

Paper title: Cancer Cell Segmentation Based on Unsupervised Clustering and Deep Learning

Paper link: [Cancer Cell Segmentation Based on Unsupervised Clustering and Deep Learning](#)

1 Summary

1.1 Motivation

The paper addresses the need for an effective methodology to segment, recognize, classify, and detect different types of cancer cells in medical images such as RGB, CT, and MRI. The motivation stems from the importance of early cancer detection and classification for better treatment outcomes.

1.2 Contribution

Proposed a methodology involving partial contrast stretching, Otsu method for image enhancement, K-means clustering for segmentation, median filtering for noise reduction, and Multi-Support Vector Machine (M-SVM) and Convolutional Neural Network (CNN) for classification.

Used various datasets including Brain MRI Images, SN-AM dataset for leukemia, and LIDC database for lung cancer to experiment with the methodology.

Achieved an overall accuracy of approximately 93% in detecting and classifying cancer cells.

1.3 Methodology

Preprocessing: Conversion to grayscale, noise removal using Gaussian filtering, partial contrast stretching, and enhancement using Otsu method.

Segmentation: Employed K-means clustering to divide the image into clusters, selecting the best region for segmentation.

Median Filtering: Applied to the segmented image to further reduce noise.

Classification: Used M-SVM and CNN for feature extraction and classification of the segmented image.

Trace and Detect Region: Morphological operations and MSER method used to trace and detect cancerous regions.

Area Calculation: Calculated the area of the detected cancerous region as a percentage of the total image area.

1.4 Conclusion

The proposed methodology showed promising results with an overall accuracy of 93% in detecting and classifying different types of cancer cells. The methodology involved a combination of unsupervised clustering and deep learning techniques, offering a significant improvement over existing methods.

2 Limitations

2.1 First Limitation

The methodology relies on initial knowledge of the image for expected centroids of clusters, which could pose challenges when working with diverse datasets.

Awareness to train dataset might require more attention for optimal results.

2.2 Second Limitation

The paper did not delve into the computational complexity or resource requirements of the proposed methodology.

Future research could focus on addressing these limitations and improving the robustness and efficiency of the proposed approach.

3 Synthesis

The paper presents a comprehensive methodology for cancer cell segmentation and classification using a combination of image processing techniques, unsupervised clustering, and deep learning. Despite some limitations, such as the need for initial image knowledge and training dataset awareness, the methodology achieved a notable accuracy rate of 93%. Future research can build upon this approach to further enhance its effectiveness and address the identified limitations.