

Retinal Vessel Segmentation (RVS)

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Background

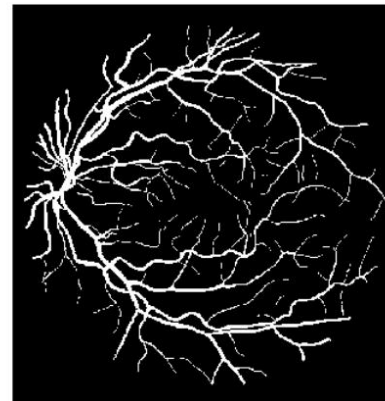
□ What is retinal vessel segmentation (RVS)?

➤ retinal images^[1]

- a digital picture of the back of your eye.
- it shows the retina, the optic disc, and blood vessels, which helps ophthalmologist find certain diseases

➤ vessel segmentation

- remove the background information and keep only the vascular information



retinal image and manual annotation segmentation image^[2]

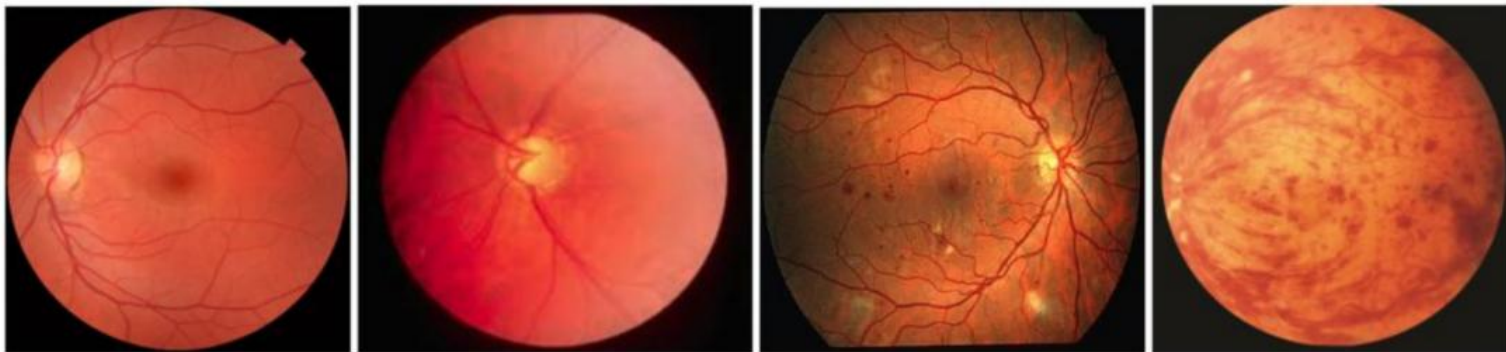
[1] <https://www.webmd.com/eye-health/what-is-retinal-imaging>

[2] J. Staal, M. D. Abramoff, M. Niemeijer, M. A. Viergever and B. van Ginneken, "Ridge-based vessel segmentation in color images of the retina," in IEEE Transactions on Medical Imaging,

Background

□ Why retinal vessel segmentation is important?

- Analyzing the length, width, curvature, bifurcation pattern and other structural characteristics of blood vessels can obtain the pathological characteristics of many diseases^{[3][4]}



(a)

(b)

(c)

(d)

Retina images under different conditions. (a) Healthy retina, (b) Glaucoma, (c) Diabetic Retinopathy, and (d) Retinal vein occlusion.

[3] WU H S, WANG W, ZHONG J F, et al. SCS-Net: a scale and context sensitive network for retinal vessel segmentation [J]. Medical Image Analysis, 2021, 70: 102025.

[4] WANG S J, YU L Q, LI K, et al. DoFE: domain-oriented feature embedding for generalizable fundus image segmentation on unseen datasets[J]. IEEE Transactions on Medical Imaging, 2020, 39(12): 4237–4248.,

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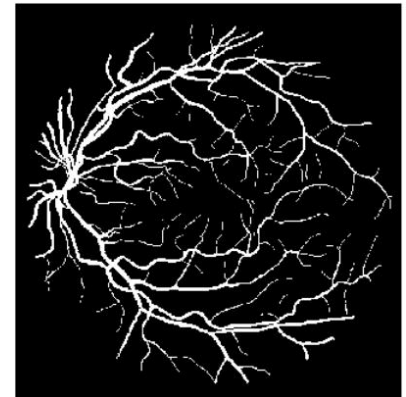
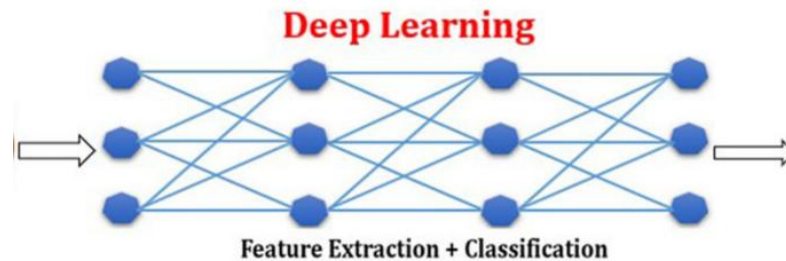
Goal statements

□ Goal

- generate the final retinal blood vessel segmentation images
 - Input: retinal images
 - Output: corresponding vessel segmentation image



Input

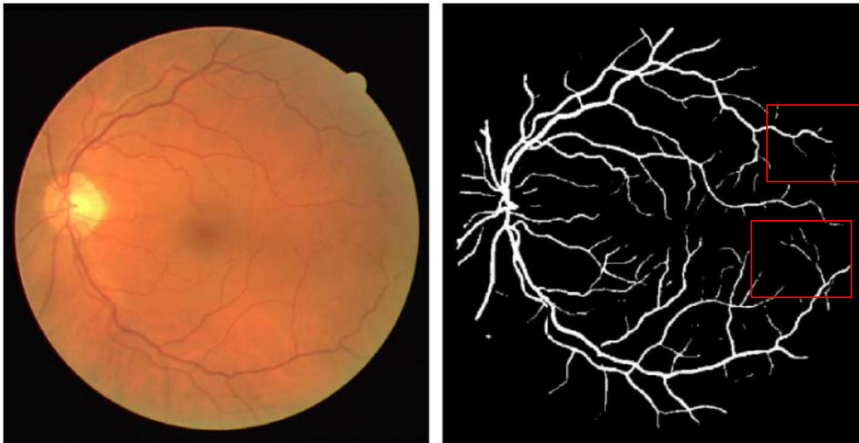


Output

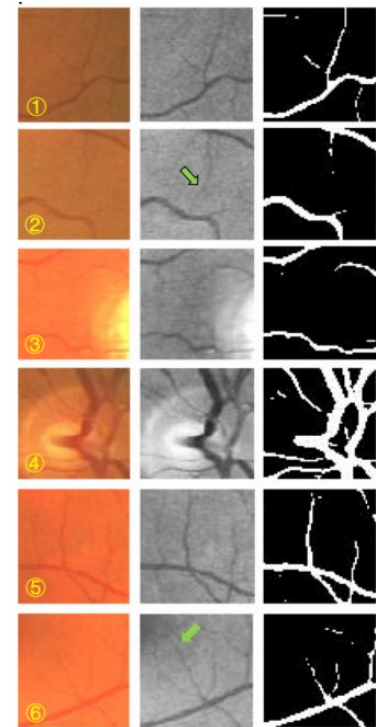
Goal statements

□ Stretch Goals

- repair some breakpoints on blood vessel segmentation images



Examples of breakpoints on the segmentation images^[5]



some patches on DRIVE dataset^[5]

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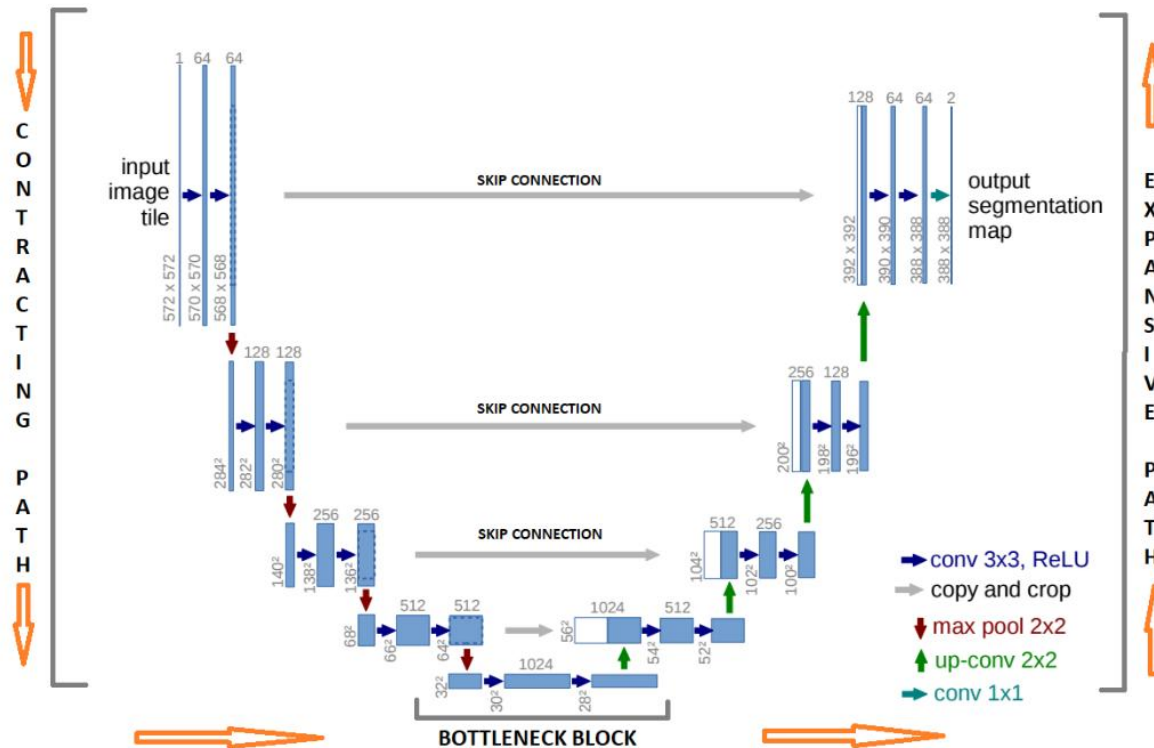
Part 3. Model

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Model I

□ U-Net^[6]

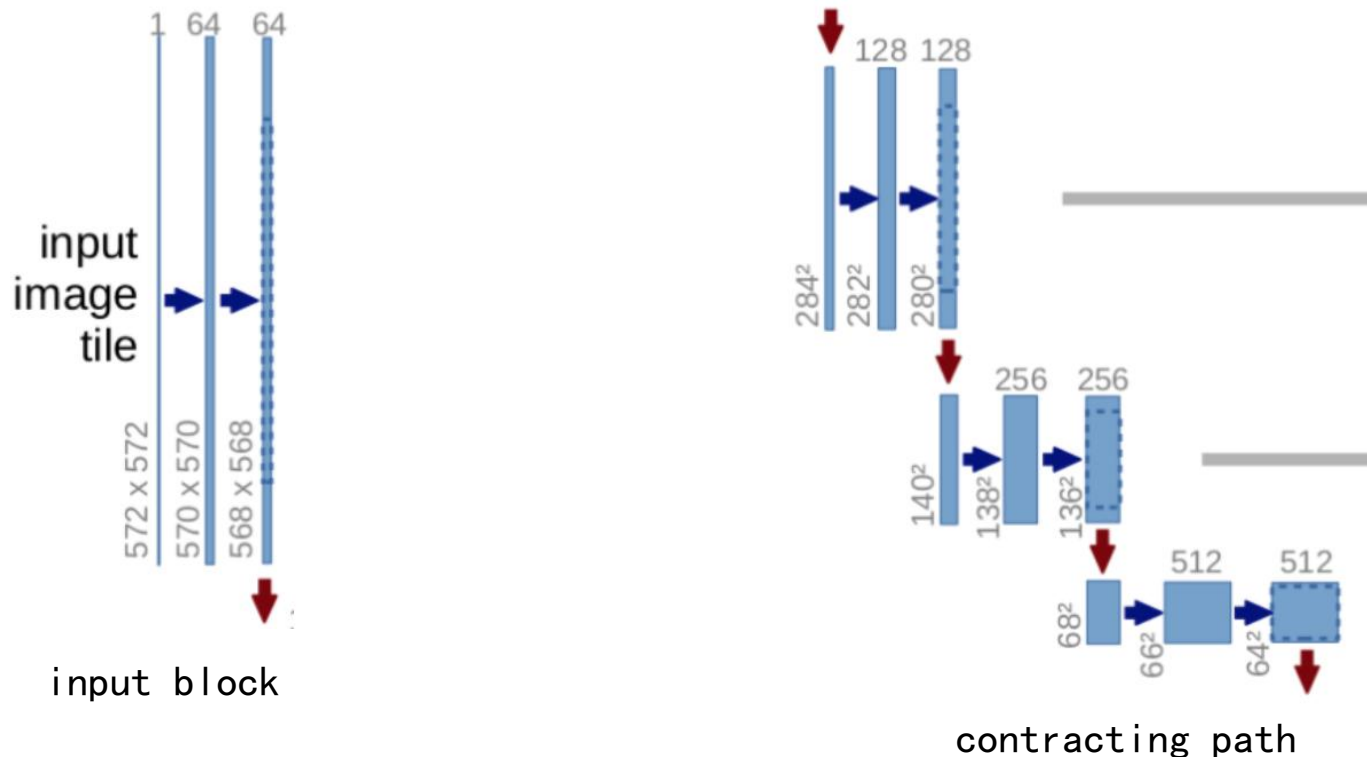
- The Contracting/Downsampling Path.
- Bottleneck Block.
- The Expansive/Upsampling Path.



Model

□ The Contracting/Downsampling Path.

- It consists of two 3x3 unpadded convolutions each followed by a rectified linear unit (ReLU) and a 2x2 max pooling operation with stride 2 for downsampling. After each downsampling operation, the number of feature channels are doubled.



Model

□ Bottleneck Block

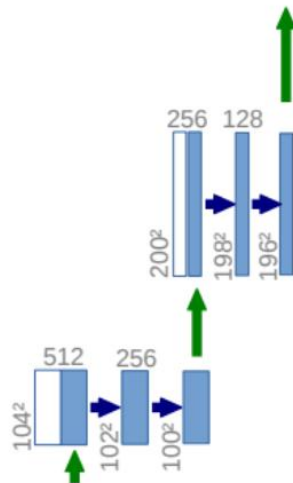
- The bottleneck block connects the contracting and the expansive paths. This block performs two unpadded convolutions each with 1024 filters and prepares for the expansive path.



Model

□ Expansive Path

- Every step in the expansive path consists of an upsampling of the feature map followed by a 2x2 convolution (“up-convolution”) using transposed convolutions, a concatenation with the correspondingly feature map from the contracting path, and two 3x3 convolutions, each followed by a ReLU. Transposed convolution is an upsampling technique to expand the size of images.



Expansive Path

Model

□ Evaluation Criteria

➤ Accuracy

- a widely used evaluation metric for the task of binary segmentation
- computes the percentage of correctly classified pixels in the whole image
- $ACC = \frac{TP + TN}{TP + FN + TN + FP}$, where TP, TN, FP and FN represent the number of true positive, true negatives, false positives and false negatives, respectively.

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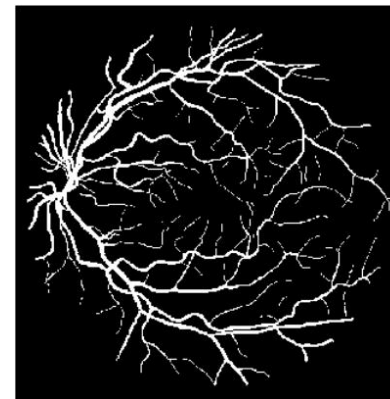
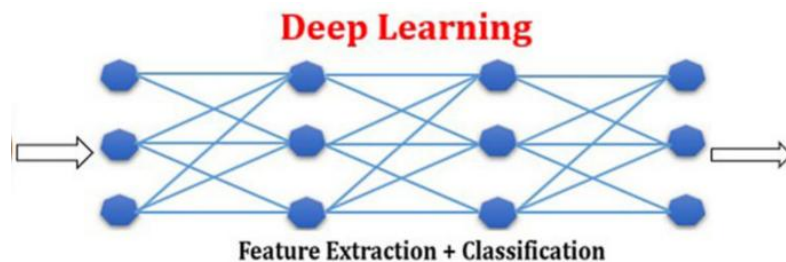
Conclusion

□ Goal

- Generate the final retinal blood vessel segmentation images
- Stretch Goals: repair some breakpoints on blood vessel segmentation images



Input



Output

□ Model

- U-Net
- Evaluation Criteria: accuracy

Questions?