

Project ID: ROBO022

Non-invasive Autonomous Anemia Screening Using Conjunctival Images

Introduction

- Anemia affects about 25% of the world population.
- Standard practice for diagnosis of anemia requires blood analysis.
- Pre-screening for possible anemia remains a challenging task.

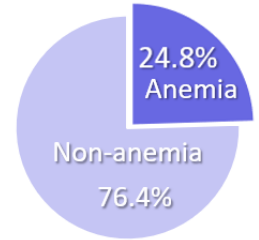


Table 1: Comparison of previous researches

Study	Sample size	Procedure	Tool	Accuracy	False Negative
Strobach et al. (1988)	50	(Physician eye check)		66.0%	60.7%
Collings et al. (2016)	101	Manual	Color palette	72.3%	43.5%
Tamir et al. (2017)	19	Manual	Fixed frame	78.9%	-
Park et al. (2020)	153	Manual	-	-	-



Fig. 1: Extraction of conjunctiva images were done manually. [1][4][7]

- **Objective:** Develop a non-invasive, autonomous and robust method for pre-screening of anemia from eye photos

Background

- WHO definition of anemia hemoglobin level
 - Female: less than 12.0 g/dl
 - Male: less than 13.0 g/dl
- Paleness of palpebral conjunctiva is commonly used for pre-screening of anemia.

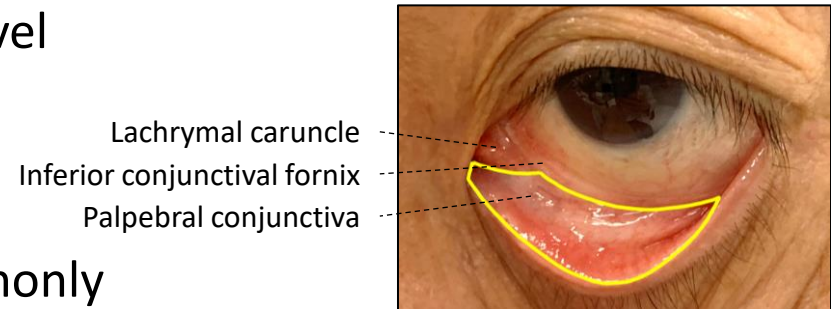


Fig. 2: Palpebral conjunctiva

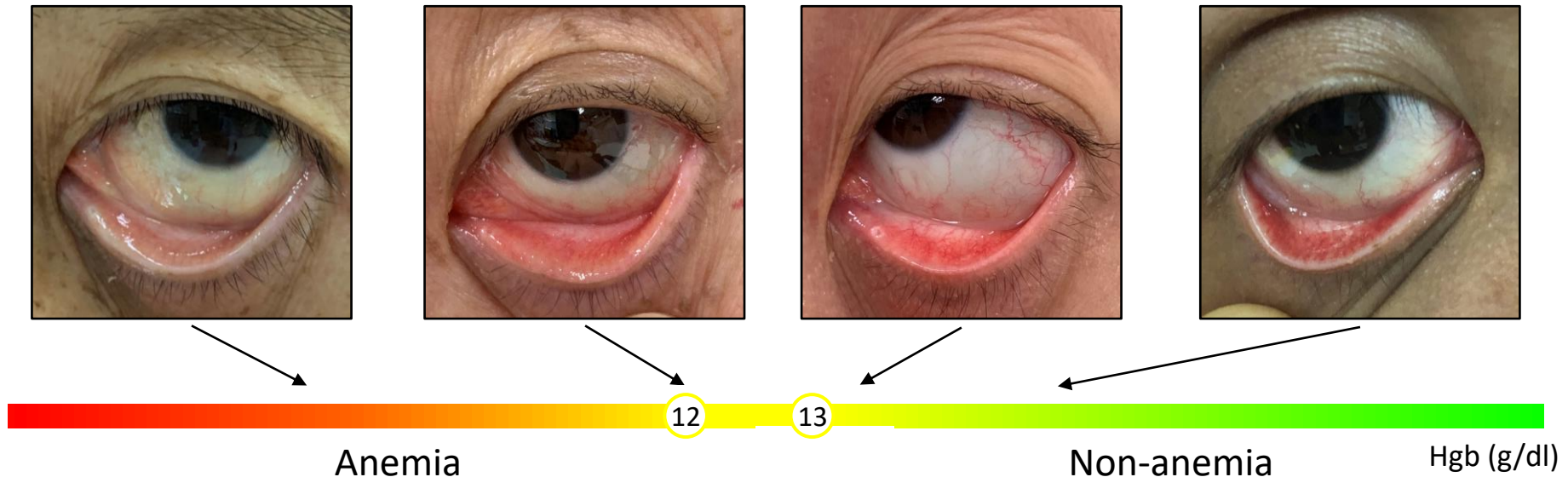


Fig. 3: Examples of palpebral conjunctiva image of anemia and non-anemia cases

Conjunctival Image Collection

- 55 subjects with 10 palpebral conjunctival images from each subject
 - blood analysis within three months before images were taken

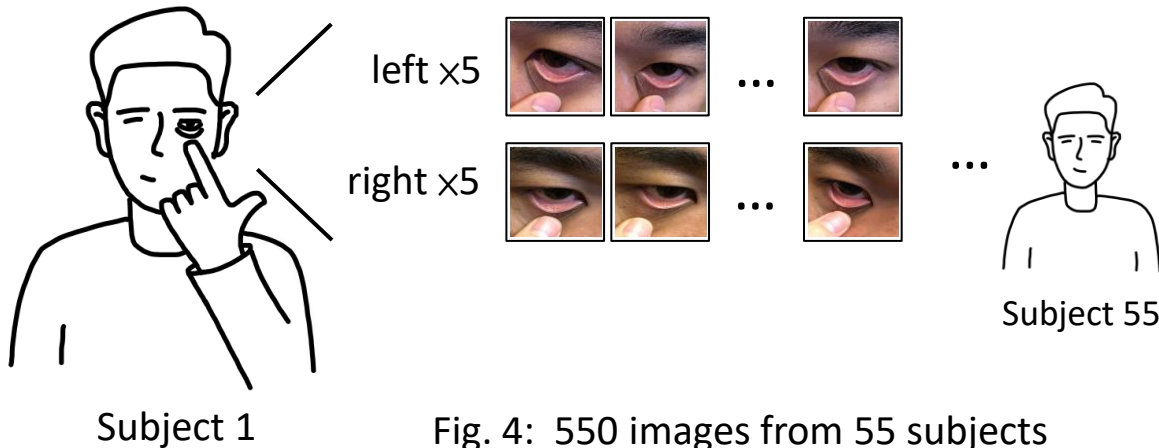


Fig. 4: 550 images from 55 subjects

Table 2: Sample size of anemia/non-anemia vs. female/male

	Anemia	Non-anemia
Female	11	14
Male	15	15

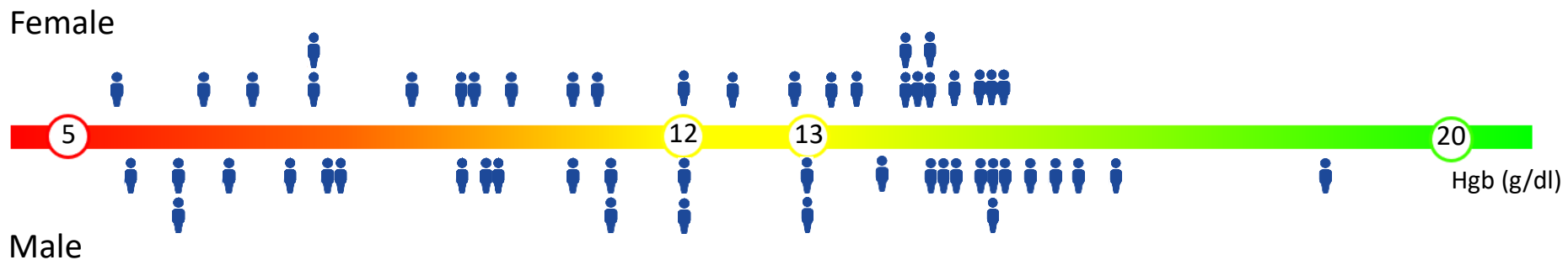


Fig. 5: Hemoglobin level of all 55 subjects

Extraction of Palpebral Conjunctiva

- Use TerausNet for palpebral conjunctiva extraction
 - UNet model with VGG11 encoder feature
 - Convolutional layers (filter size=3; stride=1, 2; padding=1)
 - Max pooling (size=2)

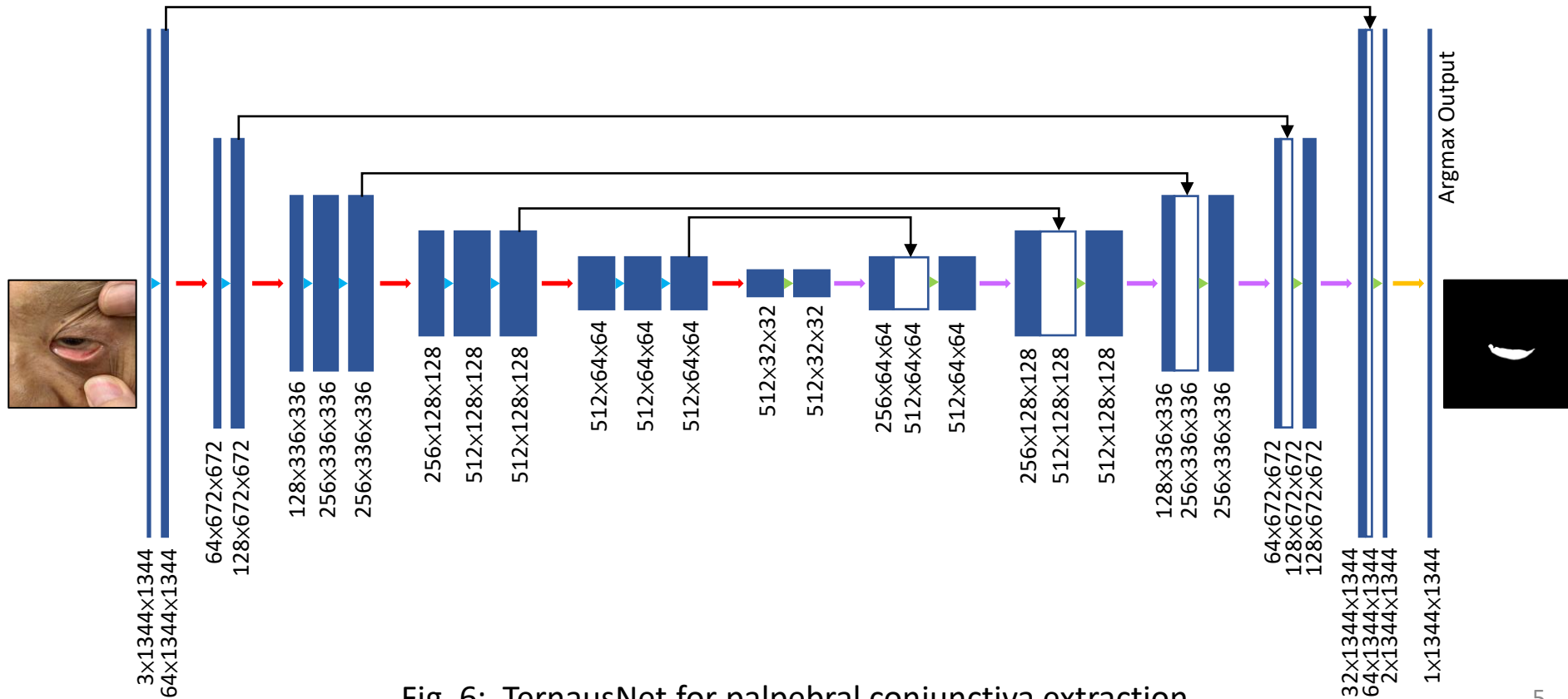
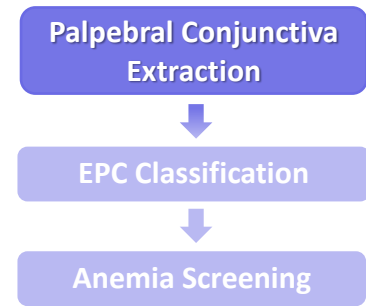


Fig. 6: TerausNet for palpebral conjunctiva extraction

Palpebral Conjunctiva Extraction Result

■ Evaluation Metric: Intersection over Union (IoU)

$$\text{IoU} = \frac{\text{Area of Overlap}}{\text{Area of Union}} = \frac{\text{RoI} \cap \text{EPC}}{\text{RoI} \cup \text{EPC}}$$

- Region of Interest (RoI) – Ground truth manually marked
- Extracted Palpebral Conjunctiva (EPC) – Output of extraction model

■ At epoch = 619, training/validation loss < 0.34, training/validation IoU > 0.87. (Fig. 7)

■ Automatic Extraction Statistics (Fig. 8)

- Best/Worst IoU = 0.973/0.676
- Average IoU = 0.903, $\sigma = 0.037$
- 92.7% of Images has IoU ≥ 0.850

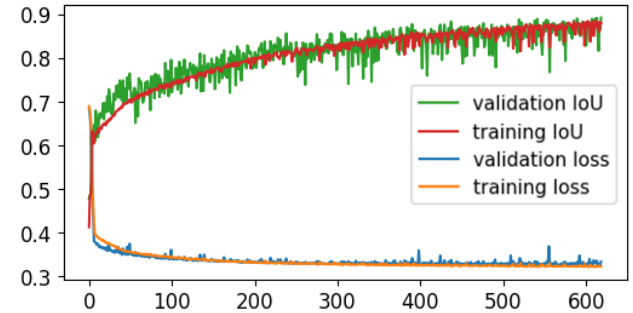


Fig. 7: Epoch vs. training loss and IoU

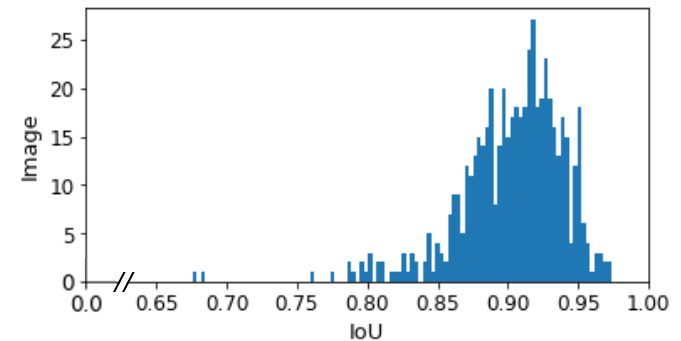


Fig. 8: IoU distribution of all EPCs

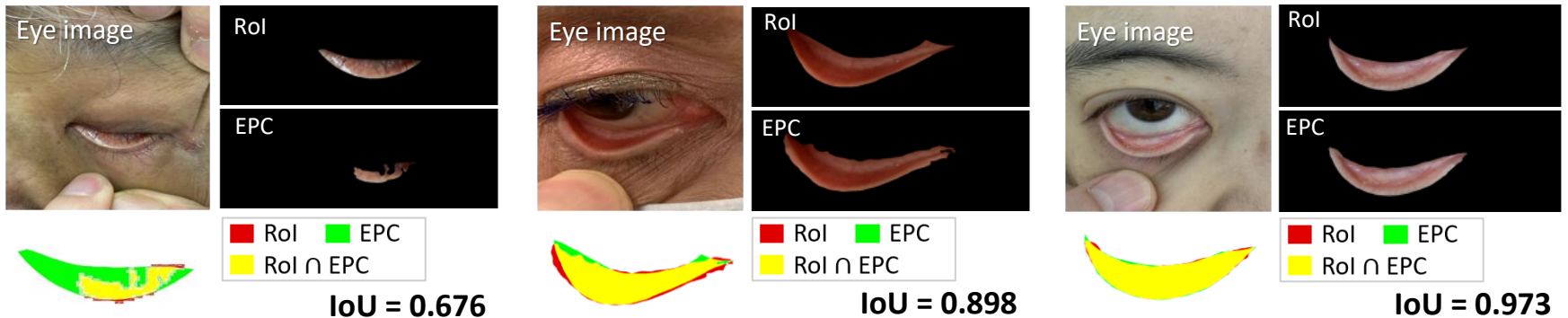


Fig. 9: Examples of palpebral conjunctiva extraction result

Feature for EPC Classification

- Use HSV color space
 - Hue indicates dominant color
 - Saturation indicates brilliance and intensity of Hue
- Equally divide the distribution profile into 10 bins
- Define 10 tuple feature as $\langle hs_1, hs_2, \dots, hs_{10} \rangle$

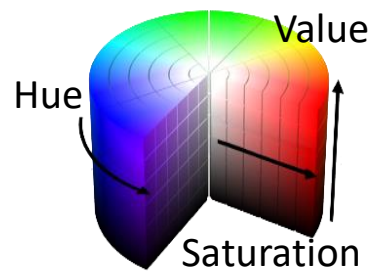
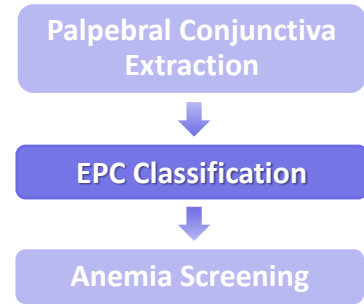


Fig. 10: HSV color space

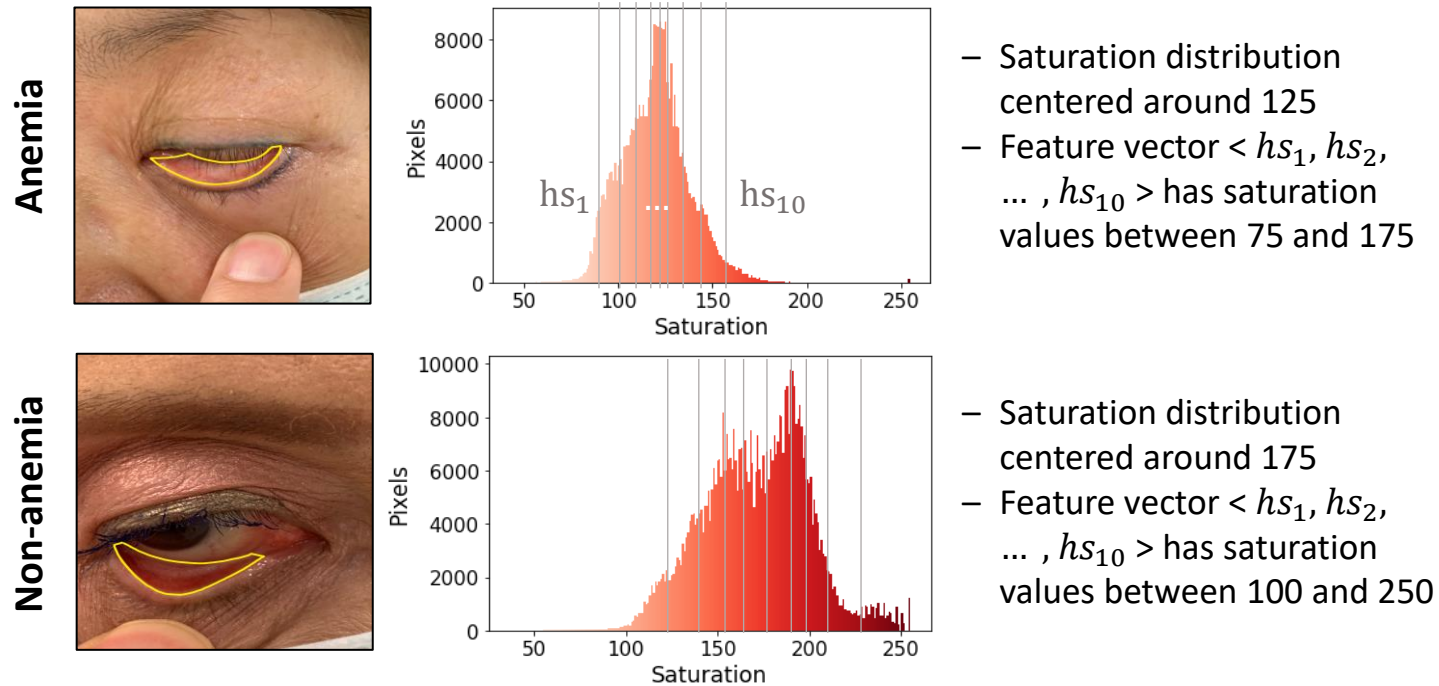


Fig. 11: 10-tuple vector computed from the HSV distribution

Supervised Classification

■ k-Nearest Neighbors (k-NN) vs. Support Vector Machine (SVM)

- k-NN with $k \in \{3, 5, \dots, \sqrt{n}\}$
- SVM with $C \in \{0.001, 0.01, 0.1, 1, 10, 100\}$,
 $\gamma \in \{0.0001, 0.001, 0.01, 0.1, 1\}$,
- Kernel functions: RBF, linear, polynomial, sigmoid

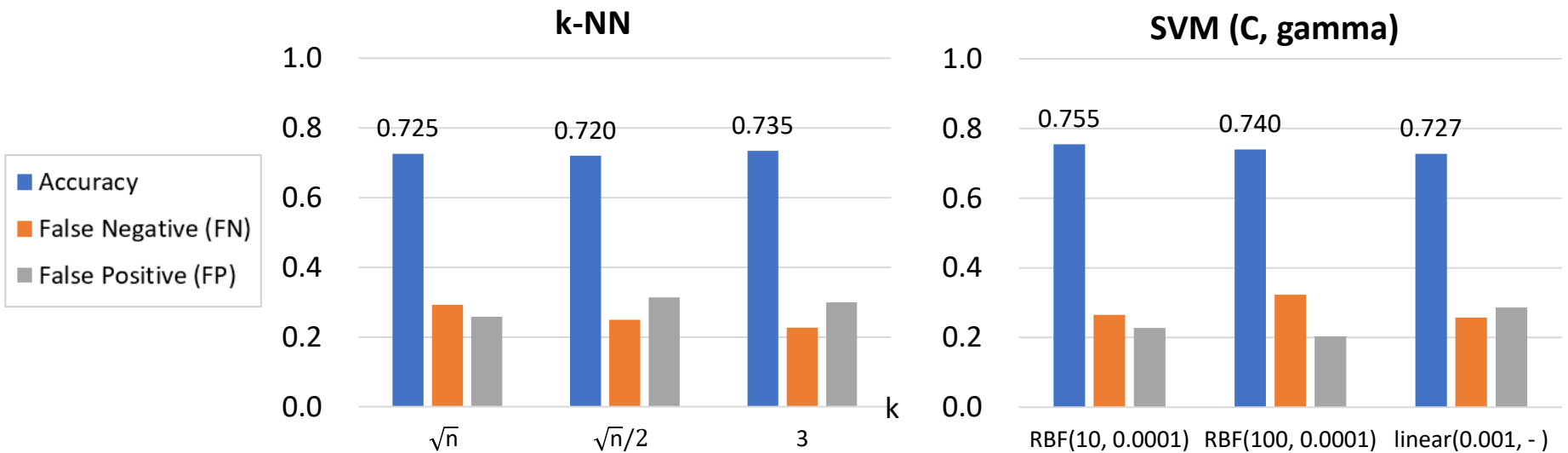
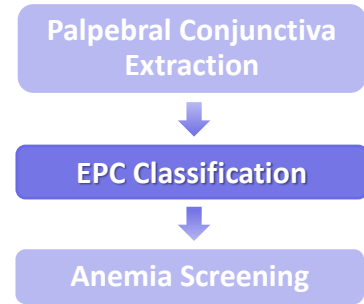


Fig. 12: Accuracy of EPC classifications

■ Best Result:

- 3-NN: Accuracy=0.735, FN=0.300, FP=0.227
- SVM(RBF, C=10, gamma=0.0001): Accuracy=0.755, FN=0.228, FP=0.265

Anemia Screening Criteria

■ Screening Criteria:

Single Image 1 Eye (SI1E)

$$P(i) \mid \exists i \in (I_L + I_R)$$

Majority Rule 1 Eye (MR1E)

$$\Sigma P(i) \geq 3 \mid i \in I_L \text{ or } \Sigma P(i) \geq 3 \mid i \in I_R$$

Majority Rule Image Set (MRIS)

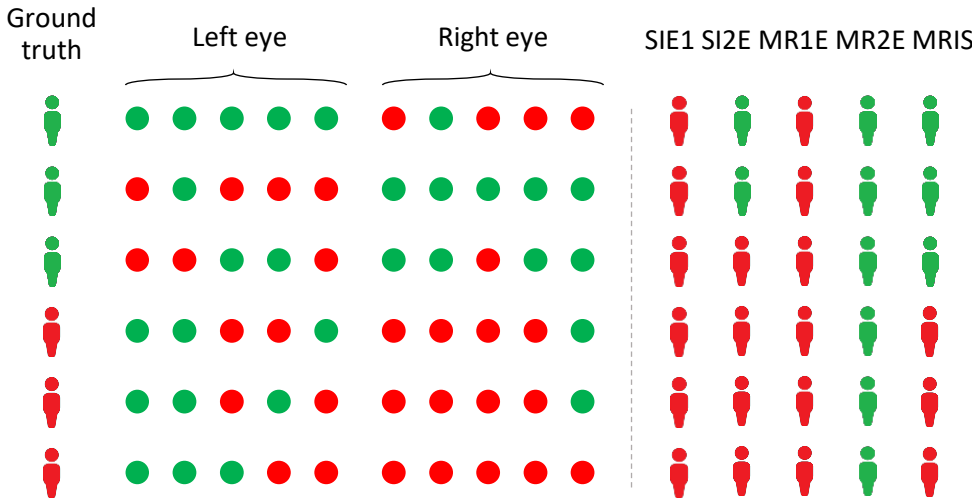
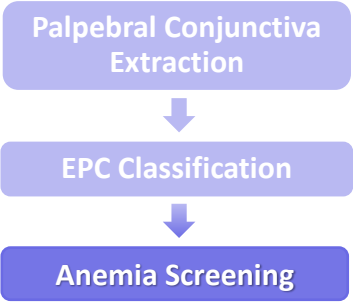
$$\Sigma P(i) > 5 \mid \forall i \in (I_L + I_R)$$

Single Image 2 Eye (SI2E)

$$P(i) \mid \exists i \in (I_L) \text{ and } P(i) \mid \exists i \in (I_R)$$

Majority Rule 2 Eye (MR2E)

$$\Sigma P(i) \geq 3 \mid i \in I_L \text{ and } \Sigma P(i) \geq 3 \mid i \in I_R$$



● Image classified as having anemia 👤 Subject classified as having anemia
● Image classified as not having anemia 👤 Subject classified as not having anemia

■ Accuracy ■ False Negative (FN) ■ False Positive (FP)

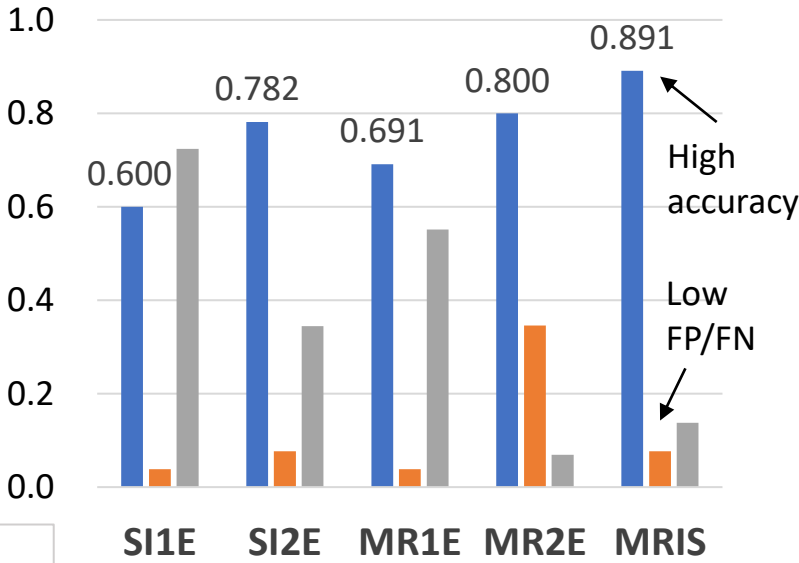


Fig. 13: Examples of applying screening criterion to classified EPCs

Fig. 14: Screening accuracies

Anemia Screening Results

Table 3: Female, male and overall experimental results

Female (Acc.=0.880)		Male (Acc.=0.900)		Overall (Acc.=0.891)	
TP = 0.909	FP = 0.143	TP = 0.933	FP = 0.133	TP = 0.923	FP = 0.138
FN = 0.091	TN = 0.857	FN = 0.067	TN = 0.867	FN = 0.077	TN = 0.862

TP: True Positive TN: True Positive FP: False Negative FN: False Negative

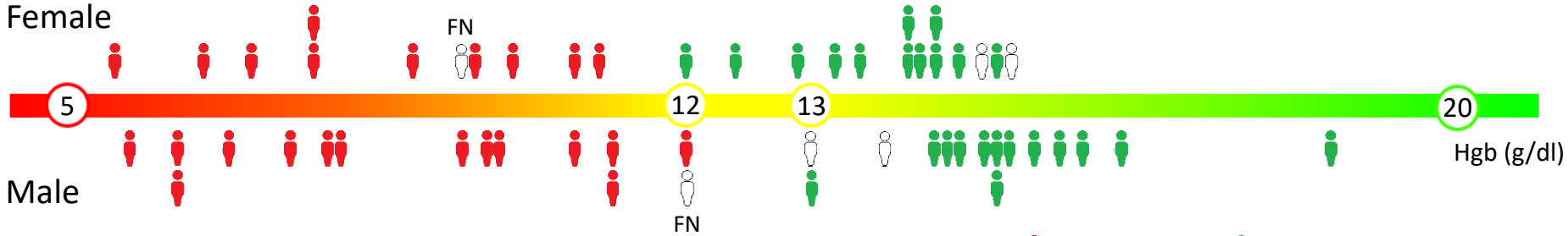


Fig. 15: Graphical representation of anemia screening results (anemia and non-anemia)

- High accuracy (TP, TN) and low FN and FP for both female and male groups.
- Two FN cases (Fig. 16) both have EPC with high IoU, but EPC mostly classified as non-anemia.

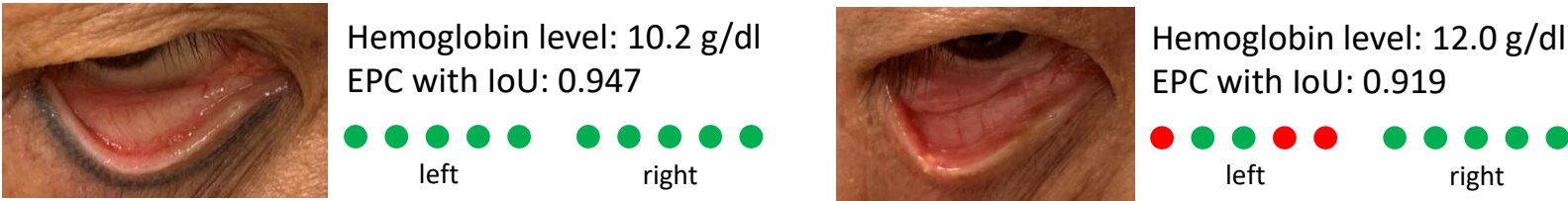


Fig. 16: Two subjects with false negative anemia screening result

Conclusions

- This research proposed an **efficient conjunctiva based anemia screening** method.
- The proposed method has anemia screening accuracy that is **10%~20% better than earlier researches**, and with much lower false positive and false negative rates.
- An anemia screening sensor can be made based on this research for use at clinics or at home.
 - To lower the false positives diagnosis and to prevent wasting medical resources
 - To facilitate early detection and early treatment before symptoms deteriorate

Automatic extraction
of palpebral
conjunctiva

Use k-NN for EPC
classification

Use MRIS for final
anemia screening

Accuracy = 0.891
False Negative = 0.077
False Positive = 0.138

References

- [1] Collings, S., et al. Non-invasive detection of anaemia using digital photographs of the conjunctiva. *PLoS One*. **2016**;11:e0153286.
- [2] Iglovikov, V., et al. TeraNet: U-Net with VGG11 encoder pre-trained on ImageNet for image segmentation. **2018**.
- [3] Mazzu, T., et al. Mobile health (mHealth) and advances in noninvasive diagnosis of anemia: an overview. *Intern J Nutrol*. **2020**;13:42-47.
- [4] Park, S.M., et al. mHealth spectroscopy of blood hemoglobin with spectral super-resolution. *Optica*. **2020**;7:563-573.
- [5] Sheth, T.N., et al. The relation of conjunctival pallor to the presence of anemia. *J Gen Intern Med*. **1997**;12:102-106.
- [6] Strobach, R.S., et al. The value of the physical examination in the diagnosis of anemia correlation of the physical findings and the hemoglobin concentration. *Arch Intern Med*. **1988**;148:831-832.
- [7] Tamir, A. et al. Detection of anemia from image of the anterior conjunctiva of the eye by image processing and thresholding. **2017 IEEE Region 10 Humanitarian Technology Conference**; 697-701.