

Lecture 6

Three Types of Adjacency (cont'd)

4-adjacency

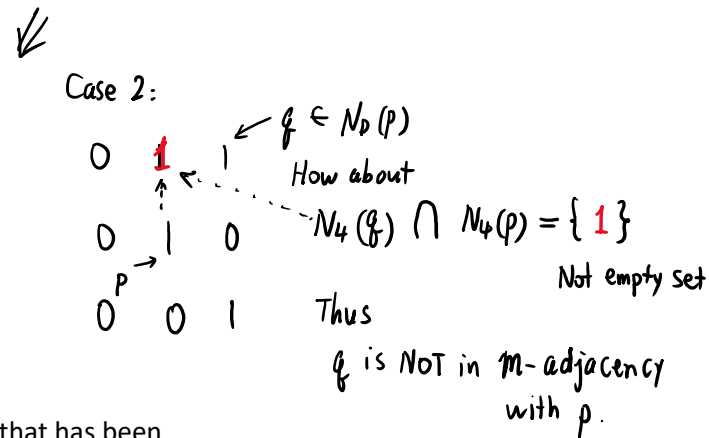
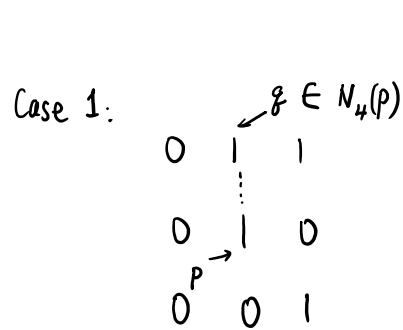
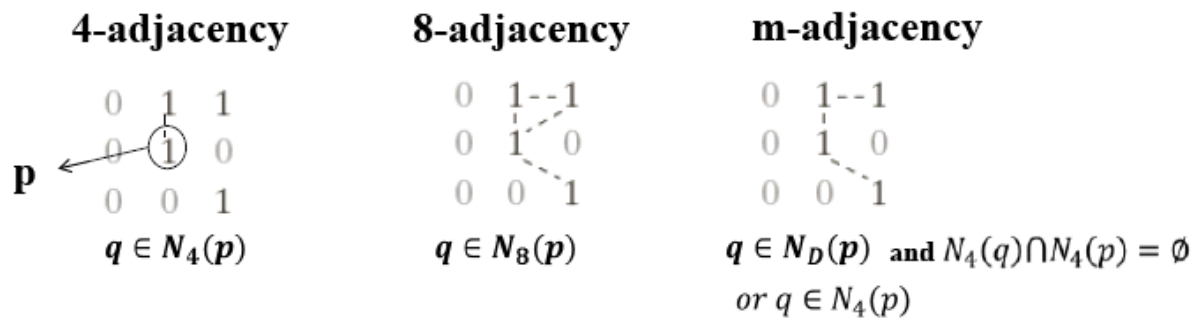
–Two pixels p and q with values from V are 4-adjacent if q is in the set $N_4(p)$.

•8-adjacency

–Two pixels p and q with values from V are 8-adjacent if q is in the set $N_8(p)$.

•m-adjacency (mixed adjacency).

Let V be the set of intensity values used to define adjacency.



m-adjacency has eliminated the multiple path connection that has been generated by the 8-adjacency.

We can specify 4-, 8- or m-paths depending on the type of adjacency specified.

```

0 1 1
0 1 0
0 0 1

```

```

0 1--1
0 1  0
0 0  1

```

```

0 1--1
0 1  0
0 0  1

```

⇓
Two alternative 8-paths

⇓
a unique m-path

Region and Boundary

- If R happens to be an entire image, then its boundary is defined as the set of pixels in the first and last rows and columns in the image.

- This extra definition is required

Digital Path

- A digital path (or curve) from pixel p with coordinate (x,y) to pixel q with coordinate (s,t) is a sequence of distinct pixels with coordinates $(x_0,y_0), (x_1,y_1), \dots, (x_n,y_n)$ where $(x_0,y_0) = (x,y)$ and $(x_n,y_n) = (s,t)$ and pixels (x_i,y_i) and (x_{i-1},y_{i-1}) are adjacent for $1 \leq i \leq n$.

- n is the length of the path.

$$D_4(p,q) = |x - s| + |y - t|$$

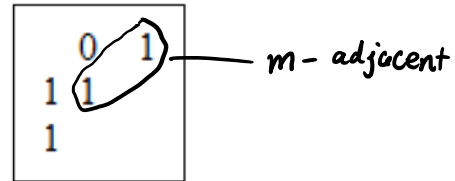
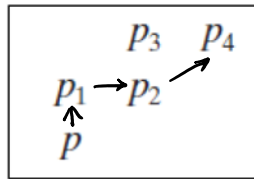
$$D_8(p,q) = \max(|x - s|, |y - t|)$$

D_m distance is defined as the shortest m-path between the points.

- **Case2:** If $p_1 = 1$ and $p_3 = 0$

now, p_2 and p will no longer be adjacent (see m -adjacency definition)

then, the length of the shortest path will be 3 (p, p_1, p_2, p_4)



Mathematical Tools

De-Noising

Assumption: Noise is uncorrelated to image and has zero mean.

Random Variable

$$x_1, x_2, \dots, x_n$$

Sample Mean:
$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

Another random variable if x_1, x_2, \dots, x_n are values taken by underlying random variables X_1, X_2, \dots, X_n

having the same distribution with the same mean \bar{x}

Define Sample Mean
$$\hat{\bar{x}} = \frac{1}{n} \sum_{i=1}^n X_i$$

```
>> sample_size = 10;
>> run = 100000;
>> Xavg = zeros(1, run);
>> for i = 1: run
    X = randn(1, sample_size);
    Xavg(i) = mean(X);
end
>> mean(Xavg)
ans =
    9.4576e-04

>> histogram(Xavg)
>> doc randn
>> var(Xavg)
ans =
    0.0999
```

