DNA Classification using Machine Learning for Detecting Genetic Disorders

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Abstract-Deoxyribonucleic acid is a double-helical molecule composed of two chains that contains genetic instructions. Genetic diseases are caused by changes in pre-existing genes. A genetic abnormality results from the alteration in chromosomes. DNA pattern recognition is a major issue in bioinformatics. DNA is classified into several categories on the basis of Structure, Location, Number of base pairs etc. Traditionally the DNA molecule is studied by extracting it from the blood sample and is then manually analysed to find out the abnormalities. To increase the accuracy, a machine learning based DNA classification is done which helps in studying the extracted DNA image using various techniques. This consumes minimal amount of time and is more efficient. The image is pre-processed using median filter and Canny edge detection. DNA sequences can be recognized correctly and effectively without any uncertainties with the help of Neural Network. The network successfully classifies an image given as input when it is trained with patterns. Thus, we can analyse if a person has a genetic disorder.

Keywords-DNA, Canny edge detection, Neural Network, Genes.

1. INTRODUCTION

A gene is a sequence of nucleotides in DNA which determines a person’s appearance and other characteristics. The different DNA sequences made up by genes are called genotypes. Some of the genetic qualities are promptly visible, such as eye colour, while others are not, such as blood type and the risk for certain diseases. To sequence the DNA into different types we classify it. DNA is classified on the basis of seven categories-
(1) Number of base pairs per turn
(2) Location
(3) Structure
(4) Coiling pattern
(5) Number of strands
(6) Coding and Non-coding DNA
(7) Nucleotide sequence

Traditionally, DNA classification was done manually by extracting the DNA from the blood sample using many steps. Then the principle component analysis is done to classify the DNA.

The major Disadvantages are:
- Inaccurate results
- Time consuming
- Not valid pattern detection for multiple images in a short time

The proposed method uses image processing and machine learning techniques. It involves basic four steps-(1) Pre-processing-The main aim is to supress undesired distortions. It involves using median filter and Canny edge detection.
(2) Dual tree complex wavelet transforms-Wavelets are used to remove noise from two dimensional signals, for example images. The decomposition of the image is done and then it is reconstructed from the modified levels.
(3) GLCM Co-occurrence Matrix-It analyses how frequently a pixel with grey-level value i occurs in any direction to adjacent pixels with the value j.
(4) Neural Network for image classification- The Neural Network contains parameters sometimes also referred to as neurons, receives multiple inputs and produces an output. It performs feature extraction and then classifies the DNA accordingly.

II. RELATED WORKS

In one of the papers, the DNA pattern recognition is done using sequence compressor.
The major disadvantage of this is that due to compression, image quality is reduced. Another method is the robust classification of DNA damage patterns in a single cell using SVM classifier which has a big drawback of low throughput. Other authors have also proposed one double-stranded DNA probes as classifier. This also has a disadvantage of its weakness to noise.

III. METHODOLOGY

In the proposed system an image is taken of the extracted DNA sample. First the image is pre-processed to remove any noise or outliers using techniques such as median filter and canny edge detection. Then the Dual-tree complex wavelet transform is applied. Wavelets are used to remove noise from two dimensional signals, for example images. The decomposition of the image is done and then it is reconstructed from the modified levels. Finally, neural network classification is done. DNA sequences can be reconstructed from the modified levels. Finally, neural network is trained with patterns.

![Fig.3 Block Diagram](image)

A. Pre-Processing

Image Pre-processing refers to the operation performed on any image at the lowest level of abstraction. Intensity images act as both input and output in this case. Its main aim is to improve the image by removing unwanted distortions and enhancing features important for further processing. Corruption can be of many forms such as camera misfocus, noise and motion blur. Image restoration is also done but it is different from image enhancement because it emphasizes on that features of the image which make it more pleasing to the viewer, but do not produce realistic data. Pre-processing is done using two techniques Median Filter and Canny Edge Detection.

B. Median Filter

It removes noise from the image. It is a part of pre-processing. A Median Filter selects the median intensity in the window that it operates upon. The pattern of neighbours is known as the "window". It slides, entry by entry, over the entire signal. After that, each entry is replaced by the median of its neighbouring entries. For 1D signals, the window is defined by the first few former and following entries, whereas for 2D data the window must include all entries within a given radius.

\[ x = (3, 6, 75, 9) \]

The output signal \( y \) is:

\[ y_1 = \text{median} (3, 6, 75) = 6, \]
\[ y_2 = \text{median} (6, 75, 9) = \text{median} (6, 9, 75) = 9, \]
\[ y_3 = \text{median} (75, 9, 3) = \text{median} (3, 9, 75) = 9, \]
\[ y_4 = \text{median} (9, 3, 6) = \text{median} (3, 6, 9) = 6, \]

i.e. \( y = (6, 9, 9, 6) \)

C. Canny Edge Detection

Canny edge detection is done to detect the wide range of edges in an image. Edges isolate the main image from the noise. Thus, it becomes easy to work on the image. The steps to do so are:-

1. Use Gaussian filter to denoise the image.
2. Find out the image’s intensity gradients.
3. Perform non-maximum suppression in order to remove forged response of edge detection.
4. Apply double threshold for determining the potential edges.
5. Suppress all other weak edges and those connected to strong edges.

The edge detection is done in the following way:

Any edge direction falling within the range of (0 to 22.5 &157.5 to 180 degrees) is set to 0 degrees. Edges falling in the range (22.5 to 67.5 degrees) is set to 45 degrees. Other edges between (67.5 to 112.5 degrees) is set to 90 degrees. Finally, any edges falling within (112.5 to 157.5 degrees) is set to 135 degrees.

![Fig.3 Edge Detection](image)

Thus, the edges following in the area are set to one value for example 0. This reduces the number of edges making the working on the image easier.

D. Dual-Tree Complex Wavelet Transform

The dual-tree complex wavelet transform (CWT) is better than the discrete wavelet transform (DWT), because it has significant added properties. It is directionally selective in two and higher dimensions and also shift invariant. Wavelets are used to remove noise from two dimensional signals like images. The decomposition of the image is done and then the reconstruction of the image from the modified levels is the final step.
E. GLCM Co-Occurrence Matrix

The GLCM calculates how often a pixel with grayscale intensity value \(i\) occurs to adjacent pixels of the value \(j\) horizontally, vertically, or diagonally. GLCM direction of Analysis:

1. Horizontal (0°)
2. Vertical (90°)
3. Diagonal: Bottom left to top right (-45°) or Top left to bottom right (-135°)

It has the following properties:

- Contrast: Measures local disparities and texture of shadow depth in the GLCM.
- Correlation: Measures the joint probability existence of the defined pixel pairs.
- Homogeneity: Measures the proximity of elements in the GLCM to the GLCM diagonal.

F. Neural Network Classifier

The neural network is a set of functions, also called parameters. It works by training the computer from analyzed data. Each parameter, which is also referred to as neurons, is a function receiving one or multiple inputs and producing an output. Those outputs are then passed on to the next layer of neurons. This process continues until the last layer of neurons has received its input. The terminal neurons produce the output which is the final result for the model. So, the network successfully classifies an image given as input.

All NN networks have four layers:

1. The input layer contains only a single neuron for each predictor variable. Typically, N-1 neurons are used where N stands for the number of categories.
2. The hidden layer has one neuron for every case in the training data set. That neuron holds the values of the predictor variables along with its target value.
3. The pattern layer is different for both NN networks and GRNN. NN networks have one pattern neuron for each category of the target variable.
4. The decision layer is the last layer. In NN networks, it compares the weighted votes for each target category and uses the largest vote to predict the target category.

IV. RESULT

The proposed system was tested on DNA images taken from various samples and the classification of the DNA pattern was done successfully. First the input DNA image is fed into the system. After that, the noise and unwanted parts are removed from the image. Then, many techniques and functions are applied to the image. Finally, the neural network is used to classify the type of DNA.

V. CONCLUSION

The DNA pattern classification is a major problem in the medical field. DNA of a human being contains genes which are responsible for inherent features of a person such as appearance and also susceptibility to certain diseases. This project solves the problem of DNA classification and reveals important information about the person like his/her likelihood to catch a certain infection or develop a genetic disease. Thus, it is an efficient, accurate and time-saving method of DNA classification, thereby acting as an invaluable aid to medical science.

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Fig.4 Neural Network Architecture

First the image is taken from the extracted DNA sample. Then the image is pre-processed to remove the noise and outliers. It is done by the median filter and canny edge detection. After this the dual-tree complex wavelet transforms GLCM are applied. These help to enhance the image for feature extraction. The data set contains training examples which help the neural network to classify the type of DNA.
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