

# Deep learning HW2

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## Q3:

### 3.a:

Starting off with the definitions:

$$T(Im) : T(Im(i, j)) = Im(i - l, j - k)$$
$$f(i, j) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} h(m, n) * Im(i - m, j - n)$$

Now, we shall prove that  $T(f(Im)) = f(T(Im))$ :

$$f(T(Im)) = f(Im(i - k, j - l) \forall i, j) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} h(m, n) * Im(i - l - m, j - k - n) =$$

$$\text{We substitute } m \text{ and } n \text{ by } m' = m + l \text{ and } n' = n + k:$$
$$= \sum_{m'=-\infty}^{\infty} \sum_{n'=-\infty}^{\infty} h(m' - l, n' - k) * Im(i - m', j - n')$$

On the other hand, since  $f$  is defined by  $h$ , we can show:

$$T(f(Im)) = T(\sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} h(m, n) * Im(i - m, j - n)) =$$
$$= \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} T(h(m, n)) * Im(i - m, j - n) =$$
$$\sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} h(m - l, n - k) * Im(i - m, j - n)$$

To sum, because the only difference is the names of the parameters, we receive:

$$T(f(Im)) = f(T(Im))$$

**3.b:**

**3.c:**

let us look at the following matrix, A:

$$A = \begin{matrix} & 1 & 2 & 3 & 1 \\ 15 & 4 & 3 & 2 \\ 3 & 6 & 3 & 3 \\ 0 & 0 & 0 & 0 \end{matrix}$$

while MAX-POOLING 3x3 will give us an answer of 15 for the top square, a one pixel shift would make the answer 6 instead. Thus, we have given a counter-example for MAX-POOLING 3x3 being invariant to one pixel shift.

**3.d:**

It is easy to notice that if all weights are equal, a fully connected layer will not change for any ordering of  $x_1, x_2, \dots, x_n$ . therefore, equal weights for all inputs is a sufficient condition for the layer to be permutation invariant.