

Image Analysis

Segmentation

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- Which are the background pixels? Which are object pixels?
- Which pixels belong to different objects?





• Find finite set of disjont regions R₁, R₂... covering the image R:

$$R = \bigcup_{i=1}^{n} R_{i} \qquad \forall i, j : i \neq j \Longrightarrow R_{i} \bigcap R_{j} = \phi$$





Thresholding

Edge-based Segmentation Region-based Segmentation Motion for Segmentation Graph Cut

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Thresholding

 Input: gray-scale intensity image f(x,y) and threshold T or domain D Output: binary image g(x,y)

$$g(x, y) = \begin{cases} 1, f(x, y) \ge T \\ 0, f(x, y) < T \end{cases}$$

$$g(x, y) = \begin{cases} 1, f(x, y) \in D\\ 0, else \end{cases}$$



Human blood cells (Wikipedia)



Thresholding

• Where are the stars in the image?



http://en.wikipedia.org/wiki/File:Pleiades_large.jpg







Thresholding

- Global thresholding
- Adaptive (local) thresholding
 - Threshold selection for each region
 - Moving average
- Threshold selection for image/region
 - Based on spectrum (intensity histogram)
 - Multi-spectral method (color images)
 - Multi-dimensional spectra methods
- Avoid Noise
 - ignore small regions in the threshold image



Iterative Threshold Selection



```
B = set containing the pixels in the corners;
O = set containing all other pixels;
T = undefined, T_next = undefined;
while (T == undefined) or (T_next == undefined) or (T_next != T) {
   T = T_next;
   sum B=0;
   for all pixel p in B { sum_B = sum_B + intensity(p); }
   avg_B =sum_B/size(B);
   sum O=0;
   for all pixel p in O { sum O = sum O + intensity(p); }
   avg O = sum O/size(O);
   T_next=(avg_B+avg_O)/2;
   B = empty set, O = empty set;
   for all pixel p in the image
      if (intensity(p) > T_next) O.put(p) else B.put(p);
}
// finally T contains the selected threshold
```



Thresholding

Edge-based Segmentation

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Edge-based segmentation

- First step
 - Edge detection
- Second step "correction"
 Edge Linking and Boundary Detection
 - Local processing
 - Regional Processing
- Resulting edges separate regions of the image







Local processing

```
i_g = gradientImage(image);
```

for all pixels **p** in **i_g** {

}

```
L = local neighborhood of p // set of pixels "near" p
```

```
for all pixels p_n in L {
```

```
if (absoluteValue(magnitude(p_n)-magnitude(p)) )
```

```
&& (absoluteValue(direction(p_n)-direction(p))) {
    connect(p,p_n);
```

```
Coperander Integration (Coperander Integration (Copera
```

```
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```



Regional Processing

• Poligonal approximation of region boundaries

```
Sequence P = orderEdgePixelsCyclic();
(A, B) = specifyStartingPoints();
T = specifyThreshold();
OPEN = new Stack(), CLOSED = new Stack();
OPEN.put(B); OPEN.put(A); CLOSED.put(B);
While (not empty(OPEN))
  LINE = calculateLine(OPEN.topElement(),CLOSED.topElement());
  for each X in P {
     PixelsAndDistances = new Set();
     if (X between OPEN.topElement() and CLOSED.topElement() in P)
         d = calculateDistance(X,LINE); PixelsAndDistances.put((X,d));
     (Y, dist Y) = selectPixelWithMaximalDistance(PixelsAndDistances);
     if (dist Y > T) { OPEN.put(Y); }
     else { Z = OPEN.getAndRemoveTopElement(); CLOSED.put(Z); }
  }
```



Thresholding

Edge-based Segmentation

Region-based Segmentation

Motion for Segmentation Graph Cut

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Region-based Segmentation

- Often applied on the gradient of the image (i.e. not directly)
- Region Splitting and Merging
 - Given region predicate Q(R)
 - Recursively split image into 4 regions, as long as Q(R)=false
 - Merge adjacent regions P and R, if Q(P)=true and Q(R)=true
- Watersheds
 - Topographic interpretation of the image
 - "Flood"
 - Dam Construction (build dam, if dilations would merge two regions)
 - Avoid oversegmentation
 - Internal Markers (areas of interest, e.g. local minima areas)
 - External Markers (separate the image into regions)



Thresholding Edge-based Segmentation Region-based Segmentation

Motion for Segmentation

Graph Cut

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Motion for Segmentation





Motion for segmentation

- Suppose we are given two "similar" pictures: one with the target object, an other one without → segmentation trivial
- But
 - Who gives us such similar images?
 - Noise between "similar" images
- Videos
 - Series of images
 - Accumulate differences: Absolute, Positive, Negative Difference Images
 - Increment pixels if |R(x, y) f(x, y)| > T

(R(x, y) - f(x, y)) > T(R(x, y) - f(x, y)) < -T

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Thresholding Edge-based Segmentation Region-based Segmentation Motion for Segmentation **Graph Cut**

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Graph Cut

- Input: image *I* with some labeled pixels
 - − $O \rightarrow Object, B \rightarrow Background$
- Task: label the other pixels
- *B*(*p*,*q*)
 - Small if p and q belong to different regions
 (e.g. p should be labeled as O and q should be labeled as B)
 - Large if p and q belong to the same regions
 (e.g. both should have label as O)
- *R*(*p*,*B*) cost of labelling *p* as background
- R(p,O) cost of labelling p as object

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Graph Cut

- Construction of the flow problem
 - Image pixels \rightarrow graph nodes
 - If the pixels p and q are adjacent \rightarrow edge between the corresponding nodes with weight B(p,q)
 - Two "special" nodes: s and t
 - Weights of edges (s,p)
 - *R*(*p*,*B*) for unlabeled pixels
 - K for pixels labeled as object
 (K denotes such a big capacity that the edge can not be saturated)
 - 0 for pixels labeled as background
 - Weights of edges (p,t)
 - *R*(*p*,*O*) for unlabeled pixels
 - 0 for pixels labeled as object
 - *K* for pixels labeled as background



Graph Cut

• How to choose *R* and *B* ?

 $R(p,O) = -\ln P(I_p \mid O)$ $R(p,B) = -\ln P(I_p \mid B)$ $B(p,q) = \frac{1}{\parallel p,q \parallel} e^{-\frac{(I_p - I_q)^2}{2\sigma^2}}$

where

 I_p denotes the intensity of pixel p

 σ denotes the expected intensity variation within the object/background //p,q// denotes the distance between p and q

 After solving the flow problem, the image segmentation is determined based on the edges in min. cut

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