
 5.05050505,
 5.15151515, 5.25252525, 5.35353535, 5.45454545,
 5.5555556,
 5.65656566,
 5.75757576, 5.85858586, 5.95959596,
 6.26262626, 6.36363636, 6.46464646,
 6.06060606,
 6.16161616,
 6.76767677, 6.86868687, 6.96969697,
 6.56565657, 6.66666667,
 7.27272727, 7.37373737, 7.47474747,
 7.07070707, 7.17171717,
 7.7777778, 7.87878788, 7.97979798,
 7.67676768,
 7.57575758,
 8.08080808,
 8.18181818,
 8.28282828, 8.38383838,
 8.48484848,
 8.58585859,
 8.68686869, 8.78787879, 8.88888889, 8.98989899,
 9.09090909,
 9.19191919,
 9.29292929, 9.39393939, 9.49494949,
 9.6969697 , 9.7979798 , 9.8989899 , 10.
 9.5959596 ,
y=np.array([1.03545807,
 1.28195072, 1.00839038, 1.36913201,
 1.3169403 ,
 0.62679988, 1.00034059,
 0.92239042,
 1.59441357,
 2.97854407,
 3.06619471,
 1.71110414,
 3.08851556,
 3.16706311,
 2.72175075,
 3.66098392,
 4.19000385,
 5.07173457,
 6.689527
 5.06361123,
 7.86424009,
 5.56925392,
 6.15908147,
 5.62465897,
 8.02694204, 10.48898249,
 6.7803182 ,
 6.20675219,
 7.87006884,
 6.83874986, 11.95823263, 10.55901979,
 11.90532882, 14.07415747, 12.76850834, 15.57531864,
 14.94107618, 14.00150556, 15.13171867, 17.00442417,
 19.1028683 , 14.24767362, 21.51084622, 22.8089485 ,
 22.62978242, 23.56177019, 21.95860918, 23.66265132,
 23.74949617, 32.01727589, 29.39263196, 24.78853825,
 25.96335042, 30.36817615, 29.36061607, 30.03992341,
 34.8370933 , 34.57508206, 40.12515065, 36.80525687,
 40.54987719, 39.32043583, 42.95524361, 45.56408584,
 40.6492202 , 41.96250039, 51.33449992, 41.71904604,
 49.11857925, 41.52333409, 52.78654852, 46.61854518,
 53.16282846, 56.56414651, 58.16278177, 53.98777614,
 61.49677912, 65.5494806 , 57.2735989 , 55.9342531 ,
 68.70554848, 65.17879191, 74.70959111, 72.26875598,
 82.45815245, 80.07970789, 69.38613556, 77.99281414,
 76.15075653, 83.1599349, 85.55032887, 88.00218236,
 90.41702902, 102.15919646, 85.09058058, 96.16198339,
 103.92010228, 94.48636688, 103.10764057, 106.06465354])
 , 0.15050505, 0.2010101 , 0.25151515, 0.3020202 ,
errorbars = np.array([0.1
 0.35252525, 0.4030303 , 0.45353535, 0.5040404 , 0.55454545,
 0.60505051, 0.65555556, 0.70606061, 0.75656566, 0.80707071,
 0.85757576, 0.90808081, 0.95858586, 1.00909091, 1.05959596,
 1.11010101, 1.16060606, 1.21111111, 1.26161616, 1.31212121,
 1.36262626, 1.41313131, 1.46363636, 1.51414141, 1.56464646,
 1.61515152, 1.66565657, 1.71616162, 1.76666667, 1.81717172,
 1.86767677, 1.91818182, 1.96868687, 2.01919192, 2.06969697,
 2.12020202, 2.17070707, 2.22121212, 2.27171717, 2.32222222,
 2.37272727, 2.42323232, 2.47373737, 2.52424242, 2.57474747,
 2.62525253, 2.67575758, 2.72626263, 2.77676768, 2.82727273,
 2.87777778, 2.92828283, 2.97878788, 3.02929293, 3.07979798,
 3.13030303, 3.18080808, 3.23131313, 3.28181818, 3.33232323,
 3.38282828, 3.43333333, 3.48383838, 3.53434343, 3.58484848,
 3.63535354, 3.68585859, 3.73636364, 3.78686869, 3.83737374,
 3.88787879, 3.93838384, 3.98888889, 4.03939394, 4.08989899,
 4.14040404, 4.19090909, 4.24141414, 4.29191919, 4.34242424,
 4.39292929, 4.44343434, 4.49393939, 4.54444444, 4.59494949,
 4.64545455, 4.6959596 , 4.74646465, 4.7969697 , 4.84747475,
 4.8979798 , 4.94848485, 4.9989899 , 5.04949495, 5.1
```

```
import scipy as sp
def drwestsfunction(params,x):
 return params[0] + params[1] * x ** 2

paramslsq = sp.optimize.curve_fit(drwestsfunction, x,y)
print("A = ", paramslsq[0], " B = " , paramslsq[1])

A = [18.6243544] B = [[0.67452297]]
```

b) Perform a  $\chi^2$  fit of the same data. What values of a and b do you obtain?

```
In [123... paramschisq = sp.optimize.curve_fit(drwestsfunction, x,y, sigma=errorbars) print("A = ", paramschisq[0], " B = " , paramschisq[1])
A = [4.93403075] B = [[0.50848438]]
```

c) Plot the original data (with error bars) in black, your least-squares fit function in red, and your  $\chi^2$  fit function in green all in the same figure.

```
plt.errorbar(x,y,errorbars, color= "black", ecolor="grey")
plt.plot(x,drwestsfunction(paramslsq,x)[0], label= "Least Squares Regression", color = "
plt.plot(x,drwestsfunction(paramschisq,x)[0], label = "Chi Squared Regression", color = plt.xlabel("X")
plt.ylabel("Y")
plt.show()
```

