

Blurred Shape Models

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Keypoints
detection

- Edges, Valleys, Contours, etc.
- Grey-level Saliency (Kadir)
- Harris
- Covariance Affine Regions (Schmid)
- Complex Salient Regions (Escalera)

Description

- SIFT (Lowe)
- Feature histograms (Buhmann)
- Gabor, Haar-like features (Viola)
- Disassociated dipoles (Baró)
- Boosting Context (Amores)
- Blurred Shape Model (Escalera)**

Landmark
selection

- Mutual Information (Ullmann)
- Boosting (Schapire)
- Clustering_{NDA, LDA, MDS, ND}

Object model

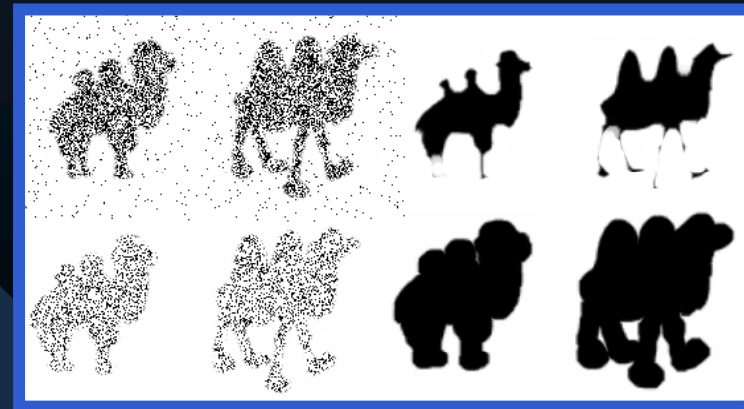
- Descriptors array (Torralba, Lowe, Ullmann)
- Probabilistic models (Buhmann, Fergus, Fei-Fei, Torralba)
- Graphical Models (Bayessian networks, Gaussian Mixture Models)
- Boosted Landmarks (Escalera)

Multi-class binary object classification

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- Rotation,
- Partial occlusions,
- Noise,
- Rigid and elastic deformations,
- Intra-class and inter-class variabilities,
- ...

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Algorithm

Given a binary image I ,
Obtain the *shape* S contained in I

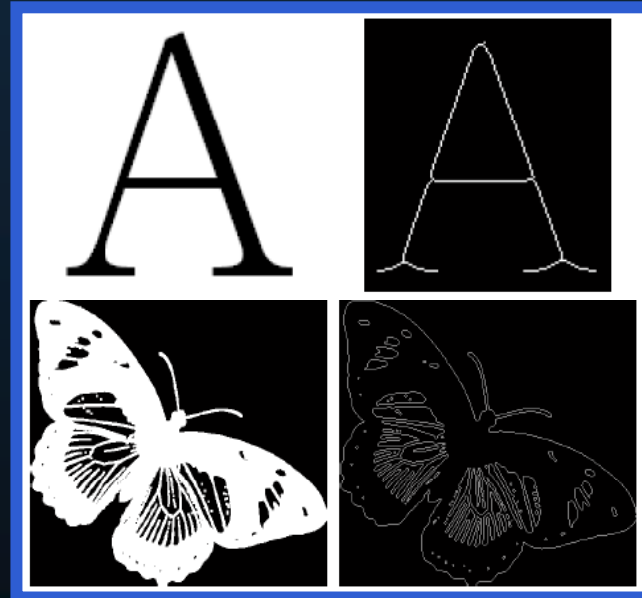
Object Shape

Grid

Points distances

Updating
histogram

BSM PDF



- Determine shape:
- Contour map
- Skeleton

...

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Algorithm

Object Shape

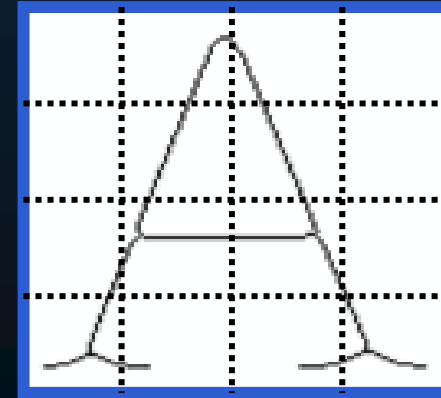
Grid

Points distances

Updating histogram

BSM PDF

Divide I in $n \times n$ equal size sub-regions $R = \{r_1, \dots, r_{n \times n}\}$, with c_i the center of coordinates for each region r_i .
Let $N(r_i)$ be the neighbor regions of region r_i , defined as $N(r_i) = \{r_k | r \in R, \|c_k - c_i\|^2 \leq 2 \times g^2\}$, where g is the cell size.



• $n \times n$ grid size

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Algorithm

Object Shape

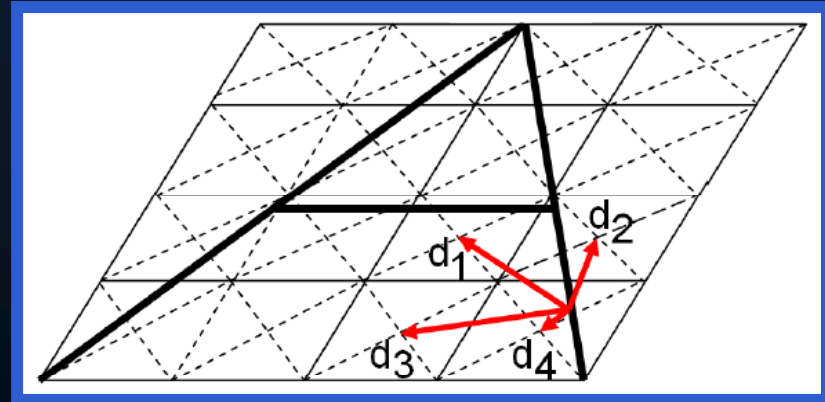
Grid

Points distances

Updating histogram

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```
For each point  $\mathbf{x} \in S$ ,  
  For each  $r_i \in N(r_{\mathbf{x}})$ ,  
     $d_i = d(\mathbf{x}, r_i) = \|\mathbf{x} - c_i\|^2$   
  End_For
```



- Distances to the nearest centroids

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Algorithm

Object Shape

Grid

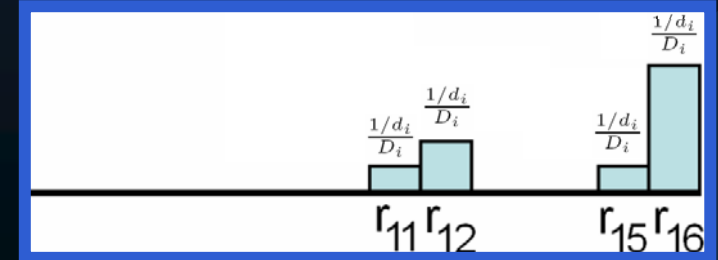
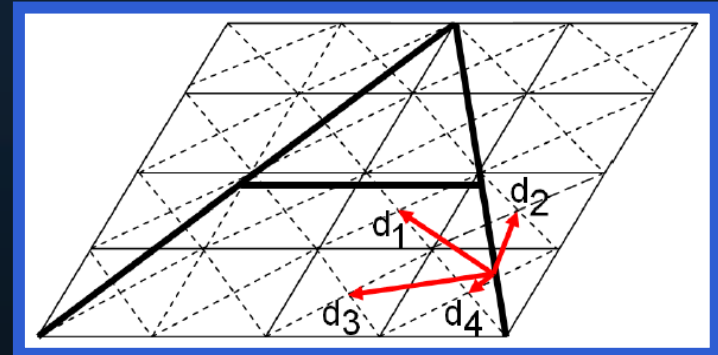
Points distances

Updating histogram

BSM PDF

Update the probabilities vector v positions as:
$$v(r_i) = v(r_i) + \frac{1/d_i}{D_i}, \quad D_i = \sum_{c_k \in N(r_i)} \frac{1}{\|\mathbf{x} - c_k\|^2}$$

End_For

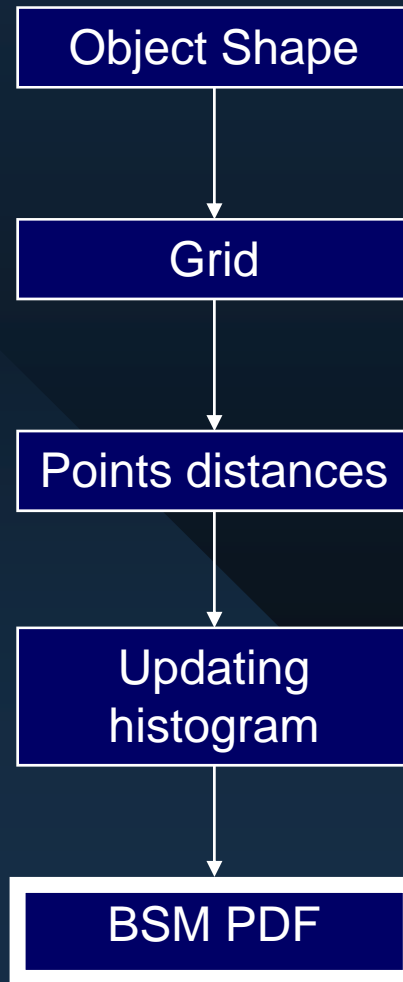


• Spatial distribution of points

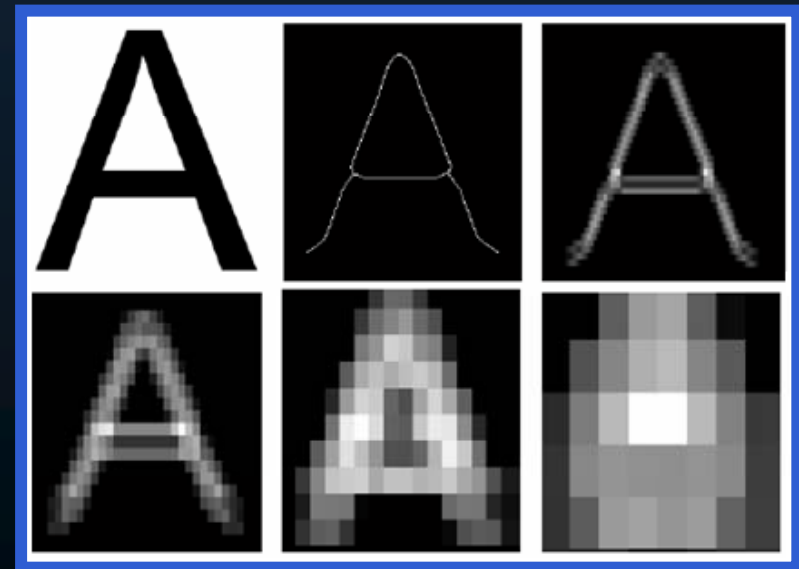
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Algorithm



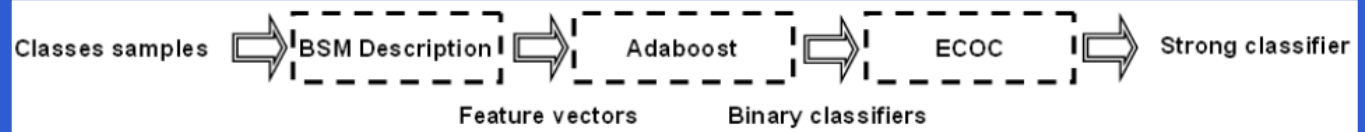
Normalize the vector v as: $v = \frac{v(i)}{\sum_{j=1}^{n^2} v(j)} \forall i \in [1, \dots, n^2]$



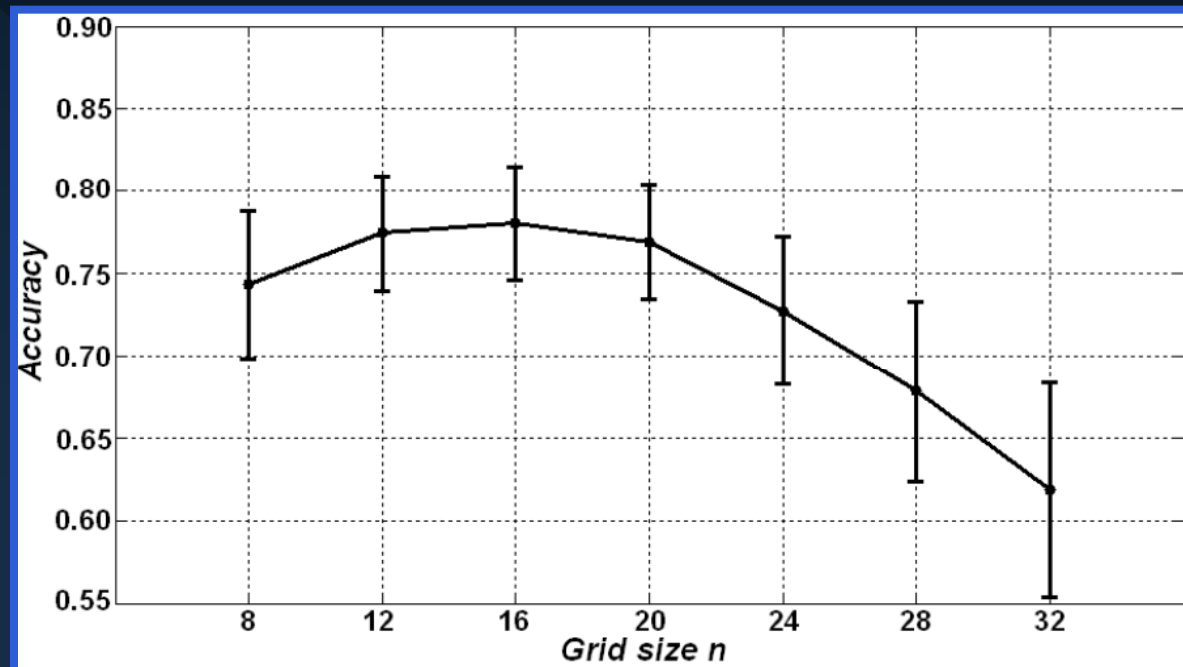
•BSM for different grid sizes

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- The method is scaling and stretching invariant.
 - The BSM descriptor is robust against rigid and elastic deformations.
- Cross-validation → optimum grid size



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ECOC CODING

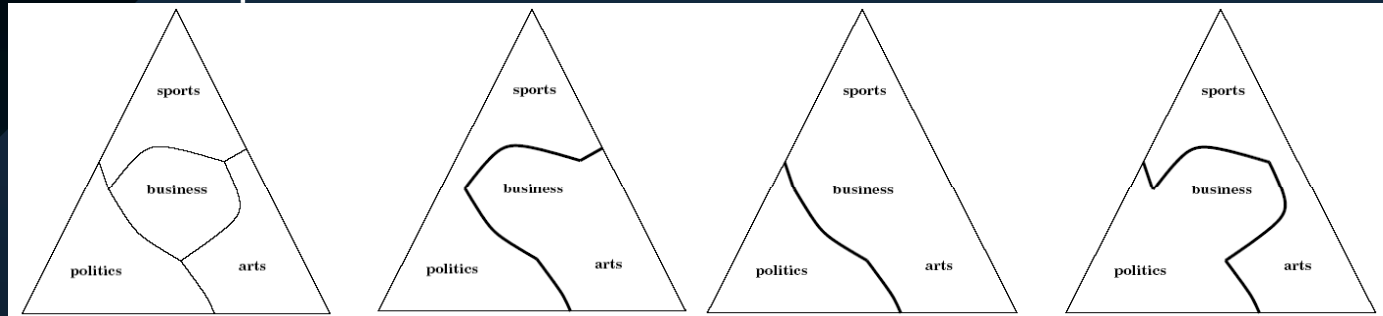
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Example

Classifier 1

Classifier 2

Classifier 3



C1= sports
C2=business
C3=politics
C4=arts

1
-1
1
-1

1
1
-1
1

1
-1
-1
1

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C1= sports
C2=business
C3=politics
C4=arts

coding matrix

1	1	1
-1	1	-1
1	-1	-1
-1	1	1

Coding

One-vs-one

One-vs-all

Dense Random

Sparse Random

Decoding: Given a test sample, we obtain a code according to the output of each classifier and find the "closest" code.

-1 1 1



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ECOC DECODING

Decoding

Hamming decoding

Euclidean decoding

Loss-based decoding

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1 versus All

Code length: N_c

1	-1	-1
-1	1	-1
-1	-1	1

Random Dense ECOC

Code length: $10 \log N_c$

1	-1	1
-1	1	-1
1	-1	-1

Three symbol codes

1 versus 1: "All pairs"

Code length: $N_c (N_c - 1) / 2$

1	1	0
-1	0	1
0	-1	-1

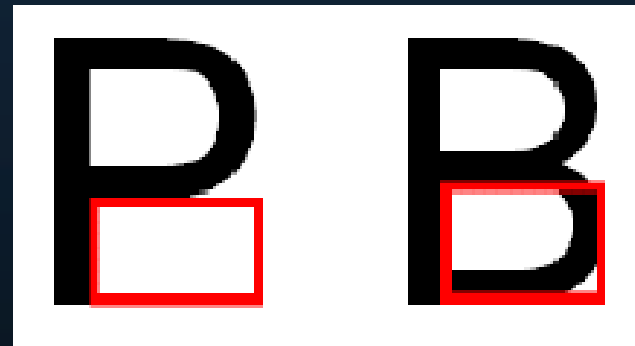
Random Sparse ECOC

Code length: $15 \log N_c$

1	0	-1
-1	1	0
0	-1	1

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- Adaboost learns difficult classes which may share features.

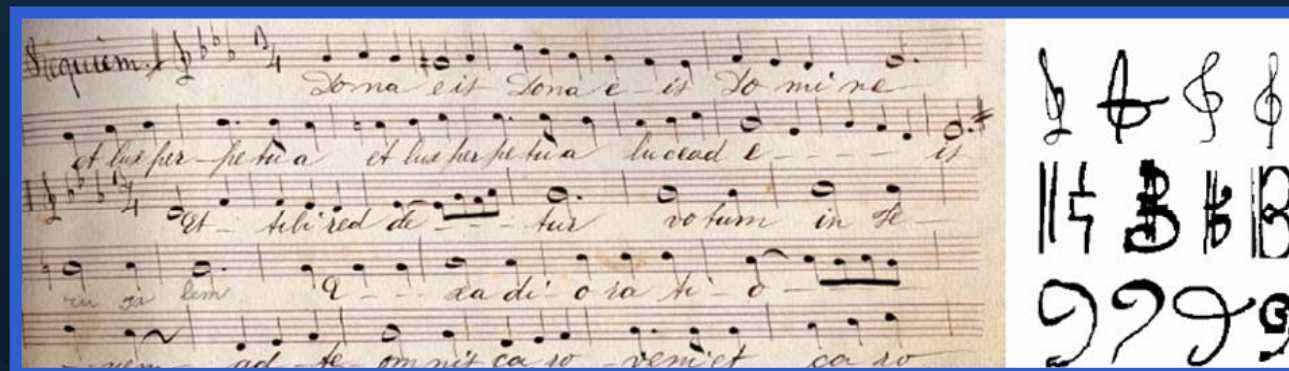


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- The multi-class ECOC framework can correct possible classification errors produced by the binary classifiers.
- When the classifiers are trained only few features are selected
→ Fast and suitable for real-time categorization problems.

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Evaluation data



- Clefs database: collection of modern and old musical scores (19th century) of the Archive of the Seminar of Barcelona. The database contains a total of 2128 samples between the three different types of clefs from 24 different authors.

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- Architectural hand-drawn symbols database: 2762 total samples organized in 14 classes. Each class consists of an average of 200 samples drawn by 13 different authors.

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Comparatives

DESCRIPTORS

Zoning	16x16 grid size
BSM	16x16 grid size
ART	Radial order with value 2 and angular order 10 with value 11
Zernique	7 moments

Measurements

50 runs Discrete Adaboost with Decision Stumps

Two optimal trees of Forest-ECOC with Beta-Density Decoding.

Classification score: stratified ten-fold cross-validation

With two-tailed t-test at 95% of the confidence interval

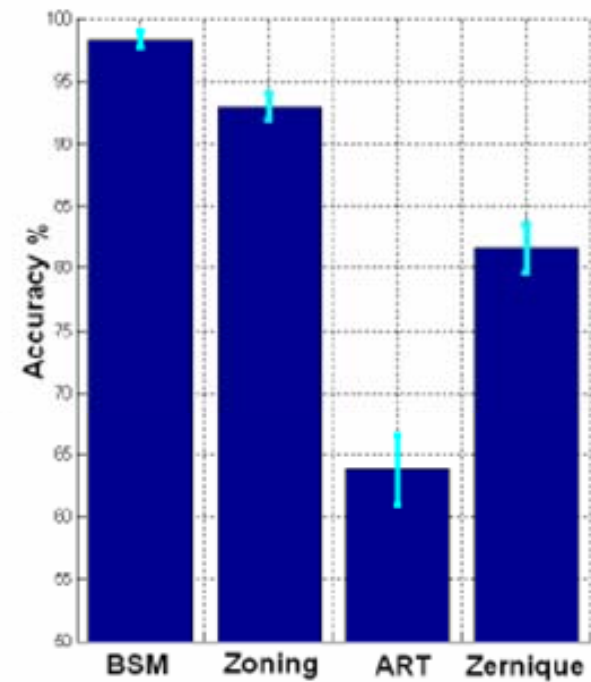
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Evaluation

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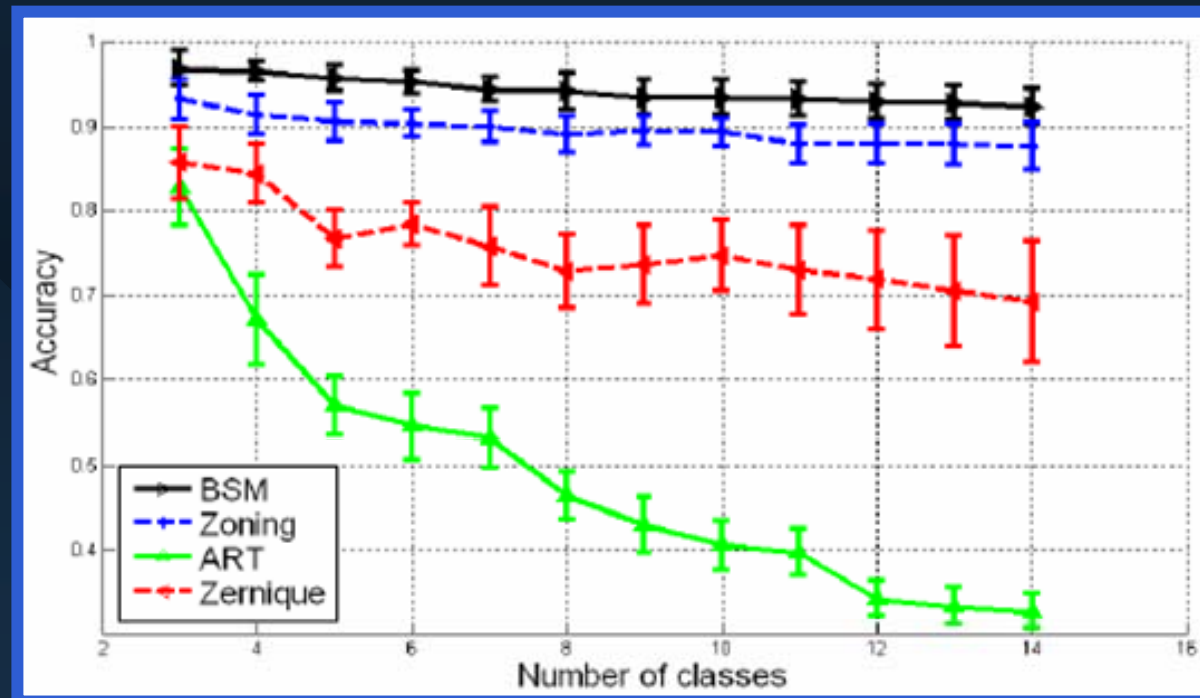
Method	Accuracy
BSM	98.11 ± 0.59
Zoning	92.45 ± 1.06
ART	63.75 ± 2.76
Zernike	81.50 ± 1.97



Clefs database

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Evaluation



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Architectural hand-drawn symbols database

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Conclusions

→ Robust to noise, occlusions, rigid and elastic deformations and inter-class and intra-class variabilities.

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