

Assignment 2

Question 1 to 2

True. True.

Question 3

1.

To use Hopfield network to retrieve noisy images, we can use the following steps:

1. suppose the noisy image is $x \in R^n$
2. each time, we update the entry x_i of x by $x_i = \text{sign}(\sum_{j=1}^n w_{ij}x_j)$ ($i=1,2,\dots,n$)
3. When the update process converges, we get the denoised image.

To use Hopfield network to retrieve masked images, we can use the following steps:

1. suppose the masked image is $x \in R^n$, and the masked part is $x_i (i \in I)$
2. each time, we update the entry x_i of x by $x_i = \text{sign}(\sum_{j=1}^n w_{ij}x_j)$ ($i \in \{1, 2, \dots, n\} - I$)
3. When the update process converges, we get the unmasked image.

2.

We can calculate that $\forall x$,

$$E(x) = -\frac{1}{2N}x^T\left(\sum_p y_p y_p^T\right)x = -\frac{1}{2N}\sum_p x^T y_p (y_p^T x) = -\frac{1}{2N}\sum_p \|y_p^T x\|^2 = -\frac{\|x\|^2}{2N} = -\frac{1}{2}$$

Thus, all the patterns have the same energy, that is, it memorizes 2^N patterns.

Question 4

2

(Prove of the gradient of the Boltzmann machine)

$$\begin{aligned}
 \nabla_W L(W) &= \nabla_W \left(-\frac{1}{|P|} \sum_{v \in P} \log \left(\sum_h \mathbb{P}(v, h) \right) \right) \\
 &= \nabla_W \left(-\frac{1}{|P|} \sum_{v \in P} \log \left(\sum_h \frac{\exp(y^T W y)}{\sum_{y'} \exp(y'^T W y')} \right) \right) \\
 &= \nabla_W \left(-\frac{1}{|P|} \sum_{v \in P} \log \left(\sum_h \exp(y^T W y) \right) \right) + \nabla_W \left(\log \sum_{y'} \exp(y'^T W y') \right) \\
 &= \nabla_W \left(-\frac{1}{|P|} \sum_{v \in P} \log \left(\sum_h \exp(y^T W y) \right) \right) + \nabla_W \left(\log \sum_{y'} \exp(y'^T W y') \right) \\
 &= -\frac{1}{|P|} \sum_{v \in P} \nabla_W \log \left(\sum_h \exp(y^T W y) \right) + \nabla_W \left(\log \sum_{y'} \exp(y'^T W y') \right) \\
 &= -\frac{1}{|P|} \sum_{v \in P} \frac{\sum_h \exp(y^T W y) (y^T y)}{\sum_h \exp(y^T W y)} + \sum_y \frac{y^T y \exp(y^T W y)}{\sum_{y'} \exp(y'^T W y')} \\
 &= -\frac{1}{|P|} \sum_{v \in P} \sum_h y^T y \mathbb{P}(v, h) + \sum_y y^T y \mathbb{P}(y) \\
 &= -\frac{1}{|P|} \sum_{v \in P} (\mathbb{E}_{h|v} [y y^T] - \mathbb{E}_{y'} [y' y'^T])
 \end{aligned}$$

Question 5

(Gaussian RBM)

Energy:

$$\mathcal{E}_{W,b}(v, h) = \frac{1}{2} (v - b)^T (v - b) - v^T W h \quad (1)$$

1.

$$\begin{aligned}
 \mathbb{P}(v|h) &= \frac{1}{Z_h} \exp(-\mathcal{E}_{W,b}(v, h)) \\
 Z_h &= \int_v \exp(-\mathcal{E}_{W,b}(v, h)) dv \\
 &= \int_v \exp\left(-\frac{1}{2} (v - b)^T (v - b) + v^T W h\right) dv \\
 &= \int_v \exp\left(-\frac{1}{2} v^T v + (v + b)^T W h\right) dv \\
 &= e^{b^T W h} \int_v \exp\left(-\frac{1}{2} v^T v + v^T W h\right) dv
 \end{aligned}$$

$$\begin{aligned}
&= e^{b^T W h + \frac{(W h)^T W h}{2}} \int_v \exp\left(-\frac{1}{2}(v - W h)^T (v - W h)\right) dv \\
&= e^{(b + \frac{W h}{2})^T W h} \int_v \exp\left(-\frac{1}{2}v^2\right) dv \\
&= e^{(b + \frac{W h}{2})^T W h} (\sqrt{2\pi})^{N_v}
\end{aligned}$$

So,

$$\begin{aligned}
\mathbb{P}(v|h) &= \frac{1}{Z} \exp(-\mathcal{E}_{W,b}(v, h)) = \frac{1}{(\sqrt{2\pi})^{N_v}} \exp\left(-\frac{1}{2}(v - b)^T (v - b) + v^T W h - (b + \frac{W h}{2})^T W h\right) \\
&= \frac{1}{(\sqrt{2\pi})^{N_v}} \exp\left(-\frac{1}{2}(v - b)^T (v - b) + v^T W h - (b^T W h + \frac{(W h)^T W h}{2})\right) \\
&= \frac{\exp\left(-\frac{1}{2}(v - b)^T (v - b) + v^T W h - b^T W h - \frac{(W h)^T W h}{2}\right)}{(\sqrt{2\pi})^{N_v}} \\
&= \frac{\exp\left(-\frac{1}{2}\|v - b - W h\|^2\right)}{(\sqrt{2\pi})^{N_v}}
\end{aligned}$$

2. Similarly to Problem 4.2, we can get

$$\nabla_b L(W, b) = -\frac{1}{|P|} \sum_{v \in P} (\mathbb{E}_{h|v}[\nabla_b \mathcal{E}] - \mathbb{E}[\nabla_b \mathcal{E}])$$

since

$$\nabla_b \mathcal{E} = v - b$$

we know

$$\begin{aligned}
\nabla_b L(W, b) &= -\frac{1}{|P|} \sum_{v \in P} (\mathbb{E}_{h|v}[v - b] - \mathbb{E}[v - b]) \\
&= -\mathbb{E}_{v \in P}[v] + \mathbb{E}_v[v]
\end{aligned}$$

Question 6

1. D and F are not independent generally, since the joint distribution of D and F is

$$\begin{aligned}
\mathbb{P}(D|F) &= \frac{\mathbb{P}(D, F)}{\mathbb{P}(F)} = \frac{\mathbb{P}(D, F)}{\sum_{D'} \mathbb{P}(D, F)} \\
&= \frac{f_{AD}(A, D) f_{AE}(A, E) f_{EF}(E, F)}{\sum_{D'} f_{AD}(A, D') f_{AE}(A, E) f_{EF}(E, F)}
\end{aligned}$$

which generally doesn't equal to $\mathbb{P}(D)$.

2.

$$\mathbb{P}(B, E|A) = \frac{\mathbb{P}(BEA)}{\mathbb{P}(A)}$$

$$\begin{aligned}
&= \frac{\sum_{C'} \mathbb{P}(BEAC')}{\sum_{C'} \mathbb{P}(AC')} \\
&= \frac{\sum_{C'} f_{BC}(B, C') f_{AE}(A, E) f_{AC}(A, C')}{\sum_{C'} f_{AC}(A, C')}
\end{aligned}$$

B and E are conditionally independent given A, because

$$\begin{aligned}
\mathbb{P}(B|A) &= \frac{\mathbb{P}(BA)}{\mathbb{P}(A)} = \frac{\sum_{C'} \mathbb{P}(BAC')}{\sum_{C'} \mathbb{P}(AC')} \\
&= \frac{\sum_{C'} f_{BC}(B, C') f_{AC}(A, C')}{\sum_{C'} f_{AC}(A, C')} \\
\mathbb{P}(E|A) &= \frac{\mathbb{P}(EA)}{\mathbb{P}(A)} = \frac{\sum_{C'} \mathbb{P}(EAC')}{\sum_{C'} \mathbb{P}(AC')} \\
&= \frac{\sum_{C'} f_{AE}(A, E) f_{AC}(A, C')}{\sum_{C'} f_{AC}(A, C')}
\end{aligned}$$

thus,

$$\begin{aligned}
&\mathbb{P}(B|A)\mathbb{P}(E|A) \\
&= \frac{\sum_{C'} f_{BC}(B, C') f_{AC}(A, C')}{\sum_{C'} f_{AC}(A, C')} \frac{\sum_{C'} f_{AE}(A, E) f_{AC}(A, C')}{\sum_{C'} f_{AC}(A, C')} \\
&= \frac{\sum_{C'} f_{BC}(B, C') f_{AC}(A, C') f_{AE}(A, E) f_{AC}(A, C')}{(\sum_{C'} f_{AC}(A, C'))^2} \\
&= \frac{\sum_{C'} f_{BC}(B, C') f_{AE}(A, E) f_{AC}(A, C')}{\sum_{C'} f_{AC}(A, C')} \\
&= \frac{\sum_{C'} f_{BC}(B, C') f_{AE}(A, E) f_{AC}(A, C')}{\sum_{C'} f_{AC}(A, C')} \\
&= \mathbb{P}(B, E|A)
\end{aligned}$$

3. That's because the joint probability

$$\mathbb{P}(A, B, C, D, E, F) = \frac{1}{Z} f_{BC}(B, C) f_{AC}(A, C) f_{AD}(A, D) f_{AE}(A, E) f_{EF}(E, F)$$

So the energy function $\mathcal{E}(A, B, C, D, E, F)$ can be expressed by $\mathcal{E}(A, C)$, $\mathcal{E}(A, D)$, $\mathcal{E}(A, E)$, $\mathcal{E}(B, C)$, $\mathcal{E}(E, F)$ (just let the energy function be the minus log of the joint probability)