

Improved Object Recognition -- The RoboCup 4 legged league

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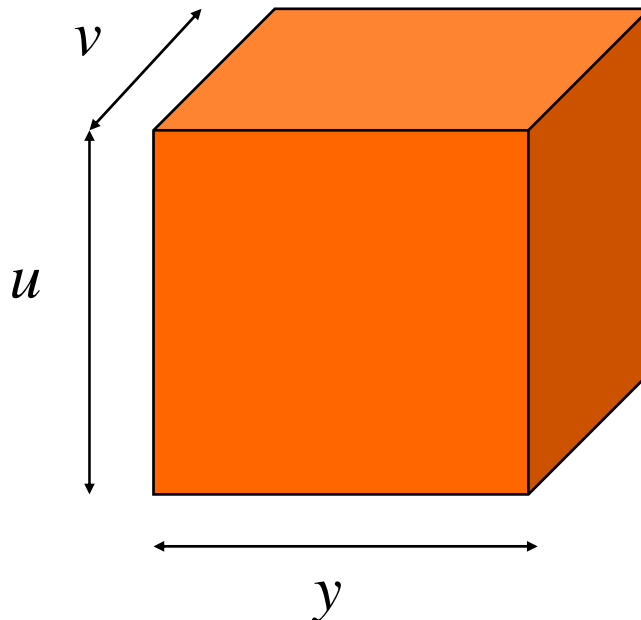
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Classifier for One Color



• Simplification for *Color_Orange*

• *If*

- $Min_{orange,y} \leq y \text{ AND } y \leq Max_{orange,y}$

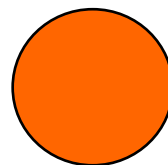
• *AND*

- $Min_{orange,u} \leq u \text{ AND } u \leq Max_{orange,u}$

• *AND*

- $Min_{orange,v} \leq v \text{ AND } v \leq Max_{orange,v}$

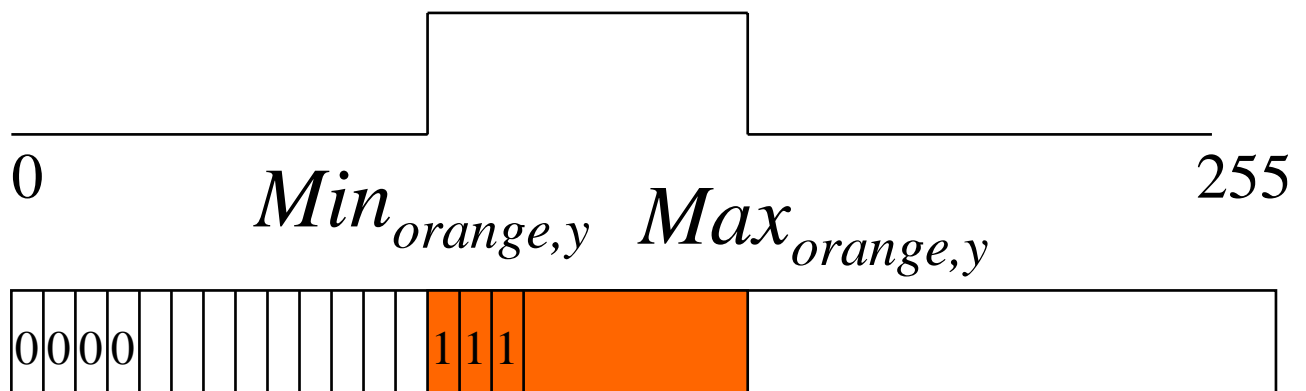
• *Then ORANGE*



The bit-Map for a color projection

► *If*

- $Min_{orange,y} \leq y \text{ AND } y \leq Max_{orange,y}$

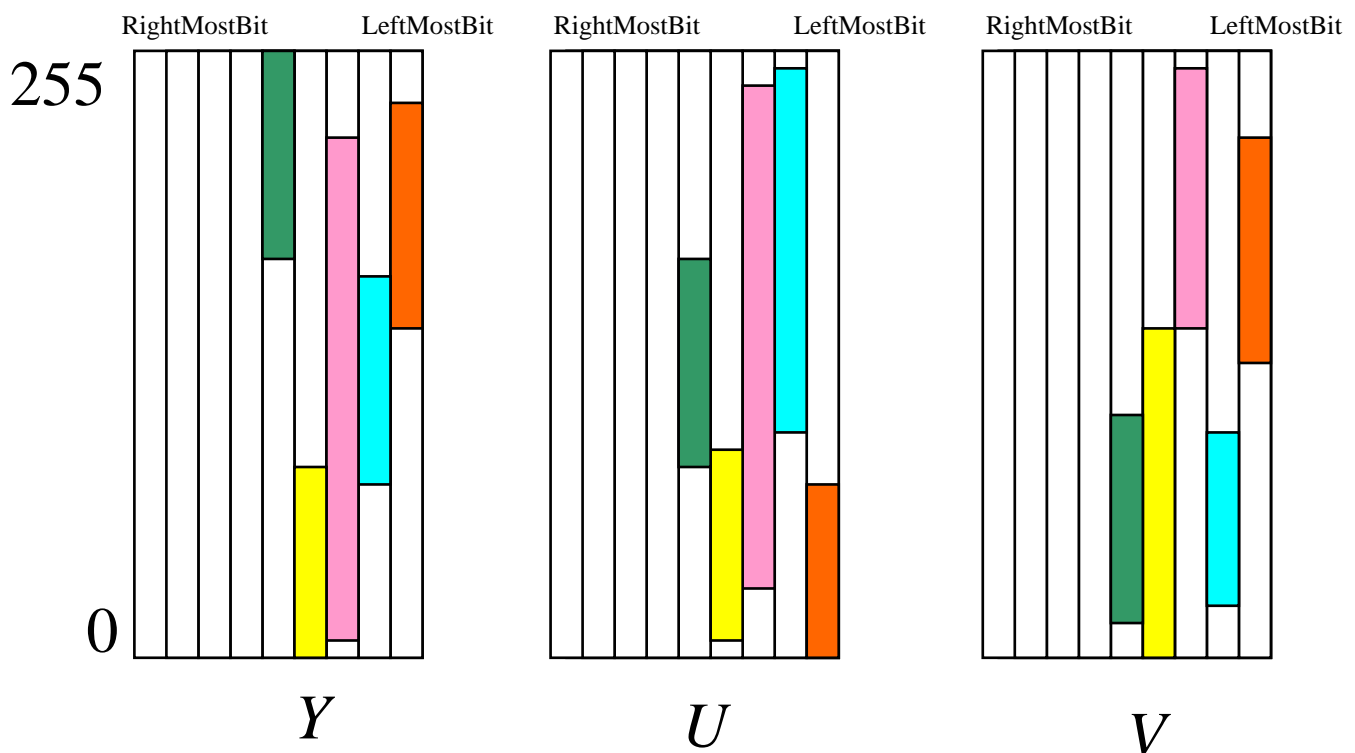


*A classifier is a table
look up*

► C/C++ bit-wise AND operation

`Color_Orange=Y[y]&U[u]&V[v]`

A Decision List is a Scan through the Bits of a Memory Word

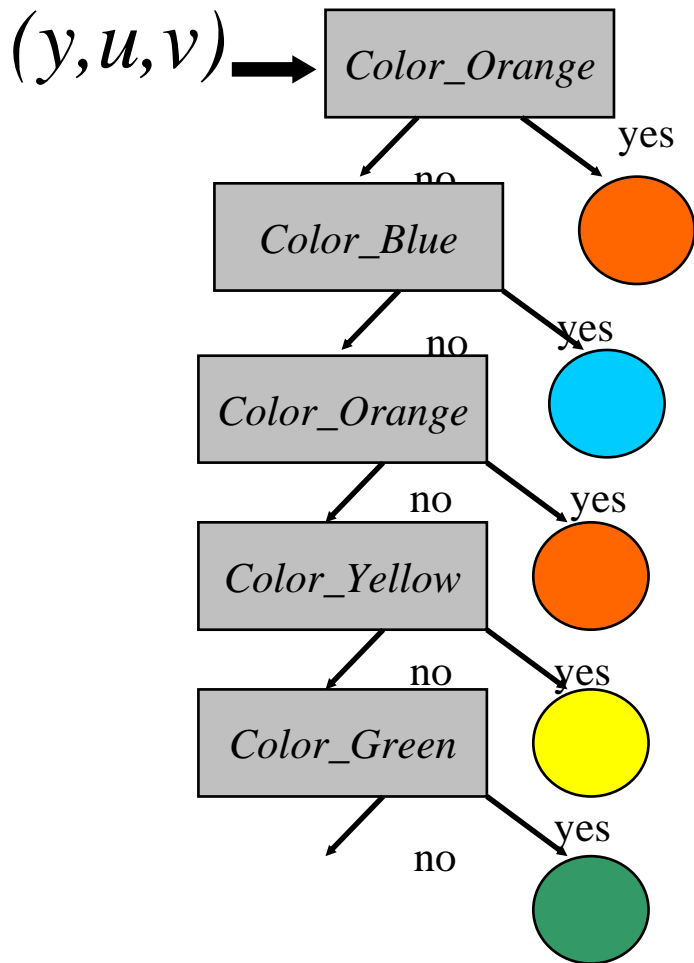


```

color=Y[y]&U[u]&V[v]
color_id=0
while (color & 1 == 0)
    { color>>=1; color_id++}

```

List Can Repeat Simple Classifiers





The RoboCup Competition



- ▶ Dynamic environment
- ▶ Very rapid analysis of images
- ▶ Legged-league
 - SONY Aibo robots
 - Image is 174x144 pixels
 - 25 frames a second
 - YUV format (3 bytes per pixel)

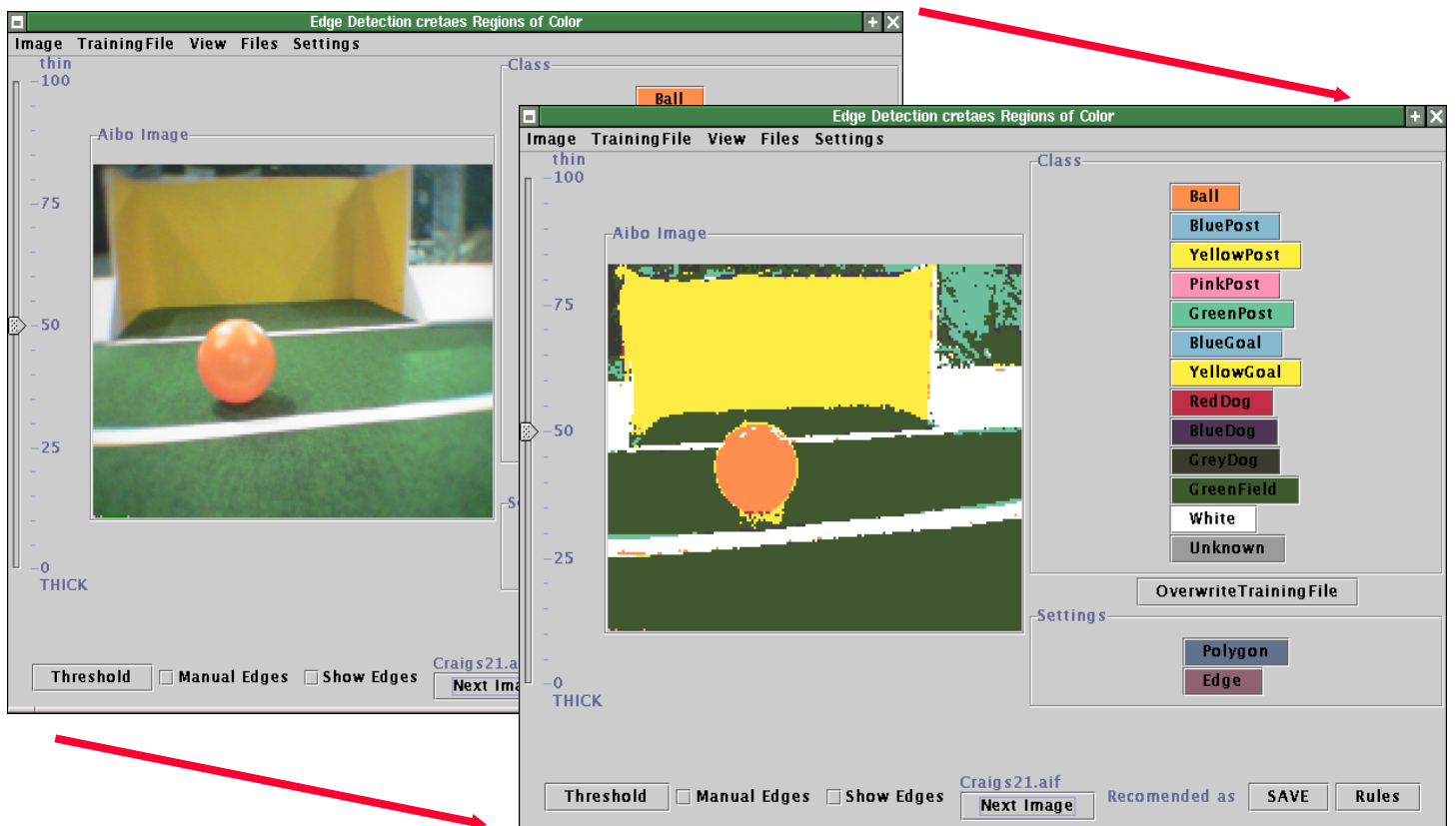
Image Segmentation

- Mapping is a classifier

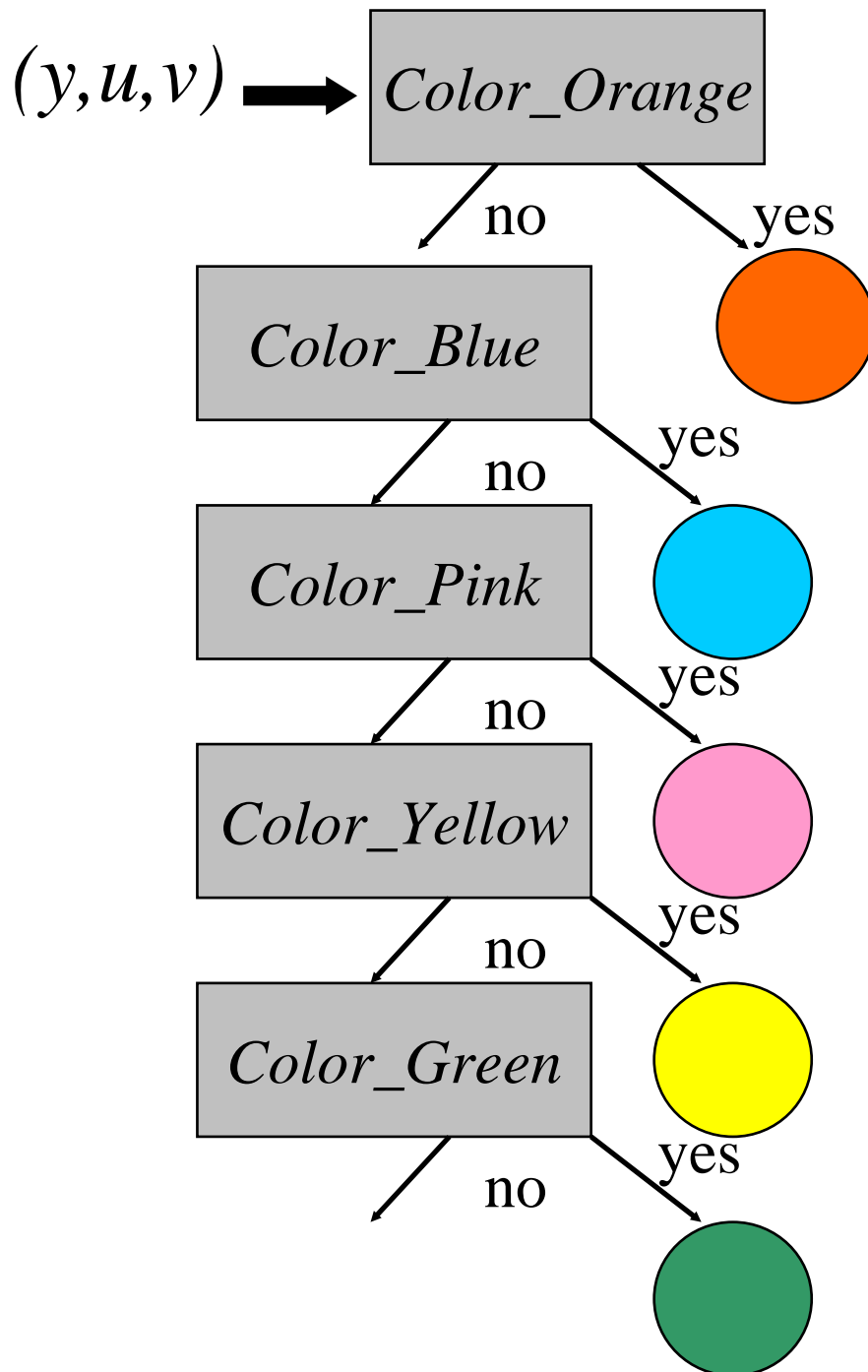
$Color_class: YxUxV \rightarrow Color$

$Color_Class(y,u,v) = Orange$

$|Y| \times |U| \times |V| = 256^3$



Decision List



Much faster than Artificial Neural Networks (ANN) or other classifiers
k- Nearest neighbors (k-NN) and
Decision Trees

- ▶ ANN using snns were 20K times slower
- ▶ *k*-NN with Quadtrees and DT (Weka) were 2K times slower

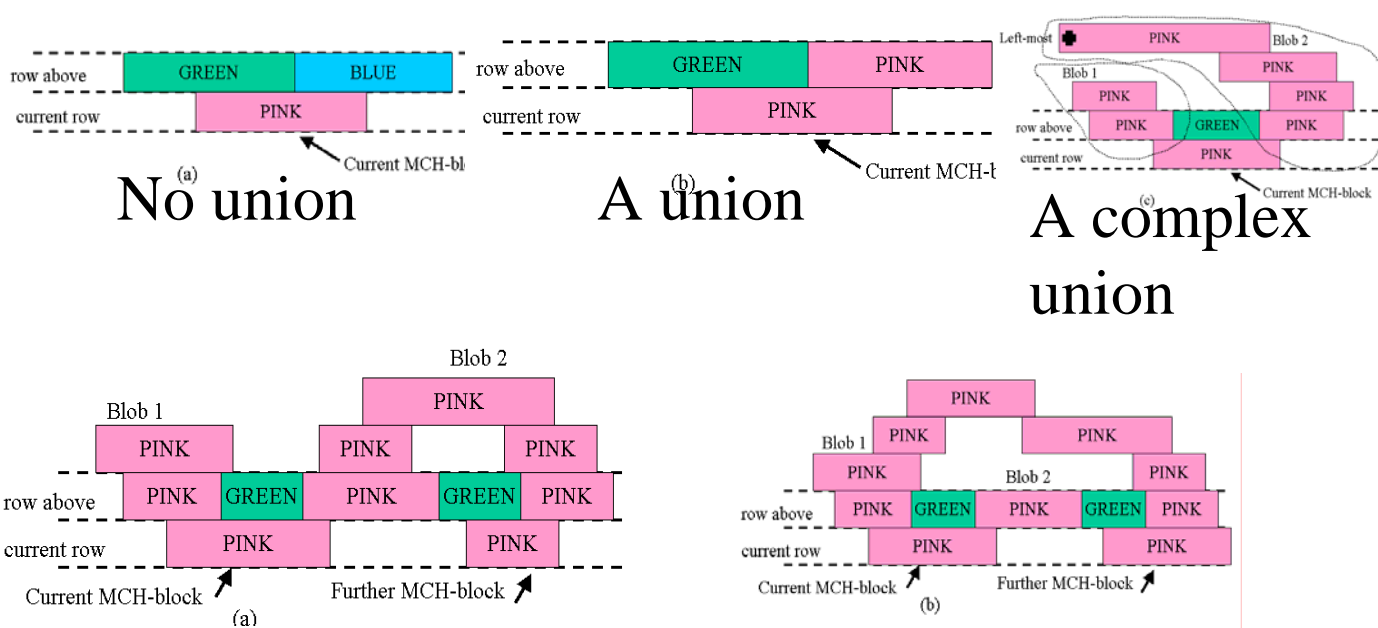
	DL	Look-up Table	Ratio
Maximum	2.87ms	2.46ms	1.16
Average	2.33ms	1.41ms	1.65
Minimum	2.08ms	1.27ms	1.63

Accuracy with Decision List is Marginally better

- ▶ ANN using snns were **20K** times slower
- ▶ k -NN with Quadrees and DT (Weka) were **2K** times slower

Algorithm	10-fold accuracy	Lowest accuracy per class	Largest 2-class confusion	size	Learning time	Test set accuracy
PART	99.0%	96% (yellow goal)	10 blue dog Vs gray dog	26 Rules	1.15s	99.3%
k -NN	99.3%	97% (blue dog)	8 red dog Vs gray dog	$k=3$ 6,226 Instances	0s	99.7%
DT	98.8%	95% (yellow goal)	10 red dog Vs gray dog	34 leaves 67 nodes	1.27s	99.6%
Look-up Table	71.6%	64% (yellow goal)	45 yellow goal vs orange ball	11 rules	manual	68.2%

Blob Forming – Bringing contiguous blocks together



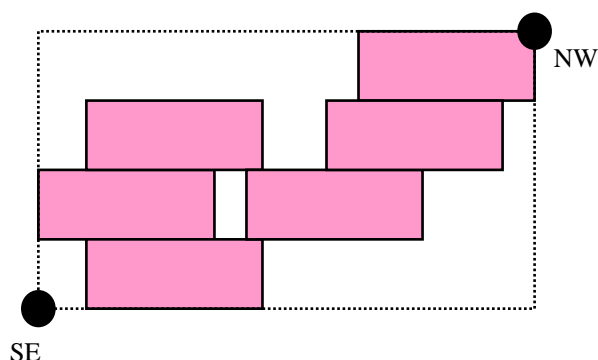
Merging needs to unify blob ids exposed in the row above

- We avoid UNION-FINF algorithms
- A data structure to process a row as we create run-length encoding

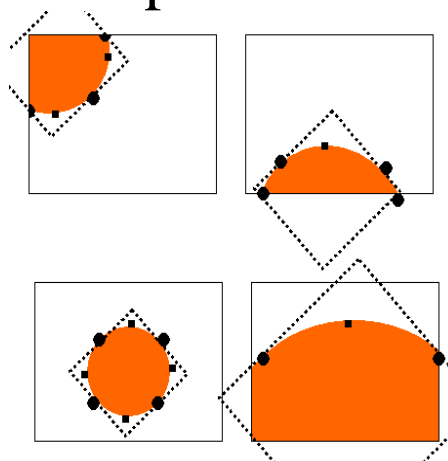
Blob information

Minimum bounding box

SE & NW (4 integers)



Special case



Data Structure for Current row

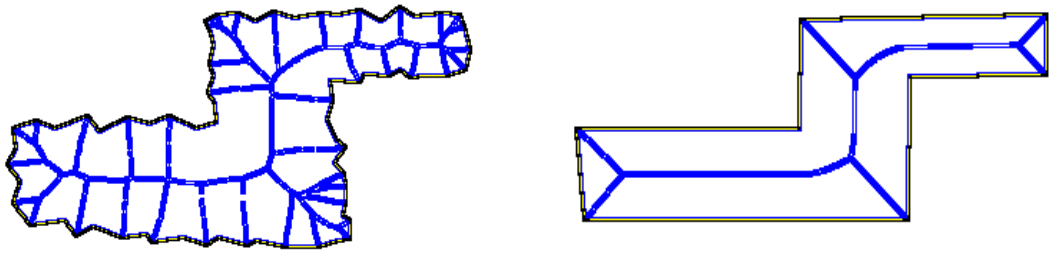
A sorted list of blobs exposed,
sorted by x-coordinate of left
most pixel exposed in the
current row

Processing a row is like merging
sorted lists

Few blobs per line

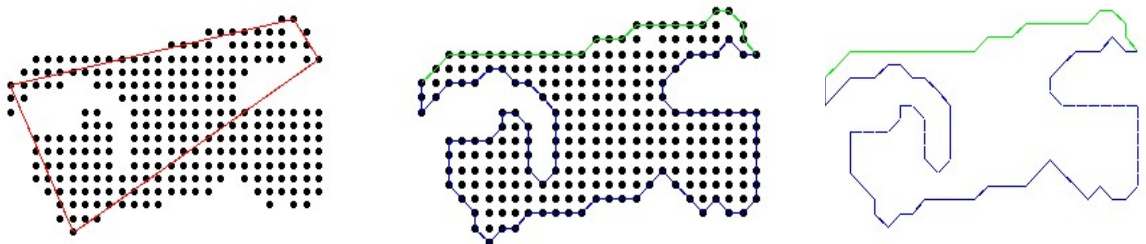
Shape Finding

- We need a better boundary to use the skeleton (medial axis)



Sinuosity in boundary produces complex skeleton

- We need to find the boundary fast



Finding the boundary

► Trace it

- Time proportional to the number b of pixels in boundary

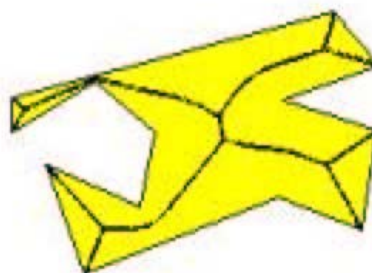
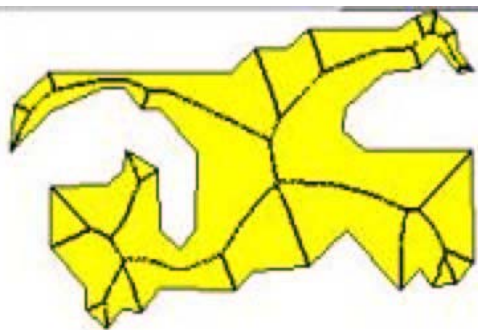
Simplify the boundary

- Douglas-Peucker Algorithm
- Time proportional to $b \log b$

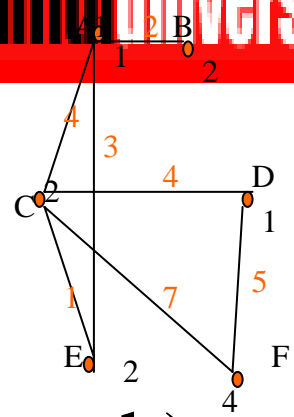
Find the skeleton

► Use Voronoi diagram of boundary segments

- Time proportional to high level description of boundary
 - (less than b)

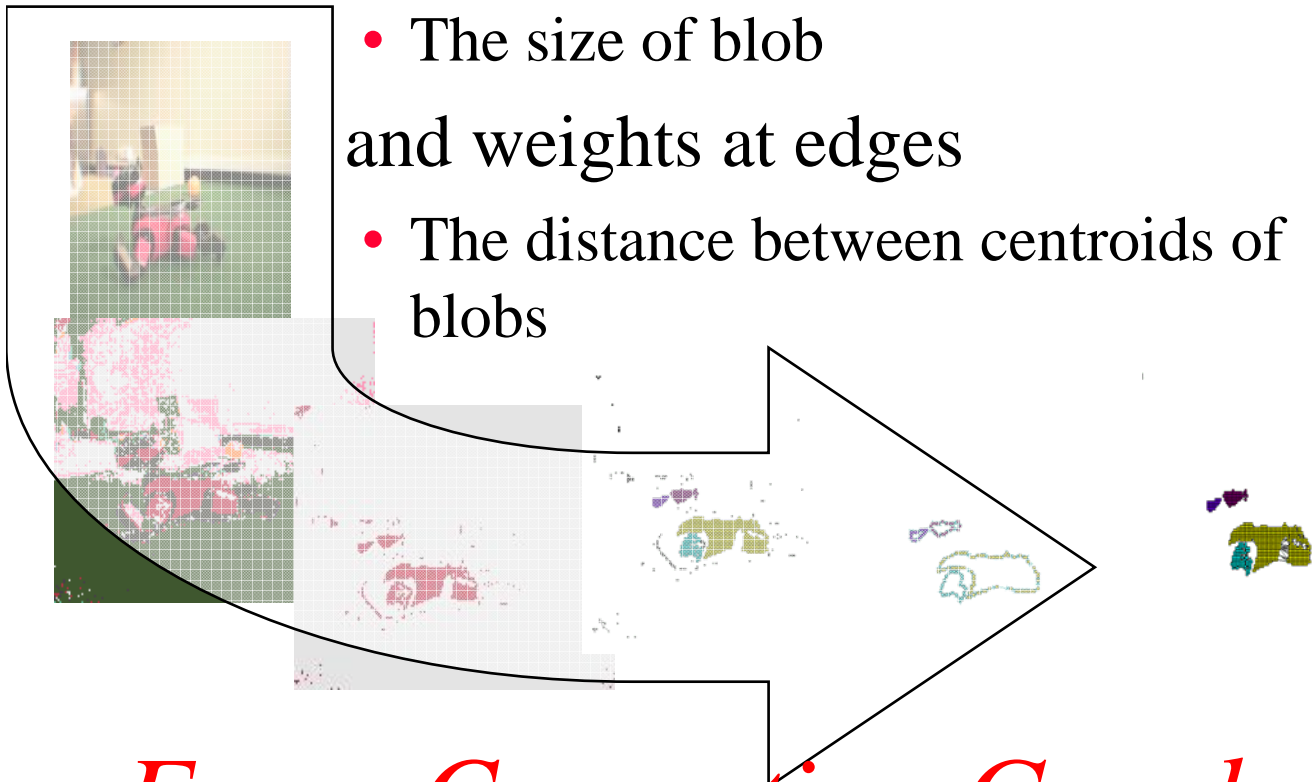


Attribute Oriented Graph



- ▶ Nodes and Edges (as a graph)
 - Weights at nodes

- The size of blob and weights at edges
- The distance between centroids of blobs



Error Correcting Graph

Isomorphism

Input: Two attributed graphs

Output:

A permutation of the nodes that minimizes total dissimilarity

Limitations of the current attribute graph approach

The robot we see

- ▶ *must not be far*
(the length of the field)
- ▶ *must not be too close*
(we must see almost all parts)
- ▶ *must be by itself*



A difficult situation for identifying the red dog in the back