

# Chapter 1

## Introduction

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### 1.1 Notation

#### Images:

An  $M \times N$  image in 256 grey scale (or grey levels) is a matrix

$$A = \begin{bmatrix} a_{00} & a_{01} & \dots & a_{0N} \\ a_{10} & a_{11} & & \\ \vdots & & & \ddots \\ a_{M0} & a_{M1} & \dots & a_{MN} \end{bmatrix}$$

also expressed as  $\{a_{jk}\}$ ,  $j = 0, 1, \dots, (M - 1)$ ,  $k = 0, 1, \dots, (N - 1)$ , and the elements are integers,  $a_{jk} \in \{0, 1, 2, \dots, 255\}$ . Images are also viewed as functions of two variables,  $f(x, y)$ , where the coordinate pair  $(x, y)$  denotes some point on the 2D plane and  $f(x, y)$  is the intensity of the image at that point.

#### Display:

Images are displayed as a rectangular array of grey squares. Each grey square is called a “pixel”, and the greyness depends on the value of  $a_{jk}$  in the matrix. For  $a_{jk} = 0$ , the pixel is black, for  $a_{jk} = 255$ , the pixel is white and for values in between various shades of grey are used.

When viewed as a function of two variables, images may be represented as a surface. However this is a mathematical convenience usually used for image filters. It normally does not make sense to display an image of a scene as a surface.

#### Types of images:

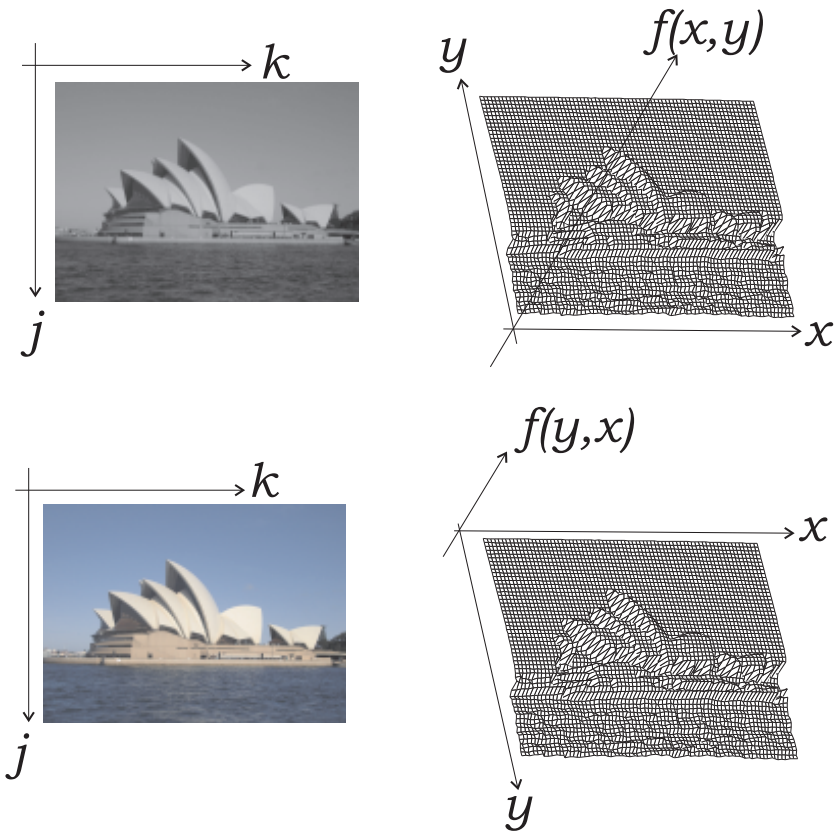
The three types of images that will be used often in this presentation, are

- Binary images, where  $a_{jk} \in \{0, 1\}$ .
- Grey scale images, where  $a_{jk} \in \{0, 1, 2, \dots, 255\}$ .
- Normalized grey scale images, where  $a_{jk} \in [0, 1]$ .
- Colour images, where each pixel is displayed in colour using three values  $(r_{jk}, g_{jk}, b_{jk})$ , the so-called RED, GREEN and BLUE colour channels. These triplets are usually packed into three different matrices,

$$\begin{aligned} R &= \{r_{jk}\}, j = 0, 1, \dots, N, \quad k = 0, 1, \dots, M, \\ G &= \{g_{jk}\}, j = 0, 1, \dots, N, \quad k = 0, 1, \dots, M, \\ B &= \{b_{jk}\}, j = 0, 1, \dots, N, \quad k = 0, 1, \dots, M, \end{aligned}$$

These matrices are referred to as the RED, GREEN and BLUE “colour planes” respectively.

The pixels in a binary image are often said to be “on” when the  $a_{jk} = 1$  and “off” when  $a_{jk} = 0$ .



**Figure 1.** TOP LEFT: A grey scale image, BOTTOM LEFT: A colour image, TOP RIGHT: Viewed as a surface plot with conventional direction for axes. BOTTOM RIGHT: A surface plot but with axes directions more convenient for images.

Pixels:

Pixels are in effect positions on the plane. The convention to denote a single pixel position by a single symbol will often be employed, for example

$$\mathbf{p} = (j, k), \quad \mathbf{q} = (j + 1, k),$$

where  $\mathbf{q}$  is the pixel “below”  $\mathbf{p}$ .

Directions in the 2D-plane:

Images are normally displayed the way matrices are written, first index  $j$  running vertically down, then the second index  $k$  running horizontally across.

When viewed as a function one normally lets  $x$  run horizontally to the right, and lets  $y$  run vertically up. This notion clashes with the ordering of the matrix indices. When working with images, it is perhaps more convenient to consider the function  $f(y, x)$ , where  $y$  is associated with  $j$  and  $x$  is associated with  $k$ , and in addition one inverts the  $y$  axis so that its positive part runs downwards (but we keep the positive part of the  $x$ -axis running to the right).

## 1.2 Pixel Neighbourhoods

Processing of images often consists of performing operations on a pixel  $\mathbf{p} = (j, k)$ , depending on the values of its neighbours. Some often used neighbourhoods are as follows:

4-adjacency:

The *4-neighbourhood* of pixel  $\mathbf{p} = (j, k)$  is a set of pixels

$$N_4(\mathbf{p}) = \{(j + 1, k), (j - 1, k), (j, k + 1), (j, k - 1)\}.$$

If  $\mathbf{q} \in N_4(\mathbf{p})$  then it is said that  $\mathbf{q}$  is *4-adjacent to*  $\mathbf{p}$ . Note the commutativity of this relation: if  $\mathbf{q} \in N_4(\mathbf{p})$  then  $\mathbf{p} \in N_4(\mathbf{q})$  and the reverse is also true.

D-diagonal-adjacency:

The *diagonal-neighbourhood* of pixel  $\mathbf{p} = (j, k)$  is a set of pixels

$$N_D(\mathbf{p}) = \{(j + 1, k + 1), (j - 1, k + 1), (j - 1, k - 1), (j + 1, k - 1)\}.$$

If  $\mathbf{q} \in N_D(\mathbf{p})$  then it is said that  $\mathbf{q}$  is *D-adjacent to*  $\mathbf{p}$ . D-adjacency is also commutative.

8-adjacency:

The *8-neighbourhood* of pixel  $\mathbf{p} = (j, k)$  is a set of pixels

$$N_8(\mathbf{p}) = N_4(\mathbf{p}) \cup N_D(\mathbf{p}).$$

If  $\mathbf{q} \in N_8(\mathbf{p})$  then it is said that  $\mathbf{q}$  is *8-adjacent to*  $\mathbf{p}$ . 8-adjacency is also commutative.

Other kinds of adjacencies such as mixed adjacency or *m-adjacency* for short, is somewhat more complicated, where the membership of a pixel in the neighbourhood of  $\mathbf{p}$  does not only depend on its position relative to  $\mathbf{p}$ , but also on some values of other neighbours of  $\mathbf{p}$ . Consider for a start binary images only. This will be discussed later.

### 1.3 Processing of an image

In all image transformations we shall refer to the *input image* (i.e the image one starts with) and the *output image* (i.e. your final result). We shall normally use  $A$  as input image and  $B$  as output image. An image transformation may then be expressed as

$$B = f(A)$$

where  $f(\cdot)$  constitutes some operations performed on the pixels of  $A$  to render the pixels of  $B$ . As used in this sentence, we shall often take the liberty to say ‘pixel’ when we actually mean ‘pixel value’.