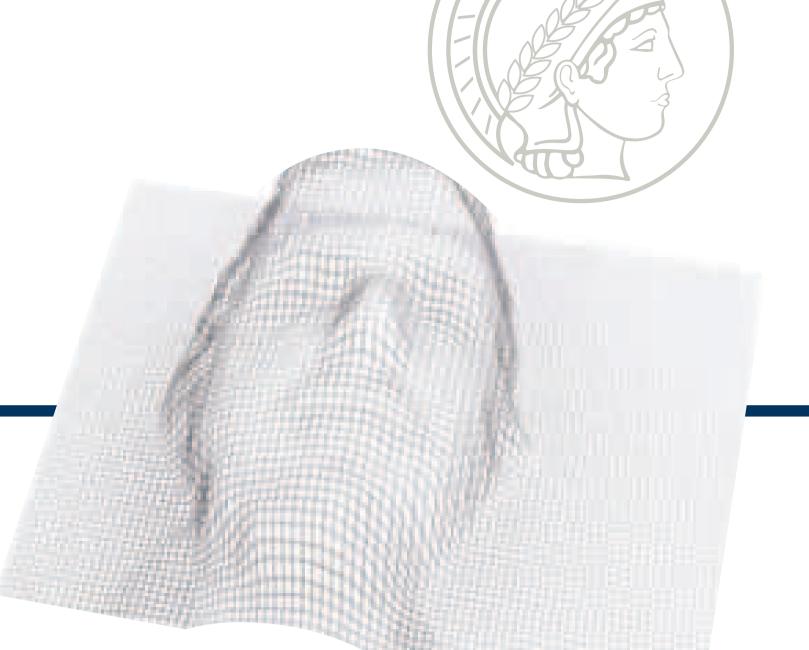
Rapid Stereo-Vision Enhanced Face Detection

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Face detection often serves as a first step in image processing applications, like, face recognition, visual surveillance, or human-machine interaction. Our approach improves state-of-theart 2D detection techniques, by additionally evaluating a disparity map, which is estimated for the face region.

1 First, faces are detected in the 2D images with a rapid object classifier based on haar-like features. 2 In a second step, we train a PCA-classifier (off-line) that helps removing falsely detected faces by analyzing the disparity map. 3 In the near field of the camera, this classifier is used, which evaluates the Eigenfaces of the normalized disparity map. Thereby, the transformation into Eigenspace is learned off-line using a principal component analysis approach. 4 In the far field, a much simpler approach determines falsepositives by evaluating the relationship between the size of the face in the image and its distance to the camera. 5 This novel combination of algorithms runs in real-time and significantly reduces the number of false-positives compared to classical 2D face detection approaches.



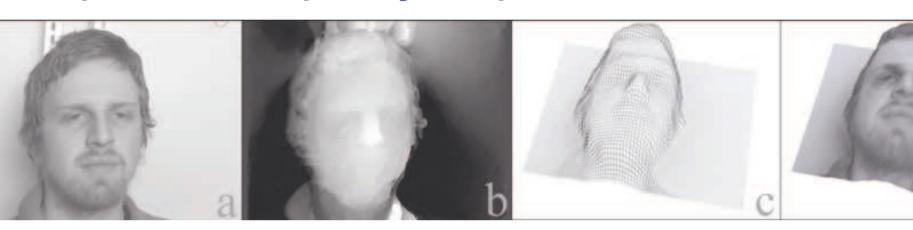


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Stereo Setup and Algorithm

- Two cameras with a resolution of 384 x 288 pixels
- Near-field stereo case: 20 cm baseline distance and 9,5° convergence-angle, far-field stereo case: 4 cm baseline distance and 0° convergence-angle
- Off-line calibration and rectification of input images
- Estimation of disparity map (variational approach, including regularization)

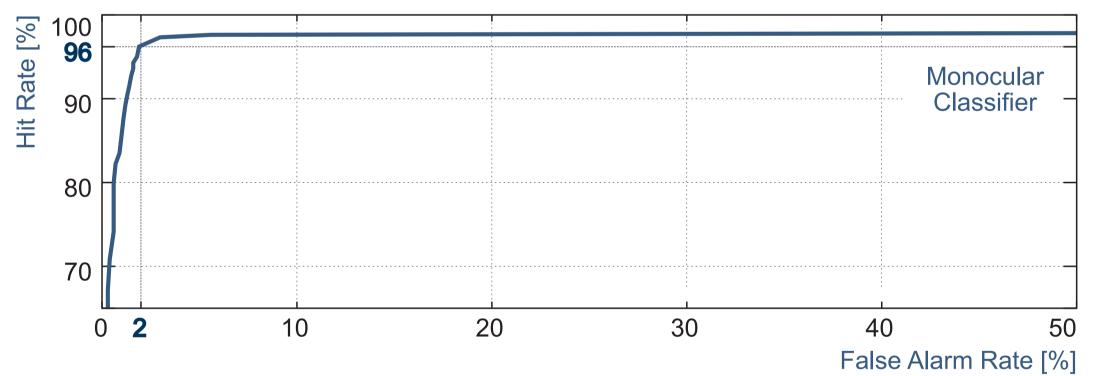
Example of a disparity map estimation:



1 2D Face Detection

- Training of a boosted classifier cascade, which uses rotated haar-like features, using discrete AdaBoost
- Input: 4700 positive samples, 3300 negative samples
- We apply the classifier on the left image of all stereo pairs to identify potential face candidates

Receiver Operator Characteristic of the classifier



2 Evaluating The Disparity Map

- We add a classification step on the disparity maps
- And use a PCA-classifier that is trained off-line on 30 facial regions

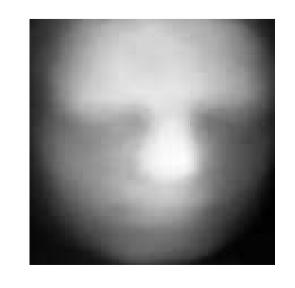
Stereo-image pairs and generated disparity maps that where used in our training set of the PCA

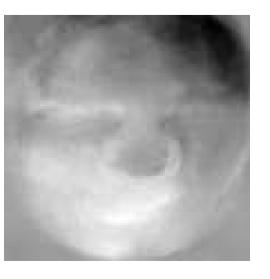


3 Near Field: 3D-Eigenfaces

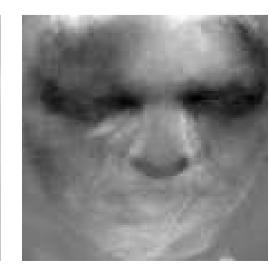
■ The PCA estimates the probability distribution of facial disparity maps around their average

Average face, and the first 3 Eigenfaces (left to right)

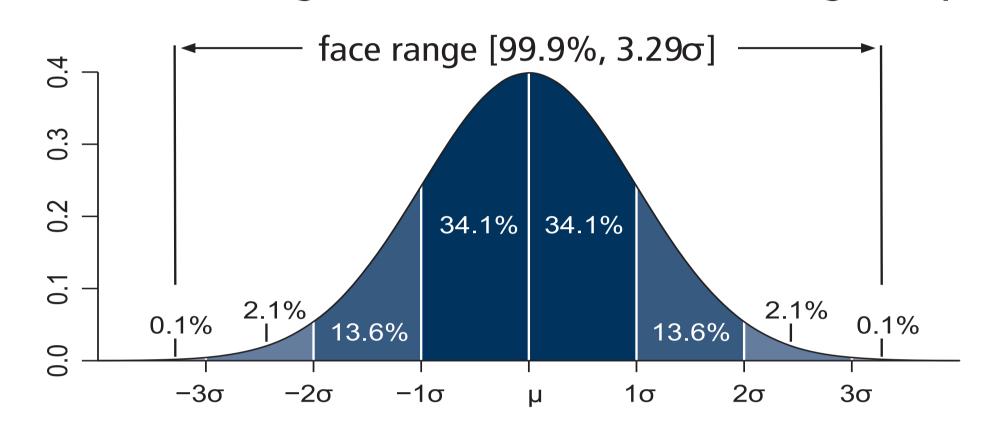








- Online detection: we project disparity map regions of potential candidates (as detected) into PCA space
- We calculate the Mahalanobis distance of current test signals to the average face
- By a threshold of σ =3.29 (in terms of standard deviation), we accept of all face candidates lying within the range of ~99,9% of all training samples



Face candidates that lie above this threshold will be rejected as non-faces

4 Far Field: Evaluating Size-Relation

- In small image patches, the above algorithm fails because a reliable disparity map cannot be estimated
- If the patch size of a face is smaller than 40x60 pixels we estimate the disparity to the right frame with the Kanade-Lucas-Tomasi approach and calculate the median disparity and the corresponding distance to the camera
- We can determine the actual size of the face in 3D and check it against an interval of sizes for veridical faces, as learned from a database of 3D face scans
- If the face size lies not in a range from 15.78 to 24.16 cm, the face is marked as a false-positive

5 Experimental Results

- Comparison of the detection rates of the monocular detection approach with our stereo enhanced method
- The two shaded columns show that the number of false-positives (FP) decreases using our method, and the number of true-positive (TP) and false-negative (FP) stays the same

	40 Faces			19 Non-Faces	
	FP	TP	FN	FP	TN
monocular	7	38	2	5	14
stereo enhanced	0	38	2	0	19

Timings for an Intel® CoreTM 2 CPU with 2.66 GHz

Tillings for all litter Core in 2 CPO with 2.00 GHZ					
Step	FF [msec]	NF [msec]			
Run monocular detector	27	27			
Estimate disparity map	_	78			
Transform into Eigenspace	_	42			
Estimate sparse disparity map	63	_			
Total	90	147			