Iris Segmentation in Challenging Conditions

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Introduction

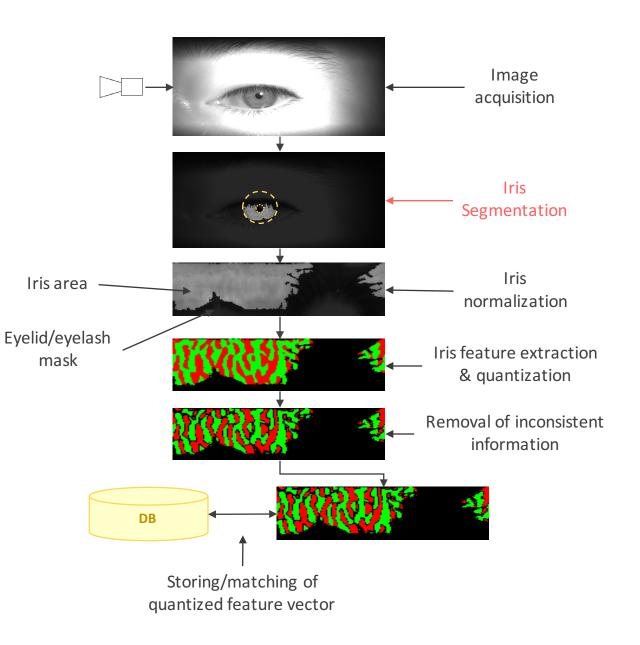
Why iris?

- It is unique for every person
- It almost does not change during the life
- Its image has high informational capacity
- It is hard to counterfeit

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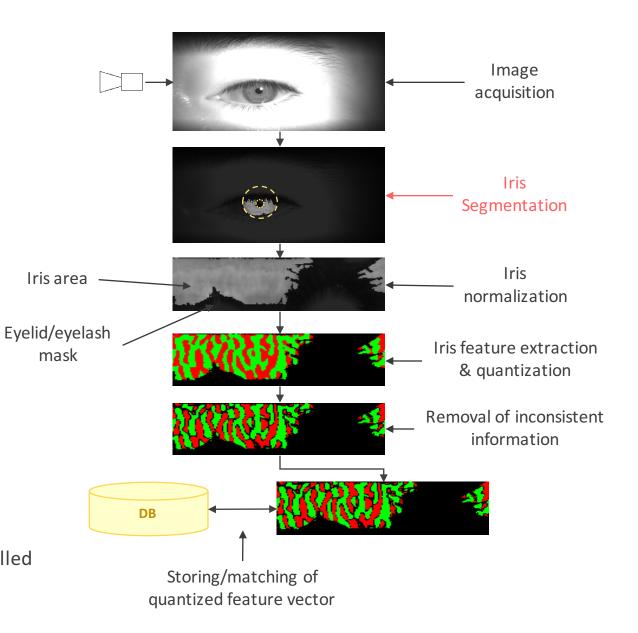
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Motivation:

- Segmentation is an irreplaceable stage of iris recognition pipeline
- Quality of segmentation hugely affects overall recognition performance
- Segmentation is still challenging under less controlled environment



Problem statement

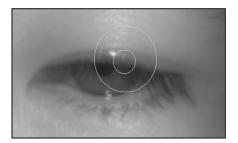
Less controlled environment:

- Conditions:
 - illumination level
 - non-glasses/glasses/contact lenses
- Eye variations:
 - pupil dilation/contraction
 - gazeaway
- Iris occlusions:
 - eyelids/eyelashes
- Device performance limitations:
 - CPU, RAM, Camera resolution (mobile)

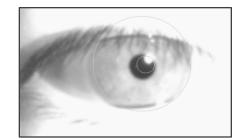


As the result:

- Image quality degradation:
 - under-/over-exposure
 - poor contrast
 - reflections from glasses
- Segmentation errors:
 - Wrong pupillary/iris/eyelid border determination
- Recognition performance degradation:
 - intra-class variations



gaze-away, eyelid occlusion



overexposure



under-illumination, glass reflection



poor contrast, glass reflection

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Existing approaches

Common (since 1993, J. Daugman):



Captured image



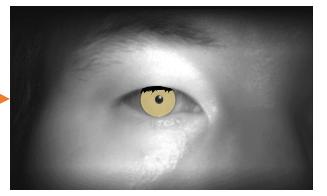
Extraction of eye region



Preprocessing



Round approximation of pupil & iris



Eyelid & eyelash detection

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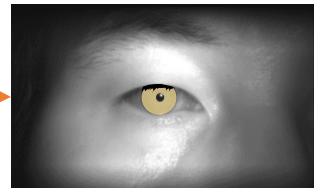
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Eyelid & eyelash detection

CNN based (end-to-end):

Captured image



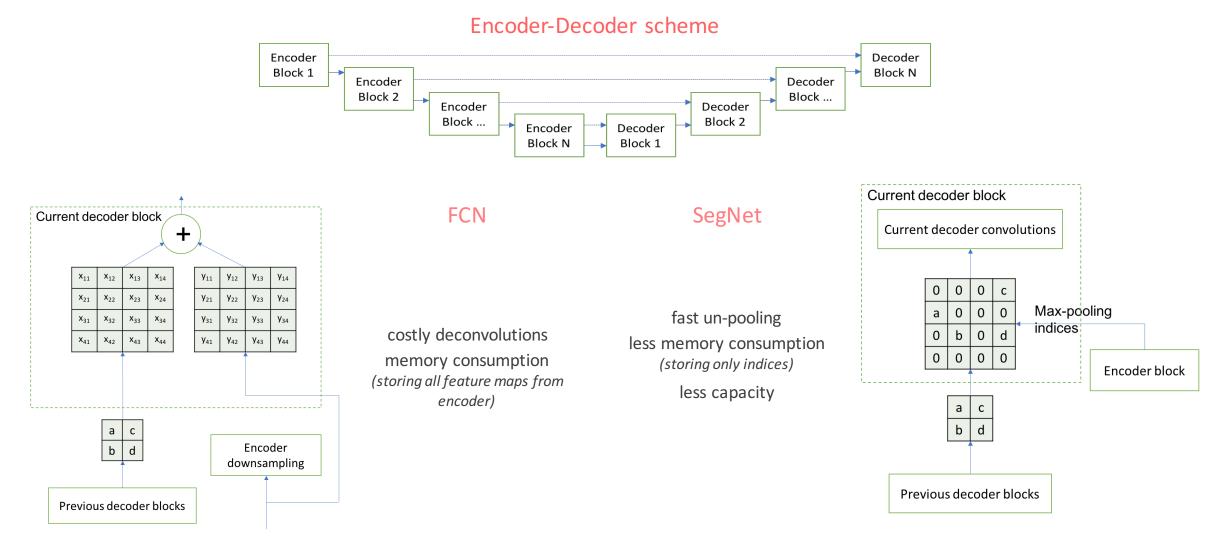
Eyelid & eyelash detection

| | UBIRIS.v2 | CASIA.v4 |
|---------------------------------------|-----------|----------|
| Method | error(%) | error(%) |
| Ours MFCNs | 0.90 | 0.59 |
| Ours HCNNs | 1.11 | 1.08 |
| Z. Zhao and A. Kumar, ICCV, 2015 [33] | 1.21 | 0.68 |
| T. Tan et al., IVC, 2009 [28] | 1.31 | - |
| C. Tan and A. Kumar, T-IP, 2013 [27] | 1.72 | 0.81 |
| H. Proença, T-PAMI, 2010 [19] | 1.87 | - |
| C. Tan and A. Kumar, T-IP, 2012 [26] | 1.90 | 1.13 |

Liu et al. 2016

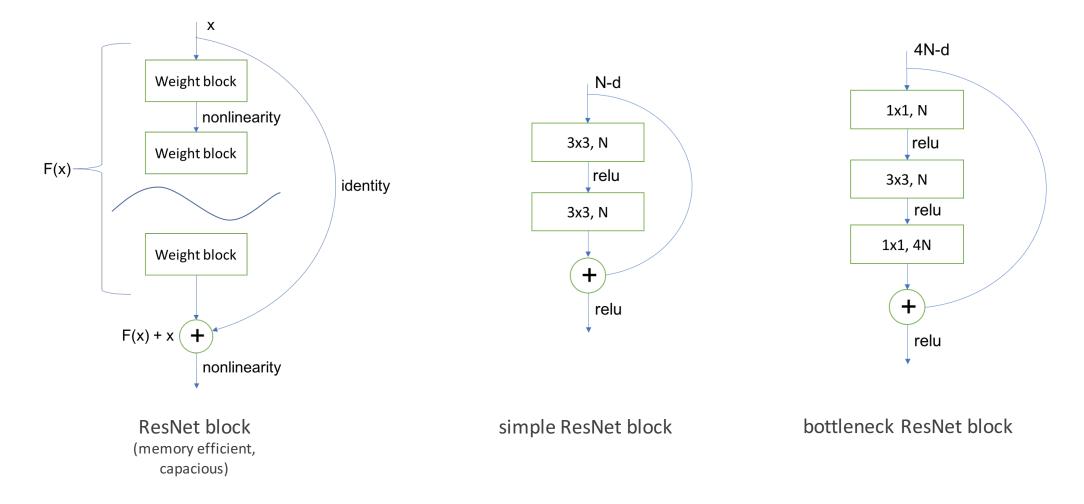
$$error = \frac{1}{N \times m \times n} \sum_{i,j \in (m,n)} G(i,j) \oplus M(i,j)$$

Review: Segmentation Networks



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Proposed approach: Internal Block Design

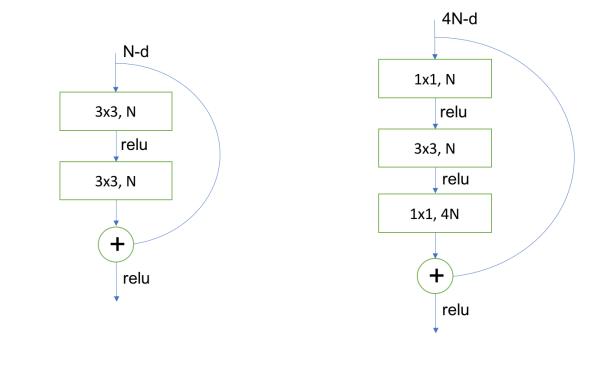


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Proposed approach: Network Design

FCN (re-designed):

- Composed with bottleneck ResNet blocks
- Encoder: ResNet-26
- Decoder: original with redesigned blocks



SegNet (re-designed):

- Composed with simple ResNet blocks
- Encoder: ResNet-18
- Decoder: reflected encoder

simple ResNet block

bottleneck ResNet block

Experimental results

Dataset description:

- Name: CASIA-Iris-Lamp-V3
- Images (used/total): 4865/16212
- Subjects (used/total): 124/411
- Labeling: marking by an expert with the condition
 - Condition: all the pixels of iris area on the image <u>except</u> <u>eyelashes</u> that overlap iris are considered as belonging to iris
- Division (train/val/test): 3386/478/1001
- Only the training subset is used for optimization
- **Modified dataset** same as above but the following changes were applied to every image in the set for training:
 - Random contrast in the range [50%;150%]
 - Random brightness in the range [-20%;20%]

Measure:

• Jacard Index (intersection over Union, IoU)

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Training details:

• Library/Epochs/Optimizer: TensorFlow/200/Adam

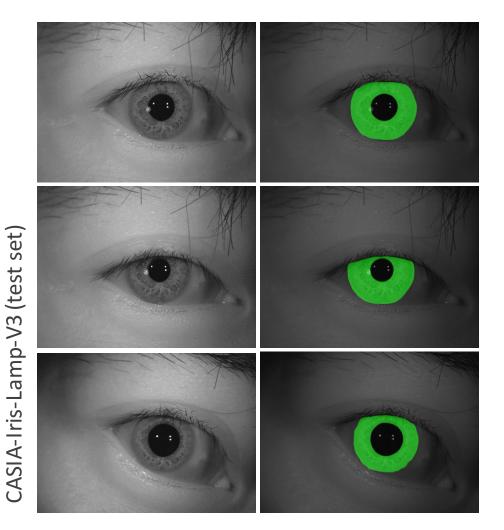
Segmentation results:

| Network | Original dataset, IoU | | Modified dataset, IoU | |
|---------|-----------------------|----------|-----------------------|----------|
| | val. set | test set | val. set | test set |
| MFCN | 0.918 | 0.919 | 0.668 | 0.676 |
| FCN | 0.930 | 0.930 | 0.884 | 0.894 |
| SegNet | 0.928 | 0.929 | 0.916 | 0.924 |

Summary:

- **Original dataset:** both FCN and SegNet showed equally well results slightly outperforming MFCN
- Modified dataset: both FCN and SegNet outperform
 MFCN by far

Experimental results: some examples



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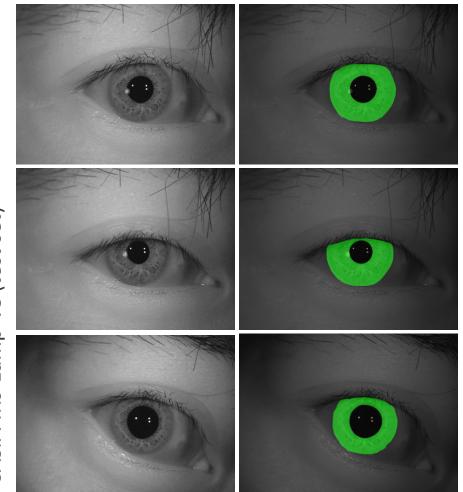
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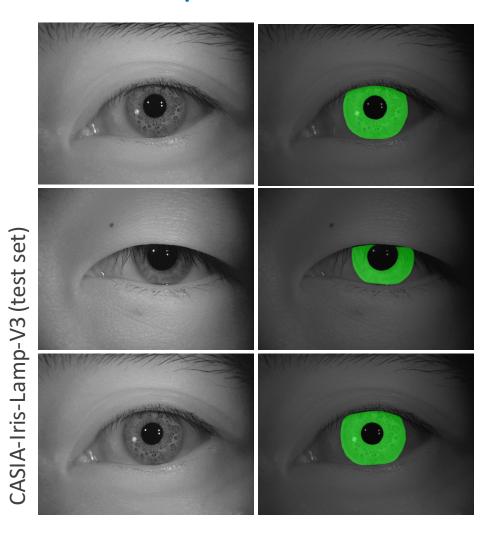
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Experimental results: more examples

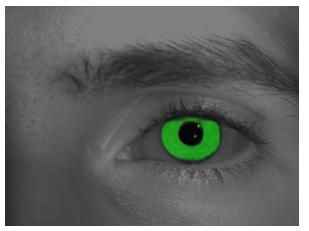




Experimental results: more examples

Random image from CASIA-Iris-V4 (yet another dataset)





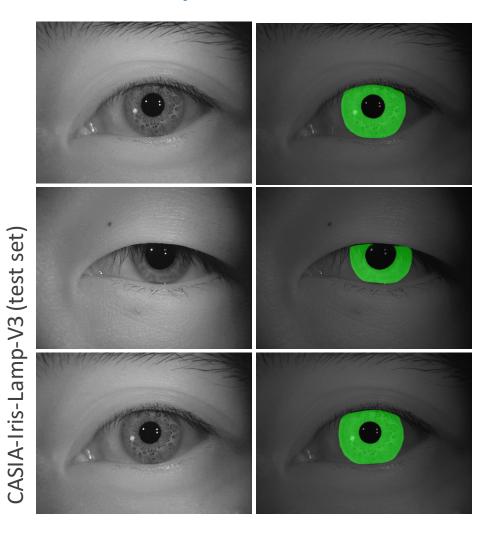


Image captured using Raspberry Pi Camera Board v2.1 (yet another dataset)



Thank you