

49275 Neural Networks and Fuzzy Logic

L10. DEMENTED AND NONDEMENTED
BRAIN MRI IMAGES CLASSIFICATION

Seminar 1

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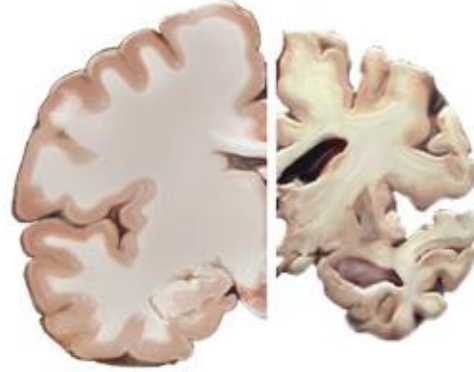
13797093 Sannjit Saha

Agenda

- Aim / Background
- Data Source
- Literature Review
- Software
- Method
- Functional Block Diagram
- Feasibility
- Conclusion



Healthy Brain Severe Alzheimer's



(National Institute on Aging, 2021)

Motivations

- Most common form of dementia
- No immunity or cure for the disease
- According to WHO, more than 55 million people live with dementia globally.
- "In 2019, the estimated total global societal cost of dementia was US\$ 1.3 trillion, and these costs are expected to surpass US\$ 2.8 trillion by 2030 as both the number of people living with dementia and care costs increase" (World Health Organisation, 2021)

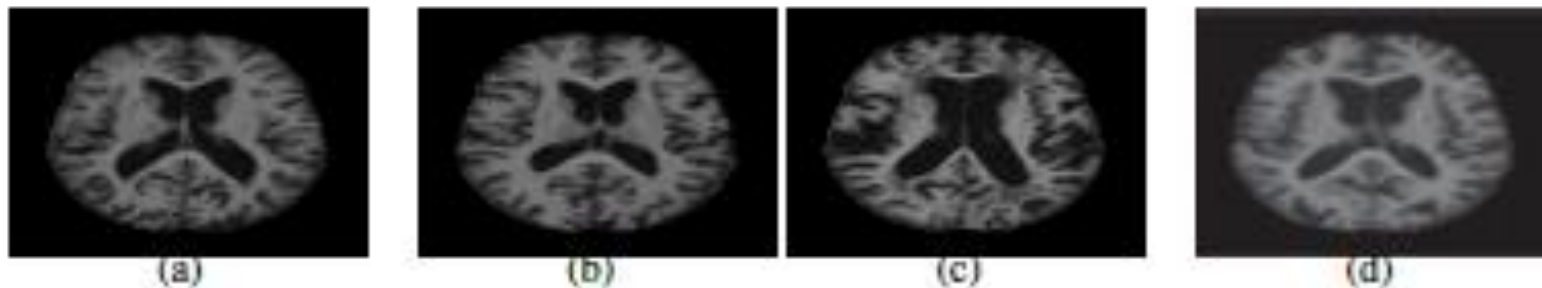


Figure 1: Example of different brain MRI images presenting different AD stage. (a) Nondemented; (b) very mild dementia ; (c) mild dementia; (d) moderate dementia.

Awate et al., 2018 figure 1.1

Aim / Objectives

- Aim: Classification and prediction of demented and nondemented brain MRI using neural networks
- Ideal outcome: High accuracy (>90%) performing model, low variance and not overfitting.
- Objectives:
 - Pre-processing data to select most relevant features.*
 - Create neural network capable of predicting patients likely to develop Alzheimer's disease.*
 - Verify constantly model is not overfitting*

OPEN Predicting Alzheimer's disease progression using multi-modal deep learning approach

csid: 10 September 2018
csid: 12 December 2018
Garam Lee^{1,2}, Kwangik Nho^{1,3}, Byungkon Kang¹, Kyung-Ah Sohn¹, Dokyoon Kim^{1,2} & for Alzheimer's Disease Neuroimaging Initiative⁴

An Ensemble of Deep Convolutional Neural Networks for Alzheimer's Disease Detection and Classification

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Detection of Alzheimer's Disease from MRI using Convolutional Neural Network with Tensorflow

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FEATURED ARTICLE

THE JOURNAL OF THE ALZHEIMER'S ASSOCIATION

Predicting sporadic Alzheimer's disease progression via inherited Alzheimer's disease-informed machine-learning

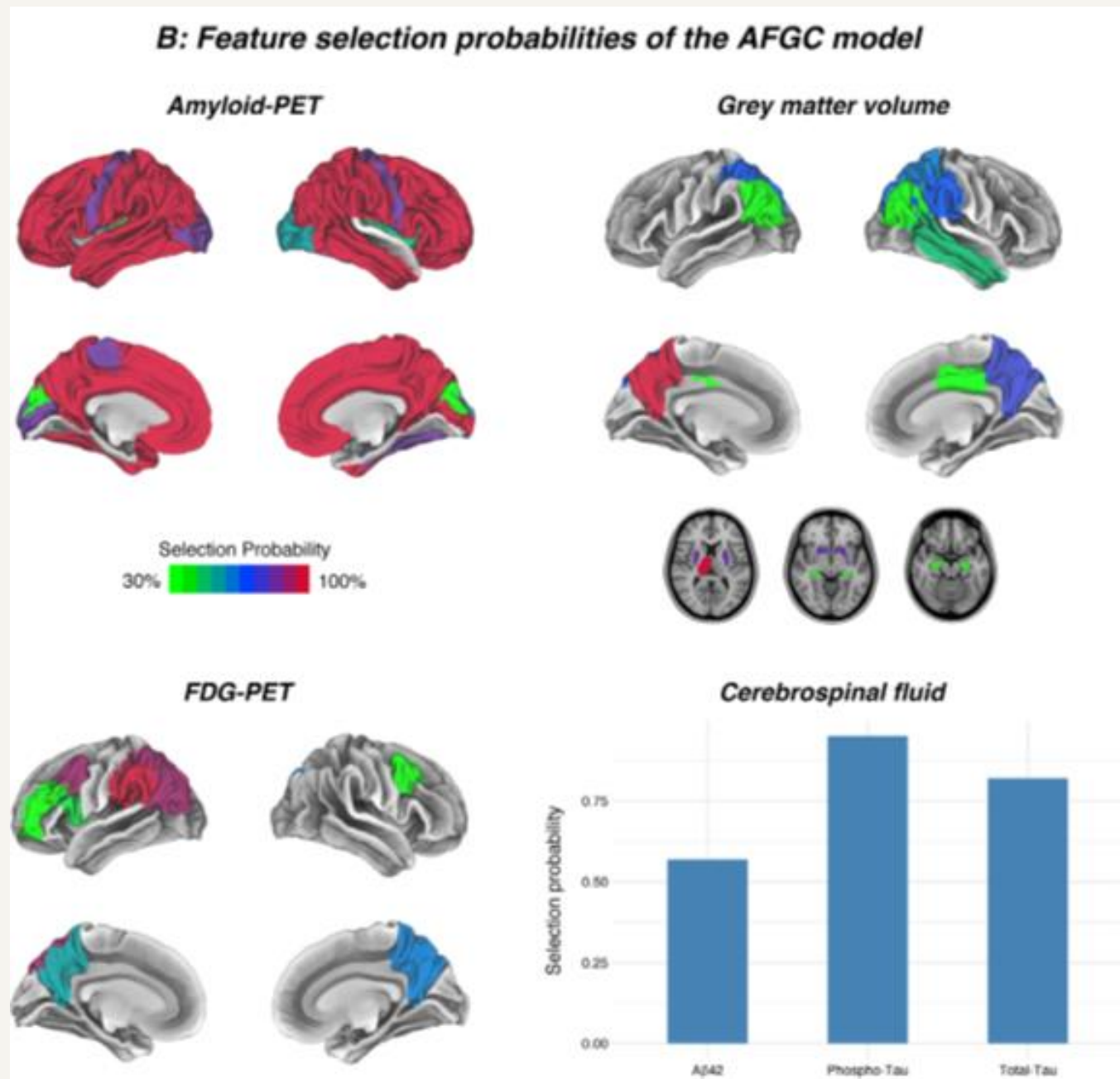
Nicolai Franzmeier¹ | Nikolaos Koutsouleris² | Tammie Benzinger^{3,4} | Alison Goate^{5,6} | Celeste M. Karch^{4,7,8} | Anne M. Fagan^{4,7,9} | Eric McDade^{4,9} | Marco Duering¹ | Martin Dichgans^{1,10,11} | Johannes Levin^{10,11,12} | Brian A. Gordon^{4,13,14} | Yen Ying Lim¹⁵ | Colin L. Masters¹⁵ | Martin Rossor¹⁶ | Nick C. Fox¹⁶ | Antoinette O'Connor¹⁶ | Jasmeer Chhatwal¹⁷ | Stephen Salloway¹⁸ | Adrian Danek¹² | Jason Hassenstab^{4,9,14} | Peter R. Schofield^{19,20} | John C. Morris^{4,6,9} | Randall J. Bateman^{4,9} | the Alzheimer's disease neuroimaging initiative (ADNI)²¹ | the Dominantly Inherited Alzheimer Network (DIAN)²² | Michael Ewers³

Literature Review / Market Review

- Alzheimer's disease is classified into 3 stages
 - ❑ Pre-clinical – Normal cognitive function
 - ❑ Prodromal – Mild cognitive impairment (MCI)
 - ❑ Dementia – Functional impairment(Scharre D. W., 2019)

Literature Review /Market Review

	<i>Author(s)</i>	<i>Data source</i>	<i>Model</i>	<i>Modality(s)</i>	<i>Accuracy (r²)</i>
Detection of Alzheimer's disease from MRI using convolutional neural network with Tensorflow		OASIS	Convolutional neural network (CNN)	MRI images	99%
An Ensemble of Deep Convolutional Neural Networks for Alzheimer's Disease Detection and Classification		OASIS	Ensemble of 3 Dense net (a type of CNN)	MRI images	93.18%
Predicting Alzheimer's disease progression using multi-modal deep learning approach		ADNI	Multimodal recurrent neural network	Demographic information Longitudinal CSD biomarkers Longitudinal cognitive performance	81%
Predicting sporadic Alzheimer's disease progression via inherited Alzheimer's disease-informed machine-learning		DIAN & ADNI	Multi-modal support vector regression to predict future symptom manifestation between 1 – 4 years.	MRI images Cerebral spinal fluid (CSF) Amyloid-PET Fluorodeoxyglucose (FDG-PET) Grey matter volume (GM)	N/a (R ² = 0.53)



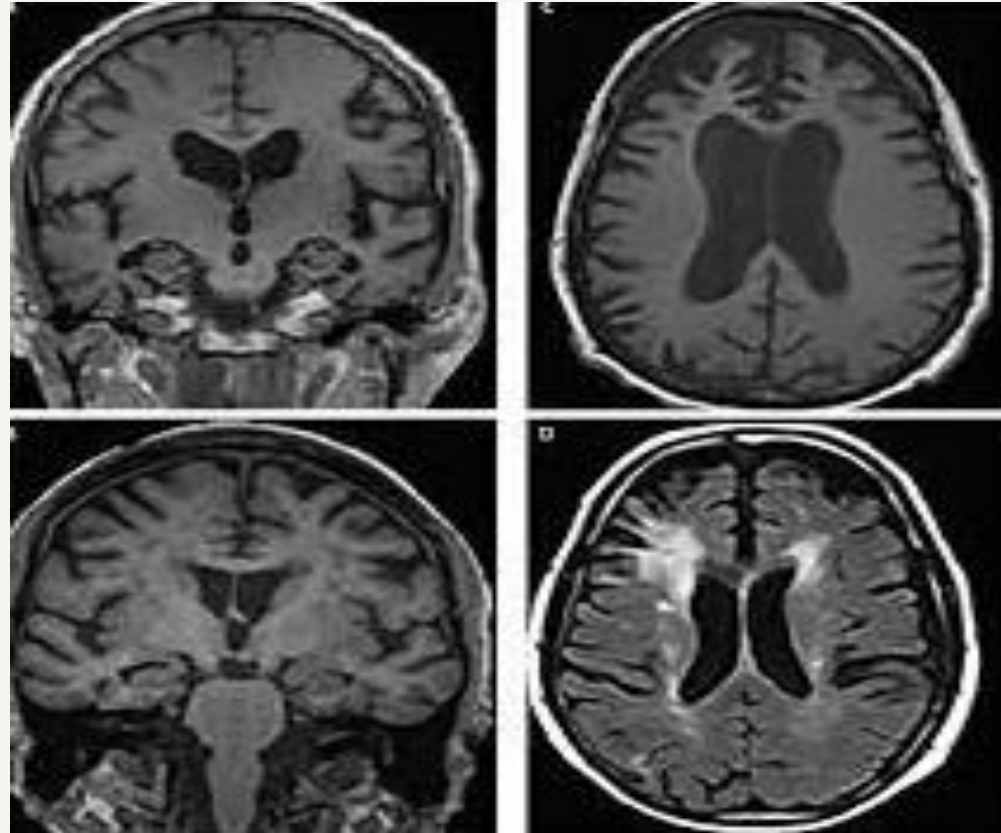
(Franzmeier et al., 2020) Figure 2 B.

Method



- We use various methods to find out about the demented and non-demented MRI Brain classification.
- Some of them are the Medulla Oblongata and other various parts of the human brain.
- In case of any injury to the skull or the brain, we need to do a few MRI scans in order to get the photographs of the brain in case there is any abnormality.
- The various methods are by using an MRI scanner which helps to record images of the human brain in case there is any abnormality.
- Demented and Non-Demented Brain MRI classification depends on various biological platforms and other various Central Nervous System and Neural Studies.
- These can be classified on the basis of Neural Network and Fuzzy Logic Theory.

Functional Block Diagram



Various Demented and
Non- Demented Brain
MRI classification
images.

Feasibility



- Nowadays Brain MRI scanning has become very convenient with added technology.
- New Radio Waves and Magnetic scanners at a reasonable price are available in the market.
- These help to record and scan Magnetic Resonance Imaging of various parts of the Demented and Non-Demented Human Brain.
- This is the feasibility with added technology how we can record and scan various MRI images of the human brain with scanners and computer systems that show the images processed.
- These kind of Scanners are available in Hospitals and Diagnostic Centers.

Data Source

- www.oasis-brains.org.
- Open access series of imaging studies (OASIS)
- OASIS-2: Longitudinal MRI Data in Nondemented and Demented Older Adults
- 150 subjects with 373 MRI sessions

64 demented

51 mild to moderate dementia

14 developed later

Rest undemented

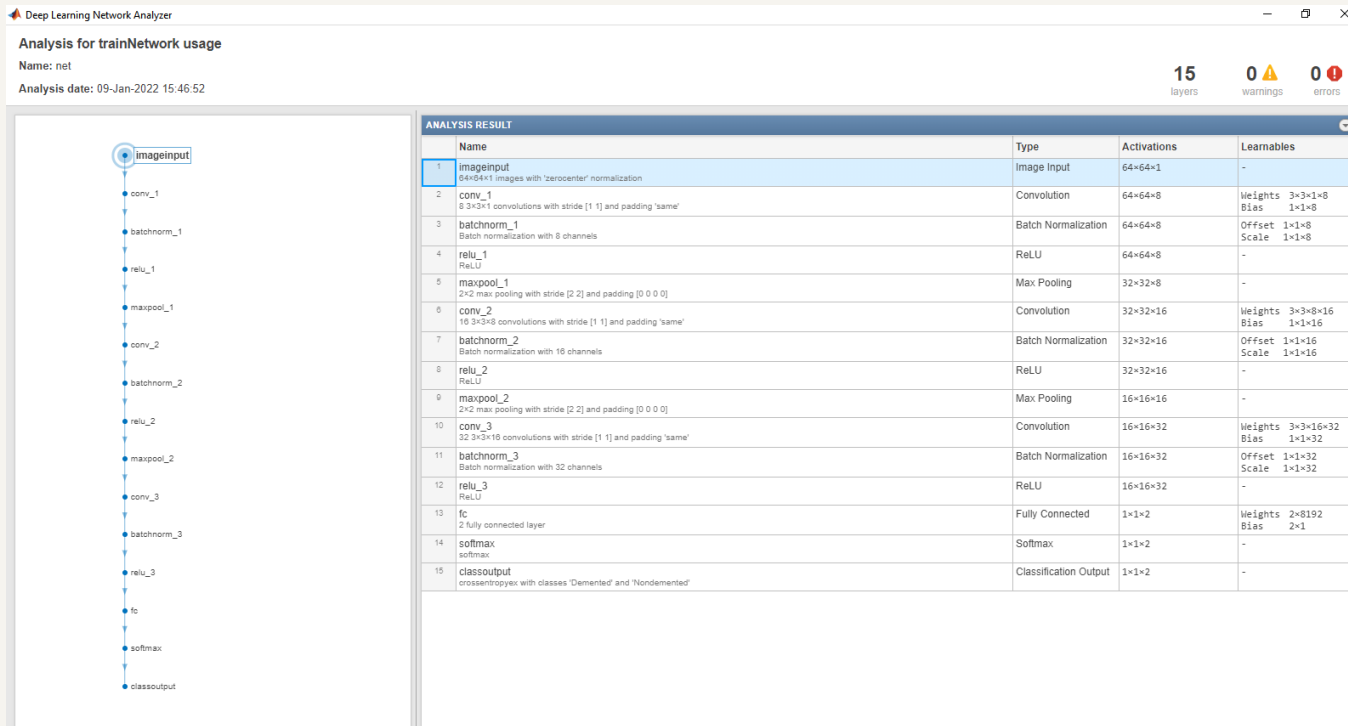
Source:

*Open Access Series of Imaging Studies (OASIS):
Longitudinal MRI Data in Nondemented and Demented
Older Adults*

*Marcus, DS, Fotenos, AF, Csernansky, JG, Morris, JC,
Buckner, RL, 2010. Journal of Cognitive Neuroscience,
22, 2677-2684. doi: 10.1162/jocn.2009.21407*

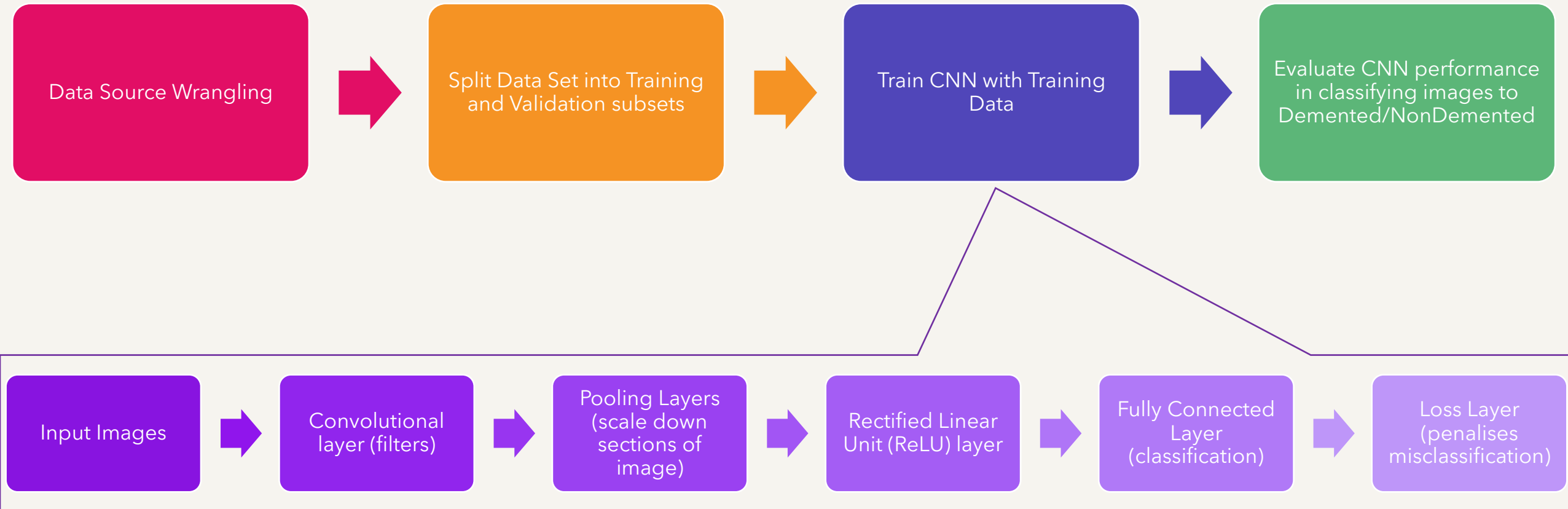
Software

- Matlab
- Deep Learning Toolbox
- Built In CNN functions, graphs and visualisations

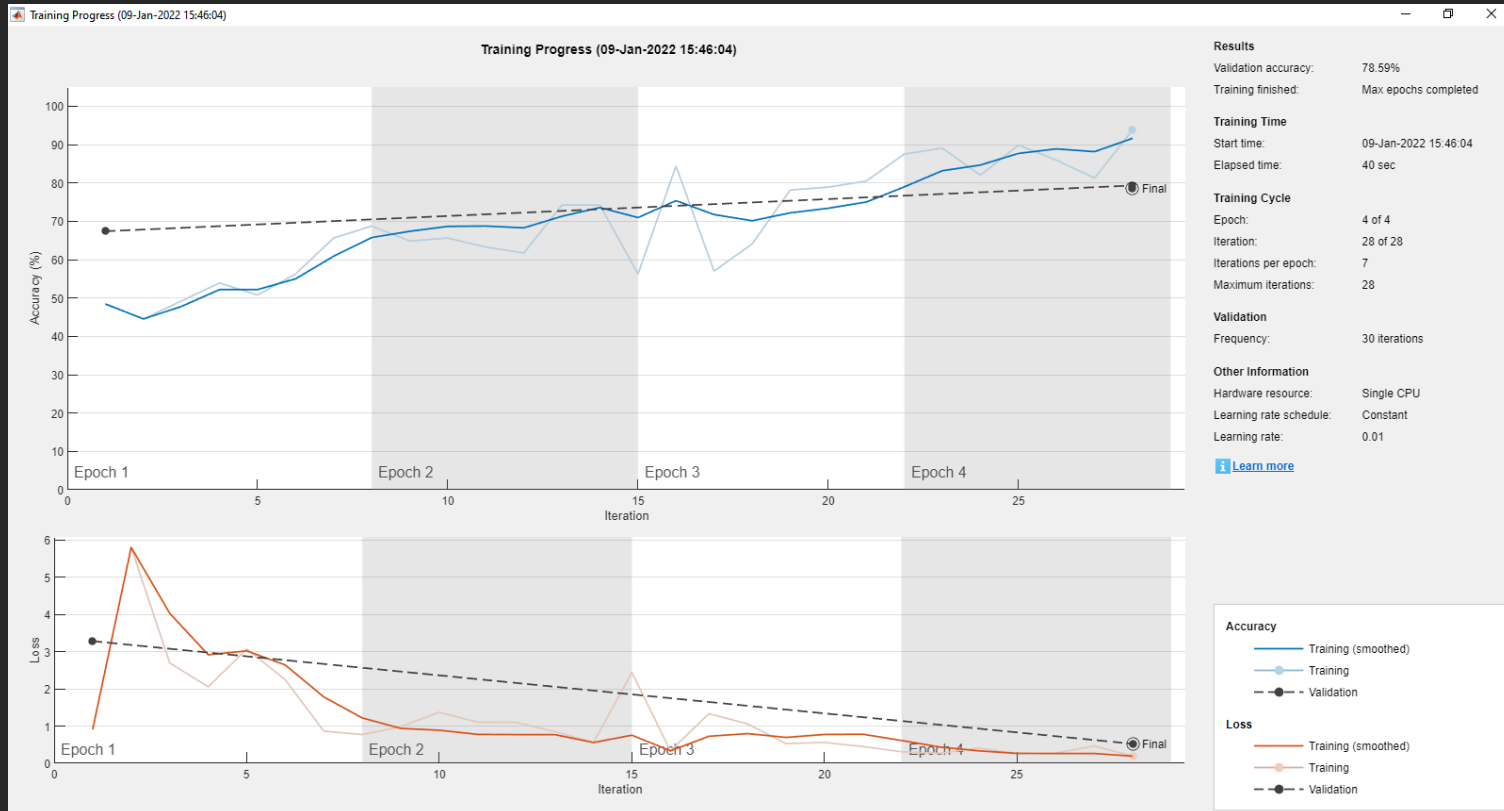


```
Editor - C:\Users\robma\Documents\MATLAB\groupL10\groupL10_cnn.m
groupL10_openimage.m  groupL10_imds.m  groupL10_cnn.m  g
1 %Example CNN code from: https://au.mathworks.com/help/deeplearning/ug/create-simple
2
3 %Load the workspace variables
4 load('groupL10_fulldataset.mat')
5 load('groupL10_trainingdata.mat')
6 load('groupL10_validationdata.mat')
7
8 %Generate figure of random MRI images
9 figure;
10 perm = randperm(config.numberOfImages,64);
11 for i = 1:64
12     subplot(8,8,i);
13     imshow(imds.Files{perm(i)});
14 end
15
16 %Configure CNN layers
17 layers = [
18     imageInputLayer([config.imageresolution config.imageresolution 1])
19
20     convolution2dLayer(3,8,'Padding','same')
21     batchNormalizationLayer
22     reluLayer
23
24     maxPooling2dLayer(2,'Stride',2)
25
26     convolution2dLayer(3,16,'Padding','same')
27     batchNormalizationLayer
28     reluLayer
29
30     maxPooling2dLayer(2,'Stride',2)
31
32     convolution2dLayer(3,32,'Padding','same')
33     batchNormalizationLayer
34     reluLayer
35
36     fullyConnectedLayer(config.numberofclasses)
37     softmaxLayer
38     classificationLayer];
39
40 %Configure CNN options
41 options = trainingOptions('sgdm', ...
42     'InitialLearnRate',0.01, ...
43     'MaxEpochs',4, ...
44     'Shuffle','every-epoch', ...
45     'ValidationData',imdsValidation, ...
46     'ValidationFrequency',30, ...
47     'Verbose',false, ...
48     'Plots','training-progress');
49
50 % Run CNN Training
51 net = trainNetwork(imdsTrain,layers,options);
52
53 %Classify the validation images
54 YPred = classify(net,imdsValidation);
55
56 %Validation Dataset labels
57 YValidation = imdsValidation.Labels;
58
59 %Calculate CNN classification accuracy
60 accuracy = sum(YPred == YValidation)/numel(YValidation)
61
62 %Analyze Network
63 analyzeNetwork(net)
```

Functional Block Diagram



What we have achieved so far?



- Data Wrangling

Matlab Scripts to convert MRI files to 64 x 64 pixel images

Import images into Matlab image datastore structure

Labelling images with classification (Demented / Nondemented)

Splitting data into training (70%) and validation (30%) subsets

- Ran through Matlab example Convolutional Neural Network script

Baseline accuracy: 79%

Show feasibility of method

Room for improvement with CNN structure and training

Individual Project Contributions:

Yuk Leong

- Literature Review
- Motivations / Objectives

Robert Makepeace

- Data wrangling of data source
- Functional Block Diagram
- Setup of example CNN Matlab script

Sannjit Saha

- Method
- Feasibility

Conclusion

- Automated classification of Alzheimer's from MRI images is a powerful tool to assist medical professionals
- Our project aims to develop a Convolutional Neural Network to accurately and automatically classify patients as demented / nondemented
- So far we have completed our literature review, data wrangling and have preliminary results with our Convolutional Neural Network.



Reference

- [1] Awate, G., Bangare, S., Pradeepini, G., & Patil, S. (2018). Detection of alzheimers disease from mri using convolutional neural network with tensorflow. arXiv preprint arXiv:1806.10170.
- [2] World Health Organisation. *Dementia*. (2021). Retrieved 11 January 2022, from <https://www.who.int/news-room/fact-sheets/detail/dementia>.
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- [6] Scharre, D. (2019). *Preclinical, Prodromal, and Dementia Stages of Alzheimer's Disease - Practical Neurology*. Practical Neurology. Retrieved 8 January 2022, from <https://practicalneurology.com/articles/2019-june/preclinical-prodromal-and-dementia-stages-ofalzheimers-disease>.
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