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In []: #Importing relevant modules for later as well as specifying directory
   import astropy
   import pandas as pd
   import numpy as np
   import matplotlib as plt
   from matplotlib import figure
   import matplotlib as plt
   import os
   import photutils
   import astroplan
   from astropy.io import ascii

# Set up matplotlib
%matplotlib inline
```

Below shows a bootstrapping technique to produce 95% upper and lower confidence intervals associated with a supernovae rate density estimation. Df refers to a pandas dataframe which contains the required values. The dataset should be taken from the BTS as a csv which is later turned into a dataframe so will contain the required headers and values such as peakabs which refers to the peak absolute magnitude a supernova was observed to have. T refers to the time over which the survey was taken, for example if the sample had supernovae that peaked bewteen 200 to 1500 days then this would yield a T = 3.56 years. n refers to a range of values of absolute magnitudes based on what was found within a dataset.

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In [ ]: Df = pd.read csv("") #Enter csv filename containing supernovae data
In [ ]: def bootstrap(n,Df,T):
          collective cumulative rates = []
           for i in range(0,4000):
              resampled data = resample(Df, n samples = len(Df) , replace = True, stratify = Non
              sliced data = resampled data.drop(resampled data[resampled data.peakabs>n].index
              #Dataset now left with data cut off by a particular apparent magnitude
              cumulative rate = np.sum(sliced data.rind)/T # Rate formula
              collective cumulative rates append (cumulative rate) #All 4000 cumulative rates c
           conf int = np.percentile(collective cumulative rates,[2.5,97.5]) # 95% Upper and Low
           print(conf int)
       # for n in np.arange(-16.5, -20.6, -0.005): <----- example implentation, ensures
       # bootstrap Ia(n,18.5)
In [ ]: def Rind PlotIa(Df, mlim, T, f rec, f ext, f cl, f sky): #Dataframe, limiting apparent Magnitud
           Df["peakabs"] = pd.to numeric(Df["peakabs"], downcast="float")
           f = f rec*f ext*f cl*f sky
           Mag1 = np.arange(-16.5, -20.6, -0.005) # Absolute Magnitude range
           muA = mlim - Df.peakabs #Essentially mu in the distance modulus formula (m-M)
           mu2A = 0.2*(muA+5-Df.A V)
           dA= pow(10, mu2A) #Distance modulus
           dGpcA = dA*(1E-9) #Conversion Parsec to GigaParsec
           VolA = (4/3)*np.pi*pow(dGpcA,3) #Volume
           IVA = 1/VolA # Invese of Volume
           RindA = IVA*(1/f) #Individual Rate
           Df["RindA"] = RindA #Adding these individual Rates to Dataframe
           CumSumA = [] #Cumulative Rates
           for n in Mag1:
              yA = Df.drop(Df[Df.peakabs>n].index) #Cutting the dataframe by absolute magnitud
```

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bA = np.sum(yA.RindA)/T #Rate Formula
    CumSumA.append(bA) #Cumulative Rates appended
########## Bootstrapping
#Add the dataset from the saved upper and lower 95% confidence intervals as a datafr
#Then use this format to plot:
# plt.pyplot.fill between(Mag1,Lower95%,Upper95%,label = "Title")
############################ Poisson Upper Limit
PLOTTING THINGS
plt.rcParams["figure.figsize"] = [10, 10]
plt.rcParams["figure.autolayout"] = True
plt.pyplot.figure()
plt.pyplot.plot(Mag1,CumSumA,color = "black",label = "SN Ia rate")
plt.pyplot.gca().invert xaxis()
plt.pyplot.legend(loc = "lower left")
manager = plt.pyplot.get current fig manager()
manager.full_screen_toggle()
plt.pyplot.yscale("log")
#plt.pyplot.ylim(0,1664)
plt.pyplot.xlim(-16,-20.7)
plt.pyplot.xlabel("Limiting Absolute Magnitude")
plt.pyplot.ylabel("Cumulative Sne Rate ($Gpc^{-3}$ $yr^{-1}$)")
#plt.pyplot.axhline(y=5.12E+02,color = "black")
\#plt.pyplot.text(x = -18, y = 1.0E + 03, s = r"SN Ia - 91T Rate = $5.12 \setminus times 10^{2} \setminus Gp
fig = plt.pyplot.gcf()
fig.set size inches(8, 7)
plt.pyplot.show(block = True)
print("SN Rate Density is:", format(max(CumSumA),".2E"))
```