



An Approach for the Detection and Classification of Bone Tumor using Convolutional Neural Network

^[1]Dr. G Manjula, ^[2]Anusha, ^[3]Divya H C, ^[4]Nayana U S, ^[5]Shwetha K.R

^[1] Associate Professor, Department of Information and Science, DayanandaSagar Academy of Technology and Management, Bengaluru – 560082

^{[2][3][4]} BE Students, Department of Information and Science, DayanandaSagar Academy of Technology and Management, Bengaluru – 560082

ABSTRACT

Cancer may be a dangerous disease among different sorts of diseases. Almost 70% of individuals will be suffering from some type of cancer. Bone cancer may be a reason for major death among all sorts of cancers. Bone cancer is additionally called as bone neoplasm or bone tumor. First, we need to focus on the affected part and stage of disease by using medical imaging techniques like X-ray, CT, PET scan accurately. The primary step is to preprocess the image using median filter as the scanned images does not contain high resolution because of the total number of slices per pixel and noise. Then specific feature are chosen using genetic algorithm and extracted using GLCM from the preprocessed images. The CNN is used to classify the extracted images and storing purpose, this classification is useful to identify the stage of disease which helps physician to give remedial suggestions. 95-98% of accuracy can be achieved using proposed methodology.

Keywords: Feature extraction, Bone tumor, CNN, Tumor size, MRI, CT or PET scan, Image segmentation, Edge detection.

1. Introduction

Bone cancer may be a reason for major death among every type of cancer. Bone cancer is additionally called as bone neoplasm or bone tumor. The target is to seek out the injured part. i.e. unrestricted growing and stages of infection in an precise method by analyzing two medical imaging techniques like XRay, PET and CT scan. The examined figure might not have high level determination because the number of segment per pixel and noise, so the beginning steps to pre-processing the

pictures with the help of median filter and that is employed to get rid of the rumpus in a picture. In the preprocessed figure specific features are pick out with the help of genetic algorithms and remove using GLCM. The removed figures are analyze using CNN and stocked, supported grouping must recognize the phase of infection which is beneficial to the physician to allow therapeutically proposition. The advanced method outcome appear to enhance the rate of initial perception of Bone tumor. A bone tumor may be a malleable extension of cells in bone. Unypical extension that integrate bone will be either innocuous or malignant." Neoplasm of the bronchus, pituitary& kidneys are the initial sarcoma that usually refine to the bone. The physical subdivision of the MRI bone image is divided into two problems: 1) It is a monotonous and time consuming work. So that can be performed only by specified doctors; and 2) It is capable of real fidelity. The certain problems might be overcome with the help of automated bone figure partitioning systems, and that has the capability to enhance the flow of work in a distant place and reduce the flexibility between userpartition.

1. In this work proposes a deep learning method for automatic detection and visual analysis of malignancy regions in whole slide images of bone tumor.
2. The outlook is the same as to how the human brain uses layers of most representative and useful features, which leads to hierarchical learned representation.

Corresponding author.

E-mail address: divyackmhc@gmail.com

3. These techniques show traditional approaches of most difficult issues in different areas such as speech recognition and object detection. The convolutional neural networks (CNNs) is the deep learning framework used here.

2. Literature Review

The paper[1] - proposed a methodology for recognizing the bone cancer cell size and phases of the identified disease by utilizing GGD investigation. Bone sarcoma is a kind of malignancy that alludes to a sporadic development of some tissue within the bone, with high likelihood to spread to different pieces of the body. It influences youngsters, teens and youthful grown-ups. There are no recognized reasons for bone malignancy with respect to any remaining kinds of disease (bosom, lung, prostate, stomach, cerebrum...). Along these lines, just an early identification assists with expanding the odds to endure a bone tumor. The interrelation of clinical imaging modalities, with picture preparing methods can give more exactness while discovery inevitable bone tumors. GGD is utilized to make sub-pictures from the handled bone MRI and play out a GGD investigation of the sub-pictures. From that point onward, a locale of interest identified with the sub-pictures with the most elevated estimation of the shape boundary α is chosen from the first MRI. Locale of Interest is differed utilizing Euclidean distance measures and afterward the associated part examination calculation is utilized to distinguish the bone tumor." This method was tried on different bones MRI and has given wonderful tumor identification.

The paper[2] - Bone malignant growth is quite possibly the most perilous and fundamental explanations behind early passing far and wide. This procedure proposes a technique for the identification of bone malignancy utilizing fluffy C- mean grouping. Confirmed 120 patient MRI of bones utilized for the checking exactness of the proposed technique. Versatile neuro fluffy surmising framework (ANFIS) is utilized in this investigation for the arrangement of amiable and threatening bone malignant growth. GLCM highlights have been taken from the MRI pictures for the preparation and testing of the ANFIS organization. A reasonable get approval is continued the gathered bone pictures are separated into preparing and testing pictures. The arrangement result has been tried dependent on the exhibition measurements precision, affectability and particularity. The proposed procedure gives 93.75% exactness in bone tumor order.

The paper[3] - The danger recognition at beginning phases is one of the exploration theme. A great deal of work has been done on the tumor location and recognizable proof. Along these lines, the purpose of conversation is to discover the recognizable proof and identification framework which is quick and solid. In this paper a methodology of tumor location utilizing AI have been talked about and the informational collection for the presentation investigation is MRI pictures. At that point calculation have been actualized in open CV to make the framework all the more quick and helpful.

The paper[4] - Bone scintigraphy is utilized to analyze tumor metastases. It yields high hugeness to precisely find and portion hotspots from bone scinti graphs. Prior PC supported analysis strategies generally center around recognizing variations from the norm. This paper propose another system that deals with the two assignments simultaneously. Right off the bat we utilize inadequate auto encoders and CNN to prepare a picture level divider that marks input pictures as typical. For suspected pictures, MIL is applied to prepare a fix level classifier. This classifier will be utilized to create a likelihood guide of hotspots. In this manner, level set division is conveyed with the likelihood map as beginning condition. The test results show that our strategy is more exact with hearty than different strategies. Our proposed approach utilizes Convolutional neural organization which has a preferred position of less expense, and it includes least human obstruction which prompts more noteworthy precision. It is useful for specialists to analyze the malignancy in prior stage with better proficiency which will accommodate for the simple fix. Helpful in various element and therapeutic executions, it is valuable in PC guided a medical procedure.

The Paper[5] - A proposed approach incorporates some pre-preparing strategies, for example, the normal channel and the reciprocal channel to eliminate commotion and to smooth pictures. This will build the nature of the pictures so they are reasonable for division just as morphological tasks that will be utilized to kill bogus fragments. The MRI bone malignant growth pictures are investigated to distinguish the presence of bone disease and to decide its stage dependent on the calculations of mean force and tumor size utilizing k-implies algorithm. The exploratory outcomes illustrate, the proposed framework could get the smooth picture with edge exhibiting the illness affected part without the spatial and ghostly commotions.

The paper[6] - Skeletal bone age appraisal as a basic imagine assessment, targets assessing the organic and underlying development of human bones. In the clinical practice, Tanner and Whitehouse (TW2) strategy is a broadly utilized technique for radiologists to perform BAA. The TW2 strategy parts the hands into Region Of Interests (ROI) and analyzes every one of the anatomical ROI independently to appraise the bone age. Be-reason for thinking about the investigation of neighborhood data, the TW2 strategy shows precise outcomes practically speaking. Following the soul of TW2, they supportive of represent a novel model called Anatomical Local-Aware Network (ALA-Net) for programmed bone age evaluation. In ALA-Net, anatomical nearby extraction module is acquainted with gain proficiency with the hand structure and concentrate neighborhood data. Also, we plan an anatomical fix preparing procedure to give additional regularization during the preparation cycle. In this model can identify the anatomical ROIs with gauge bone age mutually in a start to finish way. The exploratory outcomes show that our ALA-Net accomplishes another best in class single model execution of 3.91 mean total mistake (MAE) on the public accessible RSNA dataset. Since the plan of our model is well steady with the all-around perceived TW2 technique, it is interpretable and dependable for clinical utilization.

The Paper[7] - They applied significant convolutional neural organizations (CNN networks) to recognize periodontal bone misfortune (PBL) on all encompassing dental radiographs. They arranged a lot of 2001 picture partitions from sweeping radiographs. their reference test was the deliberate % of

PBL. A profound feed-forward CNN was prepared what's more, approved through 10-times repeated pack shuffling. Model designs and hyper parameters were tuned utilizing framework search. The final model was a seven-layer profound neural organization, defined by a absolute number of 4,299,651 loads. For assessment, six dental specialists were evaluated the picture portions for PBL. Arrived at the midpoint of more than 10 approval overlap the mean (SD) classification exactness of the CNN was 0.81 (0.02). Mean (SD) affectability and specificity were 0.81 (0.04), 0.81 (0.05), individually. The mean (SD) exactness of the dental specialists was 0.76 (0.06), yet the CNN was not factually significant better thought about than the inspectors ($p=0.067/t$ -test). Mean affectability and specificity of the dental specialists was 0.92 (0.02) and 0.63 (0.14), individually. A CNN prepared on a restricted measure of radiographic picture fragments appeared at least comparable segregation capacity as dental specialists for evaluating PBL on all encompassing radiographs. Dental experts' analytic efforts when utilizing radiographs might be diminished by implementing AI based advances.

The Paper[8] - Bone malignant growth which may happen inside or on the bone can be perilous like the other kinds of harm. The point of this paper isto improve the accuracy of the discovery cycle. At the present time, the disclosure measure is done using information(data) mining procedures and picture pre-preparing techniques as a feature of a clinical picture investigation measure, utilizing a non- computerized(non-electronic) structure which incorporates picture obtaining, picture separating, picture division, the zone of interest (force of the foundation or the portioned cuts) and characterization strategies to assess the choice. In spite of the fact that these techniques are successful somewhat, the existing strategies have a few restrictions through lack discovery regards, an expansion in the preparing time what's more, precision. The outcome demonstrates that by utilizing eigenvalues and eigenvectors, the preparing time can be diminished by executing standardization, though educating identification exactness. This paper researches the reasonability of utilizing surface grounded attractive reverberation imaging (MRI) to find various groups and arrange zones for deciding bone malignancy. This division and arrangement measures are done by utilizing eigenvalues and eigenvectors. By using these methods.

This paper presents a thresholding strategy for picking the most noteworthy eigenvectors (in the wake of breaking down pictures into bunches) and computing the mean normal qualities to segregate between malignancy/no disease. A way to deal with sectioning a tumor or disease part from a picture is actualized utilizing eigenvector disintegration for the extraction of the bone tumor part. The extricated sectioned picture is further prepared to assess the mean normal in the chose bunches (area of interest). With the mean normal esteem edge known, developed precision charges are accomplished in the recognition of bone malignancy. The planned strategy accomplishes a higher pace of precision and recovers the swiftness (preparing time) with which the tumor is recognized. The impediments of the proposed work lies in the chance of bogus worth location when achieved on the wrecked bone pictures. Additional investigations center on utilizing eigenvector deterioration procedures in various sorts of tumors of the mind, spine and so on and furthermore endeavor executions on bone injuries.

3. Objective of the project

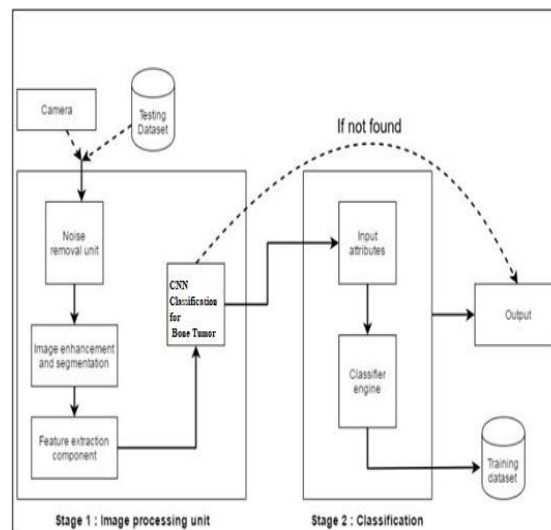
This work proposes a deep learning method for automatic detection and visual analysis of malignancy regions in whole slide images of bone tumor. The outlook is the same as to how the human brain uses layers of most representative and useful features, which leads to hierarchical learned representation. These techniques show traditional approaches of most difficult issues in different areas such as speech recognition and object detection.

The convolutional neural networks (CNNs) is the deep learning framework used here.

4. System Overview

4.1 System Design

Overall system architecture:



4.2 System Implementation

The framework can be comprehensively sorted into following significant stages:

Acquisition of image: Images are obtained either by lens or by secretly deleting them from the contraction. Whatever the source may be, it is very important that the image of the data is transparent and cautious. An incredible picture is needed for this.

Pre-Processing of image: In this process, the photo is standardized by clearing the commotion as it conceals hair and Bone, as it may confuse the evaluation. Similarly, the image given as the information may not be of standard size as required by the figure, so it is vital that the image size needed is obtained.

Data storage aspect to preserve information images for testing and training: if controlled learning will occur, as is the case here, it is important to prepare data sets. The sample database is the images collected during the photo procurement process.

Classifier to classify the type of Bone disease: The classifier used here is the last layer of the system which gives the true probability of each experience. The project involves two major parts: the Image preparation unit and grouping unit. The object processing system enhances the image by removing the clatter and noisy bits. The Bone and the image will then be isolated into different segments to isolate the Bone from running the mill after the image features are evacuated to check whether or not the Bone is contaminated.

Noise reduction unit: removes from the photo the unwanted colors.

Image enhancement unit and segmentation: carries the affected part to the middle by improving the area and dividing the area into different segments in order to isolate it from the normal ScannedImage.

Feature Extraction Component: One of the notable developments in any gathering-centered issues is highlighting extraction. Looks are the cornerstone for both purposes of planning and screening. This feature contains noteworthy image information that will be used to identify the disease. Identification unit for Cancer disease: See if the cancer is considerate or hazardous.

Input Attributes: For example, all noteworthy attributes, asymmetry, edge, concealment, distance, progression, etc. that have been expelled from the image are now provided as a dedication to Part II, which is the classifier part.

Classifier engine: characterizes the images by grouping the calculation into one of the predefined diseases. The proposed system used to extract the features of bone and it used to increase the resolution and efficiency level of an input. Here, the view of bone density gives the exact level of a density and it also finds all supplements in bone. To find a bone cancer and fracture it gives high resolution to find the level of a cancer and fracture in the bone. It is the combined process of a bone feature extraction.

- Preprocessing
- Feature Selection
- Feature Extraction
- Classification

The current methodology explicates about the reasons of Bone tumor and describes the management of respiratory motion. Even though PET/CT causes some difficulties for staging the disease and distant metastasis, it is normally used in best diagnosis. Comparison is made between PET and CET to understand which gives the leading and accurate result. Our proposed methodology is implemented for recognize the Bone tumor at initial stages.

Preprocessing:

We need to extract the concerned part of the images without any noises. and blurriness from the images, then images need to be preprocessed by applying pre- processing techniques like Filtering, Histogram equalization etc. Python software is used to preprocess the images. The main aim is to preprocess the images is to remove the extravagant existing in scanned images. Every image is preprocessed to remove the noise and to enhance its quality. The techniques used for preprocessing are: 1. Histogram equalization method is to magnify the contrast of the image.

2. The effect of deficient contrast because of noise, blaze is to be removed by median filter.

In this system, CT and PET scanned images are taken primarily. normally the medical images does not contain high resolution because quantity of slices per pixel and contain noise. After that image should be preprocessed by using preprocessing techniques. The infected region should be selected from the segmented preprocessed image. GLCM algorithm is used to extract the selected features and stored in the database. The stage of the tumor should be identified based on that.

Feature Selection:

Small set of relevant features for future use are selected in this process. After preprocessing, features are selected from the preprocessed image using genetic algorithms.

Feature Extraction:

The amount of resources required from a large set of data accurately can be determined by feature extraction. Features are required to be extracted after features are selected. It plays a major role that uses algorithms and techniques to find out the different parts and the selected features must be extracted. The GLCM describes how various associations of pixel values carried out in an image. Initially create gray-level co-occurrence matrix utilizing gray co matrix function in CNN. A GLCM shows the second order tentative joint probability densities which is the probability of occurrence of grey levels i and j inside a given distance „ d “ and along the direction „ θ “. 9 features are chosen for the specified method.

Classification:

The dataset should be segregated into training and testing sets, training and testing set shouldn't contain any irrelevant data. one unlike distribution of image size in the dataset also reduces the efficiency of neural network.

The conversion of the images to black and white format should be done before providing it to the convolution layer because scanning the images in

RGB which need a 3-D numPy matrix which reduces the executiontime. Before providing itinto the neural network all blurred images should be trimmed.

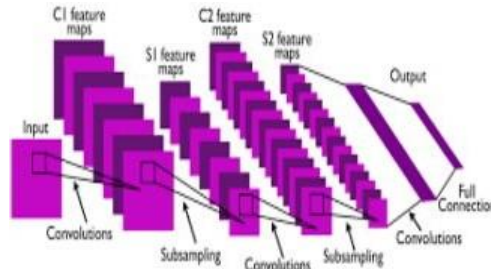


Fig 3: CNN layers

Convolution layer

Convolutional layer scans the entire image for patterns and map it in the form of a 3x3 matrix. Kernel is the convolved feature matrix of the image. Every value in the kernel is termed as a weightvector.

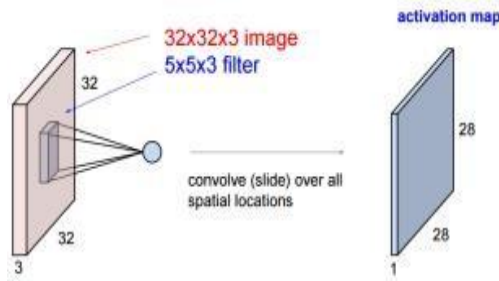


Fig 4: Convolution layer

Pooling layer

After the convolution come to the pooling, Image matrix is split into the sets of 4 rectangular segments. There are two types of pooling Max pooling and average pooling. Max pooling gives the larger value present in the relative matrix region and average pooling gives the average value. The advantage of the pooling layer is that it decreases over-fitting possibilities and raises computer performance.

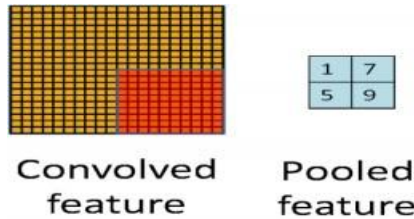


Fig 5: Pooling layer

Activation layer

Values are normalized in the activation layer and they are suited in a specific range. ReLU allows the positive values and then rejects the negative values. It is of low computational cost.

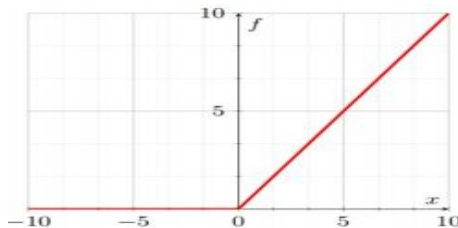


Fig 6: ReLU function

Fully connected layer

In this layer, the features are compared with the features of the test image and find out the similar features with the detailed label. For computational ease, labels are encoded in the form of numbers, then it will be transformed into their respective strings.

5. Conclusion

The systematic processing technique is examined as a vital step to enhance the grossgraphical representation of clinical images, and provides better diagnosis outcomes.

The proposed system of bone tumor detection with superpixel segmentation is implemented using Open CV Python. Also the detection of bone cancer is carried out with the given set of images. The proposed system is specially dedicated for bone cancer detection. The same system can be further extended to identifying the stages of cancer. effortless, fast and steady technique to detect cancerous tissue in bone by utilizing various image pre-processing techniques like contrast enhancement, edge detection and image fusion. The experimental outcomes shows, the proposed method could obtain a smooth image with an edge showing the disease affected part without the spatial and spectral noises.

6. Result



Fig: This page helps to train and test the data

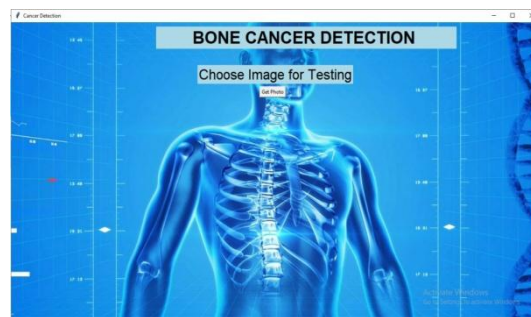


Fig: This page is for selecting the image for testing the disease.



Fig: This page will display the result of cancer.

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