Kernel

Widgets

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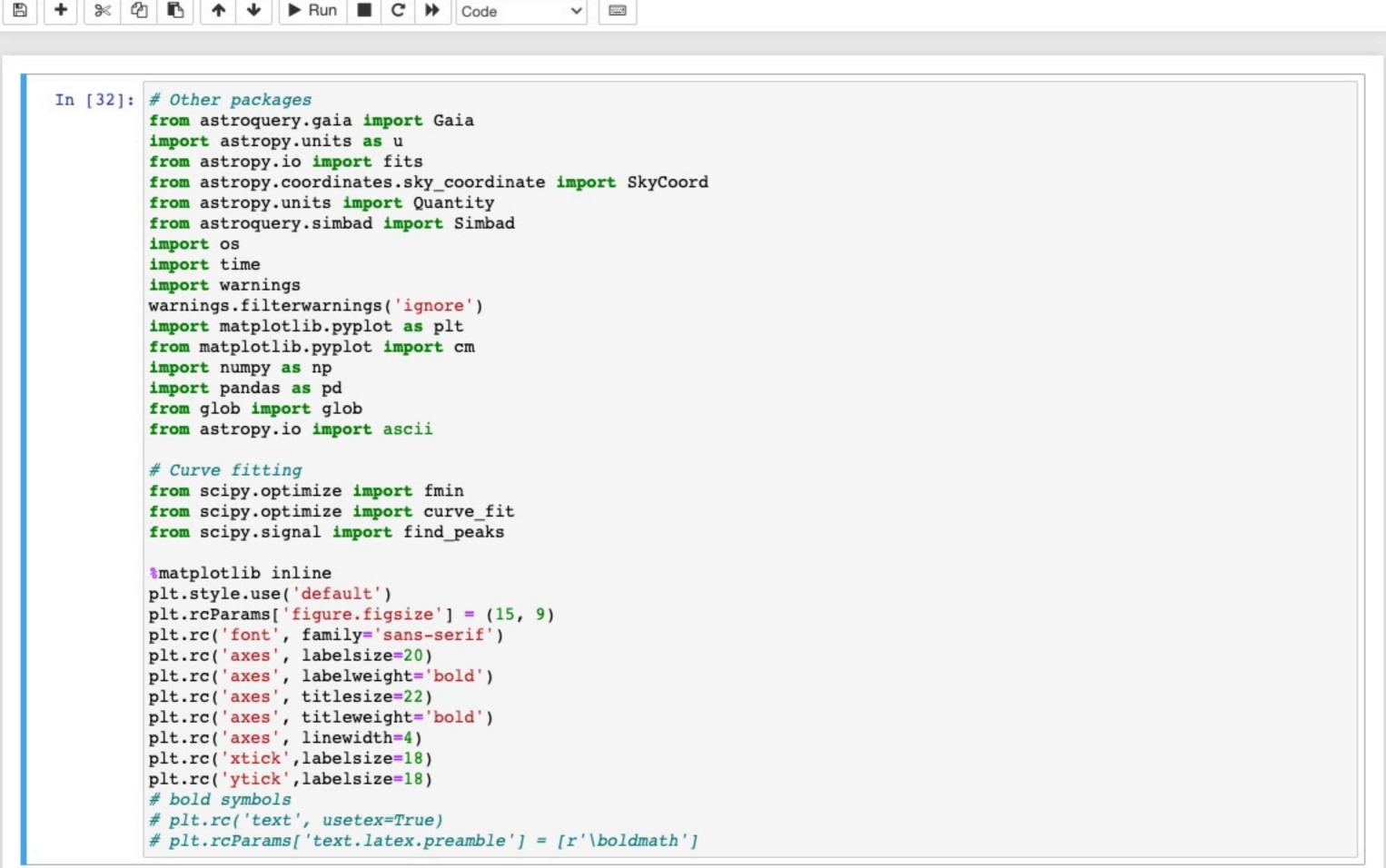


Not Trusted

25

Logout

Python 3 O



Problem 1

```
In [17]: bc03_models = ascii.read('bc03_models.txt')
bc03_models
```

Out[17]: Table length=4771

Edit

View

File

WAVE float64	LUM1 float64	LUM10 float64	LUM100 float64	LUM1000 float64	LUM5000 float64	LUM10000 float64
3400.78	0.0311214	0.012052	0.00102177	0.000107551	1.4226e-05	6.75832e-06
3401.57	0.0312522	0.0122631	0.00101318	0.000103973	1.39435e-05	6.62804e-06
3402.35	0.0313686	0.0123845	0.00101097	9.99131e-05	1.35252e-05	6.41025e-06
3403.13	0.0313466	0.0123375	0.00101234	9.61526e-05	1.30177e-05	6.15075e-06
3403.92	0.0312484	0.012252	0.00102121	9.31796e-05	1.2489e-05	5.96055e-06
3404.7	0.0312424	0.0121545	0.00102427	9.16673e-05	1.21851e-05	5.81448e-06
3405.48	0.0312688	0.0120859	0.00102324	9.14053e-05	1.20233e-05	5.67934e-06
3406.27	0.0312632	0.0120198	0.00101669	9.15303e-05	1.19122e-05	5.55208e-06
3407.05	0.0311945	0.0119456	0.00100434	9.12223e-05	1.17664e-05	5.4287e-06
10176.0	0.000659013	0.00241996	0.000248923	8 75406e-05	2.98996e-05	1.94551e-05

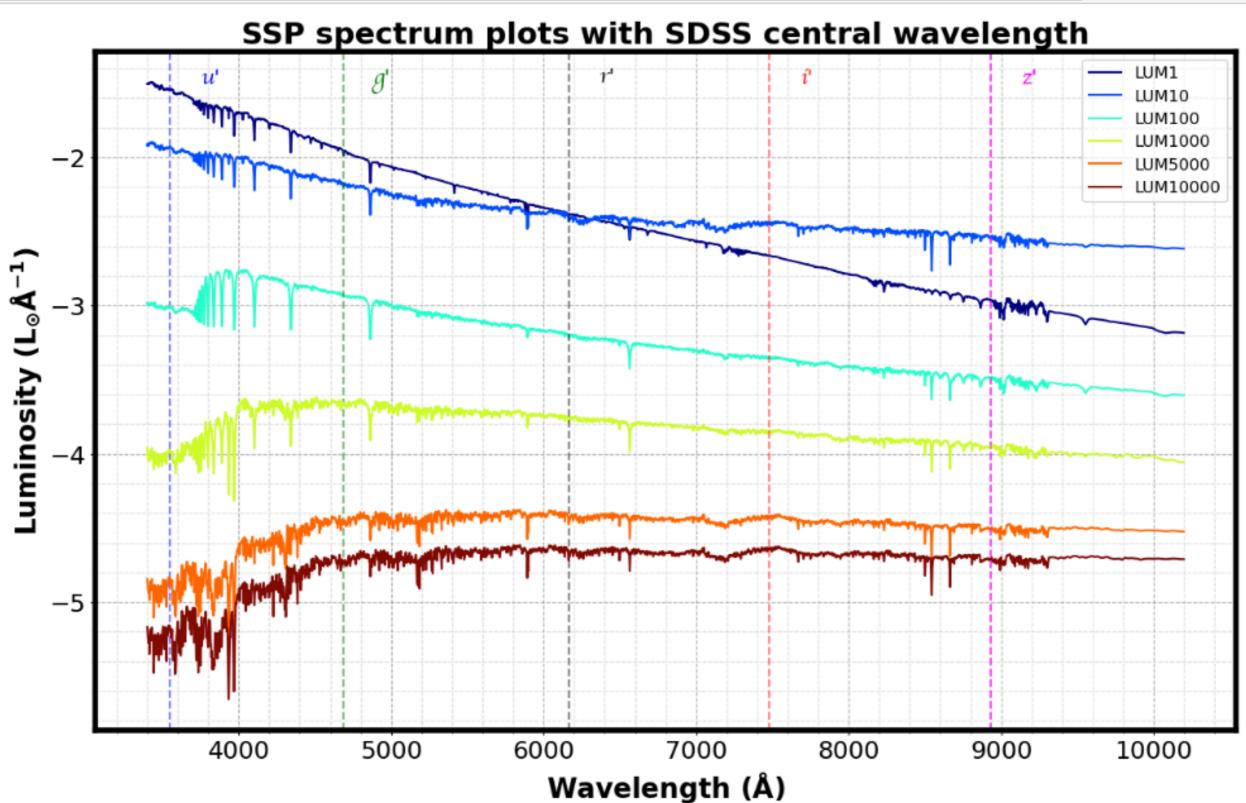
```
10176.0 0.000659013 0.00241996 0.000248923 8.75406e-05 2.98996e-05 1.94551e-05
10178.4 | 0.000658759 | 0.00241903 | 0.000248798 | 8.74202e-05 | 2.98874e-05 | 1.94485e-05
10180.7
          0.0006582 | 0.00241882 | 0.000248743 | 8.74602e-05
                                                             2.989e-05
                                                                        1.9452e-05
10183.1
        0.000657646 0.00241862 0.000248689
                                              8.74999e-05 2.98925e-05 1.94555e-05
                                              8.75399e-05
10185.4 0.000657087 0.00241841 0.000248634
                                                            2.9895e-05
                                                                        1.9459e-05
10187.8 0.000656522
                      0.0024182 0.000248579
                                              8.75804e-05 2.98976e-05 1.94626e-05
10190.1 0.000655969 0.00241799 0.000248525
                                              8.76201e-05 | 2.99001e-05 | 1.94661e-05
10192.5 | 0.000655404 | 0.00241778
                                              8.76605e-05 2.99026e-05 1.94696e-05
                                  0.00024847
10194.8 | 0.000654845 | 0.00241758 | 0.000248415 | 8.77006e-05 | 2.99051e-05 | 1.94731e-05
10197.2 0.000654286 0.00241737
                                  0.00024836 8.77407e-05 2.99077e-05 1.94766e-05
```

```
In [18]: filter_curves = fits.open('filter_curves.fits')
# print(filter_curves.info())
# filter_curves[0].header
# u_filt = filter_curves['U'].data
# # u_filt
# age_model = bc03_models.colnames[1:]
```

Central SDSS wavelength obtained from: SDSS-III Filters

Filter	Central Wavelength
u	3551 Å
g	4686 Å
r	6166 Å
i	7480 Å
z	8932 Å

```
In [23]: age_model = bc03_models.colnames[1:]
         SDSS_wav = [3551, 4686, 6166, 7480, 8932]
         colors = ['blue', 'green', 'black', 'red', 'magenta']
         filter_names = ["u'", "g'", "r'", "i'", "z'"]
         color2=iter(cm.jet(np.linspace(0,1,len(age_model))))
         for i in range(0, len(age_model)):
             c=next(color2)
             plt.plot(bc03_models['WAVE'], np.log10(bc03_models[age_model[i]]), label=age_model[i], color = c)
             plt.axvline(SDSS_wav[i-1], ls = '--', color=colors[i-1], alpha=0.5)
             plt.title('SSP spectrum plots with SDSS central wavelength')
             plt.xlabel('Wavelength $\mathbf{(\AA)}$')
             plt.ylabel('Luminosity $\mathbf{(L_{\odot}\AA^{-1}))$')
             plt.text(SDSS_wav[i-1]+200, -1.5, filter_names[i-1], fontfamily='cursive', fontsize='xx-large', fontweight='bold',
             plt.legend(fontsize=12)
             plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
             plt.minorticks_on()
             plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1)
```



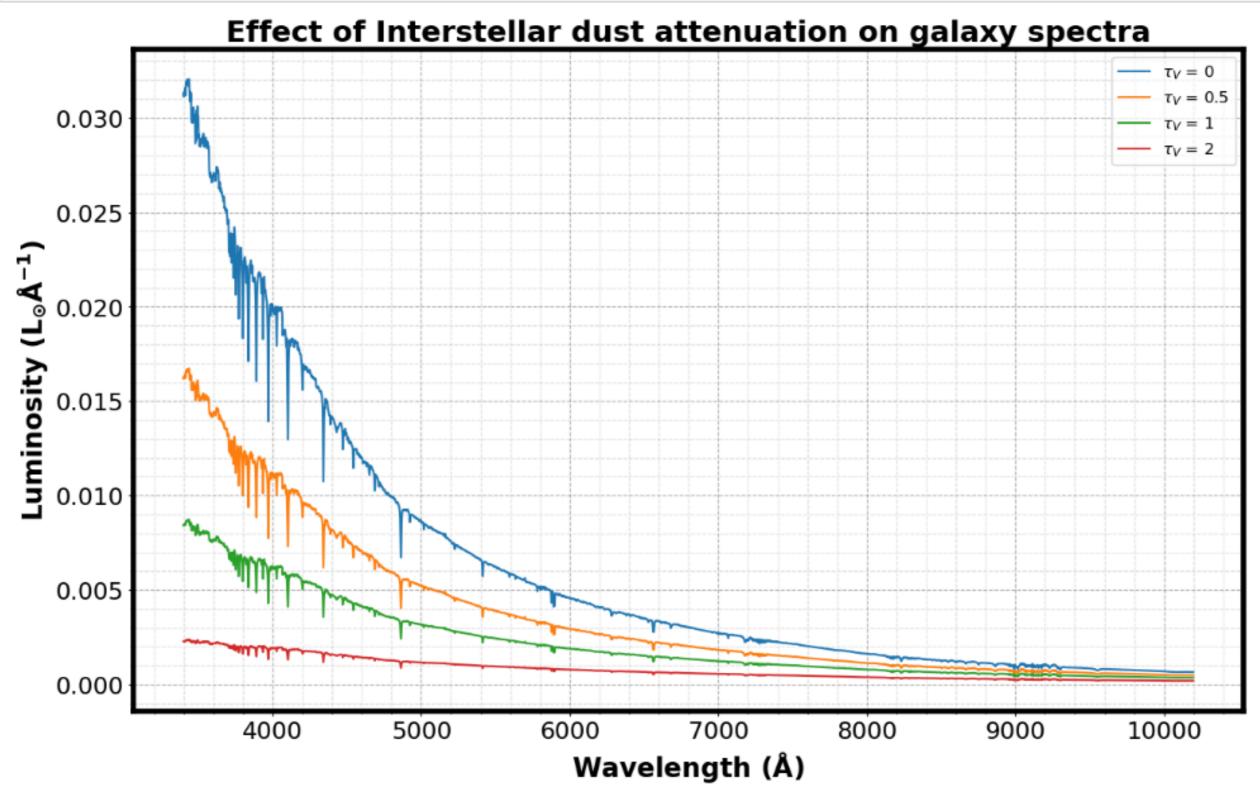
```
In [24]:

def L_dusty(tau_V, L_0, wave):
    tau_wav = tau_V * (wave/5000)**(-0.7)
    return L_0 * np.exp(-tau_wav)

for i in (0, 0.5, 1, 2):
    LUM1_dusty = L_dusty(i, bc03_models[age_model[0]], bc03_models['WAVE'])
    plt.plot(bc03_models['WAVE'], LUM1_dusty, label=r'$\tau_V$ = {}'.format(i))
    plt.title('Effect of Interstellar dust attenuation on galaxy spectra')
    plt.xlabel('Wavelength $\mathbf{(\AA)}$')
    plt.ylabel('Luminosity $\mathbf{(\L_{\Odot}\AA^{-1}))}$')

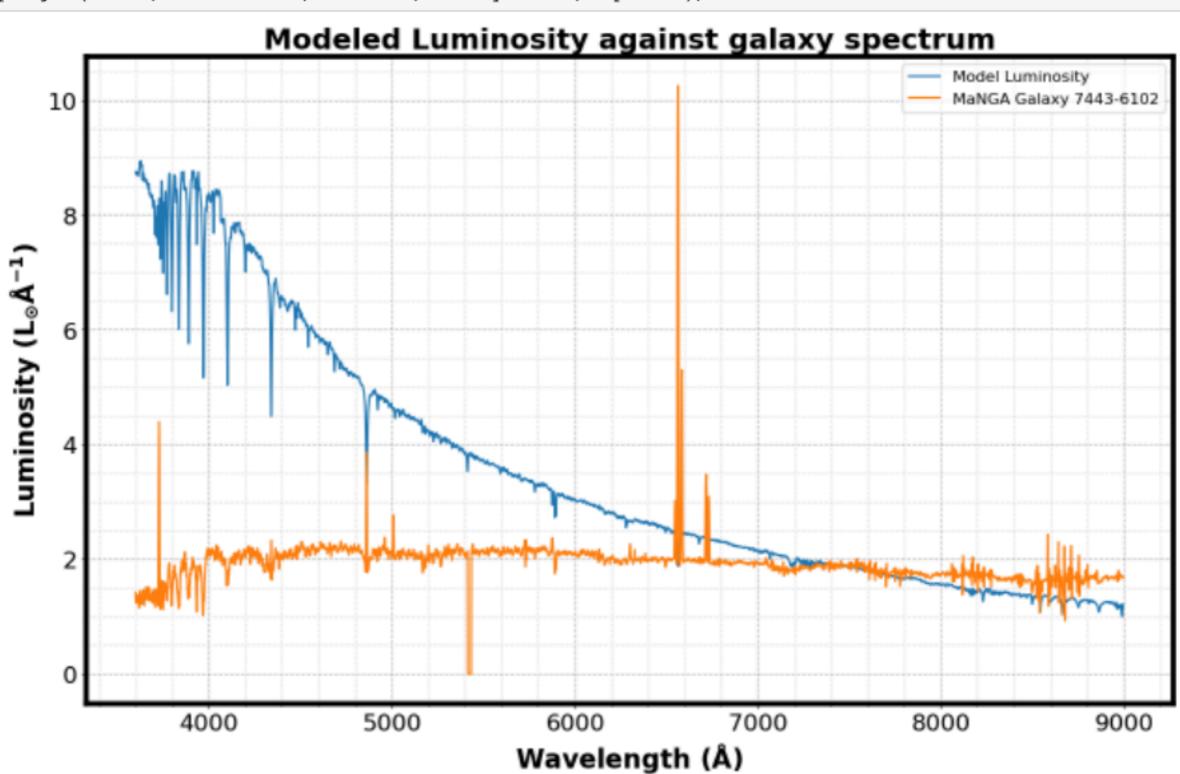
# plt.text(SDSS_wav[i-1]+200, -1.5, filter_names[i-1], fontfamily='cursive', fontsize='xx-large', fontweight='bold'

plt.legend(fontsize=12)
    plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
    plt.minorticks_on()
    plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1)
```



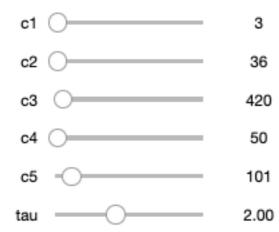
Problem 2

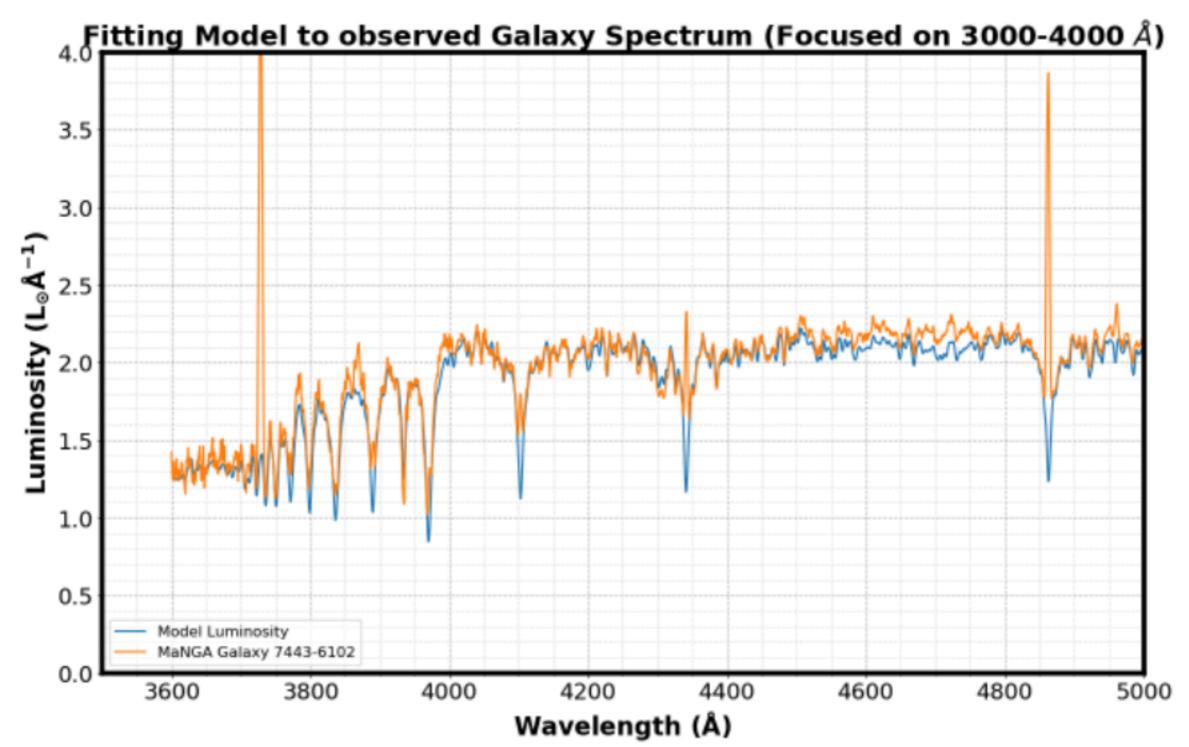
```
In [6]: from ipywidgets import interact, interactive, fixed, interact_manual, FloatSlider
         import ipywidgets as widgets
         import matplotlib.pyplot as plt, random
         from IPython.display import display
In [7]: manga_7443_6102 = ascii.read('manga_7443-6102.txt')
         # manga_7443_6102
         # Removing the wavelength offset
         bc03_offset = bc03_models[(bc03_models['WAVE']>=min(manga_7443_6102['WAVE'])) & (bc03_models['WAVE']<=max(manga_7443_61
In [40]: def L model(c1, c2, c3, c4, c5, tau v):
             model = ((c1*bc03_offset['LUM1'] + c2*bc03_offset['LUM100']
                      + c3*bc03_offset['LUM1000'] + c4*bc03 offset['LUM10000'])
                      * (c5*np.exp(-tau_v *(bc03_offset['WAVE']/5000)**(-0.7))
             return model
         plt.plot(bc03_offset['WAVE'], L_model(50,200,10,20,20,1), label='Model Luminosity')
         plt.plot(manga 7443 6102['WAVE'], manga 7443 6102['LUMINOSITY'], label='Manga Galaxy 7443-6102')
         plt.title('Modeled Luminosity against galaxy spectrum')
         plt.xlabel('Wavelength $\mathbf{(\AA)}$')
         plt.ylabel('Luminosity $\mathbf{(L_{\odot}\AA^{-1}))$')
         plt.legend(fontsize=12)
         plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
         plt.minorticks_on()
         plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1);
```



```
In [86]:

def series(c1, c2, c3, c4, c5, tau):
    plt.plot(bc03_offset('WAVE'), L_model(c1,c2,c3,c4,c5,tau), label='Model Luminosity')
    plt.plot(manga_7443_6102['WAVE'], manga_7443_6102['LUMINOSITY'], label='ManGA Galaxy 7443-6102')
    plt.title(r'Fitting Model to observed Galaxy Spectrum (Focused on 3000-4000 $\AA$)')
    plt.xlabel('Wavelength $\mathbf{(\AA})$')
    plt.ylabel('Luminosity $\mathbf{(\AA})$')
    plt.legend(fontsize=12)
    plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
    plt.minorticks_on()
    plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1)
    plt.xlim(3500, 5000)
    plt.ylim(0,4)
    plt.show();
    return()
    interact(series, c1 = (1,1000,1), c2 = (1,10000,5), c3 = (1,10000,5), c4 = (1,10000,5), c5 = (1,1000,1), tau = (0,5,0.0)
```



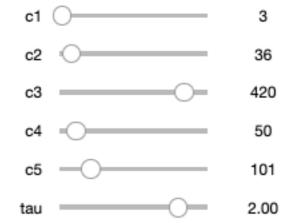


Best Values: c1: 3, c2: 36, c3: 420, c4:50, c5:101, tau=2.0

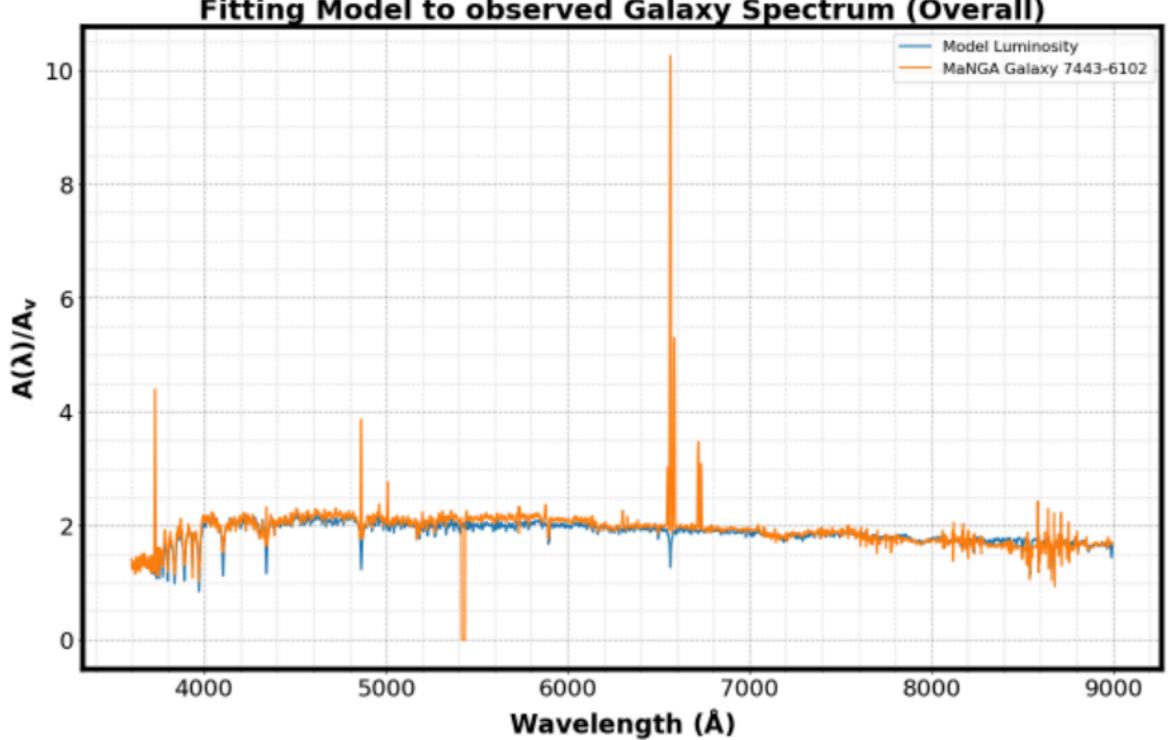
Values that also gave good fit:

```
c1: 4, c2: 1151, c3: 6471, c4:9906, c5:6, tau=2.0
c1: 8, c2: 1196, c3: 2761, c4:2676, c5:8, tau=2.0
c1: 3, c2: 71, c3: 180, c4:820, c5:100, tau=2.0
```

```
In [87]: def series(c1, c2, c3, c4, c5, tau):
             plt.figure(figsize=(15,9))
             plt.plot(bc03_offset['WAVE'], L_model(c1,c2,c3,c4,c5,tau), label='Model Luminosity')
             plt.plot(manga_7443_6102['WAVE'], manga_7443_6102['LUMINOSITY'], label='Manga_Galaxy 7443-6102')
             plt.title('Fitting Model to observed Galaxy Spectrum (Overall)')
             plt.xlabel('Wavelength $\mathbf{(\AA)}$')
             plt.ylabel('A$\mathbf{(\lambda) / A_v\$')
             plt.legend(fontsize=12)
             plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
             plt.minorticks on()
             plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1);
             plt.show()
             return()
         interact(series, c1 = (1,500,1), c2 = (1,500,1), c3 = (1,500,1), c4 = (1,500,1), c5 = (1,500,1), tau = (0,2.5,0.05);
```







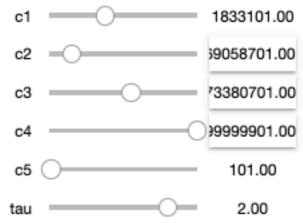
In the case we multiply the galaxy spectrum by its 1e6 scale. The variables change to:

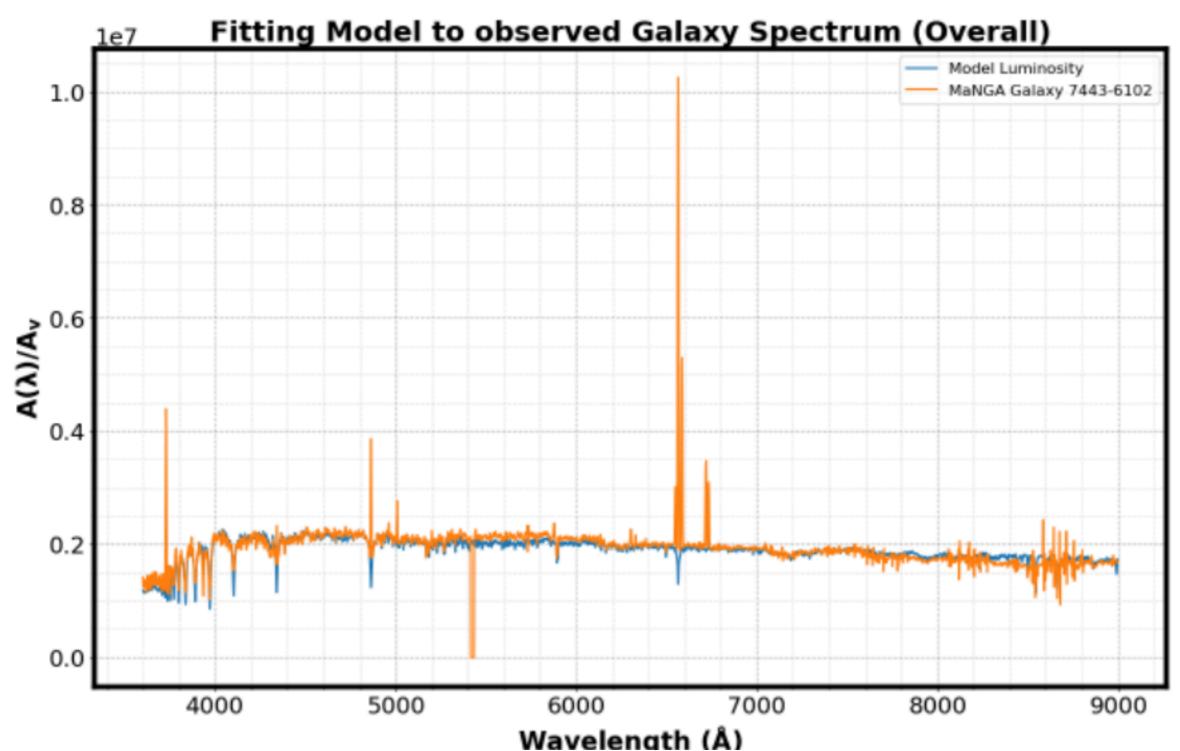
```
Option 1: c1: 1833101.00, c2: 69058701.00, c3: 273380701.00, c4: 499999901.00, c5: 101, tau: 2

Option 2: c1: 20001.00, c2: 10090601.00, c3: 20167601.00, c4: 50945001.00, c5: 1001, tau: 2
```

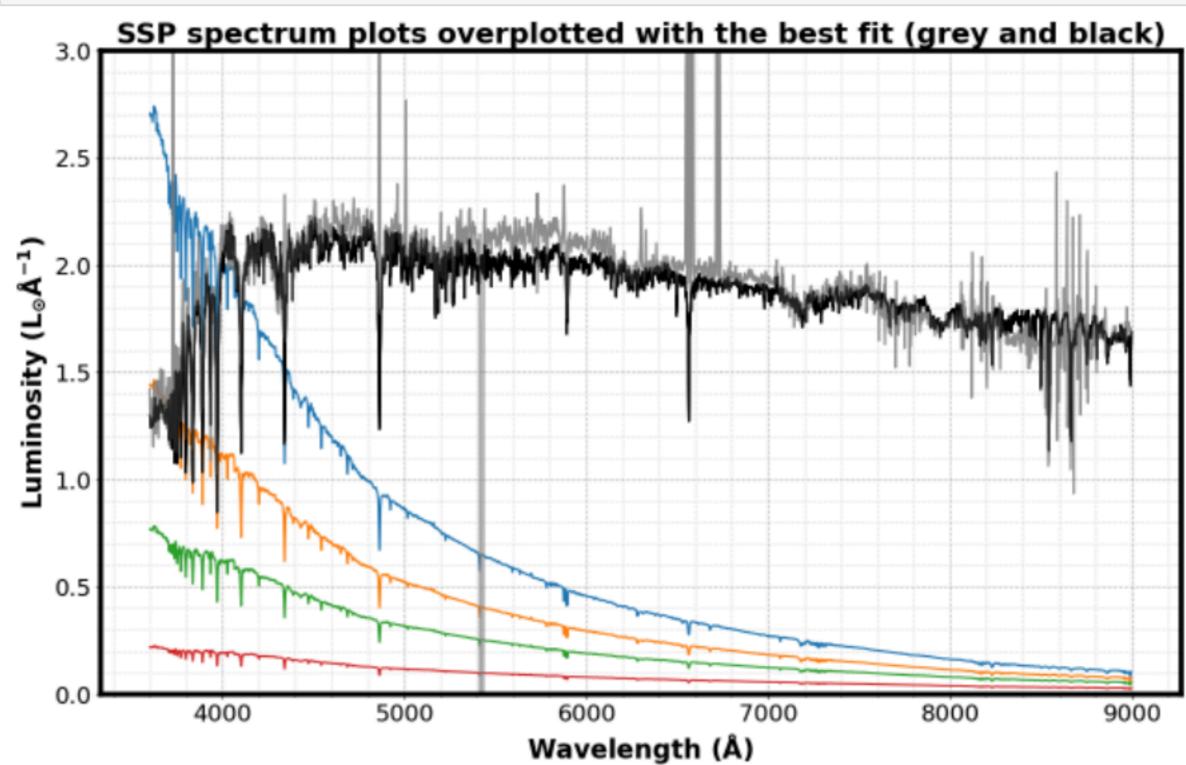
```
In [88]:

def series(c1, c2, c3, c4, c5, tau):
    plt.plot(bc03_offset['WAVE'], L_model(c1,c2,c3,c4,c5,tau), label='Model Luminosity')
    plt.plot(manga_7443_6102['WAVE'], manga_7443_6102['LUMINOSITY']*le6, label='Manga Galaxy 7443-6102')
    plt.title('Fitting Model to observed Galaxy Spectrum (Overall)')
    plt.xlabel('Wavelength $\mathbf{(\AA)}$')
    plt.ylabel('A$\mathbf{(\Ambda) / A_v)$')
    plt.legend(fontsize=12)
    plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
    plt.minorticks_on()
    plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1);
    plt.show()
    return()
    interact(series, c1 = (1,5e6,le2), c2 = (1,5e8,le2), c3 = (1,5e8,le2), c4 = (1,5e8,le2), c5 = (1,5e6,le2), tau = (0,2.5)
```





```
In [81]: for i in (0, 0.5, 1, 2):
             LUM1_dusty = L_dusty(i, bc03_offset[age_model[0]], bc03_offset['WAVE'])
             plt.plot(bc03_offset['WAVE'], LUM1_dusty*le2)
             plt.xlabel('Wavelength $\mathbf{(\AA)}$')
             plt.ylabel('Luminosity $\mathbf{(L_{\odot}\AA^{-1}))$')
             plt.ylim(0, 3)
             plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
             plt.minorticks_on()
             plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1)
         for i in range(0, len(age_model)):
             LUM1_dusty = L_dusty(2, bc03_offset[age_model[i]], bc03_offset['WAVE'])
             plt.plot(bc03_offset['WAVE'], LUM1_dusty)
             plt.plot(bc03_offset['WAVE'], L_model(3,36,420,50,101,2), 'black')
             plt.plot(manga_7443_6102['WAVE'], manga_7443_6102['LUMINOSITY'], 'grey', alpha =0.3)
             plt.title('SSP spectrum plots overplotted with the best fit (grey and black)')
             plt.xlabel('Wavelength $\mathbf{(\AA)}$')
             plt.ylabel('Luminosity $\mathbf{(L_{\odot}\AA^{-1}))$')
             plt.grid(b=True, which='major', color='k', linestyle='--', alpha=0.3)
             plt.minorticks_on()
             plt.grid(b=True, which='minor', color='k', linestyle='--', alpha=0.1)
```



Observations

I observed that when scaling the different SSP constants (c1, c2, c3, c4 & c5) each had an unique effect on the model luminosity. Some of the obervations are listed below:

- The c1 or LUM1 affected mainly the shorter wavelength (4000 5000 Angstrom) causing the start of the spectrum to be steeper or flatter. The spectrums were very sensitive to this variable (small change would cause significant shift in the modeled spectrum)
- 2) The c2 or LUM10 acted as a translation variable (shifting entire curve) up or down.
- 3) The c3 allowed me to control the mid-wavelengths (5500-7500 Anstrom) and spectrum was not as sensitive to th is variable.
- 4) The c4 moved the tail end part of the spectrum (8000+ Ansgtrom) and spectrum was not as sensitive to this variable.
- 5) The c5 acted as a scale factor affecting the overall sensitivity of modelded spectrum towards rest of the va riables. High value of c5 meant, small changes in each of the c1-c4 have significant changes, while a smaller v alue meant that spectrum was not as sensitive to the constants. It acted as a scale parameter.
- 6) The tau (dust attenuation) affected where the overall strength of each of the absorption and emission features of the spectrum. Smaller the tau, more significant the features (deeper absorption troughs and emission peaks), while larger the tau, less pronounced the features.

This indicates that each luminosity (LUM1-1000) contributes differently and is depended on specific mass to luminosity ratio. This enables us to estimate distribution of stellar mass and how that affects the overall spectrum of a galaxy. Using these features to estimate the constant value that best fit the model spectrum to actual galaxy spectrum enables us to constrain stellar population/distribution of the galaxy.

The Stellar mass from option 1 is 5.141e+10 solar mass

Compared to Milky way $5\times 10^{10}~M_\odot$ The mass of this galaxy is very similar coming at about $5.141\times 10^{10}~M_\odot$