# 5 Model Multiplies Result by using Small Scale Dataset

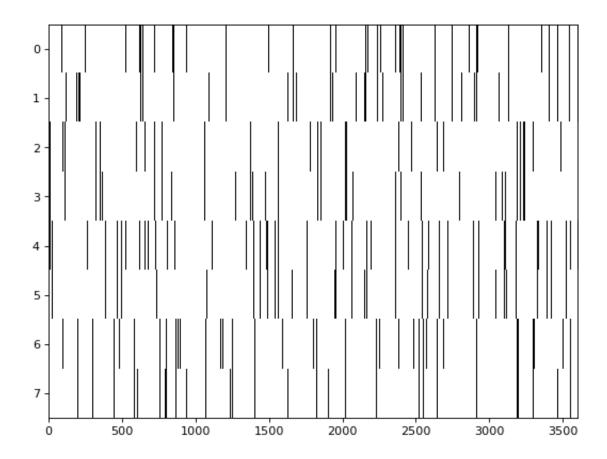
#### August 3, 2023

```
[1]: import warnings
        context
     with warnings.catch_warnings():
         warnings.filterwarnings("ignore")
         from coniii import *
     import numpy as np
     from tqdm.auto import tqdm
     from scipy.spatial.distance import cdist
     from scipy.stats import multivariate_normal
     import matplotlib.pyplot as plt
     import random
     import Jangenerate_assembly #In the same dir
     import Jangenerate_SpikeCount #In the same dir
     from scipy.stats import poisson
     import itertools
     import time
     import math
     warnings.filterwarnings("ignore")
     # Define Parameters
     T = 3600 \# time of simul"ation
     dT = 0.5 \# time step
     params_assembly_num =4 # number of assemblies
     params_point_into_neuron_distance = 0.5
     # Length of an active event as a number of timesteps
     eventDur = np.random.randint(1, 10)
     # Probability with which a unit is particularly active in a single timestep
     eventProb = np.random.uniform(0.01, 0.05)
     # Firing rate multiplier at active events
     eventMult = np.random.uniform(6, 10) # random number between 1 and 5
     showPlot = True
     def binaryOutput(original_list):
```

```
# Create a new list to hold the tuples
   tuples_list = []
    # Generate all possible combinations of two elements for each sublist
   for sublist in original_list:
        combinations = itertools.combinations(sublist, 2)
        # Convert the combinations into tuples and add them to the list
       tuples_list.extend(tuple(sorted(combination)) for combination in_
 ⇔combinations)
    \# Remove duplicates by converting the list to a set then back to a list
   unique_tuples = list(set(tuples_list))
   return unique_tuples
N = 8
params_assembly_density = 2 # size of neurons in each assembly
assemblies list = []
spikeCount_list = []
binary list = []
fire_rate_background = np.random.uniform(1, 6, N)
\#assemblies = Jangenerate\_assembly.generate\_assembly\_solve(N, \sqcup
→params_assembly_num, params_assembly_density)
assemblies = [[0, 1], [2, 3], [4, 5], [6, 7]]
# Output 0, 1 type spikes
spikeCount = Jangenerate_SpikeCount.generateSpikeCountSolve(N, T, dT, u
⇒assemblies, (1, 6), eventDur, eventProb, eventMult, showPlot)
# Transform to -1, 1 distribution
spikeCount[spikeCount == 0] = -1
assemblies_list.append(assemblies)
spikeCount_list.append(spikeCount)
print(assemblies)
print("_____")
binary_list.append(binaryOutput(assemblies))
```

```
[[0, 1], [2, 3], [4, 5], [6, 7]]
```

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```
[80]: multipliers.shape

[80]: (36,)

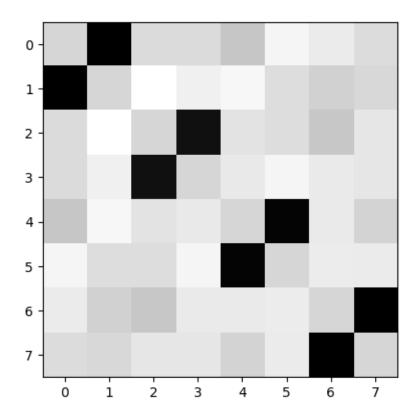
[81]: final_matrix.shape

[81]: (8, 8)
```

### 1 MCH Model

```
matrix[i, j] = mch[index]
         index += 1
upper_matrix = np.triu(matrix)
lower_matrix = np.transpose(upper_matrix)
lower_matrix = np.tril(lower_matrix, -1)
final_matrix = upper_matrix + lower_matrix
#final_matrix = np.where(final_matrix < 1, 0, final_matrix)</pre>
print(final_matrix)
plt.imshow(final_matrix, cmap='gray_r')
[[ 0.
               1.01303113 -0.02407972 -0.02494639 0.07435438 -0.14818347
  -0.10307371 -0.03224052]
                           -0.19892855 -0.12350964 -0.15794094 -0.03631786
 Γ 1.01303113 O.
   0.01942806 -0.01032601]
 [-0.02407972 -0.19892855 0. 0.93407268 -0.06607223 -0.03311937
   0.06769251 -0.07662336]
 [-0.02494639 -0.12350964 0.93407268 0.
                                                     -0.09157319 -0.15035112
 -0.09925145 -0.07970615]
 [ 0.07435438 -0.15794094 -0.06607223 -0.09157319 0.
                                                              0.99581694
 -0.09822713 0.00715672]
  \begin{bmatrix} -0.14818347 & -0.03631786 & -0.03311937 & -0.15035112 & 0.99581694 & 0. \end{bmatrix} 
 -0.10593199 -0.10179064]
  \begin{bmatrix} -0.10307371 & 0.01942806 & 0.06769251 & -0.09925145 & -0.09822713 & -0.10593199 \end{bmatrix} 
                1.01550275]
  \begin{bmatrix} -0.03224052 & -0.01032601 & -0.07662336 & -0.07970615 & 0.00715672 & -0.10179064 \end{bmatrix} 
   1.01550275 0.
                           11
```

[2]: <matplotlib.image.AxesImage at 0x145180400>



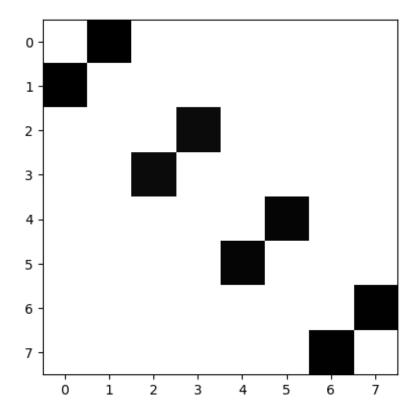
### 2 ACE Model

```
[3]: solver = ClusterExpansion(spikeCount)
     multipliers, ent, clusters, deltaSdict, deltaJdict= solver.solve(threshold = 0.
     ⇔01, full_output=True)
     ace = multipliers
     matrix = np.zeros((N, N))
     index = N
     for i in range(N):
         for j in range(i+1, N):
             matrix[i, j] = ace[index]
             index += 1
     upper_matrix = np.triu(matrix)
     lower_matrix = np.transpose(upper_matrix)
     lower_matrix = np.tril(lower_matrix, -1)
     final_matrix = upper_matrix + lower_matrix
     #final_matrix = np.where(final_matrix < 1, 0, final_matrix)</pre>
     print(final_matrix)
```

```
plt.imshow(final_matrix, cmap='gray_r')
```

```
adaptiveClusterExpansion: Clusters of size 2
adaptiveClusterExpansion: Clusters of size 3
              1.01545824 0.
[[0.
                                      0.
                                                  0.
                                                               0.
  0.
              0.
 [1.01545824 0.
                          0.
                                      0.
                                                  0.
                                                               0.
  0.
              0.
 [0.
                          0.
                                      0.97035554 0.
                                                               0.
              0.
  0.
              0.
 [0.
                          0.97035554 0.
                                                               0.
              0.
                                                  0.
  0.
              0.
                         ]
 [0.
              0.
                          0.
                                      0.
                                                  0.
                                                               0.9941811
  0.
              0.
                         ]
 [0.
              0.
                          0.
                                      0.
                                                  0.9941811
                                                              0.
  0.
              0.
 [0.
              0.
                          0.
                                      0.
                                                  0.
                                                               0.
  0.
              1.00594956]
 [0.
              0.
                          0.
                                      0.
                                                  0.
                                                               0.
                         ]]
  1.00594956 0.
```

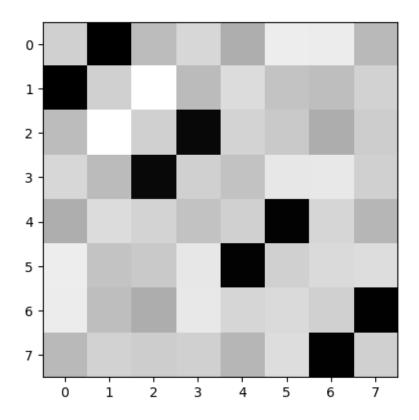
[3]: <matplotlib.image.AxesImage at 0x144f04f40>



### 3 Pseudo Model

```
[4]: solver = Pseudo(spikeCount)
    pse= solver.solve()
    matrix = np.zeros((N, N))
    index = N
    for i in range(N):
        for j in range(i+1, N):
            matrix[i, j] = pse[index]
            index += 1
    upper matrix = np.triu(matrix)
    lower matrix = np.transpose(upper matrix)
    lower_matrix = np.tril(lower_matrix, -1)
    final_matrix = upper_matrix + lower_matrix
    #final_matrix = np.where(final_matrix < 1, 0, final_matrix)</pre>
    print(final_matrix)
    plt.imshow(final_matrix, cmap='gray_r')
    [[ 0.00000000e+00 1.02186963e+00 1.00177565e-01 -3.15808332e-02
       1.65447549e-01 -1.39980914e-01 -1.36105633e-01 1.12089627e-01]
     -5.68304628e-02 5.78230594e-02 9.05254518e-02 -8.95997432e-03]
     [ 1.00177565e-01 -2.31713124e-01 0.0000000e+00 9.79936502e-01
      -1.52974581e-02 3.60193953e-02 1.74557223e-01 1.79062457e-02]
     [-3.15808332e-02 1.04891596e-01 9.79936502e-01 0.00000000e+00
      7.11390400e-02 -1.12311063e-01 -1.16178363e-01 -8.21162190e-04]
     [ 1.65447549e-01 -5.68304628e-02 -1.52974581e-02 7.11390400e-02
       0.00000000e+00 1.00232449e+00 -2.84193351e-02 1.30428231e-01
     [-1.39980914e-01 5.78230594e-02 3.60193953e-02 -1.12311063e-01
       1.00232449e+00 0.00000000e+00 -4.64647508e-02 -6.49127954e-02]
     [-1.36105633e-01 9.05254518e-02 1.74557223e-01 -1.16178363e-01
      -2.84193351e-02 -4.64647508e-02 0.00000000e+00 1.00878942e+00]
     [ 1.12089627e-01 -8.95997432e-03 1.79062457e-02 -8.21162190e-04
       1.30428231e-01 -6.49127954e-02 1.00878942e+00 0.00000000e+00]]
```

[4]: <matplotlib.image.AxesImage at 0x144f42860>



## 4 MPF Model

```
[6]: solver = MPF(spikeCount)
mpf= solver.solve()

matrix = np.zeros((N, N))
index = N
for i in range(N):
    for j in range(i+1, N):
        matrix[i, j] = mpf[index]
        index += 1
upper_matrix = np.triu(matrix)

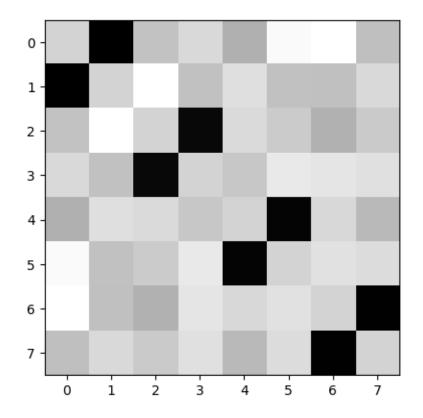
lower_matrix = np.transpose(upper_matrix)
lower_matrix = np.tril(lower_matrix, -1)

final_matrix = upper_matrix + lower_matrix
#final_matrix = np.where(final_matrix < 1, 0, final_matrix)
print(final_matrix)

plt.imshow(final_matrix, cmap='gray_r')</pre>
```

```
[[ 0.
             1.02397964 0.08558408 -0.02643278 0.17544872 -0.18313193
 -0.20591132 0.10247674]
                        [ 1.02397964 0.
  0.09810859 -0.02267283]
[0.08558408 - 0.20976937 0. 0.97941143 - 0.02743291 0.04268213
  0.16993604 0.04699703]
[-0.02643278 \quad 0.09081944 \quad 0.97941143 \quad 0. \quad 0.06239857 \quad -0.10342833
 -0.0797715 -0.05658309]
[ 0.17544872 -0.05107257 -0.02743291 0.06239857 0.
                                                           1.00305477
 -0.01842449 0.12931304]
[-0.18313193 0.09142167 0.04268213 -0.10342833 1.00305477 0.
 -0.06487989 -0.03863889]
 \begin{bmatrix} -0.20591132 & 0.09810859 & 0.16993604 & -0.0797715 & -0.01842449 & -0.06487989 \end{bmatrix} 
              1.00977897]
[ \ 0.10247674 \ -0.02267283 \ \ 0.04699703 \ -0.05658309 \ \ 0.12931304 \ -0.03863889 
  1.00977897 0.
```

[6]: <matplotlib.image.AxesImage at 0x144b0e290>

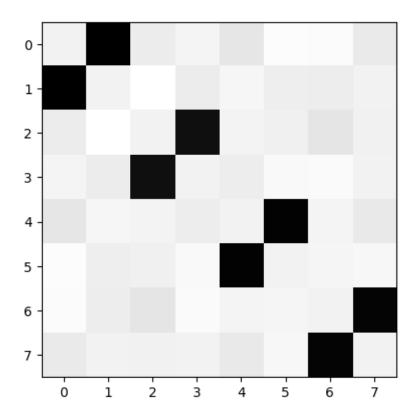


#### 5 RMF Model

```
[7]: solver = RegularizedMeanField(spikeCount)
    rmf= solver.solve()
    matrix = np.zeros((N, N))
    index = N
    for i in range(N):
        for j in range(i+1, N):
            matrix[i, j] = rmf[index]
            index += 1
    upper_matrix = np.triu(matrix)
    lower_matrix = np.transpose(upper_matrix)
    lower_matrix = np.tril(lower_matrix, -1)
    final_matrix = upper_matrix + lower_matrix
    #final_matrix = np.where(final_matrix < 1, 0, final_matrix)</pre>
    print(final_matrix)
    plt.imshow(final_matrix, cmap='gray_r')
    coocSampleCovariance: WARNING: using ad-hoc 'Laplace' correction
    [[ 0.00000000e+00 7.32738824e+00 1.89250048e-01 -6.29562465e-02
      3.82075117e-01 -2.94658366e-01 -2.62923594e-01 2.33380823e-01
     -1.12281068e-01 1.20558126e-01 1.62064147e-01 7.82364845e-03]
     [ 1.89250048e-01 -4.01543379e-01 0.00000000e+00 6.86932428e+00
     -3.73725328e-02 7.45435555e-02 3.97575172e-01 4.89358045e-02]
     [-6.29562465e-02 1.96712921e-01 6.86932428e+00 0.00000000e+00
       1.49723835e-01 -2.09000852e-01 -2.37038170e-01 6.25890826e-03]
     [ 3.82075117e-01 -1.12281068e-01 -3.73725328e-02 1.49723835e-01
```

0.00000000e+00 7.22428926e+00 -5.73914043e-02 2.90577883e-01]
[-2.94658366e-01 1.20558126e-01 7.45435555e-02 -2.09000852e-01 7.22428926e+00 0.00000000e+00 -8.51147500e-02 -1.39791116e-01]
[-2.62923594e-01 1.62064147e-01 3.97575172e-01 -2.37038170e-01 -5.73914043e-02 -8.51147500e-02 0.00000000e+00 7.18914652e+00]
[ 2.33380823e-01 7.82364845e-03 4.89358045e-02 6.25890826e-03

2.90577883e-01 -1.39791116e-01 7.18914652e+00 0.00000000e+00]]
[7]: <matplotlib.image.AxesImage at 0x1446e34f0>



[]: