Robotika in računalniško zaznavanje (RRZ)

Regije

Danijel Skočaj Univerza v Ljubljani Fakulteta za računalništvo in informatiko

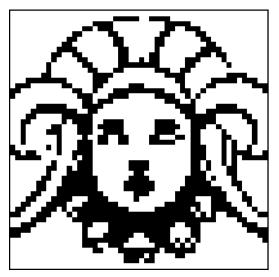
Literatura: W. Burger, M. J. Burge (2008).

Digital Image Processing, poglavja 10, 11

v1.0

Morfološki filtri

- Linearni filtri ne spreminjajo topologije slike
- Medianin filter deloma spremeni strukturo slike:



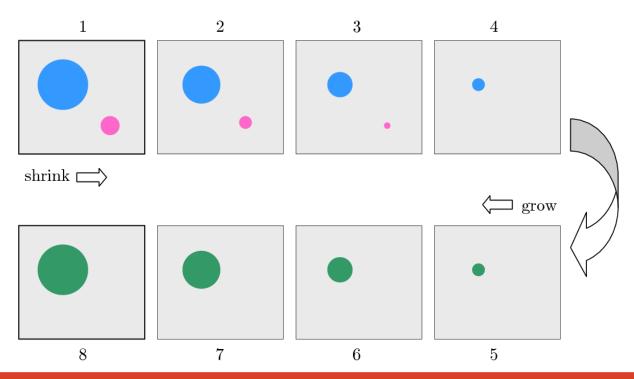




- Morfološki filtri so namenjeni ravno spreminjanju lokalne strukture slike
 - Odstranjevanje majhnih elementov slike
 - Polnjenje lukenj
 - Iskanje obrisov

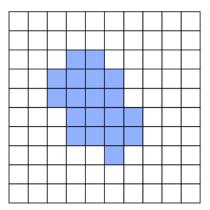
Krčenje in rast

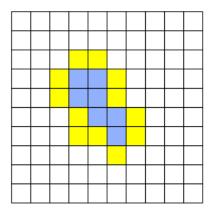
- Odstranjevanje majhnih elementov na sliki
- Algoritem:
 - 1. Vse strukture skrči, tako da se odstranijo zunanji deli struktur
 - Na ta način se izgubijo majhne strukture
 - 2. Preostale struture naj spet zrastejo se dodajajo zunanje plasti
 - Dosežejo približno enako velikost in obliko kot na začetku

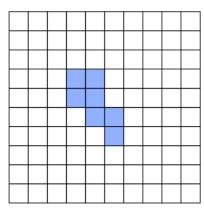


Krčenje in rast

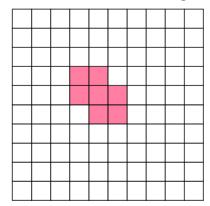
- Krčenje
 - Odstrani se zunanja plast slikovnih elementov

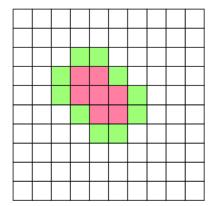


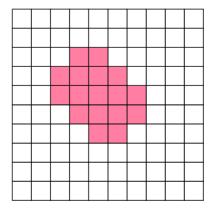




- Rast
 - doda se zunanja plast slikovnih elementov

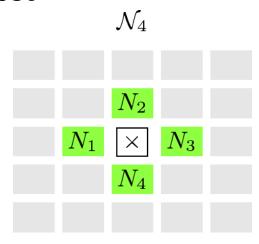




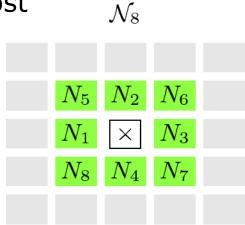


Sosednost slikovnih elementov

4-sosednost



8-sosednost



Osnovne morfološke operacije

- Krčenje in Rast (Shrinking in Growing)
- Erozija in Širitev (Erosion, Dilation)
 - Bolj splošne
 - Definirane s strukturnim elementom
- Strukturni element
 - Definira lastnosti morfološkega filtra
 - Matrika binarnih vrednosti

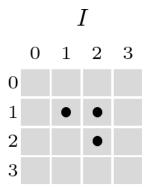
$$H(i,j) \in \{0,1\}$$

$$H = egin{array}{c|c} \bullet & \bullet & \bullet \\ \hline \bullet & \bullet & \bullet \\ \hline \bullet & \bullet & \end{array}$$

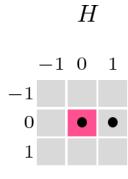
Množice točk

- Morfološke operacije ponavadi izvajamo na binarnih slikah
 $I(u,v) \in \{0,1\}$
- Binarne slike in strukturni element lahko predstavimo z množico točk (parov koordinat p = (u, v)):

$$Q_I = \{ \boldsymbol{p} \mid I(\boldsymbol{p}) = 1 \}$$



$$I \equiv Q_I = \{(1,1), (2,1), (2,2)\}$$



$$H \equiv \mathcal{Q}_H = \{(0,0), (1,0)\}$$

Osnovne binarne operacije

Invertiranje - komplementarna množica

$$\mathcal{Q}_{ar{I}} = ar{\mathcal{Q}}_I = \{oldsymbol{p} \in \mathbb{Z}^2 \mid oldsymbol{p}
otin \mathcal{Q}_I \}$$

ALI – unija

$$\mathcal{Q}_{I_1\vee I_2}=\mathcal{Q}_{I_1}\cup\mathcal{Q}_{I_2}$$

Translacija – prištevanje

$$I_{\boldsymbol{d}} \equiv \{(\boldsymbol{p} + \boldsymbol{d}) \mid \boldsymbol{p} \in I\}$$

Zrcaljenje - negacija

$$H^* \equiv \{ -\boldsymbol{p} \mid \boldsymbol{p} \in H \}$$

Širitev

 Iz Rasti - strukturni element se razmnoži na vsakem slikovnem elementu ospredja

$$I \oplus H \equiv \left\{ (\boldsymbol{p} + \boldsymbol{q}) \mid \text{for some } \boldsymbol{p} \in I \text{ and } \boldsymbol{q} \in H \right\}$$

$$I \oplus H \equiv \bigcup_{\boldsymbol{p} \in I} H_{\boldsymbol{p}} = \bigcup_{\boldsymbol{q} \in H} I_{\boldsymbol{q}}$$

$$I \qquad H \qquad I \oplus H$$

$$0 \quad 1 \quad 2 \quad 3 \qquad -1 \quad 0 \quad 1 \qquad 0 \quad 1 \quad 2 \quad 3$$

$$0 \qquad -1 \qquad 0 \qquad 1 \quad 2 \quad 3$$

$$0 \qquad 0 \qquad \bullet \qquad \bullet \qquad = \quad 1 \qquad \bullet \quad \bullet \quad \bullet$$

$$2 \qquad 1 \qquad 2 \qquad 0 \qquad \bullet$$

$$1 \equiv \{(1,1),(2,1),(2,2)\}, \ H \equiv \{(\boldsymbol{0},\boldsymbol{0}),(\boldsymbol{1},\boldsymbol{0})\}$$

$$I \oplus H \equiv \{ (1,1) + (\boldsymbol{0},\boldsymbol{0}),(1,1) + (\boldsymbol{1},\boldsymbol{0}),(1,1) + (\boldsymbol{1},\boldsymbol{0}),(1,1)$$

 $(2,1) + (\mathbf{0},\mathbf{0}), (2,1) + (\mathbf{1},\mathbf{0}),$

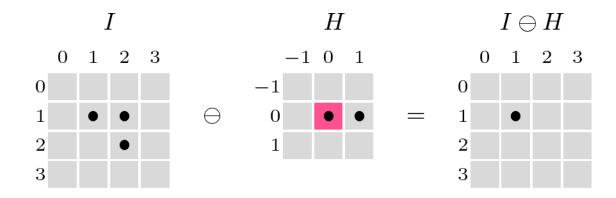
 $(2,2) + (\mathbf{0},\mathbf{0}), (2,2) + (\mathbf{1},\mathbf{0})$

Erozija

Iz Krčenja - kvazi-inverz Širitve

$$I \ominus H \equiv \left\{ \boldsymbol{p} \in \mathbb{Z}^2 \mid (\boldsymbol{p} + \boldsymbol{q}) \in I, \text{ for every } \boldsymbol{q} \in H \right\}$$

 $I \ominus H \equiv \left\{ \boldsymbol{p} \in \mathbb{Z}^2 \mid H_{\boldsymbol{p}} \subseteq I \right\}$



$$I \equiv \{(1,1),(2,1),(2,2)\}, H \equiv \{(\mathbf{0},\mathbf{0}),(\mathbf{1},\mathbf{0})\}$$

$$I\ominus H\equiv\{\,(1,1)\,\}\mbox{ because}$$

$$(1,1)+({\bf 0},{\bf 0})=(1,1)\in I \quad \mbox{and}\quad (1,1)+({\bf 1},{\bf 0})=(2,1)\in I$$

Lastnosti Širitve in Erozije

Komutativnost

$$I \oplus H = H \oplus I$$
$$I \ominus H \neq H \ominus I$$

Asociativnost

$$(I_1 \oplus I_2) \oplus I_3 = I_1 \oplus (I_2 \oplus I_3)$$

$$H_{\text{big}} = H_1 \oplus H_2 \oplus \ldots \oplus H_K$$

$$I \oplus H_{\text{big}} = (\ldots ((I \oplus H_1) \oplus H_2) \oplus \ldots \oplus H_K)$$

$$(I_1 \ominus I_2) \ominus I_3 = I_1 \ominus (I_2 \oplus I_3)$$

Nevtralni element

$$I \oplus \delta = \delta \oplus I = I$$
, with $\delta \equiv \{(0,0)\}$

Dualnost Širitve in Erozije

Širitev ospredja je enaka Eroziji ozadja in obratno

$$I\oplus H \equiv \overline{(\bar{I} \oplus H^*)}$$

$$I \oplus H \equiv \overline{(\bar{I} \oplus H^*)}$$

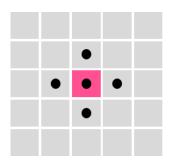
$$I \oplus H = \overline{(\bar{I} \oplus$$

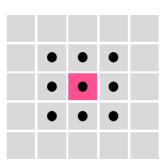
Algoritem

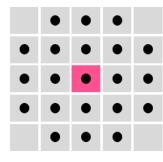
```
DILATE (I, H)
               I: binary image of size w \times h
               H: binary structuring element defined over region \mathcal{R}_H
               Returns the dilated image I' = I \oplus H
           I' \leftarrow \text{new binary image of size } w \times h
           I'(u,v) \leftarrow 0, for all (u,v)
                                                                                                     \triangleright I' \leftarrow \emptyset
                                                                                                   \triangleright (i,j) = q
           for all (i, j) \in \mathcal{R}_H do
 4:
                                                                                                       \triangleright q \in H
                 if H(i,j) = 1 then
 5:
                                                                                               \triangleright I' \leftarrow I' \cup I_a
 6:
                       Merge the shifted I_q with I':
                       for u \leftarrow 0 \dots (w-1) do
 7:
 8:
                             for v \leftarrow 0 \dots (h-1) do
                                                                                                   \triangleright (u,v) = p
                                                                                                          \triangleright p \in I
                                  if I(u,v)=1 then
 9:
                                                                                       \triangleright I' \leftarrow I' \cup (m{p} \!+\! m{q})
                                        I'(u+i,v+j) \leftarrow 1
10:
            return I'.
11:
       Erode (I, H)
12:
                                                                                                      \triangleright \bar{I} \leftarrow \neg I
           \bar{I} \leftarrow \text{INVERT}(I)
13:
           H^* \leftarrow \text{Reflect}(H)
14:
                                                                                      \triangleright I \oplus H = \overline{(\bar{I} \oplus H^*)}
           return Invert(Dilate(\bar{I}, H^*)).
15:
```

Morfološki filtri

- Morfološki filter je točno definiran s
 - Tipom operacije
 - Strukturnim elementom
- Ponavadi se uporabljajo okrogli strukturni elementi (izotropičen filter)
 - Pri Širitvi strukturni element z radijem r doda r slikovnih elementov plasti okrog ospredja
 - Pri Eroziji strukturni element z radijem r zbriše r plasti slikovnih elementov z ospredja

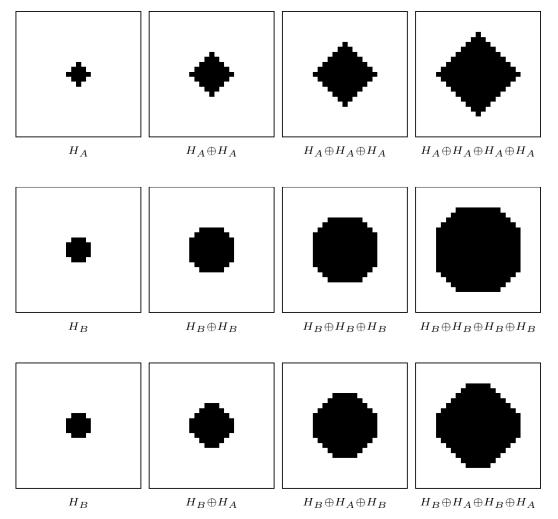


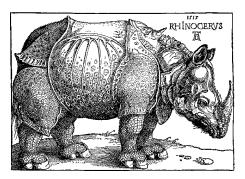




Iterativnost filtrov

 Aproksimacija izotropičnih filtrov z ustreznim vrstnim redom manjših filtrov







Dilation



Erosion



= 1.0



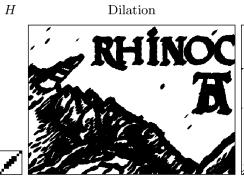


-25



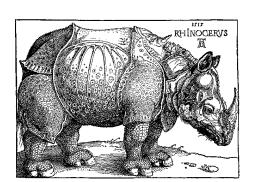


r = 5.0





Erosion













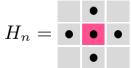


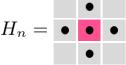


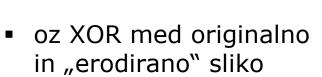
Obrisi

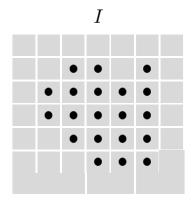
- Določanje mejnih slikovnih elementov
 - Presek med originalno in invertirano "erodirano" sliko

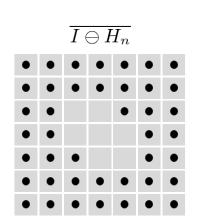
$$I' = I \ominus H_n$$
$$B = I \cap \overline{I'} = I \cap \overline{(I \ominus H_n)}$$

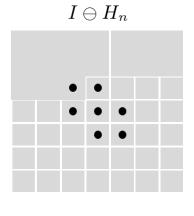


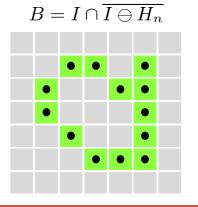




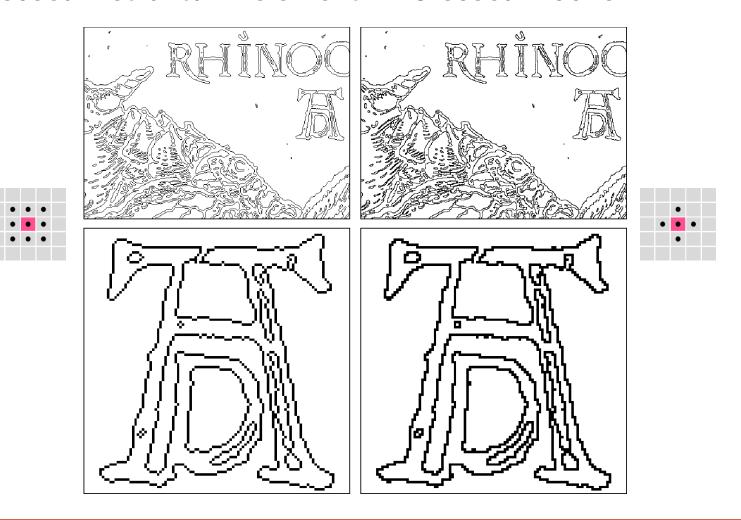








- 8-sosedni strukturni element -> 4-sosedni obris
- 4-sosedni strukturni element -> 8-sosedni obris



Kompozitni operatorji

- Odprtje (opening):
 - Erozija, nato Širitev z istim strukturnim elementom
 - Za odstranjevanje majhnih elementov

$$I \circ H = (I \ominus H) \oplus H$$

- Zaprtje (closing)
 - Širitev, nato erozija
 - Za polnjenje lukenj

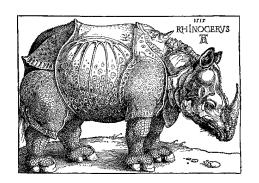
$$I \bullet H = (I \oplus H) \ominus H$$

Operaciji Odprtje in Zaprtje sta

$$\hbox{ Idempotentni:} \begin{array}{l} I\circ H=(I\circ H)\circ H=((I\circ H)\circ H)\circ H=\dots \\ I\bullet H=(I\bullet H)\bullet H=((I\bullet H)\bullet H)\bullet H=\dots \end{array}$$

■ Dualni:
$$I \circ H = \overline{(\bar{I} \bullet H)}$$
 and $I \bullet H = \overline{(\bar{I} \circ H)}$

Primer Odprtja in Zaprtja

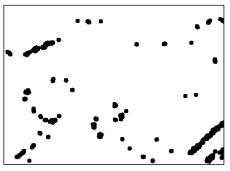














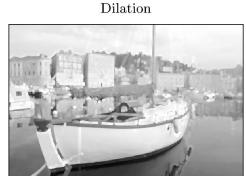




Morfološki operatorji na sivinskih slikah

 Širitev in Erozija sta tudi posplošeni za uporabo na sivinskih slikah

Primer:











r = 5.0

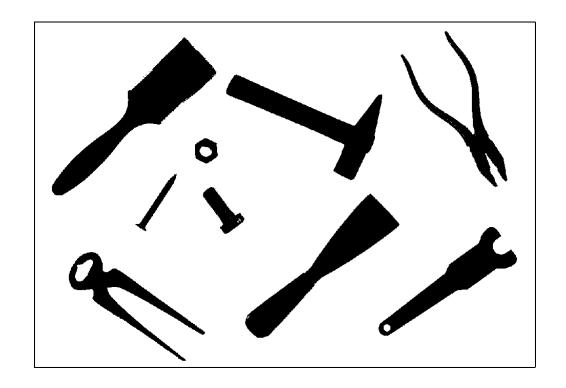




r = 10.0

Regije v binarnih slikah

- Ospredje (foreground)
- Ozadje (background)

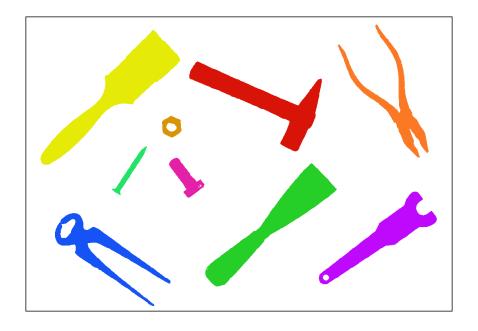


Število in vrsta predmetov na binarnih slikah

Iskanje regij na sliki

- Labeliranje (barvanje) regij
 - Kateri slikovni elementi pripadajo kateri regiji?
 - Koliko regij je na sliki?
 - Kje so regije locirane?

$$I(u, v) = \begin{cases} 0 & background \text{ pixel} \\ 1 & foreground \text{ pixel} \\ 2, 3, \dots \text{ region } label. \end{cases}$$



- Iskanje povezanih slikovnih elementov
 - 4-sosednost
 - 8-sosednost
- Več algoritmov:
 - Poplavljanje (flood filling)
 - Zaporedno označevanje regij (Sequential region marking)
 - Kombinacija obeh

Barvanje regij s poplavljanjem

- Enostaven algoritem:
 - 1. Poišči en neoznačen slikovni element ospredja
 - 2. Označi vse sosednje slikovne elemente ospredja in tako naprej
- Kot poplavljanje pokrajine
- Različni načini poplavljanja:
 - 1. Rekurzivno
 - zelo prostorosko zahtevno
 - Iterativno z uporabo sklada (najprej v globino)
 - 3. Iterativno z uporabo vrste (najprej v širino)
 - najbolj priporočljivo

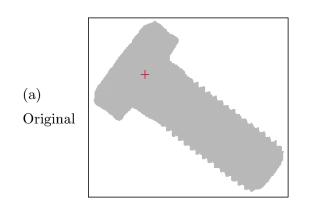
Rekurzivni algoritem

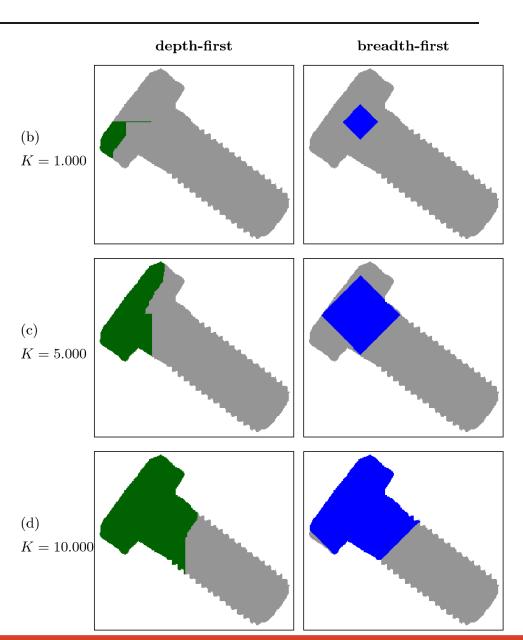
```
1: RegionLabeling(I)
          I: binary image (0 = background, 1 = foreground)
          The image I is labeled (destructively modified) and returned.
 2:
       Initialize m \leftarrow 2 (the value of the next label to be assigned).
 3:
       Iterate over all image coordinates (u, v).
           if I(u,v)=1 then
 4:
               FLOODFILL(I, u, v, m)
                                           ▶ use any of the 3 versions below
 5:
 6:
               m \leftarrow m + 1.
 7:
       return the labeled image I.
    FLOODFILL(I, u, v, label)
                                                      ▶ Recursive Version
9:
       if coordinate (u, v) is within image boundaries and I(u, v) = 1 then
10:
           Set I(u, v) \leftarrow label
           FLOODFILL(I, u+1, v, label)
11:
           FLOODFILL(I, u, v+1, label)
12:
13:
           FLOODFILL(I, u, v-1, label)
14:
           FLOODFILL(I, u-1, v, label)
15:
       return.
```

Iterativna algoritma

```
16: FLOODFILL(I, u, v, label)
                                                             ▷ Depth-First Version
17:
         Create an empty stack S
         Put the seed coordinate \langle u, v \rangle onto the stack: Push(S, \langle u, v \rangle)
18:
         while S is not empty do
19:
20:
              Get the next coordinate from the top of the stack:
                  \langle x, y \rangle \leftarrow \text{Pop}(S)
21:
              if coordinate (x,y) is within image boundaries and I(x,y)=1
                  then
22:
                  Set I(x,y) \leftarrow label
                  PUSH(S, \langle x+1, y \rangle)
23:
                  PUSH(S, \langle x, y+1 \rangle)
24:
                  PUSH(S, \langle x, y-1 \rangle)
25:
                  PUSH(S, \langle x-1, y \rangle)
26:
27:
         return.
28: FLOODFILL(I, u, v, label)
                                                          ▷ Breadth-First Version
29:
         Create an empty queue Q
30:
         Insert the seed coordinate \langle u, v \rangle into the queue: ENQUEUE(Q, \langle u, v \rangle)
31:
         while Q is not empty do
32:
              Get the next coordinate from the front of the queue:
                  \langle x, y \rangle \leftarrow \text{Dequeue}(Q)
33:
              if coordinate \langle x,y\rangle is within image boundaries and I(x,y)=1
                  then
34:
                  Set I(x,y) \leftarrow label
                  ENQUEUE(Q, \langle x+1, y \rangle)
35:
                  ENQUEUE(Q, \langle x, y+1 \rangle)
36:
                  ENQUEUE(Q, \langle x, y-1 \rangle)
37:
38:
                  ENQUEUE(Q, \langle x-1, y \rangle)
39:
         return.
```

- Iterativna algoritma
 - S skladom
 - Z vrsto





Zaporedno označevanje regij

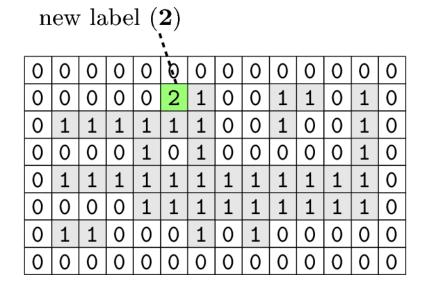
- Sliko sprocesiramo zaporedno od zgornjega levega do spodnjega desnega vogala v dveh korakih
- Korak 1:
 - Če je trenutni slikovni element del ospredja preverimo del njegovih sosedov

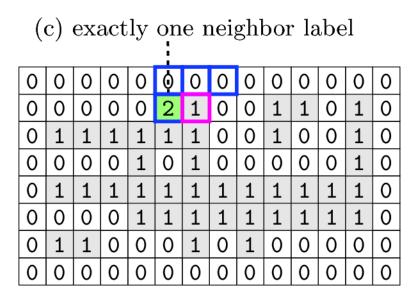
- Če so vsi sosedi del ozadja, inicializiramo novo labelo
- Če imajo vsi ospredni sosedi enako labelo jo priredimo tudi trenutnemu slikovnemu elementu
- Če imajo različne labele, si zapomnimo to kot konflikt
 - Gradimo grafe med seboj povezanih label
- Korak 2:
 - Razrešimo konflikte: poiščemo povezane komponente v grafih konfliktov in popravimo labelo ustreznih slikovnih elementov

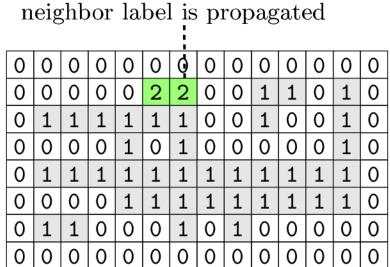
(a)

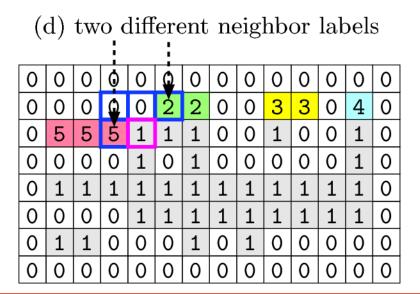
- 0 Background
- 1 Foreground

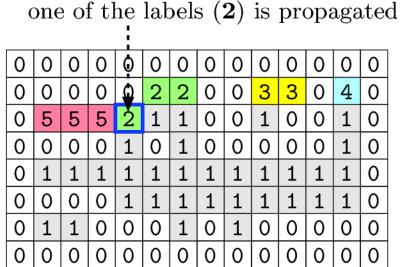
(b) only background neighbors													
0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	1	1	0	0	1	1	0	1	0
0	1	1	1	1	1	1	0	0	1	0	0	1	0
0	0	0	0	1	0	1	0	0	0	0	0	1	0
0	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	0	0	1	1	1	1	1	1	1	1	1	0
0	1	1	0	0	0	1	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0



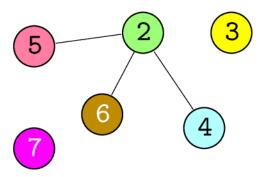








0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	9	2	2	0	0	3	3	0	4	0
0	5	15	15	2	2	2	0	0	3	0	0	4	0
0	0	0	0	2	0	2	0	0	0	0	0	4	0
0	6	ω	Ω	2	2	2	2	2	2	2	2	ဂ	0
0	0	0	0	2	2	2	2	2	2	2	2	2	0
0	7	7	0	0	0	2	0	2	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0



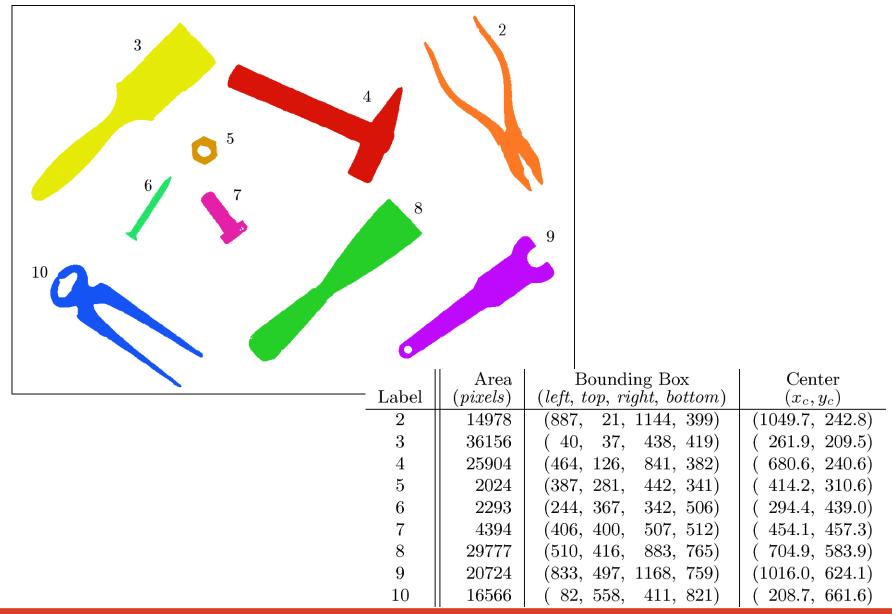
0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	2	2	0	0	3	3	0	2	0
0	2	2	2	2	2	2	0	0	3	0	0	2	0
0	0	0	0	2	0	2	0	0	0	0	0	2	0
0	2	2	2	2	2	2	2	2	2	2	2	2	0
0	0	0	0	2	2	2	2	2	2	2	2	2	0
0	7	7	0	0	0	2	0	2	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0

Algoritem – korak 1

```
1: SequentialLabeling(I)
           I: binary image (0 = background, 1 = foreground)
           The image I is labeled (destructively modified) and returned.
         Pass 1—Assign Initial Labels:
 2:
         Initialize m \leftarrow 2 (the value of the next label to be assigned).
 3:
         Create an empty set \mathcal{C} to hold the collisions: \mathcal{C} \leftarrow \{\}.
         for v \leftarrow 0 \dots H - 1 do
                                                            \triangleright H = \text{height of image } I
 4:
 5:
             for u \leftarrow 0 \dots W - 1 do
                                                            \triangleright W = \text{width of image } I
                 if I(u,v)=1 then do one of:
 6:
 7:
                      if all neighbors of (u, v) are background pixels (all n_i = 0)
                          then
                          I(u,v) \leftarrow m.
 8:
 9:
                          m \leftarrow m + 1.
                      else if exactly one of the neighbors has a label value
10:
                          n_k > 1 then
11:
                          set I(u,v) \leftarrow n_k
12:
                      else if several neighbors of (u, v) have label values n_i > 1
                          then
                           Select one of them as the new label:
13:
                               I(u,v) \leftarrow k \in \{n_i\}.
                          for all other neighbors of u, v) with label values n_i > 1
14:
                               and n_i \neq k do
                               Create a new label collision c_i = \langle n_i, k \rangle.
15:
16:
                               Record the collision: C \leftarrow C \cup \{c_i\}.
         Remark: The image I now contains label values 0, 2, \dots m-1.
```

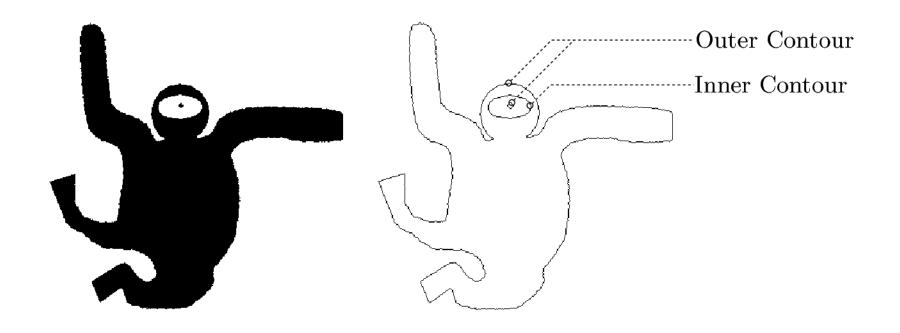
Algoritem – korak 2

```
Pass 2—Resolve Label Collisions:
17:
          Let \mathcal{L} = \{2, 3, \dots m-1\} be the set of preliminary region labels.
18:
           Create a partitioning of \mathcal{L} as a vector of sets, one set for each label
               value: \mathcal{R} \leftarrow [\mathcal{R}_2, \mathcal{R}_3, \dots, \mathcal{R}_{m-1}] = [\{2\}, \{3\}, \{4\}, \dots, \{m-1\}],
               so \mathcal{R}_i = \{i\} for all i \in \mathcal{L}.
          for all collisions \langle a, b \rangle \in \mathcal{C} do
19:
20:
               Find in \mathcal{R} the sets \mathcal{R}_a, \mathcal{R}_b containing the labels a, b, resp.:
                    \mathcal{R}_a \leftarrow the set that currently contains label a
                    \mathcal{R}_b \leftarrow \text{the set that currently contains label } b
               if \mathcal{R}_a \neq \mathcal{R}_b (a and b are contained in different sets) then
21:
22:
                     Merge sets \mathcal{R}_a and \mathcal{R}_b by moving all elements of \mathcal{R}_b to \mathcal{R}_a:
                         \mathcal{R}_a \leftarrow \mathcal{R}_a \cup \mathcal{R}_b
                          \mathcal{R}_b \leftarrow \{\}
          Remark: All equivalent label values (i.e., all labels of pixels in the
          same region) are now contained in the same set \mathcal{R}_i within \mathcal{R}.
          Pass 3—Relabel the Image:
23:
          Iterate through all image pixels (u, v):
24:
               if I(u,v) > 1 then
25:
                     Find the set \mathcal{R}_i in \mathcal{R} that contains label I(u, v).
26:
                     Choose one unique representative element k from the set \mathcal{R}_i
                          (e.g., the minimum value, k = \min(\mathcal{R}_i)).
27:
                     Replace the image label: I(u, v) \leftarrow k.
28:
          return the labeled image I.
```



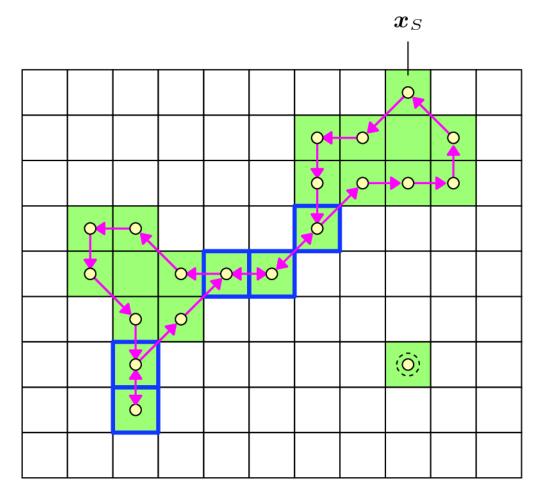
Obrisi regij

- Morfološki operatorji označijo slikovne elemente na obrisu
- Potrebujemo tudi urejeno zaporedje robnih slikovnih elementov
- Zunanji obris in notranji obris



Iskanje obrisa

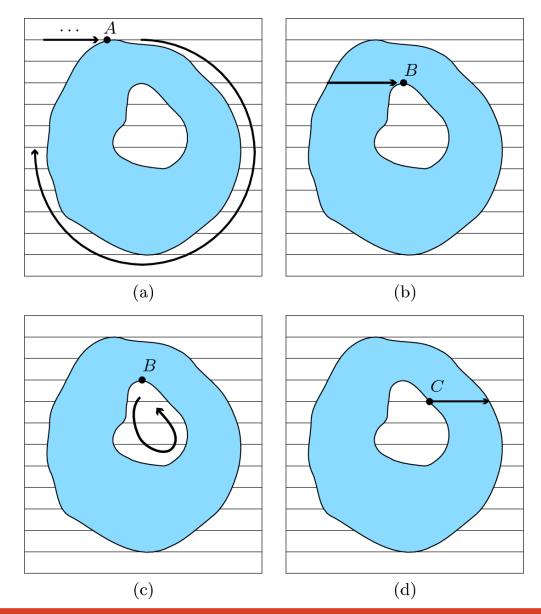
- 1. Poišči povezano regijo
- 2. Poišči rob regije in mu sledi naokrog
- Težave:
 - Notranji obrisi
 - Deli regij debeline en sl. element
 - Izolirani s- elementi



Kombinirano iskanje regij in obrisov

- Hkrati označi regije in poišče obrise
- Algoritem preišče celotno sliko od zgornjega levega do spodnjega desnega slikovnega elementa
- V nekem slikovnem elementu so možni tri primeri:
 - Primer A: BG->neoznačen FG => zunanji obris
 - Dodeli novo oznako regiji
 - Prepotuj celoten zunanji obris in ustrezno označi sl. elemente
 - Sosednje BG sl. Elemente označi z -1
 - Primer B: FG->neoznačen BG => notranji obris
 - Prepotuj celoten notranji obris in ustrezno označi sl. elemente
 - Mejne BG sl. Elemente označi z -1
 - Primer C: označen FG
 - Prenesi oznako na sl. elemente na desni

Kombinirano iskanje regij in obrisov



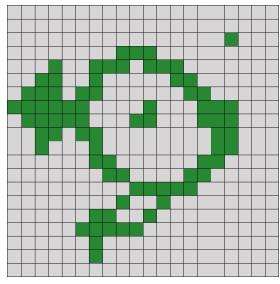
Algoritem

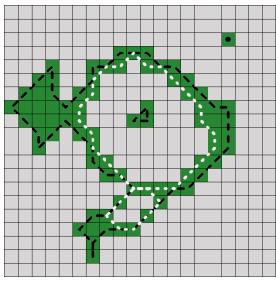
```
1: CombinedContourLabeling (I)
            I: binary image
            Returns a set of contours and a label map (labeled image).
 2:
         Create an empty set of contours: \mathcal{C} \leftarrow \{\}
 3:
         Create a label map LM of the same size as I and initialize:
 4:
         for all (u, v) do
              LM(u,v) \leftarrow 0
 5:
                                                                            \triangleright label map LM
                                                                        \triangleright region counter R
 6:
          R \leftarrow 0
 7:
         Scan the image from left to right and top to bottom:
 8:
         for v \leftarrow 0 \dots N-1 do
              L_k \leftarrow 0
 9:
                                                                          \triangleright current label L_k
              for u \leftarrow 0 \dots M-1 do
10:
11:
                   if I(u,v) is a foreground pixel then
12:
                        if (L_k \neq 0) then
                                                                13:
                             LM(u,v) \leftarrow L
14:
                        else
15:
                             L_k \leftarrow LM(u,v)
                             if (L_k = 0) then
16:
                                                                  ▶ hit new outer contour
17:
                                  R \leftarrow R + 1
                                  L_k \leftarrow R
18:
                                 \boldsymbol{x}_S \leftarrow (u,v)
19:
                                  c_{\text{outer}} \leftarrow \text{TraceContour}(x_S, 0, L_k, I, LM)
20:
21:
                                  \mathcal{C} \leftarrow \mathcal{C} \cup \{c_{\mathrm{outer}}\}
                                                                     ⊳ collect new contour
22:
                                  LM(u,v) \leftarrow L_k
                                                          \triangleright I(u,v) is a background pixel
23:
                   else
24:
                        if (L \neq 0) then
                             if (LM(u,v)=0) then
25:
                                                                   ▶ hit new inner contour
26:
                                  \boldsymbol{x}_S \leftarrow (u-1,v)
                                  c_{\text{inner}} \leftarrow \text{TraceContour}(x_S, 1, L_k, I, LM)
27:
28:
                                  \mathcal{C} \leftarrow \mathcal{C} \cup \{c_{\mathrm{inner}}\}
                                                                     ⊳ collect new contour
29:
                             L \leftarrow 0
30:
          return (C, LM).
                                    > return the set of contours and the label map
```

```
1: TraceContour(x_S, d_S, L_k, I, LM)
               x_S: start position, d_S: initial search direction,
              L_c: label for this contour
              I: original image, LM: label map.
              Traces and returns the contour starting at x_S.
           (x_T, d_{\text{next}}) \leftarrow \text{FINDNEXTPOINT}(x_S, d_S, I, LM)
 2:
 3:
           oldsymbol{c} \leftarrow [oldsymbol{x}_T]
                                                          \triangleright create a contour starting with x_T
 4:
           \boldsymbol{x}_p \leftarrow \boldsymbol{x}_S
                                                              \triangleright previous position \boldsymbol{x}_p = (u_p, v_p)
                                                                \triangleright current position x_c = (u_c, v_c)
 5:
           x_c \leftarrow x_T
 6:
           done \leftarrow (\boldsymbol{x}_S \equiv \boldsymbol{x}_T)
                                                                                       ▷ isolated pixel?
 7:
           while (\neg done) do
 8:
                LM(u_c, v_c) \leftarrow L_c
 9:
                d_{\text{search}} \leftarrow (d_{\text{next}} + 6) \mod 8
                (x_n, d_{\text{next}}) \leftarrow \text{FINDNEXTPOINT}(x_c, d_{\text{search}}, I, LM)
10:
11:
                 \boldsymbol{x}_{p} \leftarrow \boldsymbol{x}_{c}
12:
                 \boldsymbol{x}_c \leftarrow \boldsymbol{x}_n
13:
                 done \leftarrow (\boldsymbol{x}_p \equiv \boldsymbol{x}_S \wedge \boldsymbol{x}_c \equiv \boldsymbol{x}_T)
                                                                              ▶ back at start point?
14:
                if (\neg done) then
                      APPEND(c, x_n)
15:
                                                                     \triangleright add point x_n to contour c
                                                                                ⊳ return this contour
16:
           return c.
17: FINDNEXTPOINT(x_c, d, I, LM)
              x_c: start point, d: search direction,
              I: original image, LM: label map.
           for i \leftarrow 0 \dots 6 do
18:
                                                                             ⊳ search in 7 directions
                x' \leftarrow x_c + \text{Delta}(d)
                                                                                          \triangleright x' = (u', v')
19:
                if I(u',v') is a background pixel then
20:
21:
                      LM(u',v') \leftarrow -1
                                                           \triangleright mark background as visited (-1)
                      d \leftarrow (d+1) \bmod 8
22:
                                                        \triangleright found a nonbackground pixel at x'
23:
                 else
24:
                      return (x', d)
25:
           return (\boldsymbol{x}_c, d).
                                                  ▶ found no next point, return start point
```

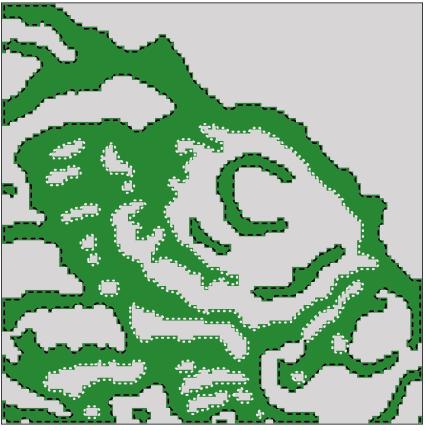
```
26: Delta(d) = (\Delta x, \Delta y), with \begin{vmatrix} d & 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 \\ \Delta x & 1 & 1 & 0 & -1 & -1 & -1 & 0 & 1 \\ \Delta y & 0 & 1 & 1 & 1 & 0 & -1 & -1 & -1 \end{vmatrix}
```

Primer







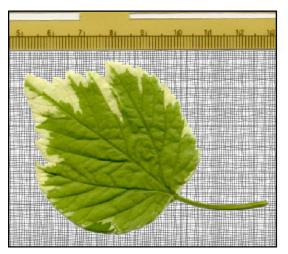


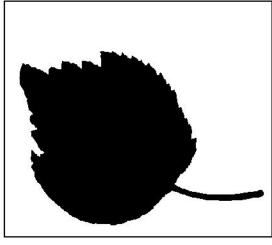
Predstavitve slikovnih regij

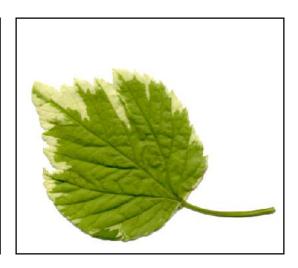
- Predstavitev z matriko
- Run Length Encoding (RLE)
- Absolutna verižna koda
- Diferenčna verižna koda

Predstavitev z matriko

- Najbolj klasična predstavitev
- Binarna maska določa regijo:







- Predstavitev ni odvisna vsebine slike
- Pogosto neučinkovita (prostorsko potratna)

Run Length Encoding (RLE)

D:1----

Kodiranje zaporedij slikovnih elementov ospredja s tremi parametri:

$$Run_i = \langle row_i, column_i, length_i \rangle$$

Primer:

Bitmap										RLE			
	0	1	2	3	4	5	6	7	8	/ 1 1 1			
0										− ⟨row, column, length⟩			
1			×	×	×	×	×	×		$\begin{array}{ccc} \langle 1,2,6 \rangle \\ \langle 3,4,4 \rangle \\ \langle 4,1,3 \rangle \\ \langle 4,5,3 \rangle \\ \langle 5,0,9 \rangle \end{array}$			
2													
3					×	×	X	×					
4		X	×	×		X	X	X					
5		×	×	×	×	×	×	×	×				
6										(0,0,0)			

DID

- Prvi parameter lahko spustimo, če je vrstica ista
- Enostavna brezizgubna metoda za kompresijo
- Že zelo dolgo, veliko uporabljana (faks, algoritmi za kompresijo slik, itd.)

Absolutna verižna koda

- Absolute Chain Code, tudi Freemanove kode
- Kodiranje obrisa regije

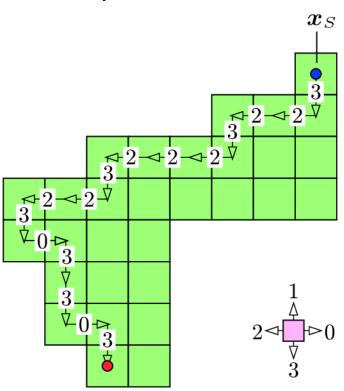
$$\boldsymbol{c}_{\mathcal{R}} = [\boldsymbol{x}_0, \, \boldsymbol{x}_1, \dots \boldsymbol{x}_{M-1}] \text{ with } \boldsymbol{x}_i = \langle u_i, v_i \rangle$$

Začnemo v nekem robnem slikovnem elementu in gremo po obrisu, pri čemer vsak korak zakodiramo:

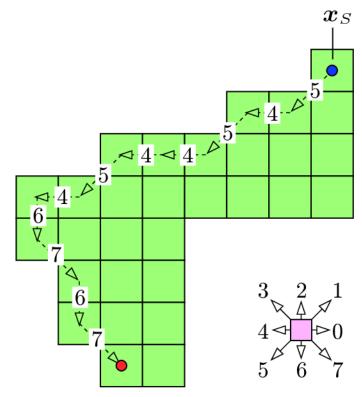
$$\begin{aligned} c_{\mathcal{R}}' &= [c_0', c_1', \dots c_{M-1}'] \\ c_i' &= \text{Code}(\Delta u_i, \Delta v_i) \\ (\Delta u_i, \Delta v_i) &= \begin{cases} (u_{i+1} - u_i, v_{i+1} - v_i) & \text{for } 0 \leq i < M-1 \\ (u_0 - u_i, v_0 - v_i) & \text{for } i = M-1, \end{cases} \\ \frac{\Delta u}{\Delta v} & 1 & 1 & 0 & -1 & -1 & 0 & 1 \\ \frac{\Delta v}{\Delta v} & 0 & 1 & 1 & 1 & 0 & -1 & -1 & -1 \\ \hline Code(\Delta u, \Delta v) & 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 \end{aligned}$$

Primer

Lahko upoštevamo 4- ali 8-sosednost



4-Chain Code 322322232303303...111 length = 28



8-Chain Code 54544546767...222 $length = 16 + 6\sqrt{2} \approx 24,5$

Koda je odvisna od začetne točke in orientacije obrisa

Diferenčna verižna koda

- Differential chain code
- Namesto kodiranja razlike pozicij, kodiramo razlike smeri

$$c_i'' = \begin{cases} (c_{i+1}' - c_i') \mod 8 & \text{for } 0 \le i < M - 1 \\ (c_0' - c_i') \mod 8 & \text{for } i = M - 1 \end{cases}$$

Primer:

$$c_{\mathcal{R}}' = [5, 4, 5, 4, 4, 5, 4, 6, 7, 6, 7, \dots 2, 2, 2]$$

 $c_{\mathcal{R}}'' = [7, 1, 7, 0, 1, 7, 2, 1, 7, 1, 1, \dots 0, 0, 3]$

- Sedaj lahko regijo zarotiramo za 90 stopinj, pa se koda ne spremeni
- Še vedno pa je koda odvisna od začetne pozicije

Števila oblik

- Shape numbers
- Verižne kode moramo poravnati, da imajo enako začetno točko, potem jih lahko primerjamo
- Elemente kode interpretiramo kot števke v bazi b (8 ali 4):

$$VAL(\mathbf{c}_{\mathcal{R}}'') = c_0'' \cdot b^0 + c_1'' \cdot b^1 + \dots = \sum_{i=0}^{M-1} c_i'' \cdot b^{i-1}$$

- Poiščemo maksimum in kodo zamaknemo za k mest $k_{\max} = \arg\max_{0 \le k \le M} \mathrm{VAL}(\boldsymbol{c}''_{\mathcal{R}} \rhd k)$
- Ni potrebno dejansko tega računati, lahko samo sortiramo
- Primer: $m{c}''_{\mathcal{R}} = [\,0,1,3,2,\ldots 9,3,7,4\,]$ $m{c}''_{\mathcal{R}} \rhd 2 = [\,7,4,0,1,3,2,\ldots 9,3\,]$
- Zdaj koda ni več odvisna od začetnega položaja, lahko je regija rotirana za 90 stopinj
- Še vedno pa predstavitev ni invariantna na poljubne rotacije in skaliranje
 - Rešitev: Fourierjevi deskriptorji

Lastnosti binarnih regij

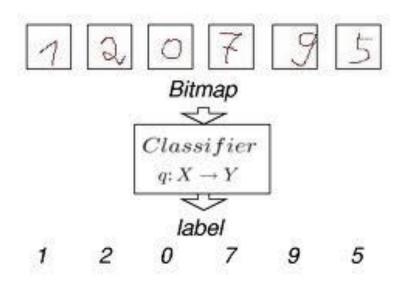
- Kako opisati sliko
 - Matrika števil: zaporedje slikovnih elementov
 - "Pretežno bel krog na pretežno zeleni podlagi."
 - "Nogometna žoga na travi."



- Zelo težko je priti do semantičnega opisa slike
- Lažje posamezne regije na sliki opišemo z enostavnimi lastnostmi

Značilnice

- Ko imamo regije, jih opišemo z značilnicami
- Na osnovi vektorjev značilnic lahko regije primerjamo med seboj
 - Lahko iščemo podobnosti med regijami
 - Lahko merimo in preverjamo dimenzije in oblike
- Primer: OCR:





Geometrične značilnice

Geometrične značilnice definiramo za binarno regijo

$$\mathcal{R} = \{ \boldsymbol{x}_0, \boldsymbol{x}_1 \dots \boldsymbol{x}_{N-1} \} = \{ (u_0, v_0), (u_1, v_1) \dots (u_{N-1}, v_{N-1}) \}$$

- Obseg (Perimeter)
- Površina (Area)
- Kompaktnost oz. okroglost (Compactness and roundness)
- Obsegajoč pravokotnik (Bounding box)
- Konveksna ovojnica (Convex hull)

Obseg

- Perimeter
- Dolžina zunanjega obsega povezane regije
- V primeru 4-sosednosti bo dolžina večja od dejanske
- Ponavadi uporabljamo 8-sosedno Verižno kodo

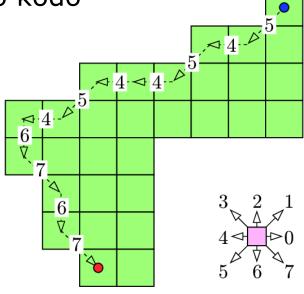
$$\mathbf{c}_{\mathcal{R}}' = [c_0', c_1', \dots c_{M-1}']$$

$$\mathsf{Perimeter}(\mathcal{R}) = \sum_{i=0}^{M-1} \operatorname{length}(c_i')$$

with length(c) =
$$\begin{cases} 1 & \text{for } c = 0, 2, 4, 6 \\ \sqrt{2} & \text{for } c = 1, 3, 5, 7 \end{cases}$$

- To nam vrne nekoliko prevelik obseg
 - Zaradi diskretizacije poševnih ravnih črt
 - Normaliziramo:

$$P(\mathcal{R}) \approx \mathsf{Perimeter}_{\mathsf{corr}}(\mathcal{R}) = 0.95 \cdot \mathsf{Perimeter}(\mathcal{R})$$



8-Chain Code

$$54544546767...222$$
 $length = 16 + 6\sqrt{2} \approx 24.5$

 \boldsymbol{x}_S

Površina

- Area
- Število slikovnih elementov v regiji:

$$A(\mathcal{R}) = |\mathcal{R}| = N.$$

- Lahko tudi ocenimo iz obrisa
 - če regija nima lukenj
 - Uporabimo zapis v Verižni kodi:

$$A(\mathcal{R}) \approx \frac{1}{2} \cdot \left| \sum_{i=0}^{M-1} \left(u_i \cdot v_{(i+1) \bmod M} - u_{(i+1) \bmod M} \cdot v_i \right) \right|$$

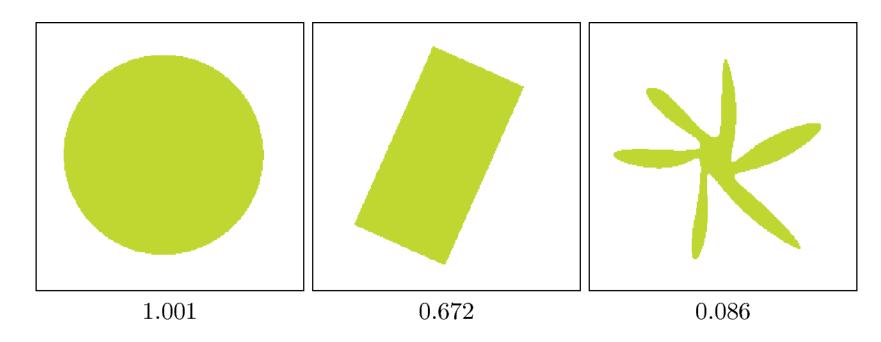
- Tako obseg kot površina sta
 - Neodvisna od translacije in rotacije
 - Odvisna od skale (velikosti, oddaljenosti)

Kompaktnost oz. okroglost

- Compactness and circularity
- Meri kako kompaktna oz. okrogla je regija

$$Circularity(\mathcal{R}) = 4\pi \cdot \frac{A(\mathcal{R})}{P^2(\mathcal{R})}$$

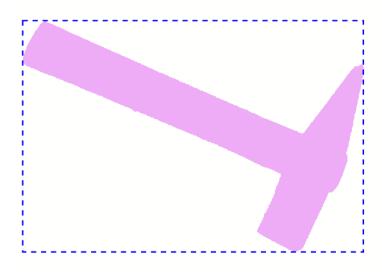
- Invariantna na translacijo, rotacijo in skalo
- Primer:



Obsegajoč pravokotnik

- Bounding box
- Najmanjši pravokotnik z osmi vzporednimi s koordinatnimi osmi, ki vsebuje celotno regijo

$$\mathsf{BoundingBox}(\mathcal{R}) = \langle u_{\min}, u_{\max}, v_{\min}, v_{\max} \rangle$$



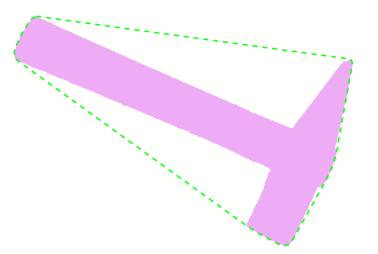
Konveksna ovojnica

Convex hull

Najmnjši poligon, ki vsebuje vse elemente regije

Računamo jo lahko z algoritmom QuickHull (kompleksnost)

O(NH)



- Konveksnost = dolžina konveksne ovojnice / obseg regije
- Gostota = površina regije / površina konveksne ovojnice
- Premer = razdalja me dvema maksimalno oddaljenima točkama na konveksni ovojnici

Statistične značilnice oblike

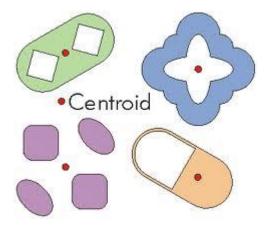
- Elemente regije obravnavamo kot točke točke porazdeljne v 2D prostoru
- Centroid
- Momanti
- Centralni momenti
- Normalizirani centralni momenti
- Geometrične lastnosti, ki temeljijo na momentih
 - Orientacija
 - Ekscentričnost
 - Invariantni momenti

Centroid

- Težišče regije
- Aritmetična vsota koordinat v smereh x in y

$$\bar{x} = \frac{1}{|\mathcal{R}|} \sum_{(u,v)\in\mathcal{R}} u$$
 and $\bar{y} = \frac{1}{|\mathcal{R}|} \sum_{(u,v)\in\mathcal{R}} v$

Primeri:



Momenti

Splošni statistični koncept:

$$m_{pq} = \sum_{(u,v)\in\mathcal{R}} I(u,v) \cdot u^p v^q$$

Za binarne slike:

$$m_{pq} = \sum_{(u,v)\in\mathcal{R}} u^p v^q$$

Površina:

$$A(\mathcal{R}) = |\mathcal{R}| = \sum_{(u,v)\in\mathcal{R}} 1 = \sum_{(u,v)\in\mathcal{R}} u^0 v^0 = m_{00}(\mathcal{R})$$

Centroid:

$$\bar{x} = \frac{1}{|\mathcal{R}|} \cdot \sum_{(u,v)\in\mathcal{R}} u^1 v^0 = \frac{m_{10}(\mathcal{R})}{m_{00}(\mathcal{R})} \qquad \bar{y} = \frac{1}{|\mathcal{R}|} \cdot \sum_{(u,v)\in\mathcal{R}} u^0 v^1 = \frac{m_{01}(\mathcal{R})}{m_{00}(\mathcal{R})}$$

Centralni momenti

 Vzamemo centroid za referenčno točko – središče koordinatnega sistema

$$\mu_{pq}(\mathcal{R}) = \sum_{(u,v)\in\mathcal{R}} I(u,v) \cdot (u-\bar{x})^p \cdot (v-\bar{y})^q$$

Za binarne slike (regije):

$$\mu_{pq}(\mathcal{R}) = \sum_{(u,v)\in\mathcal{R}} (u - \bar{x})^p \cdot (v - \bar{y})^q$$

- Momenti tako niso več odvisni od položaja regije na sliki
- Normalizirani centralni momenti:
 - Normalizirati moramo za faktor $s^{(p+q+2)}$

$$\bar{\mu}_{pq}(\mathcal{R}) = \mu_{pq} \cdot \left(\frac{1}{\mu_{00}(\mathcal{R})}\right)^{(p+q+2)/2}$$

- Centralni momenti so tako invariantni tudi na skalo
- Primerni za klasifikacijo regij

Orientacija

Smer glavne osi

$$\tan(2\theta_{\mathcal{R}}) = \frac{2 \cdot \mu_{11}(\mathcal{R})}{\mu_{20}(\mathcal{R}) - \mu_{02}(\mathcal{R})}$$

$$\theta_{\mathcal{R}} = \frac{1}{2} \tan^{-1} \left(\frac{2 \cdot \mu_{11}(\mathcal{R})}{\mu_{20}(\mathcal{R}) - \mu_{02}(\mathcal{R})} \right)$$

$$-x$$

$$-x$$

$$+y$$

$$+y$$

Orientacija

Vizualizacija vektorja smeri orientacije:

$$m{x} = ar{m{x}} + \lambda \cdot m{x}_d = \begin{pmatrix} ar{x} \\ ar{y} \end{pmatrix} + \lambda \cdot \begin{pmatrix} \cos(heta_{\mathcal{R}}) \\ \sin(heta_{\mathcal{R}}) \end{pmatrix}$$

$$\tan(2\theta_{\mathcal{R}}) = \frac{2 \cdot \mu_{11}(\mathcal{R})}{\mu_{20}(\mathcal{R}) - \mu_{02}(\mathcal{R})} = \frac{A}{B} = \frac{\sin(2\theta_{\mathcal{R}})}{\cos(2\theta_{\mathcal{R}})}$$

$$x_d = \cos(\theta_{\mathcal{R}}) = \begin{cases} 0 & \text{for } A = B = 0\\ \left[\frac{1}{2} \left(1 + \frac{B}{\sqrt{A^2 + B^2}}\right)\right]^{\frac{1}{2}} & \text{otherwise,} \end{cases}$$

$$y_d = \sin(\theta_{\mathcal{R}}) = \begin{cases} 0 & \text{for } A = B = 0\\ \left[\frac{1}{2} \left(1 - \frac{B}{\sqrt{A^2 + B^2}}\right)\right]^{\frac{1}{2}} & \text{for } A \ge 0\\ -\left[\frac{1}{2} \left(1 - \frac{b}{\sqrt{A^2 + B^2}}\right)\right]^{\frac{1}{2}} & \text{for } A < 0, \end{cases}$$

Ekscentričnost

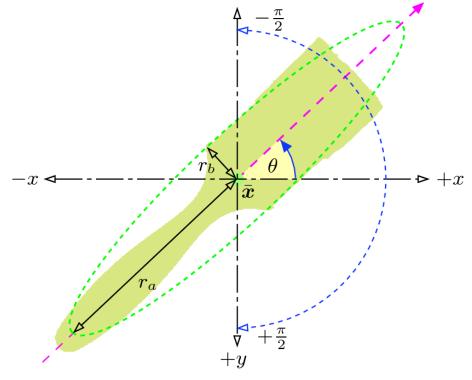
Poldogovatost regije

$$\mathsf{Ecc}(\mathcal{R}) = \frac{a_1}{a_2} = \frac{\mu_{20} + \mu_{02} + \sqrt{(\mu_{20} - \mu_{02})^2 + 4 \cdot \mu_{11}^2}}{\mu_{20} + \mu_{02} - \sqrt{(\mu_{20} - \mu_{02})^2 + 4 \cdot \mu_{11}^2}}$$

• $a_1=2\lambda_1$, $a_2=2\lambda_2$ sta večkratnika lastnih vrednosti matrike

$$m{A} = egin{pmatrix} \mu_{20} & \mu_{11} \\ \mu_{11} & \mu_{02} \end{pmatrix}$$

- Vrednosti med 1 in ∞
 - Ecc=1 => krog
 - Ecc>1 => podolgovata regija
- Invariantna na orientacijo in velikost



Ekscentričnost

- Vizualizacija elipse, ki ponazarja podolgovatost
- Dolžine osi elipse:

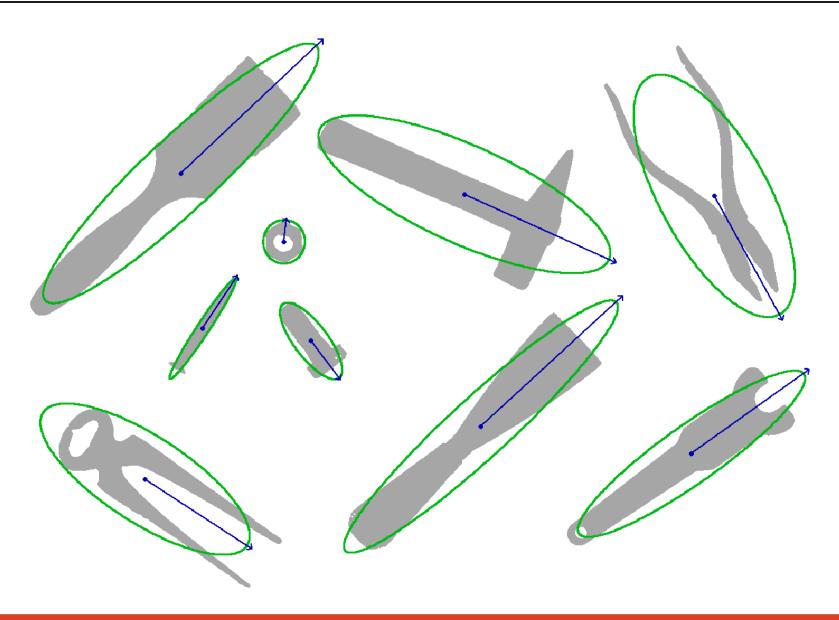
$$r_a = 2 \cdot \left(\frac{\lambda_1}{|\mathcal{R}|}\right)^{\frac{1}{2}} = \left(\frac{2a_1}{|\mathcal{R}|}\right)^{\frac{1}{2}}$$
$$r_b = 2 \cdot \left(\frac{\lambda_2}{|\mathcal{R}|}\right)^{\frac{1}{2}} = \left(\frac{2a_2}{|\mathcal{R}|}\right)^{\frac{1}{2}}$$

Parametrična enačba elipse:

$$\begin{pmatrix} x(t) \\ y(t) \end{pmatrix} = \begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix} + \begin{pmatrix} \cos(\theta) - \sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix} \cdot \begin{pmatrix} r_a \cdot \cos(t) \\ r_b \cdot \sin(t) \end{pmatrix}$$

$$= \begin{pmatrix} \bar{x} + \cos(\theta) \cdot r_a \cdot \cos(t) - \sin(\theta) \cdot r_b \cdot \sin(t) \\ \bar{y} + \sin(\theta) \cdot r_a \cdot \cos(t) + \cos(\theta) \cdot r_b \cdot \sin(t) \end{pmatrix}$$

Primer



Invariantni momenti

- Hujevi momenti oz. Invariantni momenti
- So invariantni tudi na orientacijo

$$H_{1} = \bar{\mu}_{20} + \bar{\mu}_{02}$$

$$H_{2} = (\bar{\mu}_{20} - \bar{\mu}_{02})^{2} + 4\bar{\mu}_{11}^{2}$$

$$H_{3} = (\bar{\mu}_{30} - 3\bar{\mu}_{12})^{2} + (3\bar{\mu}_{21} - \bar{\mu}_{03})^{2}$$

$$H_{4} = (\bar{\mu}_{30} + \bar{\mu}_{12})^{2} + (\bar{\mu}_{21} + \bar{\mu}_{03})^{2}$$

$$H_{5} = (\bar{\mu}_{30} - 3\bar{\mu}_{12}) \cdot (\bar{\mu}_{30} + \bar{\mu}_{12}) \cdot \left[(\bar{\mu}_{30} + \bar{\mu}_{12})^{2} - 3(\bar{\mu}_{21} + \bar{\mu}_{03})^{2} \right]$$

$$+ (3\bar{\mu}_{21} - \bar{\mu}_{03}) \cdot (\bar{\mu}_{21} + \bar{\mu}_{03}) \cdot \left[3(\bar{\mu}_{30} + \bar{\mu}_{12})^{2} - (\bar{\mu}_{21} + \bar{\mu}_{03})^{2} \right]$$

$$H_{6} = (\bar{\mu}_{20} - \bar{\mu}_{02}) \cdot \left[(\bar{\mu}_{30} + \bar{\mu}_{12})^{2} - (\bar{\mu}_{21} + \bar{\mu}_{03})^{2} \right]$$

$$+ 4\bar{\mu}_{11} \cdot (\bar{\mu}_{30} + \bar{\mu}_{12}) \cdot (\bar{\mu}_{21} + \bar{\mu}_{03})$$

$$H_{7} = (3\bar{\mu}_{21} - \bar{\mu}_{03}) \cdot (\bar{\mu}_{30} + \bar{\mu}_{12}) \cdot \left[(\bar{\mu}_{30} + \bar{\mu}_{12})^{2} - 3(\bar{\mu}_{21} + \bar{\mu}_{03})^{2} \right]$$

$$+ (3\bar{\mu}_{12} - \bar{\mu}_{30}) \cdot (\bar{\mu}_{21} + \bar{\mu}_{03}) \cdot \left[3(\bar{\mu}_{30} + \bar{\mu}_{12})^{2} - (\bar{\mu}_{21} + \bar{\mu}_{03})^{2} \right]$$

Ponavadi se uporablja logaritme vrednosti

Projekcije

- Enodimenzionalna predstavitev vsebine slike
- Slikovne elemente projiciramo na x in y koordinato:

$$P_{\text{hor}}(v_0) = \sum_{u=0}^{M-1} I(u, v_0) \qquad \text{for } 0 < v_0 < N$$

$$P_{\text{ver}}(u_0) = \sum_{v=0}^{N-1} I(u_0, v) \qquad \text{for } 0 < u_0 < M$$

- Včasih tako lahko ločimo dokument na smiselne celote
- Projiciramio lahko tudi na glavni osi regije



Topološke lastnosti

- Zajamejo strukturne lastnosti
- So invariantne tudi na zelo močne transformacije slike
- Konveksnost regije
- Število lukenj: $N_L(\mathcal{R})$
- Eulerjevo število: $N_E(\mathcal{R}) = N_R(\mathcal{R}) N_L(\mathcal{R})$