

CLASSIFYING DYSLEXIA BASED ON HANDWRITING

ATRAIN

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Children's Dyslexia Center - Boston
North, Diana Baum (Director)

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App

A web application that detects
likeliness of user being dyslexia
based on handwriting

06

Future

Implications & future improvements
for our project





01

Introduction

What Is Dyslexia?

• Solution



Issue

Dyslexia is a general term for disorders that affect your writing, speaking, reading, and spelling skills, impacting a student's ability to read and write.

This is why it is important that it is diagnosed and treated early.



Strategy

Generate and optimize a machine learning model that can classify if an individual is likely to have dyslexia in order to initiate further potential consultations.



Aim

To create an application using our machine learning model to predict if an individual has dyslexia based on their handwriting.





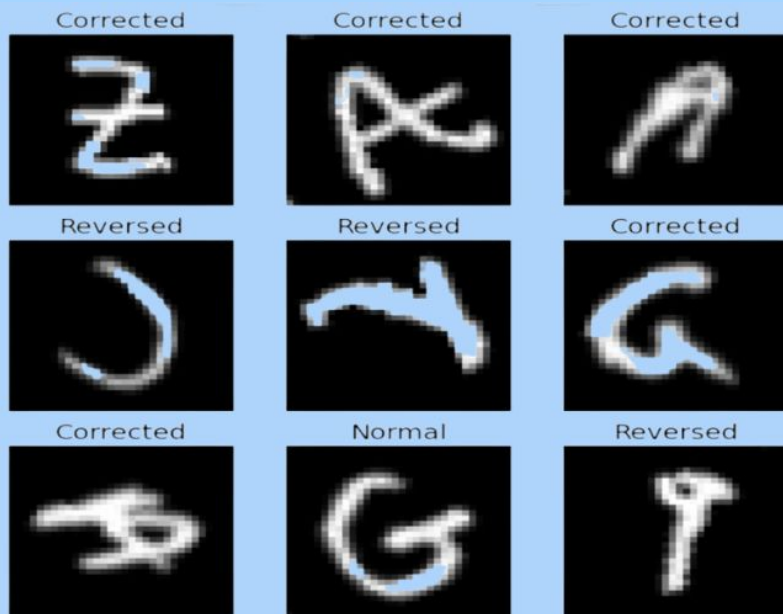
02

Dataset

What are we working with?

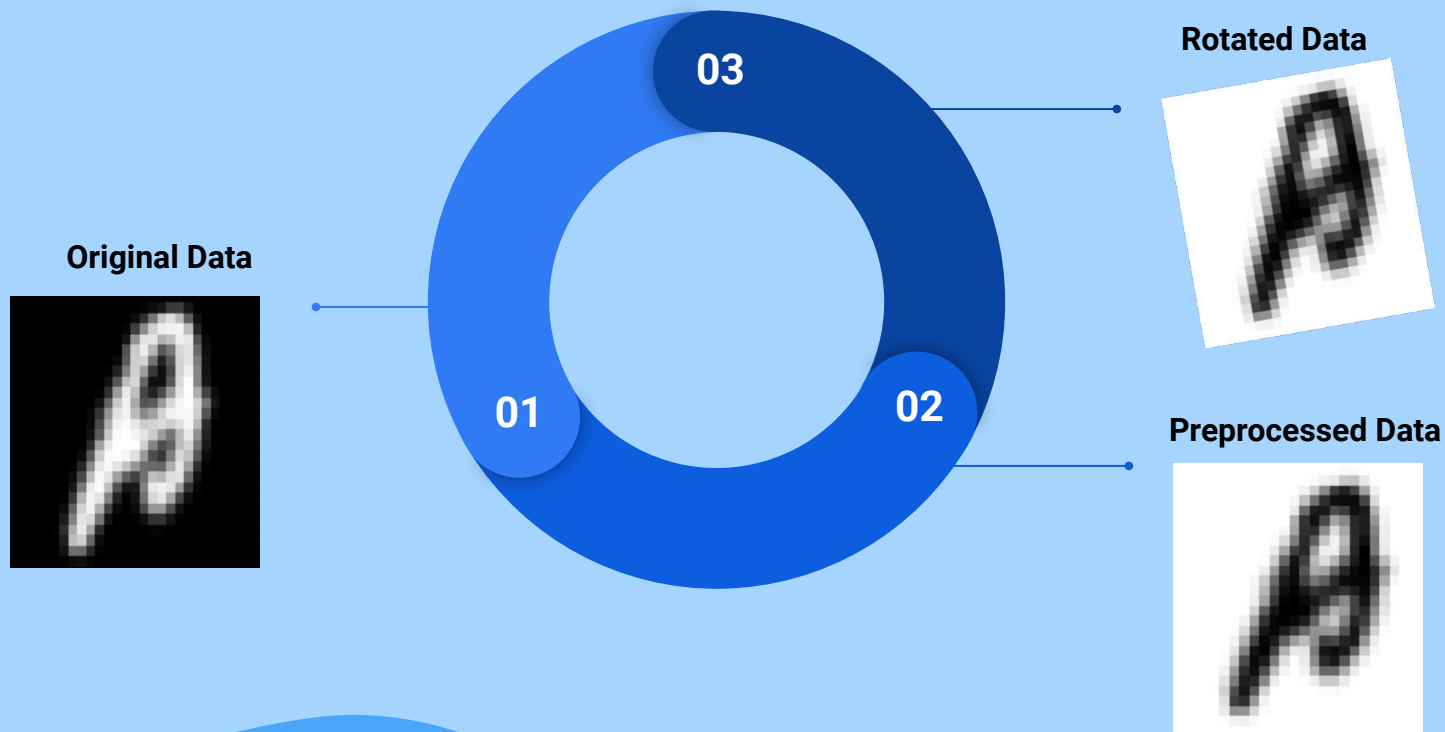
Our Data

<https://www.kaggle.com/datasets/drizasazanitaisa/dyslexia-handwriting-dataset>

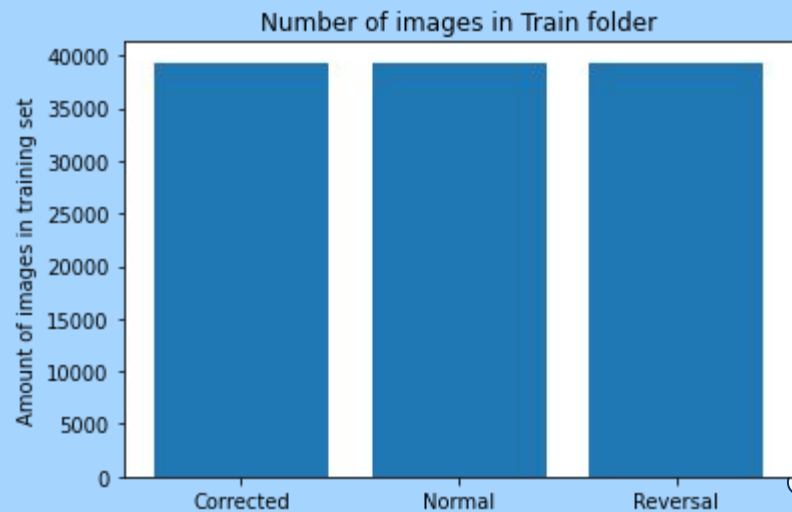
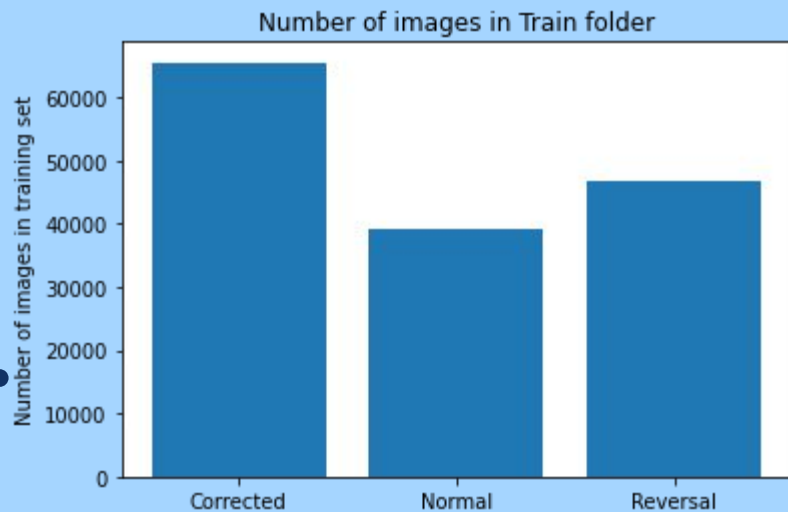


| Classes | |
|------------------|---|
| Corrected | Characterized as messy handwriting |
| Normal | No issues with legibility and orientation of letters |
| Reversal | Letters are in the opposite orientation as normal letters should be and are therefore indicating dyslexia |

Preprocessing the Data



Undersampling The Data





03

ML Models

What tools can we use?

How Are Models Evaluated?

Accuracy

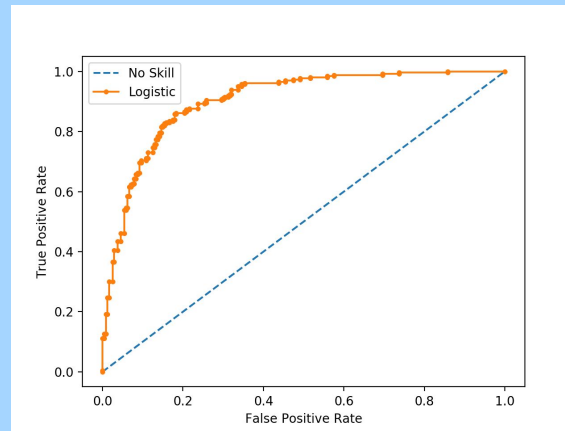
- Percentage of predictions that were correct on our test set.

Confusion Matrix

| | | Predicted Class | |
|------------|---|-----------------|----------------|
| | | 0 | 1 |
| True Class | 0 | True Negative | False Positive |
| | 1 | False Negative | True Positive |

Receiver Operating Characteristic (ROC) Curve

- The greater the area under the curve (AUC), the better the model performs



$$\text{True Positive Rate (TPR)} = \frac{TP}{TP + FN}$$

$$\text{False Positive Rate (FPR)} = \frac{FP}{FP + TN}$$

Models



K-Means

Our first image classifier



Random Forest

Used to predict accuracy and classification



CNN

A type of neural network with convolutional layers



LeNet-5

A type of CNN model for handwritten digit recognition



Inception V3

Uses the pre-trained convolutional layers of a model

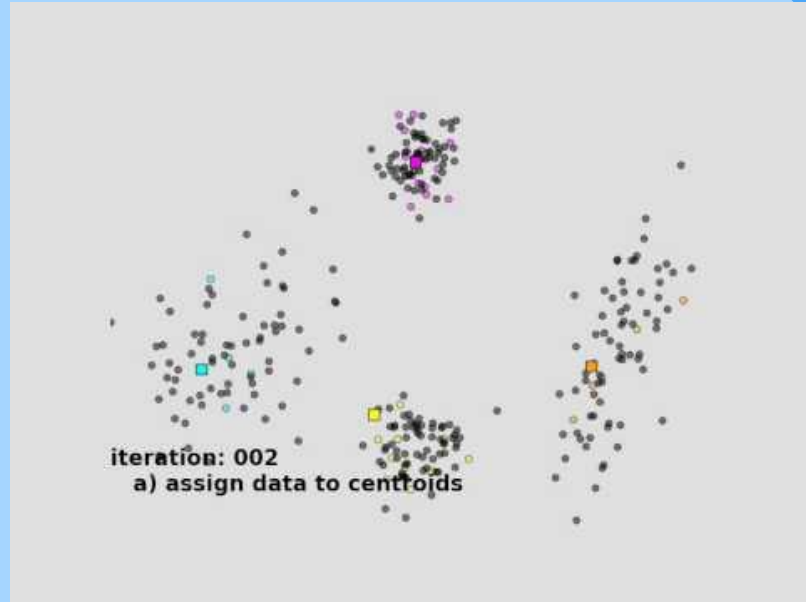
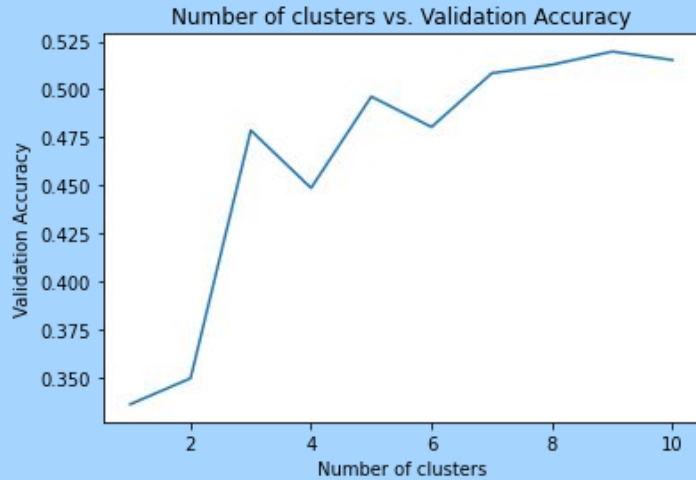


VGG16

Another pre-trained model that performed very well with our dataset

K-Means

Poor accuracy no matter the number of clusters because the images in our dataset aren't easily clusterable.



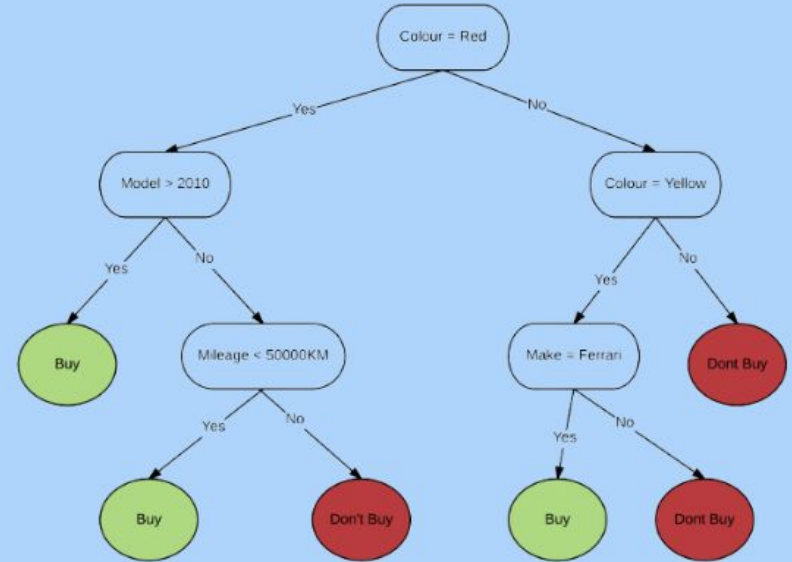


Random Forest

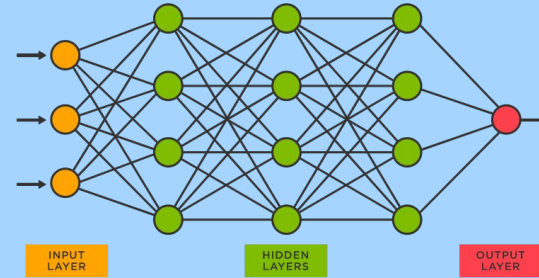
THE CONCEPT

A combination of multiple decision trees that extracts the most important features.

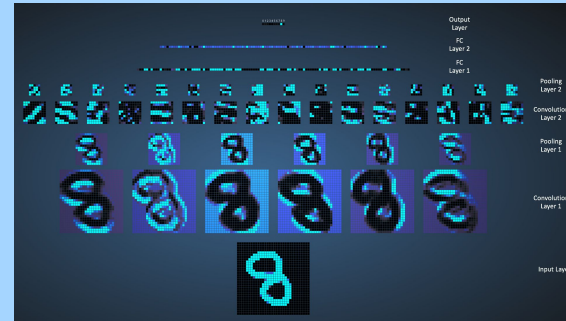
RandomizedSearchCV



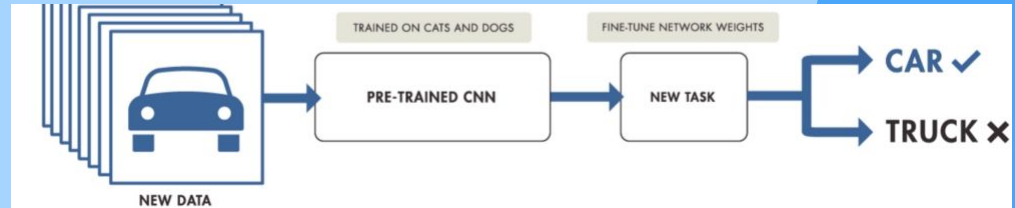
Neural Network



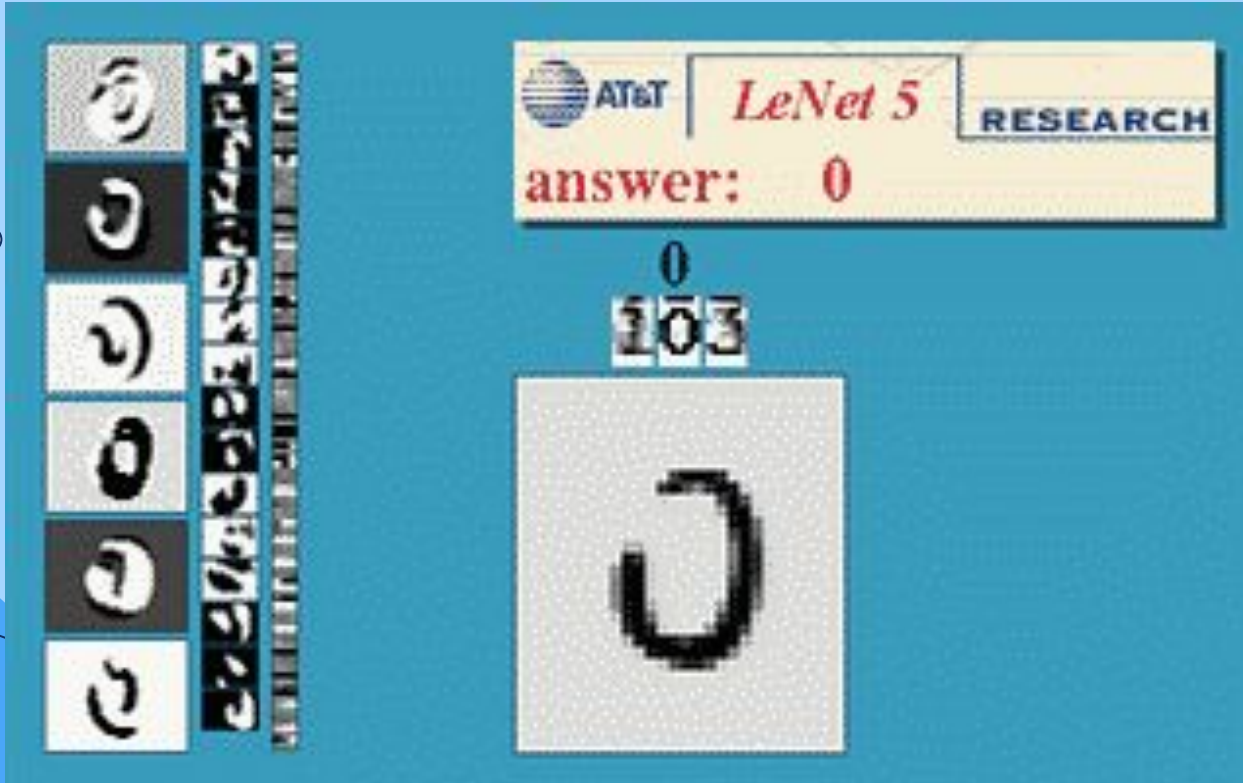
- ## Convolutional Neural Network (CNN)



Transfer Learning



CNN: LeNet-5



- Image size: 32 x 32
- 2 Convolutional Layers
- 2 Average pooling Layers
- 2 Dense Layers
- RMSprop Optimizer
- Learning Rate: 0.001

Transfer Learning: InceptionV3

1

Image Size

75 x 75

2

Layers

1 Global Average Layer, 2 Prediction Layers (One has 64 units and the other had 3 units both has relu activation functions)

3

Loss Function and Optimizer

Categorical_Crossentropy and Adam Optimizer

4

Learning Rate

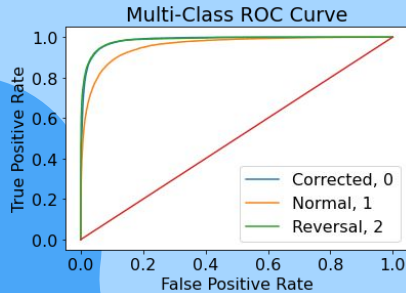
0.0001

The Best Model - VGG16 Model

Avoids Overfitting

The training accuracy (87.8%) and validation accuracy (89.4%) are similar indicating minimum overfitting

ROC Curve

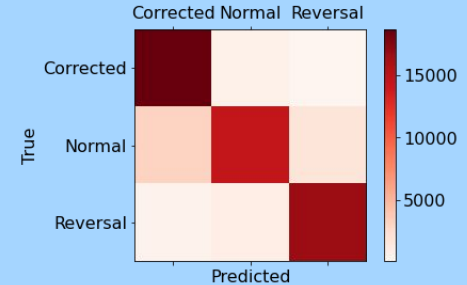


Uses Preprocessed Data

Data utilized is realistic after being preprocessed

Confusion Matrix

Confusion Matrix for Test Data



04

Collaboration



How Collaborating Helped Us?



- Letter reversal can occur in younger individuals without dyslexia. Not all individuals with dyslexia perform letter reversals
- "Although students with dyslexia produced a higher proportion of reversal errors than those without dyslexia in both letter writing and letter naming" Diane Baum, Director of Children's Dyslexia Centers - Boston North.

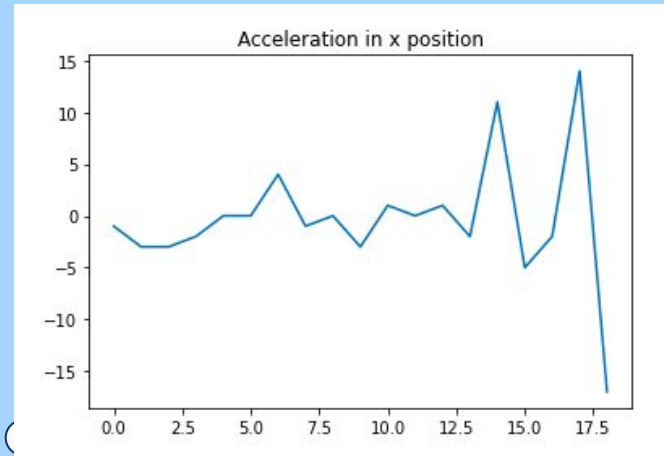
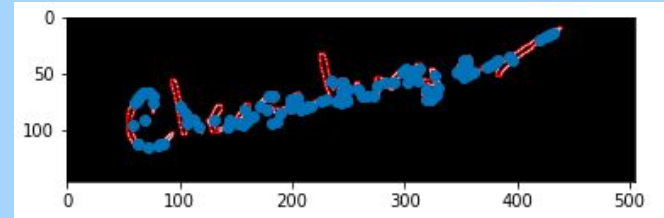


- Adding a disclaimer for user age (the user must be above the age of 7 in order for the results to be more accurate)
- Adding an algorithm to determine if there is an abnormal spacing between letters and words (another indicator of dyslexia).

Abnormal Spacing Algorithm

- Algorithm detects abnormal spacing between letters in a word
 - OpenCV
 - Contour Centers
- Would like to use a machine learning dataset to improve upon this algorithm in the future

Cheeseburger



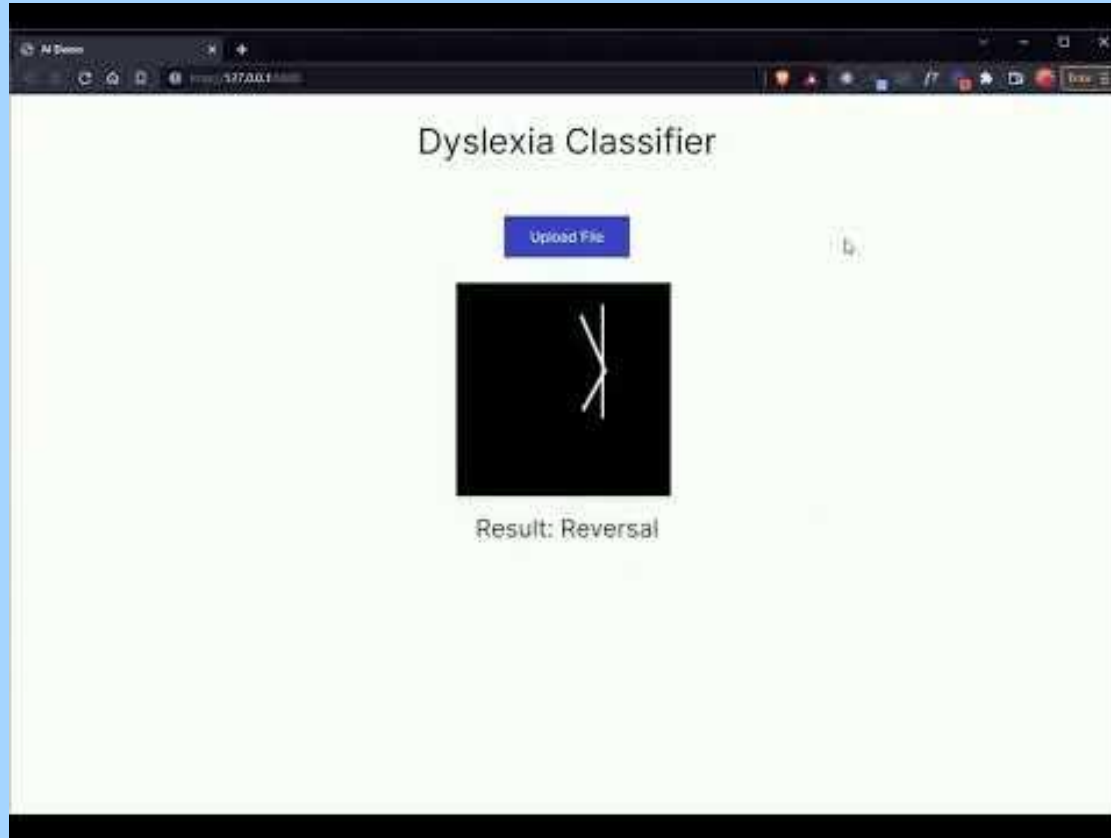


05

Web App

Demo of Our Application
Built with Flask

Web Application Demo Video





06

Future

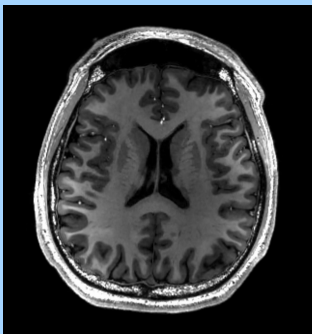
Improvements &
Implications

Further Implications

01

More Features

Including more detections for features in dyslexia handwriting (ex: abnormal mix of upper and lower case)



03

Datasets

Supplemental MRI scans

02

Collaboration

Continue working with and learning from Children's Dyslexia Centers

04

Mobile App

Provide an application with a friendlier and more versatile user interface





Thank You For Listening!

— **Angela, Amishi, Isha, Neta, Ritvik,
& Tyler**



Email TylerWestland@gmail.com regarding any questions!

