



Viability of Optical Coherence Tomography for Iris Presentation Attack Detection

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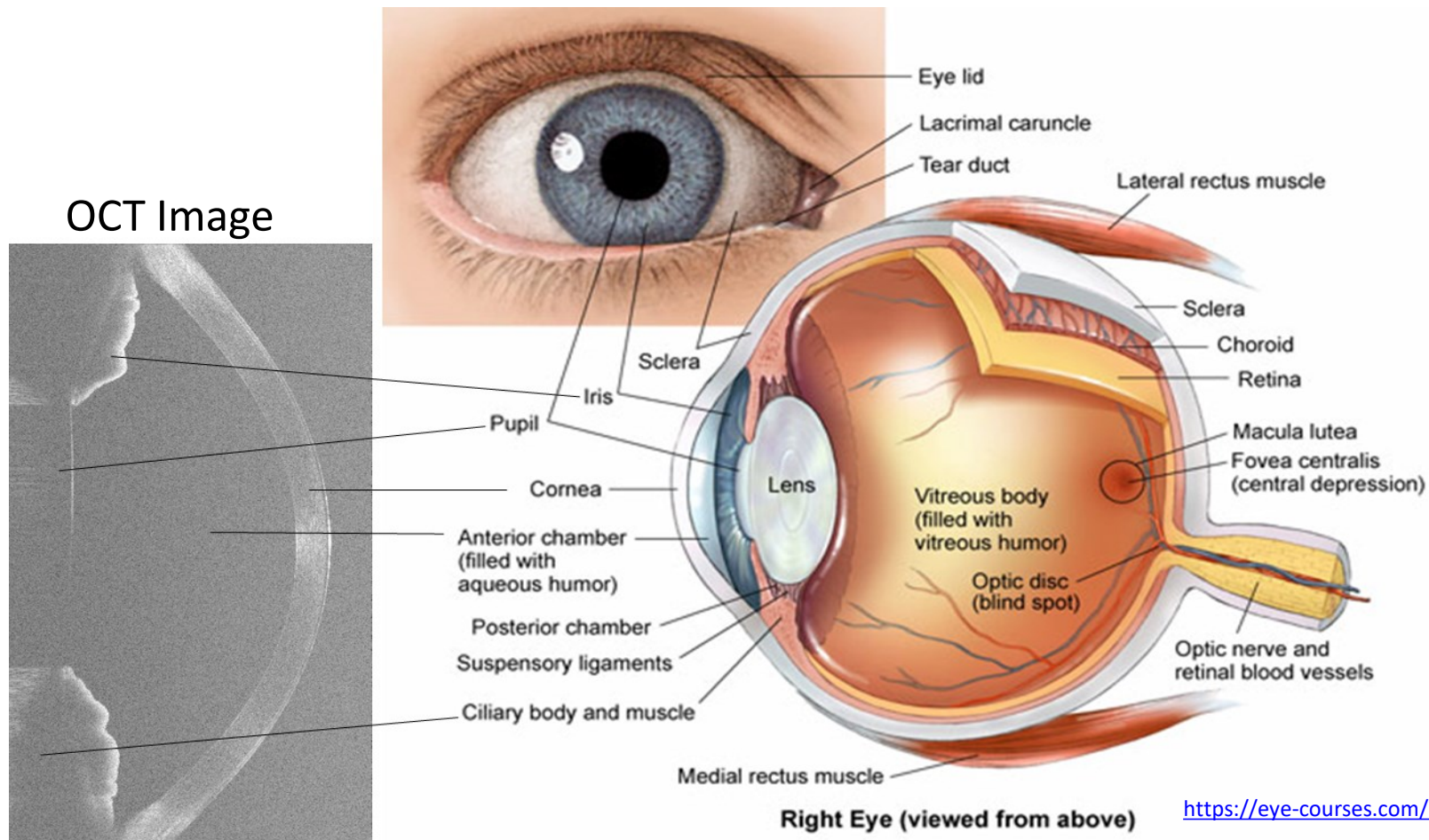
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Optical Coherence Tomography (OCT) Imaging

Optical Coherence Tomography (OCT) is non-invasive, micrometer-resolution imaging that captures **2-D cross-sectional** or 3-D volumetric images of an object in the NIR spectrum



Iris Presentation Attack

A presentation attack (PA) occurs when an adversarial user presents a fake or altered biometric sample to the sensor in order to **spoof another user's identity, obfuscate their own identity, or create a virtual identity.**

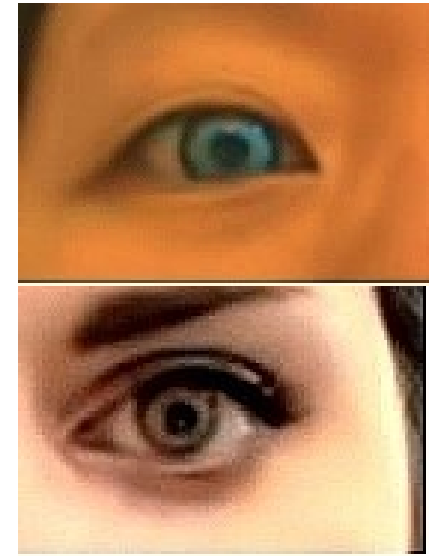
Print



Plastic and Glass



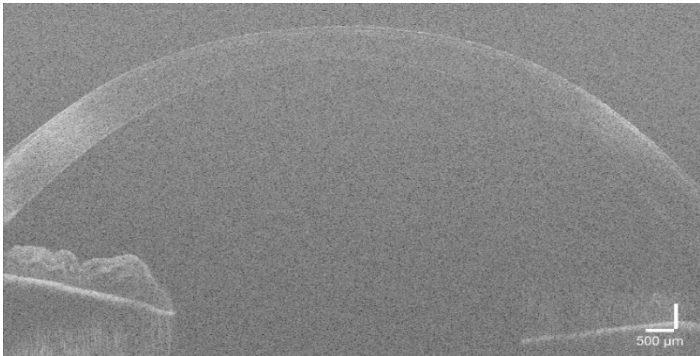
Cosmetic Lens



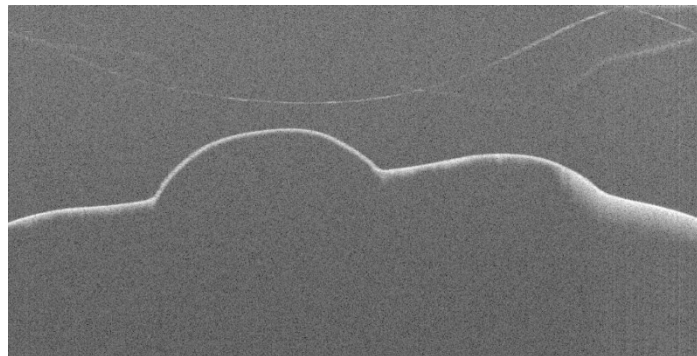
Motivation

- Bonafide OCT images show a **cross-sectional view** of cornea and iris
- Fake eye OCT images show a thin line corresponding to **the fake eye outer structure**
- The cosmetic contact lens also blocks light, which creates **gaps in the cornea** structure

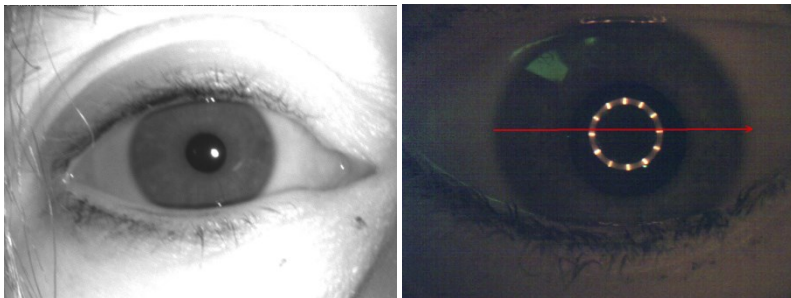
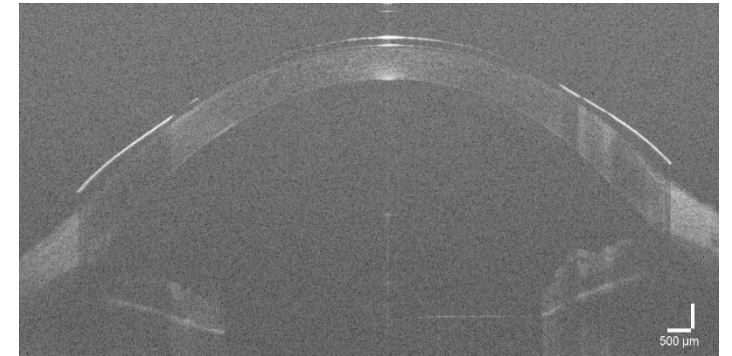
Bonafide



Fake Eye

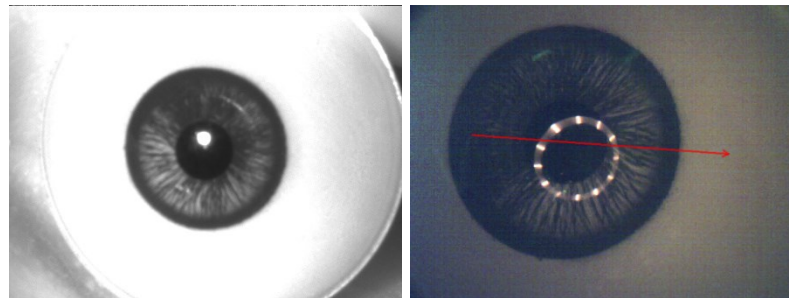


Cosmetic Contact Lens



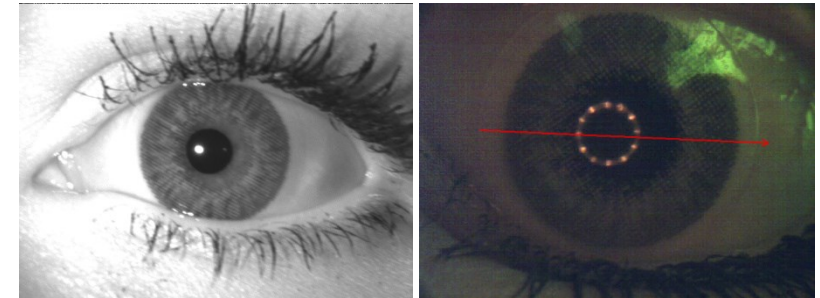
NIR

RGB



NIR

RGB



NIR

RGB

Main Contributions

- Propose an **iris PA detection** technique based on **OCT imaging** modality
- Compare the performance of **OCT**, Near-Infrared (**NIR**), and Visible (**RGB**) imaging modalities using **VGG19**, **ResNet50**, and **DenseNet121** deep architectures
- Perform the experiments on a **proprietary dataset** of 370 subjects collected from each modality
- Explain the results using **CNN visualizations**: Grad-CAM heatmaps and t-SNE plots

Implementation Details

- Utilize **pre-trained VGG19**[1], **ResNet50** [2] and **DenseNet121**[3] models trained on ImageNet dataset [4] which further **fine-tuned** with OCT, NIR and RGB iris PA data
- Images are resized to 224 x 224 and **normalized using z-score** before input
 - This helps in regulating the weight parameters
- Learning rate = 0.005, batch size= 20, and no. of epochs = 50

1. K. Simonyan and A. Zisserman. Very deep convolutional networks for large-scale image recognition. ICLR, 2015.
2. K. He, X. Zhang, S. Ren, and J. Sun. Deep residual learning for image recognition. CVPR, 2016.
3. G. Huang, Z. Liu, L. v. d. Maaten, and K. Q. Weinberger. Densely connected convolutional networks. CVPR, 2017.
4. J. Deng, W. Dong, R. Socher, L.-J. Li, K. Li, and L. Fei-Fei, ImageNet: A Large-Scale Hierarchical Image Database. CVPR, 2009.

Dataset and Experimental Setup

- The dataset collected under the Odin program of IARPA from **370 subjects** contains **844 bonafide**, **61 fake eyes**, and **120 cosmetic contact (CC)** images
- OCT, NIR and VIS images are captured using **Telesto series OCT sensor**, **RGB camera**, and **iCAM7000 NIR sensor**, respectively
- Experiments are performed under **intra** and **cross-attack** scenarios

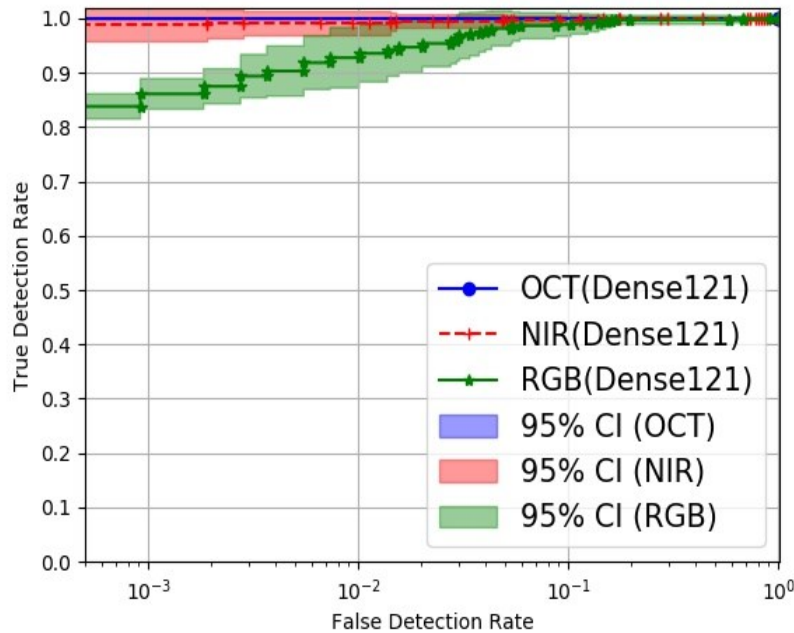
Experiments	Train Set		Validation Set		Test Set	
	Bonafide	PAs	Bonafide	PAs	Bonafide	PAs
Intra-EXP (Fold 01-05)	404	100	101	25	218	54
Cross-EXP 1 (CC Lenses are unknown)	435	41 (Fake Eyes)	145	18 (Fake Eyes)	146	120 (CC)
Cross-EXP 2 (Fake eyes are unknown)	435	84 (CC)	145	36 (CC)	146	59 (Fake Eyes)

Experimental Results

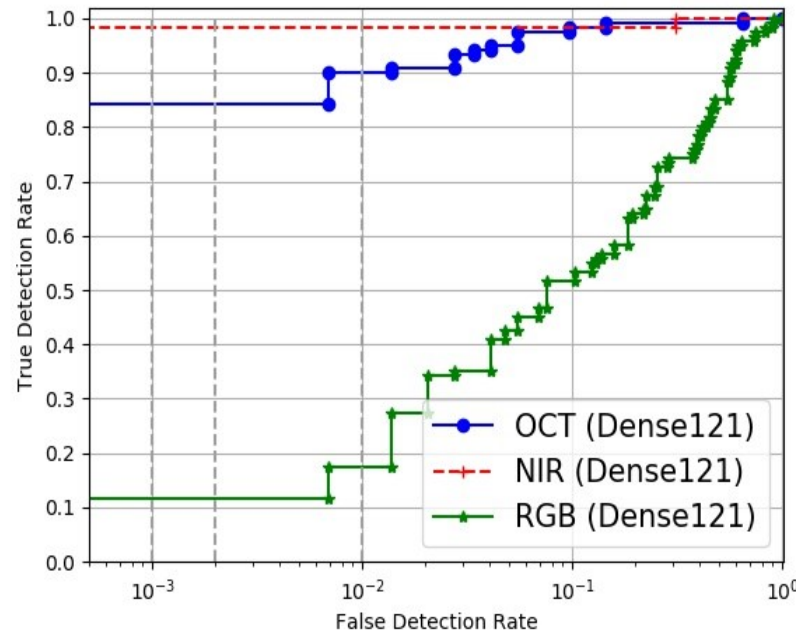
	True Detection Rate (%) at 0.2% False Detection Rate								
Experiments	VGG19			ResNet50			DenseNet121		
	OCT	NIR	RGB	OCT	NIR	RGB	OCT	NIR	RGB
Intra-EXP	100 ± 0.00	97.99 ± 2.66	82.58 ± 6.88	100 ± 0.00	97.33 ± 3.88	89.62 ± 3.62	100 ± 0.00	97.66 ± 3.26	86.66 ± 3.59
Cross-EXP 1 (CC are unknown)	21.66	97.58	26.66	92.50	98.38	15.00	84.16	98.38	11.66
Cross-EXP 2 (Fake eyes are unknown)	86.44	98.38	93.22	94.91	96.77	81.35	94.91	96.77	91.52

OCT images better discriminate between bonafide and PAs under **known PAs scenario**, whereas **NIR images** perform better across **unknown PAs scenario**

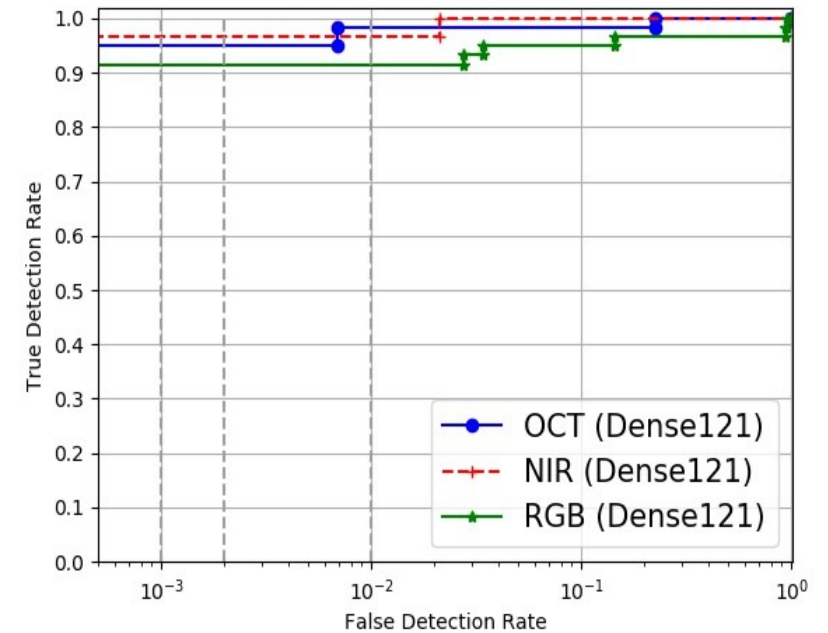
Experimental Results: ROC Curves [DenseNet121]



Intra-EXP



Cross-EXP 1
(CC are unknown)

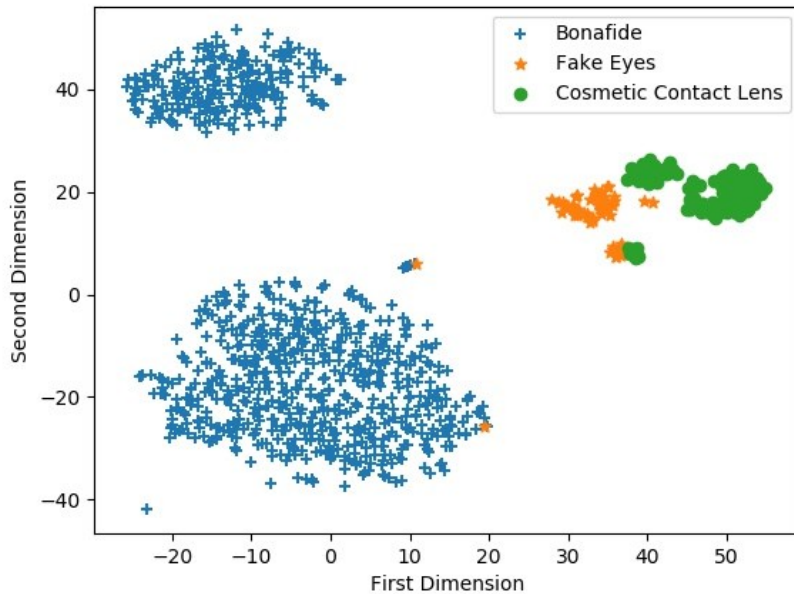


Cross-EXP 2
(Fake eyes are unknown)

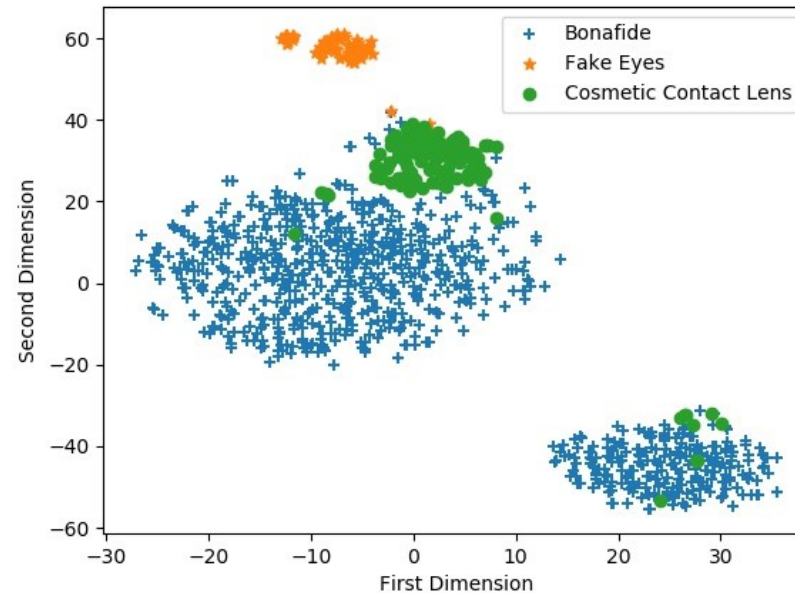
OCT imaging performs better under intra-attack scenario and NIR imaging performs better under cross-attack scenario

Explainability: t-SNE Plots

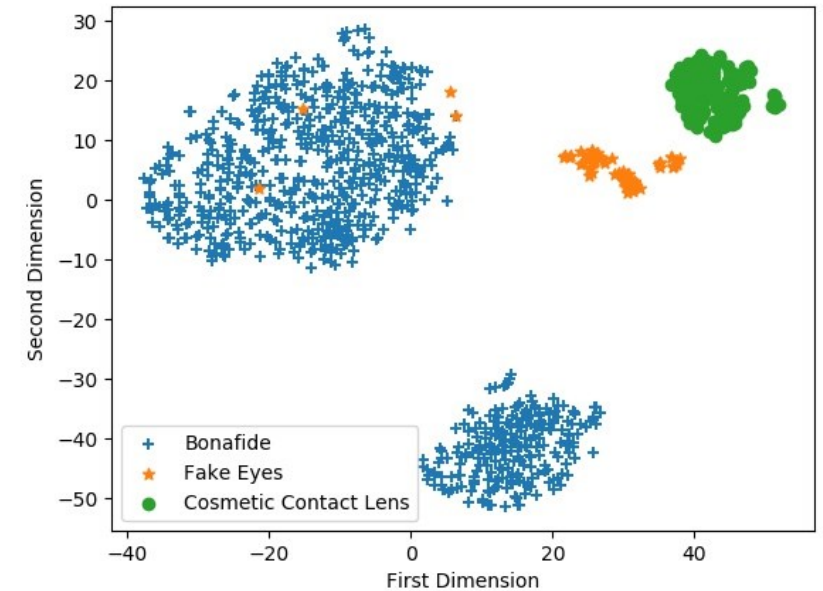
t-SNE helps in **visualizing** the high-dimensional features in a scatter plot



Intra-EXP



Cross-EXP 1
(CC are unknown)

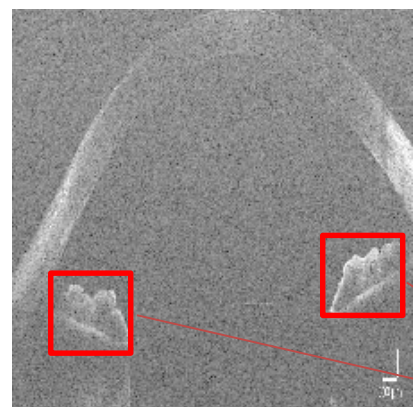


Cross-EXP 2
(Fake eyes are unknown)

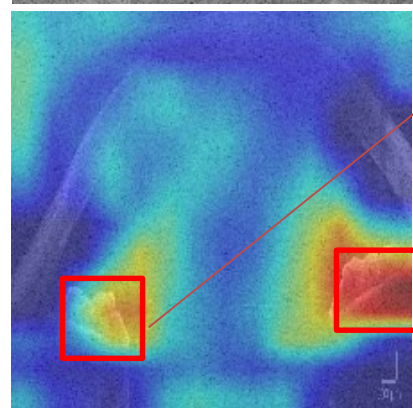
Distribution of bonafide and PAs are better **separated** in **Intra-EXP** and **Cross-EXP 2**

Explainability: Grad-CAM Heatmaps

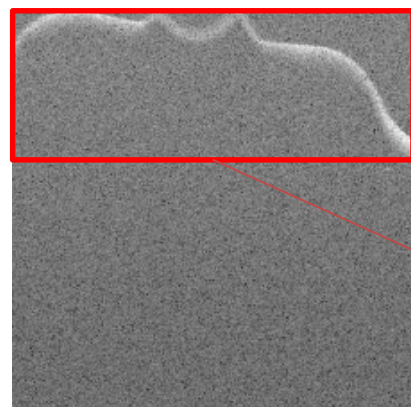
Heatmaps emphasize the **salient regions** utilized by the network to detect PAs



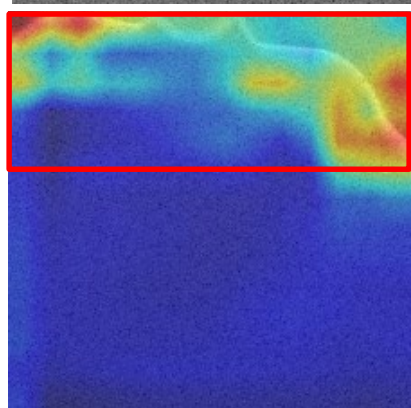
Iris
region



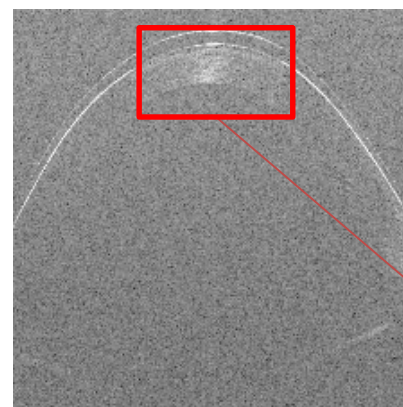
Bonafide



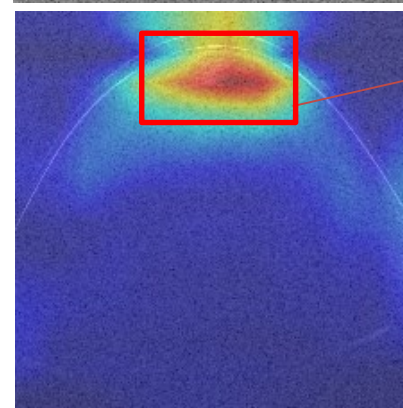
Outer
structure
of fake
eyes



Fake Eye



Pupillary area
of Cornea



Cosmetic Contact

Red-colored
regions
represent
highly focused
regions,
whereas the
blue region
represents low
priority
regions

Distinctive region of focus in each category helps in distinguishing bonafides and PAs

Conclusion and Future Work

- Proposed a **viable** solution for iris PA detection based on **OCT imaging**
- It gives **near-perfect performance** on a small dataset set under intra-attack scenario, whereas NIR imaging generalizes better under cross-attack scenario
- The performance is **explained** using t-SNE and Grad-CAM plots
- Future work: to collect **more OCT iris data** along with other types of PAs using different types of OCT hardware
- **Hardware cost** is one of the current limitations

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Any
questions?

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