```
from scipy.integrate import quad
         #Nested sampling package
         import ultranest
         import corner
In [2]: | # Import Hubble H(z) data
        hubble_data = pd.read_csv('hubble_data.csv', header=0)
         z H = np.array(hubble data['z'])
         H = np.array(hubble data['H'])
         dH = np.array(hubble_data['dH'])
         plt.figure()
         plt.errorbar(z H, H, yerr=dH, marker = '.', color='blue', ecolor='black', capsize=2, ls='none')
         plt.ylabel(r'$H(z)$')
         plt.xlabel(r'$z$')
         plt.show()
         # Import apparent magnitude m(z) data
         m_data = pd.read_csv('m_data.txt', sep = ' ', header = 0)
         m sys unc = pd.read csv('m sys unc.txt', sep = ' ', header = 0)
        m_sys_unc = np.array(m_sys_unc['40']).reshape(40, 40)
         tot = m_sys_unc + np.diag(m_data['dmb']**2)
         z m = np.array(m data['zcmb'])
         m = np.array(m_data['mb'])
         dm = np.sqrt(np.diag(tot))
         plt.figure()
         plt.errorbar(z m, m, yerr=dm, marker = '.', color='blue', ecolor='black', capsize=2, ls='none')
         plt.ylabel(r' m(z) )
         plt.xlabel(r'$z$')
         plt.show()
         # Combine redshifts for likelihood computation later
         combined z = []
         combined z.append(z H)
         combined_z.append(z_m)
         # Combine data
         combined data = []
         combined_data.append(H)
         combined_data.append(m)
         # Combine uncertainties
         combined unc = []
         combined unc.append(dH)
         combined unc.append(dm)
           250
                                                      ł
           200
                       0.5
                               1.0
                                               2.0
           26
           24
           22
         (z)
E 20
           18
           16
              0.0
                   0.2
                        0.4
                            0.6
                                 0.8
                                      1.0
                                          1.2
                                               1.4
                                                    1.6
In [3]:
        'Define LCDM Hubble model'
         def LCDM(z, params):
            H0 = params[0]
            OM = params[1]
            OL = params[2]
             return H0*np.sqrt(OM*(1+z)**3 + OL)
         'Define Domain Walls Model'
         def DomainWalls(z, params):
            H0 = params[0]
            OM = params[1]
            OD = params[2]
            return H0*np.sqrt(OM*(1+z)**3 + OD*(1+z)**(1/3))
         'Define Cosmic Strings Model'
         def CosmicStrings(z, params):
            H0 = params[0]
            OM = params[1]
            OS = params[2]
            return H0*np.sqrt(OM*(1+z)**3 + OS*(1+z)**(1/3))
         'Define Phantom Energy Model'
         def PhantomEnergy(z, params):
            H0 = params[0]
            OM = params[1]
            OP = params[2]
            w_p = params[3]
             return H0*np.sqrt (OM* (1+z) **3 + OP* (1+z) ** (3* (1+w p)))
         'Define Inverse Monomial model'
         def InverseMonomial(z, params):
            H0 = params[0]
            B = params[1]
            return H0*np.sqrt((1/(1+B*z))*(1+z)**3)
         'Define Exponential Model'
         def Exponential(z, params):
            H0 = params[0]
            B = params[1]
            return H0*np.sqrt(np.exp(B*((1/(1+z))-1))*(1+z)**3)
         'Define Logarithmic Model'
         def LogarithmicModel(z, params):
            H0 = params[0]
             B = params[1]
             return H0*np.sqrt((B*np.log(1/(1+z)) + 1)*(1+z)**3)
         'Define backreaction Model'
         def backreaction(z, params):
            H0 = params[0]
            OM = params[1]
            n = params[2]
             return H0*np.sqrt(OM*((1+z)**3)+(1-OM)*((1+z)**n))
         'Define apparent magnitude function'
         def ApparentMagnitude(z, Hubble, params):
             def integrand dl(z, Hubble, params):
             #integrand of luminosity distance formula
                 return params[0]/Hubble(z, params)
             def dl(z, Hubble, params):
             #dimensionless luminosity distance at redshift z (input array)
                 rz array = np.zeros(len(z))
                 for i in np.arange(len(z)):
                     rz each = quad( integrand dl, 0, z[i],
                                 args = (Hubble, params))[0]
                     rz array[i] = rz each
                 return (1+z) *rz array
             return 5*np.log10((c*100/params[0])*dl(z, Hubble, params)) - 19.25
In [7]: params lcdm = [72.21, 0.26, 0.74]
         params domainwalls = [73.37, 0.22, 0.73]
         params cosmicstrings = [74.96, 0.21, 0.75]
         params phantomenergy = [74.50, 0.25, 0.73, -1.03]
         params InverseMonomial = [72.21, 1]
         params Exponential = [73.37, 1.5]
         params LogarithmicModel = [68.47, 0.6834 ]
        params_backreaction = [73.37, 0.22, 1]
         plt.figure()
         plt.errorbar(z_H, H, yerr=dH, marker = '.', color='blue', ecolor='black', capsize=2, ls='none')
        plt.plot(z_H, LCDM(z_H, params_lcdm), color='red', ls='--', label=r'$\Lambda$CDM')
         plt.plot(z H, DomainWalls(z H, params domainwalls), color='orange', label='Domain Walls')
        plt.plot(z H, CosmicStrings(z H, params cosmicstrings), color='green', ls='--', label='Cosmic Strings')
        plt.plot(z H, PhantomEnergy(z H, params phantomenergy), color='yellow', label='Phantom Energy')
        plt.plot(z_H, InverseMonomial(z_H, params_InverseMonomial), color='blue', ls='--', label='InverseMonomi
        al')
         plt.plot(z H, Exponential(z H, params Exponential), color='brown', label='Exponential')
         plt.plot(z H, LogarithmicModel(z H, params LogarithmicModel), color='pink', ls='--', label='Logarithmic
         plt.plot(z H, backreaction(z H, params backreaction), color='purple', label='Back Reaction')
         plt.legend(loc='best')
        plt.ylabel(r'$H(z)$')
        plt.xlabel(r'$z$')
         plt.savefig('H(z) vs z.pdf')
         plt.show()
           250
           200
         ¥ 150
                                            Cosmic Strings
                                            Phantom Energy
           100
                                            InverseMonomial
                                            Exponential
                                            LogarithmicModel
                                            Back Reaction
                                               2.0
              0.0
                       0.5
                               1.0
In [8]: plt.figure()
        plt.errorbar(z m, m, yerr=dm, marker = '.', color='blue', ecolor='black', capsize=2, ls='none')
         z m2 = np.linspace(0.01, 2, 100)
        plt.plot(z_m2, ApparentMagnitude(z_m2, DomainWalls, params_domainwalls), color='orange', label='Domain
         Walls')
         plt.plot(z m2, ApparentMagnitude(z m2, LCDM, params lcdm), color='red', ls='--', label=r'$\Lambda$CDM')
         plt.plot(z m2, ApparentMagnitude(z m2, PhantomEnergy, params phantomenergy), color='green', label='Phan
         tom Energy')
        plt.plot(z_m2, ApparentMagnitude(z_m2, CosmicStrings, params_cosmicstrings), color='yellow', ls='--', l
         abel='Cosmic Strings')
        plt.plot(z m2, ApparentMagnitude(z m2, InverseMonomial, params InverseMonomial), color='blue', label='I
        nverse Monomial')
         plt.plot(z_m2, ApparentMagnitude(z_m2, Exponential, params_Exponential), color='brown', ls='--', label=
         'Exponential')
         plt.plot(z_m2, ApparentMagnitude(z_m2, LogarithmicModel, params_LogarithmicModel), color='pink', label=
         'LogarithmicModel')
         plt.plot(z_m2, ApparentMagnitude(z_m2, backreaction, params_backreaction), color='purple', ls='--', lab
         el='Back Reaction')
         plt.legend(loc='best')
        plt.ylabel(r'$m(z)$')
         plt.xlabel(r'$z$')
         plt.savefig('m(z) vs z.pdf')
```

plt.show()

26

24

22

18

16

14

In [9]: plt.figure()

0.00

tom Energy')

0.25

abel='Cosmic Strings')

nverse Monomial')

'LogarithmicModel')

el='Back Reaction')
plt.legend(loc='best')

plt.ylabel(r'\$m(z)\$')
plt.xlabel(r'\$z\$')

plt.savefig('m(z) vs z zoomed.pdf')

1.0

1.2

0.8

plt.xlim(0.6,1.6) plt.ylim(23,26)

plt.show()

26.0

25.5

25.0

24.0

23.5

23.0

0.6

(E) 24.5

'Exponential')

0.50 0.75 1.00

(Z) 20 A STATE OF THE PARTY OF THE PAR

Domain Walls

Phantom Energy Cosmic Strings

Inverse Monomial

LogarithmicModel

Domain Walls

Phantom Energy Cosmic Strings

Inverse Monomial Exponential

LogarithmicModel Back Reaction

1.6

1.4

ΛCDM

plt.errorbar(z m, m, yerr=dm, marker = '.', color='blue', ecolor='black', capsize=2, ls='none')

plt.plot(z\_m2, ApparentMagnitude(z\_m2, DomainWalls, params\_domainwalls), color='orange', label='Domain

plt.plot(z\_m2, ApparentMagnitude(z\_m2, LCDM, params\_lcdm), color='red', ls='--', label=r'\$\Lambda\$CDM') plt.plot(z m2, ApparentMagnitude(z m2, PhantomEnergy, params phantomenergy), color='green', label='Phan

plt.plot(z\_m2, ApparentMagnitude(z\_m2, CosmicStrings, params\_cosmicstrings), color='yellow', ls='--', l

plt.plot(z m2, ApparentMagnitude(z m2, InverseMonomial, params InverseMonomial), color='blue', label='I

plt.plot(z\_m2, ApparentMagnitude(z\_m2, Exponential, params\_Exponential), color='brown', ls='--', label=

plt.plot(z\_m2, ApparentMagnitude(z\_m2, LogarithmicModel, params\_LogarithmicModel), color='pink', label=

plt.plot(z m2, ApparentMagnitude(z m2, backreaction, params backreaction), color='purple', ls='--', lab

ACDM

Exponential

--- Back Reaction

1.25 1.50 1.75 2.00

In [1]: import numpy as np

import pandas as pd

import matplotlib.pyplot as plt

from scipy.constants import c