40 Glaucoma Detection Using Advanced Image Segmentation Techniques

소속 정보컴퓨터공학부

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Project Introduction

Introduction

- Glaucoma is a progressive eye disease affecting the optic nerve and is a leading cause of blindness.
- Early detection through **optic nerve head (ONH) analysis** is crucial.
- Cup-to-Disc Ratio (CDR) is an important metric used in diagnosing glaucoma.
- The need for **automated segmentation** arises from the challenges of manual segmentation.

Objective

Develop a machine learning-based system for automatic segmentation of optic cup and disc to measure the Cup-to-Disc Ratio
 (CDR) for early glaucoma diagnosis.

Model Architecture

Base Architecture:

• **DC U-Net:** Combines dense connectivity with the U-Net encoder-decoder design, allowing for efficient feature reuse and precise region identification, particularly beneficial for limited datasets.

input image tile 0.055 / 865 /

Transformer

Nx

Add & Norm

Add & Norm

National

Input

Embedding

Output

Add & Norm

Multi-Head

Attention

Add & Norm

Multi-Head

Attention

Add & Norm

Masked

Attention

Outputs

(shifted right)

CBAM

Channel Attention Module

AvgPool

Spatial Attention Module

Spatial Attention Module

Channel-refined [MaxPool, AvgPool]

Spatial Attention Module

Spatial Attention Module

Spatial Attention Module

Encoder Enhancements:

- DC Blocks: Utilize dilated convolutions to expand the receptive field and capture multi-scale information efficiently.
- **CBAM:** An attention mechanism that focuses on informative regions through channel and spatial attention.

Transformer Integration:

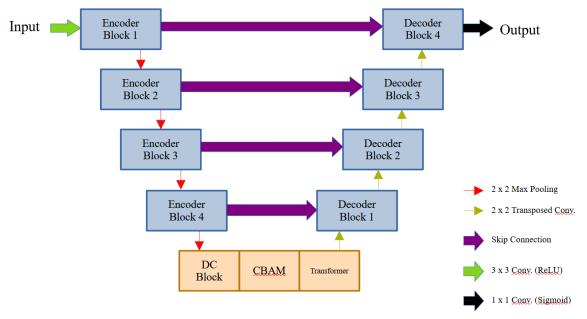
- Patch Embedding: Divides feature maps into patches for transformer processing.
- Transformer Blocks: Multi-head attention captures global context, and feed-forward networks improve representation and stability.

Decoder Enhancements:

- **Skip Connections:** Concatenate encoder features with decoder layers to retain spatial information.
- Up-sampling: Transposed convolutions progressively restore spatial dimensions.

Application of CBAM and a hybrid of the DC-UNet with a Transformer-based mode

Output: A final convolutional layer with softmax activation generates pixel-wise class probabilities for segmentation.



Results and Conclusion

Original Image

Results:

We evaluated our model against U-Net using accuracy, Intersection over Union (IoU), and Dice coefficient, and the results indicated:

Models	Pixel-Accuracy	loU	Dice Coefficient
U-Net	0.9001	-	-
U-Net with CBAM	0.9064	0.8289	0.8812
ResUNet (Used ResNet34)	0.9001	0.7928	0.8722
DC-UNet with CBAM	0.8756	0.7788	0.8467
SegNet with CBAM	0.9043	-	-
TransDC-UNet with CBAM	0.8498	0.7403	0.7765

Original Image True Mask Predicted Mask Original Image True Mask Original Image True Mask Predicted Mask

Conclusion:

This research presented advanced techniques for glaucoma segmentation but did not achieve accuracy improvements over U-Net. Notable contributions include the integration of CBAM and transformer blocks, which enhance feature attention and capture long-range dependencies.

Future Directions:

- Increase the dataset size, as 2,000 samples were insufficient.
- Explore new preprocessing and data augmentation methods to enhance model performance.

