Deep Learning Based Covid-19 Detection from CT Images

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ABSTRACT: The increasing number of cases of confirmed coronavirus disease covid-19 in China is striking. The purpose of this study was to investigate the relation between chest CT findings and the clinical conditions of covid-19 pneumonia. Among those who develop symptoms, most (about 80%) recover from the disease without needing hospital treatment. About 15% become seriously ill and require oxygen and 5% become critically ill and need intensive care. Complications leading to death may include respiratory failure, acute respiratory distress syndrome (ARDS), sepsis and septic shock, thrombo embolism, and/or multiorgan failure, including injury of the heart, liver or kidneys. In rare situations, children can develop a severe inflammatory syndrome a few weeks after infection. Proposed method not only detects the availability of NOVEL CORONA but also it tracks the treatment progress. In Second generation, number of architectures or algorithms is present for classification problem. In other languages we have to start from scratch, but for Matlab and Python this is another case. Simply calling those functions and changing the input argument, you test. Due to available built in commands, design and development time get reduced. With minimal Mathematics behind deep learning, we can design and test various architectures of neural network.

KEYWORDS: Covid-19, CNN, CT images.

I. INTRODUCTION

1. Background

Novel Corona virus has taken large attention of the entire globe. Every one joined the battle to fight the Corona virus. As a part of society we develop the software for Corona detection using AI; specially designed for front-line use to help doctors to detect and monitor the disease efficiently and effectively. Patients with confirmed covid-19 pneumonia have typical imaging features that can be helpful in early screening of highly suspected cases and in evaluation of the severity and extent of disease. Most patients with covid-19 pneumonia have ground-glass opacities or mixed ground-glass opacities and consolidation and vascular enlargement in the lesion. Lesions are more likely to have peripheral distribution and bilateral involvement and be lower CT images predominant and multifocal. CT involvement score can help in evaluation of the severity and extent of the disease [1].

Some survey [2] analyzed that the sensitivity of RT-PCR testing at various tissue sites, Broncho alveolar lavage fluid specimens demonstrated the highest positive rates of at 93% (n = 14). This was followed by sputum at 72% (n = 75), nasal swabs at 63% (n = 5), fibro bronchoscope brush biopsy at 46% (6/13), pharyngeal swabs at 32% (n = 126), feaces at 29% (n = 44) and blood at 1% (n = 3). The authors of that study pointed out that testing of specimens from multiple sites may improve the sensitivity and reduce false-negative test results. The letter examined 1070 specimens that were collected from 205 hospitalized patients with confirmed COVID-19 in China.

In another study published in Radiology, [3] investigators found chest CT achieved higher sensitivity for diagnosis of COVID- 19 as compared with initial RT-PCR from pharyngeal swab samples. This retrospective study analyzed 1014 hospitalized patients with suspected COVID-19 in Wuhan, China with patients undergoing both serial RT- PCR testing and chest CT. Using RT-PCR results as reference standard, the sensitivity, specificity, and accuracy of chest CT in diagnosing COVID-19 were 97% (n = 580), 25% (n = 105), and 68% (n = 685), Respectively.
2. **Motivation**

The Corona outbreak has put significant pressure on imaging departments, to test hundreds of peoples per day. Patients and doctors typically have to wait a few hours to get the CT results, but our system is improving the CT diagnosis speed for each case; and each minute saved is critical to decrease the chance of cross-contamination at the hospital.

The shortage of strict laboratory requirements for the use of the RT-PCR detection kit, to confirm the 2019-nCoV diagnosis, is a major problem. Proposed system can help with limited medical resources to immediately screen out suspected Coronavirus-infected patients for further diagnosis and treatment.

The battle against this epidemic is one being fought by all clinicians and countries, and we as a part of society is fully committed to support these efforts, wherever needed, and aspires to “Use the most advanced AI technology to serve the most fundamental needs.”

3. **Need**

To develop a system that detects Novel Corona symptoms with maximum precision and with minimum processing time to avoid the spread of covid-19 virus in community.

4. **Organization of Project Paper:**

In this paper, chapter 1 includes background of the project, Motivation, Need and project undertaken. Chapter 2 includes background and literature survey for this project concept. Chapter 3 includes specifications, problem statement, and software and hardware requirements. Chapter 4 includes design part. In that System Analysis architecture diagram, all UML diagrams are there. In chapter 5 there is implementation part of project. In that, all required algorithm for project such Convolutional Neural Network (CNN).

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**II. LITERATURE REVIEW**

2.1. Background

A novel corona virus has resulted in person-to-person transmission but as far as we know, it causing corona virus disease 2019 (COVID-19) can also be from an asymptomatic carrier with no covid symptoms. Till now there is no Paper about any clinically approved antiviral medicine or vaccines that are effective against covid-19. For novel corona Patients CT scan have some symptoms such as, Reticulation, Ground Glass Opacities and Consolidation of CT images tissue. To detect such symptoms is challenging. Traditional methods such as segmenting the Region of interest t, then extracting features of those part and then classify using some retrained classifier is time consuming process. Recent studies on Deep learning change the process of detection by providing the image to this second generation neural networks which are capable of extracting features and classifying itself. Also CORONA is spreading fast, so this overall development should be completed within little duration. The Corona outbreak has put significant pressure on imaging departments, to test hundreds of peoples per day. Patients and doctors typically have to wait a few hours to get the CT results, but our system is improving the CT diagnosis speed for each case; and each minute saved is critical to decrease the chance of cross-contamination at the hospital. It has spread rapidly across the world, bringing massive health, economic, environmental and social challenges to the entire human population. At the moment, WHO recommends that people should wear face masks to avoid the risk of virus transmission and also recommends that a social distance of at least 2m is maintained between individuals to prevent person-to-person spread of disease.

2.2. Literature Survey

Wei Zhao, Zheng Zhong, Xingzhi Xie, Qizhi Yu and Jun Liu “Relation Between Chest CT Findings and Clinical Conditions of Coronavirus Disease (COVID-19) Pneumonia: A Multicenter Study,” American Journal of Roentgenology: 1-6. 10.2214/AJR.20.22976 Wei Zhao et.al [1] states that In December 2019, a series of cases of pneumonia of unknown causation emerged in Wuhan, Hubei, China, and quickly raised intense attention around the world [1]. A novel bat-origin coronavirus, 2019 novel corona-virus, was identified by means of deep sequencing analysis [2]. The virus, named severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) [3], is phylogenetically closest to bat SARS-like coronavirus but in a separate clade, which means that a novel coronavirus
is spreading [2]. As of February 17, 2020, 72,436 laboratory-confirmed cases were consecutively papered in 31 provinces (municipalities and regions) in China, including 11,741 severe cases, 1868 fatal cases, and 6242 suspected cases [4]. The study had several limitations. First, only 101 patients with confirmed COVID-19 were included; negative results and infections with other viruses were not included in the analyses. Comprehensive investigation of the imaging features of patients with negative results and other virus infections may help us to differentiate COVID-19 pneumonia from other CT images infections and then to screen patients with highly suspected cases. Second, we did not evaluate follow-up CT findings in our study. Exploring the CT changes and comparing them with the clinical parameters may help us monitor and predict outcome and support clinical decision making.

Wang W, Xu Y, Gao R, “Detection of SARS-CoV-2 in Different Types of Clinical Specimens, JAMA” Published online March 11, 2020. doi:10.1001/jama.2020.3786 Wang W et.al [2] proposed that in this examination, SARS-CoV-2 was recognized in examples from numerous destinations of 205 patients with COVID-19, with lower respiratory lot tests regularly testing positive for the infection. Significantly, the live infection was identified in defection, inferring that SARS-CoV-2 might be sent by the fecal course. A little level of blood tests had positive PCR test results, proposing that contamination now and then might be fundamental. Transmission of the infection by respiratory and extra respiratory courses may help clarify the fast spread of illness. What's more, trying of examples from different destinations may improve the affectability and diminish bogus negative test outcomes. Two more modest investigations detailed the presence of SARS-CoV-2 in butt-centric or oral swabs and blood from 16 patients in Hubei Province, and viral burden in throat swabs and sputum from 17 affirmed case.

Shashi Yadav “Deep Learning based Safe Social Distancing and Face Mask Detection in Public Areas for COVID-19 Safety Guidelines Adherence” International Journal for Research in Applied Science & Engineering Technology (IJRASET) 2020 Shashi Yadav et.al [5] introducing to establish safe climate that adds to public security, we propose a productive PC vision put together methodology centered with respect to the continuous mechanized observing of individuals to recognize both safe social removing and face veils in broad daylight puts by executing the model on raspberry pi4 to screen action and identify infringement through camera. After identification of penetrate, the raspberry pi4 imparts ready sign to control focus at state police base camp and furthermore offer alert to public. In this proposed framework current profound learning calculation have been blended in with mathematical methods for building a vigorous modular which covers three parts of location, following, and approval. Accordingly, the proposed framework favors the general public by saving time and helps in bringing down
the spread of Covid. It tends to be actualized adequately in current circumstance when lockdown is facilitated to investigate people in open social events, shopping centers, and so on. Robotized investigation decreases labor to assess people in general and furthermore can be utilized in any place.

**List of Modules and Algorithm:**

**Methodology:**
A breakthrough in building models for Covid-19 Lung CT Images classification came with the discovery that a convolution neural network (CNN) could be used to progressively extract higher- and higher-level representations of the Lung image content. Instead of pre-processing the CT image to derive features like textures and shapes, a CNN takes just the CT image's raw pixel data as input and "learns" how to extract these features, and ultimately infer what object they constitute.

To start, the CNN receives an input feature map: a three-dimensional matrix where the size of the first two dimensions corresponds to the length and width of the images in pixels. The size of the third dimension is 3 (corresponding to the 3 channels of a CT image: red, green, and blue) though it is gray (by replicating the same channel three times).

Proposed system contains four modules viz., User and System:-

**System Modules:**
- Input train data
- Preprocessing
- Segmentation
- Update weight and bias
- Score Calculation
- Prediction
- Prediction result

**User Modules:**
- Input test data
- Prediction
- Prediction result

**III. ALGORITHM**

**Convolutional Neural Networks**
A breakthrough in building models for Covid-19 CT images classification came with the discovery that a convolutional neural network (CNN) could be used to progressively extract higher and higher-level representations of the CT images content. Instead of pre-processing the CT image to derive features like textures and shapes, a CNN takes just the CT image's raw pixel data as input and "learns" how to extract these features, and ultimately infer what object they constitute.
3.1 Convolutional Layer:-
Convolutional layers are the layers where filters are applied to the original image, or to other feature maps in a deep CNN. This is where most of the user-specified parameters are in the network. The most important parameters are the number of kernels and the size of the kernels.

3.2 RELU Function:--
ReLU is the max function(x,0) with input x e.g. matrix from a convolved image. ReLU then sets all negative values in the row and all other values are kept constant.

3.3 Pooling Function:--
Max pooling is a pooling operation that selects the maximum element from the region of the feature map covered by the filter. Thus, the output after max-pooling layer would be a feature map containing the most prominent features of the previous feature map.

3.4 Fully Connected Layer:--
Fully Connected Layer is simply, feed forward neural networks. Fully Connected Layers form the last few layers in the network. The input to the fully connected layer is the output from the final Pooling or Convolutional Layer, which is flattened and then fed into the fully connected layer.

IV. SYSTEM ANALYSIS PROPOSED ARCHITECTURE

Proposed method takes CT images as input. It process on input image using median filter. After that it extracts the region of interest. Then our deep dense network will look for any symptoms for corona such as glass opacity. If it found any of the trained symptom then it will gives result for COVID costiveness. The accuracy of any Deep Network depends on the training dataset. For our model we used normal CT images from LIDC Dataset and Corona image are taken from web. As there are privacy issues of corona images, also in this situation no one is ready to make those dataset public.

A new artificial intelligence-powered deep learning model will help radiologists to distinguish COVID-19 from community-acquired pneumonia and other CT images diseases in chest CT imaging. Proposed model will help with a growing workload to be able to focus on complex clinical cases. With its deep learning algorithms, it automatically highlights abnormalities, segments anatomies. Proposed model gives better accuracy for Dataset. For real time imagery large dataset is needed.
Registration Form:
V. CONCLUSION

- Faster, more informative CT scans: Proposed method not only detects the availability of novel corona but also it tracks the treatment progress.
- Ability to experiment with new deep learning architects
- Highly reduced programming time: Due to the available built-in commands, design and development time get reduced.

REFERENCES

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