

Introduction to Machine Learning

Homework 8: Convolutional Neural Networks

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1. Let X and W be arrays,

$$X = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 3 & 3 & 0 \\ 0 & 3 & 3 & 3 & 0 \\ 0 & 3 & 2 & 3 & 0 \\ 0 & 3 & 2 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}, \quad W = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}.$$

Let Z be the 2D convolution (without reversal):

$$Z[i, j] = \sum_{k_1, k_2} W[k_1, k_2] X[i + k_1, j + k_2]. \quad (1)$$

Assume that the arrays are indexed starting at $(0, 0)$.

- (a) What are the limits of the summations over k_1 and k_2 in (1)?
 - (b) What is the size of the output $Z[i, j]$ if the convolution is computed only on the *valid* pixels (i.e. the pixel locations (i, j) where the summation in (1) does not exceed the boundaries of W or X).
 - (c) What is the largest positive value of $Z[i, j]$ and state one pixel location (i, j) where that value occurs.
 - (d) What is the largest negative value of $Z[i, j]$ and state one pixel location (i, j) where that value occurs.
 - (e) Find one pixel location where $Z[i, j] = 0$.
2. Suppose that a convolutional layer of a neural network has an input tensor $X[i, j, k]$ and computes an output via a convolution and ReLU activation,

$$Z[i, j, m] = \sum_{k_1} \sum_{k_2} \sum_n W[k_1, k_2, n, m] X[i + k_1, j + k_2, n] + b[m],$$
$$U[i, j, m] = \max\{0, Z[i, j, m]\}.$$

for some weight kernel $W[k_1, k_2, n, m]$ and bias $b[m]$. Suppose that X has shape $(48, 64, 10)$ and W has shape $(3, 3, 20, 10)$. Assume the convolution is computed on the *valid* pixels.

- (a) What are the shapes of Z and U ?

- (b) What are the number of input channels and output channels?
 - (c) How many multiplications must be performed to compute the convolution in that layer?
 - (d) If W and b are to be learned, what are the total number of trainable parameters in the layer?
3. Suppose that a convolutional layer is a linear convolution followed by a sigmoid activation,

$$Z[i, j, m] = \sum_{k_1} \sum_{k_2} \sum_n W[k_1, k_2, n, m] X[i + k_1, j + k_2, n] + b[m],$$

$$U[i, j, m] = 1 / (1 + \exp(-Z[i, j, m])).$$

Suppose that during back-propagation, we have computed the gradient $\partial J / \partial U$ for some loss function J . That is, we have computed $\partial J / \partial U[i, j, m]$. Show how to compute the following:

- (a) The gradient components $\partial J / \partial Z[i, j, m]$.
- (b) The gradient components $\partial J / \partial W[k_1, k_2, n, m]$.
- (c) The gradient components $\partial J / \partial X[i, j, n]$.