

Exercise 1:

$$x_0, x_1, x_2 \in \{0, 1\}$$

$$E(x_0, x_1, x_2) = \phi_0(x_0) + \phi_1(x_1) + \phi_2(x_2) + \phi_p(x_0, x_1) + \phi_p(x_1, x_2) + \phi_p(x_0, x_2)$$

a) Evaluate $E(x_0, x_1, x_2)$ by hand for all possible configurations of x_0, x_1, x_2

i) all down-state $(0, 0, 0)$

$$E(0, 0, 0) = 0,1 + 0,8 + 0,9 + 0 = 1,8$$

ii) all up-state $(1, 1, 1)$

$$E(1, 1, 1) = 0,9 + 0,1 + 0,1 + 0 = 1,1$$

iii) x_1 up-state, x_0, x_2 different arbitrary

$$E(1, 0, 1) = 0,9 + 0,8 + 0,1 + 2 = 3,8$$

$$E(1, 1, 0) = 0,9 + 0,1 + 0,9 + 2 = 3,9$$

$$E(1, 0, 0) = 0,9 + 0,8 + 0,9 + 2 = 4,6$$

iv) x_1 down state

$$E(0, 0, 1) = 0,1 + 0,8 + 0,9 + 2 = 3,8$$

$$E(0, 1, 0) = 0,1 + 0,1 + 0,9 + 2 = 3,1$$

$$E(0, 1, 1) = 0,1 + 0,1 + 0,1 + 2 = 2,3$$

b) Which configuration of x_0, x_1, x_2 minimizes $E(x_0, x_1, x_2)$?

$E(0, 1, 1)$ and $E(0, 0, 0)$ are the 2 lowest energies.

They minimize $E_{\text{pot}} = \sum \phi_p$. Because the upper state is more likely on 2 particles $E(0, 1, 1)$ minimizes E .

Exercise 2:

$$\psi_{ij}(z_i, z_j) = \begin{cases} \alpha & \text{if } z_i = z_j \\ \beta & \text{if } z_i \neq z_j \end{cases}$$

$$E(z) = \sum_{i,j} \psi_{ij}(z_i, z_j)$$

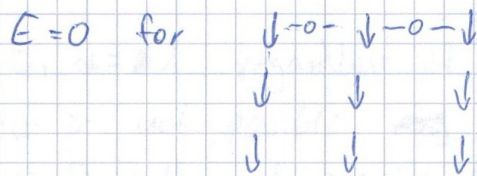
Find the configuration which minimizes E for:

a) $\alpha = 0, \beta = 1$:

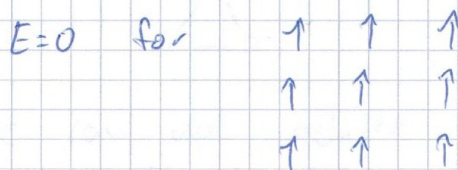
for minimal E , ψ needs to be 0.

ψ is 0 for $z_i = z_j$

↳ neighbors need to be in the same state



all particles in down state (0)



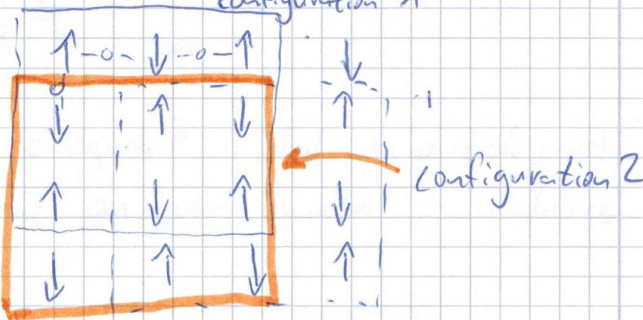
all particles in up state (1)

b) $\alpha = 1, \beta = 0$:

$\psi = 0$ for $z_i \neq z_j$

↳ neighbors need to be in different states

$E=0$ for



Exercise 3: $\psi_{ij}(z_i, z_j) = \begin{cases} \alpha & \text{if } z_i = z_j \\ \beta & \text{if } z_i \neq z_j \end{cases}$

Is it possible to find z such that $E(z) = 0$ for ~~any~~

a) $\alpha = 0, \beta = 1$ regardless of the structure.

It is possible, because if all particles are in the same state ψ for any neighbor is

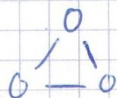
0. z all up-state

or

z all down-state ^{fulfills} ~~minimizes~~ $E(z) = 0$

b) $\beta = 0, \alpha = 1$ regardless of structure.

This is not possible ~~as~~ i.e. for the ψ configuration of exercise 1 ~~as~~



we can not achieve that all neighbors are in different states.

The "chain" which connects particle a to the neighbor b needs to be odd:



Maybe the consequence is that a particle needs to have an even number of neighbors

Untitled

November 8, 2017

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In [1]: import numpy as np
import matplotlib.pyplot as plt
from math import fabs as abs

def prior_l0(n_states, alpha):
    prior = np.ones((n_states, n_states))*alpha
    np.fill_diagonal(prior,0)
    return prior

def prior_l1(n_states, alpha):
    prior = np.zeros((n_states, n_states))
    for n in range(n_states):
        for m in range(n_states):
            prior[n,m] = alpha*abs(n-m)
    return prior

def calc_proba(lattice, x, y, prior):
    (xMax, yMax) = lattice.shape
    (horiz_neighbors, vertic_neighbors) = np.array([x-1,x+1]),np.array([y-1,y+1])
    states = []
    neighbors = []
    for xN in horiz_neighbors[np.in1d(horiz_neighbors,np.arange(xMax))]:
        neighbors.append(lattice[xN,y])
    for yN in vertic_neighbors[np.in1d(vertic_neighbors,np.arange(yMax))]:
        neighbors.append(lattice[x,yN])
    energies = np.exp(-np.sum(prior[:,neighbors], axis= 1))
    for i in range(len(prior)):
        states.append( energies[i]/np.sum(energies))
    return states

def gibbs_update(lattice, x, y, prior):
    probabilities = calc_proba(lattice, x, y, prior)
    new_state = np.random.choice(len(prior), 1, p=probabilities)
    lattice[x,y] = new_state

def sweep_scanlines(lattice, prior):
    (x_len,y_len) = lattice.shape
    for x in range(x_len):
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        for y in range(y_len):
            gibbs_update(lattice, x, y, prior)
    return 0

```

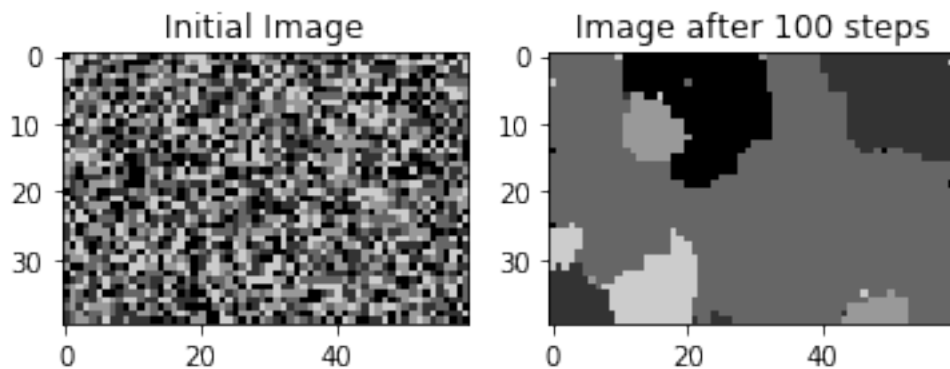
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In [2]: print('Prior 1, 100 steps, alpha = 2')
        n_iter = 100
        n_states = 5
        n_x = 40
        n_y = 60
        prior= prior_l0(n_states,2)
        lattice = np.random.randint(n_states, size = (n_x, n_y))
        plt.figure()
        plt.subplot(121)
        plt.imshow(lattice, cmap='gray', vmax=n_states)
        plt.title('Initial Image')
        for i in range(n_iter):
            sweep_scanlines(lattice, prior)

        plt.subplot(122)
        plt.imshow(lattice, cmap='gray', vmax=n_states)
        plt.title('Image after %d steps'%n_iter)
        plt.show()

```

Prior 1, 100 steps, alpha = 2



```

In [3]: print('Prior 1, 10 steps, alpha = 2')
        n_iter = 10
        n_states = 5
        n_x = 40
        n_y = 60
        prior= prior_l1(n_states,2)
        lattice = np.random.randint(n_states, size = (n_x, n_y))
        plt.figure()

```

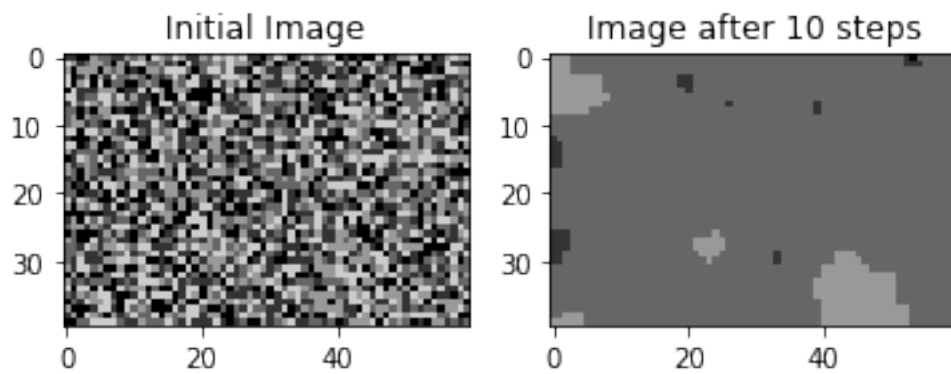
```

plt.subplot(121)
plt.imshow(lattice, cmap='gray', vmax=n_states)
plt.title('Initial Image')
for i in range(n_iter):
    sweep_scanlines(lattice, prior)

plt.subplot(122)
plt.imshow(lattice, cmap='gray', vmax=n_states)
plt.title('Image after %d steps'%n_iter)
plt.show()

```

Prior 1, 10 steps, $\alpha = 2$



Prior 2 converges much faster