### Detection of Iris Boundaries PROPOSED APPROACH: SEGMENTATION USING THE GENERALIZED STRUCTURE TENSOR (GST) THEORY How filter output looks like with different filter radius  $I_{11} = \sum |c| [p] | (f_x [p] + i f_y [p$  $I_{20} = \sum_{r} \left( \left[ p \right] \int_{r} f_{x} \left[ p \right] + i f_{y} \left[ p \right] \right)^{2}, \qquad I_{11} = \sum_{r} \left| c \left[ p \right] \right| \left( f_{x} \left[ p \right] + i f_{y} \left[ p \right] \right)^{2},$ 2  $\sum_{11}$  =  $\sum$  |  $c[p]$ || $(f_x[p]+if_y[p])^2$ PUPIL DETECTION SCLERA DETECTION  $J_{20} = \sum_{x} c \left[ p \right] \int_{x} f_{x} \left[ p \right] + i f_{y}$  $\left[p\right] \left[f_{x}\right] p\left]+if_{y}\right[p\right]$ Finding iris circles: *p p* radius=45 radius=100 Detection of maxima in  $|I_{20}|$ detected boundaries radius=55 radius=110 image derivatives (complex) Maximum magnitude of I<sub>20</sub> radius=65 radius=120 circular filter  $|$ Circular $|$ filter **Pupil**  $\vert$ radius $\vert$

 $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ 

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## Recognition Using Periocular Information

PROPOSED APPROACH: RETINOTOPIC SAMPLING AND GABOR DECOMPOSITION OF THE SPECTRUM

## Facial Detection and Recognition Resilient to Physical Image Deformations

## DETECTION OF FACIAL LANDMARKS





Circular sampling grid (5 concentric circles, 16 points/circle)

face region in the immediate vicinity of the eye, (including eyes, eyelids, eyelashes and eyebrows)



- □ Existing periocular systems have mostly used **Local Binary Patterns** (LBPs) and, to a lesser extent, Gradient Orientation (GO) histograms and Scale-Invariant Feature Transform (SIFT) keypoints for recognition. The best performance in the few published studies is consistently given by GO from SIFT keypoints (rank-one accuracy: 81-94%, EER: 7%), LBPs (rank-one: 74-87%, EER: 19%) and finally GO (rank-one: 67-90%, EER: 22%).
- □ Comparison and/or fusion of periocular features with face and iris modalities is also done in some studies.

### PERIOCULAR REGION



- Easily obtained with existing setups for face and iris acquisition
- $\checkmark$  Available over a wide range of acquisition distances, even when iris cannot be reliably obtained (low resolution, off-angle, etc.) or face is partially occluded (close distances, obstacles, criminal scenarios, etc.)

## BACKGROUND

- □ Rotation compensation: negligible effect in the Gabor magnitude, but greatly improvement with **phase vectors** (up to 50-70% in EER)
- □ Gabor magnitude vs. phase vectors: magnitude without rotation compensation still performs equal or better than phase vectors with rotation compensation, so this step could be supressed
- $\Box$  Size of the grid: adapted to dimensions of the target eye vs. constant size  $\checkmark$  Performance not affected substantially with grids of fixed dimensions, so no accurate iris detection is needed, meaning that the only requirement is the availability of the eye center and (in some cases) of the pupil radius (both expected to be available even when the iris texture is difficult to extract)

## OUR RESULTS

Our system achieves competitive verification rates compared with other periocular recognition approaches (EER ~15%)

A number of aspects are also studied, including:

- $\Box$  Apart from **correlation of edge magnitudes**, the circular filter used takes into account the direction of edges (by encoding its expected orientation), so any disagreement in the direction will be **penalized**
- □ This is not exploited by other popular edge-based methods (Daugman, Wildes), where all boundary pixels contribute equally to (do not penalize) the detection of circles
- $\Box$  Reported results show the **effectiveness of the GST algorithm**, with similar performance than the others in pupil detection, and clearly better performance in sclera detection. These results show the validity of our proposed approach and demonstrate that the GST constitutes an alternative to classic iris segmentation approaches.
- □ Segmentation experiments under different degrees of *image defocus and motion blur* also show that comparatively, our GST algorithm always gets top performance for all levels of degradation.



- □ To detect facial landmarks (eyes, nose, mouth) and recognize individuals under image perturbations found in real-world conditions (scale, rotation)
	- $\checkmark$  Facial detection with families of symmetric patterns resilient to these perturbations
	- $\checkmark$  Use of scale-invariant patterns to avoid time-consuming image pyramids, which is the traditional approach to deal with scale changes
	- $\checkmark$  Separate detection of each facial landmark, coping efficiently with facial occlusion and view changes, which is where current face detection methods fail
	- User authentication by using separate machine experts that extract Gabor features in the facial region surrounding each (available) landmark

□ This will contribute to increase user convenience and comfort by reducing the cooperation level required and by allowing the use of their own device and natural interaction patterns to communicate with the system





## OUR RESULTS

### OUR TARGET: FACE TECHNOLOGIES APPLICABLE TO REAL-WORLD DATA

LOG-POLAR SAMPLING GRIDS AND GABOR FILTERS MIMICKING RETINA PHOTORECEPTORS AND VISUAL CORTEX CELLS



## PERSONAL RECOGNITION

## LOW COST DEVICES DISTANT ACQUISITION UNCONTROLLED BACKGROUND AND LIGHTNING

