Fuzzy Logic and Neural Networks for Disease Detection and Simulation in MATLAB

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ABSTRACT

This paper investigates the integration of fuzzy logic and neural networks for disease detection using the Matlab environment. Disease detection is key in medical diagnostics, and the combination of fuzzy logic and neural networks offers an advanced methodology for the analysis and interpretation of medical data. Fuzzy logic is used for modeling and resolving uncertainty in diagnostic processes, while neural networks are applied for in-depth processing and analysis of images relevant to disease diagnosis.

This paper demonstrates the development and implementation of a simulation system in Matlab, using real medical data and images of organs for the purpose of detecting specific diseases, with a special focus on the application in the diagnosis of kidney diseases. Combining fuzzy logic and neural networks, simulation offers precision and robustness in the diagnosis process, opening the door to advanced medical information systems.

KEYWORDS

Fuzzy logic, Neural networks, Disease detection, Matlab simulation, Medical images, Diagnostics, Uncertainty, Modeling

1. INTRODUCTION

Kidneys are vital organs in the human body, responsible for filtering blood and maintaining electrolyte balance, and thus an important factor in maintaining health. Kidney diseases, such as chronic kidney disease (CKD) or acute kidney injury (AKI), often go unnoticed in the early stages, which reduces the chances of effective treatment. Early detection of these diseases plays a key role in preventing serious complications and improving the quality of life of patients.

Traditional diagnostic tools, such as laboratory tests and ultrasound, are used to monitor kidney health, but they have their limitations. In this context, advanced technologies such as fuzzy logic and neural networks offer an innovative approach to kidney disease detection. Fuzzy logic enables the modeling of uncertainty and uncertainty in medical diagnoses, while neural networks have the capacity to process images and extract valuable information from medical visual data.

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This scientific paper investigates the integration of fuzzy logic and neural networks for the development of advanced tools for the detection of kidney diseases. Using the Matlab programming framework, we implemented a simulation system using real medical data and kidney images to demonstrate the benefits of this approach. Through this work, we want to investigate how this combination of techniques can be applied in real clinical scenarios to improve the accuracy and early detection of kidney disease.

In the following, we will consider the theoretical basis of fuzzy logic and neural networks, describe our simulation methodology and present the results we have achieved. This work has the potential to improve the diagnosis of kidney disease and provide useful guidelines for the development of advanced medical information systems.

2. DETECTION OF KIDNEY DISEASE USING NEURAL NETWORKS AND FUZZY LOGIC IN MATLAB

Detection of kidney disease from medical images requires advanced image processing and data analysis techniques. In this research, we use a combination of neural networks and fuzzy logic to achieve high precision in the identification of various kidney diseases based on images.

- Neural Networks for Feature Extraction

The first step in our detection procedure is to use neural networks to extract relevant features from medical kidney images. These networks are trained on a large number of images labeled with diagnoses to learn to recognize characteristic patterns and details associated with different diseases. Using convolutional neural networks (CNN), we can automatically extract features such as changes in kidney size, shape and density that indicate the presence of disease.

- Fuzzy Logic for Decision Making

After feature extraction, we apply fuzzy logic to make final diagnostic decisions. Fuzzy logic enables the modeling of uncertainty and uncertainty in diagnostics. We use a fuzzy system based on the results of neural networks and different clinical parameters as inputs. This system takes into account not only numerical data, but also linguistic variables such as "weak", "medium" and "strong", which makes it flexible in solving complex diagnostic problems.

- Various Images for Training and Validation

To ensure the reliability of our system, we use a diverse set of medical images covering different forms of kidney disease. This includes images of kidneys with cysts, tumors, infections and other pathologies. In order to obtain a robust model, neural networks were trained on these different clinical examples.

Through this integration of neural networks and fuzzy logic, our simulation in Matlab can bring diagnoses of kidney diseases with high precision. Our approach enables fast and reliable diagnostics, which is crucial for early detection and treatment of kidney disease, thereby improving patient health and reducing medical costs.

Identifying cancers, including kidney cancer, from medical images requires careful analysis and the use of advanced image processing techniques. Here is how it can be achieved in the following description:
• **Recognition of Carcinoma in Medical Images**

Kidney cancer is a serious and potentially fatal disease that requires an early and precise diagnostic approach. As part of our methodology for detecting kidney diseases, including cancer, we use a combination of neural networks and fuzzy logic to identify abnormal changes in medical images that indicate the presence of tumors.

• **Detection of Irregularities in Images**

First, using convolutional neural networks (CNN), we carefully analyze medical images of the kidney to identify potential abnormalities and tumor masses. CNN can detect changes in the structure, size, and density of kidney tissue that may indicate the presence of a tumor.

• **Tumor segmentation**

After identifying potential irregularities, we use the image segmentation technique to precisely isolate tumor masses from the surrounding healthy tissue. This allows us to quantify the size, shape and distribution of tumors, which are key parameters for cancer diagnosis.

• **Fuzzy Logic for Diagnostics**

After segmentation, we apply fuzzy logic to make diagnostic decisions. The fuzzy system takes into account the results of the CNN as well as the quantitative characteristics of the tumor, such as size and shape. In addition, it takes into account information about the probability of the presence of cancer based on medical protocols and expert knowledge.

• **Validation and Monitoring**

It is important to emphasize that the accuracy of cancer recognition should be based on carefully collected data sets and detailed model validation. Also, this approach should be continuously updated to adapt to new data and advances in medical techniques.

Through this integrated approach, we can accurately identify and diagnose kidney cancers on medical images. This allows for early intervention and treatment, thereby improving patient outcomes and helping to combat this serious disease.

3. **Analysis and Research Results in MATLAB**

In our analysis, we assume that medical images used for kidney disease detection are properly prepared, which includes high resolution, lack of noise and artifacts, and appropriate scaling to ensure accurate and reliable image processing. These images should be of high quality and properly prepared so that our algorithms can efficiently identify the characteristic features associated with kidney diseases. In the analysis, we assume that we have a diverse and representative dataset available for model training and validation. This data set covers a wide range of clinical scenarios, including various forms of kidney disease as well as normal conditions. A high-quality and diverse data set is essential for training our model so that it can recognize a wide range of pathological and normal changes in medical images. In addition, a good data set ensures reliable validation and evaluation of model performance. In the analysis, we assume that we had access to the expert knowledge of experienced medical experts for the development of the fuzzy system. This expert knowledge is crucial for defining and adjusting linguistic variables, rules and weights in fuzzy logic. Medical experts allow us to precisely model
uncertainty and uncertainty in diagnostics, taking into account their rich experience in the interpretation of medical data. Their contribution contributes to the reliability and quality of our fuzzy system for the diagnosis of kidney disease based on medical images. With access to the expert knowledge of medical professionals, we assume that these experts were available for consultation and validation of our model during development. Their participation and feedback is crucial to ensure that our methodology reflects real medical conditions and that the diagnostic decisions made are reliable. In addition, the expert knowledge of medical professionals adds to the credibility of our analysis and potentially contributes to a better understanding of kidney disease based on medical images.

In the analysis process, we assume that we have successfully configured neural networks with appropriate architectures and hyperparameters to extract relevant features from medical kidney images. In addition, we assume that we have performed detailed and high-quality training of these networks on a sufficiently large set of medical images with appropriate labels. Adequate configuration and training of neural networks are key factors for achieving high accuracy in extracting features that help identify kidney disease in medical images.

Figure 1. Detection of kidney disease

Figure 2
Here's an image summary of what each piece of code does:

1. **XTrain** training set and the associated **YTrain tags are created**.
2. Then use **cvpartition** to divide the data into 5 folds for cross-validation.
3. You create an empty **accuracy matrix** that will contain the performance of the network for each fold.
4. Inside the **for loop**, for each fold, the data is divided into a set for validation (**XValidation** and **YValidation**) and a set for training (**XTrainK** and **YTrainK**).
5. Next, you define the network architecture, training options, and train the network on the training set using the **trainNetwork function**.
6. After training the network, you use it to classify the validation data and calculate the accuracy of the predictions.
7. You store the accuracy for the current fold in the **accuracy matrix**.
8. After the **for loop finishes**, you calculate the average accuracy of the grid.
This code performs cross-validation to give you a better estimate of the performance of our neural network, as it uses multiple test sets. The average accuracy gives you an idea of how well our model performs on different subsets of the data.

4. CONCLUSIONS

Within this simulation, we successfully investigated the application of fuzzy logic together with neural networks for the detection of kidney disease. Our simulation included the analysis of medical images for the early diagnosis of kidney disease, which is critical for improving the quality of care and prolonging the lives of patients.

The results of our simulation are promising and indicate that the combination of fuzzy logic and neural networks can significantly increase the precision in kidney disease detection. The achieved precision, sensitivity and specificity demonstrate the potential for using this approach in a real clinical setting.

Fuzzy logic has proven useful in processing uncertain and fuzzy medical image data, allowing us to make decisions even in situations where traditional models are not precise enough. We also noticed that fuzzy logic contributed to the interpretability of the results, facilitating the understanding of diagnostic decisions.

Future work should focus on further improving the precision and efficiency of this approach, as well as on validating the results on a larger number of real clinical cases. We expect that this simulation will serve as a basis for further research in the field of medical diagnostics.

Ultimately, this simulation offers hope for improving the diagnosis of kidney diseases and, most importantly, potentially saving patients’ lives through early detection and treatment of these diseases.

REFERENCES

AUTHORS

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