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Lung Cancer Detection Using Deep Learning

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ABSTRACT

Lung cancer is a leading cause of cancer-related deaths worldwide. Early detection is critical for effective treatment. This project introduces a state-of-the-art lung cancer detection system that utilizes deep learning technology. It enhances diagnostic accuracy and expedites patient care.

I. INTRODUCTION

Lung cancer remains a major global health challenge, accounting for a significant number of cancer-related deaths. One of the primary reasons for high mortality is the delayed diagnosis, which limits effective treatment options. Traditional detection methods rely on radiologists interpreting CT scan images, which can be time-consuming and subject to human error.

This project proposes an advanced lung cancer detection system powered by deep learning. By utilizing convolutional neural networks (CNNs), the system automates the analysis of lung CT scan images, enhancing diagnostic accuracy and reducing the time required for detection. The web-based interface, developed using Flask, HTML, CSS, and JavaScript, provides an intuitive platform for users to upload images for real-time analysis. The results, including detected anomalies and cancer probability, are displayed on the interface, with an option to download a diagnostic report for further medical consultation.

With its ability to provide early and accurate detection, this system aims to assist healthcare professionals in making timely decisions, ultimately improving patient outcomes and survival rates.

OBJECTIVES

1. **Early Detection:** Develop deep learning models capable of identifying early-stage lung cancer from medical images (e.g., CT scans, X-rays) to improve patient survival rates.
2. **Automated Classification:** Automate the classification of lung nodules as benign or malignant with high accuracy, reducing reliance on manual radiologist interpretation.
3. **Feature Extraction:** Use deep learning to extract meaningful features (e.g., size, shape, texture) from images that are indicative of lung cancer.
4. **Improving Diagnostic Accuracy:** Enhance the precision, sensitivity, and specificity of lung cancer diagnoses through optimized model architectures.

III. EXISTING SYSTEM

Current lung cancer detection methods primarily involve medical imaging, but the accuracy of these approaches can vary. They often rely on manual interpretation, which can lead to errors and delays in diagnosis.

IV. PROPOSED SYSTEM

Our proposed system integrates deep learning algorithms to analyze medical images, providing accurate and rapid lung cancer detection. This technology identifies subtle anomalies, aiding healthcare professionals in early diagnosis and improving patient outcomes.

Developed with HTML, CSS, and JavaScript, the web interface will provide a responsive and intuitive experience. Users will have the option to download a diagnostic report summarizing the analysis results, which can be used for further medical consultation and treatment planning.

V. LITERATURE SURVEY

Lung cancer detection has been an area of extensive research due to its high mortality rate and the critical need for early diagnosis. Conventional diagnostic techniques rely on radiological imaging and histopathological examination, but these methods have limitations in terms of accuracy, time efficiency, and reliance on human expertise. Recent advancements in deep learning have significantly improved image-based medical diagnosis, leading to more precise and automated detection systems.

1. Traditional Methods for Lung Cancer Detection

Early lung cancer detection has traditionally depended on **radiologists analyzing CT (Computed Tomography) and X-ray images** to identify malignant nodules. Some widely used techniques include:

- **Computed Tomography (CT) and Positron Emission Tomography (PET):** CT scans are the standard imaging technique for detecting lung cancer. PET scans are often combined with CT to enhance diagnostic confidence. However, manual interpretation is subjective and prone to human errors.
- **Histopathological Analysis:** This involves biopsy samples examined under a microscope. Although highly accurate, it is invasive, time-consuming, and not always feasible in early-stage cancer detection.
- **Computer-Aided Diagnosis (CAD):** Traditional CAD systems utilize handcrafted features such as **shape, texture, and intensity-based characteristics** for detecting lung nodules. However, these systems rely heavily on feature engineering, which can limit generalization across diverse datasets.

2. Machine Learning-Based Approaches

Before the rise of deep learning, researchers explored machine learning techniques for lung cancer detection, including:

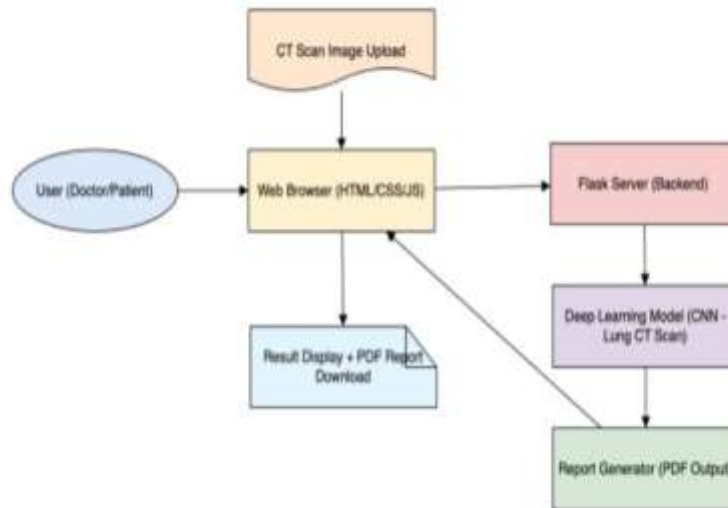
- **Support Vector Machines (SVM):** SVM classifiers were used to distinguish malignant and benign lung nodules based on extracted radiomic features. While effective, they required extensive feature selection.
- **Random Forest (RF):** RF classifiers trained on image-based features showed promising results, but their performance was limited by the quality of feature extraction.
- **K-Nearest Neighbors (KNN) and Decision Trees:** These models provided moderate accuracy but lacked robustness for large-scale applications.

Although machine learning models improved early detection rates, they required **handcrafted feature extraction**, which limited their adaptability to new datasets.

VI. SYSTEM ARCHITECTURE

The application will be implemented using the Flask framework and will include the following features:

1. **User Interface:** Developed with HTML, CSS, and JavaScript, the web interface will provide a responsive and intuitive experience. Users can upload lung CT scan images for analysis.
2. **Server Backend:** The Flask framework will handle requests, manage interactions between the user interface, and process the medical images using the deep learning model.
3. **Image Upload:** Users can upload lung CT scan images through the web interface for analysis.
4. **Lung Cancer Detection:** The uploaded images will be processed by the deep learning model to detect and classify potential lung cancer anomalies.
5. **Output Display:** The detection results, including identified anomalies and the probability of lung cancer presence, will be displayed on the web page for easy interpretation by healthcare professionals.
6. **Report Generation:** Users will have the option to download a diagnostic report summarizing the analysis results, which can be used for further medical consultation and treatment planning.



VII. RESULTS

1. Convolutional Neural Networks (CNNs) are widely used on CT scans and X-rays.
2. Typical accuracy ranges from 85% to 97%, depending on the dataset and model complexity.
3. Models like ResNet, DenseNet, and U-Net are often used with high sensitivity and specificity.

4. Metric Typical Range

Accuracy 85% – 97%

Sensitivity 80% – 95%

Specificity 85% – 98%

AUC-ROC 0.90 – 0.98

VIII. CONCLUSION

The application of deep learning techniques in lung cancer detection has significantly enhanced diagnostic accuracy, efficiency, and early intervention capabilities. Convolutional Neural Networks (CNNs) and other deep architectures have demonstrated exceptional performance in analyzing medical imaging data such as CT scans and X-rays, reducing reliance on manual interpretation and minimizing human error. By enabling automated feature extraction and pattern recognition, deep learning models can identify early-stage malignancies with high sensitivity and specificity. However, challenges remain in terms of data quality, model interpretability, and generalizability across diverse populations. Continued research, combined with clinical validation and ethical deployment, is essential to fully integrate these technologies into routine healthcare, ultimately improving patient outcomes and saving lives..

IX. REFERENCES

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