

Corneal Reflection Detection and Matching Using Deep Learning for Eye Tracking

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Introduction

What is Eye Tracking?

Ability to determine where a person is looking on a display or in the real world.

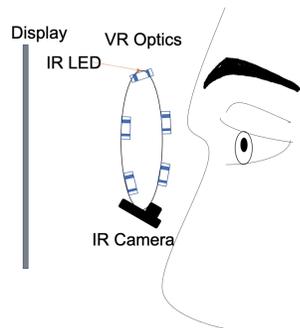


Fig 1. Eye Tracking in virtual reality (VR) Systems

Method of detecting gaze [1]

Model based approach: Uses two or more corneal reflections and pupil center to estimate the point-of-gaze. These estimates are insensitive to headset shifts.

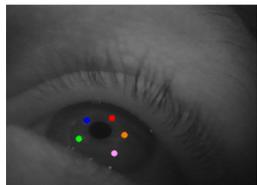


Fig 1. Labelled Corneal Reflections

Corneal reflection Requirements

- Precise locations of corneal reflections in eye images (Detection).
- Information about which reflection came from which light source (Matching)

Challenges:



Fig 2: Natural Spurious Reflections



Fig 4: Missing Reflections with change in gaze angle

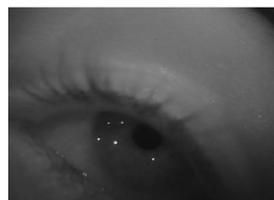


Fig 3: Missing reflections due to eyelid occlusion

Methods

Convolutional Neural Network

- U-Net style architecture
- Input and Output share the same size
- Loss Function: Weighted combination of Cross Entropy and Soft Dice Loss

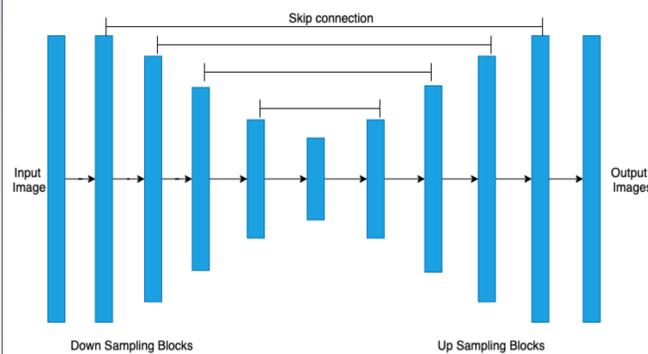


Fig 5. UNET Architecture

Dataset Collection and its Labelling

- Eye images (640x480) collected from 15 people inside an HTC Vive VR headset.
- Three of the 15 subjects wore glasses during the data collection.
- Also used unlabeled dataset from NVIDIA [3] which consists of eye images of 10 people.
- Manually labelled 4000 eye images across 25 people.

Static Data Augmentation

- Increased dataset from 4000 to 40000 eye images
- Generated 10 random crops of resolution 320x240 for every labelled eye image of resolution (640x480).

Dynamic Data Augmentation

- Added Spurious Reflections with varying intensity and sizes.
- Added Gaussian and Motion Noise.
- Adjusted Brightness and Contrast of image.

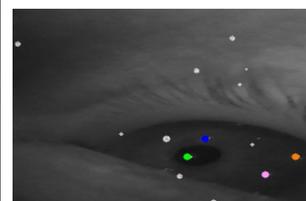


Fig 6: With Spurious Reflections

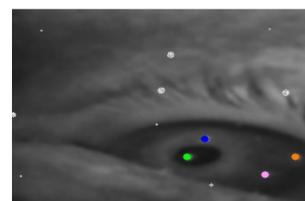


Fig 7: All techniques applied only

Results

Dataset Splitting

- Dataset was divided into training (76%), validation (12%) and test (12%) set.
- Each dataset contained images from different subjects.

Single Corneal Reflection Performance

- Accuracy for detecting true corneal reflections and finding its matching light source on the test dataset is 91%.
- One forward pass through the neural network on a RTX2060 GPU takes 5.7ms.
- Uses 3 Mb of memory.

Performance Comparison with other Systems

- Only one prior deep learning approach exists for eye tracking in VR systems [2].
- Accuracy comparable with [2].
- 10x Faster and uses 33x less space in memory than [2].

Effect of Spurious Reflections

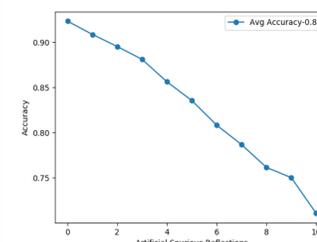


Fig 8: Corneal reflection pair accuracy

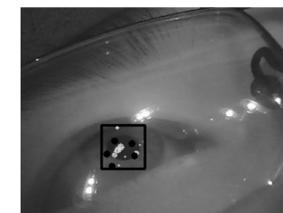


Fig 9: Addition of spurious reflections near the pupil

Eye Tracking Performance

- Integrated the trained network in a VR eye tracking system.
- Made use of a similar UNET style network for pupil estimation.
- Tested the eye tracker on 3 subjects.
- System reported mean absolute error of 1°

Table 1: Eye Tracking Performance

| Metric | Gaze Error (°) | | |
|---------------------|----------------|------------|------------|
| | EyeNet [2] | NvGaze [3] | Our System |
| Mean Absolute Error | 3 | 2.1 | 1 |

Conclusions

- Presented a neural network based on U-Net architecture for tracking multiple corneal reflection inside a VR system.
- The only prior deep learning-based solution [2], uses a series of networks for detecting and matching corneal reflections.
- Use of U-Net style network allows training with a dataset 24x smaller compared to [2].
- The proposed network is computationally efficient than [2], while achieves similar accuracy.
- Under the presence of upto 10 spurious reflections near the pupil, the accuracy of tracking multiple corneal reflections only reduces by 8%.
- Easy to retrain a U-Net style network for different hardware configurations.
- Performance of the network is improved with the use of Dilated Convolutions in the CNN layers and max pooling layer for down sampling.
- The proposed algorithm when integrated into an eye tracking system within a VR headset achieves an error of 1°.
- The gaze error is at least 100% lower than previous VR eye tracking systems.
- The presented approach is applicable to other XR systems that use Augmented Reality and Mixed Reality headsets.

Bibliography

- 1) E. D. Guestrin and M. Eizenman, "General theory of remote gaze estimation using the pupil center and corneal reflections," in IEEE Transactions on Biomedical Engineering, vol. 53, no. 6, pp. 1124-1133, June 2006.
- 2) Z.Wu, S.Rajendran, T.Van As, V.Badrinarayanan and A.Rabinovich, "EyeNet: A Multi-Task Deep Network for Off-Axis Eye Gaze Estimation," In 2019 IEEE/CVF International Conference on Computer Vision Workshop (ICCVW), pp. 3683-3687, October 2019.
- 3) J. Kim, M. Stengel, A. Majercik, S. De Mello, D. Dunn, S.Laine, M. McGuire, and D. Luebke, "Nvgaze: An anatomically-informed dataset for low-latency, near-eye gaze estimation," In Proceedings of the 2019 CHI Conference on Human Factors in Computing Systems, Scotland, pp. 1-12, May 2019.