1. Suppose we only have two neurons in the Hopfield network. Assume we have (i) \( w_{12} = w_{21} = 1 \) or (ii) \( w_{12} = w_{21} = -1 \).

In the case of asynchronous updating, show that for (i) there are two attracting fixed points namely \([1, 1]\) and \([-1, -1]\) and that all orbits converge to one of these, whereas for (ii), the attracting fixed points are \([-1, 1]\) and \([1, -1]\) and all orbits converge to one of these.

In the case of synchronous updating, show that for both (i) and (ii), the fixed points do not attract nearby points and there are orbits which oscillate forever.

**Solution:**

1. Asynchronous updating: We assume the first node is updated followed by the second node (the alternative rule would be similar):

   (i) \( w_{12} = w_{21} = 1 \): We have using the updating rule:

   \[
   11 \rightarrow 11, \text{ thus } 11 \text{ is a fixed point; } -1-1 \rightarrow -1-1, \text{ thus } -1-1 \text{ is a fixed point}
   \]

   \[
   1-1 \rightarrow -1-1, \text{ thus } 1-1 \text{ is in the basin of attraction of } -1-1
   \]

   -11 \rightarrow 11, \text{ thus } -11 \text{ is in the basin of attraction of } 11

   (ii) \( w_{12} = w_{21} = -1 \):

   \[
   1-1 \rightarrow 1-1, \text{ thus } 1-1 \text{ is a fixed point; } 1-1 \rightarrow 1-1, \text{ thus } 1-1 \text{ is a fixed point}
   \]

   \[
   11 \rightarrow -11, \text{ thus } 11 \text{ is in the basin of attraction of } -11
   \]

   \[
   -1-1 \rightarrow 1-1, \text{ thus } -1-1 \text{ is in the basin of attraction of } 1-1
   \]

2. Synchronous updating:

   (i) \( w_{12} = w_{21} = 1 \):

   \[
   11 \rightarrow 11, \text{ thus } 11 \text{ is a fixed point; } -1-1 \rightarrow -1-1, \text{ thus } -1-1 \text{ is a fixed point}
   \]

   \[
   1-1 \rightarrow -11, \quad -11 \rightarrow 1-1, \text{ thus } \{-11, 1-1\} \text{ is an orbit of period 2}
   \]
(ii) \( w_{12} = w_{21} = -1 \):

\[ 11 \rightarrow -1-1, \quad -1-1 \rightarrow 11, \text{ thus } \{11, -1-1\} \text{ is an orbit of period 2} \]

\[ 1-1 \rightarrow 1-1, \text{ thus } 1-1 \text{ is a fixed point; } -11 \rightarrow -11, \text{ thus } -11 \text{ is a fixed point} \]

2. Consider the energy function

\[ E = -\frac{1}{2} \sum_{i,j=1}^{N} w_{ij} x_i x_j \]

in the Hopfield network. We update \( x_m \) to \( x'_m \) and denote the new energy by \( E' \). Show that \( E' - E = \sum_{i \neq m} w_{mi} x_i (x_m - x'_m) \).

**Solution:** We have:

\[ E' - E = -\frac{1}{2} \sum_{i,j=1}^{N} w_{ij} x'_i x'_j + \frac{1}{2} \sum_{i,j=1}^{N} w_{ij} x_i x_j, \]

where \( x'_i = x_i \) for \( i \neq m \). Thus, all terms for which \( i \neq m \) and \( j \neq m \) will cancel out. All other terms will have \( i = m \) or \( j = m \). Since \( w_{ij} \) is symmetric, we have \( w_{im} x_i x_m = w_{mi} x_m x_i \) and also \( w_{im} x'_i x'_m = w_{mi} x'_m x'_i \) for \( i \neq m \). Thus, terms that have \( i = m \) or \( j = m \) come in pairs of equal value and we obtain:

\[ E' - E = -\sum_{i \neq m} w_{mi} x_i x'_m + \sum_{i \neq m} w_{mi} x_i x_m \]

\[ = \sum_{i \neq m} w_{mi} x_i (x_m - x'_m). \]

3. Suppose we have a Hopfield network with \( N \) nodes that has stored \( p \) patterns \( \vec{x}^k \), for \( k = 1, \ldots, p \). Consider for any three values \( 1 \leq k_1, k_2, k_3 \leq p \), the mixture state

\[ \vec{x}^{\text{mix}} = \text{sgn}(\vec{x}^{k_1} + \vec{x}^{k_2} + \vec{x}^{k_3}) \]

(i) Show that on average \( x^{\text{mix}}_i \) has the same sign as \( x^{k_1}_i \) three times out of four.

**Solution:** They will only fail to have the same sign when both \( x^{k_2}_i \) and \( x^{k_3}_i \) have the opposite sign of \( x^{k_1}_i \) which happens one out of four.

(ii) Deduce that the Hamming distance of \( \vec{x}^{\text{mix}} \) from \( \vec{x}^{k_1} \) is \( N/4 \).
Solution: On average $x_i^{k_2}$ and $x_i^{k_3}$ will differ from each other one out of four times. Since the patterns are random and we have $N$ nodes, the average number of components that are different in $\vec{x}^{\text{mix}}$ from $\vec{x}^{k_1}$ is $N/4$.

(iii) Show also that $\sum_{i=1}^{N} x_i^{k_1} x_i^{\text{mix}} = N/2$ on average.

Solution: On average $3N/4$ of the components of $\vec{x}^{\text{mix}}$ and $\vec{x}^{k_1}$ are the same, which means that on average $3N/4$ of the products $x_i^{k_1} x_i^{\text{mix}}$ are $+1$. Also on average $N/4$ of the components of $\vec{x}^{\text{mix}}$ and $\vec{x}^{k_1}$ are different, which means that on average $N/4$ of the products $x_i^{k_1} x_i^{\text{mix}}$ are $-1$. Therefore on average $\sum_{i=1}^{N} x_i^{k_1} x_i^{\text{mix}} = (3N/4) \times (1) + (N/4) \times (-1) = N/2$.

(iv) Compute $h_i^{\text{mix}}$ as on page 13 of the notes to derive:

$$h_i^{\text{mix}} = \frac{1}{N} \sum_{\ell} x_i^{\ell} x_j^{\ell} x_j^{\text{mix}} = \frac{1}{2} x_i^{k_1} + \frac{1}{2} x_i^{k_2} + \frac{1}{2} x_i^{k_3} + \text{cross terms}$$

Solution: We have:

$$h_i^{\text{mix}} = \frac{1}{N} \sum_{1 \leq j \leq N, 1 \leq \ell \leq p} x_i^{\ell} x_j^{\ell} x_j^{\text{mix}}$$

in which we separate the contributions of the three terms $\ell = k_1$, $\ell = k_2$ and $\ell = k_3$ from the rest to obtain:

$$h_i^{\text{mix}} = \frac{1}{N} \left( \sum_{j=1}^{N} x_i^{k_1} x_j^{k_1} x_j^{\text{mix}} + x_i^{k_2} x_j^{k_2} x_j^{\text{mix}} + x_i^{k_3} x_j^{k_3} x_j^{\text{mix}} \right) + \text{cross terms}$$

$$= \frac{1}{N} \left( x_i^{k_1} \sum_{j=1}^{N} x_j^{k_1} x_j^{\text{mix}} + x_i^{k_2} \sum_{j=1}^{N} x_j^{k_2} x_j^{\text{mix}} + x_i^{k_3} \sum_{j=1}^{N} x_j^{k_3} x_j^{\text{mix}} \right) + \text{cross terms}$$

By part (iii) each of the three sums above is $N/2$ and we obtain:

$$h_i^{\text{mix}} = \frac{1}{2} x_i^{k_1} + \frac{1}{2} x_i^{k_2} + \frac{1}{2} x_i^{k_3} + \text{cross terms}$$

(v) Conclude that $\vec{x}^{\text{mix}}$ is indeed an attractor of the network in most cases.

Solution: Similar to the case of a single pattern in the notes, here, assuming that the cross talks will be less than $1/2$ in absolute value, which will be for the majority of cases for sufficiently small $p$, we get:

$$\text{sgn}(h_i^{\text{mix}}) = \frac{1}{2} \text{sgn}(x_i^{k_1} + x_i^{k_2} + x_i^{k_3}) = \text{sgn}(x_i^{\text{mix}})$$

and it follows that $x_i^{\text{mix}}$ does not change sign and hence $\vec{x}^{\text{mix}}$ is in most cases an attractor of the network.