

Lung Cancer Detection Using Machine Learning and Deep Learning Techniques

¹Nazia Fatima, ²Ayonija Pathre, ³Mukesh Kumar

¹Student, Department of Computer Science, Rabindranath Tagore University, Bhopal, MP, India

^{2,3}Assistant Professor, Department of Computer Science, Rabindranath Tagore University, Bhopal, MP, India

Abstract - Early recognition of Cell disintegration in the lungs cells can help in a sharp lessening in the lung cancer death rate subsequently it is a forceful infection which conveying a bleak forecast with a 5-year endurance rate at 18%. Several PC supported determination substructures have been created to help diminish Cell disintegration in the lungs death rates. Thus, structural co-occurrence matrix - based methodology is utilized to separate the component & to characterize tumors into dangerous or considerate tumors & furthermore into their threat level. The computed tomography examines from the lung picture details base consortium & picture data set asset activity datasets give details regarding clot locations & their harm levels is been conveyed here as a model. Support vector machine & CNN is being utilized as a classifier which seems to be 1) to categories the clot pictures into threatening or benevolent clots & 2) to arrange the lung clots into harm levels. These test results uncover that the SCM effectively removed highlights of the clots from the pictures &, in this manner might be regarded as a propitious device to help clinical expert to make a more exact discover regarding the danger of lung clots.

Keywords: Cell Disintegration, Computed Tomography, CNN, Support Vector Machine, Tumor.

I. INTRODUCTION

Cell disintegration in the lungs is one of the primary driver of the demise & medical problem in many nations with a 5-year endurance pace of just 10–16%. To help radiologists in the detection of beginning phase obsessive articles, around multi decade prior, specialists began the advancement of CAD techniques to be applied to CT assessments. Albeit most tests & methods should be possible inside the space of minutes or hours, unavoidable however significant holding up occasions can happen between the tests & strategies. Indeed, days or even a very long time of deferrals are normal. Such deferrals for a possibly perilous disease not just lead to terrible experience to both the sufferers & care suppliers, yet additionally might be connected to unfriendly endurance rate. In this manner, without giving up consideration quality,

accelerating the conclusion to-treatment measure is basically imperative to improve persistent result.

Computed Tomography has been appeared as the touchiest imaging methodology for the location of little pneumonic knobs, especially since the presentation of the multi detector-column & helical CT innovations. It encourages radiologists to survey early hazard variables of malignant growth which is fundamental in Cell disintegration in lung research. The rate of passing caused because of Cell disintegration in the lungs is more than prostate, colon, & breast cancers consolidated. Likewise, most sufferers identified with Cell disintegration in the lungs today are as of now at a high-level stage as a Cell disintegration in the lungs is difficult to distinguish in the beginning phases. In these highlights, dimensions, appearances, & volumetric developing esteem have been taken as the most dependable highlights for a dangerous finding of knobs. Be that as it may, the greater part of the above highlights needs a more precise division of knobs aside from the surface. Hence, we take more thoughtfulness regarding the surface highlights for helping the threat analysis of knobs in this project.

Computed Tomography was created by the designer Godfrey N. Hounsfield & the physicist Allan M. Comark in 1972. The two researchers were granted the Nobel Prize, to make such CAD substructures, there is consistently a requirement for a reference-quality details set that can be utilized to get ground facts & can likewise go about as a reason for correlation of various CAD calculations. LIDC, Image-based procedures for breaking down injuries are typically performed with location, division hand-made component designing & classification marking Zinoviev et al. received a conviction choice tree way to deal with anticipate knob semantic ascribes. Related examinations on the Lung Image Details-base Fraternity & Image Database Resource Initiative details set & at concurrent the CADx substructure gives a subsequent assessment to help in dynamic.

Nonetheless, every one of these strategies depend on knob division as an essential. Quiet, programmed knob division may impact arrangement results since strategies, for example, area groining & level set normally rely upon introduction.

Chipping away at these fragmented areas may yield wrong highlights that lead to incorrect yields. The utilizations of programmed lump division are expansive, including estimating therapy reaction, arranging of radiation therapy, & to encourage extraction of strong highlights for high-throughput radiomics by the MRRN. Henceforth, great outcomes got with the proposed strategy would be a possibility to improve clinical findings & help with settling on more exact & effective choices for these two significant medical issues.

The rest of this paper continues as follows. Segment 2 dissects the literature survey. In segment 3 the proposed framework is presented. Area 4 is dedicated to the proposed shrewd storing technique for reserving choice. The trials and results are introduced in Section 5. At last, this examination is closed in Section 6.

II. RELATED WORK

There are many research works have been done on Lung Cancer Detection using machine learning.

In [1], Timor Kadir et al. have given a diagram of the primary methodologies utilized for knob grouping & Cell disintegration in the lung's expectation from CT imaging details. Here, adequate preparing details has been giving, the present status of-the-craftsmanship is accomplished utilizing CNN's prepared with Deep Learning accomplishing a characterization execution in the area of low 90s AUC focuses. While assessing substructure execution, it is imperative to know about the constraints or in any case of the preparation & approval details indexes utilized, i.e., were the smokers or non-smokers, or sufferers with a current or earlier history of harm included. The 15 highlights were chosen from a palette of more than 1,300 old-style surface highlights including Haralick, Gabor, alongside straightforward estimates, mean, SD, & volume. They used a fully mechanized component determination system that meant to choose a little subset of highlights that upgraded arrangement execution over an in-house preparing detail set. Since it is reckoning infeasible to test all blends of the full palette of highlights, they used a successive "eager" calculation that, beginning with the ideal pair of highlights found by thorough inquiry overall sets of highlights, chosen includes individually to augment the exhibition over the preparation details set at each progression. At last, SVM relapse estimation with a blocky piece was drawn up to utilize the libSVM library. The end product of this continuance is a number somewhere in the orbit of 0 & 1 that reflects the likelihood that a peculiar knob is harmful.

In [3], Goran Jakimovski et al. propose Cell disintegration in the lungs clinical picture classifier that depends on a CDNN. To prepare & test the substructure, they utilized CT

pictures of lungs that were recently arranged by clinical subject matter experts & placed into heaps of yes/no. In this, the creators utilized the K-implies calculation to pre-arrange the pictures into heaps of same cut pictures, where the DNN can zero in on picture order of same cut pictures. The subsequent oddity is the extra convolution layer with edge honing channels, to altogether look for malignancy. At last, the principle oddity is trying our DNN with Cell disintegration in the lung's pictures from Tx stages 2, 3, & 4 & deciding at which Tx stage the two calculations can recognize the chance of malignancy. The outcomes were dissected with clinical staff from the oncology office & were set apart as agreeable to decide disease.

In [5], Joseph A. Cruz et al. think about & survey the presentation of various AI which are applied to disease expectation & forecast. In particular, the creators distinguished various patterns regarding the sorts of AI techniques being utilized, the sorts of preparing details being coordinated, the sorts of endpoint expectations being made, the kinds of malignancies being considered & the general execution of these strategies in anticipating disease helplessness or results. While ANNs actually prevail, it is obvious that a developing assortment of substitute AI methodologies are being utilized & that they are being applied to numerous sorts of malignant growths to foresee at any rate three various types of results. It is additionally evident that AI strategies by & large improve the exhibition or prescient exactness of most forecasts, particularly when contrasted with traditional factual or master-based substructures.

In [7], Dr. S. Senthil et al. coordinated system for anticipating Cell disintegration in the lungs is presented utilizing Aural Network with Particle Swarm Optimization. Aural organizations manage diverse critical thinking strategies in which the neurons are being prepared & tried by given details base. The Cell disintegration in the lung's highlights is removed for foreseeing the disease stage dependent on certain element utilized in the substructure. Highlight determination is utilized to recognize prescient subsets of malignant growth cells inside a detail base & lessen the quantity of disease cells introduced to the calculation technique. Better execution can be accomplished by disposing of certain highlights. Lung illness is the uncontrolled improvement of odd cells that influence one or the two lungs. The fundamental target is to anticipate & early discovery of Cell disintegration in the lungs by utilizing aural organization with ideal highlights. At first the Lung details base are gathered & given as contribution to the substructure. At that point details preprocessing is applied on the detail's pictures, for the upgrading the picture to get the high differentiation pictures. The upgraded pictures are prepared & tried by aural organization contrasted & test preparing details base. PSO is applied to remove the highlights

of the given details pictures & further cycle is continue to identify the Cell disintegration in the lungs. On this premise, aural organization classifier is utilized to examine those details pictures highlights are delegated carcinogenic or non-dangerous.

In [8] Jelo Salomon proposes a strategy for recognizing Cell disintegration in the lungs in a CT filter utilizing a 2D-UNet model on a web application. The creator edited 2D malignancy covers on its reference picture utilizing the focal point of the Cell disintegration in the lungs given in the details-set & prepared a model with various strategies & hyper boundaries. At last the outcome is assessed utilizing a dice coefficient & disarray grid measurement. The creator arrives at 65.7% exactness on the dice coefficient & a normal 0.88% genuine positive rate & 0.71% bogus positive rate on a test set of positive & negative examples. A web application has been created to exhibit a proof of idea. The application requires a clientele to transfer a CT scan. The application at that point measures the document & shows the pictures to the habitu . The clientele at that point picks which examine the person needs to foresee then the application pre-measures the CT scan & gathers the picture to the prescient model. The yield of the model is then shown to the clientele. The clientele has the decision to see the pictures by means of a merry go round or a display mode.

In [10] Animesh Hazra et al. talks about, one of the important & regular bases of disease passing all around the world as far as both example & fleetingness is Cell disintegration in the lungs. The primary purpose for the expanding of passing from it is recognizing the illness of late & deficiencies in viable treatment. Thus, the early location is expected to save lives from this sickness. The survivability pace of Cell disintegration in the lungs can be anticipated with the assistance of present-day AI strategies. Appropriately, it is sharp to decide the endurance prospects among the sufferers. In this investigation details cleaning, include choice, parting & arrangement procedures have been applied for foreseeing survivability of Cell disintegration in the lungs as precisely as could reasonably be expected. The clinical details taken from CGD entryway is utilized for additional preparing to get the ideal result. The proposed system worked here containing of three stages. From the start pre-preparing of details collection is finished. At that point the dataset will be part into two lays, one is utilized for preparing stage & another for testing stage. Next, grouping on the dataset utilizing SVM & LR calculations were performed.

III. PROPOSED SYSTEM

The picture handling procedures are generally utilized for the forecast of a Cell disintegration in the lungs & furthermore

for early identification & therapy to block the Cell disintegration in the lungs. In the proposed substructure, we use Python programming for examination. In picture handling systems, the strategies included are picture pre-preparing, division, & highlight extraction procedures have been talked about in detail. We are arranging to get more precise outcomes by utilizing different upgrade & division strategies. The primary co-event grid capability was an appeal to cull out highlights from pictures of knobs & classify them into dangerous or benevolent knobs & their threat levels. The CT test from the lung picture data set consortium & picture data set asset prompt datasets to give data alluding to knob positions & their threat levels. The grouping stage use the SVM & applied them to two errands:

1. To characterize the knob pictures into threatening or kindhearted knobs &
2. To characterize the lung cells into threat levels.

For picture acknowledgment applications, a few pattern structures of CNNs have been created, which have been effectively applied to convoluted errands of visual symbolism. The principle proposed substructure can upgrade helps in finding of multi CT-check pictures with Cell disintegration in the lungs & determination of Cell disintegration in the lungs utilizing profound learning method. In CNN which includes various layers that are utilized for best expectation with the code & preparing model have accomplished an exactness level of 99%.

Advantages

- Achieved a triumph pace of 99%
- High-level execution
- Low computational force
- The proposed technique performed all around contrasted with all the current strategies
- Execution time is less.

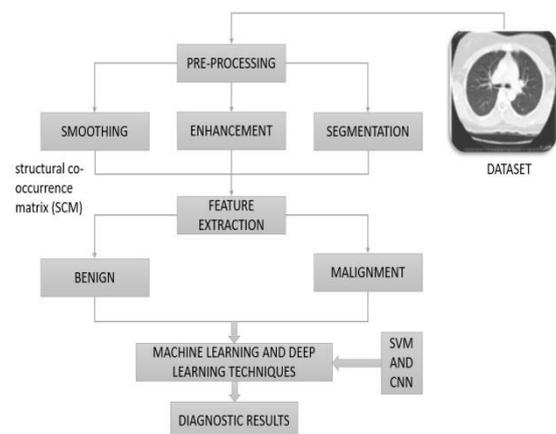


Figure 1: The system architecture substructures of ML & DL technique

IV. IMPLEMENTATION

The proposed substructure comprises of are a few techniques or modules of the accompanying strides to decipher Cell disintegration in the lungs utilizing Machine learning & profound learning strategy. The working technique comprises of the primary stages. These are separately; stacking the details collection, the plan of the convolutional aural organization, the setup of preparing alternatives, preparing of the CNN malignancy identifier with Cell disintegration in the lungs, assessment of prepared locator. These stages & traditional & techniques will be talked about in this segment.

A) Feature Extraction

In this examination, an element vector of dimension 4096 was extricated from every district proposition with Caffe profound learning system. Highlights were determined by sending the normal yield 256, 256 red-green-blue pictures with five convolution layers & two totally associated layers. To ascertain quality in an area proposition, the picture details are first changed over to a structure viable with CNN. At that point, the most straightforward of the potential changes of the irregular molded locales were chosen. Here, all the pixels in a tight jumping box around the competitor territory are settled into the necessary dimension, paying little heed to the dimension or angle proportion. Prior to dissolving, the tight jumping box was extended to give w pixels slanted picture content around the case at the slanted measurement (w = 16 was utilized). Moreover, a straightforward jumping box relapse was utilized to extend the limitation execution inside the application.

B) Training

CNN was prepared on a huge assistant details index utilizing just picture level extra labels. CNN was prepared on details index utilizing just extra labels. This preparation was done utilizing the Caffe Deep Learning system. We utilized some AI characterization calculations. We get a thought from the details perception plots which calculations will be appropriate for the arrangement issue. The ML substructure utilizes the preparation details to prepare models to see examples & utilizations the test details to assess the prescient nature of the prepared model. AI substructure assesses the prescient execution by contrasting expectations on the assessment details index with genuine qualities utilizing an assortment of measurements. Thus, for our proposal, we will assess some unique AI & profound learning calculations, example:

1. Support Vector Machine (SVM)
2. Convolutional Aural Network (CNN)

C) Lung Category Classifiers

Here, paired classifier preparing was utilized to see lung. It is a positive illustration of a picture region in which a lung is firmly encased. Along these lines, a foundation area that isn't keen on lung is a negative model. It is indistinct how a mostly covering area of the lung should be marked. The muddled state is addressed by indicating cover edge esteem. Regions underneath this limit esteem are recognized as negative & those over the edge an incentive as sure. The cover limit "0.3" was picked by leading a matrix search on the confirmation set. When the highlights are eliminated & the preparation labels are applied, CNN is applied ideally to all classes.

D) Result (lung cancer detection)

The proposed malignancy locator with lung has been effectively prepared by utilizing AI (SVM) & CNN profound learning techniques on the example lungs datasets & the Cell disintegration in the lung's discovery measure has been effectively performed by the prepared Cell disintegration in the lung's identifier being tried on the test details index. Various pictures were tried & discovered that the new method of characterization was found to show 97% precision. A few pictures tried with other details base pictures are given in the investigation of the outcomes. In Results, the examination is constant recognition in Cell disintegration in the lungs with capacities like Cell disintegration in the lungs. At the point when they perceive the cycle starts at that point think about the prepared model 'CNN classifier. Model' if the cycle model matches after that Cell disintegration in the lungs discovery measure in the showcase the outcome.

V. ALGORITHM

A) SVM Algorithm Overview

SVM is a managed AI calculation that is doing great in example acknowledgment issues & it is utilized as a preparation calculation for considering grouping & relapse rules from details. SVM is most absolutely utilized when the quantity of highlights & the quantity of occurrences is high. A twofold classifier is worked by the SVM calculation. This paired classifier is developed utilizing a hyper plane where it is a line in more than 3-measurements. The hyper plane accomplishes crafted by isolating the individuals into one of the two classes. The hyper plane of SVM is based on numerical conditions.

The condition of hyper plane is $W \cdot X = 0$ which is like the line condition $y = ax + b$. Here W & X speak to vectors where the vector W is consistently ordinary to the hyper plane. WTX speaks to the dab result of vectors. As SVM manages the dataset when the quantity of highlights is all the more thus, we

need to utilize the condition $WTX=0$ for this situation as opposed to utilizing the line condition $y= ax + b$. On the off chance that a bunch of preparing details is given to the machine, every details thing will be allocated to either clear cut factors; an SVM preparing calculation fabricates a model that plots new details things to either classification. In a SVM model, every details thing is spoken to as focuses in a n-dimensional space where n is the quantity of highlights where each element is spoken to as the estimation of a specific facilitate in the n-dimensional space. Arrangement is done by finding a hyper-plane that isolates the two-classes capably. Afterward, the new details thing is planned into similar space & its classification is anticipated dependent on the side of the hyper-plane they turn up.

B) CNN Algorithm Overview

CNN were utilized to accomplish some advancement results & win notable challenges. The use of convolutional layers comprises in complicating a sign or a picture with pieces to acquire highlight maps. In this way, a unit in an element map is associated with the past layer through the loads of the parts. The loads of the pieces are adjusted during the preparation stage by back propagation, to upgrade certain qualities of the details. Since the pieces are divided between all units of a similar component maps, convolutional layers have less loads to prepare than thick FC layers, creating CNN simpler to prepare & less inclined to over fitting.

Besides, since a similar part is convolved over all the pictures, a similar component is recognized freely of the finding—interpretation invariance. By utilizing portions, data of the area is considered, which a helpful wellspring of setting data is. Typically, a non-direct actuation work is applied to the yield of each aural unit. On the off chance that we stack a few convolutional layers, the separated highlights become more theoretical with the expanding profundity. The principal layers improve highlights, for example, edges, which are totaled in the accompanying layers as themes, parts, or items.

The accompanying ideas are significant with regards to CNN:

1. Initialization

It is imperative to accomplish assembly. We utilize the Xavier instatement. With this, the enactments & the inclinations are kept up in controlled levels; in any case back-spread slopes could disappear or detonate.

2. Activation Function

It is liable for non-directly changing the details. Rectifier straight units (ReLU), characterized as

$$f(x)= \max(0, x),$$

Were found to accomplish preferable outcomes over the older style sigmoid or exaggerated digression capacities, & accelerate preparing. Be that as it may, forcing a steady 0 can impede the angle streaming & resulting change of the loads. We adapt to these impediments utilizing a variation called flawed redressed straight unit (LReLU) that presents a little slant on the negative piece of the capacity. This capacity is characterized as:

$$f(x)= \max(0, x)+\alpha\min(0,x)$$

Where is the defectiveness boundary? In the last FC layer, we use softmax.

3. Pooling

It joins spatially close by highlights in the element maps. This blend of potentially excess highlights creates the portrayal more reduced & stable to miniature picture changes, example, unimportant subtleties; it additionally lessens the ciphering heap of the following stages. To join apex, it is more normal to utilize max-pooling or normal pooling.

4. Regularization

It is utilized to lessen over fitting. We use Dropout in the FC layers. In each preparation step, it eliminates hubs from the organization with likelihood. Along these lines, it powers all hubs of the FC layers to learn better portrayals of the details, keeping hubs from co-adjusting to one another. At test time, all hubs are utilized.

5. Data Augmentation

It very well may be utilized to expand the dimension of preparing sets & lessen over fitting. Since the class of the fix is gotten by the focal voxel, we confined the details expansion to turning activities.

VI. NETWORK ARCHITECTURE

A) Image Input Layer

A picture Input Layer is a spot you introduce the dimension of the info picture, here, 128-by- 128-by-1 is utilized. These numbers speak to the tallness, width, & number of channels. For this situation, input details is a grayscale picture, consequently the quantity of channels is 1.

B) Convolutional Layer

Details contentions for this layer are separating dimension, the quantity of channels, & cushioning. Here, the

channel of dimension 10 is utilized, which decides the 10 x 10 channel. The quantity of channels utilized is 10, which implies 10 neurons are associated. The cushioning of 1 indicates that the dimension of the yield picture is equivalent to that of an details picture.

C) ReLU Layer

ReLU (redressed direct unit) layer is a clump standardization layer, which is put subsequent to introducing a nonlinear actuation work. The significance of this layer is to diminish the affectability & increment the speed of the preparation.

D) Max Pooling Layer

It is one of the down-sampling methods which is utilized for convolutional layers. In this design, the pool dimension is set to 3 & the preparation capacity's progression dimension is 3.

E) Fully Connected Layer

In this layer, all the neurons of all layers are interconnected to the past layer. The given details contention for this layer is 10, which demonstrate 10 classes.

F) SoftMax Layer

Completely associated layers are trailed by the SoftMax layer, which is a standardization strategy. This layer produces positive numbers as the yield with the end goal that the amount of numbers is one. The grouping layer utilizes these numbers for arrangement.

G) Classification Layer

The arrangement layer is the last layer of the design. This layer characterizes the classes dependent on probabilities got from the SoftMax layer & furthermore ascertains the cost work.

VII. ARCHITECTURE (ALEXNET)

This design was one of the principal profound organizations to push Image Net Classification precision by a critical step in contrast with customary strategies. It is made out of 5 convolutional layers followed by 3 completely associated layers, as portrayed in Figure.

AlexNet, proposed by Alex Krizhevsky, utilizes ReLu for the non-straight part, rather than a Tanh or Sigmoid capacity which was the previous norm for conventional aural organizations. ReLu is given by:

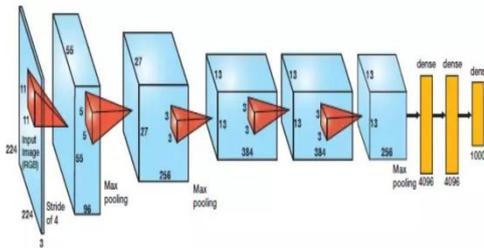
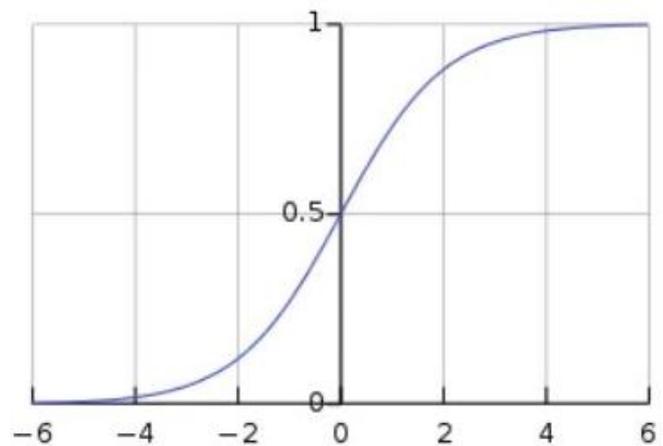


Figure 2: Architecture of Aural network (AlexNet)

$$f(x) = \max(0, x)$$

The upside of the ReLu over sigmoid is that it prepares a lot quicker than the last on the grounds that the subordinate of sigmoid turns out to be little in the immersing area & along these lines the updates to the loads nearly evaporate. This is called evaporating angle issue.

In the organization, ReLu layer is put after every single convolutional & completely associated layer (FC).



Another issue that this design settled was lessening the over-fitting by utilizing a Dropout layer after each FC layer. Dropout layer has a probability, (p), related with it & is applied at each neuron of the reaction map independently. It arbitrarily turns off the actuation with the likelihood p, as can be found in figure below.

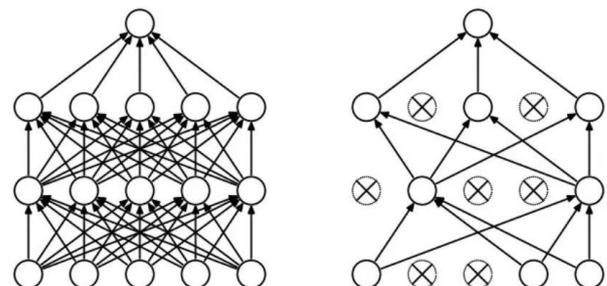


Figure 2: Arbitrarily turns off the actuation with the likelihood p

VIII. RESULTS

```

Python 3.7.4 Shell
File Edit Shell Debug Options Window Help
C:\Users\91988\AppData\Local\Programs\Python\Python37\lib\site-packages\sklearn\externals\joblib\_init_.py:15: DeprecationWarning:
sklearn.externals.joblib is deprecated in 0.21 and will be removed in 0.23. Please import this functionality directly from joblib, which
can be installed with pip install joblib. If this warning is raised when loading pickled models, you may need to re-serialize those
models with scikit-learn 0.21+.

Using TensorFlow backend.
* Serving Flask app "modules" (lazy loading)
* Environment: production
WARNING: This is a development server. Do not use it in a production deployment.
Use a production WSGI server instead.
* Debug mode: off
* Running on http://127.0.0.1:8050/ (Press CTRL+C to quit)

svm started
accuracy score for SVM 0.8840579710144928
precision score for SVM 0.8840579710144928
f1 score for SVM 0.8840579710144928
recall score for SVM 0.8840579710144928
(0.8840579710144928, 0.8840579710144928, 0.8840579710144928, 0.8840579710144928)
>>>
  
```

```

Python 3.7.4 Shell
File Edit Shell Debug Options Window Help
Python 3.7.4 (tags/v3.7.4:e09359112e, Jul 8 2019, 20:34:20) [MSC v.1916 64 bit
(AMD64)] on win32
Type "help", "copyright", "credits" or "license()" for more information.
>>>
RESTART: E:\2020_21_ML_DL_IP\VTech\Lung-Cancer\Final-Project (LungCancer)\Mach
e Learning\segment.py
Segmentation:
Area: 132.5
parameter: 74.5269113779068
>>>
  
```

```

Python 3.7.4 Shell
File Edit Shell Debug Options Window Help
Python 3.7.4 (tags/v3.7.4:e09359112e, Jul 8 2019, 20:34:20) [MSC v.1916 64 bit
(AMD64)] on win32
Type "help", "copyright", "credits" or "license()" for more information.
>>>
RESTART: E:\2020_21_ML_DL_IP\VTech\Lung-Cancer\Final-Project (LungCancer)\Mach
e Learning\predict.py
neg.png
class: [0]
Diagnosis Result is: Benign
>>>
  
```

```

Python 3.7.4 Shell
File Edit Shell Debug Options Window Help
Python 3.7.4 (tags/v3.7.4:e09359112e, Jul 8 2019, 20:34:20) [MSC v.1916 64 bit
(AMD64)] on win32
Type "help", "copyright", "credits" or "license()" for more information.
>>>
RESTART: E:\2020_21_ML_DL_IP\VTech\Lung-Cancer\Final-Project (LungCancer)\Mach
e Learning\predict.py
pos.png
class: [1]
Diagnosis Result is: Malignant
Segmentation:
Area: 132.5
parameter: 74.5269113779068
>>>
  
```

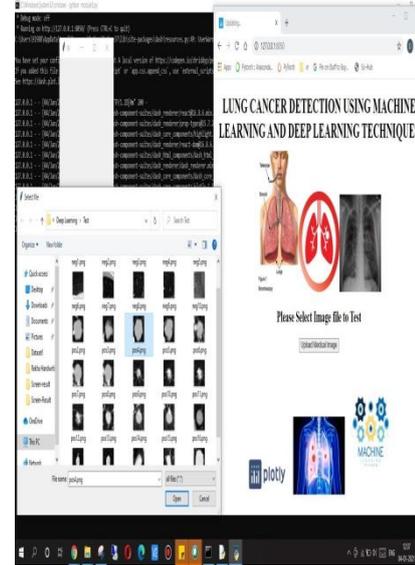
```

Python 3.7.4 Shell
File Edit Shell Debug Options Window Help
Python 3.7.4 (tags/v3.7.4:e09359112e, Jul 8 2019, 20:34:20) [MSC v.1916 64 bit
(AMD64)] on win32
Type "help", "copyright", "credits" or "license()" for more information.
>>>
threshold = 0.5
img = cv2.imread('E:\2020_21_ML_DL_IP\VTech\Lung-Cancer\Final-Project (LungCancer)\Mach
e Learning\preprocessing.py
cv2.imshow('CT-Scan Image', img)
cv2.waitKey()
print('Grey Image')
img = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
kernel = cv2.getStructuringElement(cv2.MORPH_RECT, (5, 5))
img = cv2.morphologyEx(img, cv2.MORPH_OPEN, kernel)
img = cv2.morphologyEx(img, cv2.MORPH_CLOSE, kernel)
img = cv2.morphologyEx(img, cv2.MORPH_ERODE, kernel)
img = cv2.morphologyEx(img, cv2.MORPH_DILATE, kernel)
cv2.imshow('Edge Image', img)
cv2.waitKey()
>>>
  
```

```

Select C:\Windows\System32\cmd.exe - python module.py
E:\2020_21_ML_DL_IP\VTech\Lung-Cancer\Final-Project (LungCancer)\Deep Learning\python module.py
C:\Users\91988\AppData\Local\Programs\Python\Python37\lib\site-packages\sklearn\externals\joblib\_init_.py:15: DeprecationWarning:
sklearn.externals.joblib is deprecated in 0.21 and will be removed in 0.23. Please import this functionality directly from joblib, which
can be installed with pip install joblib. If this warning is raised when loading pickled models, you may need to re-serialize those
models with scikit-learn 0.21+.

Using TensorFlow backend.
* Serving Flask app "modules" (lazy loading)
* Environment: production
WARNING: This is a development server. Do not use it in a production deployment.
Use a production WSGI server instead.
* Debug mode: off
* Running on http://127.0.0.1:8050/ (Press CTRL+C to quit)
  
```



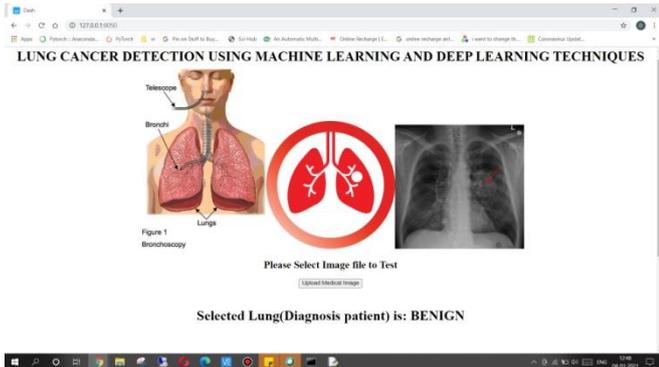
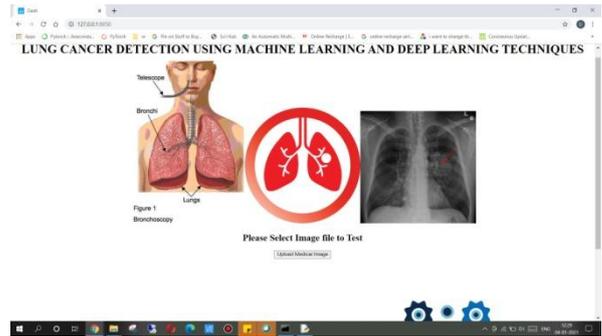
```

Select C:\Windows\System32\cmd.exe
Using TensorFlow backend.
2021-01-04 11:48:17.361530: I tensorflow/core/platform/cpu_feature_guard.cc:142] Your CPU supports instructions that this TensorFlow binary
was not compiled to use: AVX2
Found 288 images belonging to 2 classes.
Found 24 images belonging to 2 classes.
Epoch 1/10
25/25 [=====] - 4s 164ms/step - loss: 0.3111 - accuracy: 0.8684 - val_loss: 0.6974 - val_accuracy: 1.0000
Epoch 2/10
25/25 [=====] - 2s 68ms/step - loss: 0.1491 - accuracy: 0.9487 - val_loss: 0.8033 - val_accuracy: 1.0000
Epoch 3/10
25/25 [=====] - 2s 73ms/step - loss: 0.1737 - accuracy: 0.9334 - val_loss: 0.2730 - val_accuracy: 0.9706
Epoch 4/10
25/25 [=====] - 2s 77ms/step - loss: 0.0751 - accuracy: 0.9742 - val_loss: 5.8474e-04 - val_accuracy: 1.0000
Epoch 5/10
25/25 [=====] - 2s 67ms/step - loss: 0.0635 - accuracy: 0.9885 - val_loss: 2.2655e-04 - val_accuracy: 1.0000
Epoch 6/10
25/25 [=====] - 2s 80ms/step - loss: 0.0338 - accuracy: 0.9861 - val_loss: 9.0127e-04 - val_accuracy: 1.0000
Epoch 7/10
25/25 [=====] - 2s 66ms/step - loss: 0.0366 - accuracy: 0.9894 - val_loss: 7.3867e-04 - val_accuracy: 1.0000
Epoch 8/10
25/25 [=====] - 2s 70ms/step - loss: 0.0282 - accuracy: 0.9934 - val_loss: 5.8807e-04 - val_accuracy: 1.0000
Epoch 9/10
25/25 [=====] - 2s 72ms/step - loss: 0.0081 - accuracy: 1.0000 - val_loss: 3.6340e-04 - val_accuracy: 1.0000
Epoch 10/10
25/25 [=====] - 2s 62ms/step - loss: 0.0068 - accuracy: 0.9973 - val_loss: 0.0011 - val_accuracy: 1.0000
  
```

```

C:\Windows\System32\cmd.exe

E:\2020_21_III_IP\Tech(Lung-Cancer)\Final-Project(Lung-Cancer)\Deep Learning\python module2.py
3
Segmentation:
Area: 24.0
Parameter: 20.485281229019165
    
```



```

C:\Windows\System32\cmd.exe

0.0
Detected: BENIGN
Test\neg7.png

0.0
Detected: BENIGN
Test\neg8.png

0.0
Detected: BENIGN
Test\neg9.png

1.0
Detected: MALIGNANT
Test\pos10.png

1.0
Detected: MALIGNANT
Test\pos11.png

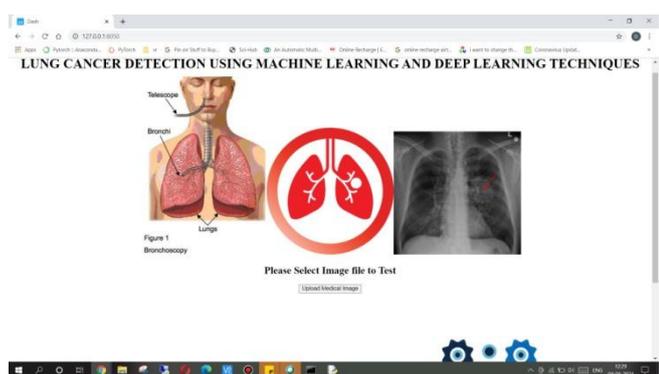
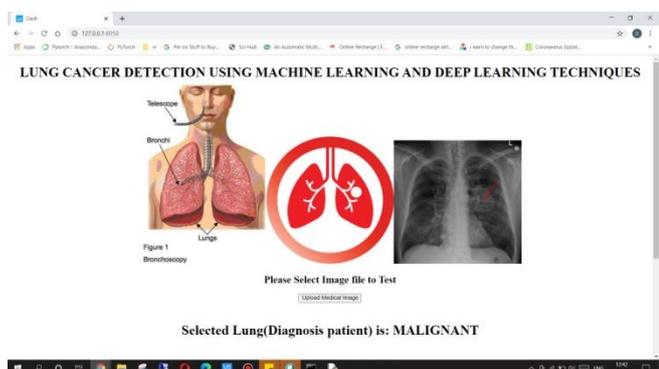
1.0
Detected: MALIGNANT
Test\pos12.png

1.0
Detected: MALIGNANT
Test\pos13.png
    
```

```

Select C:\Windows\System32\cmd.exe - python module4.py

/..11[0m] 200
127.0.0.1 - [04/Jan/2021 12:40:36] "0[37mGET /_dash-component-suites/dash_core_components/dash_core_components.min.js?v=1.3.1&m=157192980 HTTP/1.1] 200
127.0.0.1 - [04/Jan/2021 12:40:36] "0[37mGET /_dash-component-suites/dash_renderer/dash_renderer.min.js?v=1.1.2&m=1571929839 HTTP/1.1] 200
127.0.0.1 - [04/Jan/2021 12:40:36] "0[37mGET /_dash-component-suites/dash_core_components/plotly-1.50.1.min.js?v=1.3.1&m=1571929840 HTTP/1.1] 200
127.0.0.1 - [04/Jan/2021 12:40:36] "0[37mGET /_dash-layout HTTP/1.1] 200
127.0.0.1 - [04/Jan/2021 12:40:36] "0[37mGET /_dash-dependencies HTTP/1.1] 200
127.0.0.1 - [04/Jan/2021 12:40:36] "0[37mGET /_favicon.ico?v=1.4.1 HTTP/1.1] 200
127.0.0.1 - [04/Jan/2021 12:40:36] "0[37mPOST /_dash-update-component HTTP/1.1] 200
5/2809_21_III_IP\Tech(Lung-Cancer)\Final-Project(Lung-Cancer)\Deep Learning\Test4\pos2.png
2021-01-04 12:41:33.621154: I tensorflow/core/platform/cpu_feature_guard.cc:142] Your CPU supports instructions that this TensorFlow binary was not compiled to use: AVX2
1.0
Segmentation:
Area: 120.0
Parameter: 64.28427052497864
127.0.0.1 - [04/Jan/2021 12:41:35] "0[37mPOST /_dash-update-component HTTP/1.1] 200
    
```



IX. CONCLUSION

Here we handled the CT scan pictures to separate the favorable & the harmful clot & its degree of the development of the malignancy cells by the AI, profound learning substructure & distinguishing the development of the destructive cell in the underlying stage can be made in my undertaking. Here is introduced a way to deal with separate the pneumonic clots into threatening & considerate clots to help the radiologist & for the future upgrade. Further loads should be aimed at improving the SVM & CNN for arranging harm levels of tumors through tests with different other options.

REFERENCES

- [1] N.Camarlinghi, "Automatic detection of lung clots in computed tomography images: Training & validation of algorithms using public research databases," *Eur. Phys. J. Plus*, vol. 128, no. 9, p. 110, Sep. 2013.
- [2] R. L. Siegel, K. D. Miller, & A. Jemal, "Cancer statistics" *CA, Cancer J. Clin.*, vol. 66, no. 1, pp. 730, Jan. 2016.
- [3] D. Kumar, A. Wong, & D. A. Clausi, "Lung clot classification using deep features in CT images," in *Proc. 12th Conf. Comput. Robot Vis. (CRV)*, Jun. 2015, pp. 133138.
- [4] P. P. RebouçasFilho, E. D. S. Rebouças, L. B. Marinho, R. M. Sarmiento, J. M. R. Tavares, & V. H.

- C. de Albuquerque, "Analysis of human tissue densities: A new approach to extract features from medical images," *Pattern Recognit. Lett.*, vol. 94, pp. 211218, Jul. 2017.
- [5] W. Shen et al., "Multi-crop convolutional aural networks for lung clot malignancy suspiciousness classification," *Pattern Recognit.*, vol. 61, pp. 663673, Jan. 2017.
- [6] Fangfang Han, Guopeng Zhang, Huafeng Wang, Bowen Song, Hongbing Lu, Dazhe Zhao, Hong Zhao & Zhengrong Liang "A Texture Feature Analysis for Diagnosis of Pulmonary Clots Using LIDC-IDRI Database" 2013.
- [7] Hyo Kyung Lee, Student Member, IEEE, Fengju, Member, IEEE, Raymond U. Osarogiagbon, Nicholas Faris, Xinhua Yu, Fedoria Rugless, Shan Jiang, & Jingshan Li, Fellow, "A System-Theoretic Method for Modeling, Analysis, and Improvement of Lung Cancer Diagnosis-to-Surgery Process", 2017.
- [8] Jue Jiang, Yu-chi Hu, Chia-Ju Liu, Darragh Halpenny, Matthew D. Hellmann, Joseph O. Deasy, Gig Mageras & Harini Veeraraghavan "Multiple Resolution Residually Connected Feature Streams For Automatic Lung Lump Segmentation From CT Images", 2018.
- [9] I.R.S. Valente, P.C. Cortez, E.C. Neto, J.M. Soares, V.H.C. de Albuquerque, & J.M.R. Tavares, "Automatic 3D pulmonary clot detection in CT images: A survey," *Comput. Methods Programs Biomed.*, vol. 124, pp. 91107, Feb. 2016.
- [10] P. Lambin, E. Rios-Velazquez, R. Leijenaar, S. Carvalho, R.G. van Stiphout, P. Granton, C.M. Zegers, R. Gillies, R. Boellard, A. Dekker, et al., Radiomics: extracting more details from medical images using advanced feature analysis, *Eur. J. Cancer* 48 (4) (2012) 441–446.
- [11] M.M. Wahidi, J.A. Govert, R.K. Goudar, M.K. Gould, D.C. Mc Crory, Evidence for the treatment of sufferers with pulmonary clots: when is it lung cancer?: *Accp evidence-based clinical practice guidelines, CHEST J* 132 (3_suppl) (2007) 94S–107S.
- [12] Edson Cavalcanti Neto, Edson Cavalcanti Neto, V.H.C. Albuquerque, Joao Manuel R. S. Tavares, "Automatic 3D pulmonary clot detection in CT images", February 2016.

AUTHORS BIOGRAPHY



Nazia Fatima has completed BE computer Science and now she is submitting the paper for M Tech Computer science from Rabindranath Tagore University, Bhopal, MP, India.



Ms. Ayonija Pathre, MTech, Asst Professor, Rabindranath Tagore University, Bhopal, MP, India.
Research Area: Network Security, Data Mining



Prof. Mukesh Kumar, MTech Asst Professor, Rabindranath Tagore University, Bhopal, MP, India.
Research Area: Image Processing & Info Retrieval & Network Security.

Citation of this Article:

Nazia Fatima, Ayonija Pathre, Mukesh Kumar, "Lung Cancer Detection Using Machine Learning and Deep Learning Techniques" Published in *International Research Journal of Innovations in Engineering and Technology - IRJIET*, Volume 5, Issue 2, pp 34-42, February 2021. Article DOI <https://doi.org/10.47001/IRJIET/2021.502006>
