EFFICIENT PNEUMONIA DETECTION IN DIGITAL CHEST XRAY IMAGES USING DEEP TRANSFER LEARNING

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INTRODUCTION

- Pneumonia is one of the largest infectious diseases that cause death in children and elderly people across the globe.
- Pneumonia is ranked eight in the list of the top 10 causes of death in the United States.
- Due to pneumonia, every year, 3.7 lakh children die in India, which constitutes a total of fifty percent of the pneumonia deaths that occur in India.

INTRODUCTION 3/23

EXISTING SYSTEM

- Chest X-rays are primarily used for the diagnosis of this disease.
- However, even for a trained radiologist, it is a challenging task to examine chest X-rays.

• There is a need to improve the diagnosis accuracy.

EXISTING SYSTEM 4/2:

PROPOSED SYSTEM

- To develop an efficient application to detect Pneumonia by uploading digital Chest X-ray .
- Efficient model for the detection of pneumonia
- Trained on digital chest X-ray images
- Transfer learning is used to fine-tune the deep learning models

• Supervised learning approach.

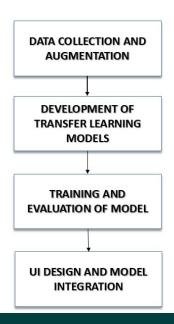
PROPOSED SYSTEM 5/23

PROPOSED SYSTEM

- Several deep learning models are customised and trained in order to get the best detection model.
- Various models used are as follows :
 - DenseNet121
 - MobileNetV3
 - InceptionV3
 - ResNet152V2
 - Sequential
 - VGG16

PROPOSED SYSTEM 6/23

ARCHITECTURE DIAGRAM



WORK FLOW

- Implementation is supposed to have the following steps :
 - Data Collection
 - Data Augmentation
 - Development of transfer learning models (customization)
 - Train models
 - Evaluation
 - Choose the best model
 - UI Design and model Integration
 - Generate predicted results

WORK FLOW 8/23

REQUIREMENT SPECIFICATIONS

- Software Requirements :
 - Operating System : Windows 8 or above
 - Front End : Android , Java
 - Back End: python
 - IDE : Google Colab and Android studio

- Hardware Requirements:
 - Processor : core i3 or above
 - Hard Disk Space: 320 GB
 - Memory: 4 GB or above

- DATA COLLECTION AND AUGMENTATION
- DEVELOPMENT OF TRANSFER LEARNING MODELS
- TRAINING AND EVALUATION OF MODELS
- UI DESIGN AND MODEL INTEGRATION

MODULES 10/23

DATA COLLECTION AND AUGMENTATION

- The Data set is taken from kaggle repository for training.
- Some Data are collected from hospital for testing purpose.
- Split images as 80% for training and 20% for validation.
- Images are grouped into 2 classes normal and pneumonia.
- The training set contains 5889 labeled images
- The test set contains 1749 images
- All the images are resized to [224,224]
- Data augmentation is done to significantly increase the diversity of data available for training models

• ImageDataGenerator class is used for Data Augmentation

MODULES 11/23

DEVELOPMENT OF TRANSFER LEARNING MODELS

- Model for training is built using keras.
- Add preprocessing layer to the front of the model.
- Here we will be using imagenet weights which is large visual database designed for use in visual object recognition
- The layers in the model are set "trainable = false", thus not to train the existing weights.
- The last layer in the model is flattened and then dense using ReLU Activation function
- Rectified Linear Unit activation function is a linear function that will output the input directly if it is positive, otherwise, it will output zero

MODULES 12/23

TRAINING AND EVALUATION OF MODELS

- Model thus created is compiled.
- Adam is used as the optimizer with binary-crossentropy as loss and accuracy as the metrics
- optimizer is a function or an algorithm that modifies each epoch's weights and minimize the loss function.
- Binary cross entropy compares each of the predicted probabilities to actual class output which can be either 0 or 1
- Finally the model is trained with the required number of epochs and is saved.

MODULES 13/23

UI DESIGN AND MODEL INTEGRATION

- The model thus created is converted to tflite_model for integrating to android UI.
- TFLiteConverter in Tensorflow Library is used for conversion
- Upload a chest X-ray image to test the model.
- When the user clicks the predict button, it will return the corresponding class to which the image belongs to.

MODULES 14/23

Pneumonia Detection



UPLOAD

PREDICT

RESULT 15/2:



RESULT 16/23

Pneumonia Detection



UPLOAD

PREDICT

Pneumonia

RESULT 17/2:

Pneumonia Detection



UPLOAD

PREDICT

Normal

RESULT 18/2

EVALUATION MEASURES

- ACCURACY :
 - Computes the count of correct predictions.

- ACCURACY OF VARIOUS TRAINED MODEL ARE AS FOLLOWS:
 - DenseNet121 50%
 - MobileNetV3 49%
 - InceptionV3 50%
 - ResNet152V2 85%
 - Sequential 88%
 - VGG16 90%

EVALUATION MEASURES 19/23

FUTURE ENHANCEMENT

- In the future, this work could be extended to detect and classify X-ray images consisting of lung cancer and pneumonia.
- In the future, this work could be extended to explore more accurate classification architectures to diagnose two types of pneumonia, viruses, and bacteria.
- The future works can involve developing an algorithm which can localize the parts of the lung affected by pneumonia.

FUTURE ENHANCEMENT 20/23

CONCLUSION

- An android application using a deep transfer learning model is built that can detect pneumonia from digital chest X-ray images with an accuracy above 90
- With this model ,it is possible for early diagnosis of Pneumonia, that helps to save the patients from being severe.
- Minimise human intervention in detecting pneumonia.
- The result shows that the proposed method has good performance that helps the trained radiologist for early detection of pneumonia.

CONCLUSION 21/23

REFERENCE



Hashmi, Mohammad Farukh et al.

Efficient Pneumonia Detection in Chest Xray Images Using Deep Transfer Learning

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Appendix 22/23

Thanks

THANK YOU

Appendix 23/2: