COSFIRE

A trainable features approach to pattern recognition

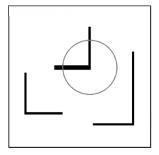




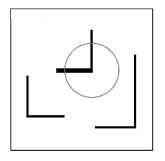
George Azzopardi

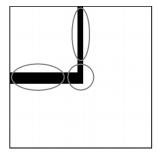
Nicolai Petkov

University of Groningen The Netherlands Select a pattern of interest



- A filter is automatically configured
 - Sub-units: orientation-selective detectors

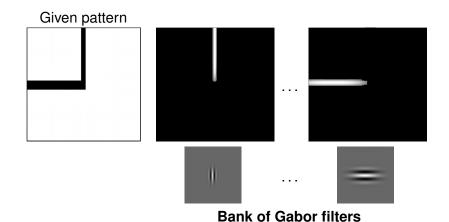




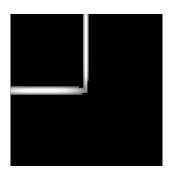
Conclusions

Step 1: detect oriented stimuli

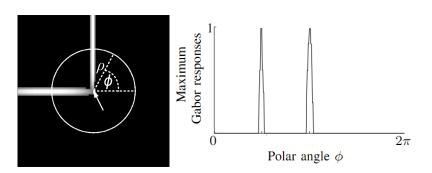
2D Gabor filters



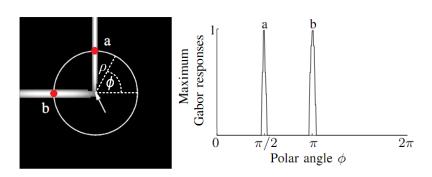
How? Consider a superposition of Gabor responses



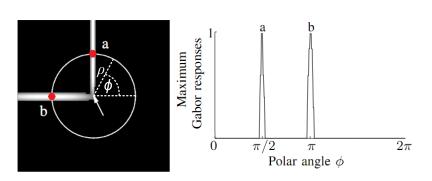
• Consider Gabor responses along a circle of a given radius



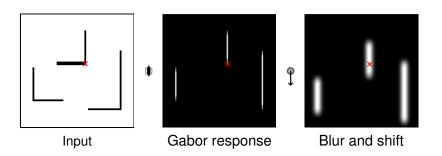
Identify significant local maxima



- Result of analysis
 - (λ, θ) line properties
 - (ρ, ϕ) position w.r.t. filter center

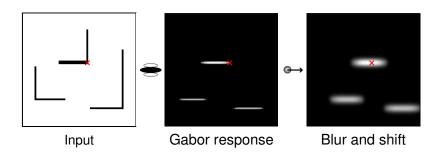


Step 3: Response of Sub-unit 1



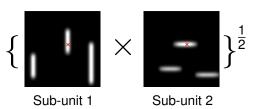
• If this filter responds strongly in the center point this means: "There is a (nearly) vertical line at (or near) point (ρ, ϕ) from the center of the filter."

Step 3: Response of Sub-unit 2

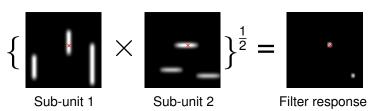


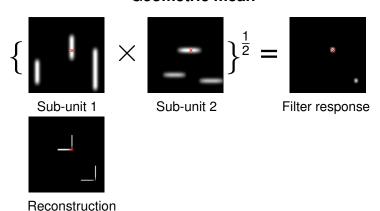
• If this filter responds strongly in the center point this means: "There is a (nearly) horizontal line at (or near) point (ρ, ϕ) from the center of the filter."

Step 4: Filter response Geometric mean

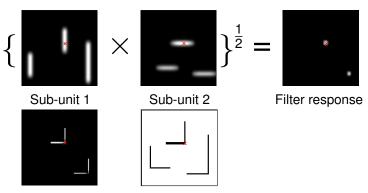


Step 4: Filter response Geometric mean



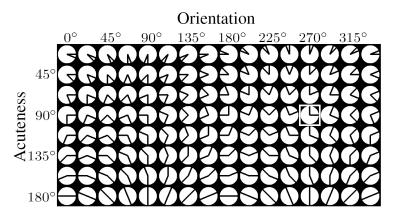


Step 4: Filter response Geometric mean



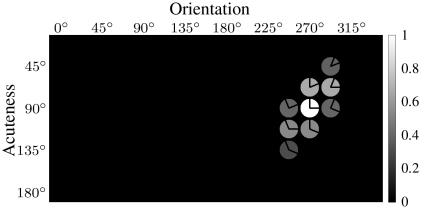
Reconstruction

 If the product of the two blurred and shifted Gabor filters is substantial in a given point, this means: "There is a (nearly) vertical line at a certain distance above and a (nearly) horizontal line at a certain distance to the left of the concerned point".

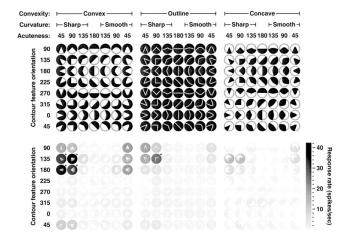


Set of elementary features used in neurophysiological study [Pasupathy and Connor, 1999]





Filter responses



[Pasupathy and Connor, 1999]



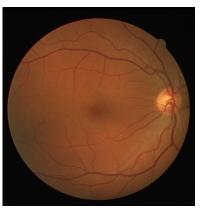
Summary of trainable feature detector

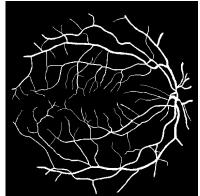
- COSFIRE filter (Combination Of Shifted Filter REsponses)
 - Gabor filters → blur and shift → multiply

- Trainable
 - can be configured with any given local contour pattern

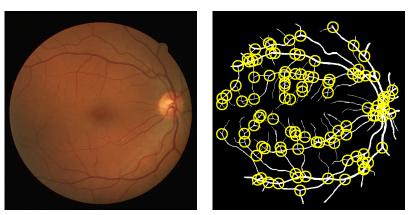
1. Detection of bifurcations in retinal fundus images

 The structure of the retinal vascular tree can reveal signs of cardiovascular diseases



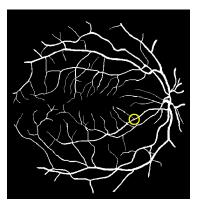


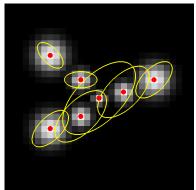
1. Detection of bifurcations in retinal fundus images



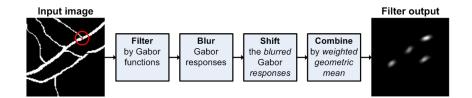
This image contains 101 bifurcations

Trainable filter configuration

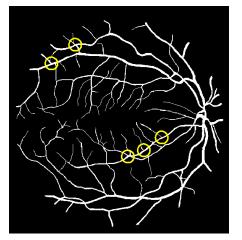




Apply the configured filter



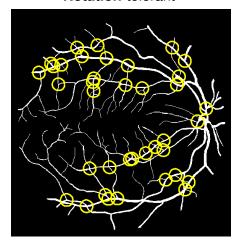
Non-tolerant



5 (out of 101) detected bifurcations



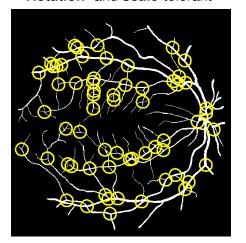
Rotation-tolerant



38 (out of 101) detected bifurcations



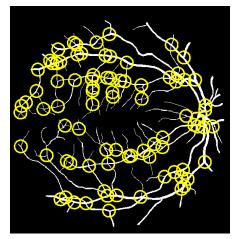
Rotation- and scale-tolerant



62 (out of 101) detected bifurcations



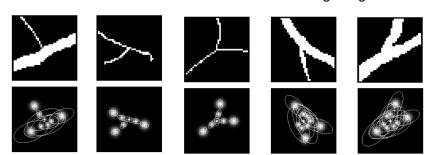
Rotation-, scale- and reflection-tolerant



85 (out of 101) detected bifurcations

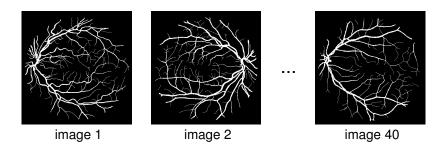
Training

100% Precision and 100% Recall for training image



Testing

DRIVE data set: (40 images, more than 5000 bifurcations) [Staal et al., 2004]



Results for the whole DRIVE data set **Recall**: 97.88% and **Precision**: 96.94%

http://matlabserver.cs.rug.nl/RetinalVascularBifurcations

Conclusions

Detection of vascular bifurcations in retinal fundus images using trainable COSFIRE filters N. Petkov, G. Azzopardi and Zhe Sun, University of Groningen, Department of Computing Science, Intelligent Systems

Other on-line image processing and computer vision applications are available at http://matlabserver.cs.rug.nl.

On this site you can: process a binary retinal image taken from the DRIVE data set [11] with some predefined COSFIRE filters, visualize the output of the filters rendered as a gray-level image, visualize the detected vascular bifurcations as encircled features in the input image, use the slider below the images to control the output of the COSFIRE filters; the number of detected features decreases with an increasing threshold value (i.e. moving the slider to the right), and download the output image with the detected features encircled as well as the coordinates of these detected features.

We are currently extending this application by including a facility to upload coloured/binary retinal images, and an algorithm for automatic vessel tree segmentation of the uploaded images. In the near future, we will also include detailed explanations on the parameters of the COSFIRE filters.



Ground truth data of vascular bifurcations in retinal images of the DRIVE database In http://www.cs.rug.nl/~imaging/databases/retina_databases/, we provide the (row.col) coordinates of the vascular bifurcations that were manually selected as ground truth data in 40 binary images of the

DRIVE data set



2. Handwritten digit recognition

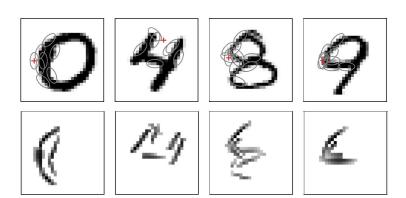
MNIST data set [LeCun et al., 1998]

60,000 training and 10,000 test digits



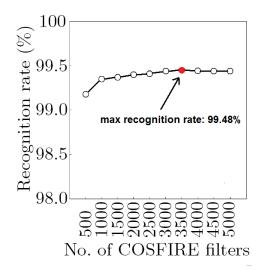
Configuration of COSFIRE filters

- Choose a subset of random images from each digit class
- Choose a random location in each selected digit image



Recognition rate: 99.48%

Conclusions





2. Handwritten digit recognition

Farsi data set: 99.33% [Khosravi and Kabir, 2007]

60,000 training and 20,000 test digits



3. Detection and recognition of traffic signs

RuG data set [Grigorescu and Petkov, 2003]







Configuration of COSFIRE filters















Data set of 48 grayscale images Detection and recognition rates: **100%**







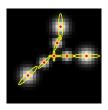


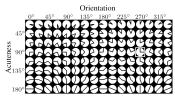


Conclusions

- COSFIRE filter
 - Inspired by shape-selective V4 neurons



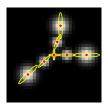


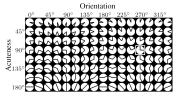


Conclusions

- COSFIRE filter
 - Inspired by shape-selective V4 neurons







- Simple implementation:
 - Gabor filters → blur and shift → multiply
 - http://matlabserver.cs.rug.nl/













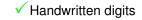
















√ Camera calibration

















√ Retinal bifurcations



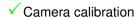


Signatures

Traffic signs



Architectural symbols



Outlook/Work in progress

- Extend it with color properties
- Extend it to 3D images and video streams
- Investigate it with non-visual time series data; e.g. sound
- Quantitative comparison with SIFT keypoint detector

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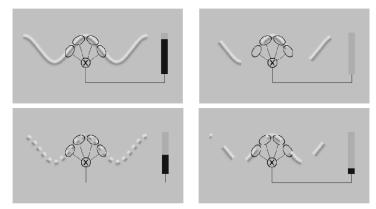
Publication



G. Azzopardi and N. Petkov, **Trainable COSFIRE filters for keypoint detection and pattern recognition**, *IEEE Transactions on Pattern Analysis and Machine Intelligence*, vol. 35(2), pp. 490-503, 2013

Psychophysics

Curved contour parts are detected by multiplication



[Gheorghiu and Kingdom, 2009]



References



Gheorghiu, Elena and Frederick A. A. Kingdom [2009]. "Multiplication in curvature processing". In: *Journal of Vision* 9.2.



Grigorescu, C and N Petkov [2003]. "Distance sets for shape filters and shape recognition". In: *IEEE Transactions on Image Processing* 12.10, 1274–1286.



Khosravi, Hossein and Ehsanollah Kabir [2007]. "Introducing a very large dataset of handwritten Farsi digits and a study on their varieties". In: Pattern Recognition Letters 28.10. 1133–1141.



LeCun, Y. et al. [1998] "Gradient-based learning applied to document recognition". In: *Proceedings of the IEEE* 86.11, pp. 2278–2324.



Pasupathy, A. and C. E. Connor [1999]. "Responses to contour features in macaque area V4". In: *Journal of Neurophysiology* 82.5, pp. 2490–2502.



Staal, J et al. [2004]. "Ridge-based vessel segmentation in color images of the retina". In: *IEEE Transactions on Medical Imaging* 23.4, 501–509.