ASTRO C
207 Radiative Processes in Astrophysics Lydia Lee

Problem Set 8

1. Saha and the Redshift of Recombination

- electron de Broglie wavelength: $\lambda = \frac{h}{\sqrt{2\pi m k_B T}}$
- electron degeneracy: $\xi \equiv n_e \lambda^3 \approx 4.1 (10^{-16}) n_e T^{-\frac{3}{2}}$
- $\gamma \equiv \ln\left(\frac{1}{\xi}\right) \approx 10-30$
- Assume $U_x(T) \propto T^0, \approx 1$

(1)

From Saha

$$\frac{n_{j+1}n_e}{n_j} = \left(\frac{2\pi m_e k_B T}{h^2}\right)^{\frac{3}{2}} \left(\frac{2U_{j+1}(T)}{U_j(T)}\right) e^{-\frac{\chi}{k_B T}}$$

$$\frac{n_{j+1}}{n_j} = \frac{1}{\xi} \frac{2U_{j+1}(T)}{U_j(T)} e^{-\frac{\chi}{k_B T}}$$

$$0 \approx \gamma - \frac{\chi}{k_B T} \longleftarrow n_{j+1} \approx n_j$$

$$k_B T \approx \frac{\chi}{\gamma}$$

(2)

$$R \equiv \frac{n_{j+1}}{n_j} \approx \frac{1}{\xi} e^{-\frac{\chi}{k_B T}}$$

$$\frac{d \ln(R)}{d \ln(T)} = \frac{d \ln(R)}{dT} \left(\frac{d \ln(T)}{dT} \right)^{-1}$$

$$\approx \frac{d}{dT} \left[\gamma - \frac{\chi}{k_B T} \right] \cdot T$$

$$= \left\{ \frac{d}{dT} \left[\ln \left(\frac{T^{\frac{3}{2}}}{4.1(10^{-16}n_e)} \right) \right] + \frac{\chi}{k_B T^2} \right\} \cdot T$$

$$= \frac{3}{2} + \frac{\chi}{k_B T}$$

$$\approx \frac{3}{2} + \frac{\chi}{\chi/\gamma}$$

$$= \frac{3}{2} + \gamma \leftarrow \gamma \approx 10-30$$

$$\approx \gamma$$

$$\frac{\Delta T}{T} \approx \left(\frac{d \ln(R)}{d \ln(T)} \right)^{-1} \approx \frac{1}{\gamma}$$

(3)

$$\frac{n_{j+1}}{n_j} = \frac{g_{j+1}}{g_j} e^{-\frac{E_{j+1} - E_j}{k_B T}}$$

$$\approx e^{-\gamma \frac{E_{j+1} - E_j}{\chi}}$$

$$\approx e^{-\gamma}$$

$$\ll 1 \longleftarrow \gamma \gg 1$$

2. Pulsar Dispersion Measure

- (1) i. Find data where, for a given frequency, there is a clear time at which it peaks. The selection criterion was that the difference between the maxima was some percentage of the absolute maximum for that frequency.
 - ii. Using

$$t_1 - t_0 = \frac{e^2}{2\pi m_e c} (DM) \left(\frac{1}{v_1^2} - \frac{1}{v_0^2} \right)$$

estimate the dispersion measure between each pairing of "good" data points. Depending on the severity of the selection criteria, it's possible to end up with a negative calculated dispersion ratio; increasing the stringency of criteria for what constitutes a clean peak resolves this.

iii. Find the geometric mean of the data.

$$DM \approx 10^{21} \frac{electrons}{cm^2}$$

(2) •
$$n_e \approx 0.03 \frac{\text{electrons}}{\text{cm}^3}$$

$$DM = \int_0^d n_e d\ell \approx n_e d$$

 $d \approx 10,000$ pc—roughly the distance to the center of the Milky Way.

$$v_p = \frac{1}{2\pi} \sqrt{\frac{4\pi n_e e^2}{m_e}}$$

 $v_p \approx 1.5 \text{kHz}$ —quite a lot lower than the observation frequency.

(4)
$$\sigma_{\text{Thomson}} = \frac{8\pi}{3} \left(\frac{e^2}{m_e c^2} \right)^2$$

 $\tau_{\rm Thomson} \approx 7(10^{-4}) \ll 1$ so Thomson scattering isn't significant

 $\tau_{\text{Thomson}} = n_e \sigma_{\text{Thomson}} d$

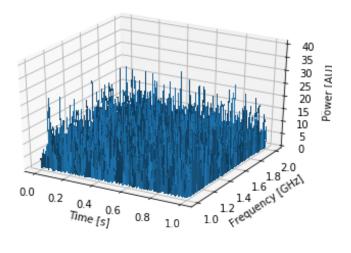
q_pulsar_dispersion_measure_ipynb

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```
In [1]: %matplotlib inline
        import numpy as np
        import matplotlib.pyplot as plt
        from matplotlib import cm
        from mpl_toolkits.mplot3d import Axes3D
       from scipy.stats.mstats import gmean
       import time
        import pylab as pl
       from IPython import display
       from pprint import pprint
In [2]: float_formatter = "{0:0.2e}".format
        np.set_printoptions(formatter={'float_kind':float_formatter})
In [3]: q = 5e-10 \# esu, electron charge
       m_e = 1e-27 \# g, electron mass
       c = 3e10 # cm/s, speed of light
        cm2pc = 3.24077929e-19
        prefactor = q**2 / (2*np.pi * m_e * c)
        def estimate_DM(nu0, nu1, t0, t1):
            Inputs:
               nu0/1: Float. Frequencies (in Hz) of peaks.
                t0/1: Float. Time of arrival of the peaks.
               DM: An estimate of the dispersion measure (i.e. column density of electrons)
               in electrons/cm^2.
            prefactor = 4140 * 1e6**2 * cm2pc
           prefactor = q**2 / (2*np.pi * m_e * c)
           DM = (t1-t0) / (prefactor * (1/nu1**2 - 1/nu0**2))
           return DM
  2.1
```

1

```
In [4]: fname_pulsar = './pulsar.dat'
        # Getting frequencies
        with open(fname_pulsar) as f:
            column_headers = f.readlines()[0].strip().split('] [')
        nu_vec = np.array([1e9*float(nu_str.replace('GHz', '').replace(']', '')) for nu_str in
        # Data
        data_raw = np.loadtxt(fname_pulsar)
       t_vec = data_raw[:,0]
        t_mesh, nu_mesh = np.meshgrid(t_vec, nu_vec/1e9)
        data_pulsar = data_raw[:,1:].transpose()
        # Plot the surface
       fig, ax = plt.subplots(subplot_kw={"projection": "3d"})
        surf = ax.plot_surface(t_mesh, nu_mesh, data_pulsar, linewidth=0, antialiased=False)
        ax.set_xlabel('Time [s]')
        ax.set_ylabel('Frequency [GHz]')
        ax.set_zlabel('Power [AU]')
        # for i in range(0, 100, 10):
             plt.figure()
             plt.plot(t_vec, data_pulsar[i])
             plt.title(f'{nu_vec[i]} GHz')
        # Data too noisy for cross-correlations of raw data
        # for d in data_pulsar[11:20]:
             plt.figure()
             cross_corr = np.convolve(data_pulsar[0], d[::-1], mode='same')
            cross_corr_norm = cross_corr/max(cross_corr)
             plt.plot(t_vec, cross_corr_norm)
        # print(len(column_headers))
        # print(len(set(column_headers)))
        # print(column_headers.index('1.022 GHz'))
        # print(column_headers[23])
        # print(column_headers[24])
Out[4]: Text(0.5,0,'Power [AU]')
```

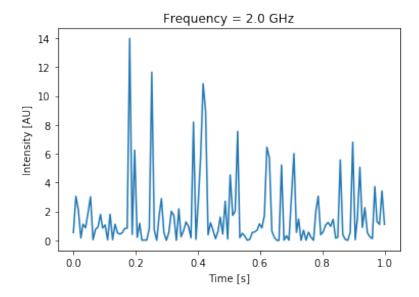


```
In [5]: # Indices of top 2 peaks for a given frequency
        idx_pk_dict = dict()
        for i,nu in enumerate(nu_vec):
            idx_pk = np.argsort(data_pulsar[i])[::-1][0:2]
            idx_pk_dict[i] = idx_pk
        # Filter to get (indices of) frequencies which have a single "clearly" distinguishable
        # (defined by a minimum difference in two max peaks' respective heights)
        # The keys are the indices matching up to nu_vec.
        # The threshold for what constitutes a single clearly distinguishable peak
        diff_threshold_dict = {i:np.max(data_pulsar[i])*0.6 for i,_ in enumerate(nu_vec)}
        # The difference between peaks at a given time
        diff_pk_dict = {i:(data_pulsar[i][idx_pk[0]]-data_pulsar[i][idx_pk[1]]) for i,idx_pk i1
        # The indices (in nu_vec) for which the max is sufficiently large relative to other til
        idx pk visible vec = [i for i,v in diff pk dict.items() if v>=diff threshold dict[i]]
        print(len(idx_pk_visible_vec))
        # print(idx_pk_visible_vec)
```

Initial Estimate Method 1: Find the Time of Peak for a Given Frequency

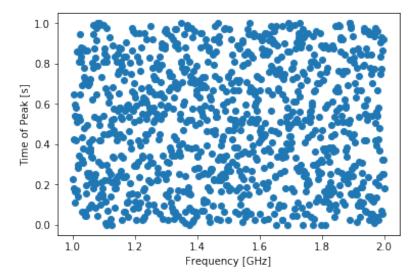
3

7



4

```
t0 = t_vec[np.argmax(data_pulsar[idx0])]
               t1 = t_vec[np.argmax(data_pulsar[idx1])]
               DM_array[i][j] = estimate_DM(nu0, nu1, t0, t1)
       #
                 print(f'{\{(0:2e\})''.format(nu0)\}/{\{(0:2e\})''.format(nu1)\}} \rightarrow \{DM_array[i][j]\}')
       #
                 if DM_array[i][j] > 0:
       #
                     plt.figure()
       #
                     for n in (idx0, idx1):
                        plt.legend()
       #
                        plt.xlabel('Time [s]')
                         plt.ylabel('Power [AU]')
       pprint(DM_array)
       DM_array_frequency = np.array(DM_array, copy=True)
array([[nan, nan, nan, nan, nan, nan, nan],
      [5.51e+20, nan, nan, nan, nan, nan, nan],
      [-1.57e+19, -5.00e+19, nan, nan, nan, nan, nan],
      [1.60e+21, 1.66e+21, 3.27e+22, nan, nan, nan, nan],
      [3.21e+20, 3.12e+20, 1.08e+21, -3.13e+21, nan, nan, nan],
      [6.14e+19, 4.50e+19, 1.63e+20, -2.23e+21, -1.12e+21, nan, nan],
       [6.59e+20, 6.63e+20, 1.55e+21, -7.40e+20, 2.19e+21, 4.64e+23, nan]])
  Initial Estimate Method 2: Use Prior Knowledge (t=0 @ Start of Pulse Arrival)
In [8]: t_pk_vec = [t_vec[np.argmax(data_pulsar[i])] for i,_ in enumerate(nu_vec)]
       plt.figure()
       plt.plot(nu_vec*1e-9, t_pk_vec, 'o')
       plt.xlabel('Frequency [GHz]')
       plt.ylabel('Time of Peak [s]')
Out[8]: Text(0,0.5,'Time of Peak [s]')
```



There's not enough rhyme or reason to the above to comfortably throw something like least squares at the raw data.

```
In [9]: DM_vec_frequency = []
    for idx in idx_pk_visible_vec:
        nu0 = np.max(nu_vec)
        nu1 = nu_vec[idx]
        t0 = t_vec[np.argmax(data_pulsar[-1])]
        t1 = t_vec[np.argmax(data_pulsar[idx])]
        DM_vec_frequency.append(estimate_DM(nu0, nu1, t0, t1))
        DM_vec_frequency = np.array(DM_vec_frequency)
        pprint(DM_vec_frequency)

array([7.53e+20, 7.60e+20, 1.68e+21, -3.97e+20, 2.37e+21, 1.82e+22, 3.21e+21])
```

General curiosity: You could rotate this and treat it as tracking a moving object (where the object's position is the the frequency of the peak) with variable acceleration, then throw an extended Kalman filter at it to figure out the DM with the initial estimate calculated using either of the methods above

```
In [10]: n_e = .03 # electrons/cm^2
```

 ${\it \# Filtering out intentionally introduced NaNs and impossible DMs}$

6

2.2

```
# caused by noise overwhelming the real pulse
        DM_vec = DM_array_frequency.flatten() # DM_array_frequency.flatten()/DM_vec_frequency.
        DM_vec = DM_vec[~np.isnan(DM_vec)]
        DM_vec = DM_vec[DM_vec > 0]
        DM = gmean(DM_vec) # np.mean/median(DM_vec)
        print(f'DM Estimate:\t\t{"{0:2e}}".format(DM)} electrons/cm^2')
        d = DM/n_e \# cm
        DM Estimate:
                           1.032229e+21 electrons/cm<sup>2</sup>
Distance Estimate:
                       3.440765e+22 cm = 11150.758494193946 pc
  The value for distance seems high.
  2.3
In [11]: omega_p = np.sqrt(4*np.pi * n_e * q**2 / m_e)
        print(f'Plasma Frequency: {"{0:2e}".format(omega_p/(2*np.pi))} Hz')
Plasma Frequency: 1.545097e+03 Hz
  2.4
In [12]: sigma_Thomson = 8*np.pi/3 * (q**2/(m_e*c**2))**2
        {\tt alpha\_Thomson} \; = \; {\tt n\_e} \; * \; {\tt sigma\_Thomson}
        tau\_Thomson = alpha\_Thomson * d
        print(f'Optical Depth: {tau_Thomson}')
Optical Depth: 0.0006672518908541399
```