# **Vision-Based Magnification of Corneal Endothelium Frames**

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## **O**VERVIEW

*Objective* — To compute a high-resolution (confocal quality) image of the corneal endothelium from a low-resolution video sequence obtained with a general purpose biomicroscope.



## Approach

- 1. Automatic segmentation of the visible endothelium within each frame;
- 2. Selection of the best quality endothelium

#### DATA SELECTION

A rectangular region enclosing the endothelium region is segmented by analyzing the shape and slope of cumulative horizontal and vertical histograms of pixel intensity.



absent, but cannot provide any quality ranking among the images within the same class. On the other hand, the Laplacian operator is a powerful sharpness indicator, but is unreliable when applied to images without endothelium.



A subsequence of 60 good consecutive frames (2 sec of video) is used to obtain the HR image. The best SVM performance is obtained by using just two labels, "useful" and "not useful", Laplacian-based ranking. These methods are and concatenating the CLD (Color Layout), CSD (Color Structure) and HTD (Homogeneous Texture) MPEG-7 descriptors. content and images where the endothelium is

A quality measurement is required in order to select the best images to be used. For this pur-

complementary. On the one hand, SVMs can

discriminate between segments with endothelial

- subsequence; pose, SVM-based classification is combined with
- 3. Image alignment and mosaicing of the selected visible endothelium segments;
- 4. High-resolution image generation by using all the pixels from the aligned images.

#### FRAME MOSAICING

A robust, graph-based technique is used to align Multiple trees arise when it is not possible to the frames of the selected 60-frame subsequence. merge all the chains in a unique connected A first, raw alignment based on affine transfor- graph. When this happens, each tree corremations is carried out, producing several distinct sponds to a different mosaic, one of which is chochains of linked images. Each chain is a tree sen to produce the HR image. By default the secomposed by a set of subsequent image nodes. lected tree is the largest, unless a second one has Chains are then merged together to build wider a better average quality. trees.





A finer image alignment process then starts. For this purpose, a node is selected as root, and acts as the reference frame of an image mosaic based on full projective warping transformations (2D homographies).

## THE HR IMAGE

The HR image is a magnified version of the reference frame. It can be recovered in closed form as a linear combination of several pixels coming from the LR images. The transformation  $W_k$ mapping the HR image onto each LR image  $I_k$ is  $W_k = \text{diag}(\rho^{-1}, \rho^{-1}, 1) H_k^{-1}$ , where  $\rho > 1$  is the magnification factor and  $H_k$  is the homography mapping  $I_k$  onto the reference frame.



This method is better than using image in-

terpolation, in which case non-realistic highfrequency image artifacts would arise.

#### RESULTS

MPEG-7 descriptors	validation	runtime
CLD	76.08%	70.86%
CSD	92.11%	82.61%
SCD	80.86%	83.30%
EHD	87.80%	82.89%
HTD	89.47%	88.87%
CLD+CSD	89.71%	89.16%
CLD+SCD	80.62%	80.32%
CLD+EHD	88.28%	81.36%
CLD+HTD	90.19%	84.83%
CSD+SCD	90.19%	80.51%
CSD+EHD	89.47%	88.12%
CSD+HTD	91.39%	89.91%
SCD+EHD	90.19%	88.01%
SCD+HTD	90.67%	88.20%
EHD+HTD	91.63%	84.23%
CLD+CSD+SCD	89.00%	88.25%
CLD+CSD+EHD	90.67%	86.56%
CLD+CSD+HTD	92.58%	91.86%
CLD+EHD+HTD	88.73%	85.61%
CSD+SCD+HTD	92.82%	90.58%
CSD+EHD+HTD	91.39%	88.59%
SCD+CLD+EHD	87.61%	86.12%
SCD+EHD+HTD	91.63%	87.58%
CLD+CSD+SCD+EHD	90.67%	88.16%
CLD+CSD+EHD+HTD	89.87%	87.01%
CLD+SCD+EHD+HTD	90.01%	88.72%
CLD+CSD+SCD+HTD	92.34%	91.16%
CSD+SCD+EHD+HTD	91.13%	88.76%
CLD+CSD+SCD+EHD+HTD	91.63%	88.51%

#### SVM-based classification





Laplacian-based ranking



















