Exercise Sheet No. 5

Exercise 1: Spin correlations of the one-dimensional Heisenberg modeL

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Here we focus on the isotropic Heisenberg model in dimension d = 1 with open boundary conditions

$$H = -J \sum_{i=1}^{N-1} \vec{S}_i \vec{S}_{i+1}$$

with J = -1 (the antiferromagnetic case). We are interested in the spin correlations between sites 1 and n,  $\chi_{1n} = \langle \vec{S}_1 \cdot \vec{S}_n \rangle$ , for both zero and finite temperature.

a)Calculate  $\chi_{1n} = \langle \psi_g | \vec{S}_1 \cdot \vec{S}_n | \psi_g \rangle$  for the ground state  $| \psi_g \rangle$  of a Heisenberg chain with N = 10 sites and n = 1, . . . , 10. Note that for an even number of sites N, the ground state of the antiferromagnetic Heisenberg chain is non-degenerate

## In [1]:

```
import numpy as np
import matplotlib.pyplot as plt
from numpy import linalg as LA
import pandas as pd
```

# In [2]:

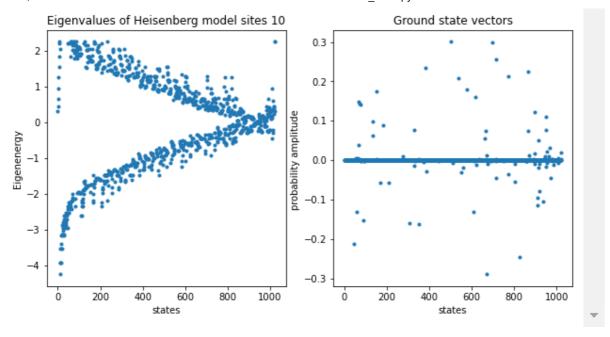
```
def Hmatrix(N,J_x,J_y,J_z):
    n_states = 2**N
    matrix = np.zeros((n_states,n_states))
    for n in range(n_states):
        z = np.zeros(N,dtype = int)
        nz = n
        for z i in range(N):
            if nz//2 >= 0:
                z[z_i] = nz\%2
                nz = nz//2
        for p in range(1,N+1):
            for q in range(p+1,N+1):
                s = (2*z[p-1]-1)*(2*z[q-1]-1)
                1 = n + (1-2*z[p-1])*2**(p-1) + (1-2*z[q-1])*2**(q-1)
                x_{link} = 0.25* J_x[p-1,q-1]
                y_{link} = -0.25*s*J_y[p-1,q-1]
                z link = 0.25*s*J z[p-1,q-1]
                matrix[1,n]-= x_link
                matrix[1,n]-= y_link
                matrix[n,n]-= z_link
    return(matrix)
```

To calculate the ground state we use the code from last assignment and pay attention that the H only consider the neighboring sites

$$H = -J \sum_{i=1}^{N-1} \vec{S}_i \vec{S}_{i+1}$$

## In [3]:

```
# Isotropic Heisenberg model with neighboing sites, set N = 10, J=-1
N = 10
J \times model = np.zeros((2**N,2**N))
J_y_model = np.zeros((2**N,2**N))
J_z_{model} = np.zeros((2**N,2**N))
for i in range(N-1):
    J_x_{\text{model}[i,i+1]} = -1
    J y model[i,i+1] = -1
    J_z_{model[i,i+1]} = -1
H_model = Hmatrix(N,J_x_model,J_y_model,J_z_model)
w_model, v_model = LA.eig(H_model)
print(H_model)
ground_indice = np.argmin(w_model)
fig,ax = plt.subplots(1,2,figsize=(10,5))
ax[0].plot(w_model,'.')
ax[0].set_title('Eigenvalues of Heisenberg model sites '+str(N))
ax[0].set_xlabel('states')
ax[0].set ylabel('Eigenenergy')
ground_state = v_model[ground_indice]
print('The ground state is ',ground_state)
ax[1].plot(ground_state,'.')
ax[1].set_title('Ground state vectors')
ax[1].set_xlabel('states')
ax[1].set ylabel('probability amplitude ')
[[2.25 0.
            0.
                 ... 0.
                           0.
                                0.
                                    ]
       1.75 0.5 ... 0.
 [0.
                           0.
                                0.
                                    ]
 [0.
       0.5 1.25 ... 0.
                                0.
                                    ]
 . . .
 [0.
       0.
            0.
                 ... 1.25 0.5 0.
 [0.
            0.
                 ... 0.5
                          1.75 0.
                                    1
                  ... 0.
 [0.
       0.
            0.
                           0.
                                2.25]]
The ground state is [0.
                                 +0.j 0.
                                                +0.j 0.
                                                              +0.j ... 0.018
93372+0.j
           +0.j 0.
0.
                           +0.il
C:\Users\Xiongxiao Wang\anaconda3\lib\site-packages\matplotlib\cbook\ init
_.py:1333: ComplexWarning: Casting complex values to real discards the imagi
nary part
  return np.asarray(x, float)
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_.py:1333: ComplexWarning: Casting complex values to real discards the imagi
nary part
  return np.asarray(x, float)
Out[3]:
Text(0, 0.5, 'probability amplitude ')
```



To calculate  $\chi_{1n} = \langle \psi_g | \vec{S}_1 \cdot \vec{S}_n | \psi_g \rangle$ , firstly we define matrix  $\vec{\chi}$  as  $\vec{S}_1 \cdot \vec{S}_n$ . Using the similar algorithm of calcualting Hmatrix, it's easy to find the matrix of  $\vec{\chi}$  by setting  $J_{1n}$  = -1 and the other J = 0.

But there is one special case we should pay attention to, when the n = 1 which means  $\vec{S}_1 \cdot \vec{S}_n = \vec{S}_1 \cdot \vec{S}_1$ , the algorithm of calculating Hmatrix() doesn't include this case any more because Hmatrix() only concerns about two different spins. So we should modify the algorithm to include this case

## In [4]:

```
def Cmatrix(N,J_x,J_y,J_z):
    n states = 2**N
    matrix = np.zeros((n_states,n_states))
    for n in range(n_states):
        z = np.zeros(N,dtype = int)
        nz = n
        for z_i in range(N):
            if nz//2 >= 0:
                z[z_i] = nz\%2
                nz = nz//2
        for p in range(1,N+1):
            for q in range(p,N+1):
                s = (2*z[p-1]-1)*(2*z[q-1]-1)
                1 = n + (1-2*z[p-1])*2**(p-1) + (1-2*z[q-1])*2**(q-1)
                #print(n,l)
                x_{link} = 0.25* J_x[p-1,q-1]
                y_{link} = -0.25*s*J_y[p-1,q-1]
                z_{link} = 0.25*s*J_z[p-1,q-1]
                matrix[(l+n_states)%n_states,n]-= x_link
                matrix[(l+n_states)%n_states,n]-= y_link
                matrix[n,n]-= z_link
    return(matrix)
```

## In [5]:

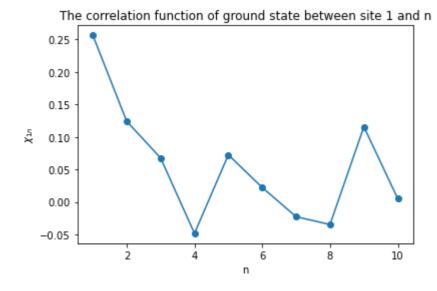
```
# The chi matrix
chi_1n = []
for n in range(N):
    J_x = np.zeros((N,N))
    J_y = np.zeros((N,N))
    J_z = np.zeros((N,N))
    J_x[0,n] = -1
    J_y[0,n] = -1
    J_z[0,n] = -1
    chi_matrix = Cmatrix(N,J_x,J_y,J_z)
    #print(chi_matrix)
    chi_1n.append(np.matmul(ground_state,chi_matrix),ground_state.T))
```

## In [6]:

```
n_list = [i for i in range(1,N+1)]
plt.plot(n_list,chi_1n,'-o')
plt.title(r'The correlation function of ground state between site 1 and n')
plt.xlabel('n')
plt.ylabel(r'$\chi_{1n}$')
```

## Out[6]:

Text(0, 0.5, '\$\\chi\_{1n}\$')



b) Calculate the temperature dependence of the spin correlation

$$\chi_{1n}(T) = \frac{1}{Z} \sum_{l} < l |\vec{S}_1 \cdot \vec{S}_n| l > e^{-\beta E_l}$$

with  $Z = \sum_i e^{-\beta E_l}$  the partition function,  $\beta$  = 1/(kBT) ( $k_B$  can be set to 1), and |I> the eigenstates of H with eigenenergies EI , for temperatures T = 0.5, 2,and 10

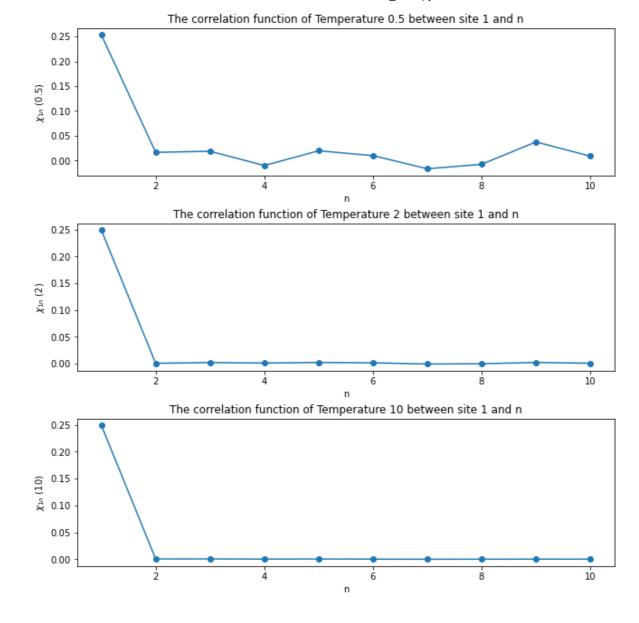
## In [7]:

```
eigenvalue, eigenvector = w_model,v_model
T = [0.5, 2, 10]
index = 0
fig,ax = plt.subplots(len(T),1,figsize = (9,9),constrained_layout=True)
chi_1n_T_dataframe = pd.DataFrame({'site':n_list})
for Ti in T: #loop temperature
   beta = 1/Ti
   Z = np.sum(np.exp(-beta*eigenvalue))
   chi 1n T = []
   for n in range(N): #loop site n from 1 to N
        J_x = np.zeros((N,N))
        J_y = np.zeros((N,N))
        J_z = np.zeros((N,N))
        J_x[0,n] = -1
        J_y[0,n] = -1
        J_z[0,n] = -1
        chi_matrix = Cmatrix(N,J_x,J_y,J_z)
        sum = 0
        for 1 in range(2**N):#loop eigenstates
            sum+=np.matmul(np.matmul(eigenvector[1],chi_matrix),eigenvector[1].T)*np.exp(-b
        chi 1n T.append(sum/Z)
   ax[index].plot(n_list,chi_1n_T,'-o')
   ax[index].set_title(r'The correlation function of Temperature '+str(Ti)+' between site
   ax[index].set_xlabel('n')
   ax[index].set_ylabel(r'$\chi_{1n}$ ('+str(Ti)+')')
   index+=1
    chi_1n_T_dataframe['Temperature'+str(Ti)] = chi_1n_T
print('The spin correlation chi(1n) in three different Temperature')
chi 1n T dataframe.set_index('site')
```

The spin correlation chi(1n) in three different Temperature

#### Out[7]:

	Temperature0.5	Temperature2	Temperature10
site			
1	0.253419-0.000000j	0.249331-0.000000j	0.248840-0.000000j
2	0.016281+0.000050j	0.000542+0.000145j	0.000414+0.000149j
3	0.018577+0.000048j	0.001902+0.000006j	0.000500-0.000020j
4	-0.010254+0.000138j	0.000987+0.000188j	0.000063+0.000181j
5	0.019535-0.000124j	0.002064-0.000165j	0.000388-0.000178j
6	0.009689-0.000129j	0.001406-0.000126j	0.000055-0.000117j
7	-0.016775+0.000087j	-0.000839+0.000141j	-0.000166+0.000122j
8	-0.007781+0.000143j	-0.000265+0.000196j	-0.000006+0.000188j
9	0.037356+0.000127j	0.002105+0.000200j	0.000098+0.000189j
10	0.008641+0.000134j	0.000625+0.000174j	0.000026+0.000183j



c) Show numerically that, in the limit  $T \to 0$ , the finite-temperature spin-correlation  $\chi_{1n}(T)$  of part b) corresponds to the zero-temperature spin-correlation of part a). (2 points)

## In [8]:

```
T0 = 0.01
beta = 1/T0
Z = np.sum(np.exp(-beta*eigenvalue))
chi_1n_T0 = []
for n in range(N): #loop site n from 1 to N
   J_x = np.zeros((N,N))
   J_y = np.zeros((N,N))
   J_z = np.zeros((N,N))
   J_x[0,n] = -1
   J y[0,n] = -1
   J_z[0,n] = -1
   chi_matrix = Cmatrix(N,J_x,J_y,J_z)
   sum = 0
   for 1 in range(2**N):#loop eigenstates
        sum+=np.matmul(np.matmul(eigenvector[1],chi_matrix),eigenvector[1].T)*np.exp(-beta*
   chi_1n_T0.append(sum/Z)
```

# In [11]:

```
print('The spin correlation chi(1n) of T = 0.01 and ground state')
chi_1n_g0_df = pd.DataFrame({'site':n_list,'$\chi_{1n}$ of T = 0.01':chi_1n_T0,'$\chi_{1n}$
chi_1n_g0_df.set_index('site')
```

The spin correlation chi(1n) of T = 0.01 and ground state

#### Out[11]:

## $\chi_{1n}$ of T = 0.01 $\chi_{1n}$ of ground state

#### site

0.256334+0.000000j 0.256334+0.000000j 2 0.124048+0.000028j 0.124048+0.000028j 0.067674+0.000037j 0.067674+0.000037j **4** -0.048182+0.000037j -0.048182+0.000037j 5 0.072508-0.000038j 0.072508-0.000038j 6 0.022278+0.002280j 0.022278+0.002280j -0.022621-0.000049i -0.022621-0.000049j -0.034672+0.000041j -0.034672+0.000041j 0.115693+0.000041j 0.115693+0.000041j 0.005170+0.000041j 0.005170+0.000041j 10

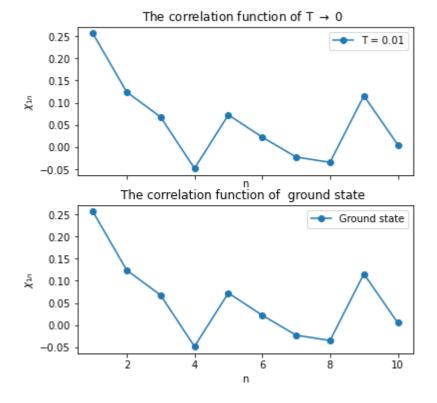
## In [10]:

```
fig,ax1 = plt.subplots(2,figsize = (6,6),sharex = True)
ax1[0].plot(n_list,chi_1n_T0,'-o',label = r'T = 0.01')
ax1[1].plot(n_list,chi_1n,'-o',label = 'Ground state')
ax1[0].set_title(r'The correlation function of T $\rightarrow$ 0 ')
ax1[1].set_title(r'The correlation function of ground state ')
ax1[0].set_xlabel('n')
ax1[0].set_ylabel(r'$\chi_{1n}$')
ax1[1].set_xlabel('n')
ax1[1].set_ylabel(r'$\chi_{1n}$')
ax1[0].legend()
ax1[1].legend()
```

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\_.py:1333: ComplexWarning: Casting complex values to real discards the imagi
nary part
 return np.asarray(x, float)

#### Out[10]:

<matplotlib.legend.Legend at 0x23eaae43550>



These show in the limit  $T\rightarrow 0$ , the finite-temperature spin-correlation  $\chi 1n(T)$  of part b) corresponds to the zero-temperature spin-correlation of part a).