Time: 30 minutes

Max marks: 10

#### **Instructions:**

- Do not plagiarize. Do not assist your classmates in plagiarism.
- Show your full solution for the questions to get full credit.
- Attempt all questions that you can.
- In the unlikely case a question is not clear, discuss it with an invigilating TA. Please ensure that you clearly include any assumptions you make, even after clarification from the invigilator.
- 1. Fig. 1 shows the convolution operation on an input array using a single kernel (filter) to generate the output feature map. Answer the following questions.

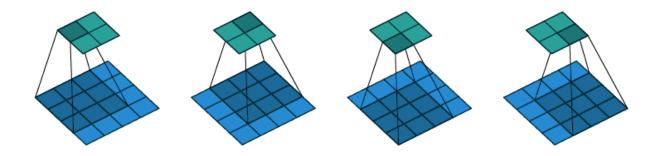


Figure 1: Steps of the convolution operation. The  $4 \times 4$  patch is the input. The dark-shaded  $3 \times 3$  patch overlaid on the input is the convolution kernel (filter). The  $2 \times 2$  grid is the output feature map.

- (a)  $(\frac{1}{2} \text{ point})$  What are the values of stride and padding used in Fig. 1?
- (b) (½ point) What is the total number of learnable parameters in this case?
- (c) ( $\frac{1}{2}$  point) If the input was an RGB image of  $6 \times 6$ , and we used a  $3 \times 3$  kernel with a stride of 1 with no padding, what would be the spatial dimensions (height  $\times$  width) of the output feature map?
- (d) ( $\frac{1}{2}$  point) The feature map from part (c) above is passed through a max-pooling layer with a 2 × 2 filter and a stride of 2 and no padding (Usually pooling layers don't use padding). What will be the spatial dimensions of the feature map at the output of this pooling layer?
- (e) (1 point) Say you have the convolution layer with 8 such filters (as in part (c)) to process the  $6 \times 6$  RGB image. What would be the total number of *learnable* parameters for this layer?
- (f) (2 points) For Fig. 1, let the weights of the  $3 \times 3$  kernel be  $w_{i,j}$ ,  $i,j \in \{0,1,2\}$ . Write the convolution operation as a matrix multiplication when the input is the  $4 \times 4$  single-channel image and the output is the  $2 \times 2$  feature map. You may ignore the bias term for this part, however, for full credit, describe how would the input / output need to be processed in order to implement the convolution operation as a matrix multiplication, e.g., processing may need operations like reshape etc.

Total for Question 1: 5

# **Solution:**

(a)  $(\frac{1}{2} \text{ point})$  stride = 1, padding = 0.

Rubric: binary.

(b)  $(\frac{1}{2} \text{ point})$  9 for the kernel (filter) weights and 1 for the bias = 10.

Rubric: binary.

(c)  $(\frac{1}{2} \text{ point})$  The output feature map size O is given by the formula

$$O = \left| \frac{(I - K + 2P)}{S} \right| + 1$$

where I is the input dimension (calculated separately if height and width are not the same, similarly if kernel height and width are not the same; usually the kernel dimensions are the same, the input not often), K is the kernel dimension, P is the amount of padding (in terms of rows / columns, with the multiplication of 2 accounting for top & bottom rows / left & right columns), and S is the stride and  $\lfloor x \rfloor$  indicates the *floor* function (largest integer smaller than x). Using this formula, the output dimension of the feature map will be  $4 \times 4$ .

Rubric: binary.

(d) ( $\frac{1}{2}$  point) We apply the 2 × 2 max-pooling filter to the 4 × 4 feature map, which gives us a 2 × 2 output feature map.

Rubric: binary.

(e) (1 point) Since it is an RGB image, there are 3 input channels, and we have a  $3 \times 3$  kernel. Therefore we have  $3 \times 3 \times 3 = 27$  weights for each kernel and 1 bias term. For eight filters, the total number of learnable parameters is  $28 \times 8 = 224$ .

Rubric:

- +1 if calculation fully correct
- +0.5 if bias not considered
- (f) (2 points) Let **I** be the  $4 \times 4$  input image shown in Fig. 1. We *vectorize* this image row-wise to obtain the 16-dimensional column vector  $\mathbf{x} = [I_{0,0}, I_{0,1}, \dots, I_{0,3}, I_{1,0}, \dots, I_{3,3}]^{\top}$ . We then write the convolution operation as a matrix multiplication as shown below.

$$C_{4 \times 16} =$$

Then the 4-dimensional vectorized output  $\mathbf{y} = \mathbf{C}\mathbf{x}$  and is then reshaped to obtain a  $2 \times 2$  matrix that is the feature map.

# Rubric:

- +1 point for correct matrix
- +0.5 points for mentioning vectorizing input
- +0.5 points for mentioning reshaping output vector
- 2. (a) (1 point) Is  $\mathbf{R} = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$  a valid rotation matrix? Explain why or why not?
  - (b) (1 point) If **R** above is not a valid rotation matrix, convert it into one by only changing the sign of the non-zero entries. Find the axis and angle of rotation, for the corrected rotation matrix, or for the original if it already was a valid rotation matrix. {*Hint*: This should be possible just by inspection, even if you don't recall the formula.}

Total for Question 2: 2

#### Solution:

(a) (1 point) Yes, **R** is a valid rotation matrix, as it is orthogonal and has a determinant of 1. (Edit) Comparing it to the standard form of the rotation matrix is also valid.

#### Rubric:

0.5 points for correctness; 0.5 points for justification

(b) (1 point) The axis of rotation is the third axis, or Z, i.e.,  $[0,0,1]^{\top}$ , and the angle of rotation is  $-90^{\circ}$  or equivalently  $270^{\circ}$ . One could also alternatively claim that the axis of rotation is the negative Z axis, i.e.,  $[0,0,-1]^{\top}$  and the angle of rotation is  $90^{\circ}$ .

# Rubric:

0.5 points for correct axis; 0.5 points for correct angle

- 3. (a) (1 point) Write the most general form of the intrinsic camera parameter matrix  $\mathbf{K}$  and identify the name of each parameter.
  - (b) (1 point) Given two vectors  $\mathbf{n} = [n_1, n_2, n_3]^{\top}$  and  $\mathbf{x} = [x_1, x_2, x_3]^{\top}$ , give the expression for the cross product  $\mathbf{n} \times \mathbf{x}$ . Write the form of the  $3 \times 3$  cross-product matrix denoted by  $\mathbf{N}$  (also denote by  $[\mathbf{n}]_{\times}$ ), such that  $\mathbf{N}\mathbf{x} = \mathbf{n} \times \mathbf{x}$ . {*Hint*: You may use of the determinant-based approach for computing cross-products of 3D vectors.}
  - (c) (1 point) Show that  $\mathbf{N}\mathbf{n} = (\mathbf{N}\mathbf{x})^{\top}\mathbf{n} = 0$  for any arbitrary non-zero  $\mathbf{x}$  and  $\mathbf{n}$ .

Total for Question 3: 3

### **Solution:**

(a) (1 point)

$$\mathbf{K} = \begin{bmatrix} f_x & s & p_x \\ 0 & f_y & p_y \\ 0 & 0 & 1 \end{bmatrix}$$

#### where

- $f_x$  and  $f_y$  are the **focal length** parameters (measured in #pixels along the x and y axes respectively; recall that a pixel may not be square and can have different horizontal and vertical side lengths.)
- $p_x$  and  $p_y$  are the pixel coordinates of the **principal point**, which is the projection of the center of projection on the image plane.
- s is the **skew parameter** that captures the correlation in the pixels horizontal and vertical axes (non-zero for non-rectangular pixels).

### **Rubric:**

+0.5 points for correct matrix; 0 if skew not present

+0.5 points for naming the parameters

(b) (1 point) We can compute  $\mathbf{n} \times \mathbf{x}$  as:

$$\mathbf{n} \times \mathbf{x} = \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ n_1 & n_2 & n_3 \\ x_1 & x_2 & x_3 \end{vmatrix}$$

$$= \begin{bmatrix} n_2 x_3 - n_3 x_2 \\ n_3 x_1 - n_1 x_3 \\ n_1 x_2 - n_2 x_1 \end{bmatrix}$$

$$[\mathbf{n}]_{\times} = \begin{bmatrix} 0 & -n_3 & n_2 \\ n_3 & 0 & -n_1 \\ -n_2 & n_1 & 0 \end{bmatrix}$$
(1)

**Rubric:** 

+1 for correct matrix

(c) (1 point) We can see that for any non-zero **n**, we necessarily have

$$[\mathbf{n}]_{\times} \mathbf{n} = \mathbf{N} \mathbf{n} = \begin{bmatrix} 0 - n_3 n_2 + n_2 n_3 \\ n_3 n_1 + 0 - n_1 n_3 \\ -n_2 n_1 + n_1 n_2 + 0 \end{bmatrix}$$
$$= [\mathbf{n}]_{\times}^{\top} \mathbf{n} = \mathbf{N}^{\top} \mathbf{n} = \begin{bmatrix} 0 + n_3 n_2 - n_2 n_3 \\ -n_3 n_1 + 0 + n_1 n_3 \\ n_2 n_1 - n_1 n_2 + 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$
 (2)

$$([\mathbf{n}]_{\times}\mathbf{x})^{\top}\mathbf{n} = (\mathbf{N}\mathbf{x})^{\top}\mathbf{n} = \mathbf{x}^{\top}[\mathbf{n}]_{\times}^{\top}\mathbf{n} = \mathbf{x}^{\top}\mathbf{N}^{\top}\mathbf{n} = \mathbf{x}^{\top}\begin{bmatrix}0\\0\\0\end{bmatrix} = 0$$

# Rubric:

+0.5 points for showing Nn=0

+0.5 points for showing  $(Nx)^T n = 0$