

## Exercise Sheet No. 5

## Exercise 1: Spin correlations of the one-dimensional Heisenberg model

Xiong Xiao Wang, Pulkit Kukreja

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

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

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  $n = 1, \dots, 10$ . 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)
                l = 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[l, n] -= x_link
                matrix[l, 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

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



In [3]:

```
# Isotropic Heisenberg model with neighboring sites, set N = 10, J=-1
```

```
N=10
```

```
J_x_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_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. ]
```

```
[0. 1.75 0.5 ... 0. 0. 0. ]
```

```
[0. 0.5 1.25 ... 0. 0. 0. ]
```

```
...
```

```
[0. 0. 0. ... 1.25 0.5 0. ]
```

```
[0. 0. 0. ... 0.5 1.75 0. ]
```

```
[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. +0.j 0. +0.j]
```

```
C:\Users\Xiong Xiao Wang\anaconda3\lib\site-packages\matplotlib\cbook\__init_
_.py:1333: ComplexWarning: Casting complex values to real discards the imagi
nary part
```

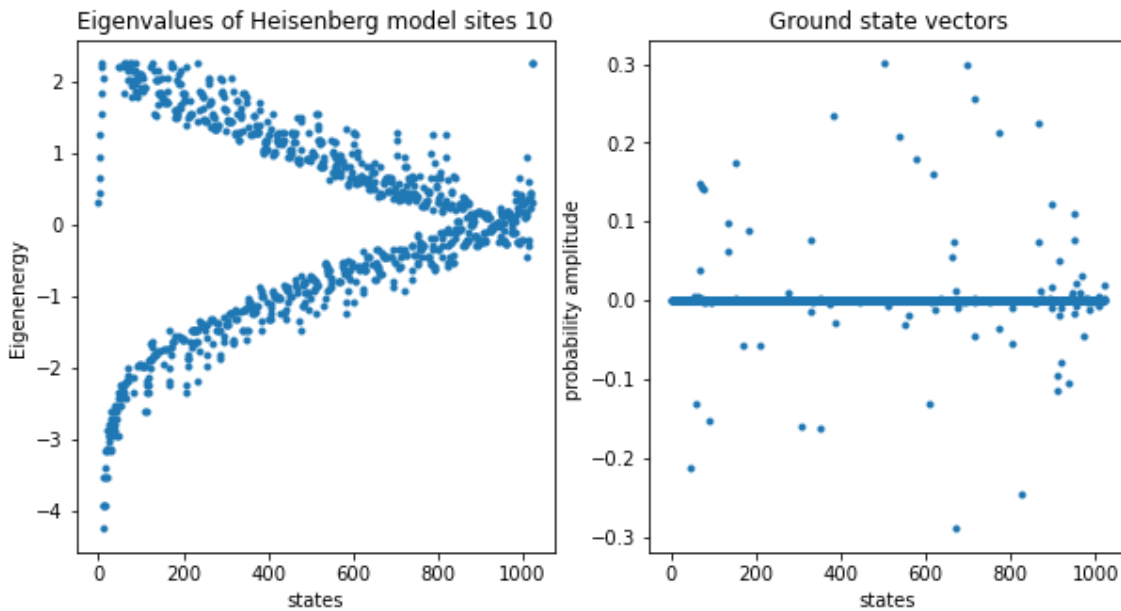
```
return np.asarray(x, float)
```

```
C:\Users\Xiong Xiao 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)
```

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 calculating 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)
                l = 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]:

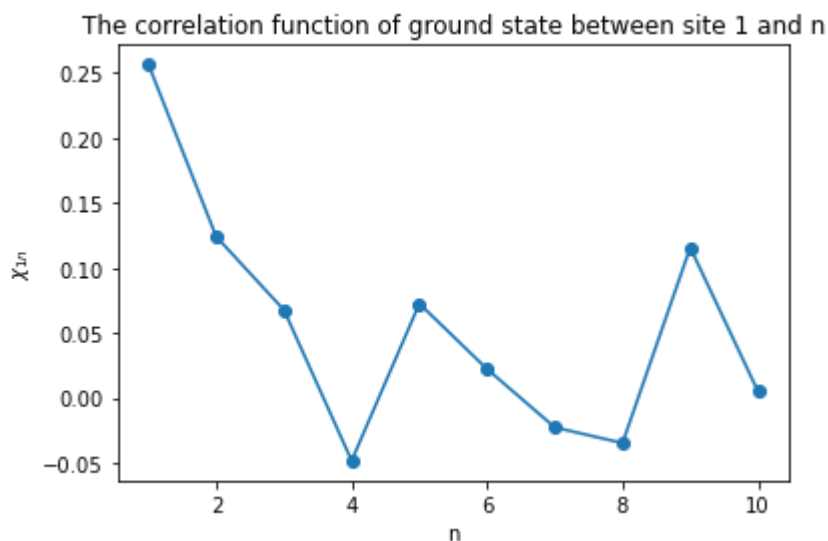
```
# The chi matrix
chi_1n = []
for n in range(N):
    J_x = np.zeros((N,N))
    J_y = np.zeros((N,N))
    J_z = np.zeros((N,N))
    J_x[0,n] = -1
    J_y[0,n] = -1
    J_z[0,n] = -1
    chi_matrix = Cmatrix(N,J_x,J_y,J_z)
    #print(chi_matrix)
    chi_1n.append(np.matmul(np.matmul(ground_state,chi_matrix),ground_state.T))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

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

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

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

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 l in range(2**N):#Loop eigenstates
            sum+=np.matmul(np.matmul(eigenvector[l],chi_matrix),eigenvector[l].T)*np.exp(-b

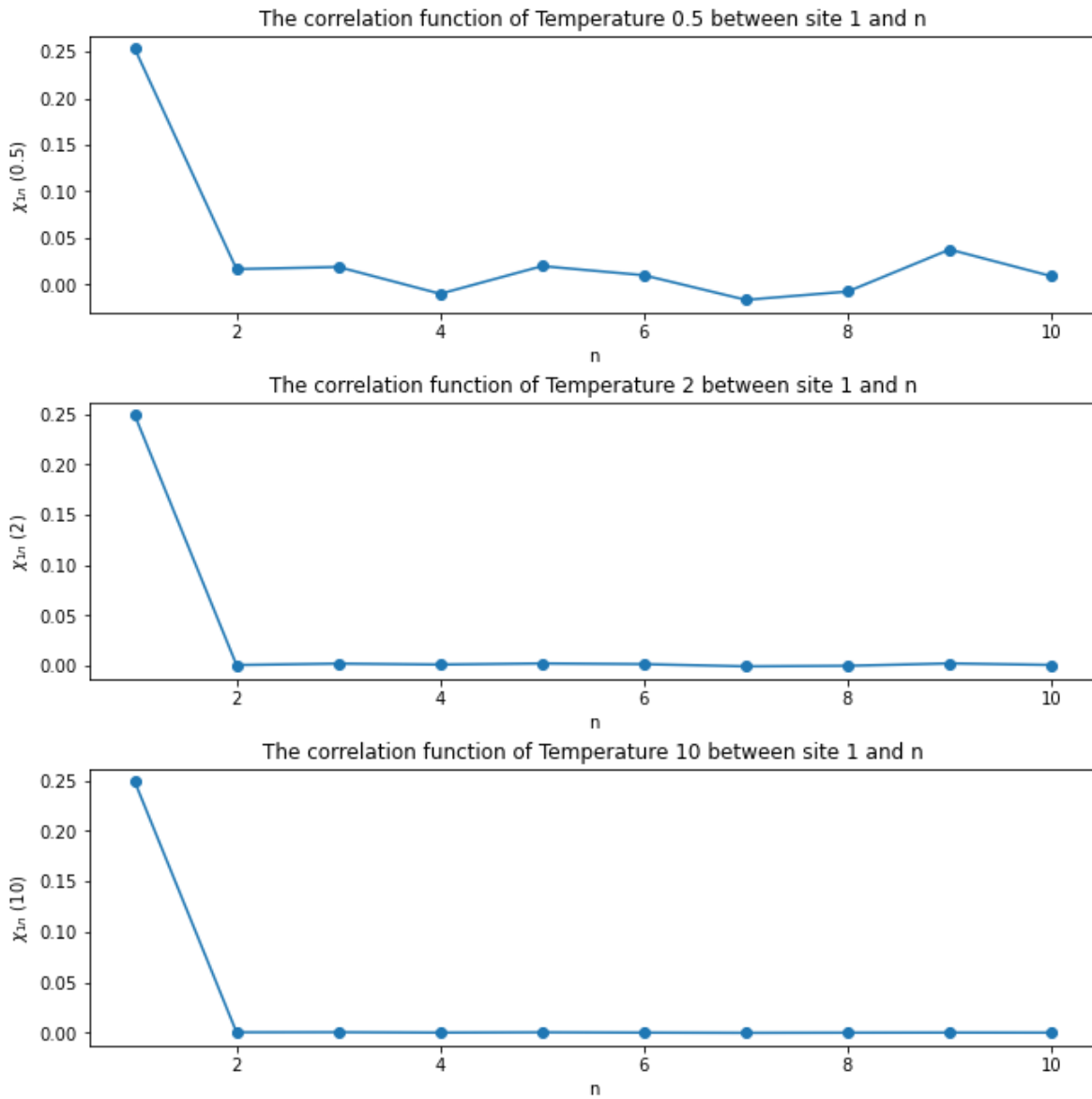
        chi_1n_T.append(sum/Z)
    ax[index].plot(n_list,chi_1n_T,'-o')
    ax[index].set_title(r'The correlation function of Temperature '+str(Ti)+' between site
    ax[index].set_xlabel('n')
    ax[index].set_ylabel(r'$\chi_{1n}$ ('+str(Ti)+'))
    index+=1
    chi_1n_T_dataframe['Temperature'+str(Ti)] = chi_1n_T
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  $T \rightarrow 0$ , 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 l in range(2**N):#Loop eigenstates
        sum+=np.matmul(np.matmul(eigenvector[l],chi_matrix),eigenvector[l].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		
1	0.256334+0.000000j	0.256334+0.000000j
2	0.124048+0.000028j	0.124048+0.000028j
3	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
7	-0.022621-0.000049j	-0.022621-0.000049j
8	-0.034672+0.000041j	-0.034672+0.000041j
9	0.115693+0.000041j	0.115693+0.000041j
10	0.005170+0.000041j	0.005170+0.000041j



In [10]:

```

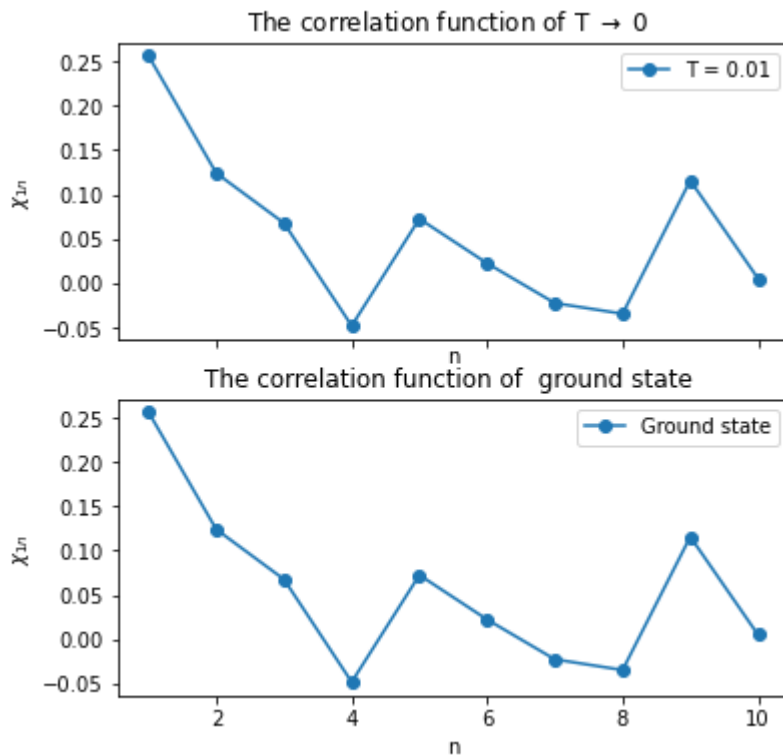
fig,ax1 = plt.subplots(2,figsize = (6,6),sharex = True)
ax1[0].plot(n_list,chi_1n_T0,'-o',label = r'T = 0.01')
ax1[1].plot(n_list,chi_1n,'-o',label = 'Ground state')
ax1[0].set_title(r'The correlation function of T  $\rightarrow$  0 ')
ax1[1].set_title(r'The correlation function of ground state ')
ax1[0].set_xlabel('n')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

```

C:\Users\Xiong Xiao Wang\anaconda3\lib\site-packages\matplotlib\cbook\\_\_init\_\_.py:1333: ComplexWarning: Casting complex values to real discards the imaginary 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).



日期:

## Exercise 2: Reduced density matrix and entanglement entropy

$$a) |\psi\rangle = \sum_{i=1}^2 \sum_{j=1}^2 \psi_{ij} |i\rangle |j\rangle$$

$$|\psi_1\rangle = |\uparrow_A\rangle |\downarrow_B\rangle$$

$$|\psi_2\rangle = \frac{1}{\sqrt{2}} (|\uparrow_A\rangle |\downarrow_B\rangle - |\downarrow_A\rangle |\uparrow_B\rangle)$$

$$|\psi_3\rangle = \frac{1}{2} (|\uparrow_A\rangle + |\downarrow_A\rangle) (|\uparrow_B\rangle + |\downarrow_B\rangle)$$

$$|\psi\rangle = \sum_{\sigma_1=\uparrow,\downarrow} \sum_{\sigma_2=\uparrow,\downarrow} \psi_{\sigma_1,\sigma_2} |\sigma_1\rangle |\sigma_2\rangle$$

$$= (|\uparrow_A\rangle |\downarrow_A\rangle) \begin{pmatrix} \psi_{\uparrow\uparrow} & \psi_{\uparrow\downarrow} \\ \psi_{\downarrow\uparrow} & \psi_{\downarrow\downarrow} \end{pmatrix} \begin{pmatrix} |\uparrow_B\rangle \\ |\downarrow_B\rangle \end{pmatrix}$$

$$|\psi_1\rangle = (|\uparrow_A\rangle |\downarrow_A\rangle) \cdot \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} |\uparrow_B\rangle \\ |\downarrow_B\rangle \end{pmatrix}$$

$$|\psi_2\rangle = (|\uparrow_A\rangle |\downarrow_A\rangle) \cdot \begin{pmatrix} 0 & \frac{\sqrt{2}}{2} \\ -\frac{\sqrt{2}}{2} & 0 \end{pmatrix} \begin{pmatrix} |\uparrow_B\rangle \\ |\downarrow_B\rangle \end{pmatrix}$$

$$|\psi_3\rangle = (|\uparrow_A\rangle |\downarrow_A\rangle) \cdot \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} |\uparrow_B\rangle \\ |\downarrow_B\rangle \end{pmatrix}$$

日期:  $\hat{P} \equiv \hat{T}_{\text{B}}(|\psi\rangle\langle\psi|) = \sum_{j=1}^2 |\psi_j\rangle\langle\psi_j|$

with the matrix elements  $P_{ii'} = \langle i | \hat{P} | i' \rangle = \sum_j \psi_{ij} \psi_{ij'}$

$$\hat{P}_A = C \begin{pmatrix} |\uparrow_A\rangle & |\downarrow_A\rangle \end{pmatrix} \underbrace{\begin{pmatrix} P_{\uparrow\uparrow} & P_{\uparrow\downarrow} \\ P_{\downarrow\uparrow} & P_{\downarrow\downarrow} \end{pmatrix}}_{P_{\sigma_1\sigma_2}} \begin{pmatrix} \langle\uparrow_A| \\ \langle\downarrow_A| \end{pmatrix}$$

$$\begin{aligned} \hat{P}_{A1} &= \hat{T}_B(|\psi_1\rangle\langle\psi_1|) = \underbrace{\langle\uparrow_B|\psi_1\rangle\langle\psi_1|\uparrow_B\rangle}_{=\langle\uparrow_B|\downarrow_B\rangle\langle\uparrow_A|} + \underbrace{\langle\downarrow_B|\psi_1\rangle\langle\psi_1|\downarrow_B\rangle}_{=\langle\downarrow_B|\downarrow_B\rangle\langle\uparrow_A|} \\ &= \underbrace{\langle\uparrow_B|\downarrow_B\rangle}_{=0} \langle\uparrow_A| = \underbrace{\langle\downarrow_B|\downarrow_B\rangle}_1 \langle\uparrow_A| = \underbrace{\langle\uparrow_A|\langle\downarrow_B|\downarrow_B\rangle}_1 \end{aligned}$$

$$\hat{P}_{A1} = |\uparrow_A\rangle\langle\uparrow_A| \quad \overline{P_{\sigma_1\sigma_2}} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\hat{P}_{A2} = \hat{T}_B(|\psi_2\rangle\langle\psi_2|) = \langle\uparrow_B|\psi_2\rangle\langle\psi_2|\uparrow_B\rangle + \langle\downarrow_B|\psi_2\rangle\langle\psi_2|\downarrow_B\rangle$$

$$\langle\uparrow_B|\psi_2\rangle = -\frac{1}{\sqrt{2}} \langle\uparrow_B|\uparrow_B\rangle |\downarrow_A\rangle = -\frac{1}{\sqrt{2}} |\downarrow_A\rangle \quad \langle\psi_2|\uparrow_B\rangle = -\frac{1}{\sqrt{2}} \cdot \langle\downarrow_A|$$

$$\langle\downarrow_B|\psi_2\rangle = \frac{1}{\sqrt{2}} \langle\uparrow_A| \quad \langle\psi_2|\downarrow_B\rangle = \frac{1}{\sqrt{2}} \cdot \langle\uparrow_A|$$

$$\hat{P}_{A2} = \frac{1}{2} |\downarrow_A\rangle\langle\downarrow_A| + \frac{1}{2} |\uparrow_A\rangle\langle\uparrow_A| \quad \overline{P_{\sigma_1\sigma_2}} = \begin{pmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{pmatrix}$$

$$\hat{P}_{A3} = \hat{T}_B(|\psi_3\rangle\langle\psi_3|) = \langle\uparrow_B|\psi_3\rangle\langle\psi_3|\uparrow_B\rangle + \langle\downarrow_B|\psi_3\rangle\langle\psi_3|\downarrow_B\rangle$$

$$\langle\uparrow_B|\psi_3\rangle = \frac{1}{2} (|\uparrow_A\rangle + |\downarrow_A\rangle) \quad \langle\psi_3|\uparrow_B\rangle = \frac{1}{2} (\langle\uparrow_A| + \langle\downarrow_A|)$$

$$\langle\downarrow_B|\psi_3\rangle = \frac{1}{2} (|\uparrow_A\rangle + |\downarrow_A\rangle) \quad \langle\psi_3|\downarrow_B\rangle = \frac{1}{2} (\langle\uparrow_A| + \langle\downarrow_A|)$$

$$\hat{P}_{A3} = \frac{1}{2} (|\uparrow_A\rangle + |\downarrow_A\rangle) (\langle\uparrow_A| + \langle\downarrow_A|) \quad \overline{P_{\sigma_1\sigma_2}} = \begin{pmatrix} \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} \end{pmatrix}$$

日期:

Entanglement entropy  $S_e$ :

$$S_e = -\text{Tr}_A [\hat{\rho}_A \ln \hat{\rho}_A] = -\sum_i w_i \ln w_i$$

For state  $|\psi_1\rangle$   $\overline{\Psi}_{0102} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$  eigenvalue:  $w_1 = 1$   
 $w_2 = 0$

$$S_e = -w_1 \ln w_1 - w_2 \ln w_2 = \underbrace{-0 \ln 0}_{\lim_{x \rightarrow 0} x \cdot \ln x = 0} - 1 \cdot \ln 1 = 0$$

$$\lim_{x \rightarrow 0} x \cdot \ln x = 0$$

For state  $|\psi_2\rangle$   $\overline{\Psi}_{0102} = \begin{pmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{pmatrix}$   $w_1 = \frac{1}{2}$   $w_2 = \frac{1}{2}$

$$S_e = -\frac{1}{2} \ln\left(\frac{1}{2}\right) - \frac{1}{2} \ln\left(\frac{1}{2}\right) = \ln 2$$

For state  $|\psi_3\rangle$   $\overline{\Psi}_{0102} = \begin{pmatrix} \frac{1}{4} & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} \end{pmatrix}$

$$\det(\overline{\Psi}_{0102} - \lambda E) = \begin{vmatrix} \frac{1}{4} - \lambda & \frac{1}{4} \\ \frac{1}{4} & \frac{1}{4} - \lambda \end{vmatrix} = \left(\frac{1}{4} - \lambda\right)^2 - \frac{1}{16} = 0 \Rightarrow \lambda_1 = \frac{1}{2} \quad \lambda_2 = 0$$
$$\Rightarrow w_1 = \frac{1}{2} \quad w_2 = 0$$

$$S_e = -\frac{1}{2} \ln\left(\frac{1}{2}\right) - 0 \ln 0 = \frac{1}{2} \ln 2$$

日期: /