

# Introduction to Machine Learning

## Homework 8: Convolutional Neural Networks

Prof. Sundeep Rangan

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  $(3, 3, 10, 20)$ . 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  $\partial J / \partial X[i, j, n]$ .