Optimizing color information processing inside an SVM network

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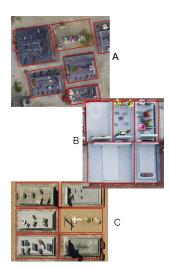
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Outline

- Motivations
 - Problem
 - Used features
 - The SVM Network
- 2 How taking into account color information
 - Color space
 - Design proposal for SVM Network
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Urban Object Detection In Aerial Images



Urban object definition:

- The appearance of the objects varies : color, size, orientation...
- Multiple distortions and occlusions due to shadows, vegetation..
- Multiple-object detection

About the databases of aerial images :

- 19 images for the training database
- 3 images for the validation database
- 2 images for the testing database

Features Extraction

Motivations

Extraction of multiple Histograms of Oriented Gradients

- Calculate HOG within cells and blocks
- Accumulate features to construct HOG descriptor of different sizes



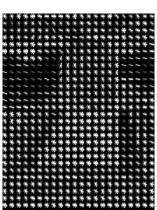


Figure: HOG representation for 9x9 cells size.



Color Features Extraction

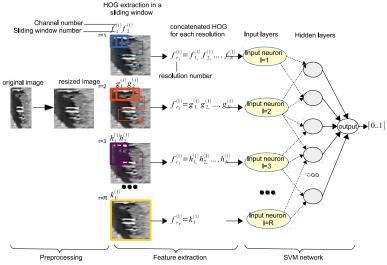
How can we extract the best descriptor based on the HOG descriptor when there are 3 channels (color image)?

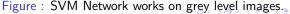
- Transform the color image input into grey level
- Extract HOG descriptor on each channel and concatenate them
- Only take into account the highest gradient
- Reduce the dimension (PCA...)

 Motivations
 How taking into account color information
 Experimental results
 Summary
 Appendix

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HOG and SVM Network







HOG and SVM Network

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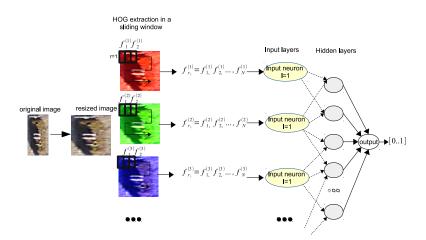


Figure: SVM Network works on color images called *separate architecture*.

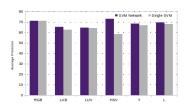


Color space

What is the best discriminant color space?

- RGB
- \bullet 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

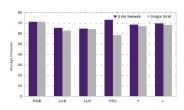


SVM Network vs SVM Single:

- 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



SVM Network vs SVM Single:

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=> But the colors are threaten separately in the SVM Network...

Design the network

1- Fusion of channels per resolution

Concatenate the different channels from the same resolution.

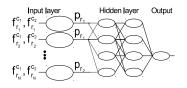


Figure: Representation of the Fusion Network

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

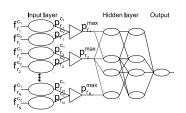


Figure : Representation of the Maximum Network

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)

=> The quantification may lose important information.

Design the network

3- Stacking of channels per resolution

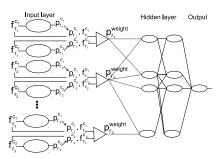


Figure: Representation of the Stack Network

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

Results

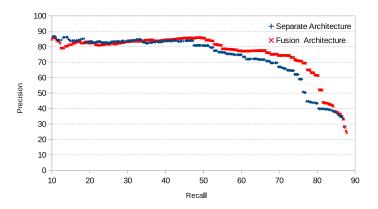


Figure: ROC curves for the separate and fusion architectures

Results

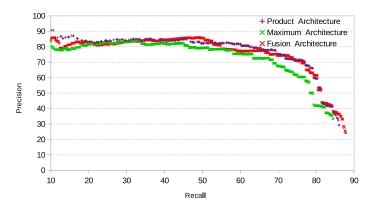


Figure: ROC curves for the fusion, product and maximum architectures

Results

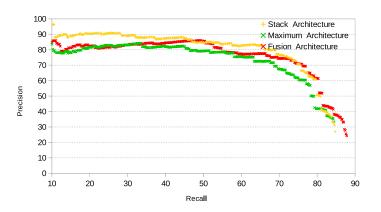


Figure: ROC curves for the fusion, maximum and stack architectures

Overview of the Network performance

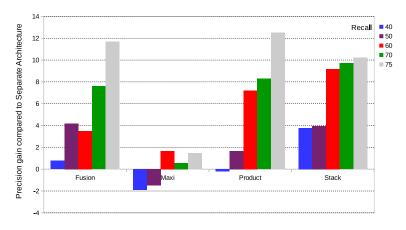


Figure: Precision gain is compared to the separate architecture.

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!





How taking into account color information Experimental results Summary Appendix

The best Parameters for SVM Network

Motivations

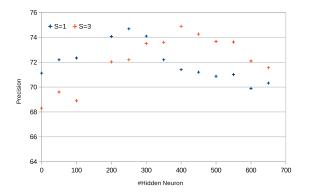


Figure: Tuning of the SVM Network on a validation database

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