

전산천문학 HW6 Solution

June 24, 2020

```
[1]: import numpy as np
import matplotlib.pyplot as plt
from scipy import linalg
from scipy.odr import *

import warnings
warnings.filterwarnings('ignore')
```

1.

```
[2]: lam=np.array([[0.      ,88.356E4,32.661E4,4931.1,0.      ],
                  [88.356E4,0.      ,51.814E4,4958.9,2321.0],
                  [32.661E4,51.814E4,0.      ,5006.9,2331.4],
                  [4931.1 ,4958.9 ,5006.9 ,0.      ,4363.2],
                  [0.      ,2321.0 ,2331.4 ,4363.2,0.      ]]).T

Eng=np.array([[0.      ,0.014,0.038,2.514,5.300],
              [0.014,0.      ,0.024,2.500,5.314],
              [0.038,0.024,0.      ,2.476,5.318],
              [2.514,2.500,2.476,0.      ,2.842],
              [5.300,5.314,5.318,2.842,0.      ]]).T

A=np.array([[0.      ,2.7E-05,3.1E-11,1.7E-06,0.      ],
            [0.      ,0.      ,9.7E-05,6.8E-03,2.3E-01],
            [0.      ,0.      ,0.      ,2.0E-02,6.1E-04],
            [0.      ,0.      ,0.      ,0.      ,1.6      ],
            [0.      ,0.      ,0.      ,0.      ,0.      ]]).T

Om1=np.array([[0.      ,0.55,0.27,0.254,0.032],
              [0.      ,0.      ,1.29,0.763,0.097],
              [0.      ,0.      ,0.      ,1.272,0.161],
              [0.      ,0.      ,0.      ,0.      ,0.58 ],
              [0.      ,0.      ,0.      ,0.      ,0.      ]]).T

b0=np.array([0.,0.,0.,0.,1.])
omega=np.array([1.,3.,5.,5.,1.])
```

(a)

```
[3]: def 02plus_pop(Ne,T): # Ne:cm-3, T:K
    Ne,T=np.atleast_1d(Ne),np.atleast_1d(T)
    N1,N2=len(Ne),len(T)
    pop=np.zeros([N1,N2,5])
    for jt in range(N1):
        ne=Ne[jt]
        for it in range(N2):
            t=T[it]/1e4
            q=np.zeros([5,5])
            M=np.zeros([5,5])
            for i in range(4):
                for j in range(i,5): #i<j
                    q[j,i]=8.629e-8*Om1[j,i]/(np.sqrt(t)*omega[j])

            for i in range(1,5):
                for j in range(i): #i>j
                    coef2 = np.exp(-1.1605*Eng[i,j]/t)
                    q[j,i]=q[i,j]*omega[i]/omega[j]*coef2

            for i in range(4):
                qsum,asum=0,0
                for k in range(i): # k<i
                    M[i,k]=q[k,i]*ne
                    qsum += q[i,k]
                    asum += A[i,k]

                for k in range(i+1,5): #k>i
                    M[i,k]=q[k,i]*ne + A[k,i]
                    qsum += q[i,k]
                k=i
                M[i,k]=-(qsum*ne + asum)

            i=4
            for k in range(5):
                M[i,k]=1. # 전체의 합이 1이 되도록 함
            LU=linalg.lu_factor(M) # calculate pivoted LU decomposition
            x =linalg.lu_solve(LU, b0) # solve Ax=b for given LU factorization
            pop[jt,it,:]=x

    return pop #pop[i,j,k]=i번째 Ne, j번째 T에서의 n_k

[4]: Ne1=[1.0e2,1.0e2,1.0e1]
    T1=[1.0e4,2.0e4,1.0e4]

    for i in range(3):
        print ('Ne : {:.1e} cm-3, T : {:.1e} K'.format(Ne1[i],T1[i]))
```

```

    pop_i = O2plus_pop(Ne1[i],T1[i])[0,0]
    print ('=> (n1,n2,n3,n4,n5) = ({:.4e}, {:.4e}, {:.4e}, {:.4e}, {:.4e})'.
    →format(pop_i[0],pop_i[1],pop_i[2],pop_i[3],pop_i[4]))

```

```

Ne : 1.0e+02 cm-3, T : 1.0e+04 K
=> (n1,n2,n3,n4,n5) = (7.8287e-01, 1.9181e-01, 2.5313e-02, 4.4603e-06,
3.2121e-10)
Ne : 1.0e+02 cm-3, T : 2.0e+04 K
=> (n1,n2,n3,n4,n5) = (8.2436e-01, 1.5543e-01, 2.0204e-02, 1.3765e-05,
4.9272e-09)
Ne : 1.0e+01 cm-3, T : 1.0e+04 K
=> (n1,n2,n3,n4,n5) = (9.7246e-01, 2.5125e-02, 2.4131e-03, 4.4432e-07,
3.2155e-11)

```

(b)

```

[5]: def O2plus_R(Ne,T):
    Ne,T=np.atleast_1d(Ne),np.atleast_1d(T)
    N1,N2=len(Ne),len(T)
    Ratio=np.zeros([N1,N2])
    pop=O2plus_pop(Ne,T)
    for jt in range(N1):
        for it in range(N2):
            x=pop[jt,it]
            Ratio[jt,it]=(x[3]/x[4])*(A[3,2]*lam[3,2]+A[3,1]*lam[3,1])/
            →(A[4,3]*lam[4,3])
    return Ratio # Ratio[i,j]=i번째 Ne, j번째 T에서의 R

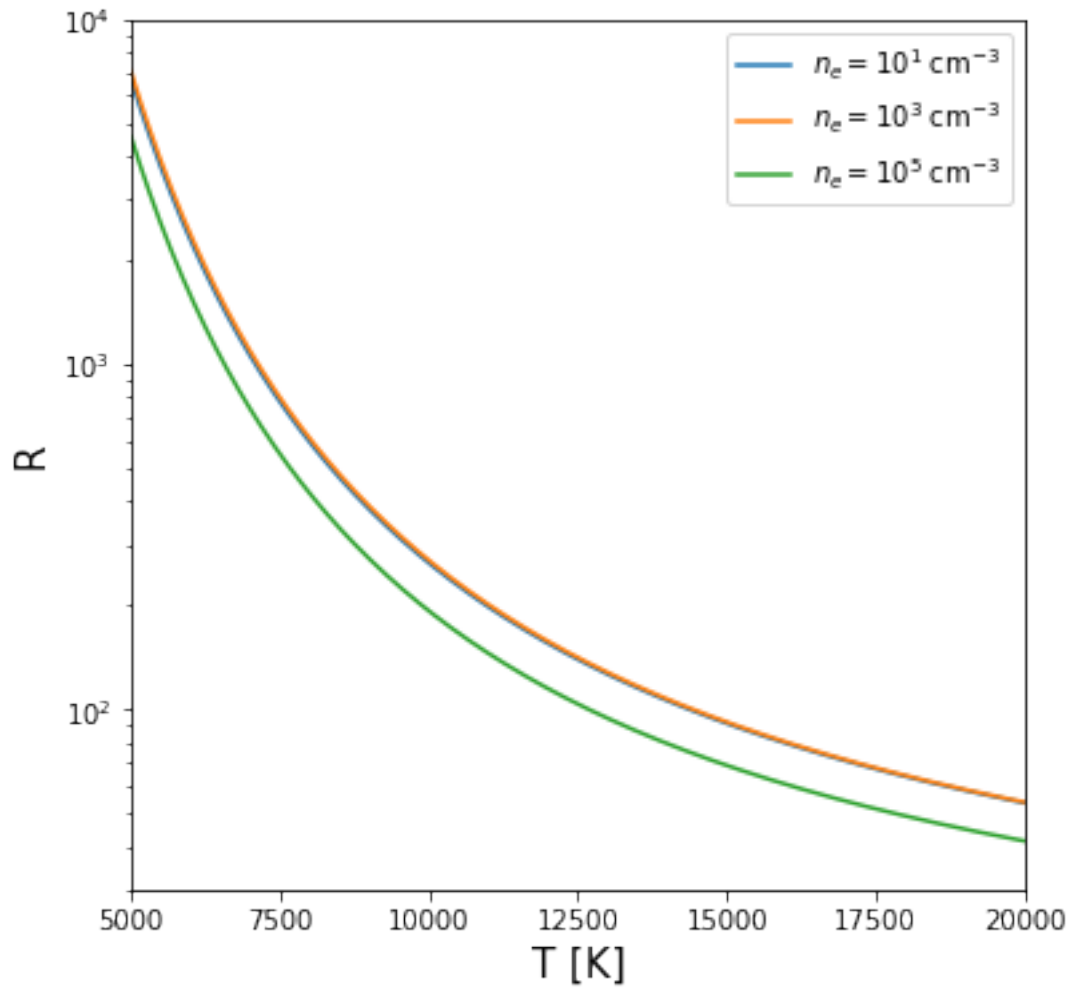
```

```

[6]: Ne,Temp=[1e1,1e3,1e5],np.linspace(5e3,2e4,400)
Ratio=O2plus_R(Ne,Temp)

plt.figure(figsize=(6,6))
plt.semilogy(Temp,Ratio[0,:],label=r'$n_e=10^1$ $\mathrm{cm}^{-3}$')
plt.semilogy(Temp,Ratio[1,:],label=r'$n_e=10^3$ $\mathrm{cm}^{-3}$')
plt.semilogy(Temp,Ratio[2,:],label=r'$n_e=10^5$ $\mathrm{cm}^{-3}$')
plt.xlabel(r'T [K]',fontsize=15)
plt.ylabel('R',fontsize=15)
plt.xlim(5000.,20000.)
plt.xticks(np.arange(5000.,20001.,2500.))
plt.ylim(3.0e1,1.0e4)
plt.legend()
plt.show()

```

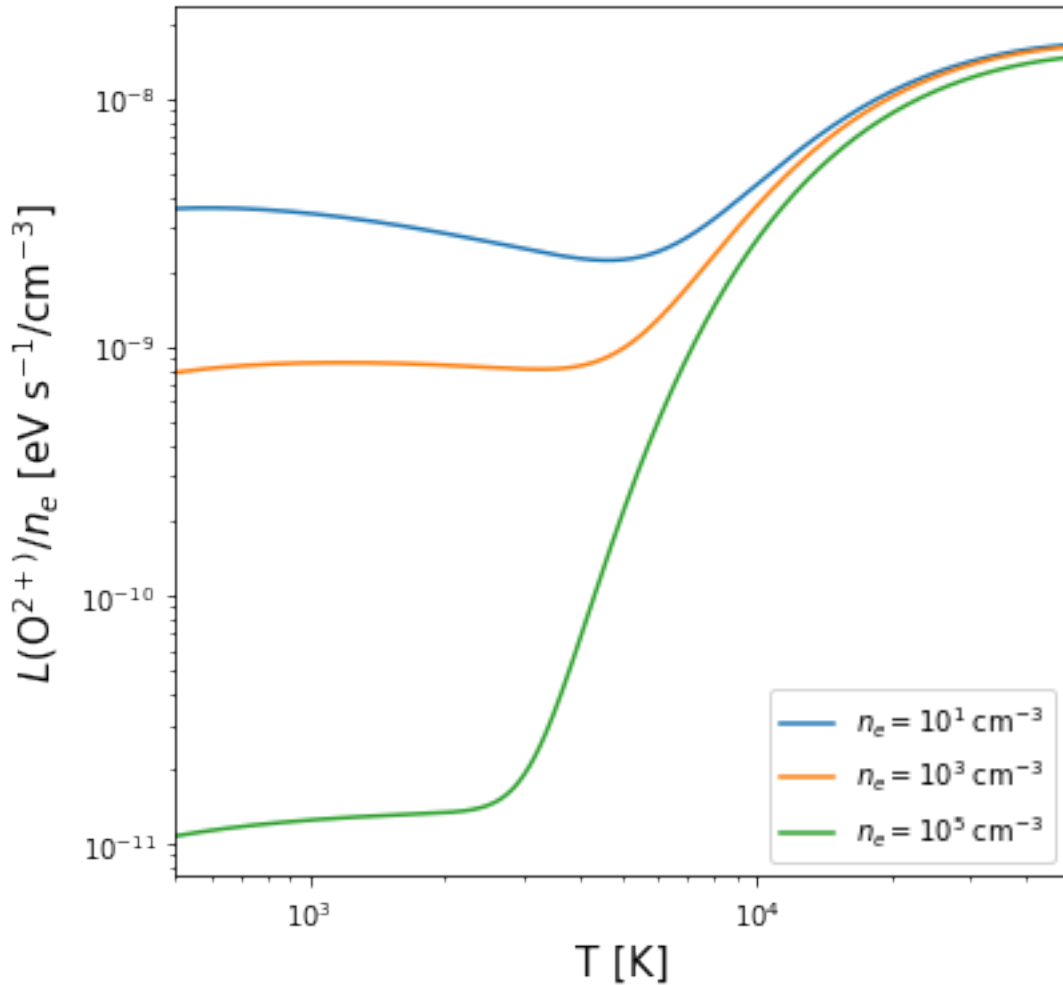


(c)

```
[7]: def O2plus_L(Ne,T):
    Ne,T=np.atleast_1d(Ne),np.atleast_1d(T)
    N1,N2=len(Ne),len(T)
    Loss=np.zeros([N1,N2])
    pop=O2plus_pop(Ne,T)
    for jt in range(N1):
        for it in range(N2):
            x=pop[jt,it]
            for i in range(1,5):
                for k in range(0,i):
                    Loss[jt,it]+=x[i]*A[i,k]*Eng[k,i]
    return Loss # Loss[i,j]=i번째 Ne, j번째 T에서의 L(O2+)
```

```
[8]: Ne,Temp=[1e1,1e3,1e5],np.logspace(2+np.log10(5.),4.+np.log10(5.),1000)
Loss=O2plus_L(Ne,Temp)

plt.figure(figsize=(6,6))
plt.loglog(Temp,Loss[0,:]/Ne[0],label=r'$n_e=10^1$ $\mathrm{cm}^{-3}$')
plt.loglog(Temp,Loss[1,:]/Ne[1],label=r'$n_e=10^3$ $\mathrm{cm}^{-3}$')
plt.loglog(Temp,Loss[2,:]/Ne[2],label=r'$n_e=10^5$ $\mathrm{cm}^{-3}$')
plt.xlabel(r'T [K]',fontsize=15)
plt.ylabel(r'$L(\mathrm{O}^{2+})/n_e$ [eV s$^{-1}$ / cm$^{-3}$]',fontsize=15)
plt.xlim(500.,50000.)
# plt.ylim(3.0e1,1.0e4)
plt.legend(loc=4)
plt.show()
```



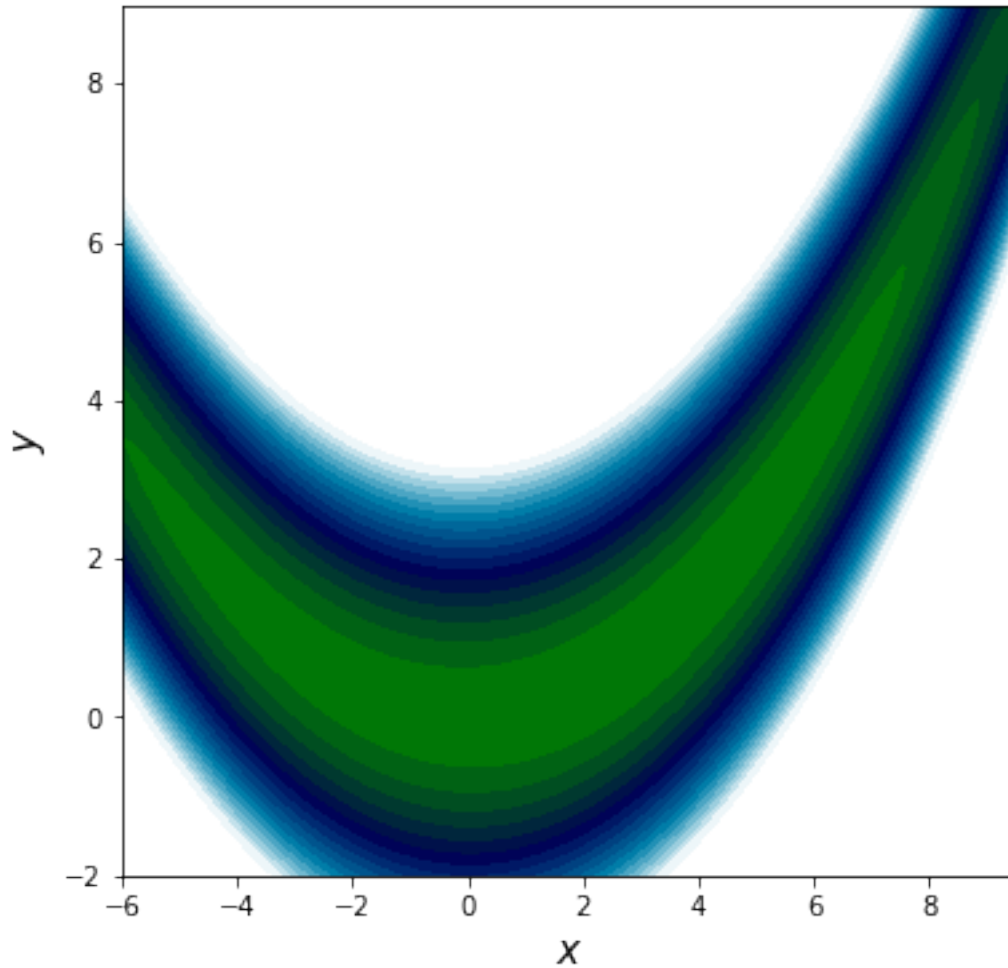
2.

```
[9]: def func(x):  
    return ((x[0]**2.)-10.*x[1])**2.+2.*(x[1]**2.)-3.*x[0]  
  
    def deriv(x):  
        f1 = 4.*x[0]*((x[0]**2.)-10.*x[1])-3.  
        f2 = -20.*((x[0]**2.)-10.*x[1])+4.*x[1]  
        return np.array([f1,f2])
```

```
[10]: def golden2D(x,dr,tol):  
    R,err,loop=0.61803399,10.,-1  
    a,b=x-dr,x+dr # initialize the range  
    while(err > TOL):  
        loop += 1  
        x1,x2=b-R*(b-a),a+R*(b-a)  
        f1,f2=func(x1),func(x2)  
        if (f2>f1):  
            b=x2  
        else:  
            a=x1  
        err=np.sqrt(sum((a-b)**2))  
    xmin=a  
    return xmin,err,loop
```

```
[11]: x_gd,y_gd=np.linspace(-6.5,9.5,500),np.linspace(-2.5,9.,500)  
X, Y =np.meshgrid(x_gd,y_gd)  
E=((X**2.)-10.*Y)**2.+2.*(Y**2.)-3.*X  
dmax,dmin=1000.,-10.  
levels=(dmax-dmin)*np.arange(20)/19.+dmin  
cmap=plt.cm.ocean  
  
plt.figure(figsize=(6,6))  
plt.contourf(x_gd, y_gd, E, levels,cmap=cmap)  
plt.xlim(-6.,9.5)  
plt.ylim(-2.,9.)  
plt.xlabel(r'$x$',fontsize=15)  
plt.ylabel(r'$y$',fontsize=15)
```

```
[11]: Text(0, 0.5, '$y$')
```



(a) The Steepest Descent Method

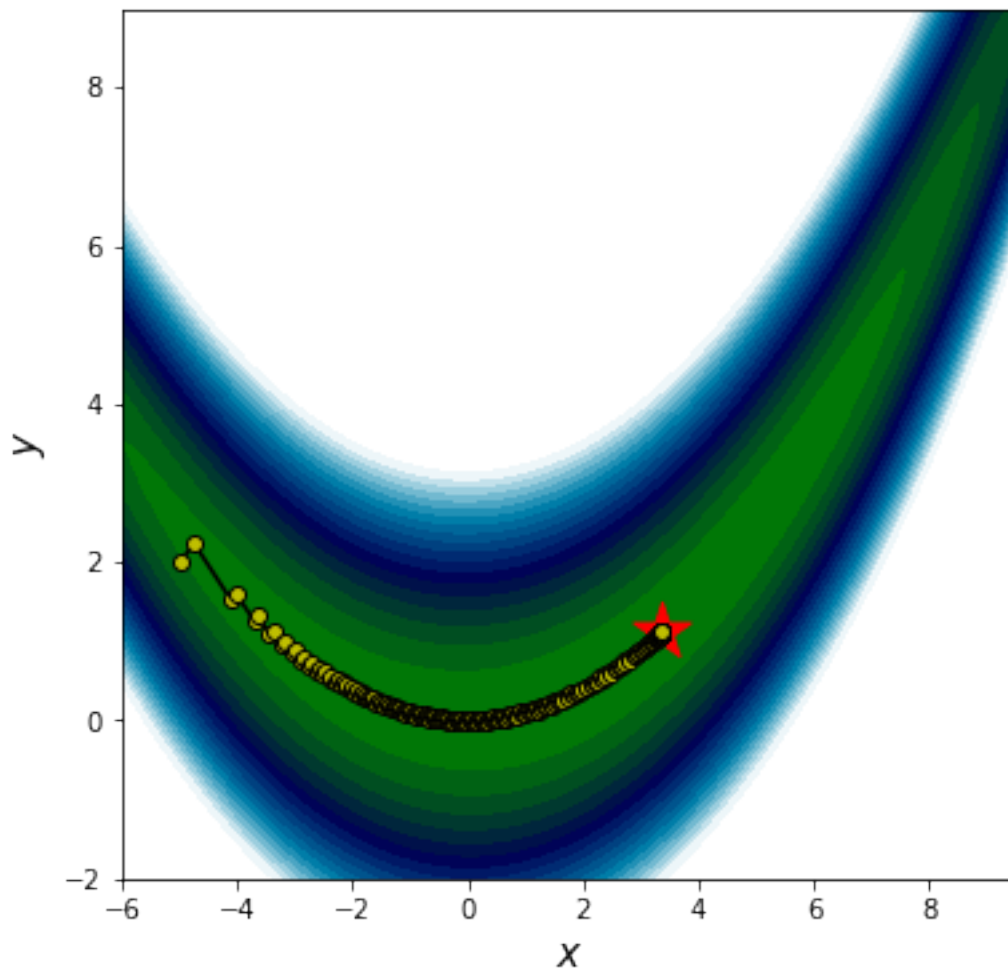
```
[12]: err, TOL = 1.0, 1.0e-6
x=np.array([-5.,2.])
xt,yt=np.array([x[0]]),np.array([x[1]])

loop=0
while err>TOL:
    loop+=1
    dr = -deriv(x)/np.sqrt(np.sum(np.power(deriv(x),2.)))
    xmin,err_lp,loop_lp = golden2D(x,dr,1.0e-5)
    err = np.sqrt(sum((xmin-x)**2.))
    x = xmin
    xt,yt=np.append(xt,x[0]),np.append(yt,x[1])
print (x,loop)
```

[3.36926465 1.11293537] 220

```
[13]: plt.figure(figsize=(6,6))
plt.plot(xt,yt,marker='o',c='k',markerfacecolor='y')
plt.contourf(x_gd, y_gd, E, levels,cmap=cmap)
plt.xlim(-6.,9.5)
plt.ylim(-2.,9.)
plt.xlabel(r'$x$',fontsize=15)
plt.ylabel(r'$y$',fontsize=15)
plt.scatter(x[0],x[1],c='r',marker='*',s=500)
```

[13]: <matplotlib.collections.PathCollection at 0x1c357329fd0>



(b) Powell's Method

```
[14]: P0=np.array([-5.,2.])
      xt,yt=np.array([P0[0]]),np.array([P0[1]])

      u=np.array([[1.,0.],[0.,1.]]) # Cartesian unit vectors
      ut = u

      loop=0
      err, TOL = 1.0, 1.0e-6
      while err>TOL:
          loop+=1
          u1,u2=u[0],u[1] # Initialize set of directions

          #Step 2
          P1,err1,loop1 = golden2D(P0,u1,1.0e-6)
          P2,err2,loop2 = golden2D(P1,u2,1.0e-6)

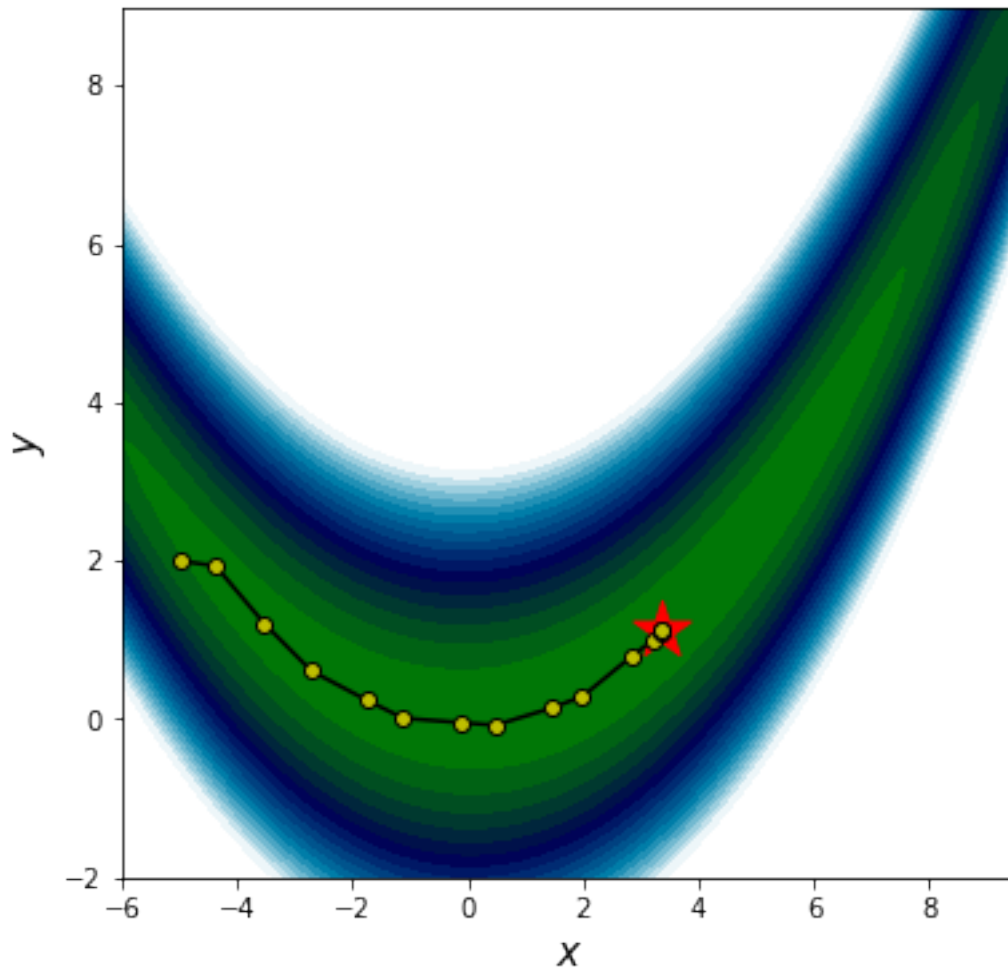
          #Step 3
          u1 = u2
          u2 = (P2-P0)/np.sqrt(sum((P2-P0)**2.)) # Unit vector로 변환
          u = np.vstack((u1,u2))
          ut = np.vstack((ut,u2))

          #Step 4
          newP0,err_lp,loop_lp = golden2D(P2,u2,1.0e-6)
          err = np.sqrt(sum((newP0-P0)**2.))
          P0 = newP0
          xt,yt=np.append(xt,P0[0]),np.append(yt,P0[1])
      print (P0,loop)
```

[3.36933163 1.11298012] 14

```
[15]: plt.figure(figsize=(6,6))
      plt.plot(xt,yt,marker='o',c='k',markerfacecolor='y')
      plt.contourf(x_gd, y_gd, E, levels,cmap=cmap)
      plt.xlim(-6.,9.5)
      plt.ylim(-2.,9.)
      plt.xlabel(r'$x$',fontsize=15)
      plt.ylabel(r'$y$',fontsize=15)
      plt.scatter(x[0],x[1],c='r',marker='*',s=500)
```

[15]: <matplotlib.collections.PathCollection at 0x1c359506be0>



3.

1)

```
[16]: def func(x):
        return 100*(x[1]-x[0]**2)**2 + (1-x[0])**2+100*(x[2]-x[1]**2)**2 +
        →(1-x[2])**2

    def deriv(x):
        f1=-400*x[0]*(x[1]-x[0]**2)-2*(1-x[0])
        f2=200*(x[1]-x[0]**2)-400*x[1]*(x[2]-x[1]**2)
        f3=200*(x[2]-x[1]**2)-2*(1-x[2])
        return np.array([f1,f2,f3])

    err,TOL=1.,1.e-6
    x=np.array([0.,2.,1.] )
```

```

xt,yt,zt=[x[0]], [x[1]], [x[2]]

loop=0
while err>TOL:
    loop+=1
    if loop==1:
        g_now=-deriv(x)
        dr_now=g_now/np.sqrt(sum(g_now**2))
        xmin,err_lp,loop_lp=golden2D(x,dr_now,1.e-5)
        err=np.sqrt(sum((xmin-x)**2))
        x=xmin
        xt,yt,zt=np.append(xt,x[0]),np.append(yt,x[1]),np.append(zt,x[2])
        g_prev=g_now # 이전 단계의 gradient 저장
        dr_prev=dr_now # 이전 단계의 direction 저장
    else:
        g_now=-deriv(x)
        l=sum(g_now**2)/sum(g_prev**2) # Fletcher-Reeves Method
        # l=sum(g_now*(g_now-g_prev))/sum(g_prev**2) # Polak-Ribière Method
        dr_now=g_now+l*dr_prev
        dr_now=dr_now/np.sqrt(sum(dr_now**2)) # dr을 단위벡터로 변환
        xmin,err_lp,loop_lp=golden2D(x,dr_now,1.e-6)
        err=np.sqrt(sum((x-xmin)**2))
        x=xmin
        xt,yt,zt=np.append(xt,x[0]),np.append(yt,x[1]),np.append(zt,x[2])
        g_prev=g_now
        dr_prev=dr_now
print (xmin,loop)

```

```
[0.99991815 0.99989568 0.9998397 ] 154
```

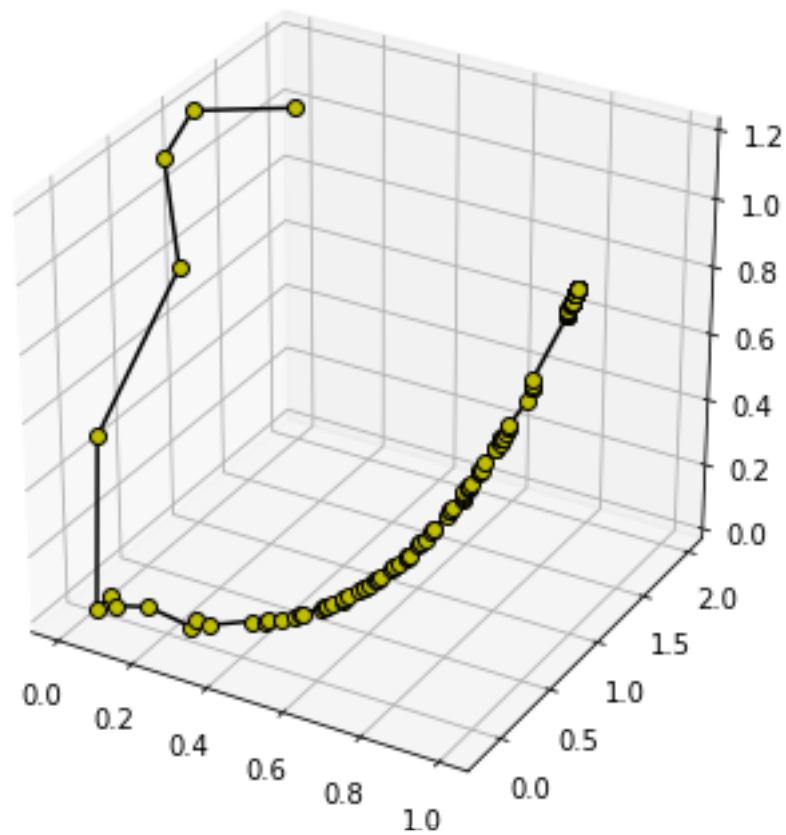
```

[17]: from mpl_toolkits import mplot3d

fig = plt.figure(figsize=(6,6))
ax = plt.axes(projection='3d')
ax.plot3D(xt,yt,zt,c='k',marker='o',markerfacecolor='y')

```

```
[17]: [<mpl_toolkits.mplot3d.art3d.Line3D at 0x1c358f3c6a0>]
```



2) scipy 사용

```
[18]: from scipy.optimize import minimize
x0=np.array([0.,2.,1.])
xt,yt,zt=[x0[0]], [x0[1]], [x0[2]]

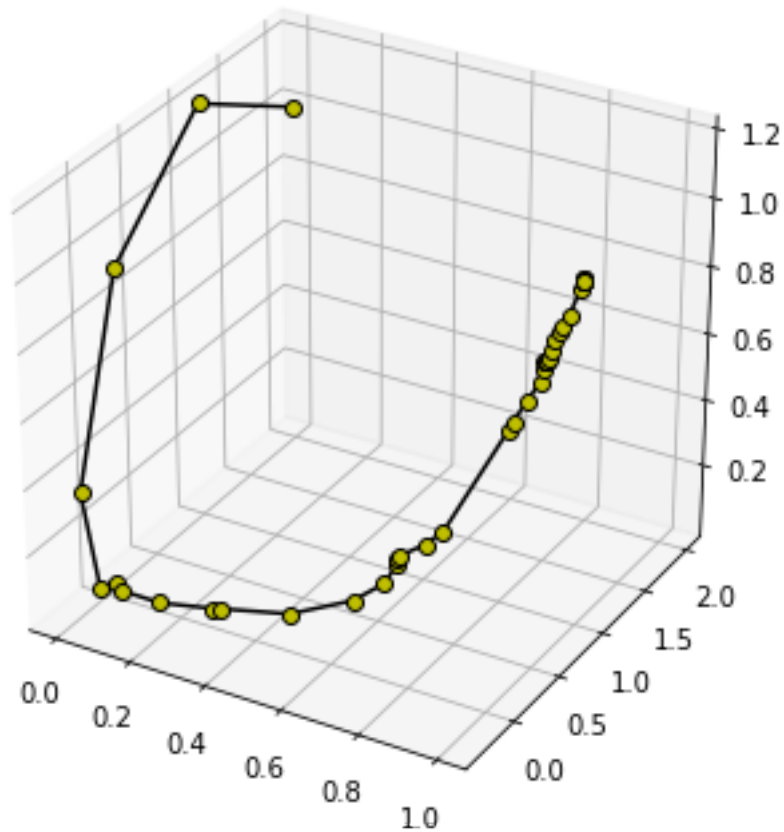
def rosen(x):
    return 100*(x[1]-x[0]**2)**2+(1-x[0])**2+100*(x[2]-x[1]**2)**2+(1-x[2])**2
def callbackF(x):
    global xt,yt,zt
    xt.append(x[0])
    yt.append(x[1])
    zt.append(x[2])

res = minimize(rosen, x0, method='CG',\
               options={'gtol': 1e-5, 'disp': True},callback=callbackF)
print (res.x)
```

Warning: Desired error not necessarily achieved due to precision loss.
Current function value: 0.000000
Iterations: 39
Function evaluations: 707
Gradient evaluations: 139
[0.99999935 0.99999874 0.99999744]

```
[19]: from mpl_toolkits import mplot3d  
  
fig = plt.figure(figsize=(6,6))  
ax = plt.axes(projection='3d')  
ax.plot3D(xt,yt,zt,c='k',marker='o',markerfacecolor='y')
```

```
[19]: [<mpl_toolkits.mplot3d.art3d.Line3D at 0x1c358f98da0>]
```



4.

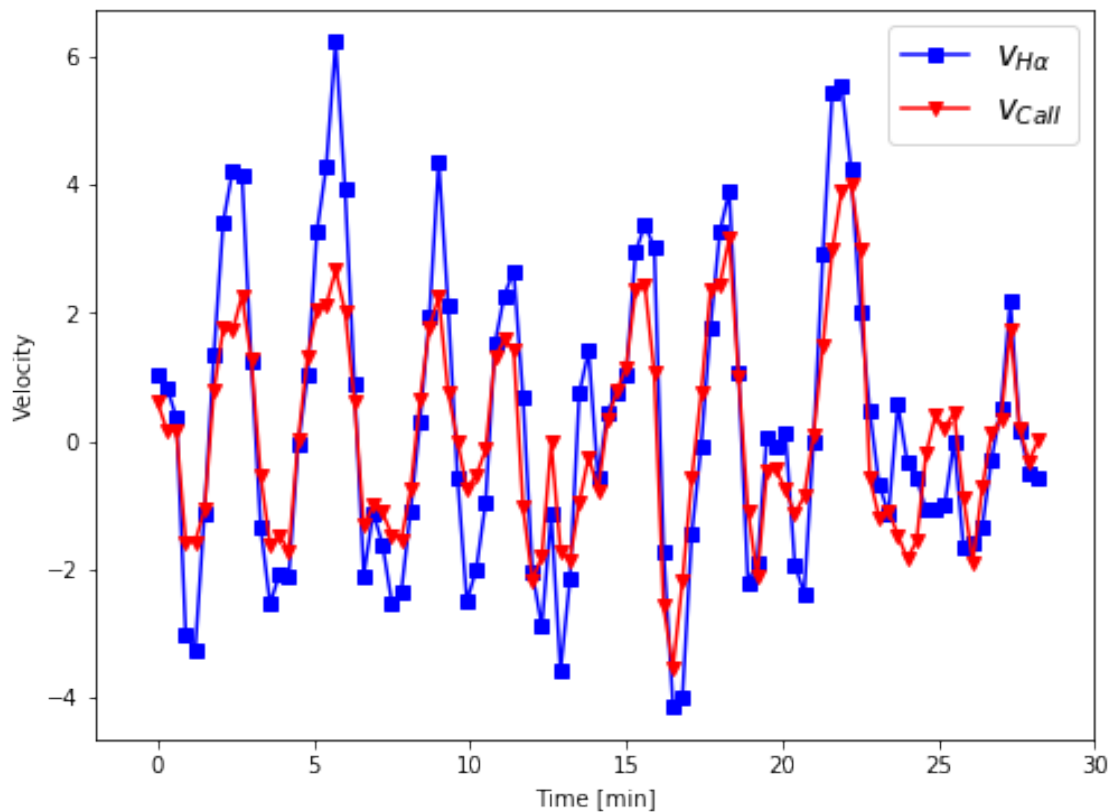
```
[20]: t,v_ha,v_caii=np.loadtxt('sol_vel.dat',unpack=True,usecols=(0,1,2))
```

(a)

```
[21]: plt.figure(figsize=(8,6))

plt.plot(t,v_ha,c='blue',marker='s',label=r'$v_{H\alpha}$')
plt.plot(t,v_caii,c='red',marker='v',label=r'$v_{CaII}$')
plt.xlabel('Time [min]')
plt.ylabel('Velocity')
plt.xlim(-2.,30.)
plt.legend(fontsize=15)
```

[21]: <matplotlib.legend.Legend at 0x1c34beff550>



(b)

```
[22]: # H alpha
ha_fft=np.fft.fft(v_ha)
ha_fft=np.fft.fftshift(ha_fft)

# Ca II
caii_fft=np.fft.fft(v_caii)
```

```

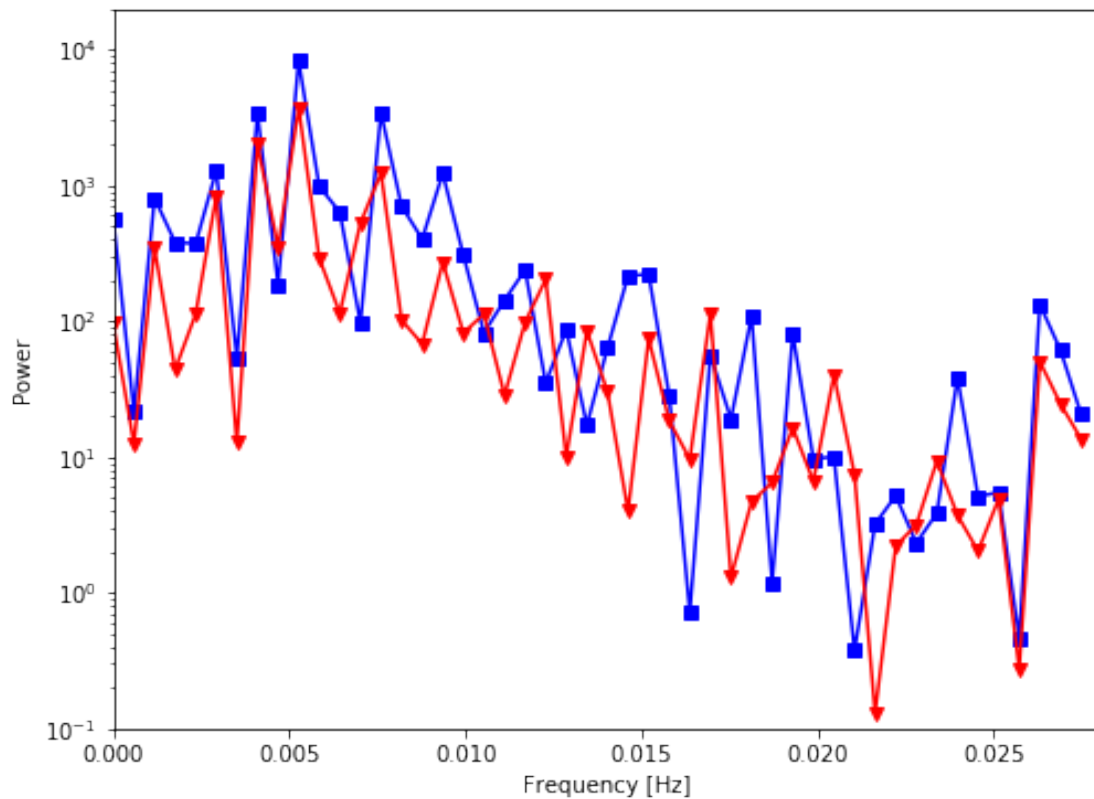
caii_fft=np.fft.fftshift(caii_fft)

# Sample Frequency
t_fft=np.fft.fftfreq(len(t),18) # 0.3분 간격으로 측정된 값!
t_fft=np.fft.fftshift(t_fft)

plt.figure(figsize=(8,6))
plt.semilogy(t_fft,np.absolute(ha_fft)**2,c='blue',marker='s',label=r'H$\alpha$')
plt.semilogy(t_fft,np.absolute(caii_fft)**2,c='red',marker='v',label=r'Ca II')
plt.xlim(0.0,0.028)
plt.ylim(1.0e-1,2.0e4)
plt.xlabel('Frequency [Hz]')
plt.ylabel('Power')

```

[22]: Text(0, 0.5, 'Power')



(c)

```

[23]: t_fft_pos=t_fft[t_fft>0.] # frequency>0만 뽑아냄!

# Power Spectrum에서 최대값을 가지는 Frequency를 주기로 변환

```

```

P_ha    =1./(t_fft_pos[np.argmax(np.absolute(ha_fft[t_fft>0.])**2)])
P_caii  =1./(t_fft_pos[np.argmax(np.absolute(caii_fft[t_fft>0.])**2)])

print ('H alpha : {:.2e} sec, Ca II : {:.2e} sec'.format(P_ha,P_caii))

```

H alpha : 1.90e+02 sec, Ca II : 1.90e+02 sec

(d)

```

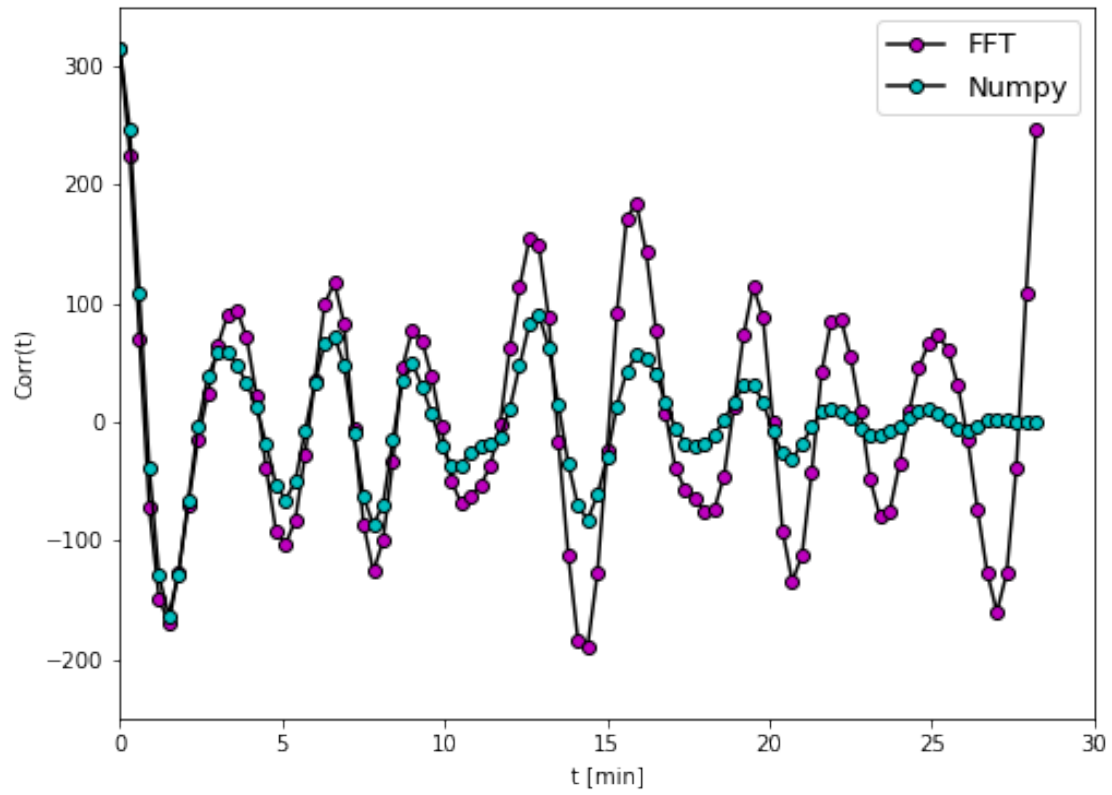
[24]: # 1. FFT를 이용
ha_fft=np.fft.fft(v_ha)
caii_fft=np.fft.fft(v_caii)
Corr=np.fft.ifft(np.conj(ha_fft)*caii_fft) # Correlation Theorem!

# 2. np.correlate 내장함수 사용
Corr2=np.correlate(v_ha,v_caii,'full')
lag=np.arange(-t[-1],t[-1]+0.01,0.3)

plt.figure(figsize=(8,6))
plt.plot(t,Corr,c='k',marker='o',markerfacecolor='m',label='FFT')
plt.plot(lag,Corr2,c='k',marker='o',markerfacecolor='c',label='Numpy')
plt.xlim(0.,30.)
plt.ylim(-250.,350.)
plt.xlabel('t [min]')
plt.ylabel('Corr(t)')
plt.legend(fontsize=13)

```

[24]: <matplotlib.legend.Legend at 0x1c3593ac8d0>



데이터의 범위가 $(-\infty, \infty)$ 가 아니기에 두 결과가 다르게 나타난다. 하지만 그 경향성은 일치하는 것을 볼 수 있다.