

# Computerized Tomography Images Processing using Artificial Intelligence Techniques

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SCHOOL OF  
MATHEMATICS SCIENCES AND  
INFORMATION TECHNOLOGY

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# Study of Images

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There is extensive research to contribute to medicine by segmentation and identifying areas of interest (e.g., organs)

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Researches has been focuses on Computerized Tomography (CT) images to apply image processing techniques due to its essential role in detecting diseases on the liver [1], lungs [2], brain [3], skull [4], and Covid-19 disease [5].

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Artificial Intelligence (AI) has different ways to automated processes like Machine Learning (ML), Artificial Neural Networks (ANN), and Convolutional Neural Networks (CNN).

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# Objective

This work aims to present state-of-the-art image segmentation using Computed Tomography images.

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# Artificial Intelligence

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One of the most considerable advantages is on medicine by detecting anomalies or pathologies from digital images.

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The reinforcement algorithms use AI agents to choose the best actions to maximize a numerical reward signal [7].

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The impact of DL is on different studies like recognition of breast cancer [10, 11, 12, 13], extraction of brain tumor [14, 15, 3], and organ segmentation [1, 16, 9].

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Application of CNNs are: image classification [18, 19], edge detection [20], decoding facial recognition [21, 22], and analyzing documents [23, 24].

# Convolutional Neural Networks

A CNN has three fundamental layers: a convolutional layer, a pooling layer, and a fully connected layer

**Figure 1:** A CNN architecture showing the layers: convolutional, pooling, and fully connected [25].

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# Computer Vision

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There are many applications of computer vision such as machine inspection, retail, 3D model building, medical imaging, automotive safety, and biometrics [26].



# Computer Vision

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Image processing applications involve the medical, agriculture, sports, and object recognition fields.

# Computer Vision

Computerized Tomography image refers to an x-ray image. It can show the skeleton, organs, and tissues as well as any abnormalities the physician is trying to identify.

Figure 2: Example of a Computerized Tomography Image from Lung [9]

# Computer Vision

Computerized Tomography image refers to an x-ray image. It can show the skeleton, organs, and tissues as well as any abnormalities the physician is trying to identify.

Several challenges arise when processing images such as noise, complex background, dataset size, and shadows.

**Figure 2:** Example of a Computerized Tomography Image from Lung [9]

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## Related Works

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Guo et al. [2] combine thresholding and the Chan-Vense algorithm to segment many types of images.

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Wang et al. [28] propose a framework for multi-organ segmentation using Organ-Attention-Network with Reverse Connections (OAN-RCs). The OAN is a deep convolutional network that reduces the complex background

Sakboonyara et al. [29] propose an efficient method for liver segmentation. They mention the use of U-Net and the mean-shift clustering algorithm. Also, they introduce statistical thresholding to improve the accuracy.

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Zhou et al. [16] made an implementation based on multi-organ segmentation. They use partially-labeled datasets of the pancreas. The authors applied the Prior-aware Neural Network (PaNN) to overcome the context background problem.

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Yan et al. [5] use a new deep convolutional neural network called COVIDSegNet to segment infected regions and the entire lung from chest CT images. The proposed network focuses on Feature Variation (FV) block and Progressive Atrous Spatial Pyramid Pooling (PASPP).

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Approaches	Year	Method	Neural Networks	Deep Learning	Machine Learning	Other Methods
Yan et al. [5]	2021	Architecture with Feature Variation and Progressive Atrous Spatial Pyramid Pooling	x			
Dabiri et al. [30]	2020	Algorithm of L3		x		
Wang et al. [28]	2019	OAN - RCs and statistical fusion	x			
Sakboonyara et al. [29]	2019	U-Net and Mean-Shift Histogram	x			
Zhou et al. [16]	2019	PaNN	x			
Xu et al. [27]	2017	Membership function convolution neural network	x			
Guo et al. [2]	2015	Iterative Chan-Vese				x
Seng et al. [4]	2011	Reinforcement Learning			x	
Susomboom et al. [1]	2006	Pixel-based, split-and-merge, and region growing algorithms				x

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Image processing is crucial in many domains such as computer sciences, industry, security, and medicine.

In recent years, the study of Computerized Tomography (CT) images reveals its importance in the medical field by helping to detect diseases like tumors, cancer, or control organs with problems.

The automation of algorithms using advanced techniques such as Convolutional Neural Networks can improve the accuracy and performance of results.

Previous work reveals that it is necessary to use traditional image processing techniques in the data preparation stage of the algorithms.

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*Thanks!*