



Hopfield Model

Computer modeling of
complex systems

OVERVIEW

- 1 | Hopfield network
first task
- 2 | Asynchronous updates
second task
- 3 | Increasing patterns
extra task

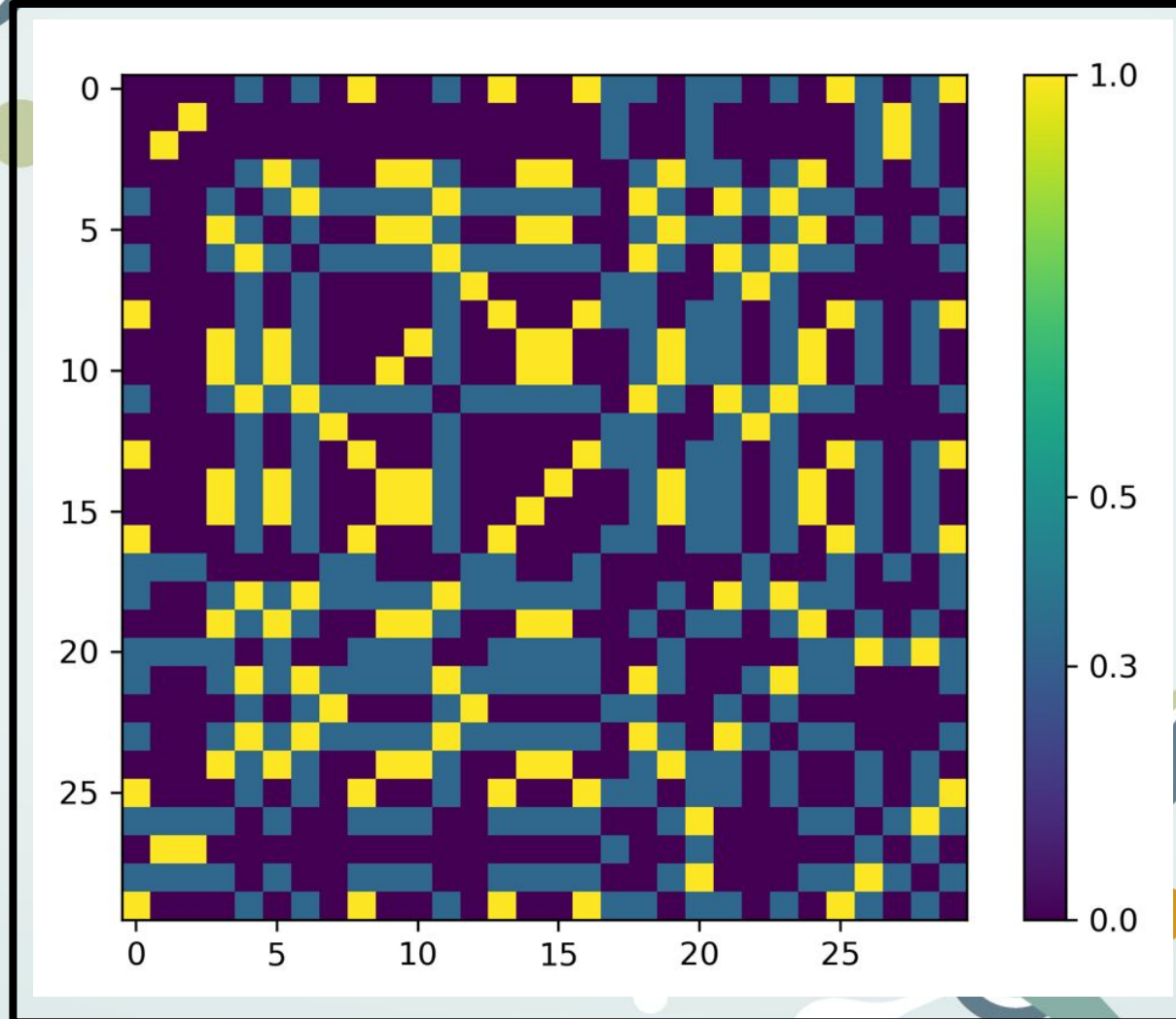
Receiving matrix: w

When learning n binary (spin) patterns x^μ :

$$\mathbf{w} = \frac{1}{n} \sum_{\mu=1}^n \mathbf{x}^\mu \otimes \mathbf{x}^\mu - \mathbf{1}$$

```
def hopfield_network(training_matrices):  
    """  
    :param training_matrices: list of matrices, which we will use to training our network.  
    :return: matrix, which describes network 'after training'.  
    """  
    n_matrix = len(training_matrices)  
    size_matrix = len(training_matrices[0].A[0])  
  
    w_matrix = np.zeros((size_matrix, size_matrix)) # network after training  
    for j in range(0, n_matrix):  
        # create training pattern  
        matrix = training_matrices[j]  
        pattern = matrix.flatten()  
        # update network  
        w_matrix = w_matrix + np.outer(pattern, pattern) / n_matrix  
  
    # Subtract identity matrix  
    identity_matrix = np.identity(size_matrix)  
    w_matrix = w_matrix - identity_matrix  
  
    return np.matrix(w_matrix)
```

Martix w itself



Updating the network

The energy of i th spin can be written as:

$$E_i = -\frac{1}{2}x_i \left(\sum_j w_{ij}x_j \right) = -\frac{1}{2}x_i h(i)$$

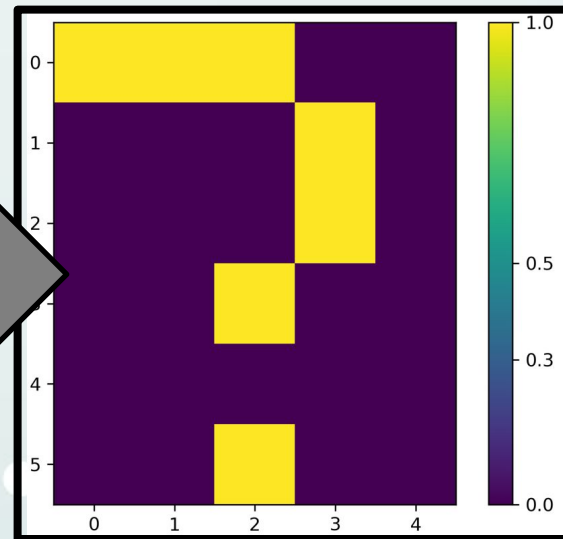
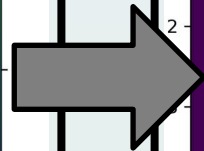
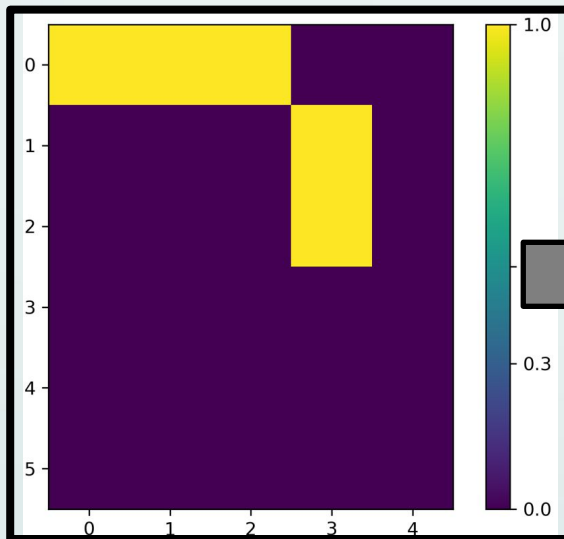
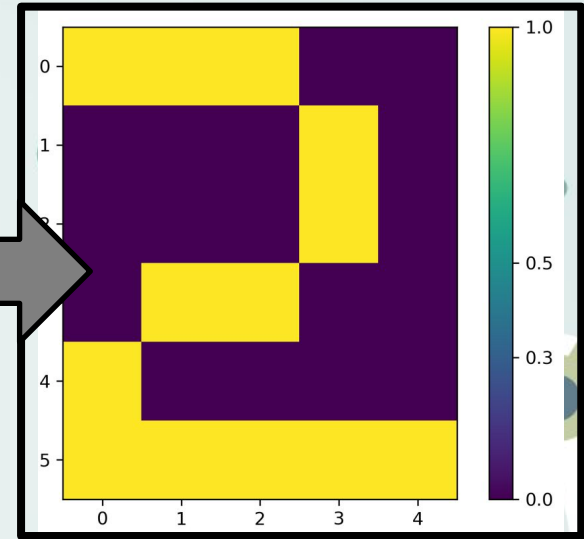
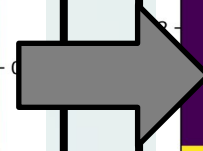
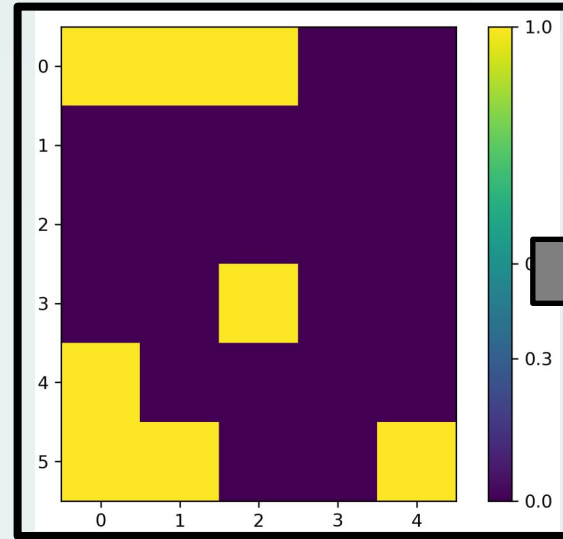
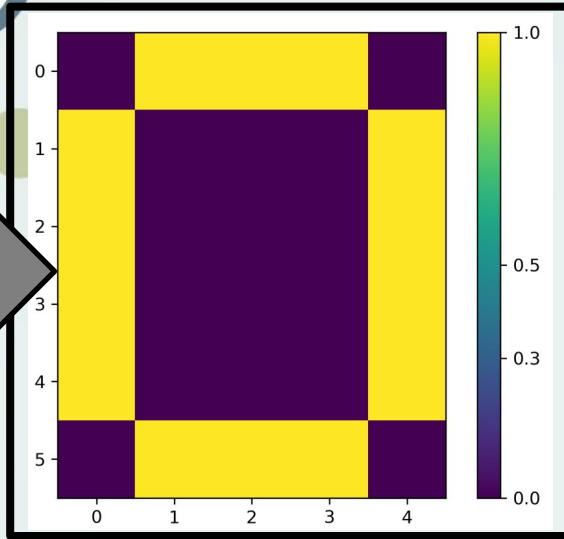
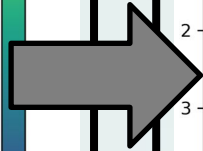
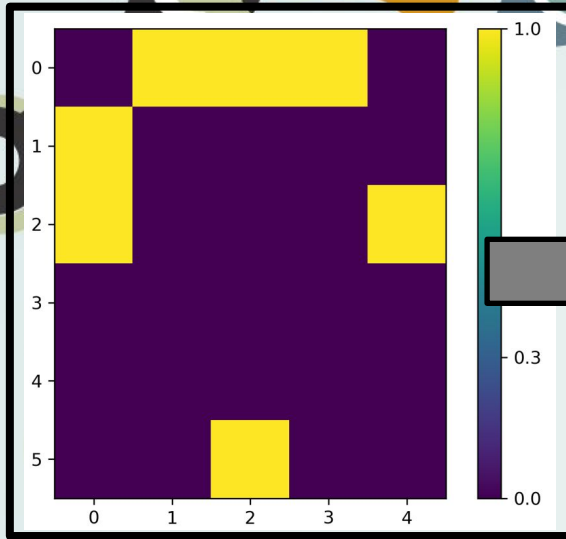
where $h(i)$ is the field acting on the spin. If the product $x_i h(i)$ is negative then the field would try to flip the spin:

$$\mathbf{x}(t+1) = \text{sgn}(\mathbf{w} \cdot \mathbf{x}(t))$$

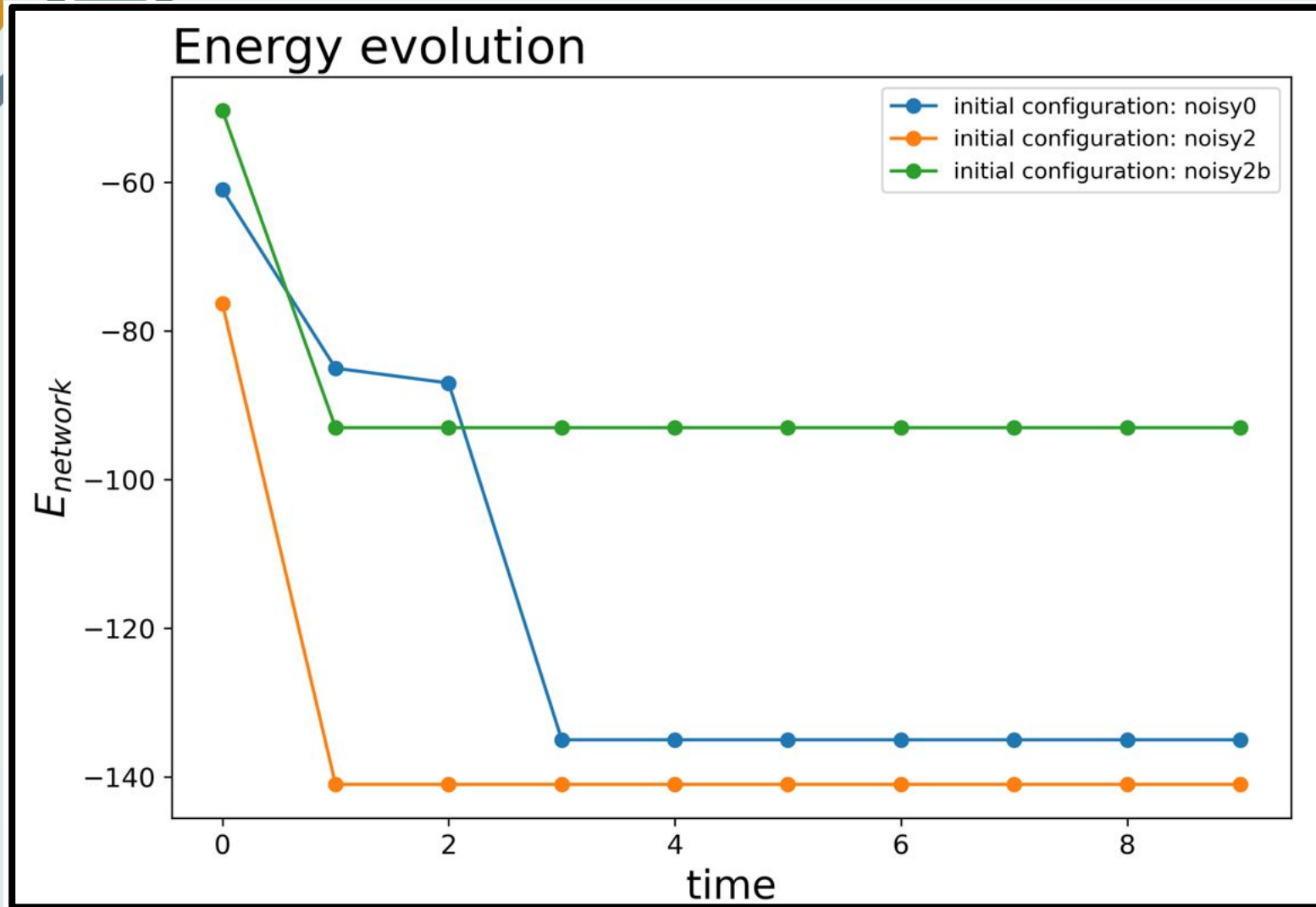
```
def test_network(pattern, w_matrix, steps=10):
    """
    :param pattern: pattern, which we will use as starting configuration
                    of network.
    :param w_matrix: pattern, which we get after training.
    :param steps: number of updates, which we want make.
    :return: ending configuration and list contain energy of whole network after
            updating the network.
    """
    n_elements = pattern.size
    E_network_list = []
    for step in range(0, steps):
        # do one-step operation of hopfield network
        E_list = []
        for j in range(0, n_elements):
            x_i = pattern[0, j]
            # calculate: sum over j of 'w_ij * x_j'
            h_i = w_matrix[j].dot(pattern.T)
            # compute energy
            E_i = float(- 0.5 * x_i * h_i)
            E_list.append(E_i)
        # update test matrix
        pattern = np.sign(w_matrix.dot(pattern.T).T)
        E_network = sum(E_list)
        E_network_list.append(E_network)

    return pattern, E_network_list
```

Results task 1



Energy Evolution



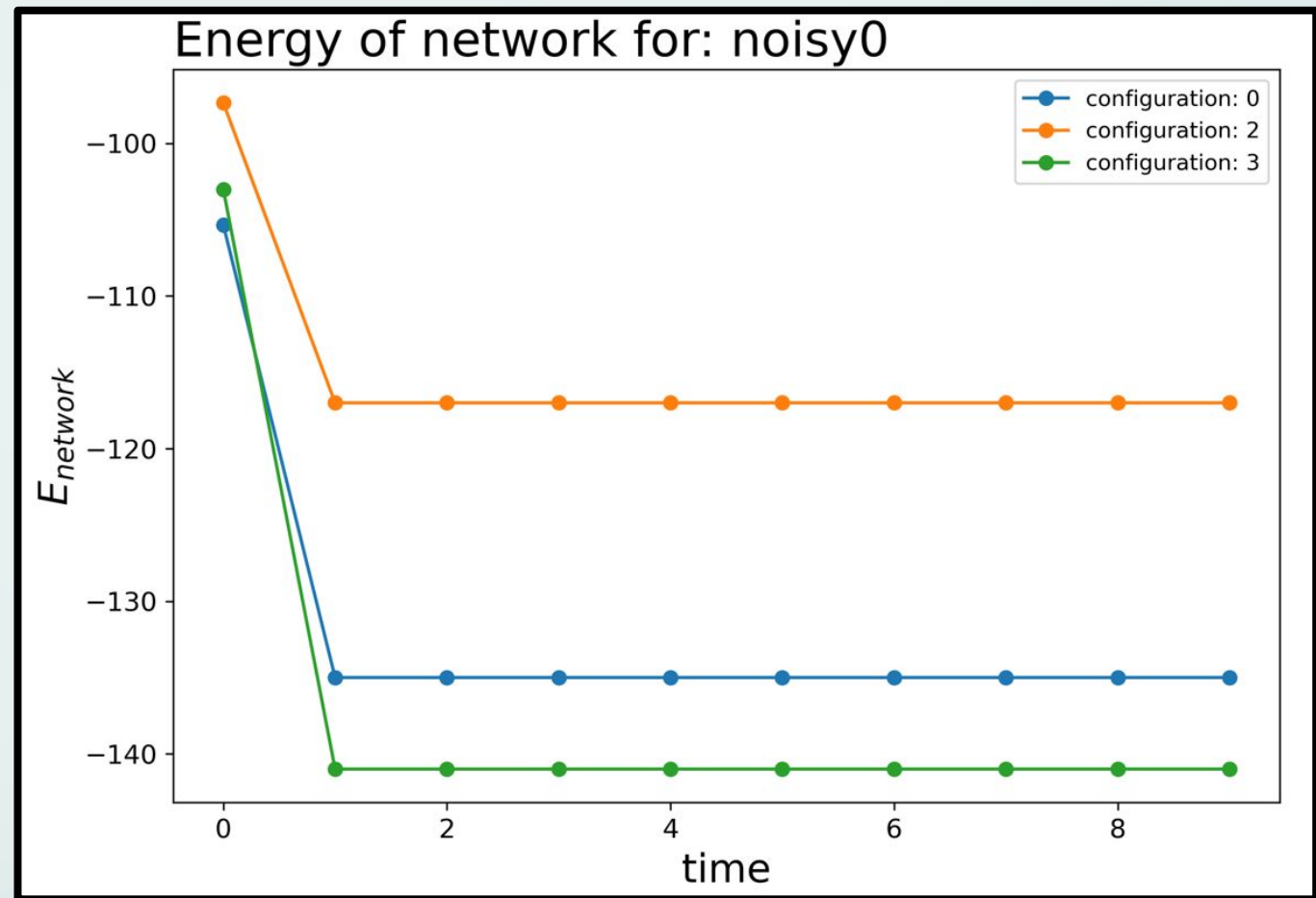
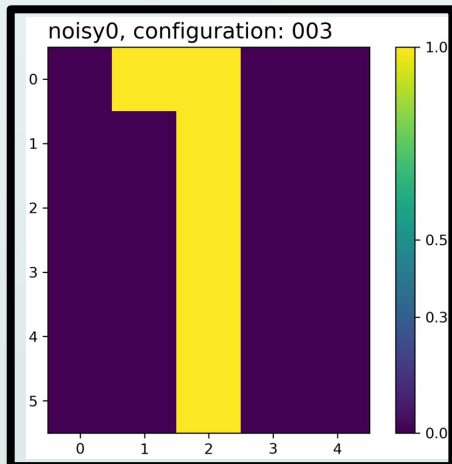
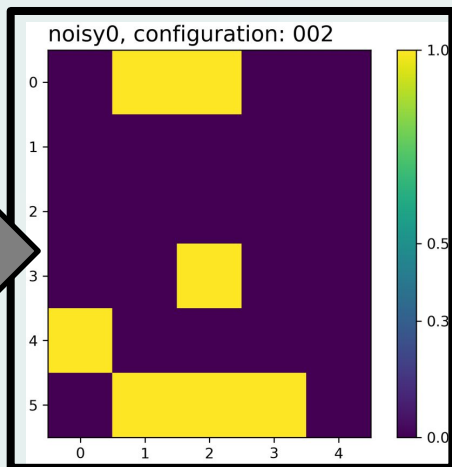
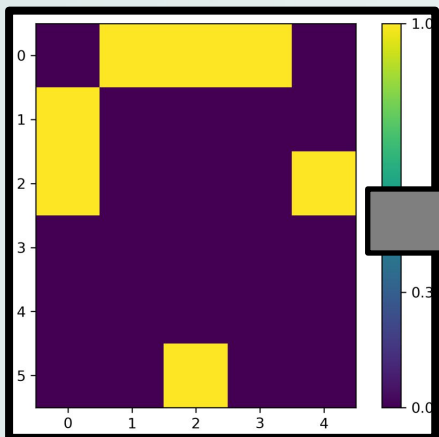
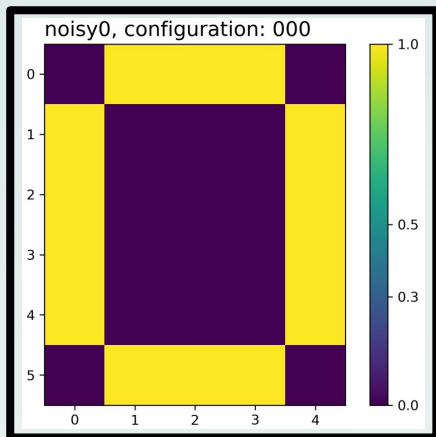
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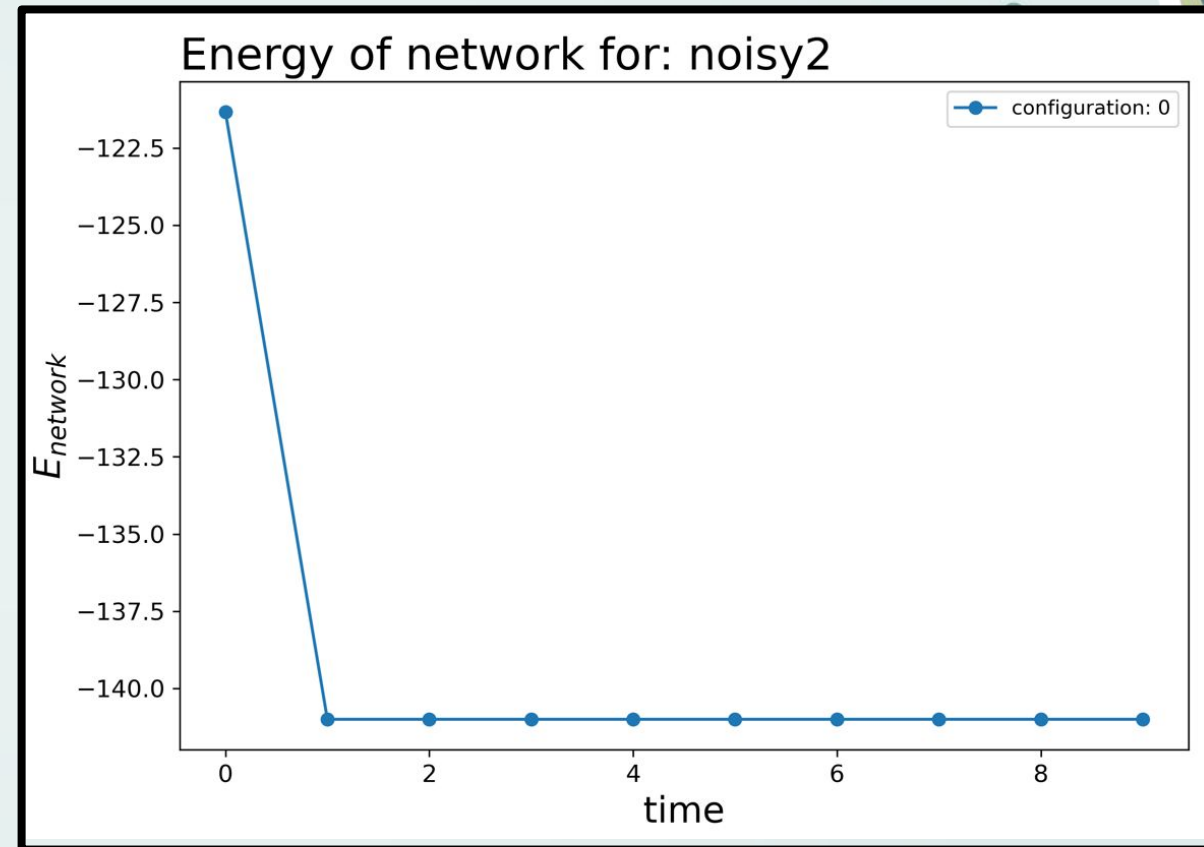
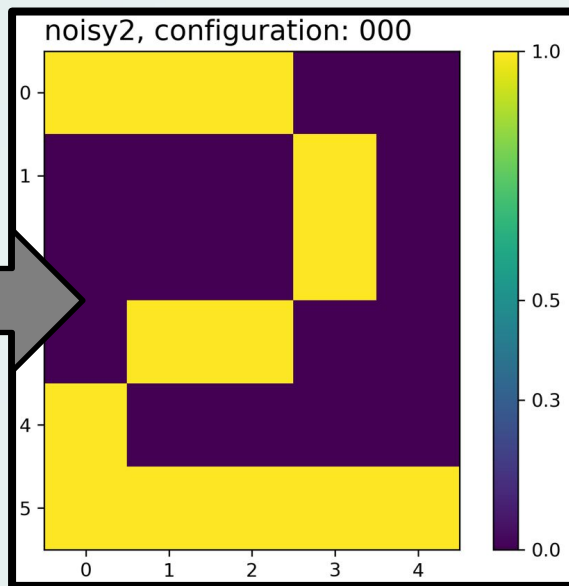
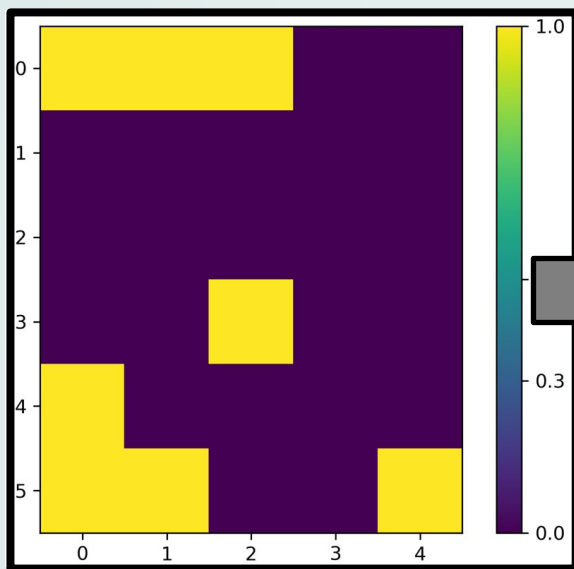
Updating the network

```
def test_network_asynchronously(pattern, w_matrix, steps=10):  
    """  
    :param pattern: pattern, which we will use as starting configuration  
        of network.  
    :param w_matrix: pattern, which we get after training.  
    :param steps: number of updates, which we want make.  
    :return: ending configuration and list contain energy of whole network after  
        updating the network.  
    """  
    n_elements = pattern.size  
    E_network_list = []  
    for step in range(0, steps):  
        # do one-step operation of hopfield network  
        E_list = []  
        # ordering of choosing spin, which are picked up randomly  
        order_index = random.sample(range(n_elements), n_elements)  
  
        for i in range(0, n_elements):  
            index = order_index[i]  
            x_i = pattern[0, index]  
            # flipped spin  
            if x_i == -1:  
                x_i_flipped = 1  
            else:  
                x_i_flipped = -1  
            pattern_flipped = pattern.copy()  
            pattern_flipped[0, index] = x_i_flipped  
            # calculate: sum over j of 'w_ij * x_j'  
            h_i = w_matrix[index].dot(pattern.T)  
            h_i_flipped = w_matrix[index].dot(pattern_flipped.T)  
            # compute energy  
            E_i = float(- 0.5 * x_i * h_i)  
            E_i_flipped = float(- 0.5 * x_i_flipped * h_i_flipped)  
            if E_i_flipped < E_i:  
                # we flip spin  
                E_list.append(E_i_flipped)  
                pattern = pattern_flipped  
            else:  
                E_list.append(E_i)  
  
        # update test matrix  
        pattern = np.sign(w_matrix.dot(pattern.T).T)  
        E_network = sum(E_list)  
        E_network_list.append(E_network)
```

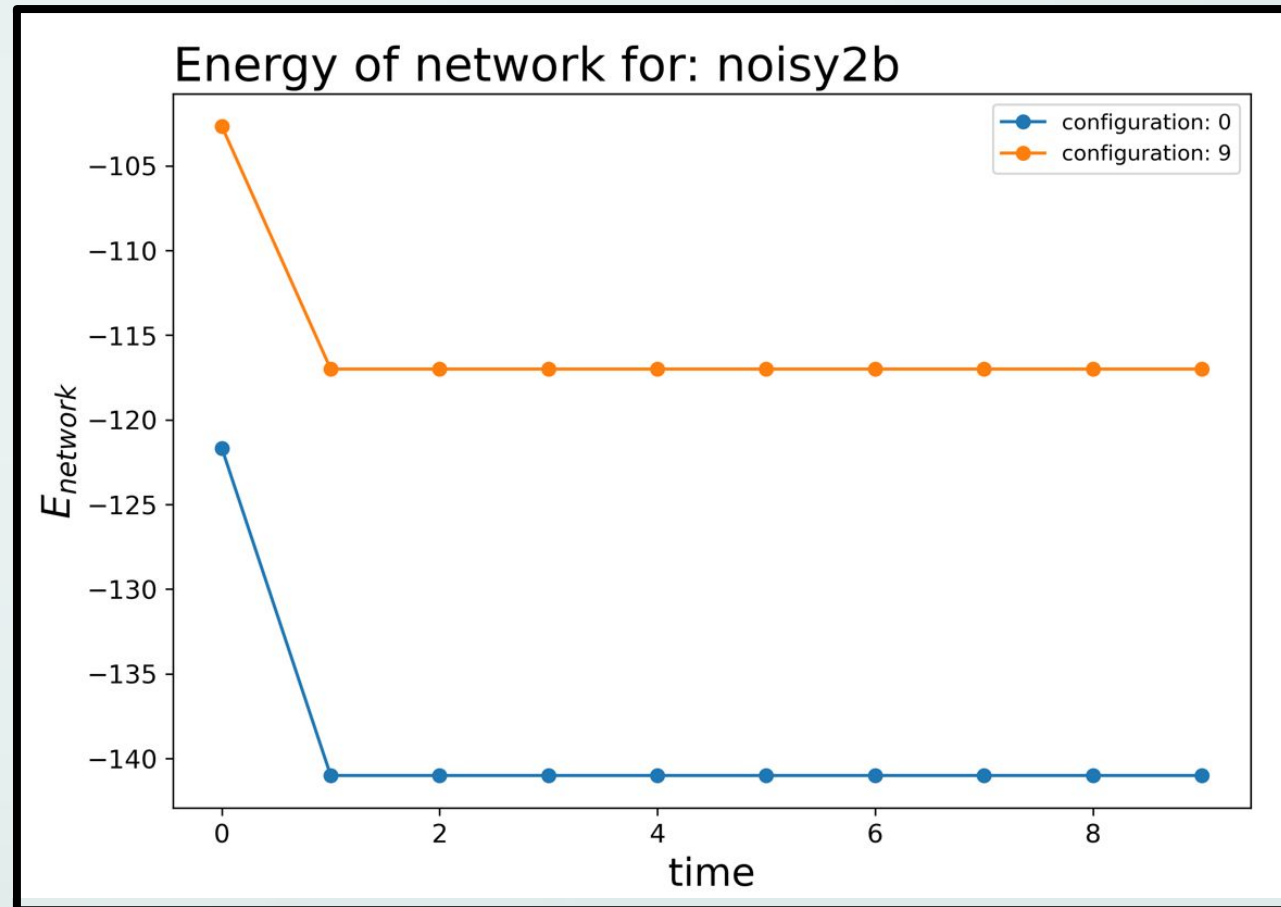
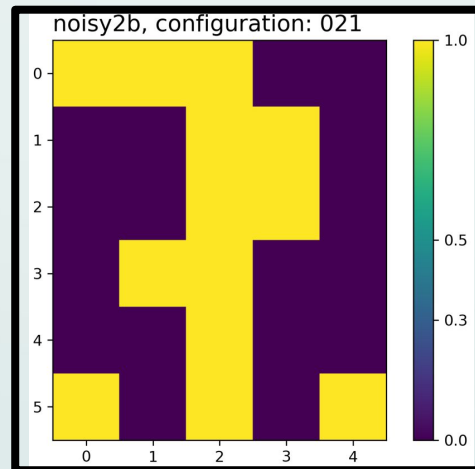
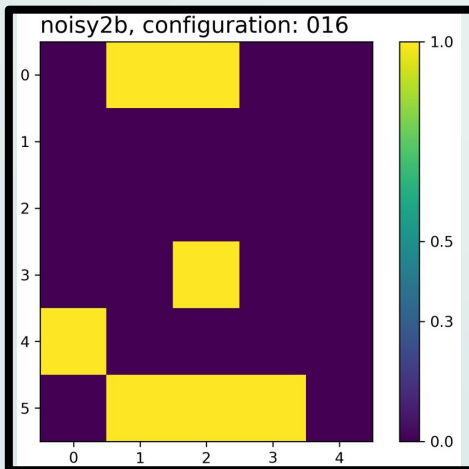
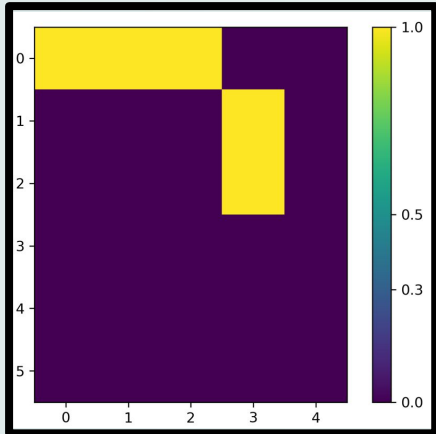
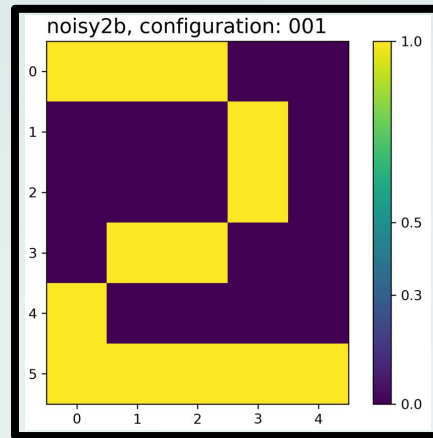
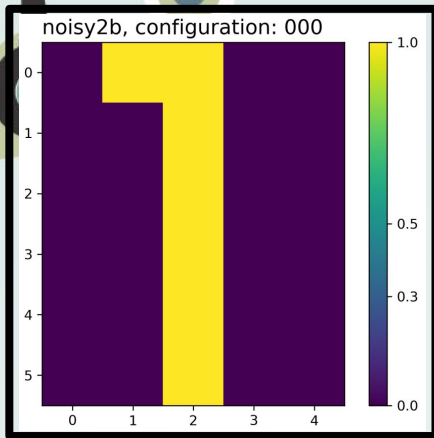
configuration: noisy-0



configuration: noisy-2



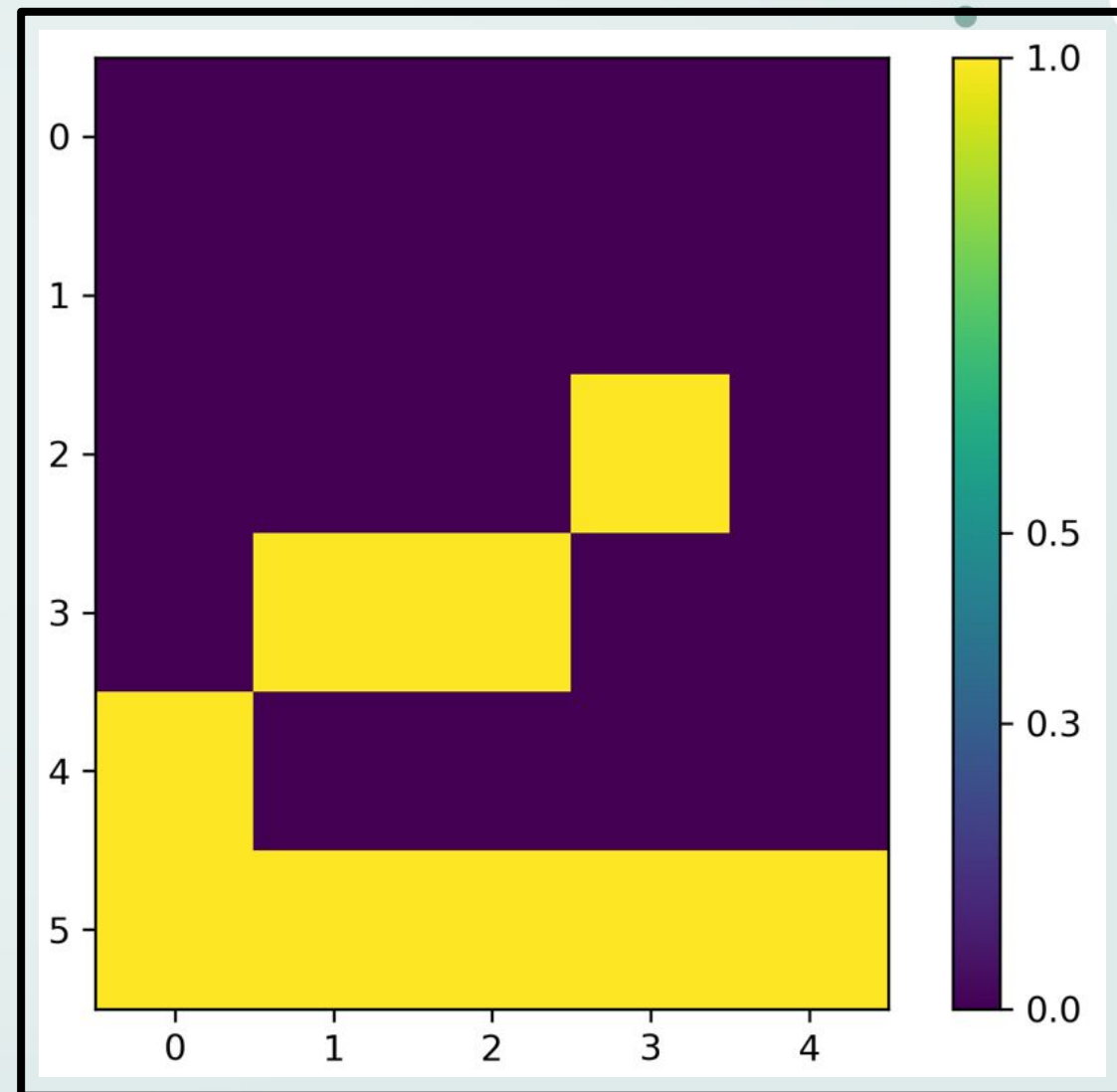
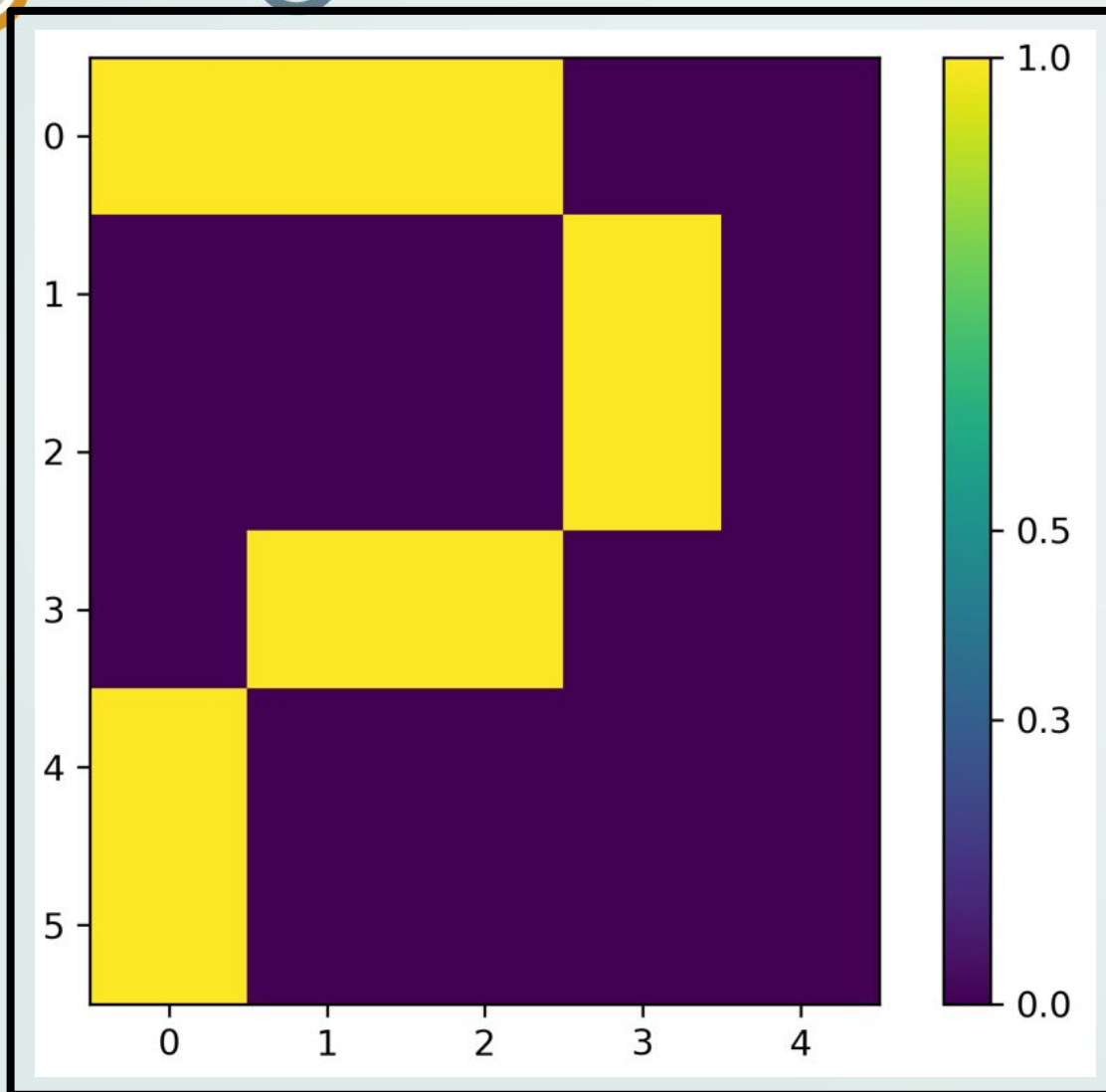
configuration: noisy-2b



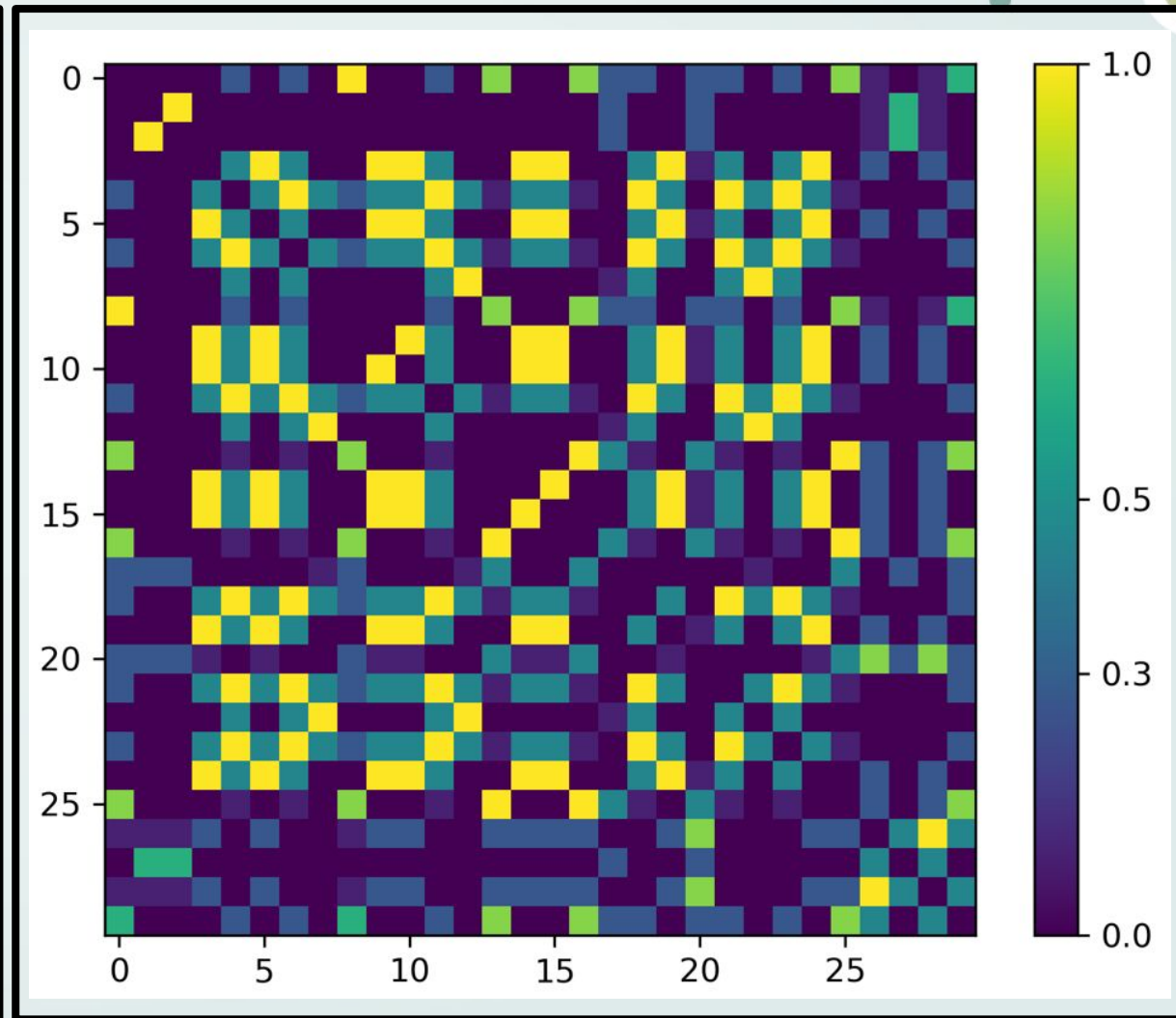
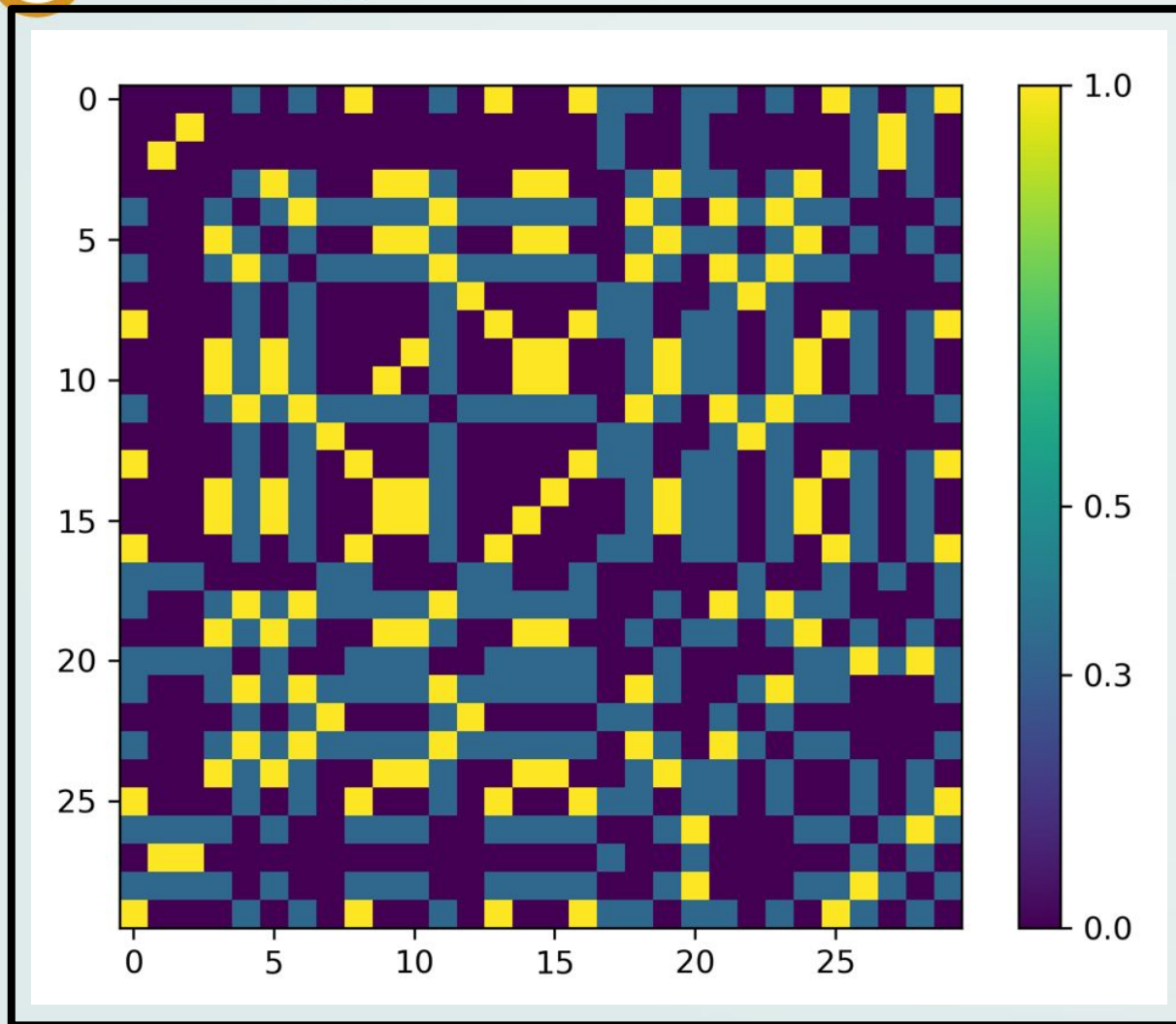
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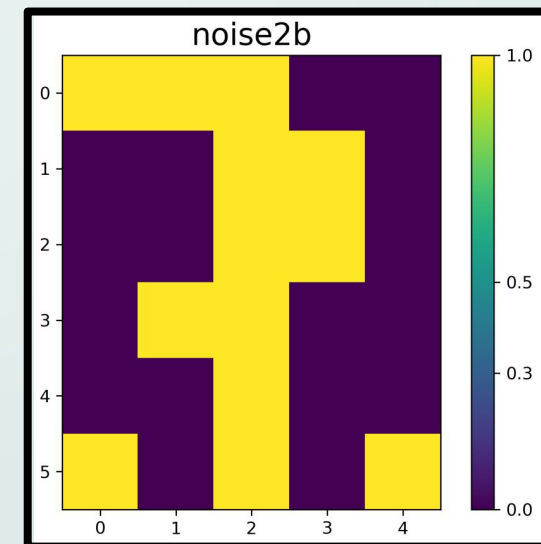
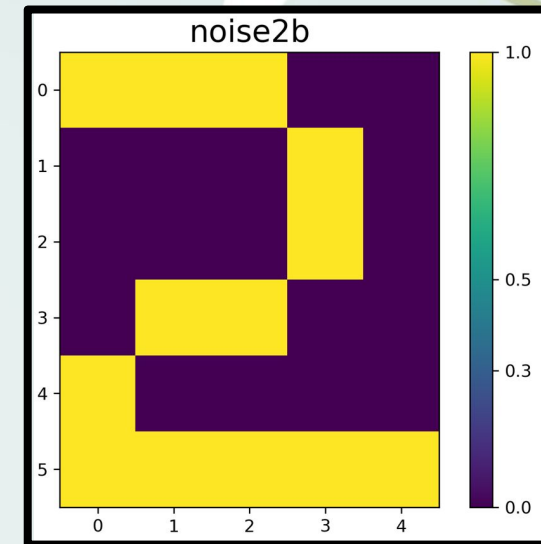
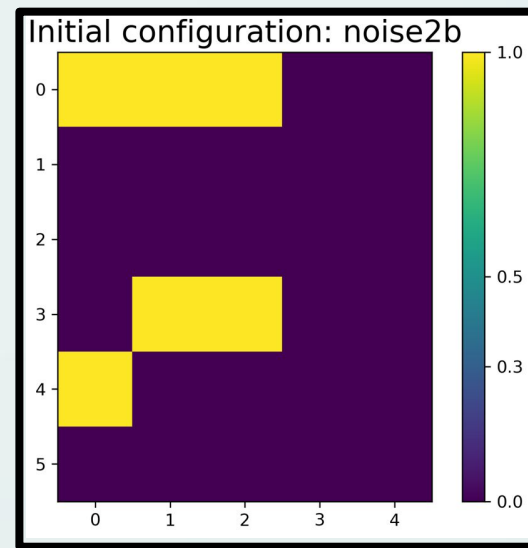
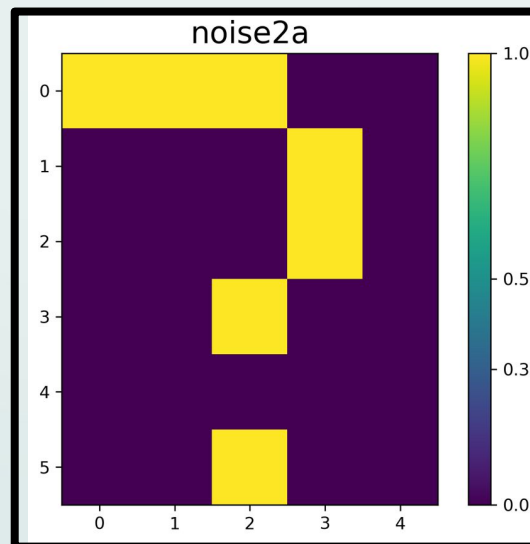
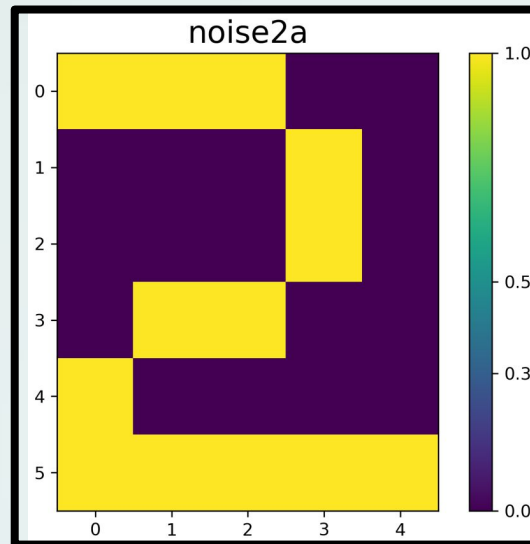
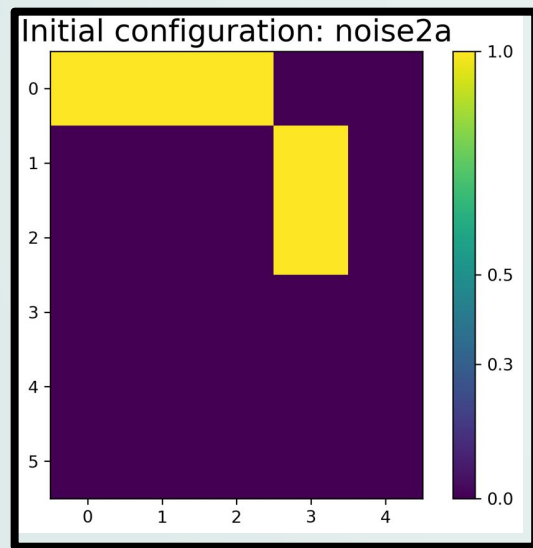
Added patterns



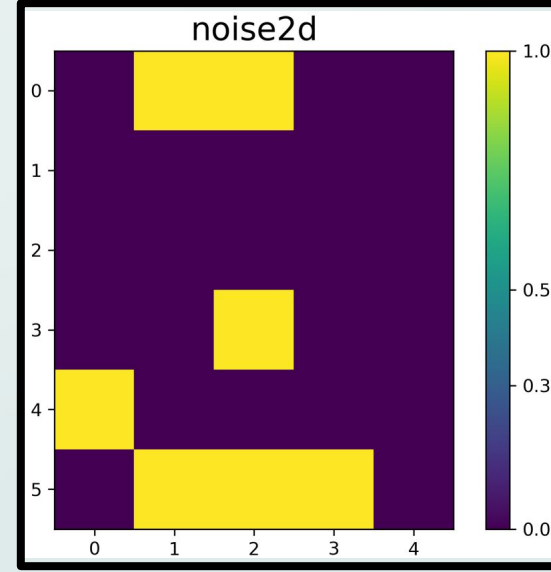
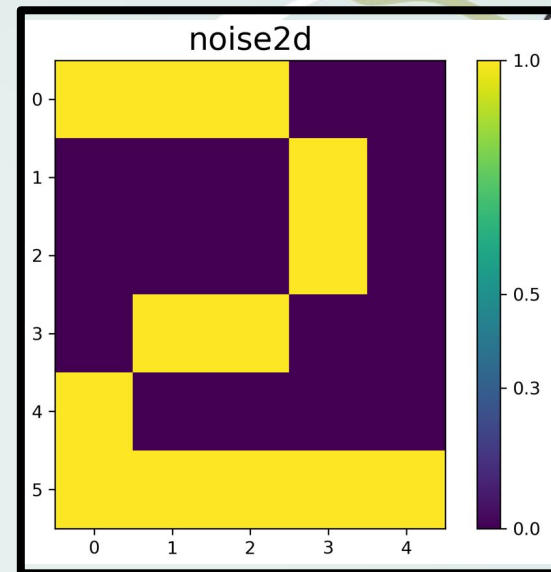
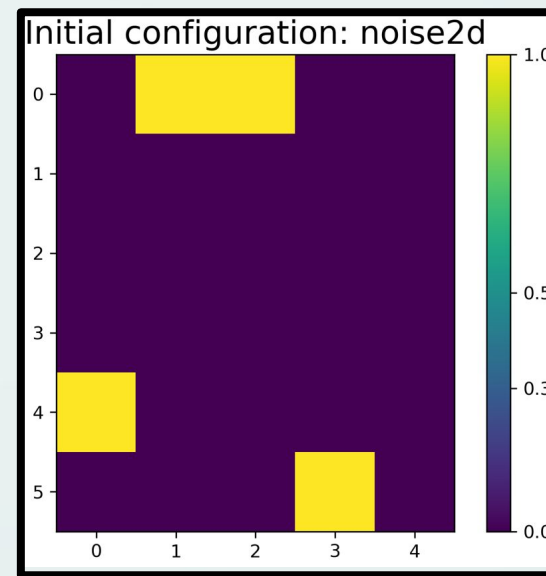
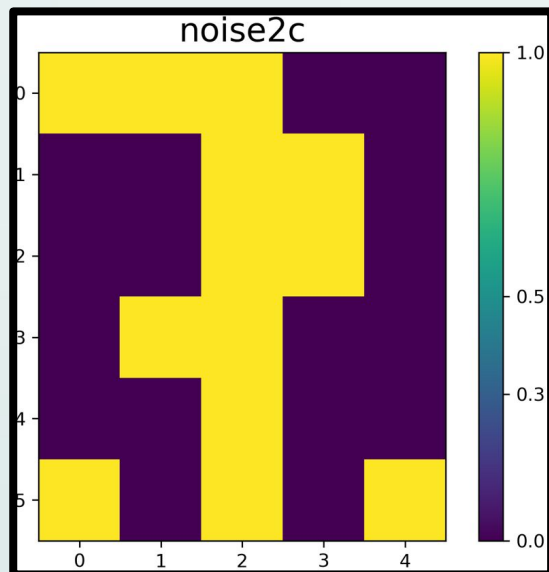
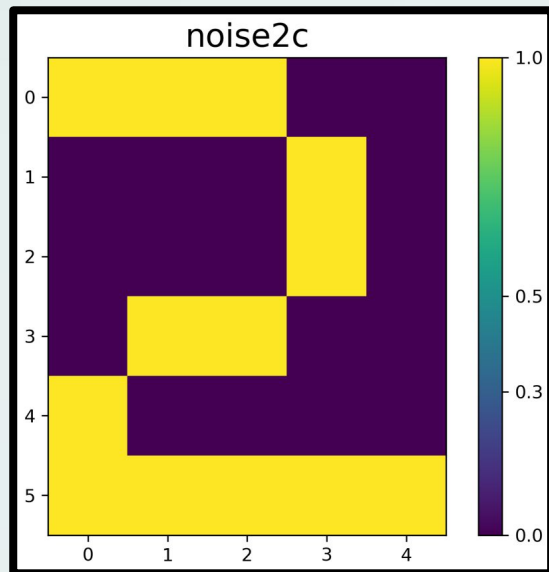
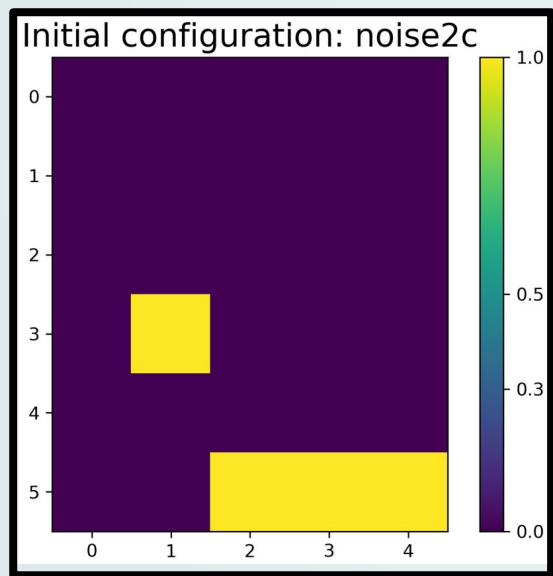
Comparison w matrix



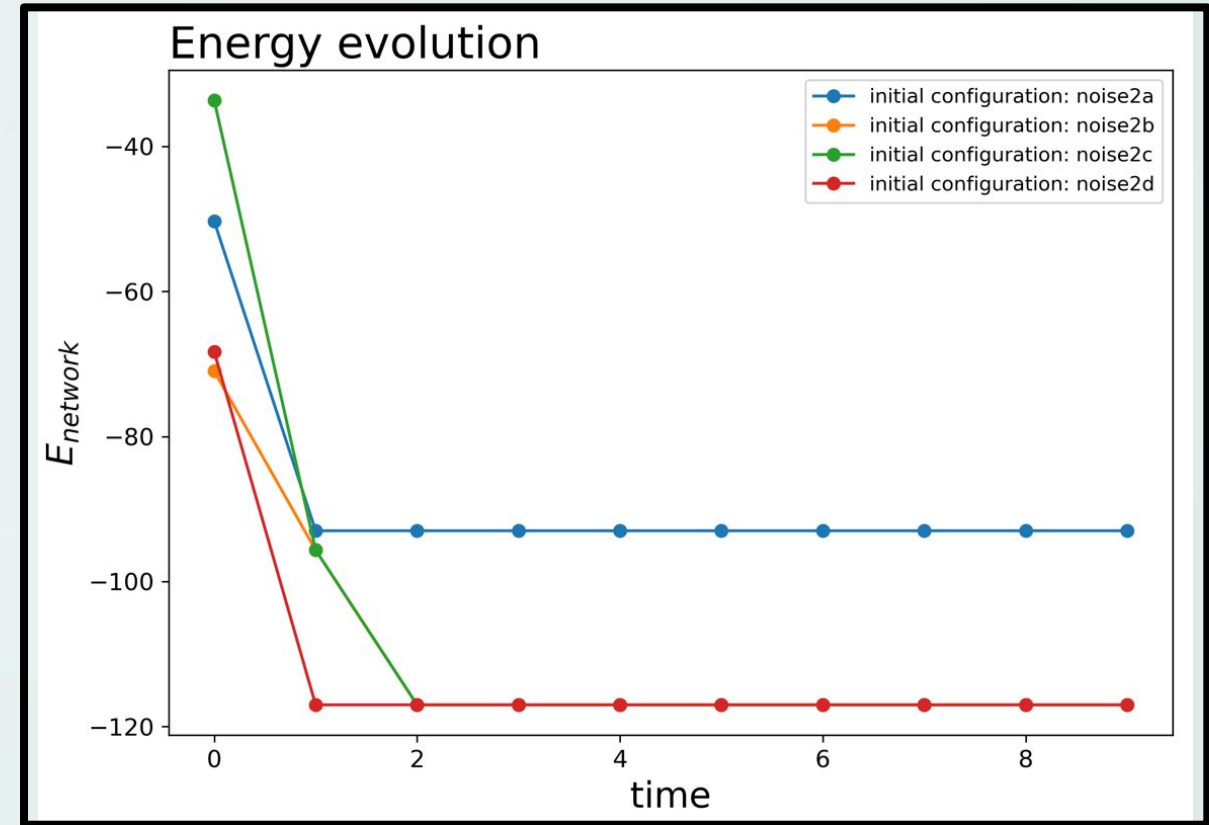
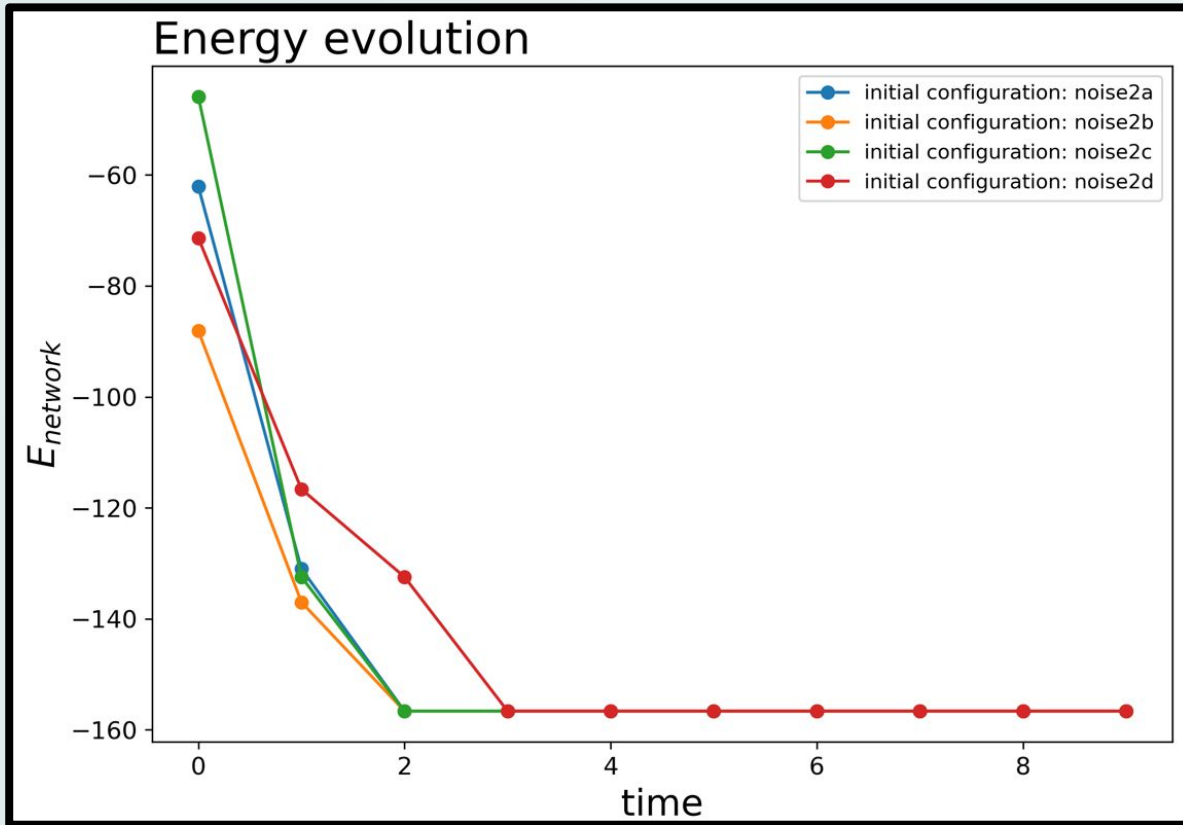
Comparison part one



Comparison part two



Energy evolution





Thank you!