00137130/00101755: Deep Learning: Algorithms and Applications Homework 2 Due: April 16, 2020

Note: Unless otherwise noted, section and equation numbers refer to those in the book by Goodfellow, Bengio, and Courville.

- 1. Represent the convolution example in Figure 9.1 (3×4 input, 2×2 kernel, "valid" convolution) as matrix multiplication with a doubly block circulant matrix.
- 2. Consider the pooling example in Figure 9.9. Design a set of filters such that the max pooling unit can learn to be invariant to (a) rotation, and (b) scaling.
- 3. The Hopfield network is a type of recurrent network consisting of n units with states s_i and update rule

$$s_i \leftarrow \sigma \left(\sum_{j \neq i} w_{ij} s_j - \theta_i \right),$$

where $\sigma(x) = 2I(x \ge 0) - 1$, $w_{ij} = w_{ji}$, and $w_{ii} = 0$. The network is updated in an asynchronous manner, so that one unit is randomly selected and updated at each time step. Prove that the network will eventually reach a stable state at a local minimum of the energy function

$$E(s) = -\frac{1}{2} \sum_{i=1}^{n} \sum_{j=1}^{n} w_{ij} s_i s_j + \sum_{i=1}^{n} \theta_i s_i.$$

4. Design a recurrent neural network to approximate the dynamics of the Lorenz 96 model

$$\frac{dx_i}{dt} = (x_{i+1} - x_{i-2})x_{i-1} - x_i + F, \quad i = 1, \dots, n,$$

where F is a forcing constant and the indices are cyclic so that $x_{-1} = x_{n-1}$, $x_0 = x_n$, and $x_{n+1} = x_1$.

- 5. Prove that an undercomplete autoencoder with linear decoder and MSE loss learns to span the principal subspace of the training data.
- 6. Consider the Boltzmann machine with state vector $x \in \{0, 1\}^d$ and energy function

$$E(\mathbf{x}) = -\mathbf{x}^T \mathbf{U}\mathbf{x} - \mathbf{b}^T \mathbf{x}.$$

- (a) Derive the conditional distributions $p(x_i | x_{-i})$.
- (b) Do the conditional distributions in part (a) uniquely determine the joint distribution of x? Prove or disprove.
- 7. Derive the expression (19.56) for the mean field approximation.
- 8. Describe an algorithm for training a generative adversarial network and comment on its convergence properties.