# AN IMPROVED APPROACH FOR AUTOMATIC DIAGNOSIS OF ADHD USING FMRIS

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# History of ADHD

- Condition that leads to behavioral changes in children.
- ADHD was first considered a mental disorder in the 1960s.
- The traditional ways to diagnose ADHD is outdated, making it underdiagnosed in many patients.
- Subjective testing for kids leads to underdiagnosing.
- Neuroimaging research in psychiatric disorders is a new field that is starting to gain popularity.
- The ENIGMA organization's mission is to bring together research in neurology and psychiatry to better understand the brain and how it functions.

#### ADHD-200

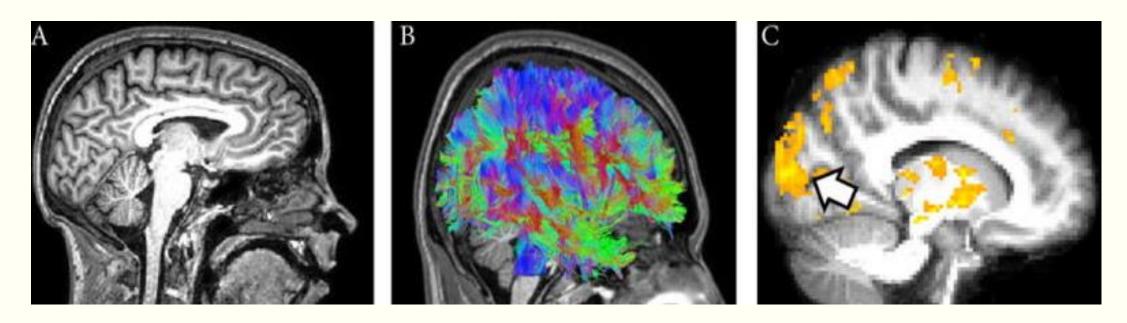
#### ADHD-200 Dataset

- The dataset consists of approximately 800 subjects MRIs provided from 8 research institutions.
- The MRI scans are preprocessed using different methods.
- The most famous and recent one is the Athena pipeline.

#### ADHD-200 Competition

- The ADHD-200 global competition started in 2011.
- Single and group entries are both allowed.
- Different methods are applied in aim to reach the best accuracy to better diagnose the subjects.

#### Difference between functional and structural MRI



- Functional MRI shows the areas of the brain that are active when a patient is asked to perform a certain task. (C)
- Structural MRI has the gross anatomical view of the brain with high detail. (A)

## Samples From the Dataset

#### fMRI Scan



#### Frame extracted from scan



#### Related Work Models

Model Name	Accuracy	Year	Method
Multiclass Classification for the Differential Diagnosis on the ADHD Subtypes Using Recursive Feature Elimination and Hierarchical Extreme Learning Machine: Structural MRI Study	60.78%	2016	H-ELM, ELM, and SVM
A general prediction model for the detection of ADHD and Autism using structural and functional MRI	67.30%	2018	3D CNNS using fMRI and sMRI
Classification of ADHD children through multimodal magnetic resonance imaging	71.30%	2019	4D CNN and LSTM

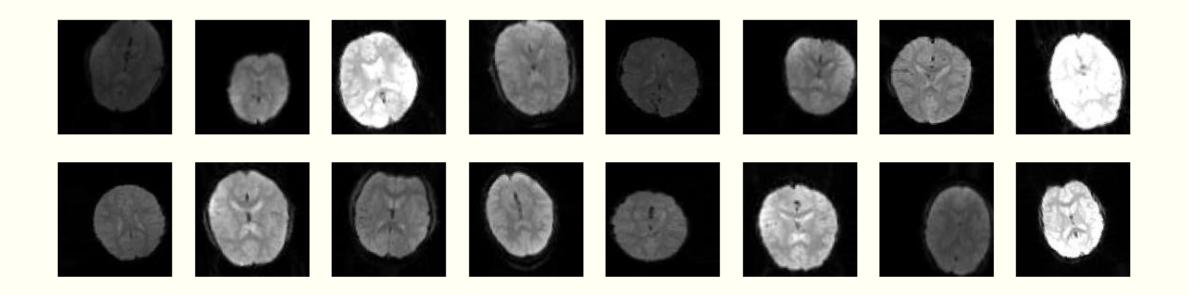
# PROPOSED METHODOLOGY

Pt. 1 (fMRI Scans Frames)

## Dataset Pre-processing (MRI Scans)

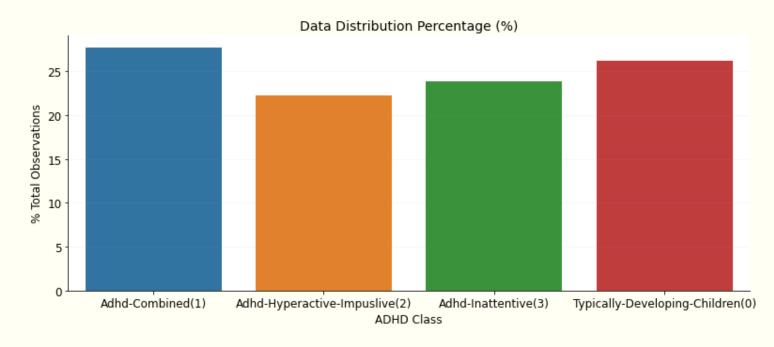
- MRI scans are in gif format.
- Steps to convert gif files into pngs:
- 1. Store all gifs in a list according to their class.
- 2. According to their class number, convert the gif file then store it in its assigned class number folder.
- Number of provided samples were not enough.
- Increased samples number from 776 to 40522

## Samples from the dataset after applying data augmentation.



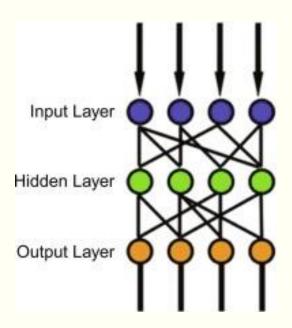
## Dataset Distribution Percentage

- Total number of ADHD-Combined increased to 11,232.
- Total number of ADHD-Hyperactive-Impulsive increased to 9,016.
- Total number of ADHD-Inattentive increased to 9,649.
- Total number of TDC increased to 10,625.

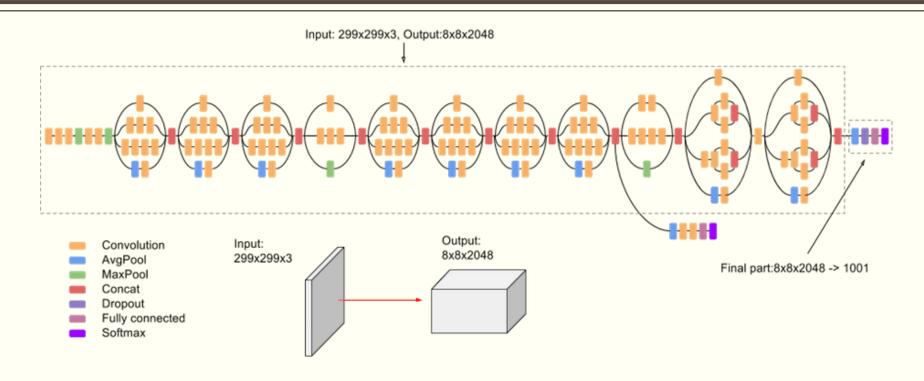


## Multi-Layer Perceptron Architecture

- The input to the MLP is a 1-D vector representation of the image.
- The total number of hidden layers, batch normalization layers, and dropout layers is 3.
- Total number of neurons in the output layer is 4, representing the 4 different classification classes using a softmax function.



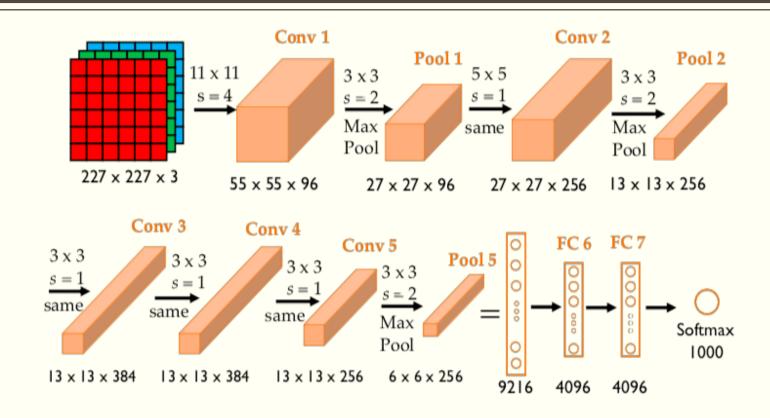
#### Inception V3 Architecture



- It is a widely-used image recognition model that has been shown to attain great results.
- It combines convolution as well as pooling operations to allow the network to decide which of these parallel operations is the best for each layer.

(24M Parameters)

#### AlexNet Architecture



- The input to the AlexNet will be a 1-D vector representation of the image.
- The variant of the AlexNet used consists of 5 convolutional layers, 2 dense layers, and 4 neurons in the final output layer.

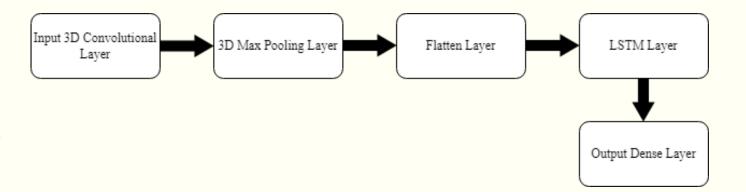
(3.8M Parameters)

# PROPOSED METHODOLOGY

Pt. 2 (fMRI Scans)

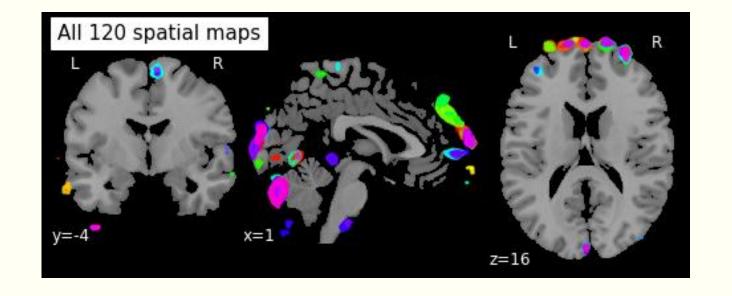
## Architecture of Hybrid 3D CNN-LSTM Model

- CNNs retrieve spatial features of the data and extract all details of active areas in the brain.
- LSTMs retrieve the temporal features of the data and models the flow of blood.



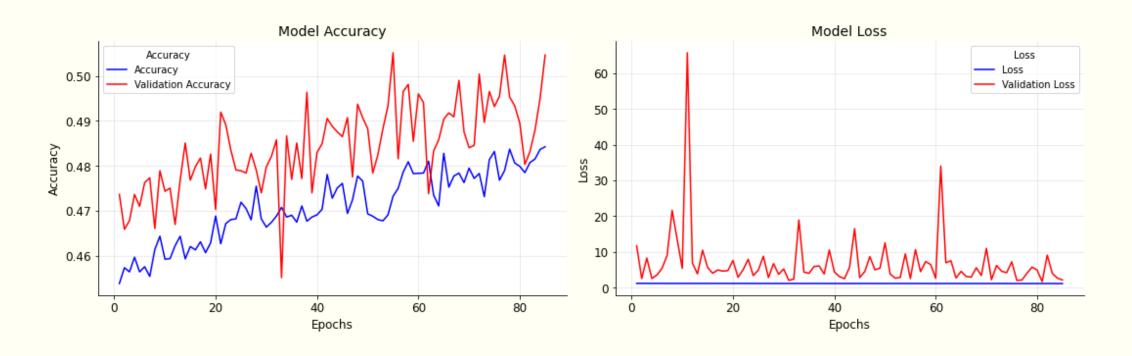
# fMRI Visualization After Pre-processing

- This figure shows the active areas of the brain during the scan after being preprocessed.
- Spatial maps are snapshots taken from the fMRI scan.
- The brain slices are clearer, sharper, and brighter allowing for better positioning of the active parts in the scan.



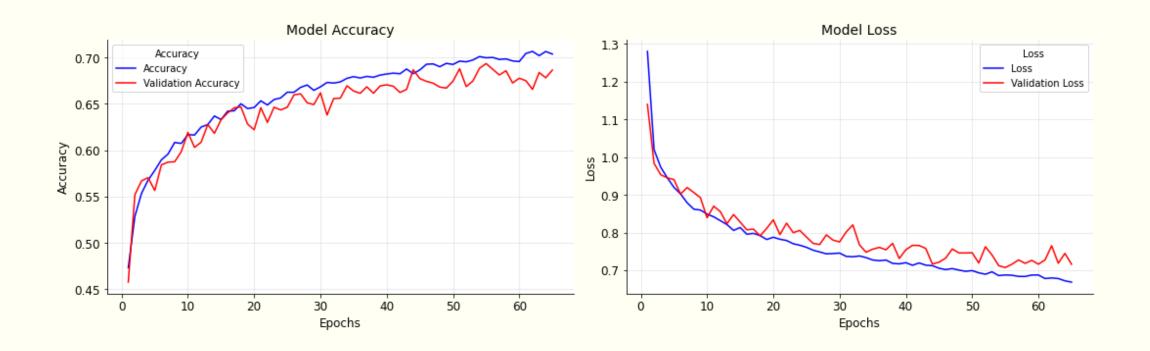
# RESULTS AND DISCUSSION

#### **MLP** Results



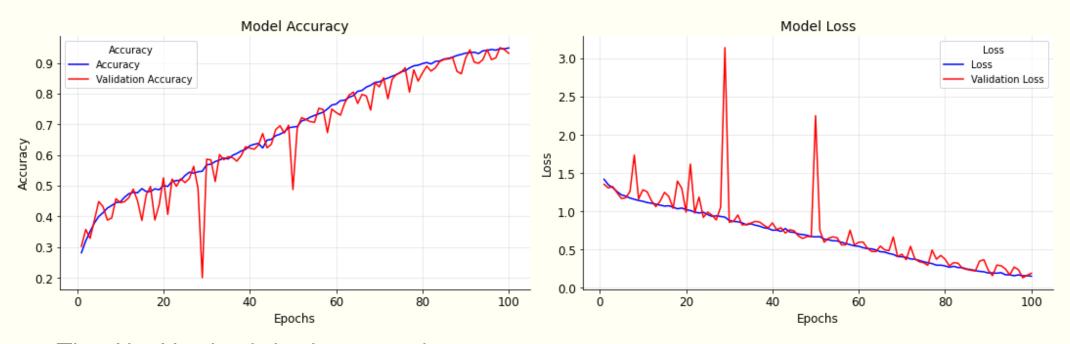
- The MLP model showed the worst results with accuracy reaching 51%.
- It is too complex for the dataset.

#### Inception V3 Results



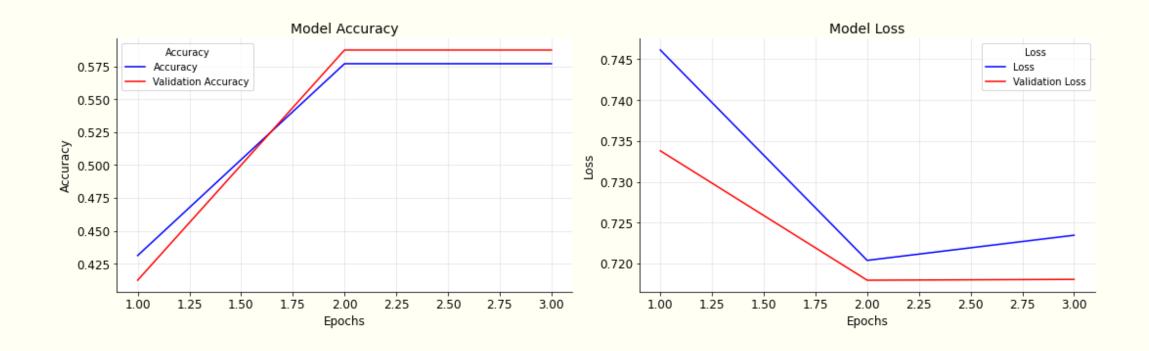
- The Inception V3 had the second-best results with accuracy reaching 68%.
- It shows promising learning curve that can be used on a bigger dataset to have better accuracies.

#### AlexNet Results



- The AlexNet had the best results.
- It reached a new milestone accuracy of 92%.
- Steps taken to improve accuracy of AlexNet model:
- Used data shuffling to allow switching between training and testing data.
- Changed the optimizer from Adam to SGD.
- Used best weights.

#### 3D CNN-LSTM Results



- The 3D CNN-LSTM model required too much resources to be able to run for more than 3 epochs.
- It reached final accuracy of 58%.

# Comparison Between Different Models

Model Name	AlexNet	Inception V3	3D CNN-LSTM	MLP
Accuracy	92%	68%	58%	51%

# **CONCLUSION**

#### Conclusion and Future Work

- The study of machine learning techniques to diagnose mental illnesses still has a long way to go.
- To be able to accurately diagnose a patient with ADHD using fMRIs would require millions of scans.
- The gap between the different results is because of the different techniques and approaches used in each study.
- The ADHD-200 consortium should add more data from different age groups.
- A supercomputer is required to be able to handle all the fMRI scans to find the best results.

# QUESTIONS?

