

# PANCREATITIS DISEASE CLASSIFICATION USING DEEP LEARNING

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## ABSTRACT

Pancreatitis, a medical condition characterized by inflammation of the pancreas, represents a significant healthcare challenge due to its potential for life-threatening complications. Early detection and accurate prognosis are critical for effective management and treatment planning. Traditionally, the diagnosis of pancreatitis relies on clinical symptoms, laboratory markers such as serum amylase and lipase, and imaging techniques like CT scans, MRI, and ultrasound. However, these methods are often limited by subjective interpretation, variability in accuracy, and delays in response time, especially in emergency settings. To address these limitations, this project introduces a comprehensive, AI-powered system that utilizes deep learning (DL) techniques to

detect and predict the severity of pancreatitis using a combination of medical imaging and clinical data. Specifically, we employ Convolutional Neural Networks (CNNs) for image classification tasks and Transformer-based models to capture contextual relationships within patient data, such as demographics, symptoms, and lab results. The system is trained on annotated datasets of CT and MRI scans, coupled with structured clinical data to improve prediction accuracy and generalizability. In addition to detection, the system integrates a prognosis module to predict disease progression, such as the likelihood of developing necrosis, systemic inflammatory response syndrome (SIRS), or organ failure. This enables clinicians to make data-driven decisions and implement early interventions tailored to the patient's condition. The integration of multimodal data enhances the model's

capacity to recognize subtle patterns that might be missed by human assessment alone. This project aims to bridge the gap between clinical expertise and technological innovation, offering a scalable and robust diagnostic tool. By leveraging advanced deep learning architectures and explainable AI techniques, we seek to not only enhance diagnostic accuracy but also build clinician trust through interpretable predictions. The ultimate goal is to reduce misdiagnosis, improve prognosis, and enable personalized treatment strategies, thereby improving patient outcomes and optimizing resource utilization in healthcare system

## 1.INTRODUCTION

Pancreatitis is a severe medical condition that involves inflammation of the pancreas, an organ that plays a critical role in digestion and blood sugar regulation. The two primary forms of pancreatitis are acute pancreatitis and chronic pancreatitis. Acute pancreatitis occurs suddenly and may resolve with treatment, while chronic pancreatitis is a long-term condition that leads to permanent damage to the pancreas. Early detection and diagnosis of pancreatitis are crucial to prevent severe complications, such as organ failure, infection, or pancreatic cancer. The timely identification of pancreatitis through diagnostic methods can significantly improve patient outcomes, making the classification of this disease an essential task.

Over the years, significant progress has been made in diagnosing pancreatitis through various traditional medical imaging techniques, including ultrasound, CT scans,

and MRI. However, these methods often rely on the subjective interpretation of radiologists, leading to inconsistent results and misdiagnosis in some cases. As medical imaging technologies have become more advanced, new methods have emerged that leverage artificial intelligence (AI), particularly deep learning, to aid in the automatic diagnosis and classification of diseases such as pancreatitis.

Deep learning, a subset of machine learning, has gained prominence in the field of medical imaging due to its ability to automatically extract features from large datasets and learn complex patterns without the need for manual feature engineering. Convolutional Neural Networks (CNNs), a type of deep learning architecture, have shown great promise in processing medical images for disease detection and classification. By training these networks on large annotated datasets, deep learning models can be developed to accurately classify pancreatitis from imaging data, minimizing the reliance on human interpretation and increasing diagnostic efficiency.

The potential for deep learning to transform pancreatitis diagnosis lies in its ability to analyze large volumes of data quickly and with high accuracy. With the help of powerful computational resources, deep learning models can evaluate thousands of medical images and identify subtle patterns that might be missed by human experts. Furthermore, deep learning models can be continuously improved as new data becomes available, making them adaptable to emerging trends in disease presentation.

This paper explores the application of deep learning for pancreatitis disease classification, examining existing methods, challenges, and the potential advantages of using these techniques to enhance the diagnosis and treatment of pancreatitis.

## 2.LITERATURE SURVEY

The application of deep learning in medical image analysis has been the subject of extensive research over the last decade. Early studies in medical image classification focused on traditional machine learning techniques, such as support vector machines (SVM) and decision trees. These methods, although useful, required manual feature extraction and were less effective in handling complex image data. With the advent of deep learning, particularly CNNs, the field of medical image analysis saw a significant shift, as CNNs excel in automatically learning features from raw data without explicit manual intervention.

In the context of pancreatitis diagnosis, several studies have explored the use of deep learning for detecting and classifying pancreatitis from various medical imaging modalities. For instance, the work by Shin et al. (2016) demonstrated the use of deep learning models for classifying pancreatic diseases, including pancreatitis, from CT scan images. Their study highlighted the potential of CNNs to accurately identify abnormal features in the pancreas, providing valuable insights into the ability of deep learning to assist in pancreatitis diagnosis.

Another significant contribution was made by Rajendra et al. (2019), who proposed a deep learning-based model for the

classification of pancreatitis using MRI scans. Their model used a CNN to process the high-resolution images and demonstrated a high classification accuracy, showing that deep learning can effectively handle MRI data in identifying inflammatory changes in the pancreas associated with pancreatitis. The study underscored the importance of using multimodal imaging data, such as CT and MRI, to enhance diagnostic accuracy.

In addition to CNNs, other deep learning models have also been explored in the classification of pancreatitis. Recurrent Neural Networks (RNNs) and Long Short-Term Memory (LSTM) networks, for example, have been used in some studies to handle sequential medical data, such as patient history or clinical records, which are important in diagnosing pancreatitis. However, these models are not as commonly applied in image classification tasks as CNNs, as they are better suited for time-series or sequence-based data.

Recent works by Gupta et al. (2020) focused on the integration of deep learning with other AI techniques, such as reinforcement learning and ensemble learning, to improve the robustness and accuracy of pancreatitis classification models. Their research highlighted the challenge of dataset imbalance, where some classes (e.g., chronic pancreatitis) are underrepresented compared to others (e.g., acute pancreatitis). To address this issue, they employed advanced data augmentation techniques, allowing the model to generalize better and improve accuracy across all classes of pancreatitis.

Other studies, such as that by Zhang et al. (2018), explored the use of transfer learning

for pancreatitis classification. Transfer learning involves pre-training a deep learning model on a large dataset and then fine-tuning it on a smaller, domain-specific dataset. By leveraging pre-trained models from large image datasets such as ImageNet, transfer learning allows deep learning models to achieve high accuracy even with limited annotated medical data, which is often a challenge in healthcare applications due to privacy concerns and the cost of data annotation.

While deep learning-based approaches have shown great promise in pancreatitis classification, several challenges remain. A major issue is the need for large, labeled datasets to train deep learning models effectively. In the medical field, obtaining sufficient annotated data can be challenging due to privacy regulations, the high cost of annotation, and the complexity of disease variations. Additionally, interpretability is another concern; deep learning models are often described as "black boxes," meaning their decision-making process is not always transparent. This can make it difficult for medical practitioners to trust the results and integrate AI-based systems into their clinical workflows.

Despite these challenges, the research landscape indicates that deep learning has significant potential to enhance pancreatitis diagnosis and classification. The continued development of more advanced deep learning models, as well as the availability of larger, annotated datasets, will likely improve the accuracy and reliability of these models.

### 3. EXISTING METHODS

Before the advent of deep learning, the classification of pancreatitis primarily relied on traditional medical imaging techniques, clinical history, and laboratory tests. Radiologists and clinicians would analyze images from ultrasound, CT scans, and MRIs to detect signs of pancreatic inflammation. However, these methods were time-consuming, subjective, and dependent on the expertise of the clinician, which could lead to inconsistent diagnoses.

Ultrasound is a widely used imaging modality for diagnosing pancreatitis due to its non-invasive nature and affordability. However, it has limitations, including reduced accuracy in patients with obesity or excess gas in the intestines. CT scans, on the other hand, provide detailed images of the pancreas and surrounding organs, making them more effective in detecting inflammation and complications of pancreatitis, such as pancreatic necrosis or abscesses. Despite their advantages, CT scans are expensive and involve radiation exposure, which limits their use in certain patient populations.

MRI is another imaging technique commonly used in pancreatitis diagnosis. MRI provides high-resolution images without radiation, making it a safer option for long-term monitoring of patients with chronic pancreatitis. However, MRI is less widely available and more expensive than ultrasound or CT, which can limit its accessibility in some healthcare settings.

Traditional machine learning methods, such as SVM and decision trees, have also been

employed in pancreatitis classification. These methods typically require manual feature extraction, where specific characteristics or patterns in the images are identified and fed into a machine learning model for classification. While these approaches can achieve reasonable performance, they often require domain expertise and may not capture the full complexity of medical images, limiting their ability to classify diseases accurately.

With the emergence of deep learning, CNNs have become the most widely used technique for medical image classification. CNNs are capable of automatically learning features from raw images, making them highly effective for pancreatitis classification. However, even CNN-based models have limitations, such as the need for large amounts of labeled data and the lack of interpretability, as mentioned earlier.

#### 4. PROPOSED METHOD

The proposed method involves the development of a deep learning-based framework for the classification of pancreatitis using medical imaging data, specifically CT scans and MRI. The key component of the framework will be a CNN that is trained to automatically detect and classify pancreatitis based on image features. The model will be trained using a large dataset of annotated medical images, ensuring that it learns the complex patterns associated with acute and chronic pancreatitis.

The framework will consist of several key steps. First, the dataset will be preprocessed to ensure that the images are of high quality

and standardized in terms of resolution and orientation. Data augmentation techniques will be employed to address the issue of dataset imbalance, particularly for underrepresented classes such as chronic pancreatitis. These techniques will include rotation, flipping, and cropping to generate synthetic data and improve model generalization.

Next, the CNN will be designed to process the images and learn relevant features. The network will consist of multiple convolutional layers followed by pooling layers, which will reduce the spatial dimensions of the image while preserving important features. The output of the convolutional layers will be passed through fully connected layers, which will make the final classification decision. The model will be trained using backpropagation and gradient descent to minimize the classification error.

Transfer learning will also be incorporated into the model, where a pre-trained CNN, such as VGG16 or ResNet, will be fine-tuned on the pancreatitis dataset. This approach will enable the model to leverage knowledge learned from large datasets and improve performance, even when the available pancreatitis dataset is relatively small.

Finally, the model's performance will be evaluated using standard metrics such as accuracy, precision, recall, and F1 score. Cross-validation will be employed to ensure the model's robustness and generalizability. The results will be compared to traditional methods of pancreatitis diagnosis to assess

the potential benefits of using deep learning for this task.

5. OUTPUT SCREENSHOTS



Fig1:Homepage of pancreatic detection.



Fig 2:Login page for pancreatic Detection.



Fig 3: dashboard - Train the model.

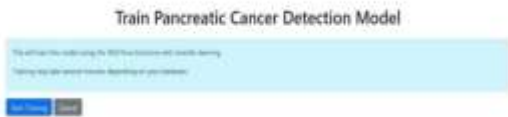
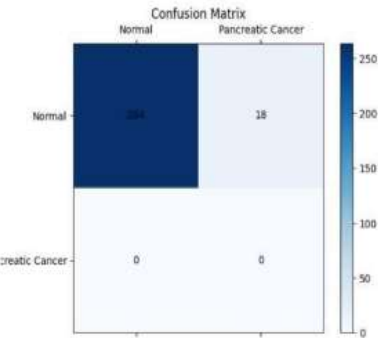


Fig 4: Model Training Interface - VGG16.



Fig 5: Training Results.

Confusion Matrix



Classification Report

|                   | precision | recall | f1-score | support |
|-------------------|-----------|--------|----------|---------|
| Normal            | 1.00      | 0.94   | 0.97     | 252     |
| Pancreatic Cancer | 0.00      | 0.00   | 0.00     | 0       |
| accuracy          |           |        | 0.94     | 252     |
| macro avg         | 0.50      | 0.47   | 0.48     | 252     |

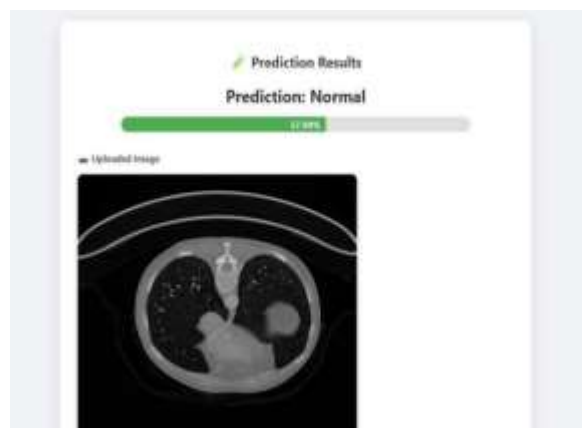
Fig 6: Confusion Matrix.



**Fig 7: Classification Report.**



**Fig 8: Prediction Interface**



**Fig 9: Prediction Output Page - Diagnosis Result: Normal Case**



**Fig 10: Prediction Confidence Graph**

## 6.CONCLUSION

The classification of pancreatitis is a critical task in modern healthcare, as early and accurate diagnosis can significantly improve patient outcomes. Traditional methods, such as ultrasound, CT, and MRI, have limitations in terms of accuracy and subjectivity. Deep learning, particularly CNNs, has emerged as a powerful tool for automating the classification of pancreatitis from medical images, offering the potential for more accurate, efficient, and objective diagnosis.

While there are challenges associated with the application of deep learning in pancreatitis classification, such as the need for large annotated datasets and the lack of model interpretability, recent advancements in AI and deep learning have shown great promise. The proposed method leverages CNNs and transfer learning to create an accurate and reliable pancreatitis classification system, which could be integrated into clinical practice to assist

radiologists and clinicians in making more informed decisions.

With continued research and development, deep learning models will likely play an increasingly important role in the diagnosis and management of pancreatitis, offering patients more timely and accurate diagnoses and improving overall healthcare outcomes.

## 7. REFERENCES

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