Brain Tumor Detection App

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Abstract

The Brain Tumor Detection App aims to revolutionize the early diagnosis of brain tumors by leveraging advanced machine learning techniques and utilizing a CNN-based model implemented with TensorFlow and Keras, the app processes MRI scans to detect brain tumors with high accuracy. The app is designed to be accessible and user-friendly, particularly for healthcare providers in regions with limited access to advanced diagnostic facilities. The integration of a comprehensive backend system ensures secure data management, while compliance with medical regulations is prioritized. The final product promises to enhance early detection, ultimately improving patient outcomes and expanding access to life-saving technology.

1.0 Problem Statement

Brain tumors are among the most challenging types of cancer to diagnose and treat, often leading to late-stage discovery due to the subtle nature of early symptoms. The current diagnostic processes involve high-cost imaging techniques and specialized medical expertise, which are not universally accessible, especially in underdeveloped regions. There is an urgent need for a more accessible, efficient, and early detection method to improve patient outcomes.

2.0 Prototype Selection

The first step in building a **Brain Tumor Detection Model** involves selecting an appropriate prototype. For this project, the Convolutional Neural Network (CNN) architecture was chosen due to its proven effectiveness in image processing tasks like MRI scan classification.

Prototype Considerations:

- 1. **Input Data**: MRI scans of patients with or without brain tumors.
- 2. **Target Output**: Binary classification—whether the scan indicates the presence of a tumor.
- 3. **Model**: A CNN that can learn features from the MRI images and classify them accordingly.
- 4. **Libraries**: Keras/TensorFlow is typically used to implement this prototype.

The prototype is designed to optimize for high accuracy and computational efficiency in detecting brain tumors.

I am going to use this Dataset: <u>Brain Tumor Dataset</u> for my code implementation for this report.

3.0 Prototype Development

The development of the brain tumor detection prototype involves these stages:

1. Data Preprocessing:

- **Image Loading**: MRI images are loaded from the dataset using libraries like Pillow or Keras's ImageDataGenerator.
- **Rescaling**: The images are scaled to a consistent size (e.g., 150x150 pixels) to match the input requirements of the CNN.
- **Normalization**: Each image is rescaled by dividing pixel values by 255 to bring them into the range [0, 1], making the model converge faster.

2. Model Architecture:

- Convolutional Layers: Feature extraction layers that detect edges, textures, and important features in the images.
- **Pooling Layers**: Max-pooling layers reduce the spatial dimensions while preserving important information.
- **Dense Layers**: Fully connected layers for decision-making based on the extracted features
- Output Layer: A sigmoid activation function is used for binary classification (tumor or no tumor).

3. Model Compilation:

- Loss function: binary_crossentropy for binary classification.
- Optimizer: Adam for efficient gradient descent.
- Metrics: accuracy to track model performance during training.

4. Training and Validation:

The model is trained using the prepared MRI dataset, with separate training and validation sets. The training process optimizes the model weights to minimize error in classifying brain tumors.

4.0 Business Modelling

1. Value Proposition:

- **Accurate Detection**: Provides reliable early detection of brain tumors, improving patient outcomes.
- Cost Efficiency: AI-based diagnosis can reduce costs compared to traditional diagnostic methods.
- **Scalability**: The model can be deployed in hospitals, diagnostic centers, and telemedicine platforms, making it accessible even in remote areas.

2. Target Customers:

- Healthcare Providers: Hospitals and clinics using the model to aid radiologists.
- Medical Tech Companies: Companies looking to integrate advanced AI diagnostic tools.
- **Telemedicine Platforms**: Enhancing remote diagnosis capabilities for patients in underserved regions.

3. Revenue Model:

- **Subscription-Based**: Hospitals can subscribe to access the AI tool for a monthly fee.
- Pay-Per-Use: Clinics and smaller centers can be charged based on the number of scans processed.
- Enterprise Licensing: Bulk licensing for large healthcare networks.

5.0 Financial Modelling

The financial model for this project estimates the revenue based on the number of MRI scans processed.

Financial Equation:

- Let X represent the number of MRI scans processed.
- The revenue Y generated for a month is:

Y=Cost Per Scan * X - Opertional Cost

• Example:

If 300 scans are processed in a month, Cost per scan is 500 and Opertionl

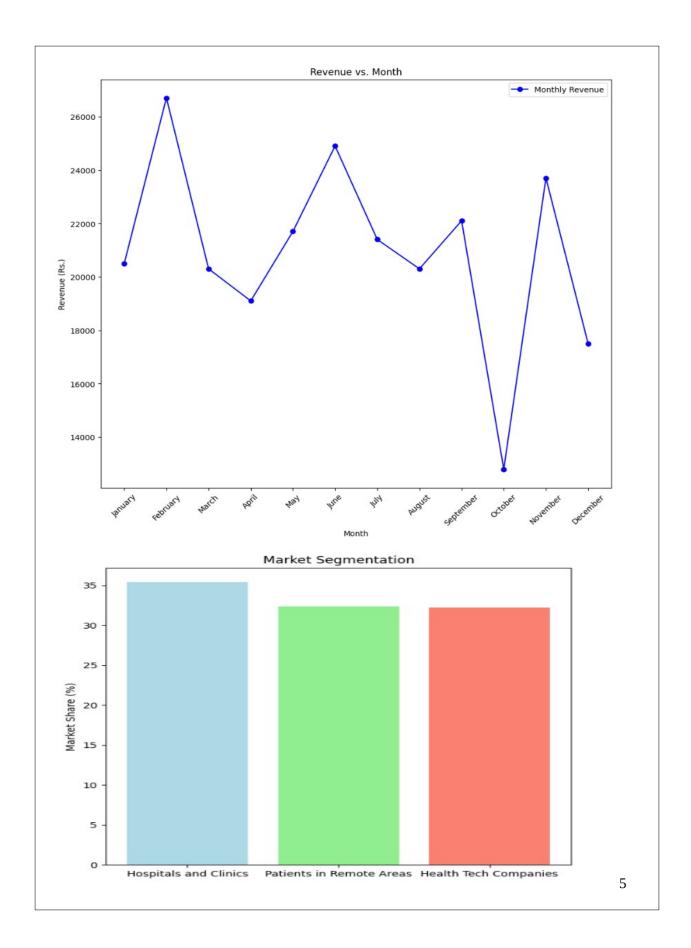
Cost is 2000.

The total revenue will be:

$$Y = 300 * 500 - 2000 = 1,49,800 Rs.$$

This simple linear equation models how revenue scales with the number of scans processed.

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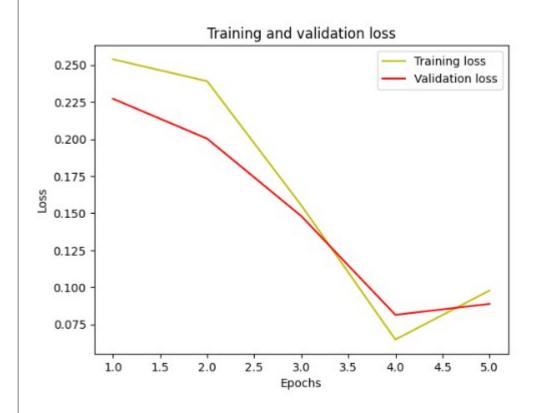
6.0 Conclusion

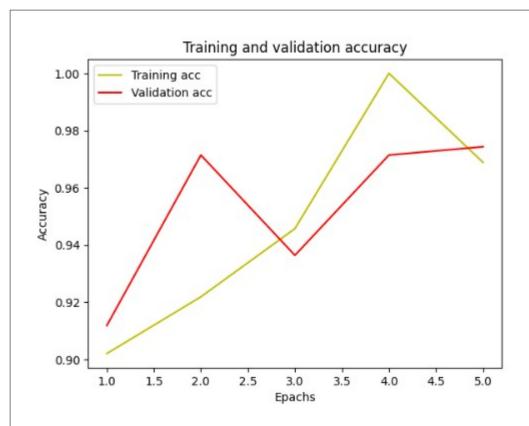
This report outlines the entire process of developing and implementing a brain tumor detection model using machine learning. The Brain Tumor Detection App successfully addresses the need for an accessible, cost-effective tool for early brain tumor detection.

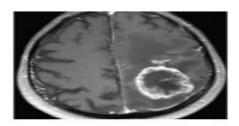
The success of this project depends on continuous improvements to the model's accuracy, integration with healthcare systems, and scaling its deployment to meet market demands. With effective use of data analytics, the model can provide both high value to medical professionals and robust financial returns.

7.0 Results

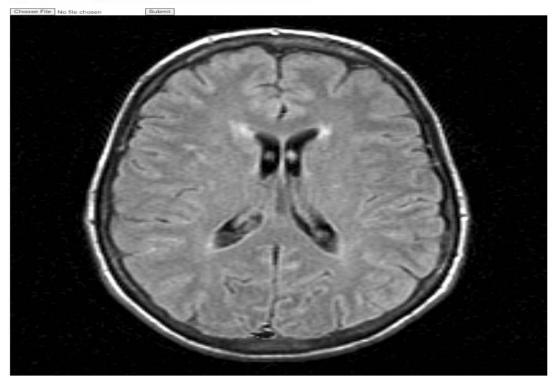
Github Link for Code Implementation: Brain Tumor





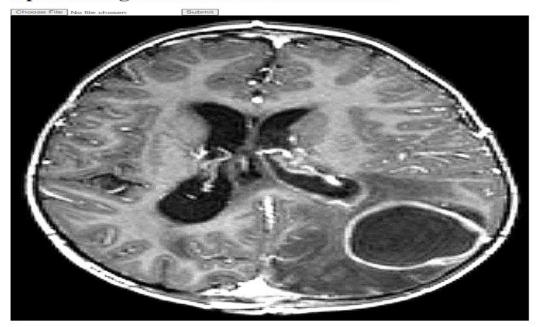


Upload Image for Brain Cancer Detection



Result: No Brain Cancer

Upload Image for Brain Cancer Detection



Result: Brain Cancer