

Optimizing color information processing inside an SVM network

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Outline

- 1 Motivations
 - Problem
 - Used features
 - The SVM Network
- 2 How taking into account color information
 - Color space
 - Design proposal for SVM Network
- 3 Experimental results

Color Features Extraction

How can we extract the best descriptor based on the HOG descriptor when there are 3 channels (color image)?

- 1 Transform the color image input into grey level
- 2 Extract HOG descriptor on each channel and concatenate them
- 3 Only take into account the highest gradient
- 4 Reduce the dimension (PCA...)

HOG and SVM Network

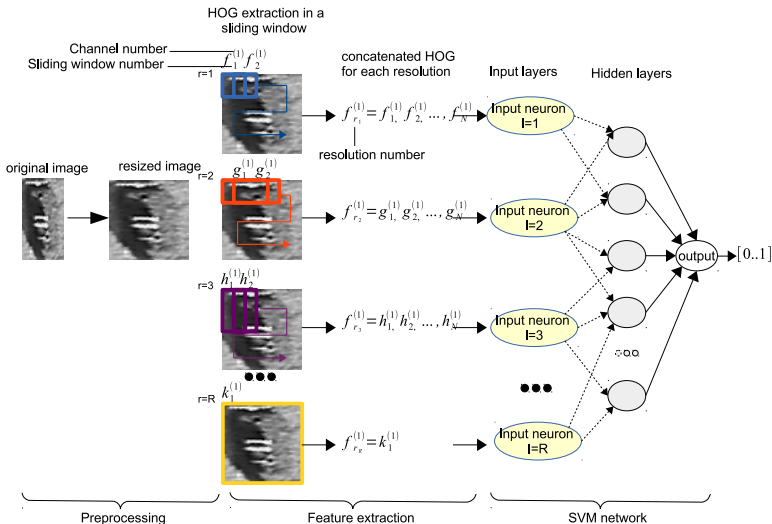


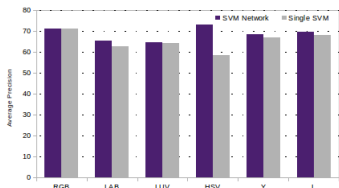
Figure : SVM Network works on grey level images.

Color space

What is the best discriminant color space?

- RGB
- $Y = 0.21R + 0.71G + 0.07B$
- CIE – LAB and CIE – LUV (based on human perception)
- HSV (cylindrical-coordinate representations of RGB)
- Use a PCA transformation

Experimental results

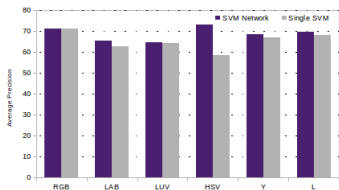


SVM Network vs SVM Single :

- SVM Network always improves the precision
- The performance from different color spaces are very closed.
- HSV and RGB : the 2 bests

⇒ But the colors are threaten separately in the SVM Network...

Experimental results



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Design the network

1- Fusion of channels per resolution

Concatenate the different channels from the same resolution.

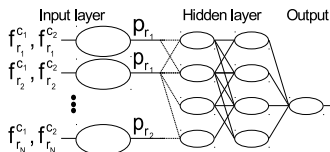


Figure : Representation of the Fusion Network

Different channels are not necessarily in the same feature space,
e.g. : HSV.

=> Normalisation problem

Design the network

2- A specific function to merge the channels per resolution

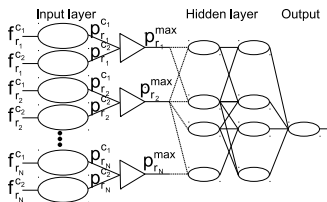


Figure : Representation of the Maximum Network

=> The quantification may lose important information.

Connect all input neurons with the same resolution r to a specific function :

- No focus on the low variations (Maximum function)
- Linear quantification (Product function)
- The first principal component (PCA transformation)

Design the network

3- Stacking of channels per resolution

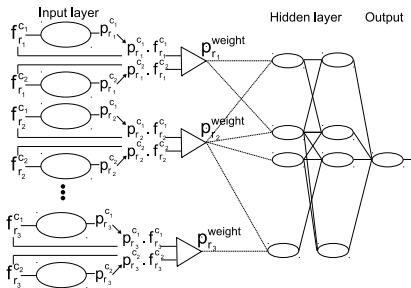


Figure : Representation of the Stack Network

- Features from different channels are scaled and used in the same neuron
- Each weight is learned by using an SVM

Results

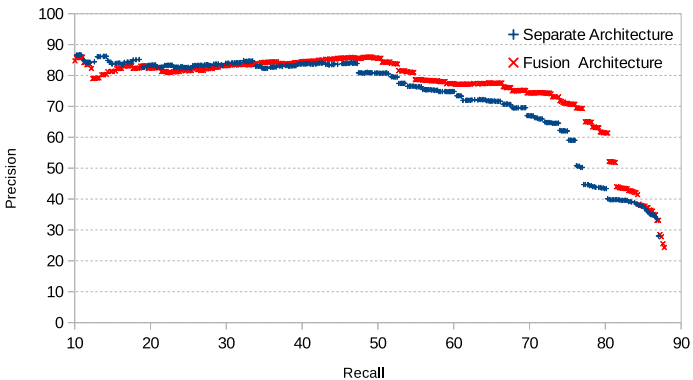


Figure : ROC curves for the separate and fusion architectures

Results

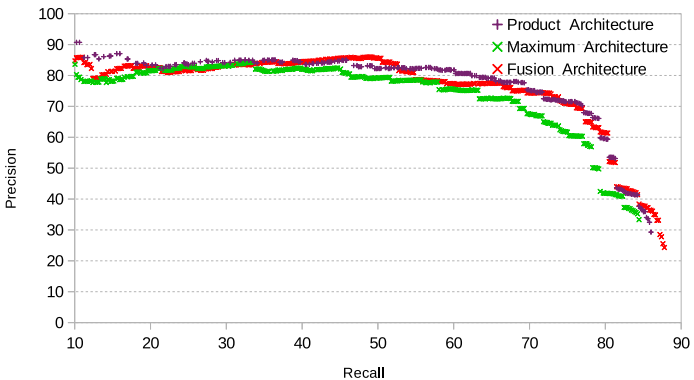


Figure : ROC curves for the fusion, product and maximum architectures

Results

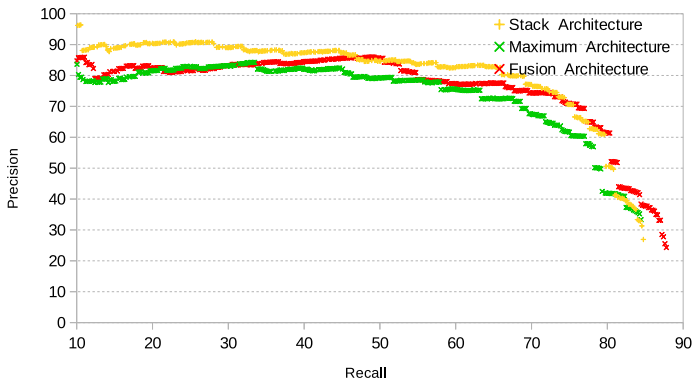


Figure : ROC curves for the fusion, maximum and stack architectures

Overview of the Network performance

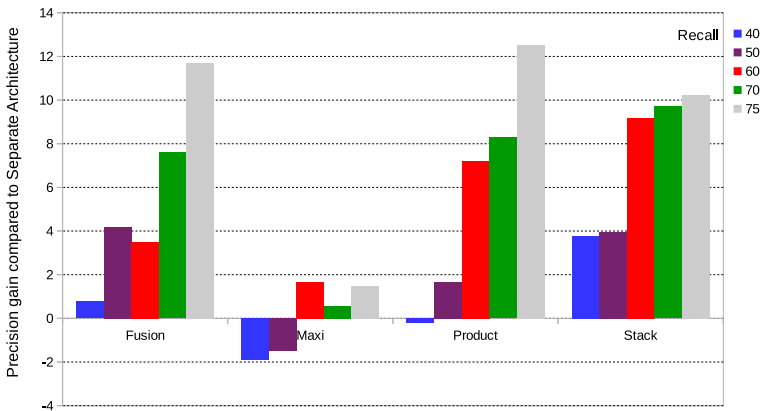


Figure : Precision gain is compared to the separate architecture.

Summary

Conclusions

- SVM Network outperforms SVM by an average precision gain ranging from 1.5% to 6%.
- SVM Network design (e.g. Stack Design) would improve to 10% the precision.

Future works

- Design an SVM Network to combine several feature types : SURF, SIFT...
- Design a CNN input neurons.

Thanks!



The best Parameters for SVM Network

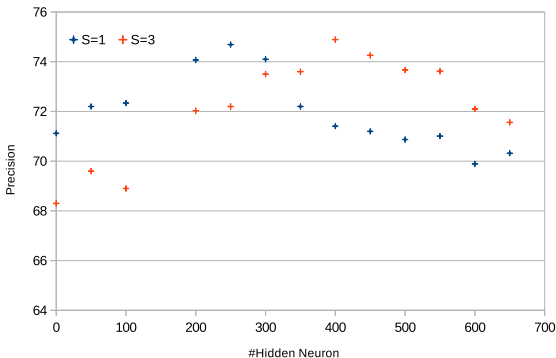


Figure : Tuning of the SVM Network on a validation database

Before each experiment the SVM network used a validation database to fix the best number of hidden and random neurons.