

Design And Analysis Of An Improved Deep Learning Algorithm On Classification Of Intervertebral Discs

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Abstract

In this paper, the classification of Intervertebral Disc after scanning of spinal vertebra. In this classification process analysis and classification is carried out using a deep learning classifier that involves Deep Learning Algorithm namely Convolutional Neural Network. This method uses ConvNet method for analyzing and reduction of dimensions that extracts the essential features and therefore it helps in classification of vertebra. The evaluation shows that the proposed method offers increased speed and accuracy with reduced computational time than the existing machine learning algorithms.

Keywords: CNN Algorithm, Intervertebral disc, Classification and Prediction

I. INTRODUCTION

Intervertebral Disc (ID) is a lint-like mattress that separates the bones, which are considered a set of spines. These discs help keep the spine column stable. The ID acts as a shock absorber between each vertebrae in the vertical of the spine. In doing so, the axons in between ID are kept wholesome and protection. The discs in the mid of the spine are filled with little fluid. If those discs are found to be dry without fluid, there are occurs some disorders. Drought is dehydration in the tissues causing a few problems. These discs absorb impact and trauma, protecting the bones without friction from overlapping. Discs the disorders caused by that drought are illustrated in the figure 1 below.



Fig.1: Disc Degeneration

Normal Disc

There are 23 discs in the spine. There are 5 discs in the pelvis, 12 discs in the thorax, and 6 discs in the cervix. Generally normal discs are healthy, and these will keep the spine very secure.

Degenerative Disc

These types of discs are age-dependent, that means a few discs in the between of the spine deteriorate; this causes the spine to break down and cause pain. Thus, the back cannot bend. This means that the spinal disc is torn and causes back pain.

Bulging Disc

When this type of disc is annually intact, a swelling occurs; this inflammation causes a pressure on the nerves. It is caused by aging, Swelling can only increase when there is enough swelling to cause components of the spinal canal.

Herniated Disc

A part of the disc nucleus is called the herniated disc. It is ejected from the annulus, and then, through decay or tears, into the spinal canal. Discs that become hernia are in the early stages of degeneration.

Thinning Disc

Reason for this type of discs is, the discs become thin due to dehydration, and thus the distance between the vertebrae begins to decline. As the distance decreases, the bones are friction with each other, and thus, the unwanted bones begin to grow. This puts pressure on the nerves and causes pain.

The purpose of this paper is therefore to analyze the scan image of spinal. Analyzing them and then categorizing the deformed discs. That is, the discs are classifying as specified in the figure 1. So, this way, the position of the discs can be detected and then treated quickly. Doing so requires more precise results to detect the state of the discs.

In paper [1], image segmentation and localization are considered important for diagnosing spinal cord disease. So they have analyzed the image here using some method of deep learning. That is, in this paper, he localized the centers of the spine. But their accuracy is not that useful. In paper [2], some injuries due to attention deficit disorder are occurring on ID. Two types of failure models have been compared in this paper. In the first failure model, there was a maximum strain applied to the embryo and annular organs. In the second failure model, in the middle of the ID, a split plane was operating between two divisions with a built-in interface. So both models have failed. In paper [3], author of this paper have attempted to separate the spinal column with confidence and accuracy, for that he used the DCNN method. But dealing with this task is a very difficult task. In paper [4], the CNN system is utilized to detach the spinal column image. However, with the usual CNN structure, it is very difficult to accurately separate the image. In paper [5], they used the integration networks to separate the MRI images of the spine; this method does not work because they have inappropriate backgrounds.

In paper [6], the FCN algorithm was used to analyze the scan image of the spine. But its results were not effective and take more time consuming. In paper [7], for classification of DDD images, the FEM model is used, but its results were not enough. In paper [8], used the HOG method to analyze the LBP image, the results were not accurate and required a high cost. In paper [9], they used the PEF method to analyze the IVD image, but their results are not perfect.

II. PROPOSED METHODOLOGY

Here the CNN algorithm is used to analyze the ID image that is the vertical of the vertebrae. ConvNet mechanism is one of the deepest learning methods and it takes an image as an input and segregates and categorizes their important features. The CNN method is very efficient for pre-processing and classification exercises. The structure of this ConvNet method is thought to be related to the adjunct of neurons in the anthropoid brain. An image is nothing but matrix of pixel values. So the only scheme to do this is to kind and classifies the image, so this CNN process is the best way to finish this.

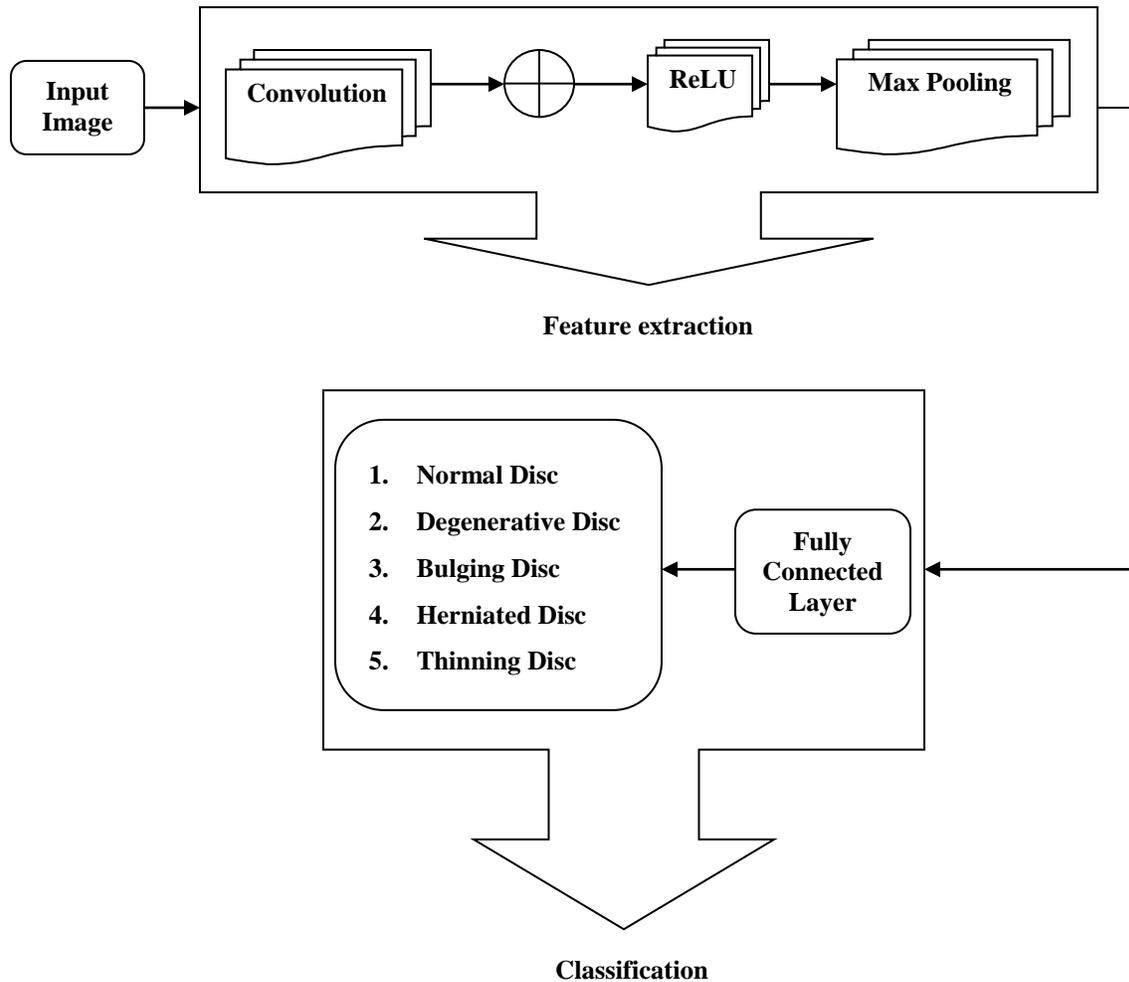


Fig.2: Flow of CNN Algorithm

This method accurately displays the mean scores when making predictions of classes of binary images. With the use of related filters, a CNN method can successfully apprehend the worldly and spatial dependencies in a figure. This structure makes a good fit because it reduces the number of parameters involved.

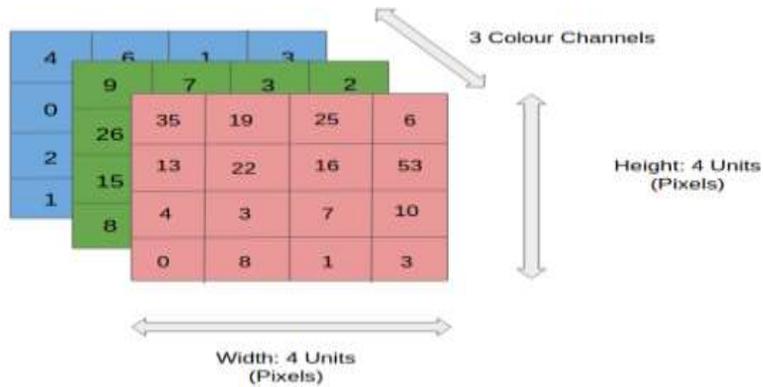


Fig.3: 4x4x 3 Red-Green-Blue Images

In the 3rd of the picture is the RGB image, which is separated by 3 color levels such as red, green and blue. Similarly, there are so many colors. Once the dimensions of the images have been reached, a good prediction is to be obtained computationally. This method is very useful for image processing without losing important features.

Convolution Layer

Figure 3, which is considered an input image and it is in the 5*5*1 matrix. In the 1st piece of a convolutional sheet, in carrying out the transformation process, the involved element is called kernel, this is referred to in yellow, and the K element is considered the 3*3*1 matrix.

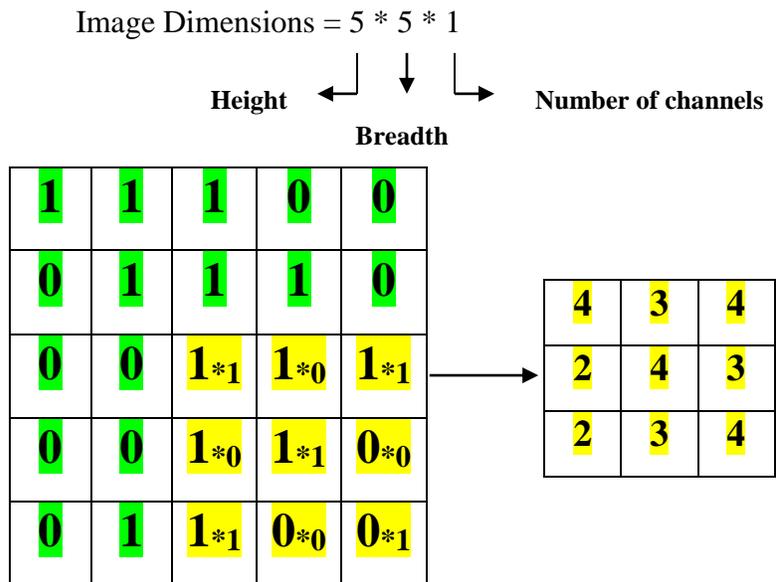


Fig.4: Geta 3x3x1 Convolved Feature

The Kernel changes nine times because the value of the stride length is 1, each time the transform multiplies between the portion and k of the image. The filter moves right to a specified stride value until it is parsed a full width. The whole picture repeats the process until it travels.

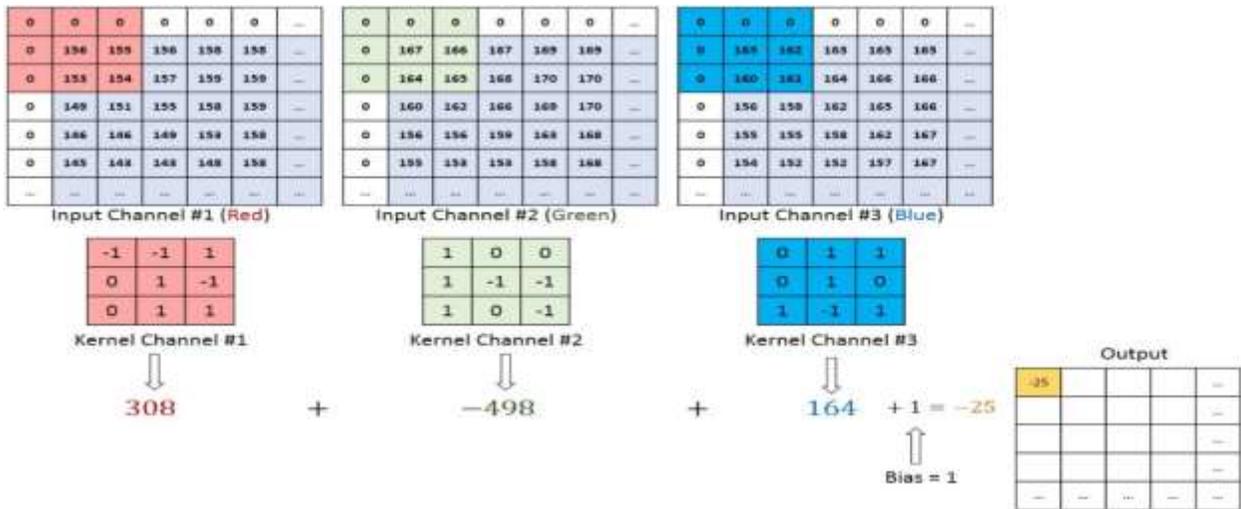


Fig.5: Convolution Operation on an $M \times N \times 3$ Picture Matrix with a $3 \times 3 \times 3$ Kernels

In the activity of images with several channels, the kernel has the identical depth as the input picture. Matrix multiplication is executed between the K_n and in stack $([K_1, R_1]; [K_2, R_2]; [K_3, R_3])$.

Pooling Layer

The pooling sheet is utilized to lessen size of the region of the desegregated attribute. It minimizes the calculation energy essential to actions information through dimension reduction. This is useful for the extraction of stable features, thus preserving the training process of the model.

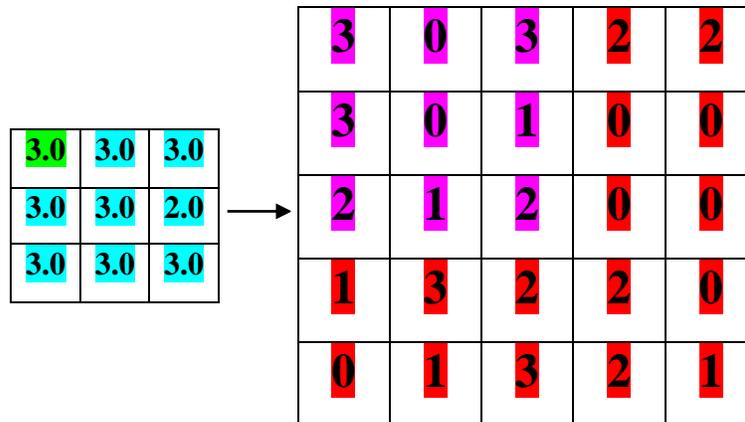


Fig.6: 3×3 Pooling above 5×5 Convolved Feature

The Max pooling method works much better than the average pooling here. Max pooling gives maximum value from part of the image, and this also suppresses noise, this reduces the dimensionality and reduces the noise. Amalgamate the pooling layer and the convolutional layer generates the r^{th} layer of a ConvNet.

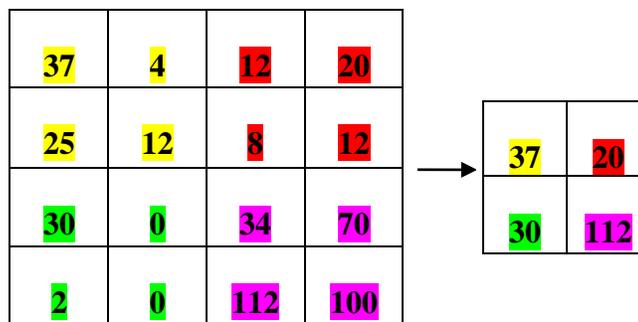


Fig.7: Max Pooling

Then, to give an input to the neural network in order to classify the final output.

Fully Connected Layer (FCL)

The fully connected part of the CNN network goes through its back propagation process to determine the most accurate weight. Each neuron receives preferential weights for the most appropriate labels, and neurons each, voting for each of the labels. Finally, Winners who receive votes will be selected for classification results. Typically in this fully connected network, there are six levels.

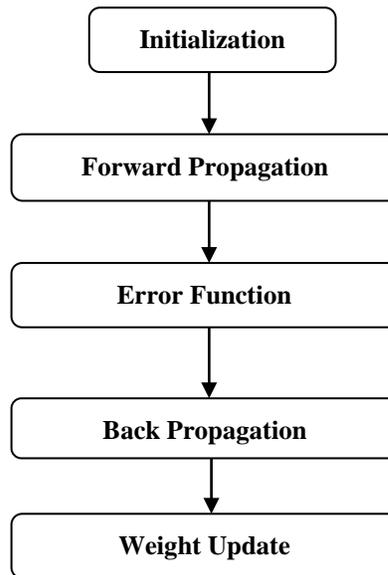


Fig.8: Flow of FCL

STEP 1: Initial weights are also used for

STEP 2: Inputs from the training set are used and an output is calculated.

STEP 3: Given the existing sample weights, finds the delta between the real outcomes of the replica and the correct output.

STEP 4: Back propagation's job is to change the weight of the neurons to reduce the error's activity.

STEP 5: The weights are shifted to optimal values according to the function of the back propagation mechanism.

STEP 6: The weights are updated at a time by a small delta; the network needs to perform many iterations to understand those actions, after each iteration, towards a lower loss function, renews weight.

The magnitude of the actions required to merge, depends on the network Meta parameters, the optimization used and the learning rate.

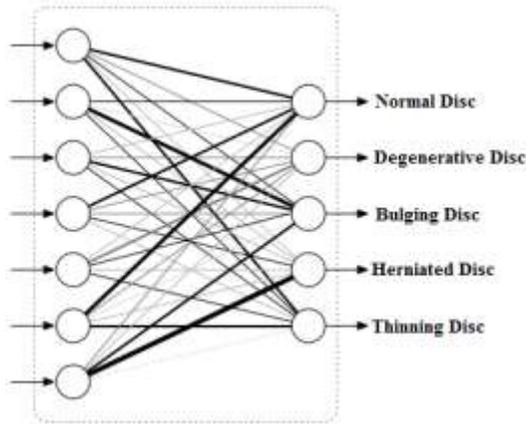


Fig.9: Output of FCL

Thus this CNN method releases the output very accurately. Thus, this method divides the disc images of man into five types of disc images categories.

III. RESULTS AND DISCUSSION

In this section, we describe in detail the performance of functionalities of the algorithm used in this paper. That is, in this paper, the CNN algorithm is used. The result of this method used is very effective. This means that the output of this process is more accurately obtained.

Accuracy calculation has the following definition:

$$Accuracy = \frac{\text{no.of exact divinations}}{\text{total no.of divinations}} \times 100$$

$$= \frac{160}{165} \times 100$$

$$Accuracy = 96.9\%$$

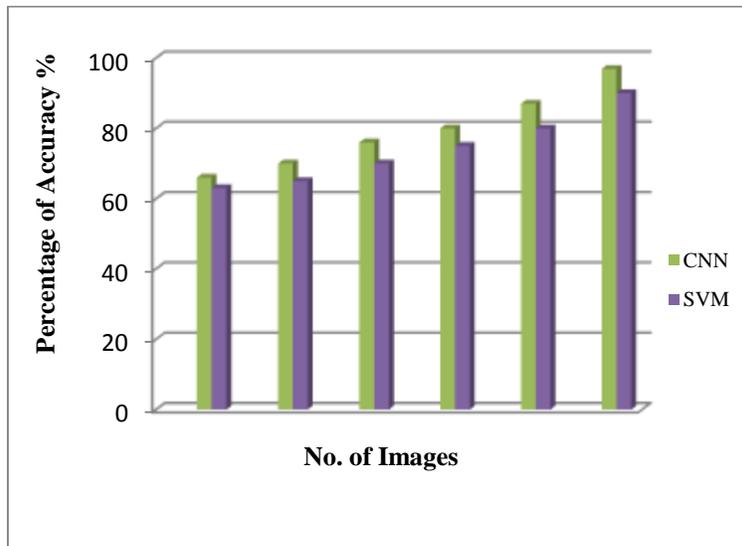


Fig.10: CNN Performance

So in this paper, the percentage of precision is 96.9%, and the percentage of the SVM method is 90 percent. So what we know from this is that the CNN system is more efficient than the SVM system.

IV. CONCLUSION

In this paper, the intervertebral discs image are analyzed and classified into various disc types using CNN ConvNet algorithm and offers improved performance than the existing SVM system. This paper, hence provides an improved model for clinicians for its improved accuracy on classification.

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