sheet08

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1 Machine Intelligence II - Team MensaNord

1.1 Sheet 08

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```
In [1]: from __future__ import division, print_function
    import seaborn as sns
    import matplotlib.pyplot as plt
    %matplotlib inline
    import scipy.stats
    import numpy as np
```

1.2 Exercise 1

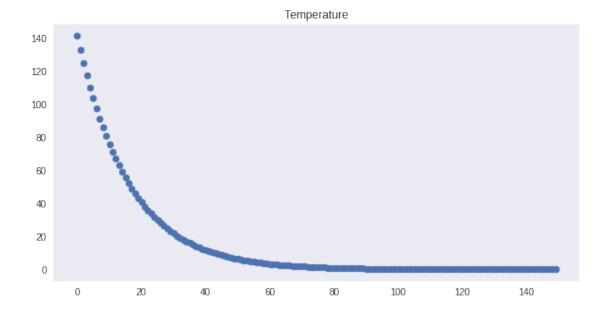
1.2.1 plotting (a) with M = 1

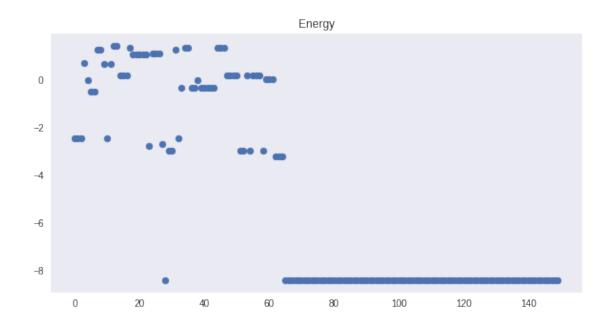
```
In [3]: # loop borders
    t_max = 150

# Temperature ( beta == noise ) decreasing with factor tau
    beta_0 = np.random.random() / 50
    beta_0 = 0.007080269862410016
    print("beta: ", beta_0)
    print("Temp: ", 1 / beta_0) # == Temperature
```

```
tau = np.random.random() / 10 + 1
        tau = 1.064168355840528
        print("tau: ", tau)
beta: 0.007080269862410016
Temp: 141.2375544199406
tau: 1.064168355840528
In [7]: # INITIALIZATION
        M = 1
        beta = beta_0
        # configuration s
        s = np.random.choice([-1, 1], 6)
        print("\ns: ", s)
        # weight matrix, symmetrical, diagonal 0
        W = np.random.random(size=(6, 6))
        W = (W + W.T)/2
        for i in range(W.shape[0]):
            W[i, i] = 0
        print("W: ", W)
        \# minimum for validation ( E >= validation\_min @ any point)
        validation_min = E_s(W, s)
        print("\nvalidation min", validation_min)
        # variables for plotting
        T_plot = [0.0 \text{ for } i \text{ in } range(t_max * M)]
        E_plot = [0.0 \text{ for } i \text{ in } range(t_max * M)]
        # ALGORITHM
        for t in range(t_max):
             for m in range(M):
                 i = np.random.randint(0, 6)
                 s_{local} = np.copy(s)
                 s_local[i] *= -1
                 E_1 = E_s(W, s)
                 E_2 = E_s(W, s_{local})
                 E d = E 2 - E 1
                 P = 1 / (1 + np.exp(beta*E_d))
                 # print("\nt:", t, " i:", i, "\n s1:", s, "\tE1:", E_1, "\n s2:", s
                 if np.random.random() < P:</pre>
```

```
s = np.copy(s_local)
                   # print("new s")
               if E_s(W, s) < validation_min:</pre>
                   validation_min = E_s(W, s)
                   print("new min")
               T_plot[M*t+m] = 1 / beta
               E \text{ plot}[M*t+m] = E \text{ s}(W, \text{ s})
           beta *= tau
       # RESULTS
       print("validation min", validation_min)
       print("\nresult: \ts:", s, "\tE:", E_s(W, s))
       # print("E_plot: ", E_plot)
       # print("T_plot: ", T_plot)
       # print(W)
       plt.figure(figsize=(10, 5))
       plt.scatter(range(t_max * M), T_plot)
       plt.title('Temperature')
       plt.grid()
       plt.show()
       plt.figure(figsize=(10, 5))
       plt.scatter(range(t_max * M), E_plot)
       plt.title('Energy')
       plt.grid()
       plt.show()
s: [-1 -1 -1 1 1 -1]
                 0.56017005 0.44456597 0.43450962 0.6157965 0.684050561
W: [[ O.
[ 0.56017005 0.
                         0.34383935 0.51386398 0.4499107 0.74056969]
 [ 0.44456597  0.34383935  0.
                                     0.89637617 0.71340219 0.462468 1
 [ 0.43450962  0.51386398  0.89637617  0.
                                                0.60596283 0.53142878]
 0.420338271
 [ 0.68405056  0.74056969  0.462468  0.53142878  0.42033827  0.
                                                                     11
validation min 0.733999749529
new min
new min
new min
validation min -8.41725264754
result:
         s: [-1 -1 -1 -1 -1]
                                           E: -8.41725264754
```

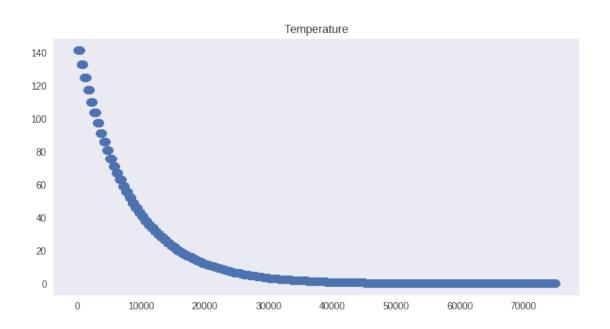


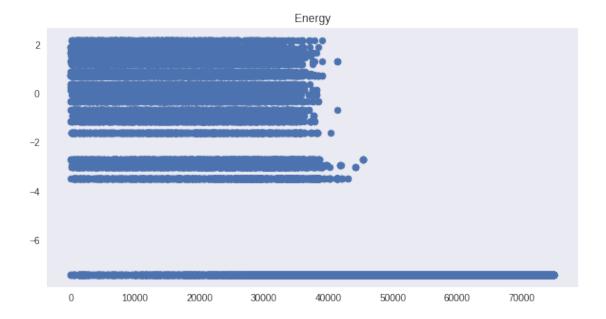


1.2.2 plotting (a) with M = 500

```
print("\ns: ", s)
# weight matrix, symmetrical, diagonal 0
W = np.random.random(size=(6, 6))
W = (W + W.T)/2
for i in range(W.shape[0]):
    W[i, i] = 0
print("W: ", W)
# minimum for validation ( E >= validation_min @ any point)
validation_min = E_s(W, s)
print("\nvalidation min", validation_min)
# variables for plotting
T_plot = [0.0 \text{ for } i \text{ in } range(t_max * M)]
E_plot = [0.0 \text{ for } i \text{ in } range(t_max * M)]
# ALGORITHM
for t in range(t max):
    for m in range(M):
        i = np.random.randint(0, 6)
        s_{local} = np.copy(s)
        s_local[i] \star = -1
        E_1 = E_s(W, s)
        E_2 = E_s(W, s_{local})
        E_d = E_2 - E_1
        P = 1 / (1 + np.exp(beta*E_d))
         # print("\nt:", t, " i:", i, "\n s1:", s, "\tE1:", E_1, "\n s2:", s
        if np.random.random() < P:</pre>
             s = np.copy(s_local)
             # print("new s")
        if E_s(W, s) < validation_min:</pre>
             validation_min = E_s(W, s)
             print("new min")
        T_plot[M*t+m] = 1 / beta
        E_plot[M*t+m] = E_s(W, s)
    beta *= tau
# RESULTS
print("validation min", validation_min)
print("\nresult: \ts:", s, "\tE:", E_s(W, s))
# print("E_plot: ", E_plot)
# print("T_plot: ", T_plot)
# print(W)
plt.figure(figsize=(10, 5))
```

```
plt.scatter(range(t_max * M), T_plot)
        plt.title('Temperature')
        plt.grid()
        plt.show()
        plt.figure(figsize=(10, 5))
        plt.scatter(range(t_max * M), E_plot)
        plt.title('Energy')
        plt.grid()
        plt.show()
s: [-1 -1 -1 -1 -1]
                   0.24163601 0.33421698 0.47590737 0.53840565 0.37582615]
W: [[ 0.
                                                    0.4797754
                           0.59755121 0.2469511
 [ 0.24163601 0.
                                                                0.64198545]
 [ 0.33421698  0.59755121  0.
                                       0.26031671 0.41562438
                                                                0.64093898]
                           0.26031671
                                                                0.70671285]
 [ 0.47590737  0.2469511
                                       0.
                                                    0.67985082
 [ 0.53840565  0.4797754
                           0.41562438 0.67985082
                                                                0.78613867]
                                                    0.
 [0.37582615 \quad 0.64198545 \quad 0.64093898 \quad 0.70671285 \quad 0.78613867 \quad 0.
                                                                          11
validation min -7.42183772977
validation min -7.42183772977
result:
                s: [1 1 1 1 1 1] E: -7.42183772977
```

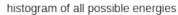


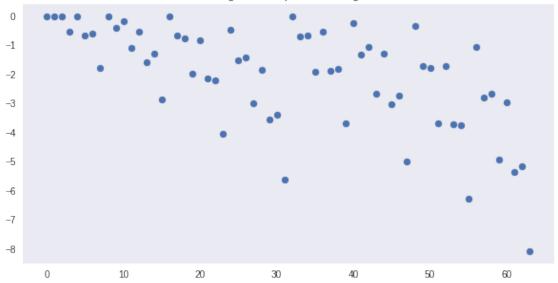


1.2.3 plotting (b)

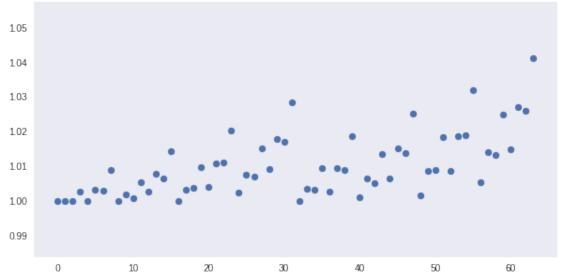
```
In [6]: # generate all posible states & energies
        all_states = [[0, 0, 0, 0, 0] for i in range (2**6)
        all_energies = [0.0 \text{ for } i \text{ in } range(2**6)]
        for si in range (2**6):
            all\_states[si] = [int(x) for x in list('{0:06b}'.format(si))]
            all_energies[si] = E_s(W, all_states[si])
        plt.figure(figsize=(10, 5))
        plt.scatter(range(2**6), all_energies)
        plt.title('histogram of all possible energies')
        plt.grid()
        plt.show()
        probab_beta = [0.005, 1, 3]
        for beta in probab_beta:
            Z = 0
            for en in all_energies:
                 Z += np.exp(-beta * en)
            all_probabilities = [0.0 \text{ for } i \text{ in } range(2**6)]
            for si in range (2**6):
                 all_probabilities[si] = np.exp(-beta * all_energies[si])
            plt.figure(figsize=(10, 5))
```

```
plt.scatter(range(2**6), all_probabilities)
plt.title('histogram of all possible probabilities for beta {}'.formate
plt.grid()
plt.show()
```

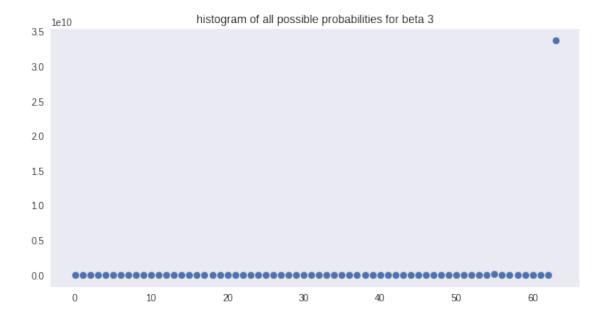




histogram of all possible probabilities for beta 0.005







In []: