

Topology of Digital Images

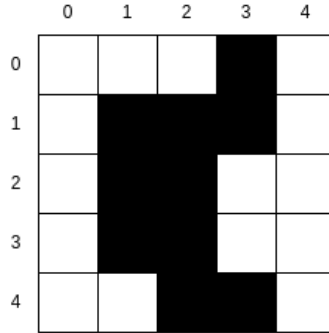
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Abstract

In this note I will give a crash course on the topology of digital images. I will define the concepts of local neighborhood, and the topology that it generates. I will discuss the open and closed sets in a digital image, as well as the closure (dilation) and the interior (erosion) operator.

The backbone of a digital image, i.e. its canvas D , can be thought of as a subset of \mathbb{Z}^2 , and any particular digital image is simply a function from this domain to $\{0, \dots, 255\}^3$ where we have assumed that the image is a three channel 8-bit RGB image. However for our purpose here, we will follow a different point of view. We fix a particular binary image (hence a particular function $f : D \rightarrow \{0, 1\}$), and then each binary image can be thought of as a set of black and white (or alternatively 0,1) pixels. Using our first point of view, we are in fact working with the product space $D \times \{0, 1\}$, and the image will be the graph of the function f . For instance, consider the following digital image.



With our point of view, this image can be thought of as the set

$$I = \{(0, 0, 0), (0, 1, 0), (0, 2, 0), (0, 3, 1), (0, 4, 0), (1, 0, 0), (1, 1, 1), (1, 2, 1), \dots, (4, 3, 1), (4, 4, 0)\}.$$

Since the white pixels and the black pixels carry the dual information (knowing the set of white pixels we can determine the black pixels and vice versa), it is more convenient to assume the white pixels to be the “background” and the black pixels to be the “foreground” and express the image I by only specifying the foreground pixels. So with this point of view we have

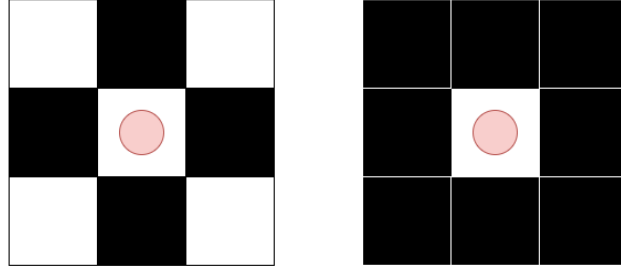
$$I = \{(0, 3), (1, 1), (1, 2), (1, 3), (2, 1), \dots, (4, 2), (4, 3)\}, \quad (1)$$

where we have also dropped the third component because it is a redundant information (as we know from the context that we are representing the foreground pixels).

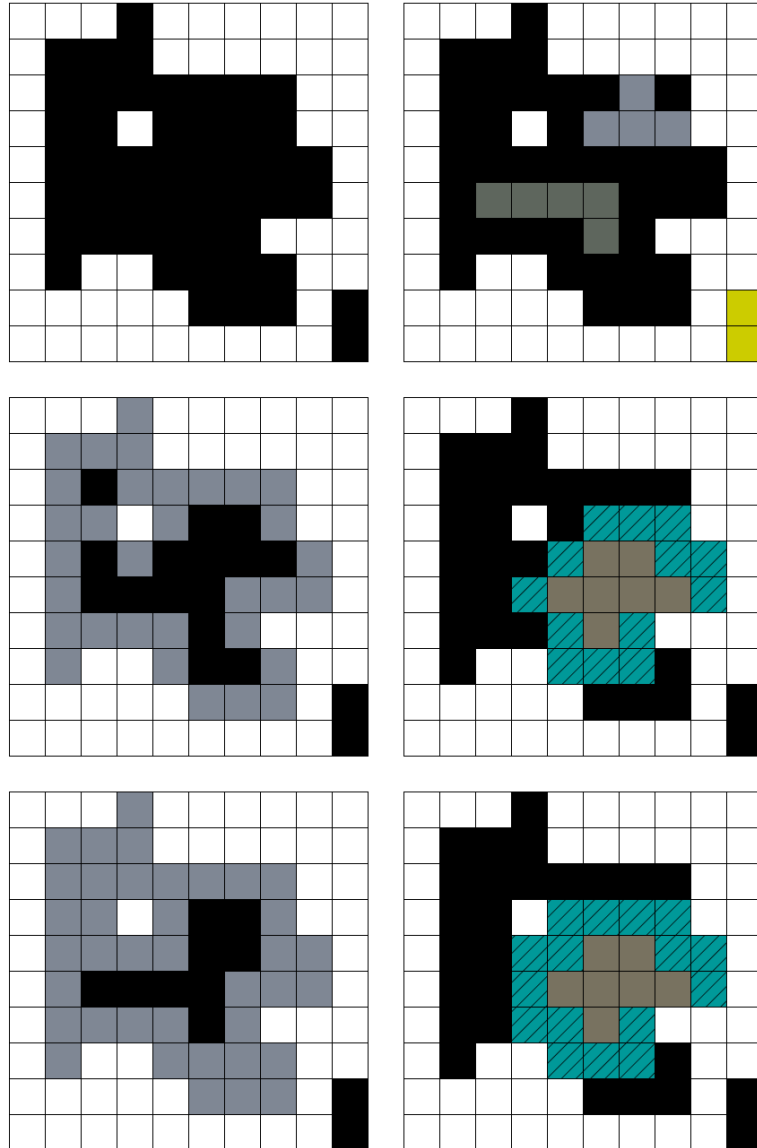
We can now equip this set with a topology, and there are multiple ways to do so. One simple way is by defining a metric on the space. Two metrics are very popular: taxicab metric (d_1), and max-metric (d_∞) given as below.

$$d_1((n_1, n_2), (m_1, m_2)) = |m_1 - n_1| + |m_2 - n_2|, \quad d_\infty((n_1, n_2), (m_1, m_2)) = \max\{|m_1 - n_1|, |m_2 - n_2|\}.$$

These metrics define the neighborhood pixels of a particular pixel, using which we can define a topology on the set. For instance, the following figure demonstrates the local neighborhood of a pixel marked with red dot with the taxicab metric (left figure) and the max-metric (right figure). We can also say that these are the unit balls of radius 1 with the corresponding metric used. These are also sometimes called 4-neighborhood, or 8-neighborhood cells.



These local neighborhoods is a neighborhood base for a topology, and we can determine this topology by determining the open sets, which are the sets where every point in the set has a local neighborhood that is contained in the set. Once we have the notion of topology, then we can talk about the boundary, interior, and the closure of a set. These concepts are depicted in the figure below.

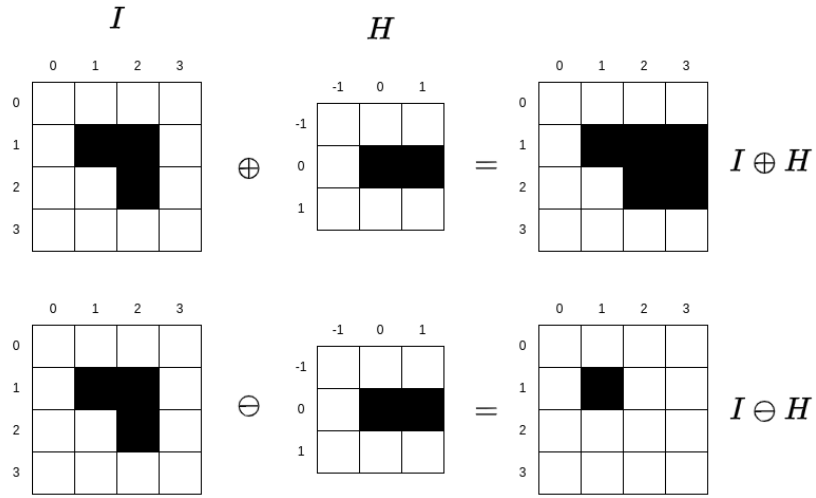


In the figure above, the subplot (Left, Up) corner represents a binary digital image where the black pixels are the foreground and the white pixels are the background. We can express this space as a set similar to (1). In the (Right, Up) subplot, three subsets are shown: gray, green and yellow. The green set is an open set, because for every point in it, there is a local neighborhood (both 4-neighborhood and 8-neighborhood) that is contained in the digital image (black region). This means that this set is open in both topologies induced by d_1 as well as d_∞ . The gray set is neither open,

nor closed. And the yellow set is closed, because it contains the neighborhoods (both 4-neighborhood, and 8-neighborhood) of all of its elements. So the yellow set is open in both topologies induced by d_1 and d_∞ . In the middle row, the left subplot shows the boundary of the black region in the topology induced by the 4-neighborhood. In a topological space a point is a boundary point every neighborhood if that point has non-empty intersection with the image and its complement complement. The plot in the right, shows a gray set, that when considered along with the dashed regions, is the closure of the gray set. The plots on the third row are the same as the ones on the second row, with the only difference that we have used the topology induced by the 8-neighborhood.

Note that to construct a topological space there is not need for any metric function (although a metric always induces a topology). This means that we can define neighborhoods as arbitrary as we want, without worrying about the metric functions that can generate those neighborhoods.

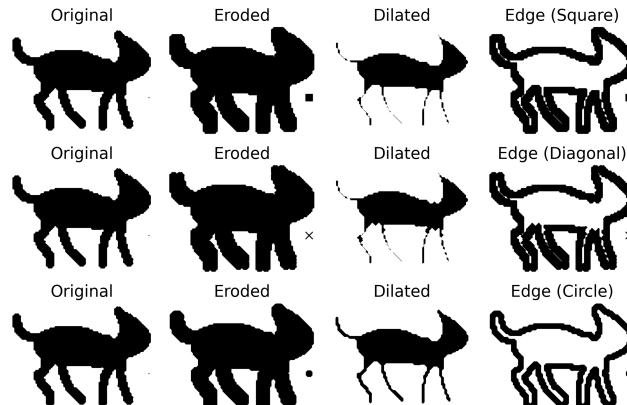
Erosion and dilation of a binary image are intuitively peeling off/adding layers from/to the boundary of regions. For any erosion and dilation operator an structural element is needed, that is often denoted as H . The following figure demonstrates erosion and dilation operations with a particular structure element.



Erosion acts like the \cdot° operator (interior) in topology literature in analysis, where as Dilation acts like the $\bar{\cdot}$ operator (closure) in the topology that its local neighborhood basis is successive dilation of the structure element with itself. With this point of view, the duality between erosion and dilation, as expressed in (9.17) in [THE TEXT BOOK], is nothing but a well know fact in analysis

$$(A^\circ)^c = (\bar{A})^\circ$$

where the structural element H should be a hermitian (symmetric in this case) matrix. The following figure shows some common structuring elements and the corresponding edge in that topological space.



Opening and Closing

Opening and closing are two morphological operations that are achieved by successive application of dilation and erosion. Fixing some structuring element H , and an image I , the opening is often written as $I \circ H$ and the closing operator is denoted by $I \bullet H$ that is defined as:

$$I \circ H = (I \ominus H) \oplus H, \quad I \bullet H = (I \oplus H) \ominus H$$

In opening, we first do an erosion and then perform a dilation. One might think that the dilation will undo the erosion and the total effect is the same as identity operator that does nothing. This is almost true, except for the islands in the image that are smaller than the structuring element (in the topological terminology, isolated sets that has no interior). These sets will vanish once the first erosion is performed, and in the followup dilation, no pixels will grow out of these regions. So the opening can be used to eliminate the scattered noisy pixels.

Closing does the opposite job compared to opening. It is easy to argue that by applying the closing operator, the wholes in the foreground that are smaller than the structuring element (or in the terminology of the topology, those whose all of whose pixels are the limit points of the set that contains them) will close after closing. Furthermore, those disjoint sets that share common limit points will merge after the closing operation.

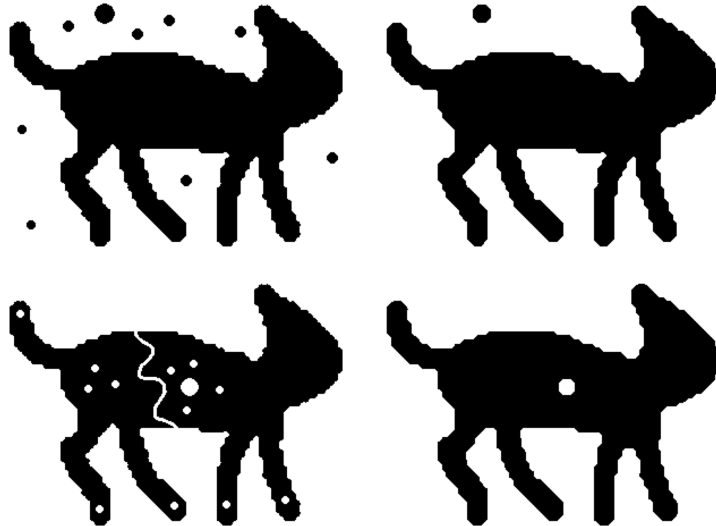


Figure 1: Caption

Connectivity detection

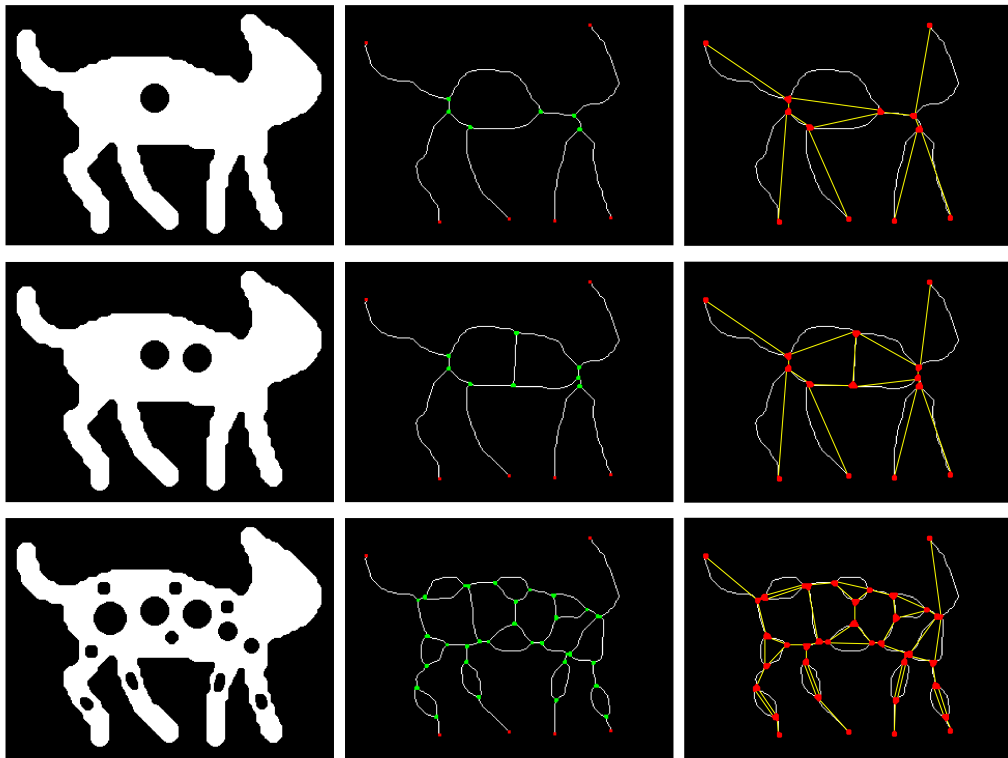


Figure 2: Caption