97%

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- 9 September 2022
- Dr. Moe

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- ASTR5420
- Please show all work. If you collaborate with other students, write their names at the top of your homework. Please also write the total number of hours you spent on this. If you use any code or plotting routines, print those out and attach them to your hand-written solutions.

1. (10%): For AFGK dwarfs, the mass-luminosity relation scales as $L \propto M^{3.5}$. If this relation continued toward earlier spectral types, then eventually the luminosity would exceed the Eddington limit. Calculate the mass of a star such that its luminosity would equal the Eddington limit assuming $L \propto M^{3.5}$. In reality, in this high-mass regime, massive O and Wolf-Rayet stars actually follow a $L \propto M$ relation limited by the Eddington luminosity.

$$L_{Edd} = 3.2 \times 10^4 \left(\frac{M}{M_{\odot}}\right) L_{\odot}$$

$$L \propto M^{3.5}$$

$$\frac{L}{L_{\odot}} = 3.2 \times 10^4 \left(\frac{M}{M_{\odot}}\right) = \left(\frac{M}{M_{\odot}}\right)^{3.5}$$

$$\left(\frac{M}{M_{\odot}}\right)^{2.5} = 3.2 \times 10^4$$

$$M = 63.4 M_{\odot}$$

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2. (20%): To first order, the density profile of the Sun can be described as $\rho(r) = \rho_c (1 - \frac{r}{R_{\odot}})^6$.

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(a) Use the mass conservation equation to calculate the central density ρ_c . Express your answer in both units of g cm⁻³ and M_{\odot} R_{\odot}⁻³.

$$\begin{split} \rho\left(r\right) &= \rho_{c} \left(1 - \frac{r}{R_{\odot}}\right)^{6} 4\pi r^{2} dr \\ 4\pi r^{2} \rho\left(r\right) dr &= \rho_{c} \left(1 - \frac{r}{R_{\odot}}\right)^{6} 4\pi r^{2} dr \\ dM &= \rho_{c} \left(1 - \frac{r}{R_{\odot}}\right)^{6} 4\pi r^{2} dr \\ \int_{0}^{M_{\odot}} dM &= 4\pi \rho_{c} \int_{0}^{R_{\odot}} \left(1 - \frac{r}{R_{\odot}}\right)^{6} r^{2} dr \\ M_{\odot} &= \frac{4\pi \rho_{c}}{R_{\odot}^{6}} \int_{0}^{R_{\odot}} \left(R_{\odot} - r\right)^{6} r^{2} dr \\ M_{\odot} &= \frac{4\pi \rho_{c}}{R_{\odot}^{6}} \int_{0}^{R_{\odot}} \left(r^{1/3} R_{\odot} - r^{4/3}\right)^{6} dr \\ M_{\odot} &= \frac{4\pi \rho_{c}}{R_{\odot}^{6}} \int_{0}^{R_{\odot}} \left(r^{1/3} R_{\odot} - r^{4/3}\right)^{6} dr \\ M_{\odot} &= \frac{4\pi \rho_{c}}{R_{\odot}^{6}} \int_{0}^{R_{\odot}} \left(r^{1/3} R_{\odot}\right)^{6} + 6 \left(r^{1/3} R_{\odot}\right)^{5} \left(-r^{4/3}\right)^{4} + 6 \left(r^{1/3} R_{\odot}\right)^{4} \left(-r^{4/3}\right)^{2} \\ &+ 20 \left(r^{1/3} R_{\odot}\right)^{3} \left(-r^{4/3}\right)^{3} + 15 \left(r^{1/3} R_{\odot}\right)^{2} \left(-r^{4/3}\right)^{4} + 6 \left(r^{1/3} R_{\odot}\right) \left(r^{4/3}\right)^{5} + \left(-r^{4/3}\right)^{6} dr \\ M_{\odot} &= \frac{4\pi \rho_{c}}{R_{\odot}^{6}} \int_{0}^{R_{\odot}} r^{2} R_{\odot}^{6} - 6 \left(r^{8/3} R_{\odot}^{3}\right) \left(r^{4/3}\right) + 15 \left(r^{4/3} R_{\odot}^{4}\right) \left(r^{8/3}\right) - 20 \left(r R_{\odot}^{3}\right) \left(r^{4}\right) + 15 \left(r^{2/3} R_{\odot}^{2}\right) \left(r^{16/3}\right) \\ &- 6 \left(r^{1/3} R_{\odot}\right) \left(r^{20/3}\right) + \left(r^{8}\right) dr \\ M_{\odot} &= \frac{4\pi \rho_{c}}{R_{\odot}^{6}} \int_{0}^{R_{\odot}} r^{2} R_{\odot}^{6} - 6 r^{3} R_{\odot}^{5} + 15 r^{4} R_{\odot}^{6} - 20 r^{5} R_{\odot}^{3} + 15 r^{6} R_{\odot}^{2} - 6 r^{7} R_{\odot} + r^{2} dr \\ M_{\odot} &= \frac{4\pi \rho_{c}}{R_{\odot}^{6}} \left(\frac{1}{3} r^{3} R_{\odot}^{6} - \frac{3}{3} r^{3} R_{\odot} + 3 r^{5} R_{\odot}^{4} - \frac{10}{3} r^{6} R_{\odot}^{3} + \frac{15}{7} r^{7} R_{\odot}^{2} - \frac{3}{4} r^{8} R_{\odot} + \frac{1}{9} r^{9} \right|_{0}^{R_{\odot}} \right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(\frac{1}{3} - \frac{3}{2} + 3 r^{5} R_{\odot}^{4} - \frac{10}{3} r^{5} R_{\odot}^{4} + \frac{1}{7} r^{8} R_{\odot}^{3} + \frac{1}{9} r^{8} \right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(\frac{3}{3} - \frac{3}{2} + 3 r^{5} \frac{15}{63} + \frac{7}{63}\right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(-\frac{3}{4} + \frac{135}{63} + \frac{7}{63}\right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(-\frac{3}{4} + \frac{135}{63} + \frac{7}{63}\right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(-\frac{3}{4} + \frac{135}{63} + \frac{7}{63}\right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(-\frac{3}{4} + \frac{135}{63} + \frac{7}{63}\right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(-\frac{3}{4} + \frac{135}{63} + \frac{7}{63}\right) \\ M_{\odot} &= 4\pi \rho_{c} R_{\odot}^{3} \left(-\frac{3}$$

 $\rho_c = 119 \text{ g cm}^{-3}, 20.1 \text{ M}_{\odot} \text{ R}_{\odot}^{-1}$

(b) Calculate the mass $M_{r<0.2}$ enclosed within radius $r<0.2R_{\odot}$ in units of M_{\odot} . Computed numerically in python.

$$M_{\odot} = \frac{4\pi\rho_c}{R_{\odot}^6} \int_0^{0.2R_{\odot}} (R_{\odot} - r)^6 r^2 dr$$

$$M_{r<0.2} = 0.262M_{\odot}$$

Code:

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```
In [67]: Rrange = np.linspace(0, .2*Ro, 1000)
         M = ((4*np.pi*pc)/(Ro**6))*(Ro-Rrange)**6*Rrange**2
         Min = trapz(M, Rrange)
         print('M r<0.2 = ' +str(format(Min/Mo, '.2E')) +' Mo')
         M_r<0.2 = 2.62E-01 Mo
```

(c) As explained in Ex. 2.4 of the textbook, the Sun's central pressure is $P_c = \langle M(r)/r^4 \rangle GM/4\pi$, where the crude approximation of $\langle M(r)/r^4 \rangle = M_{\odot}/R_{\odot}^4$ results in a central pressure that is two orders of magnitude lower than the true value. Instead, the average value $\langle M(r)/r^4 \rangle$ is closer to the evaluation of $M(r)/r^4$ toward small radii. Thus assume $\langle M(r)/r^4 \rangle = M_{r<0.2}/(0.2R_{\odot})^4$, and calculate the central pressure $P_c = GM_{\odot}M_{r<0.2}/4\pi(0.2R_{\odot})^4$ in units of dyn cm⁻².

$$P_c = \frac{GM_{\odot}M_{r<0.2}}{4\pi \left(0.2R_{\odot}\right)^4}$$

$$P_c = 1.48 \times 10^{17} \text{ dyn cm}^{-2}$$

Code:

```
In [73]: Pc = (G*Mo*Min)/(4*np.pi*(0.2*Ro)**4)
                print('Central Pressure: ' +format(Pc, '.2E') +' Pa')
                print('Central Pressure: ' +format(Pc*1000/100, '.2E') +' dyn/cm^2')
                Central Pressure: 1.48E+16 Pa
                Central Pressure: 1.48E+17 dyn/cm^2
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```

- 3. (20%) Assume that Earth's troposphere is isothermal with temperature $T=280~\mathrm{K}$ and composed of 77% N_2 and 23% O_2 .
 - (a) Compute the mean molecular weight μ of air.

$$\mu = \sum_{i} \frac{n_i}{N} m_i$$

$$\mu = 0.77 \text{ (28.014 amu)} + 0.23 \text{ (31.999 amu)}$$

$$\mu = 28.9 \text{ amu}$$

Code:

mu: 2.89E+01 amu

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(b) Calculate the pressure scale height H of our atmosphere. Assuming atmosphere is an ideal gas.

$$\frac{dP(r)}{dr} = -\rho(r)g$$

$$\rho(r) = \frac{m}{V}$$

$$\rho(r) = \frac{N\mu m_H}{V}$$

$$\rho(r) = \frac{\frac{P(r)V}{kT}\mu m_H}{V}$$

$$\rho(r) = \frac{P(r)\mu m_H}{kT}$$

$$\frac{dP}{dr} = -\frac{P(r)\mu m_H g}{kT}$$

$$\frac{dP}{P(r)} = -\frac{\mu m_H g}{kT} dr$$

$$P(r) = P_0 e^{-(\frac{\mu m_H g}{kT})r}$$

$$P(H) = P_0 e^{-1}$$

$$\frac{\mu m_H gH}{kT} = 1$$

$$H = \frac{kT}{\mu m_H g}$$

Code:

```
In [84]: T = 280
k = 1.38e-23
mh = 1.67e-27
g = 9.8
H = (k*T)/(mu*mh*g)
print('Scale height: ' +format(H, '.2E') +' m')
Scale height: 8.16E+03 m
```

(c) What is the air pressure at Jelm Mountain (altitude of 2943 m) relative to sea level?

$$P(r) = P_0 e^{-r/H}$$

 $r = 2943$

Code:

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```
In [92]: P0 = 101500
r = 2943
Pr = np.exp(-r/H)
print('Atmospheric Pressure: ' +format(Pr, '.2E') +' P_0')
```

Atmospheric Pressure: 6.97E-01 P_0

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4. (15%) The giant molecular cloud Orion A is the most active star-forming region in the solar neighborhood. The cloud has mass $M=10^5 M_{\odot}$, radius r=50 pc, and mean temperature T=60 K. Compare its mean density ρ to the critical Jeans density ρ_J (both in units of g cm⁻³) assuming the gas is completely composed of molecular hydrogen H₂. Is $\rho > \rho_J$ as expected?

$$\rho = \frac{M}{V}$$

$$\rho = \frac{M}{4/3\pi R^3}$$

$$\rho = 1.36 \times 10^{-26} \text{ g cm}^{-3}$$

$$\rho_J = \left(\frac{5kT}{\mu m_H G}\right)^3 \left(\frac{3}{4\pi M^2}\right)$$

$$\rho_J = 3.47 \times 10^{-20} \text{g cm}^{-3}$$

Code:

As expected, the density of this molecular cloud is greater than the Jean's density. This means that according to the Jean's criterion, this cloud will have active star formation.

- 5. (35%) Download from WyoCourses the Gaia table of 942,267 stars brighter than G < 19 mag and within d < 150 pc (parallaxes > 6.6667 mas). The file is comma separated values (.csv), and the columns are Gaia DR3 ID, RA, Dec, parallax (mas), G, $G_{BP} G_{RP}$, $G_{BP} G$, and $G G_{RP}$.
 - (a) Plot a color-magnitude diagram (CMD; MG versus $G_{BP} G_{RP}$). Label the main-sequence, white dwarfs, and red giant branch. Draw and label the approximate regions bracketing the different spectral types:

A dwarfs $(0.7 < M_G < 2.4, -0.3 < G_{BP} - G_{RP} < 0.6)$
F dwarfs $(2.4 < M_G < 4.2, 0.1 < G_{BP} - G_{RP} < 1.0)$
G dwarfs $(4.2 < M_G < 5.4, 0.5 < G_{BP} - G_{RP} < 1.2)$
K dwarfs (5.4 $< M_G < 8.0, 0.7 < G_{BP} - G_{RP} < 2.1$)
M dwarfs $(8.0 < M_G < 15.5, 1.6 < G_{BP} - G_{RP} < 5.0)$

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Code:

```
In [121]: # Columns: Gaia DR3 ID, RA, Dec, parallax (mas), G, GBP — GRP, GBP — G, and G — GRP.
                            gaia = pandas.read_csv('/d/users/nikhil/Downloads/Gaia_Glt19_dlt150pc.csv')
                            ras = gaia['ra']
                             decs = gaia['dec']
                           par = gaia['parallax']
Gmag = gaia['phot g mean mag']
bpmrp = gaia['bp_rp']
                            bpmg = gaia['bp g']
                            gmrp = gaia['g rp']
distances = 1.5*10**11/(np.tan(par*np.pi/(1000*60*60*180))*3.09*10**16)
                            GM = Gmag -5*(np.log10(distances)) +5
In [116]: plt.figure(figsize = [8, 6])
                            iA = (GM > 0.7) & (GM < 2.4) & (bpmrp > -0.3) & (bpmrp < 0.6)
                            iF = (GM > 2.4) & (GM < 4.2) & (bpmrp > 0.1) & (bpmrp < 1)
                            iG = (GM > 4.2) & (GM < 5.4) & (bpmrp > 0.5) & (bpmrp < 1.2)
                            iK = (GM > 5.4) & (GM < 8.0) & (bpmrp > 0.7) & (bpmrp < 2.1)
                            iM = (GM > 8.0) & (GM < 15.5) & (bpmrp > 1.6) & (bpmrp < 5.0)
                            iW = (GM > 6) & (bpmrp < 0.5)
                            iRG = (bpmrp > 1) & (GM < 1)
                           testx = np.linspace(-1, 6.0, 1000)
                            testline = 3*(testx) +1
                          testline = 3*(testx) +1
plt.plot(bpmrp[iA], GM[iA], ',b', label = 'A stars')
plt.plot(bpmrp[iF], GM[iF], ',c', label = 'F stars')
plt.plot(bpmrp[iG], GM[iG], ',y', label = 'G stars')
plt.plot(bpmrp[iK], GM[iK], ',', color = 'orange', label = 'K stars')
plt.plot(bpmrp[iM], GM[iM], ',r', label = 'M stars')
plt.plot(bpmrp[iW], GM[iW], ',', label = 'White dwarfs', color = 'gray')
plt.plot(bpmrp[iRG], GM[iRG], '.r', label = 'Red giants')
plt.plot(bpmrp[-iA & ~iF & ~iG & ~iK & ~iM & ~iM & ~iRG],\
GM[iA & ~iF & ~iG & ~iK & ~iM & ~
                                                    GM[~iA & ~iF & ~iG & ~iK & ~iM & ~iW & ~iRG]\
                           plt.plot(testx, testline, '--g', label = 'Main Sequence')
                           ax = plt.gca()
                           ax.invert_yaxis()
plt.title('Hertzsprung-Russel Diagram using Gaia data')
                           plt.xlabel(r'sG {BP}-G {RP}$')
                           plt.ylabel(r'$G$')
plt.legend()
                           plt.savefig('/d/www/nikhil/public_html/ASTR5420/images/HertsprungRusselGaia.png')
```

Plot:

(b) What fraction of your stars are M-dwarfs? Now consider the subset of stars within d < 40 pc? Within this subset, what fraction of stars are M-dwarfs? Explain why this second answer is larger and closer to the true value. Code:

```
In [133]: print('Fraction of M dwarfs in the whole set: '\
+str(np.round(100*len(GM[iM])/len(GM), 3)) +'%')
iclose = (distances < 40)
print('Fraction of M dwarfs closer than 40 pc: '\
+str(np.round(100*len(GM[iM & iclose])/len(GM[iclose]), 3)) +'%')
Fraction of M dwarfs in the whole set: 70.029%
Fraction of M dwarfs closer than 40 pc: 70.209%

\frac{M}{N_{total}} = 0.70029, \left(\frac{M}{N_{total}}\right)_{d<40} = 0.70209
```

The second fraction is slightly larger because at closer distances (d < 40 pc), M dwarfs are easier to detect than at larger distances. This is an example of Malmquist bias, which is the effect of exluding intrinsically dimmer objects at larger distances. As a result, the fraction calculated at distances closer than 40 pc is likely closer to the true value of the ratio of M dwarf stars over total stars.

(c) Identify the hot luminous star near G_{BP} - $G_{RP} = -0.500$ and MG = 7.079. Describe its location on the CMD. What are its coordinates? Plug the coordinates into Simbad (https://simbad.u-strasbg.fr/simbad/sim-fcoo). What is its name and type of object? To see the full extent of the object, make sure to select DSS in the image viewer and zoom out to a 1.5° field of view.

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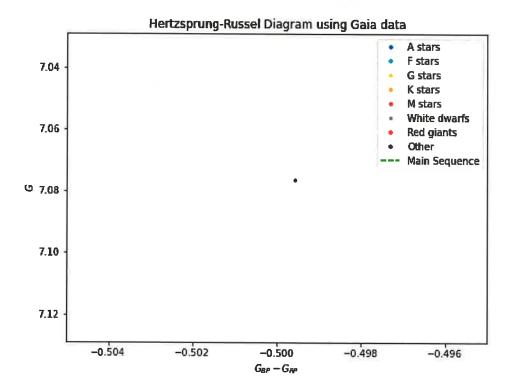
108 109 110

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SH 2-216- Planetary Nebula Plot:



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```
iK = (GM > 5.4) & (GM < 8.0) & (bpmrp > 0.7) & (bpmrp < 2.1)
                                iM = (GM > 8.0) & (GM < 15.5) & (bpmrp > 1.6) & (bpmrp < 5.0)
                                iW = (GM > 6) & (bpmrp < 0.5)
                                iRG = (bpmrp > 1) & (GM < 1)
                               testx = np.linspace(-1, 6.0, 1000)
testline = 3*(testx) +1
                               testline = 3*(testx) +1
plt.plot(bpmrp[iA], GM[iA], '.b', label = 'A stars')
plt.plot(bpmrp[iF], GM[iF], '.c', label = 'F stars')
plt.plot(bpmrp[iG], GM[iG], '.y', label = 'G stars')
plt.plot(bpmrp[iK], GM[iK], '.', color = 'orange', label = 'K stars')
plt.plot(bpmrp[iM], GM[iM], '.r', label = 'M stars')
plt.plot(bpmrp[iM], GM[iW], '.', label = 'White dwarfs', color = 'gray')
plt.plot(bpmrp[iA], GM[iK], '.r', label = 'Red giants')
plt.plot(bpmrp[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iB & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iB & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iB & -iB & -iG & -iK & -iM & -iM & -iM G],\

GM[-iA & -iB & -iB & -iB & -iM & -iM & -iM G],\

GM[-iA & -iB & -iB & -iB & -iM & -iM & -iM G],\

GM[-iA & -iB & -iB & -iM & -iM & -iM & -iM G],\

GM[-iA & -iB & -iB & -iM & -iM & -iM & -iM & -iM G],\

GM[-iA & -iB & -iB & -iB & -iM & -i
                                                           GM[-iA & -iF & -iG & -iK & -iM & -iW & -iRG]\
, 'k', label = 'Other')
                                plt.plot(testx, testline, '--g', label = 'Main Sequence')
                                ax = plt.gca()
                                ax.invert yaxis()
plt.title('Hertzsprung-Russel Diagram using Gaia data')
                                plt.xlabel(r'sG {BP}-G {RP}s')
                                plt.ylabel(r'$65')
plt.legend()
                                 #plt.savefig('/d/www/nikhil/public_html/ASTR5420/images/HertsprungRusselGaia.png')
                                plt.xlim(-.505, -.495)
plt.ylim(7.079+.05, 7.079-0.05)
                                xd = bpmrp+.5
                                yd = GM-7.079
                                 dist = (xd**2 +yd**2)**0.5
                                print('M G: ' +str(GM[np.argmin(dist)]) +', G BP-G RP: ' +str(bpmrp[np.argmin(dist)]))
print('RA: ' +str(ras[np.argmin(dist)]) +'\nDEC: ' +str(decs[np.argmin(dist)]))
print('Object classified as white dwarf by roughly defined parameters.')
                                plt.savefig('/d/www/nikhil/public_html/ASTR5420/images/5cplot.png')
                                M G: 7.076387228469315, G_BP-G_RP: -0.49958706
                                RA: 70.83873938587688
                                 DEC: 46.70156304230519
                                  Object classified as white dwarf by roughly defined parameters. 🦡
```

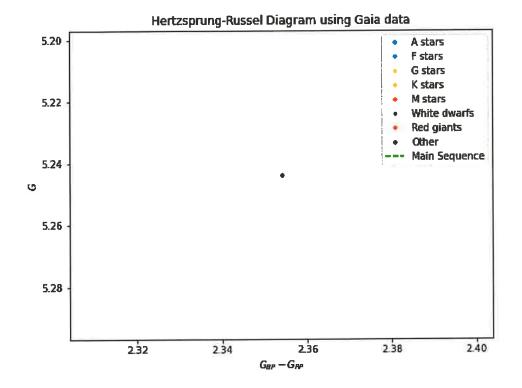


(d) Now identify the cool luminous star near G_{BP} - $G_{RP}=2.354$ and $M_G=5.247$. Describe its location on the CMD. What are its coordinates? Using Simbad, what is its name and type of object? To see the full extent of the region, make sure to zoom out to a 1.0° field of view and compare both 2MASS (near-IR) and DSS (optical). In Simbad, display the references / journal publications for this object. Read the titles of the papers - what is the name of this region?

EM* SR 3- Young Stellar Object. John Star Forming region
This region is called Ophiuchus.

Ophiuchus is jost the constellation

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```
In [147]: plt.figure(figsize = [8, 6])
                            1A = (GM > 0.7) & (GM < 2.4) & (bpmrp > -0.3) & (bpmrp < 0.6)
                            iF = (GM > 2.4) & (GM < 4.2) & (bpmrp > 0.1) & (bpmrp < 1)
                            iG = (GM > 4.2) & (GM < 5.4) & (bpmrp > 0.5) & (bpmrp < 1.2)
                           iK = (GM > 5.4) & (GM < 8.0) & (bpmrp > 0.7) & (bpmrp < 2.1) 
 <math>iM = (GM > 8.0) & (GM < 15.5) & (bpmrp > 1.6) & (bpmrp < 5.0)
                            iW = (GM > 6) & (bpmrp < 0.5)
                            iRG = (bpmrp > 1) & (GM < 1)
                            testx = np.linspace(-1, 6.0, 1000)
                            testline = 3*(testx) +1
                         testline = 3*(testx) +1
plt.plot(bpmrp[iA], GM[iA], '.b', label = 'A stars')
plt.plot(bpmrp[iF], GM[iF], '.c', label = 'F stars')
plt.plot(bpmrp[iG], GM[iG], '.y', label = 'G stars')
plt.plot(bpmrp[iK], GM[iK], '.', color = 'orange', label = 'K stars')
plt.plot(bpmrp[iM], GM[iM], '.r', label = 'M stars')
plt.plot(bpmrp[iM], GM[iM], '.r', label = 'White dwarfs', color = 'gray')
plt.plot(bpmrp[iM], GM[iM], '.r', label = 'Red giants')
plt.plot(bpmrp[-iA & -iF & -iG & -iK & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM & -iM & -iM & -iRG],\

GM[-iA & -iF & -iG & -iK & -iM &
                                                    GM[-iA & -iF & -iG & -iK & -iM & -iW & -iRG]\
                                                          '.k', label = 'Other')
                           plt.plot(testx, testline, '--g', label = 'Main Sequence')
                            ax = plt.gca()
                           ax.invert_yaxis()
                           plt.title('Hertzsprung-Russel Diagram using Gaia data')
                           plt.xlabel(r'sG {BP}-G {RP}s')
                           plt.ylabel(r'sGs')
                           plt.legend(loc = 'upper right')
                           #plt.savefig('/d/www/nikhil/public_html/ASTR5420/images/HertsprungRusselGaia.png')
                           plt.xlim(2.354-0.05, 2.354+0.05)
                           plt.ylim(5.247+.05, 5.247-0.05)
                           xd = bpmrp-2.354
                           yd = GM-5.247
                           dist = (xd**2 + yd**2)**0.5
                          print('M G: ' +str(GM[np.argmin(dist)]) +', G BP-G RP: ' +str(bpmrp[np.argmin(dist)]))
print('RA: ' +str(ras[np.argmin(dist)]) +'\nDEC: ' +str(decs[np.argmin(dist)]))
                          print('Object is unclassified by loose categorization.')
plt.savefig('/d/www/nikhil/public_html/ASTR5420/images/5dplot.png')
                           M_G: 5.243937461800611, G_BP-G_RP: 2.3540802
                           RA: 246.53881047417445
                           DEC: -24.570171196663942
                           Object is unclassified by loose categorization.
```

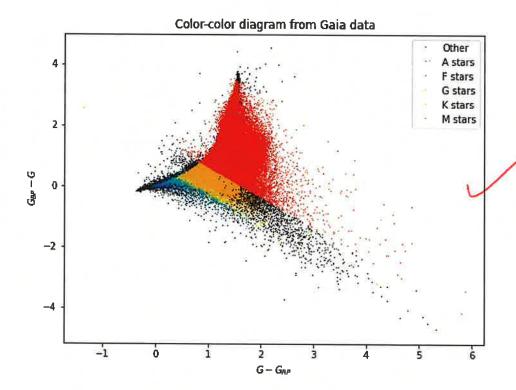
Is above MS (see plot) => pre-MS



(e) Finally, plot a color-color diagram (G_{BP} -G versus G- G_{RP}). Describe the MS relation on this diagram.

Plot:

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Code:

```
In [148]: plt.figure(figsize = [8, 6])
           plt.plot(gmrp[-iA & ~iF & ~iG & ~iK & ~iM],\
                     bpmg[~iA & ~iF & ~iG & ~iK & ~iM]\
                            ', label = 'Other')
           plt.plot(gmrp[iA], bpmg[iA],
plt.plot(gmrp[iF], bpmg[iF],
                                             ,b',
                                                  label = 'A stars')
                                                  label = 'F stars')
           plt.plot(gmrp[iG], bpmg[iG],
                                                  label = 'G stars')
           plt.plot(gmrp[iK], bpmg[iK],
plt.plot(gmrp[iM], bpmg[iM],
                                                 color = 'orange', label = 'K stars')
                                             ,r', label = 'M stars')
           plt.legend()
           plt.ylabel(r'$G_{BP}-G$')
           plt.xlabel(r'$G-G {RP}$')
           plt.title('Color-color diagram from Gaia data')
           plt.savefig('/d/www/nikhil/public_html/ASTR5420/images/colorcolorhrdiagram.png')
```

Main sequence relation appears to be a semi-U shape with the redder stars appearing in the top-right and the bluer stars in the bottom left. This makes sense as this is a color-color diagram and the positive axis represents redder colors and the negative direction represents bluer colors.

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