

Categorization and Segmentation of Covid by Ultrasonography Using U-Net

Y. Dharani Sree, B.Sony M. Tech(Ph.D), B.Lakshmi Devi M. Tech(Ph.D)

Student School of Engineering and Technology Sri Padmavati Mahila Visvavidyalayam, Tirupati
Assistant Professor School Of Engineering and Technology Sri Padmavati Mahila Visvavidyalayam, Tirupati
Assistant Professor School Of Engineering and Technology Sri Padmavati Mahila Visvavidyalayam, Tirupati

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ABSTRACT

In the wake of the current COVID19 outbreak, research has started to examine DL-based solutions for the assisted detection of lung illnesses. Deep learning (DL) has shown promise in the field of medical imaging. This work analyses the application of deep learning algorithms to analyse lung ultrasonography (LUS) data, in contrast to earlier studies that concentrated on chest computed tomography (CT) scans. With reflected labels that reveal the degree of sickness at the frame and pixel levels, this dataset of lung ultrasonography pictures is completely unique and is drawn from several hospitals and online sites. Deep networks with a Spatial Transformational Network (STN) are introduced at the frame level to automatically assess the degree of disease in lung ultrasound pictures. In the system that is being suggested, a deep network

Key words: Deep Learning, CNN, STN, LUS

I. INTRODUCTION

1.1 GENERAL

Wheat COVID-19 pneumonia has the potential to quickly turn into a life-threatening condition. respiratory distress syndrome is too common Over 1,000 COVID-19 patients' radiological images revealed symptoms similar to acute respiratory distress syndrome (ARDS), such as consensual and multi-lobar glass ground pacifications (primarily posteriorly and/or scattered peripherally). In order to help individuals who have been affected by the coronavirus, computed tomography (CT) of the chest has been suggested as a potential diagnostic method. Computed tomography (CT) examination can be completed much more quickly than real-time reverse transcriptase-polymerase chain reaction (RT-PCR), which might take up to a day and numerous tests to

produce accurate results. However, there are a number of drawbacks to adopting chest CT; it is expensive, patients will be exposed to radiation, it requires extensive image cleaning, and it depends on Ultrasound imaging has recently gained popularity, availability, affordability, security, and real-time imaging technology. For the identification and treatment of acute respiratory illnesses, lung ultrasonography (LUS) is being used increasingly often in point-of-care settings. It can sometimes be more accurate than a chest X-ray at detecting pneumonia. Recently, doctors in the emergency room have highlighted LUS imaging as a tool for identifying COVID-19. In order to determine and control the respiratory efficacy of mechanical ventilation, the findings lead to specific biomarker images and characteristics of LUS for COVID-19 affected individuals. Due to its extensive use and comparatively inexpensive cost, ultrasound imaging is an incredibly successful technology in situations where patient inflow exceeds the capacities of imaging equipment. minimal costs Due of its low price, it can be used in countries with weak to average economies as well. However, due to the steep learning curve and difficulty of the task, ultrasound image interpretation is prone to errors.

Automatic image analysis using deep learning (DL) and machine learning (ML) methods for tissue According to recent data, ultrasound is employed for reconstruction, classification, regression, and segmentation. using DL to assist doctors with identifying diseases In this study, point-of-care lung ultrasonography imaging patterns associated with COVID-19 are presented. It focuses on three major LUS imaging goals: segmentation of pathological artefacts, frame-based classification, and video-level grading. The first task comprises classifying each frame of a lung

ultrasonography picture according to the severity of the disease and its progression in one of four categories that must be determined by the scoring system. Grading aims to forecast a grade using the same grading system for the entire frame sequence Video-level. Segmentation, on the other hand, is the classification of pathological artefacts at the pixel level inside each frame.

In many aspects, the state-of-the-art automated LUS research has been advanced image processing to help clinical practitioners diagnose COVID-19 disorders. suggest creating an expanded and fully annotated database version in the ICLUS-DB. The dataset must include labels on the recommended 4-levels scale at both the frame and video levels. Additionally, it provides a pixel-level annotated subset of LUS images that can be used to develop and access semantic segmentation algorithms as well as a novel deep architecture that enables us to locate regions with abnormal artefacts and estimate the score associated with a single LUS image in a weakly supervised manner. Our network utilises Consistency losses in the Spatial Transformers Network (STN) to localise illness patterns, such Use soft ordinal regression loss as well as for reliable score estimation. provide a simple and lightweight method uniforms to compute the score linked with frame-level predictions for a video sequence as well as for robust score estimation, use soft ordinal regression loss. To compute the score associated with frame-level predictions with a video sequence and present a basic and lightweight approach uniforms. The subject of automatic problematic artefact localization is addressed by evaluating the state-of-the-art segmentation performance algorithms derived from fully convolutional architectures. Finally, carry out a detailed assessment of our methods for all jobs, demonstrating that the suggested fixes can precisely forecast and pinpoint COVID-19 imaging biomarkers. Computer vision tasks like semantic segmentation and object detection have both been accomplished successfully by DL. Deep learning offers datasets and source code. Due to these successes, Deep learning is now increasingly frequently used in medical applications, such as the segmentation of biological images and the detection of pneumonia from chest X-rays. The availability of information, the potential of deep learning to assist with and automate preliminary diagnosis, all of which plays a crucial in the sector of medical .

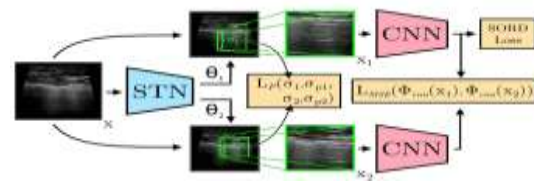


Fig. 1.1 Frame-based score prediction architecture

In the aftermath of the current epidemic, recent work has mostly concentrated on the COVID-19 detection from CT. A form of U-Net network is used on successive occasions to drop a box of bounding for each concerning coronavirus pneumonia area. To reduce the likelihood of false positive detections, CT scans are operating using quadrant-based filtering. On the other hand, a concept for a region based on thresholds uses the Inception network to categorise each recommended Region of interest (RoI) after obtaining the region of interest from the input scan. The VNET-IR-RPN model was created for diagnosis in a pre-trained for pulmonary TB[8]. A 3D version is used to classify each Region of Interest (RoI) and is used to suggest RoI input to CT. Nevertheless, there are only a few articles using DL on LUS photographs in the literature. In, a method for poorly guided localisation and categorization of lung diseases is developed. It is suggested to use LUS images for COVID-19-related pattern discovery using a weakly-supervised frame-based classification and segmentation approach. A segmentation map of the input image that is only weakly directed is produced after training Efficient net to recognise Class activation maps (CAMs) in COVID-19 LUS pictures. Our work differs in a variety of ways from all earlier efforts. To start, while CAMs are used for localization, this research's methodology and use of STN allowed for the development of a soft localization technique. When completing a classification task in the second, and scoring well on the ordinal regressions, which not only predicts a score of disease severity but also the existence of coronavirus-related artefacts. Third, a video-level prediction type constructed on top of the frame-based approach, which advances all earlier approaches. Finally, using various STN of the state of neural network designs for image segmentation, we provide a straightforward yet efficient method for anticipating segmentation masks. To make the results easier to understand, uncertainty estimations are added to the model's forecasts.

1.1.1 Detection of Covid-19 Using CNN

CNN The four layers that make up a convolutional neural network (CNN) calculation are the input layer, the convolution layer, the

pooling layer, the flatten layer, and the dense layer. Images are seen as an information layer contribution. A ready-made flexible net serves as the information layer in this scenario. In the Convolution layer, images would be transformed into matrix designs. In this case, the lattice size is 1024 x 1024. (lines X columns). The pooling layer will retain its mathematical properties. To deal with converting mathematical contributions to double information, SoftMax is used (managed learning calculation). The SoftMax layer will adjust the mathematical data to double it.

1.1.2 Wheat Disease Detection using CNN

For a set of data comprising of 227 images of wheat plants having healthy, sick and snailattacked leaves, a convolutional neural network classifier is used. Alex-Net is used to train classifiers which is based on transfer learning. The accuracy of the aforementioned architecture is 91.23 percent of success rate after training, although it can only forecast if the plant is ill or not ill. The researchers gathered 500 photos of 10 distinct wheat-leaf and stem-diseases. They created a Le-Net and Alex-Net inspired architecture that scored 95.48 percent on the test set. They employed different pre-processing steps such as image re-scaling to 512*512, normalization, PCA, and whitening because the data was so small.

1.1.2 CNN: Convolutional Neural Network

A convolutional neural network (CNN), a type of neural network specifically created for processing pixel data, is utilised in image recognition and processing. To analyse the data and separate significant aspects from the grid-like structure, it includes several layers. The primary goal of the convolution neural network approach is to extract visual features while preserving the data that each feature represents. Since the convolution neural network doesn't require additional parameters to learn the features of image filters, it saves a significant amount of time and trial-and-error effort. Neurons make up a convolution neural network, which is based on recent advances in neuroscience. When an image is presented to a convolution neural network for image processing, it typically takes the pixel values as input and selects specific visual properties. It sends the edges of the image for corner detection and colorizes portions of the image. The defined image is then given to the following layer for pooling, and so on until the image is anticipated.

Maximum pooling happens as an image's finite definition rises. Only the most important aspects for the soft maximum layer's activation function are

revealed for the dense procedure.

Convolutional Neural Networks Scan Images

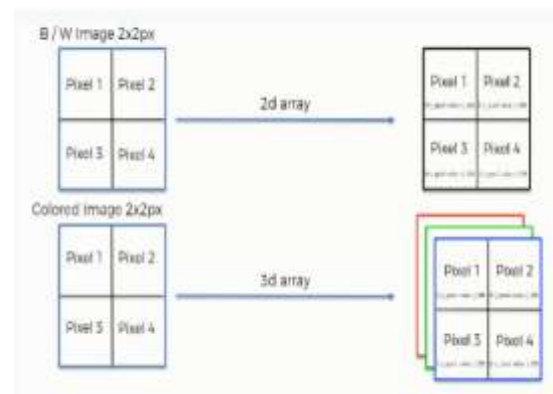


Fig 1.2 Convolution Neural Network Scan Image

1. Pooling Layer

This section will focus on pooling's actual operation while obscuring other aspects of it. In any case, we'll be using a rare form of pooling called most extreme pooling. In any case, It go through an assortment of ways, including mean (or aggregate) pooling. This episode will come to a close with a feature that appears to be straightforward but actually explains the entire topic.

Its purpose is to scale back the network's representation of spatial size as well as the number of parameters and computations.

2. Flattening

This will be a brief explanation of the evenness approach and how to manage deep learning techniques by switching from pooling to levelled layers.

3. PROBLEM DEFINITION

Using Deep Learning models to extract the information limits from lung ultrasonography images would improve the ability of coronavirus ultrasonography to predict whether an infection will be positive or negative.

4. OBJECTIVES

- 1) It is quicker and more reliable to build a Deep Learning network utilising LUS pictures to detect COVID-19.
- 2) It performs model execution after pre-processing the data to separate components from the lung ultrasonography data. Compare the effects of the convolution neural network model and test the model using information for approval of the highlights.

II. LITERATURE SURVEY

2.1 Survey on localization of Covid-19

R. Niehus, P. M. D. Salazar, A. Taylor, and M. Lipsitch., (2020) investigates the risk of coronavirus infections in Wuhan has been calculated based on imported case numbers of overseas visitors, frequently assuming that all instances in passengers have been confirmed. According to recent research, countries' detection capabilities for the outbreaks varies. Singapore has a long history of robust epidemiological monitoring and contact-tracing capabilities, as evidenced by the high dedication of case discovery during the COVID-19 pandemic. Assuming as if all countries had similar discovery abilities as Singapore, an expected 2.8 (95% HPDI 1.5–4.4) times the current number of imported cases might have been found. Utilizing the second part of the Global Health Security file to categorize likely case-identification abilities, are found that the ready to recognize imported cases comparative with Singapore is 40% (95% HPDI 22% – 67%), 37% (95% HPDI 18% – 68%), and 11% (95% HPDI 0% – 42%) among factor in ensuring locations, intermediate surveillance locations, and low surveillance locations.[1]

S. Wang et al (2020) studies Methods and Results are gathered 1,065 CT images of COVID-19 patients with pathogen confirmation (325 images), as well as those who had earlier been classified with normal bacterial pneumonia (740 images). To establish the method, updated the Inception 'mutually' trailed by inside and remotely approval. The internal approval was 89.5 including on overall, with a specificity of 0.88 and a sensitivity of 0.87. The outside testing dataset revealed a total accuracy of 79.3%, with a specificity of 0.83 and a sensitivity of 0.67. Furthermore, the system correctly predicted 46 COVID-19 positive images out of 54 while the first two nucleic acid test results were negative, with an accuracy of 85.2 percent. These findings show that employing artificial intelligence to retrieve radiological characteristics for rapid and accurate COVID-19 diagnosis is possible. Summary of the Author Screening large numbers of suspected patients for proper quarantine and treatment procedures is a priority for controlling the spread of COVID-19. Pathogenic lab tests the gold standard, but it takes time and has a high rate of false negatives. Furthermore, the system accurately predicted 46 COVID-19 positive images out of 54 while the first two nucleic testing results were negative, with an accuracy of 85.2 percent. Ours is the first study to use AI technology to effectively screen for COVID-19 using CT scans.[2]

S.Liu et al.,(2019)In clinical practice, ultrasound (US) has become one of the most regularly used imaging modalities. It is a very fast-growing technology that offers some benefits while also posing distinct obstacles, such as low image quality and considerable unpredictability. In terms of image analysis, sophisticated automatic US image analysis methods must be developed to aid in US diagnosis and/or to make such assessments more objective and accurate. Deep learning has lately risen to prominence as the most powerful machine learning technology in a variety of domains, including general image analysis and machine vision. Deep learning also has a lot of promise for different autonomous image analysis applications in the United States. This paper begins with a quick introduction to many deep learning architectures, followed by a summary and in-depth discussion of their applicability in several particular tasks in US image analysis, including as categorization, identification, and segments. The open difficulties and probable tendencies of future deep learning research in medicine image analysis in the United States are examined.[3]

J. Chen et al(2020), researches about the optimal imaging approach for identifying 2019 new corona virus (COVID19) pneumonia is computed tomography (CT). Our research aimed to develop a deep learning-based method for identifying COVID-19 pneumonia on high-resolution CT, ease radiologists' workload, and assist to the epidemic's containment. Methods In Hospital of Wuhan University (Wuhan, Hubei province, China), 46,096 anonymized images from 106 admitted patients, which include 51 patients with laboratory confirmed COVID-19 pneumonia and 55 monitoring patients with other diseases, were retroactively gathered and analyzed for model development and validation. Twenty-seven patients getting CT scans in Hospital of Wuhan University on February 5, 2020 were prospectively gathered to analyze and assess radiologists' effectiveness against 2019-CoV pneumonia with that of the modeling. The model's performance was equivalent to that of an expert radiologist for 27 prospective patients. The approach helped radiologists cut their reading time in half, saving them 65 percent of their time.[4]

L.-C. Chen, Y. Zhu, G. Papandreou, F. Schroff, and H. Adam.,(2018) studied about the Image Segmentation separable convolution for semantic images. In Deep neural networks employ a spatial pyramid pooling module or an encode-decoder structure for semantic segmentation. By probing the inbound features with filters or

allocating activities at numerous rates and multiple effective fields-of-view, the veteran networks can encode multi-scale contextual information, whereas the latter connections can capture sharper object boundaries by gradually recouping the location data. The propose combining the benefits of both strategies. DeepLabv3+, our suggested model, expands DeepLabv3 by adding a basic but effective decoder module to refine segmentation results, particularly at object borders and extend the Xception model by using depthwise separable convolution to both the Atrous Spatial Pyramid Pooling and decoder modules, yielding a faster and more robust encoder-decoder network. On the PASCAL VOC 2012 and Cityscapes datasets, shows that the suggested model is effective, with testing data performance of 89 percent and 82.1 percent, respectively, without any post-processing. A freely available reference are used to implementation of the suggested models.[5]

X.Xuet al.,(2020) In these paper studied about coronavirus disease in the early stages of determining coronavirus(designated by the World Health Organization), discovered that the real-time reverse transcription-polymerase chain reaction (RT-PCR) found of viral RNA from sputum or nasopharyngeal swab has a relatively low positive rate. Corona virus's computed tomography (CT) imaging presentations have distinct features that distinguished them from other kinds of the viral pneumonia, such as Influenza-A viral pneumonia. As a result, clinical specialists are urging the development of new early diagnostic criteria for this new kind of pneumonia as quickly as feasible.[6]

F. Mojoli, B. Bouhemad, S. Mongodi, and D. Lichtenstein.,(2019)In these studied about the use of point-of-care ultrasonography at the bedside to combine the physical examination of the critically sick is becoming more common; in particular, lung ultrasound has advanced significantly in the previous decade. Basic lung ultrasonography signals are described in this study, with an emphasis on their applicability in critical care. Artifacts (generated from the air/tissue interface) and genuine images(i.e. effusions and consolidations) make up lung semiotics, and both provide valuable information for identifying the most common respiratory infections illnesses. Lung ultrasonography indicators are useful in the diagnosis of patients with acute pulmonary disease, circulatory shock, or cardiac arrest, either alone or in combination with other point-of-care ultrasound procedures. Furthermore, in mechanically ventilated patients, a semi quantification of lung aeration may be conducted at the bedside and

utilized to guide positive end-expiratory pressure setting, assess therapy efficacy, track the progression of the respiratory of illness, and aid the weaning process

2.2 SUMMARY OF LITERATURE

- 1) In The risk of corona virus infections at Wuhan has calculated based on foreign case numbers of overseas visitors. Singapore has a long history of robust epidemiological monitoring and contact-tracing capabilities. If all nations had the same detection capabilities as Singapore, an estimated 2.8 time the present number of foreign cases may has been discovered.
- 2) Using artificial intelligence to retrieve radiological characteristics for rapid and accurate diagnosis is possible. System correctly predicted 46 COVID-19 positive images out of 54 while the first two nucleic acid test results were negative, with an accuracy of 85.2 percent. Ours study is the first one to use AI technology to effectively screen for the COID-19 using CT scans.
- 3) Ultrasound (US) is one of the most commonly used imaging modalities in medicine. Deep learning has a lot of promise for different autonomous image analysis applications in the US. The difficulties and tendencies of future deep learning research in medicine image analysis in the United States are examined.
- 4) DeepLabv3+ extends the Xception model by using depthwise separable convolution to both the Atrous Spatial Pyramid Pooling and decoder modules. Suggested model is effective with testing data performance of 89 percent and 82.1 percent, respectively, without any post-processing.
- 5) A new kind of pneumonia called COVID-19 has distinct features that distinguished them from different sorts of viral pneumonia, like Influenza-A viral pneumonia. Clinical specialists are urging the development of new early diagnostic criteria for this new type of pneumonia as quickly as possible.
- 6) Artifacts (produced from the air/tissue interaction) and true images make up lung semiotics (i.e., effusions and consolidations). Lung ultrasonography can be used to diagnose and treat respiratory issues such as the pneumothorax, ventilator-associated pneumonia, atelectasis, and pleural effusions in patients who are on mechanical ventilation.
- 7) There is minimal evidence on the effect and cost-benefit of the ultrasonography in the resource-constrained environments. Ultrasound

was used in about 70% of the research, with the majority of them coming from South-east Asia and the Sub-Saharan Africa. The paucity

of the higher-quality data and the small number of randomized clinical studies provided restrict the findings.

ANALYSIS ON PLANT DISEASE DETECTION

S.NO	TITLE	DESCRIPTION	KEYFINDINGS
1	A deep learning algorithm using ct images to screen for corona virus disease (covid-19)	The research show that artificial intelligence may be used to extract radiological characteristics in a fast and accurate manner. Diagnosis of COVID-19	While actually esteem the significance of nucleic corrosive identification in the analysis of SARS-COV-2 contamination, it should be noticed that the large number of bogus negatives because of a few factors, for example, strategic disservices, illness stages, and techniques for example assortment may postpone finding and infectious prevention.
2	Deep learning in medical ultrasound analysis	many common deep learning architectures before summarising and discussing their applicability in a variety of particular problems in US image processing, including as classification, detection, and segmentation are all steps in the classification, detection, and segmentation process.	CNN models based on ResNet inception and Xception architecture are not included for accurate results. Automated disease recognition for plants with the incorporation of its soil data and weather is not yet developed
3	Detection of Plant Leaf Diseases Using Image Segmentation and Soft Computing Techniques	Deep Learning techniques segmentation is used to detect the disease of plant leaf by disease classification system. Plant disease can be identified at an early stage	Hybrid algorithms are not included for classification of the image given. Limited plant diseases are identified as the related plant diseases are applied in system.
4	Automatic recognition of plant leaf diseases based on serial combination of two SVM classifiers	SVM classifier is used for recognizing the plant diseases and names through texture qualities. KNN, NB and SVM are tested for better classifier of plant detection.	Neural-Networks or discriminate analysis are not included in plant classification. It is a system based on parallel combinations which is a time taking process for training and testing on large datasets.
5	SVM classifier based grape leaf	It has K-means clustering segmentation	Combination of algorithms such as fusion classification

	diseasedetection	entation foris notincluded. victimizingthopatholo gicalimagedataset and feature extraction.SVMclassifi cationisusedforleafde tecton.	Embeddedsystem for spraying offungicide mixture is notdesigned. It takes more amount oftraining time for a largedataset.
6.	Areviewofneuralnetw orksinplantdisease detectionusinghyper- spectraldata	ConvolutionNeuralNe twork(CNN)procedur eisavailablein the marketplace toapproach hyper- spectral data,with aget unique pressure onsicknessfindings.At first,weprovide an assessment oftheCNN mechanism,types and classifiers thatrequires useabsolutely one-of- a-kindalgorithms to method hyper- spectralstatistics.	CNNhasnotbeenevaluatedfor SDI whereas there may be aprospecttoexemptafewinstr uctions to getconceivableimprovement. Theseprocedures are shippedvictimization earlier than the occasionofcompanion SDI. It requires high cost toperformCNNNonSDIasit requires high initialinvestmentforirrigation system compared to thatofother.

Table-1 ANALYSIS ON PLANT DISEASE DETECTION

III. PROPOSED SYSTEM

3.1. PROPOSED SYSTEM

A new deep network developed from Spatially Transformer Systems forecasts the illness affects linked with an input frame while also providing location of pathological artefacts. Trails on the provided dataset show good outcomes for all the tasks studied, clearing the path for further research

on Deep learning for coronavirus aided diagnosis using LUS data.

filter to an input data for the map activation results leads to featuremap.

Convolutional is first layer extract features from the dataset. It preserves the relationship between the pixel data through learning image features.

3.2 ARCHITECTURE

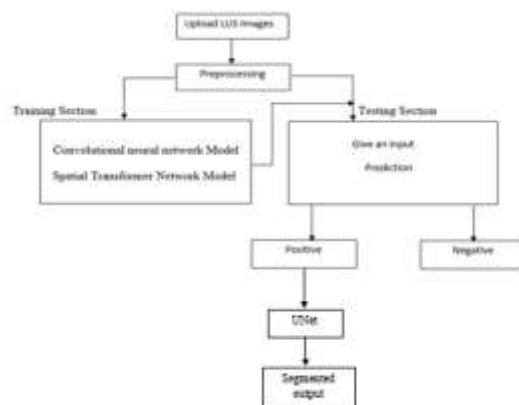


Fig 3. 1 Proposed System Architecture

The procedure to develop our system is clearly described in this section. After the splitting

of the data will pre-process the splitted dataset and then the CNN based transfer learning model is used

for the training purpose with the help of TensorFlow and Keras. After the training, the model is saved for the testing. Where creating a website where a user can upload the input image for which the classification is to be performed. The classification is performed on the uploaded image and the result will be generated as either the uploaded image is covid positive or negative. If the output is classified as the positive then the U-Net is applied on the classified image. From the U-Net applied results on the segmented part of the disease from the given input image along with that a voice will be generated for predicted results.

3.3. ALGORITHMS

U-Net

When U-Net is a biomedical image classification convolution layer developed at the University of Freiburg's Science Department. The topology of the network was upgraded and enlarged so that it could function with less training photographs and provide more precise segmentations. A 512 512 image may be segmented in as little as a second on a modern GPU.

The "completely convolutional network" was given by Long, Shelhamer, and Darrell as the foundation for the U-Net concept. The basic idea is to add layers to a standard contractual network, with up-sampling operators substituting pooling activities. As a result, these layers aid in achieving the intended outcome. Furthermore, a subsequent fully connected layer may learn to build a correct output based on this information.

U-Net's sample selection segment has a large number of feature channels, allowing the device to pass frame of reference information to appropriate complex function. This is a necessary modification.

ALGORITHM

1. For each training images of RGB format, rescale the RGB images to its largest dimension and to its HSI (Hyper Spectral Images). Normalize the original image between 0 and 1.
2. Apply histogram equalization to HSI images and train the RGB images from 2D white pad to flipped/rotated versions.
3. Train U-Net from selected training images with its iterations. Initialize the Threshold.
4. For each training model predict its skin lesion area, apply the binary fill holes, and use threshold values for optimization.
5. Use optimized threshold values to get skin lesion area for each test image and apply binary fill holes.

IV. CONCLUSION

A promising method for coronavirus screening and detection uses deep learning for computer-assisted interpretation of lung ultrasound images. An updated collection of LUS photos with labels that indicate the degree of disease at the frame and pixel levels was collected from hospitals and other online sources. Convolutional Neural Networks (CNN), which are built from Spatial Transformer Networks (STN), which forecast the affected part from the given image, are used in this sophisticated organisation. U-Net will apply for the division of the predicted disease section from the provided image after the affected area has been identified. By suggesting deep learning methods for evaluating Lung Ultrasonography (LUS) images, it is possible to detect COVID infected victims with greater accuracy and with less error.

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