

A Cheap Portable Eye-Tracker Solution for Common Setups

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Introduction Proposed Setup Contributions Experiments Results Conclusions

Outline

Introduction

Proposed Setup

Contributions

Experiments

Results

Conclusions

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► Expensive (€,€€€)

- Expertise required for scientists
- Not accessible to regular computer users
- ADVANTAGE: High accuracy (0.1°-1°) and high refresh rates (up to 600 Hz)

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Building an eye-tracker that is:

- Cheap
- Easy to use
- Scalable
- Easy to build with custom components
- Compatible with a variety of operating systems
- Accurate and robust

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Such a system can be used in:

- Remote eye-tracking studies [1]
- Parallel experiments in a laboratory
- Applications in mobile devices
- Non-critical applications such as games

[1] A. Bojko, "The Truth About Webcam Eye Tracking"

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Techniques (By Camera Type) [2]

Infrared (IR): IR methods use IR lights and cameras. Make use of the reflections of lights on the cornea.

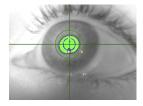


Image ©ITU Gaze Group.

[2] D. W. Hansen and Q. Ji, "In the eye of the beholder: A survey of models for eyes and gaze." IEEE TPAMI, vol. 32,

no. 3, pp. 478-500, 2010.

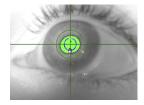
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Techniques (By Camera Type) [2]

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 Visible light: Use regular cameras and no additional light as part of the algorithm.

Image ©ITU Gaze Group.

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Techniques (By Methodology)

Model based: Fit a 2D/3D model to the eye/face. Then calculate the gaze point geometrically.



Image © Takahiro Ishikawa.

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Introduction Proposed Setup Contributions Experiments Results Conclusions

Techniques (By Methodology)

Model based: Fit a 2D/3D model to the eye/face. Then calculate the gaze point geometrically.



Appearance based: Directly map the image of the eyes to the gaze point. Mostly neural networks (NN) are trained with image pixels as input and gaze point as output.

Image ©Takahiro Ishikawa.

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There exist several open source options:

- ITU Gaze Tracker from IT Univ. of Copenhagen Uses IR cameras and IR light sources, camera image contains only an eye
- Opengazer from Univ. of Cambridge
 Uses regular webcams, no special lighting, camera image contains the entire face
- Others: EyeWriter, openEyes, Neural Network Eye Tracker (NNET), etc. [3]



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Software structure:

- Based on Opengazer [4], works under natural light
- 5 components modified
 - Point selection
 - Point tracking
 - Blink detection
 - Calibration
 - Estimation

[4] P. Zieliński, http://www.inference.phy.cam.ac.uk/opengazer/

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Hardware for portable eye-tracker:

Raspberry Pi, webcam and accessories (SD card, cables)

[4] P. Zieliński, http://www.inference.phy.cam.ac.uk/opengazer/

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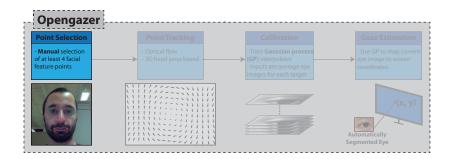
Introduction	Proposed Setup	Contributions	Experiments	Results	Conclusions

Video

[EYE-TRACKER VIDEO]

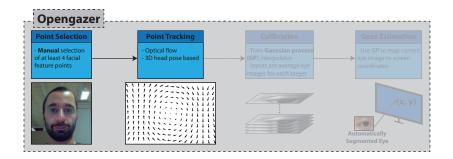
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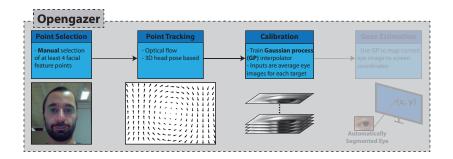
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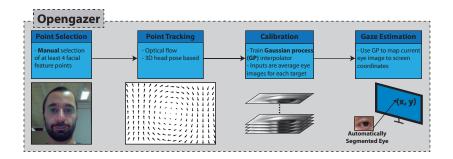
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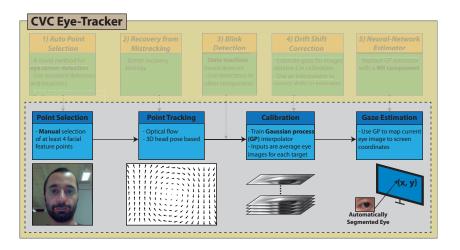
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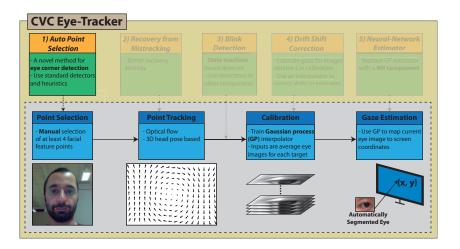
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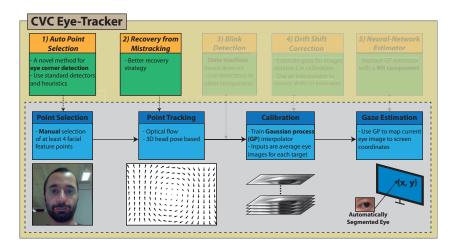
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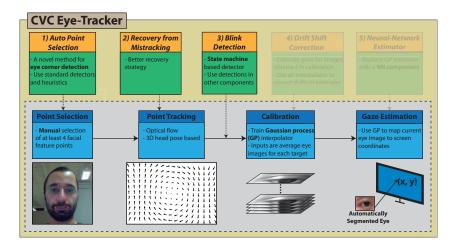
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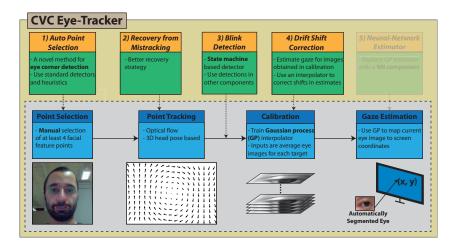
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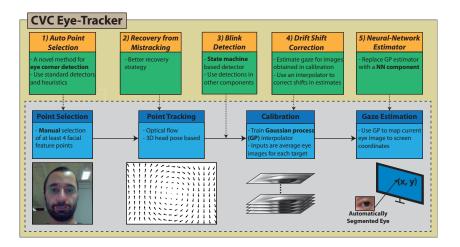
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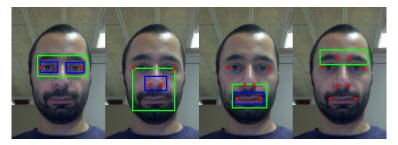
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1) Automatic Point Selection

- Based on Haar cascades [5] (detectors for eye, nose, mouth)
- Use of geometrical heuristics (e.g. location of nosetips inside the detected nose rectangle)
- Novel method for fast eye-corner detection



[5] M. Castrillon-Santana, http://mozart.dis.ulpgc.es/Gias/modesto.html?lang=0

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2) Tracking and 3) Blink Detection

- Improved strategies for point tracking and recovery from mistracking
 - In case of mistracking, use the 3D head pose based estimation
- A state-machine based blink detector
 - No estimation during blinks
 - Skip blinking frames during calibration

4) Calibration Error Correction

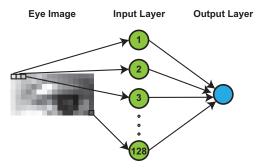
A drift shift correction method for improving training:

- 1. After calibration, calculate the gaze estimations for images acquired during calibration
- 2. Measure the errors between the average estimation for a target and actual target coordinates
- 3. Train a multivariate interpolator to correct these shifts
- 4. During testing, correct the estimations using the interpolator

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5) Neural Network Estimator

Implement a neural network based estimator [6]:



- Eye image resized to 16x8 pixels
- 4 networks for 2 eyes and 2 different outputs (X and Y coordinates)

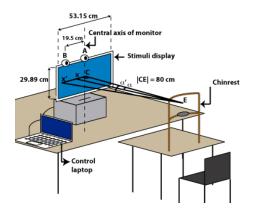
[6] O. V. Komogortsev, http://cs.txstate.edu/ok11/nnet.html

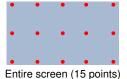
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Experimental Setup

Components of the experimental setup and target point locations:







iPad screen (12 points)

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Setup Variations

Test for 4 different cases:

Setup	Distance	Camera	Chinrest	Target	
				• • • • •	
		_		••••	
STANDARD	80 <i>cm</i>	Center		• • • • •	
				••••	
				• • • • •	
EXTREME	80 <i>cm</i>	Left			
				••••	
				• • • • •	
CHINREST	80 <i>cm</i>	Center	Yes		
iPAD	40 <i>cm</i>	Center			

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	Horizontal err.		Vertical err.	
	$^{\circ}(\sigma)$	Px	$^{\circ}(\sigma)$	Px
ORIG	1.82 (1.40)	95	1.56 (0.81)	80
2-EYE	1.56(1.28)	82	1.48(0.86)	75
TRACK	1.73(1.37)	91	1.58(0.78)	81
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 640×480 camera

 1280×720 camera

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	Horizontal	err.	Vertical err.				
	$^{\circ}(\sigma)$	Px	$^{\circ}(\sigma)$	Px			
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Extreme	1.76 (1.15) 92 1.78 (0.67		1.78(0.67)	91			
Chinrest	1.14(0.79)	60	1.36(0.53)	69			
iPad	2.15(1.32)	55	2.02(0.76)	52			

 640×480 camera

 1280×720 camera

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Experiments

Comparison of Setups (CORR)

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Standard	1.62 (1.40)	85	1.50 (0.77)	76			
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Decrease in errors by 18% hor. and 17% ver. (std. setup)

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- In iPad setup, pixel errors are smaller but angular errors appear larger

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A Portable Eye-Tracker

Feasibility analysis of our system as a portable, stand-alone eye-tracker



- The video stream is processed by Raspberry Pi
- Estimates can be accessed in real-time (ethernet) or offline (text file)
- > The **reporting** and **visualisation** tools run in the analysis computer
- The eye-tracker can estimate the gaze with ~4 fps in 640×480

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Improved the state-of-the-art (18% hor. and 17% ver.)

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- The eye-tracker (Mac OS X and Linux) is ready for download (http://bit.ly/cvc-eyetracker)

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Introduction Proposed Setup Contributions Experiments Results Conclusions

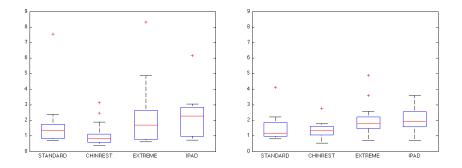
The End

Thanks for your time!

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Comparison of Subject Performances



Box plots for subject error rates horizontally (left) and vertically (right) under 1280×720 camera

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