

```
1 pip install astropy
```

Looking in indexes: <https://pypi.org/simple>, <https://us-python.pkg.dev/colab-wheels/public/simple/>  
Requirement already satisfied: astropy in /usr/local/lib/python3.8/dist-packages  
Requirement already satisfied: pyerfa>=1.7.3 in /usr/local/lib/python3.8/dist-packages  
Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.8/dist-packages

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 import math
4 import scipy as sp
5 import scipy.stats as stats
6 from astropy.io import fits
```

## ▼ PART ONE

```
1 # Saving dark, flat, and target files of different exposure times into
2 # individual holders.
3
4 dark_15s = ['s0347.fits',
5            's0348.fits',
6            's0349.fits',
7            's0350.fits',
8            's0351.fits',
9            's0352.fits',
10           's0353.fits']
11
12 dark_44s = ['s0354.fits',
13            's0355.fits',
14            's0356.fits',
15            's0357.fits',
16            's0358.fits',
17            's0359.fits',
18            's0360.fits']
19
20 dark_150s = ['s0796.fits',
21             's0797.fits',
22             's0798.fits',
23             's0799.fits',
24             's0800.fits',
25             's0801.fits',
26             's0802.fits']
27
28 flat_44s = ['s0310.fits',
29            's0311.fits',
30            's0312.fits',
31            's0313.fits',
```

```

32     's0314.fits',
33     's0315.fits',
34     's0316.fits',
35     's0317.fits',
36     's0318.fits',
37     's0319.fits']
38
39
40 tar_150s = ['s0576.fits',
41             's0577.fits',
42             's0578.fits',
43             's0579.fits',
44             's0580.fits',
45             's0581.fits',
46             's0582.fits',
47             's0583.fits',
48             's0584.fits',
49             's0585.fits']
50

```

```

1 # Finding the median values for each of the three sets of dark frames (15 sec,
2 # 44 sec, and 150 sec exposure times).
3
4 def Med(files,extension):
5     im = []
6     for x in range(len(files)):
7         im.append(fits.open(files[x])[extension].data)
8     median = np.median(im,axis = 0)
9     return median

```

```

1 # Applying function and saving values into new variables.
2
3 d15_med = Med(dark_15s,0)
4 d44_med = Med(dark_44s,0)
5 d150_med = Med(dark_150s,0)

```

```

1 # Subtracting the median dark values found above from the flat frames and target
2 # frames.
3
4 flat44s_d = []
5 for i in range(len(flat_44s)):
6     flat44s_d.append(fits.open(flat_44s[i])[0].data - d44_med)
7
8 tar150s_d = []
9 for i in range(len(tar_150s)):
10     tar150s_d.append(fits.open(tar_150s[i])[0].data - d150_med)

```

```

1 # Finding median flat frame after subtracting dark from all flat frames.
2

```

```
2
3 flats_med = np.median(flat44s_d,axis = 0)
```

```
1 # Create subsection of image above to find the median counts in the AO beam.
2 # Zooming in on the image created above and finding the median of the image.
3 # This median value is the median number of counts in the AO beam.
4
5 flat_zoom = flats_med[250:1250,600:1600]
6 zoom_med = np.median(flat_zoom)
```

```
1 # Creating a normalized flat by dividing the median flat image by the median of
2 # the subsection of the image.
3
4 flat_norm = flats_med/zoom_med
5 zoomed_norm_flat = flat_norm[250:1250,600:1600]
6 norm_med = np.median(flat_norm)
7 norm_std = np.std(flat_norm)
```

```
1 # This code ensures that no divisions by zero will occur when science frames are
2 # being divided by flat frames.
3
4 for i in range(len(flat_norm)):
5     for j in range(len(flat_norm[0])):
6         if flat_norm[i][j] < 0.5:
7             flat_norm[i][j] = 1
```

```
1 m_s_im = []
2 # This loop creates normalized science frames
3 for p in range(len(tar150s_d)):
4     m_s_im.append(tar150s_d[p][250:1250,600:1600]/zoomed_norm_flat)
```

```
1 def centerfinder(s):
2     keep = s
3
4     for i in range(8):
5         top = keep[0:int(len(keep)/2),0:len(keep[0])]
6         bottom = keep[int(len(keep)/2):len(keep),0:len(keep[0])]
7         if np.median(top)> np.median(bottom):
8             keep = top
9         else:
10            keep = bottom
11
12
13         left = keep[0:int(len(keep)),0:int(len(keep[0])/2)]
14         right = keep[0:int(len(keep)),int(len(keep[0])/2):int(len(keep[0]))]
15         if np.mean(left)> np.mean(right):
16             keep = left
17         else:
```

```

18         keep = right
19
20     for y in range(len(s)):
21         for x in range(len(s[0])):
22             if s[y][x] == np.max(keep):
23                 return [x,y]

```

```

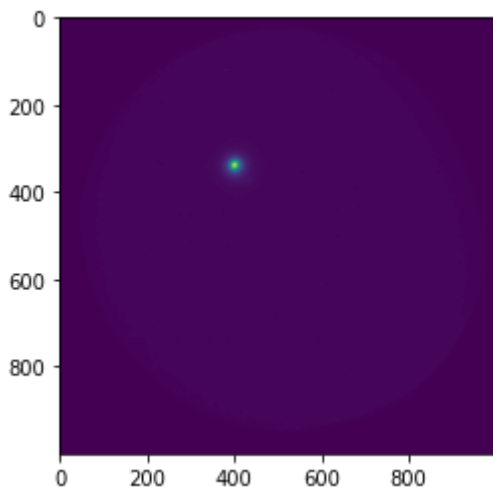
1 for i in range(len(m_s_im)):
2     for j in range(len(m_s_im[5])):
3         for k in range(len(m_s_im[5][0])):
4             if m_s_im[i][j][k] < 0:
5                 m_s_im[i][j][k] = 0
6
7 centerfinder(m_s_im[5])

```

[401, 338]

```
1 plt.imshow(m_s_im[5])
```

<matplotlib.image.AxesImage at 0x7f06b9423a30>



```

1 # Using centerfinder() function
2 # centers holds all info about where the center of each star is,
3 # snippets holds 100x100 pixels snippets centered on star center
4
5 centers = []
6 snippets = []
7 for f in range(len(m_s_im)):
8     centers.append(centerfinder(m_s_im[f]))

```

```
1 print(centers)
```

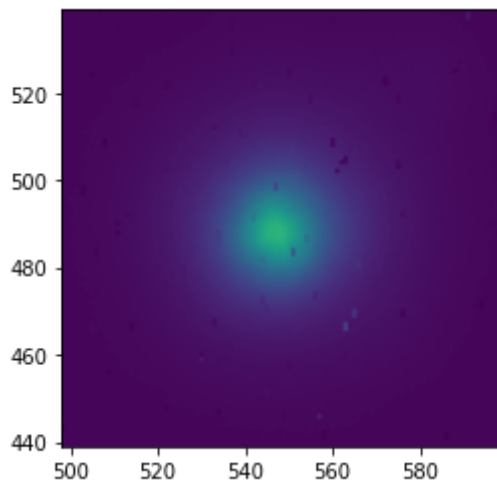
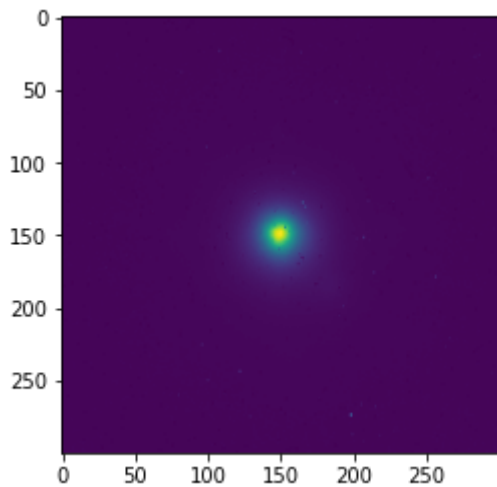
[[548, 489], [546, 487], [547, 487], [697, 334], [699, 331], [401, 338], [400, 338]]

```
1 # 'centers' is an array that holds information about the center of each
```

```

2 # normalized science image.
3
4 #append snippets with the desired sections of each normalized science image
5 for h in range(len(m_s_im)):
6     snippets.append(m_s_im[h][(centers[h][1]-150):(centers[h][1]+150),(centers[h][0]
7
8 # Displaying image to see if star is centered
9 plt.imshow(snippets[0])
10 plt.show()
11
12 plt.imshow(m_s_im[0])
13 plt.axis([498,598,439,539])
14 plt.show()
15

```

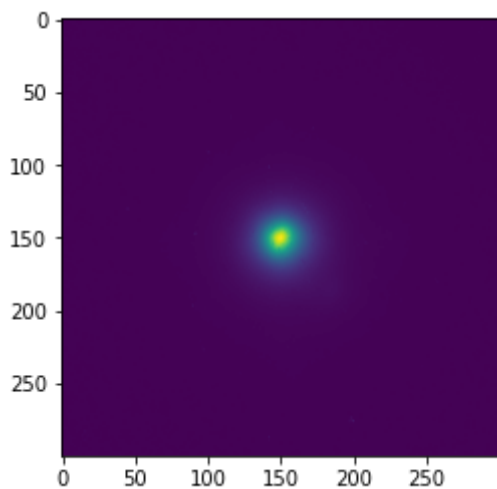


```

1 # Combining all images with centered stars into one final image.
2
3 co_add= np.zeros([300,300])
4
5 for i in range(len(snippets)):
6     co_add = co_add + snippets[i]
7
8 co_add = co_add/10
9 plt.imshow(co_add)

```

<matplotlib.image.AxesImage at 0x7f06b8e97f10>



```
1 psfx0 = snippets[0][150,0:300]
2 psfy0 = snippets[0][0:300,150]
```

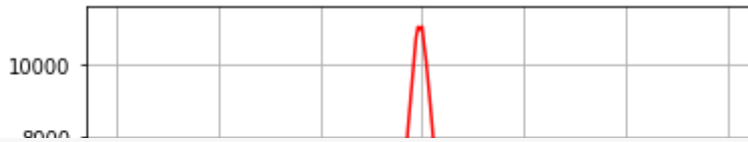
```
1 hm_psfx0 = (0.5)*psfx0.max()
2 hm_psfy0 = (0.5)*psfy0.max()
3 #these two values are the same! So are referred to as hm_psf0 in future lines
4
5 hm_psf0 = hm_psfx0
```

```
1 def FWHM(cut):
2     hm = 0.5*cut.max()
3
4     std = np.std(cut)
5     val = []
6     n = 0
7     while len(val) != 2:
8         n = n+1
9         val = []
10        for i in range(len(cut)):
11            if cut[i] < hm + std/n and cut[i] > hm - std/n:
12                val.append(i)
13
14        if len(val) == 1:
15            n = n-1
16            val = []
17            for i in range(len(cut)):
18                if cut[i] < hm + std/n and cut[i] > hm - std/n:
19                    val.append(i)
20
21            if val[1] - val[0] < 5:
22                return (val[-1] - val[-2])
23                break
24            else:
```

```
25         return (val[1] - val[0])
26         break
27
28
29
30
31
32     return (val[1] - val[0])
33
```

```
1 plt.plot(psfy0,color = 'red')
2 plt.axhline(y=hm_psf0,color = 'green')
3 print ('In the y direction the FWHM is',FWHM(psfy0))
4 plt.xlabel('Pixel Number')
5 plt.ylabel('Counts')
6 plt.grid()
7 plt.show()
8
9 plt.plot(psfx0,color = 'b')
10 plt.axhline(y=hm_psf0,color = 'orange')
11 print ('In the x direction the FWHM is',FWHM(psfx0))
12 plt.xlabel('Pixel Number')
13 plt.grid()
14 plt.ylabel('Counts')
15 plt.show()
```

In the y direction the FWHM is 21



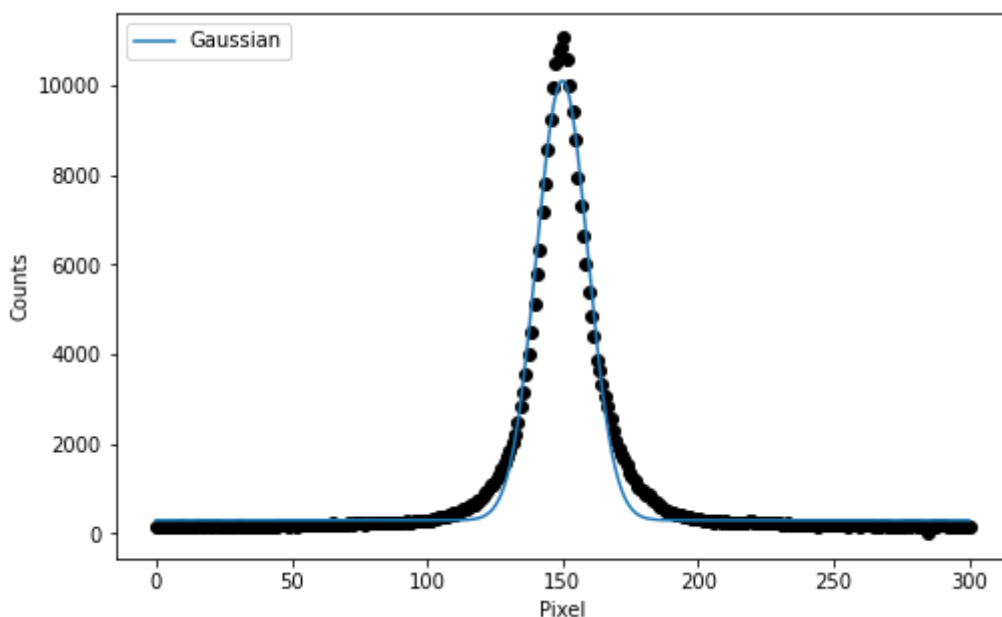
```

1 from astropy.modeling import models, fitting
2
3 xdata = np.linspace(0,300,300)
4 ydata = psfx0
5 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=np.mean(xdata), stddev=np.
6 fit_g = fitting.LevMarLSQFitter()
7 g_x = fit_g(g_init, xdata, ydata)
8 plt.figure(figsize=(8,5))
9 plt.plot(xdata, ydata, 'ko')
10 plt.plot(xdata, g_x(xdata), label='Gaussian')
11 plt.xlabel('Pixel')
12 plt.ylabel('Counts')
13 plt.legend(loc=2)
14 print('Gaussian Fit for the x direction')
15 plt.show()
16
17
18 xdata = np.linspace(0,300,300)
19 ydata = psfy0
20 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=50, stddev=np.std(xdata))
21 fit_g = fitting.LevMarLSQFitter()
22 g_y = fit_g(g_init, xdata, ydata)
23 plt.figure(figsize=(8,5))
24 plt.plot(xdata, ydata, 'ko')
25 plt.plot(xdata, g_y(xdata), label='Gaussian')
26 plt.xlabel('Pixel')
27 plt.ylabel('Counts')
28 plt.legend(loc=2)
29 print('Gaussian Fit for the y direction')
30 plt.show()

```



## Gaussian Fit for the x direction



## Gaussian Fit for the y direction



```
1 g_x
```

```
<CompoundModel(amplitude_0=9807.72937855, mean_0=149.84883466,
stddev_0=9.46911548, amplitude_1=296.61035068)>
```

```
1 g_y
```

```
<CompoundModel(amplitude_0=9987.78236481, mean_0=149.92063103,
stddev_0=9.75056617, amplitude_1=287.80551201)>
```

```
1 psfx3 = snippets[3][150,0:300]
2 psfy3 = snippets[3][0:300,150]
3 hm_psfx3 = (0.5)*psfx3.max()
4 hm_psfy3 = (0.5)*psfy3.max()
```

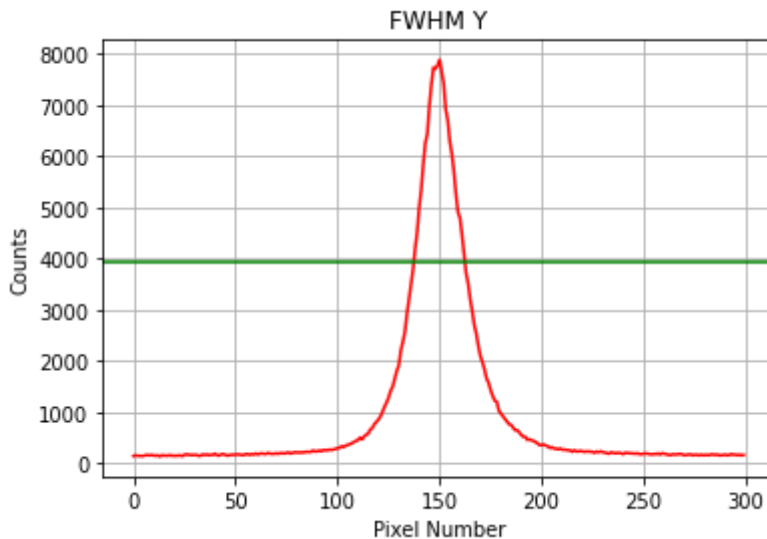
```
1 plt.plot(psfy3,color = 'red')
2 plt.axhline(y=hm_psfy3,color = 'green')
3 print('FWHM in Y-Direction: ',FWHM(psfy3))
4 plt.title("FWHM Y")
5 plt.xlabel('Pixel Number')
6 plt.ylabel('Counts')
7 plt.grid()
8 plt.show()
9
10 plt.plot(psfx3,color = 'b')
11 plt.axhline(y=hm_psfx3,color = 'orange')
12 print ('FWHM in X-Direction: ',FWHM(psfx3))
13 plt.title("FWHM X")
```

```

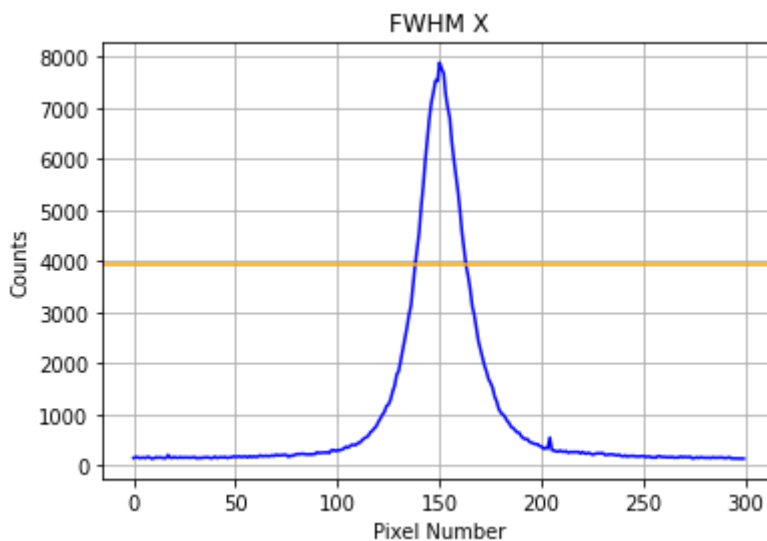
14 plt.xlabel('Pixel Number')
15 plt.ylabel('Counts')
16 plt.grid()
17 plt.show()

```

FWHM in Y-Direction: 25



FWHM in X-Direction: 25



```

1 xdata = np.linspace(0,300,300)
2 ydata = psfx3
3 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=np.mean(xdata), stddev=np.
4 fit_g = fitting.LevMarLSQFitter()
5 g_x3 = fit_g(g_init, xdata, ydata)
6 plt.figure(figsize=(8,5))
7 plt.plot(xdata, ydata,'ko', label="Emission Profile")
8 plt.plot(xdata, g_x3(xdata), label='Gaussian Fit')
9 plt.xlabel('Pixel')
10 plt.ylabel('Counts')
11 plt.legend(loc=2)
12 plt.title('Gaussian Fit: X-Direction')
13 plt.grid()

```

```
14 plt.show()
15
16
17 xdata = np.linspace(0,300,300)
18 ydata = psfy3
19 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=50, stddev=np.std(xdata))
20 fit_g = fitting.LevMarLSQFitter()
21 g_y3 = fit_g(g_init, xdata, ydata)
22 plt.figure(figsize=(8,5))
23 plt.plot(xdata, ydata, 'ko', label="Emission Profile")
24 plt.plot(xdata, g_y3(xdata), label='Gaussian Fit')
25 plt.xlabel('Pixel')
26 plt.ylabel('Counts')
27 plt.legend(loc=2)
28 plt.title('Gaussian Fit: Y-Direction')
29 plt.grid()
30 plt.show()
```

## Gaussian Fit: X-Direction



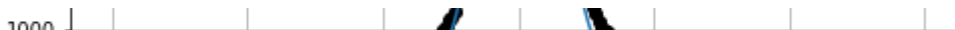
```
1 g_x3
```

```
<CompoundModel(amplitude_0=6987.39909388, mean_0=151.09815486,
stddev_0=12.07673428, amplitude_1=256.83767016)>
```



```
1 g_y3
```

```
<CompoundModel(amplitude_0=6996.54213072, mean_0=150.28753579,
stddev_0=11.99664591, amplitude_1=253.13829597)>
```



```
1 psfx5 = snippets[5][150,0:300]
```

```
2 psfy5 = snippets[5][0:300,150]
```

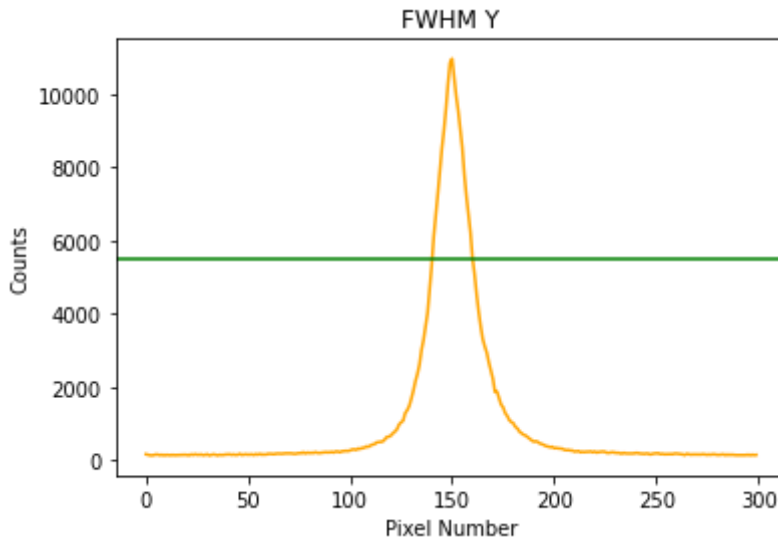
```
3 hm_psfx5 = (0.5)*psfx5.max()
```

```
4 hm_psfy5 = (0.5)*psfy5.max()
```



```
1 plt.plot(psfy5,color = 'orange')
2 plt.axhline(y=hm_psfy5,color = 'green')
3 print ('FWHM in Y-Direction: ',FWHM(psfy5))
4 plt.title("FWHM Y")
5 plt.xlabel('Pixel Number')
6 plt.ylabel('Counts')
7 plt.show()
8
9 plt.plot(psfx5,color = 'b')
10 plt.axhline(y=hm_psfx5,color = 'yellow')
11 print ('FWHM in X-Direction: ',FWHM(psfx5))
12 plt.title("FWHM X")
13 plt.xlabel('Pixel Number')
14 plt.ylabel('Counts')
15 plt.show()
```

FWHM in Y-Direction: 20

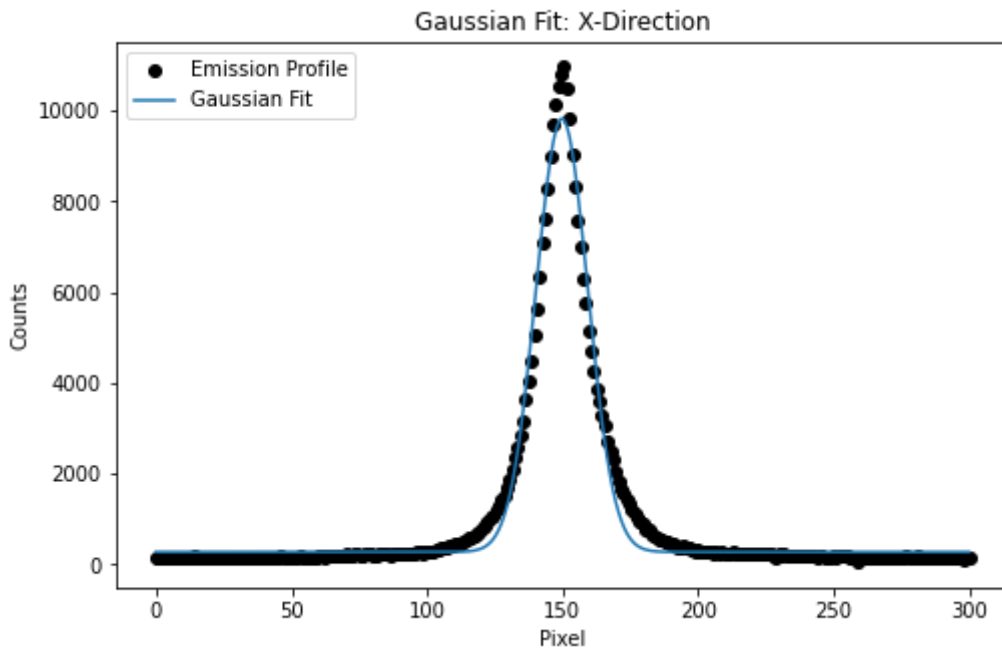


FWHM in X-Direction: 18

```

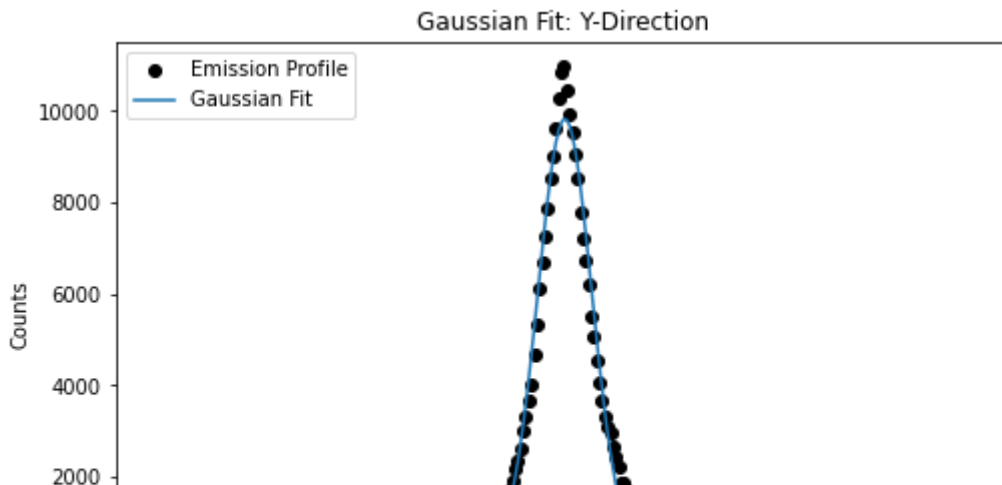
1 xdata = np.linspace(0,300,300)
2 ydata = psfx5
3 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=np.mean(xdata), stddev=np.
4 fit_g = fitting.LevMarLSQFitter()
5 g_x5 = fit_g(g_init, xdata, ydata)
6 plt.figure(figsize=(8,5))
7 plt.plot(xdata, ydata,'ko', label="Emission Profile")
8 plt.plot(xdata, g_x5(xdata), label='Gaussian Fit')
9 plt.xlabel('Pixel')
10 plt.ylabel('Counts')
11 plt.legend(loc=2)
12 plt.title('Gaussian Fit: X-Direction')
13 plt.show()
14
15
16 xdata = np.linspace(0,300,300)
17 ydata = psfy5
18 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=50, stddev=np.std(xdata))
19 fit_g = fitting.LevMarLSQFitter()
20 g_y5 = fit_g(g_init, xdata, ydata)
21 plt.figure(figsize=(8,5))
22 plt.plot(xdata, ydata,'ko', label="Emission Profile")
23 plt.plot(xdata, g_y5(xdata), label='Gaussian Fit')
24 plt.xlabel('Pixel')
25 plt.ylabel('Counts')
26 plt.legend(loc=2)
27 plt.title('Gaussian Fit: Y-Direction')
28 plt.show()

```



WARNING: The fit may be unsuccessful; check fit\_info['message'] for more information.

WARNING:astropy:The fit may be unsuccessful; check fit\_info['message'] for more information.



```
1 g_x5
```

```
<CompoundModel(amplitude_0=9555.96762134, mean_0=149.72561841,
stddev_0=9.5578744, amplitude_1=273.24176652)>
```

```
1 g_y5
```

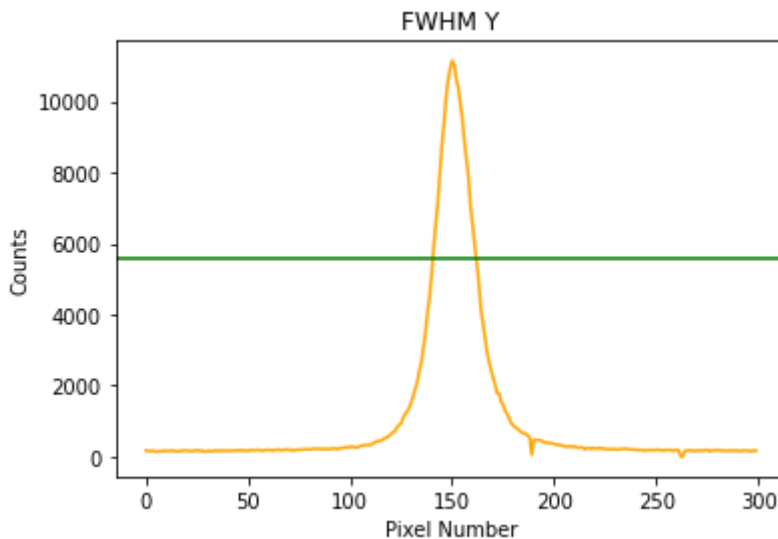
```
<CompoundModel(amplitude_0=9552.96647431, mean_0=150.80062487,
stddev_0=9.88539749, amplitude_1=275.12640495)>
```

```
1 psfx9 = snippets[9][150,0:300]
2 psfy9 = snippets[9][0:300,150]
3 hm_psfx9 = (0.5)*psfx9.max()
4 hm_psfy9 = (0.5)*psfy9.max()
```

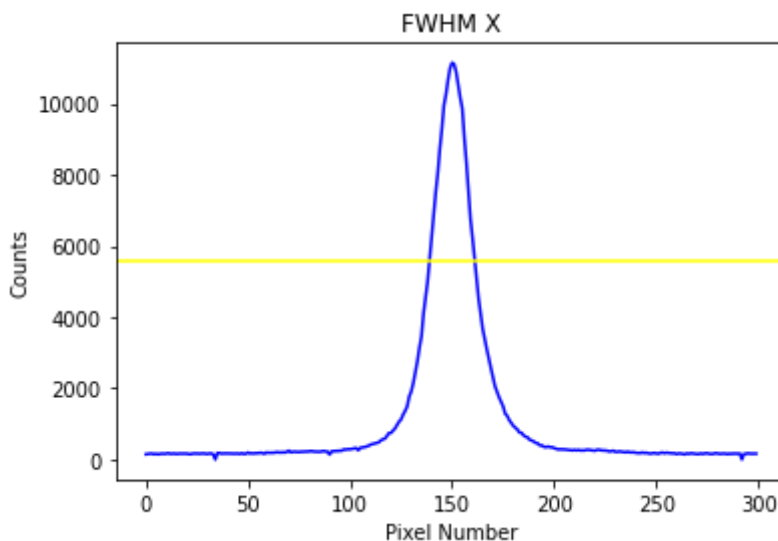
```
1 plt.plot(psfy9,color = 'orange')
```

```
2 plt.axhline(y=hm_psfy9,color = 'green')
3 print ('FWHM in Y-Direction: ',FWHM(psfy9))
4 plt.title("FWHM Y")
5 plt.xlabel('Pixel Number')
6 plt.ylabel('Counts')
7 plt.show()
8
9 plt.plot(psfx9,color = 'b')
10 plt.axhline(y=hm_psfx9,color = 'yellow')
11 print ('FWHM in X-Direction: ',FWHM(psfx9))
12 plt.title("FWHM X")
13 plt.xlabel('Pixel Number')
14 plt.ylabel('Counts')
15 plt.show()
```

FWHM in Y-Direction: 20



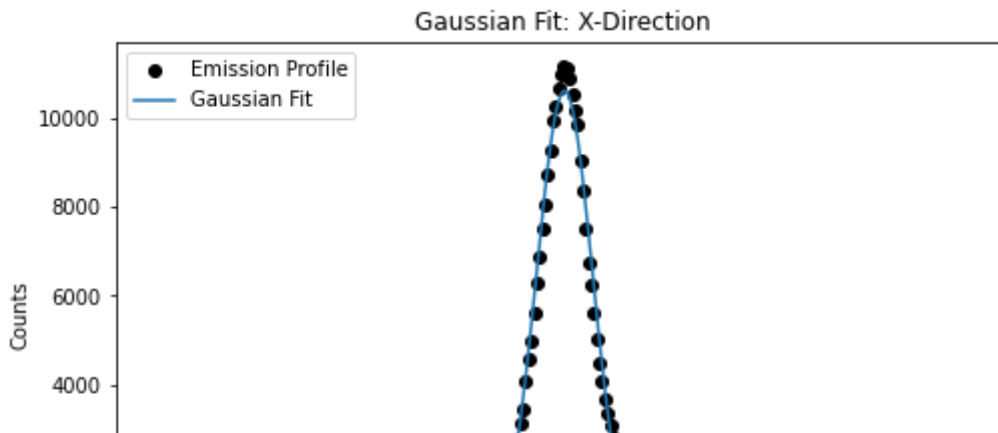
FWHM in X-Direction: 22



```
1 xdata = np.linspace(0,300,300)
2 ydata = psfx9
3 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=np.mean(xdata), stddev=np.
4 fit_g = fitting.LevMarLSQFitter()
```

```
5 g_x9 = fit_g(g_init, xdata, ydata)
6 plt.figure(figsize=(8,5))
7 plt.plot(xdata, ydata,'ko', label="Emission Profile")
8 plt.plot(xdata, g_x9(xdata), label='Gaussian Fit')
9 plt.xlabel('Pixel')
10 plt.ylabel('Counts')
11 plt.legend(loc=2)
12 plt.title('Gaussian Fit: X-Direction')
13 plt.show()
14
15
16 xdata = np.linspace(0,300,300)
17 ydata = psfy9
18 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=50, stddev=np.std(xdata))
19 fit_g = fitting.LevMarLSQFitter()
20 g_y9 = fit_g(g_init, xdata, ydata)
21 plt.figure(figsize=(8,5))
22 plt.plot(xdata, ydata,'ko', label="Emission Profile")
23 plt.plot(xdata, g_y9(xdata), label='Gaussian Fit')
24 plt.xlabel('Pixel')
25 plt.ylabel('Counts')
26 plt.legend(loc=2)
27 plt.title('Gaussian Fit: Y-Direction')
28 plt.show()
```





```
1 g_x9
```

```
<CompoundModel(amplitude_0=10349.17915377, mean_0=150.69518163,
stddev_0=10.27463531, amplitude_1=266.94409771)>
```

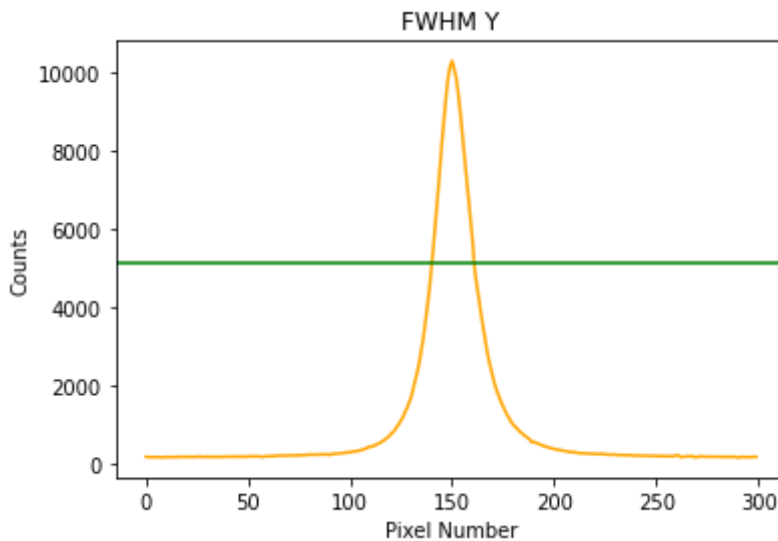
```
1 g_y9
```

```
<CompoundModel(amplitude_0=10227.75418275, mean_0=151.5645408,
stddev_0=9.76196401, amplitude_1=264.2761535)>
```

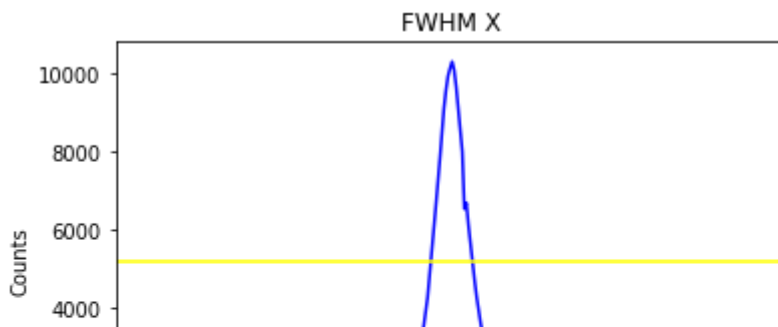
```
1 psfx = co_add[150,0:300]
2 psfy = co_add[0:300,150]
3 hm_psfx = (0.5)*psfx.max()
4 hm_psfy = (0.5)*psfy.max()
```

```
1 plt.plot(psfy,color = 'orange')
2 plt.axhline(y=hm_psfy,color = 'green')
3 print ('FWHM in Y-Direction: ',FWHM(psfy))
4 plt.title('FWHM Y')
5 plt.xlabel('Pixel Number')
6 plt.ylabel('Counts')
7 plt.show()
8
9 plt.plot(psfx,color = 'b')
10 plt.axhline(y=hm_psfx,color = 'yellow')
11 print ('FWHM in X-Direction: ',FWHM(psfx))
12 plt.title('FWHM X')
13 plt.xlabel('Pixel Number')
14 plt.ylabel('Counts')
15 plt.show()
```

FWHM in Y-Direction: 21



FWHM in X-Direction: 20

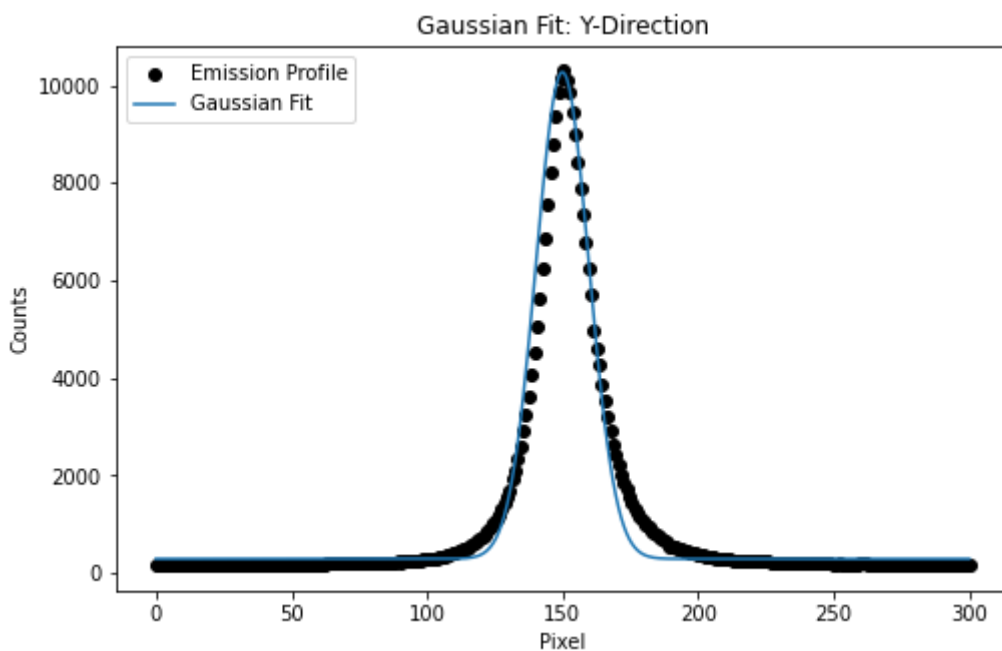
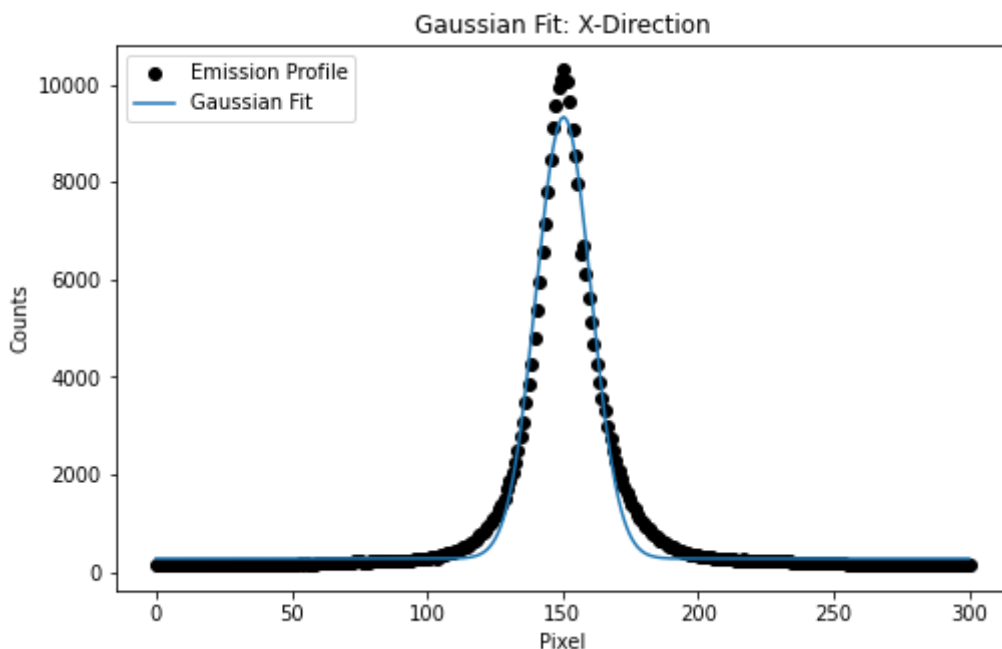


```

1 xdata = np.linspace(0,300,300)
2 ydata = psfx
3 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=np.mean(xdata), stddev=np.
4 fit_g = fitting.LevMarLSQFitter()
5 g_x = fit_g(g_init, xdata, ydata)
6 plt.figure(figsize=(8,5))
7 plt.plot(xdata, ydata,'ko', label="Emission Profile")
8 plt.plot(xdata, g_x(xdata), label='Gaussian Fit')
9 plt.xlabel('Pixel')
10 plt.ylabel('Counts')
11 plt.legend(loc=2)
12 plt.title('Gaussian Fit: X-Direction')
13 plt.show()
14
15
16 xdata = np.linspace(0,300,300)
17 ydata = psfy
18 g_init = models.Gaussian1D(amplitude=np.max(ydata), mean=50, stddev=np.std(xdata))
19 fit_g = fitting.LevMarLSQFitter()
20 g_y3 = fit_g(g_init, xdata, ydata)
21 plt.figure(figsize=(8,5))
22 plt.plot(xdata, ydata,'ko', label="Emission Profile")
23 plt.plot(xdata, g_y(xdata), label='Gaussian Fit')
24 plt.xlabel('Pixel')
25 plt.ylabel('Counts')

```

```
26 plt.legend(loc=2)
27 plt.title('Gaussian Fit: Y-Direction')
28 plt.show()
```



```
1 g_x
```

```
<CompoundModel(amplitude_0=9051.95159099, mean_0=150.33536069,
stddev_0=10.12383687, amplitude_1=279.87173984)>
```

```
1 g_y
```

```
<CompoundModel(amplitude_0=9987.78236481, mean_0=149.92063103,
stddev_0=9.75056617, amplitude_1=287.80551201)>
```

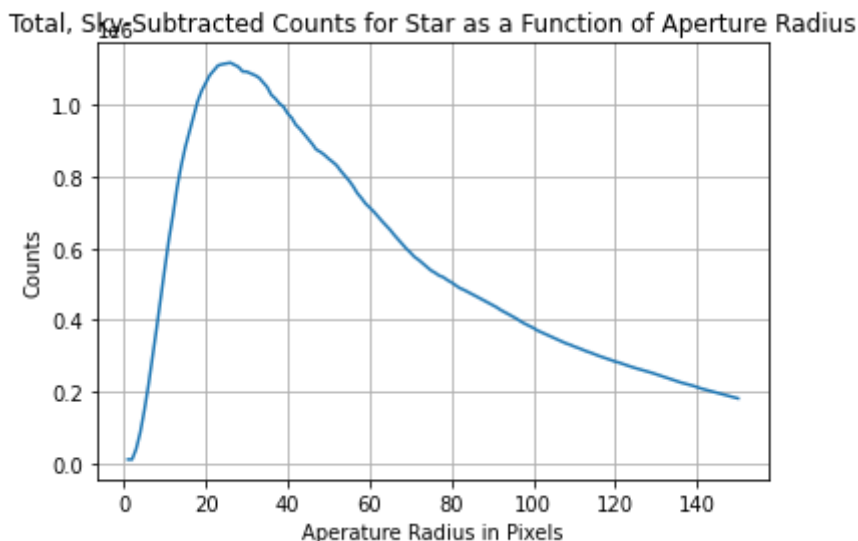
## ▼ PART TWO

```

1 def func(data,y,x,n,s):
2     pc = [data[y][x]]
3     size = n
4     local = [n]
5     sky_sum = np.median(data[0:20*s,0:20*s])
6     tot_sky = []
7     fin = [pc[0]-sky_sum]
8
9     while size < 150:
10         start = data[x-size:x+size,y-size:y+size]
11         sumz = size*size*np.median(start.flatten())
12         sky = sky_sum*size*size
13         tot_sky.append(np.abs(sky))
14         pc.append(sumz)
15         fin.append(sumz - sky)
16         size = size + n
17         local.append(size)
18
19     plt.plot(local,fin)
20     plt.grid()
21     plt.title("Total, Sky-Subtracted Counts for Star as a Function of Aperture Radi
22     plt.xlabel('Aperature Radius in Pixels')
23     plt.ylabel('Counts')

```

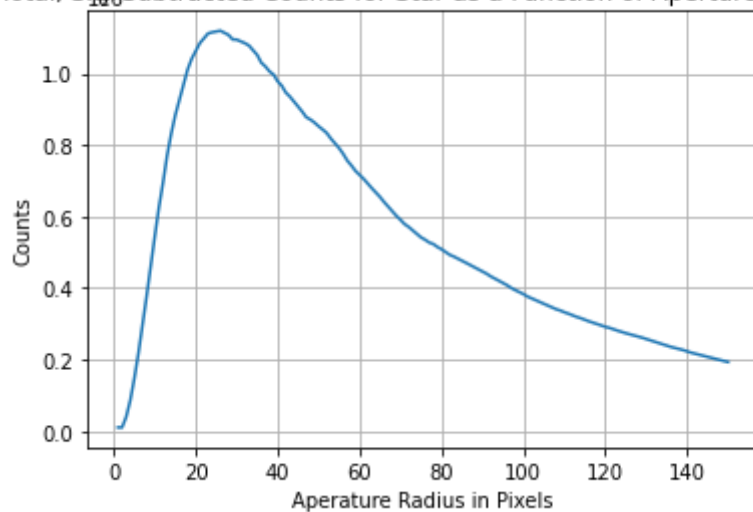
```
1 func(co_add,150,150,1,1)
```



```
1 # Varying the location and size of the aperture used to measure the sky
```

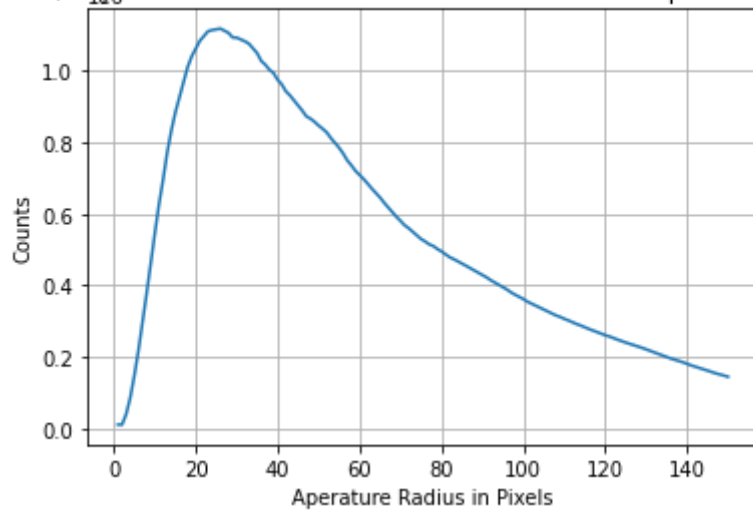
```
1 func(co_add,150,150,1,2)
```

Total, Sky-Subtracted Counts for Star as a Function of Aperture Radius



```
1 func(co_add,150,150,1,5)
```

Total, Sky-Subtracted Counts for Star as a Function of Aperture Radius

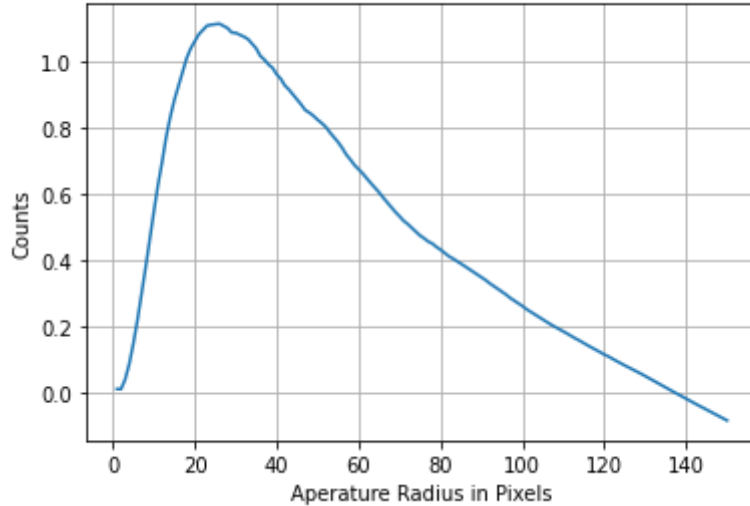


```
1 func(co_add,150,150,1,7)
```

## Total, Sky Subtracted Counts for Star as a Function of Aperture Radius

```
1 func(co_add,150,150,1,10)
```

## Total, Sky Subtracted Counts for Star as a Function of Aperture Radius



I will select an aperture of 25 within which to measure the total flux of the star. No matter how the location and size of the aperture being used to measure the sky varies, the graph has a peak around a radius of 25 pixels.

```
1 val = co_add[125:175,125:175]
2 vsum = val.sum()
3 dvsum = vsum/150
4 print("Total Counts/Second for Targeted Star: ",dvsum, "Counts/Second")
```

Total Counts/Second for Targeted Star: 45421.764173528034 Counts/Second

## ▼ PART THREE

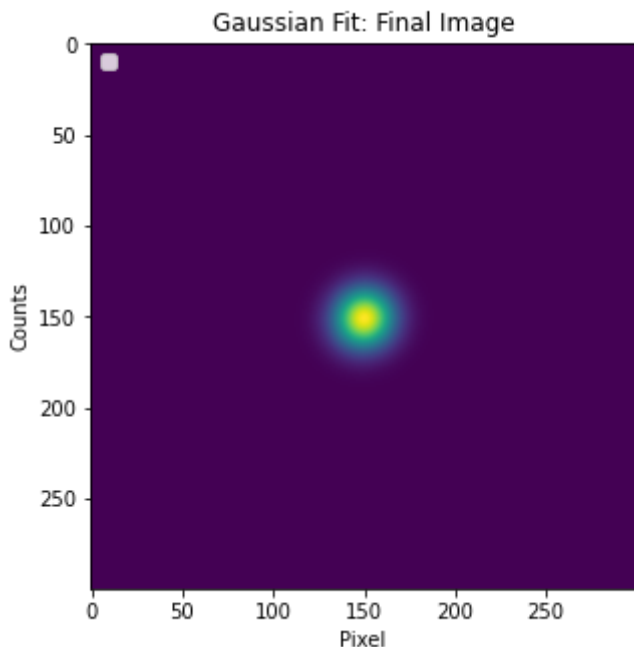
```
1 y_data = np.arange(300)
2 x_data = np.arange(300)
3
4 x, y = np.meshgrid(x_data,y_data)
5
6 z_data = co_add
7 g_init3 = models.Gaussian2D(amplitude=co_add.max(), x_mean=150., y_mean=150., x_std=
8                               )+ models.Const2D(amplitude=1.)
9 fit_g3 = fitting.LevMarLSQFitter()
10 g_3 = fit_g3(g_init3,x,y,z_data)
11 plt.figure(figsize=(8,5))
12
13 plt.imshow(g_3(x,y), label='Gaussian')
14 plt.xlabel('Pixel')
```

```

15 plt.ylabel('Counts')
16 plt.legend(loc=2)
17 plt.title('Gaussian Fit: Final Image')
18 plt.show()

```

WARNING:matplotlib.legend:No handles with labels found to put in legend.



```
1 g_3
```

```

<CompoundModel(amplitude_0=7867.28167245, x_mean_0=150.0927443,
y_mean_0=150.74957877, x_stddev_0=11.83045637, y_stddev_0=11.51437649,
theta_0=84.00115207, amplitude_1=194.30539993)>

```

```
1 (g_3(x,y)[125:175,125:175].sum() - (194.7*2500))/150
```

```
41989.58112719517
```

```
1 val.sum()/150
```

```
45421.764173528034
```



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