

# A Novel Approach for Automatic Detection of the Coronavirus Disease from CT Images Using an Optimized Convolutional Neural Network

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**Abstract:** To speed up the diagnosis of covid19 virus disease, an automatic automated approach for classifying computed tomography images of the chest is presented. The Automatic Detection Coronavirus Optimized Convolutional Neural Network (ADECOCNN) is a proposed convolutional neural network model for dividing infected, non-infected, and other pulmonary disease patients. The ADECO-CNN-optimized CNN model can categorise CT images with 99.99 % accuracy, 99.96 % sensitivity, 99.92 % precision, and 99.97 % specificity, according to extensive testing. CT imaging of the chest is highly recommended in the early detection of disease since it determines the amount and nature of the lesion. It also evaluates alterations that aren't obvious on X-rays of the chest. The shape, quantity, distribution, density, and accompanying symptoms of a lesion are all examined using CT imaging. Chest CT imaging can serve as a critical early warning indicator of being a COVID-19 carrier and can be extremely useful for patients who are experiencing any COVID-19-related symptoms.

**Keywords:** Convolutional Neural Network (CNN), COVID, Computed Tomography (CT) Images, Deep Learning, etc.

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