

## DEEP LEARNING IN COMPUTER VISION - EXERCISE 4

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### Exercise 4.1a

Possible values for  $x_1, \dots, x_8$  are

$$\begin{aligned}x_1 &= 17, & x_2 &= 13, \\x_3 &= 3, & x_4 &= 4, \\x_5 &= 13, & x_6 &= 13, \\x_7 &= 2, & x_8 &= 3.\end{aligned}$$

### Exercise 4.1c

The `tf.nn.conv2d_transpose()` requires to define the output shape since it is not necessarily unique. For example, let the input shape for a convolution be

- (a)  $4 \times 4$ ,
- (b)  $3 \times 3$ .

In both cases we use a filter of size  $3 \times 3$ , striding of 2 and the padding option SAME. Then, in both cases (a) and (b), performing a `conv2d()` yields an output of  $2 \times 2$ . Hence, if we want to go back using `tf.nn.conv2d_transpose()`, we have to tell tensorflow if we came from the  $4 \times 4$  input or the  $3 \times 3$  input.

The `tf.layers.conv2d_transpose()` uses the  $4 \times 4$  output by default. See also our file `ex_04_1.c.ipynb`, where we implemented this example.

### Exercise 4.3

Let  $g : \mathbb{R}^n \rightarrow \mathbb{R}$  be a differentiable function,  $A \in \mathbb{R}^{m \times n}$ . We consider the composite function  $E : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $x \mapsto g(Ax)$ .

- (a) The chainrule yields

$$\nabla E(x) = (E'(x))^T = (A^T g'(x))^T = \nabla g(x) A.$$

- (b) Since the action of a convolutional layer is linear, it can be viewed as a mapping  $x \mapsto Mx$  with a matrix  $M$ . At any point is the derivative of this mapping (which is necessary for the backpropagation algorithm) given by the transpose matrix  $M^T$ . This matrix  $M^T$  describes a transpose convolution.