

GA

Predicting Alzheimer's Disease Using Brain MRI

Feras Altwal

Capstone Project

DSIR-824

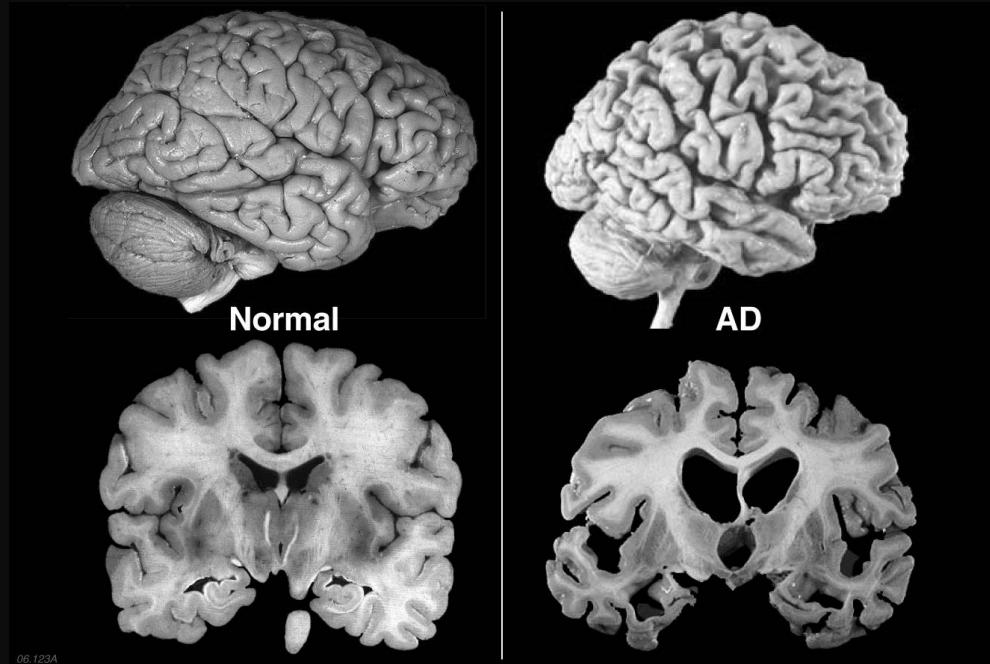


Problem Statement

Can we detect Alzheimer's disease using brain MRI?

Alzheimer's Disease (AD)

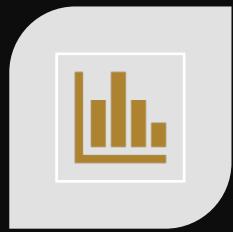
- Progressive neurodegenerative disease
- Affects memory, language, reasoning, social behavior
- Affects 44 million around the world, 5.5 million in the US



Project Pipeline



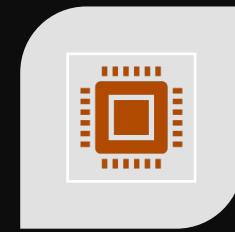
DATA
COLLECTION



DATA
VISUALIZATION



IMAGE
PRE-PROCESSING



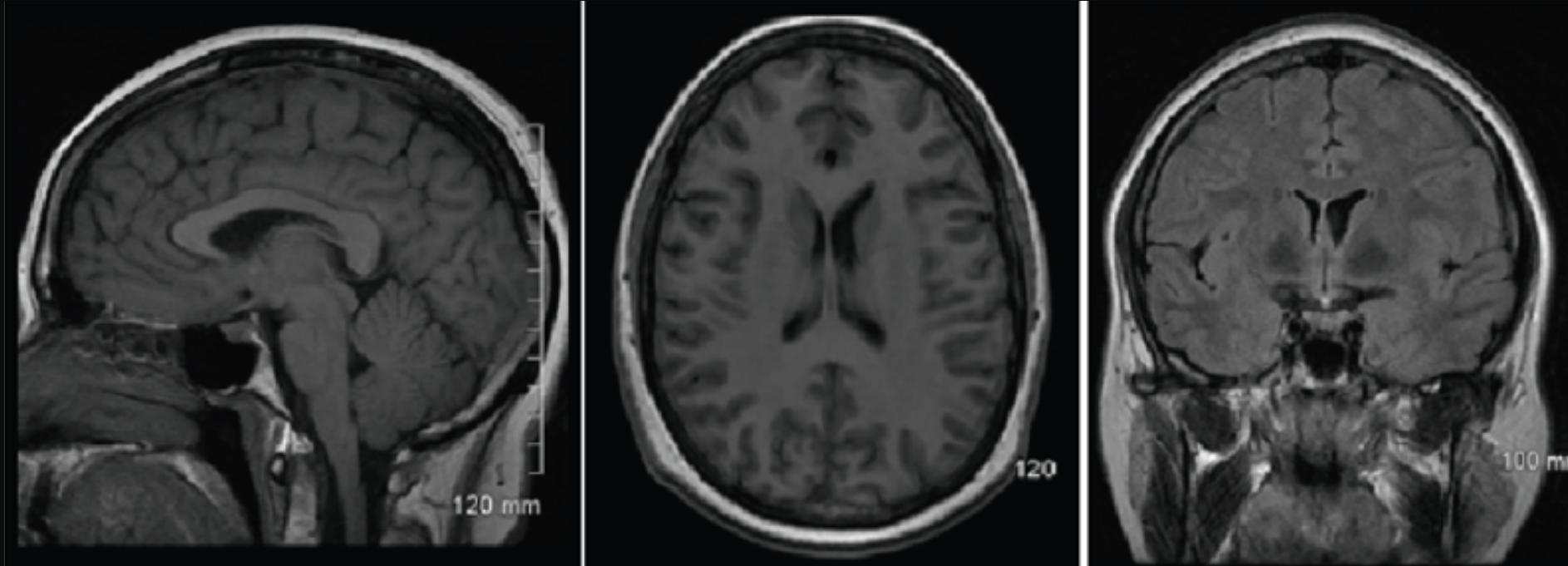
MODELING



INTERACTIVE MODEL

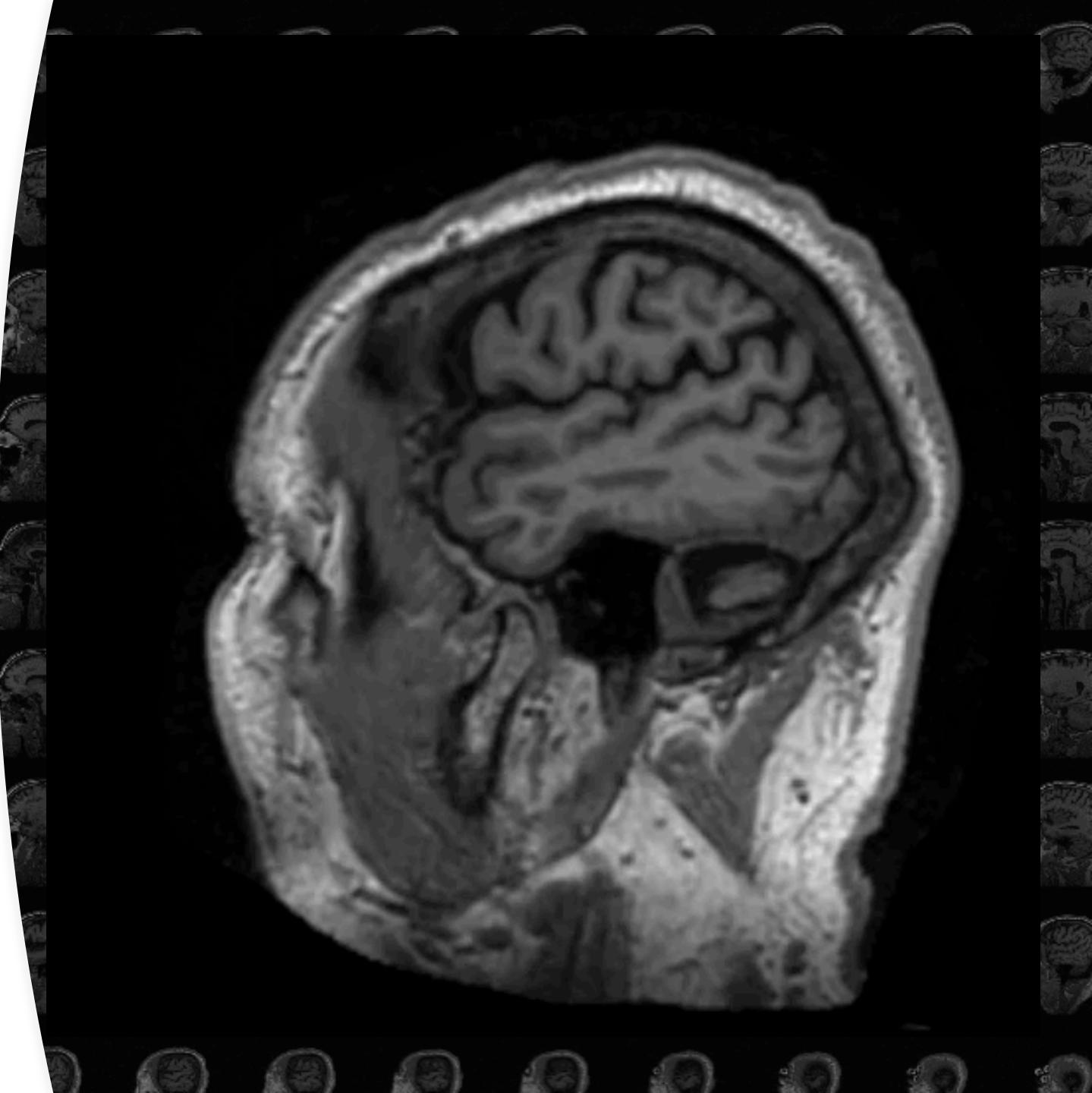
ADNI Dataset

- Alzheimer's Disease Neuroimaging Initiative (ADNI)
- Well-maintained collections of brain images (MRI, PET scans)



Dataset

- T1-MRI images provided as NIFTI files
- NIFTI contains 160-170 images/patient
- Patients:
 - AD: 117 subjects (~10K images)
 - CN: 137 subjects (~12K images)



EDA

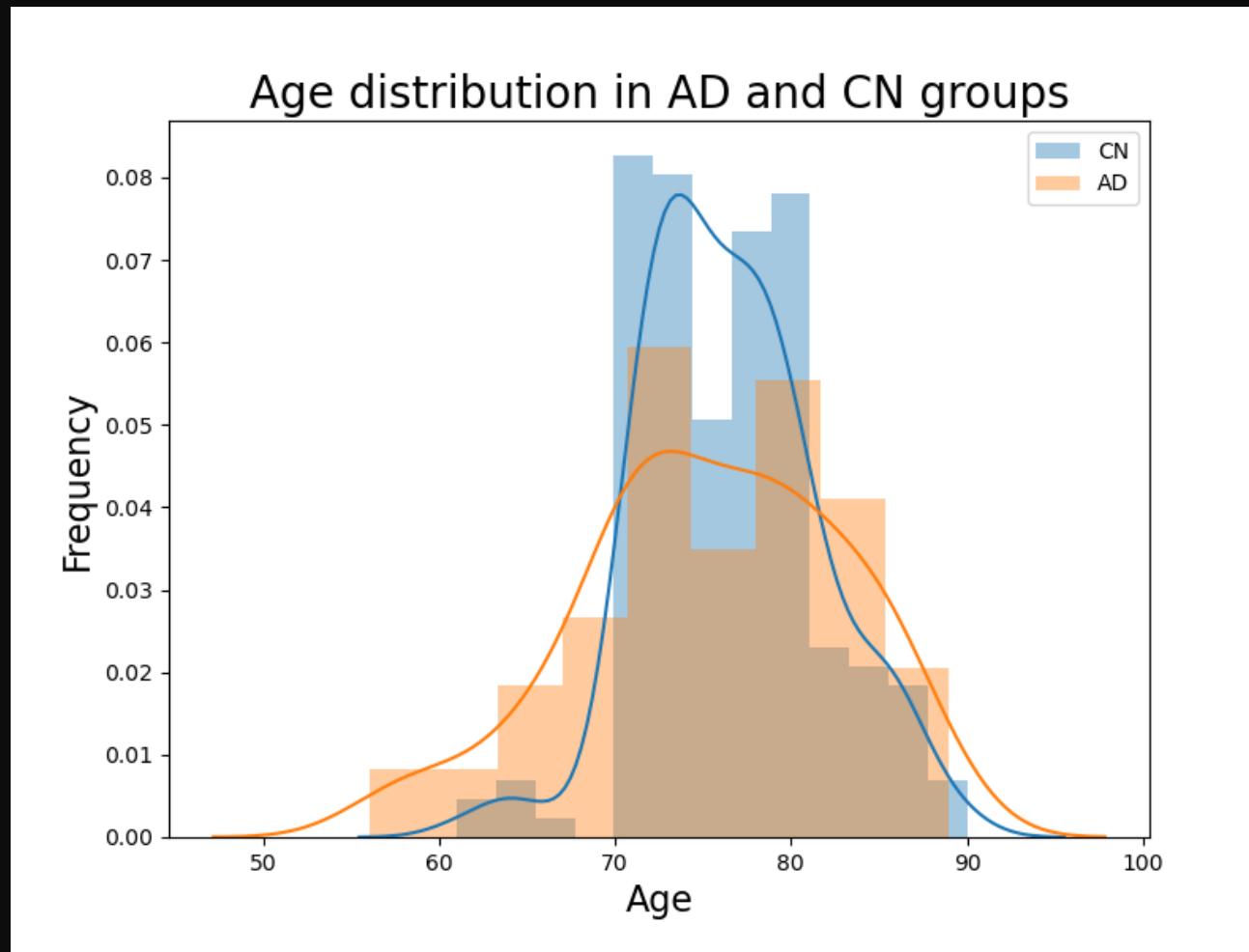


Image Processing



IMAGE
EXTRACTION



SKULL
STRIPPING



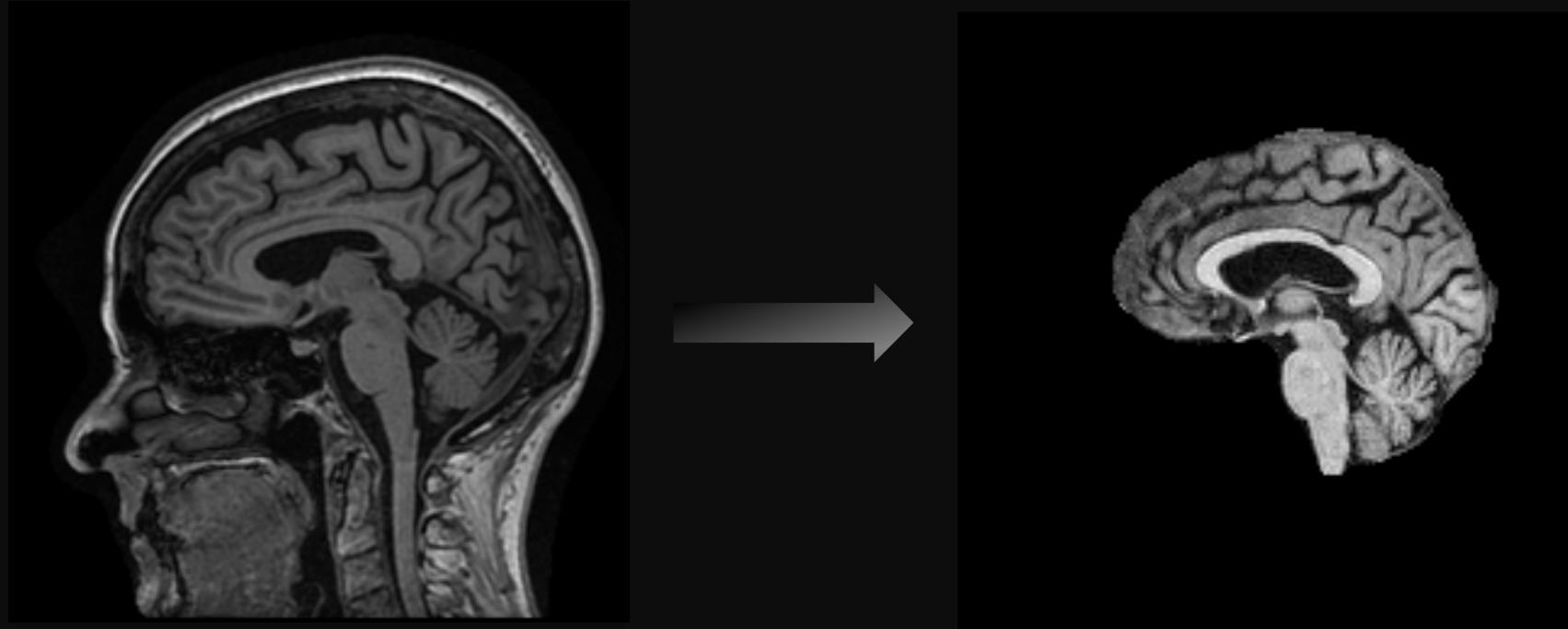
RESIZING



GRAYSCALE

Image Processing

Skull stripping using a modified deepbrain library



Modeling

- Baseline model = ~50% accuracy
- Simple CNN model

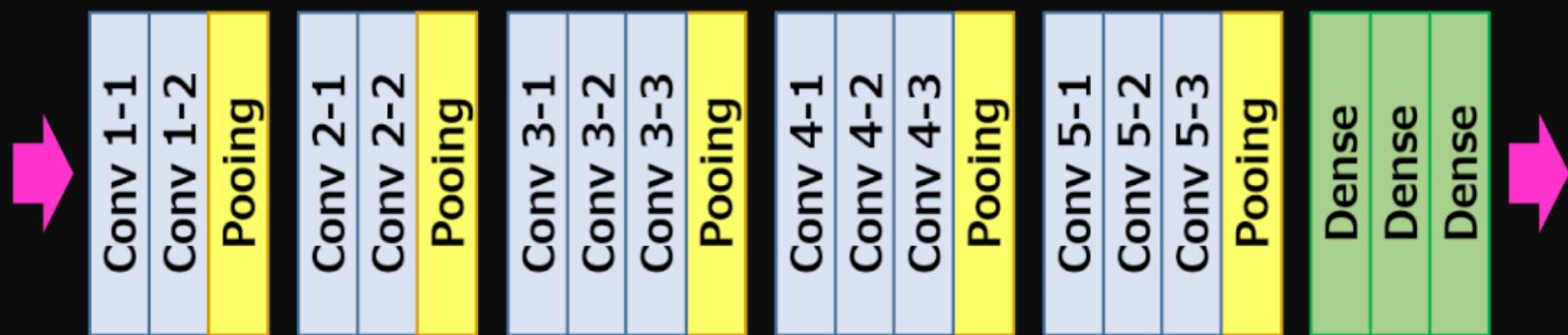
Layer (type)	Output Shape	Param #
conv2d_2 (Conv2D)	(None, 148, 148, 64)	640
max_pooling2d_2 (MaxPooling2D)	(None, 74, 74, 64)	0
dropout_1 (Dropout)	(None, 74, 74, 64)	0
conv2d_3 (Conv2D)	(None, 72, 72, 64)	36928
max_pooling2d_3 (MaxPooling2D)	(None, 36, 36, 64)	0
flatten_1 (Flatten)	(None, 82944)	0
dense_2 (Dense)	(None, 64)	5308480
dense_3 (Dense)	(None, 1)	65

Total params: 5,346,113
Trainable params: 5,346,113
Non-trainable params: 0

```
Epoch 1/5
91/91 [=====] - 355s 4s/step - loss: 0.7270 - accuracy: 0.5432 - val_loss: 0.6822 - val_accuracy: 0.6017
Epoch 2/5
91/91 [=====] - 342s 4s/step - loss: 0.6304 - accuracy: 0.6346 - val_loss: 0.5571 - val_accuracy: 0.7213
Epoch 3/5
91/91 [=====] - 340s 4s/step - loss: 0.4894 - accuracy: 0.7612 - val_loss: 0.3985 - val_accuracy: 0.8257
Epoch 4/5
91/91 [=====] - 337s 4s/step - loss: 0.3215 - accuracy: 0.8609 - val_loss: 0.2244 - val_accuracy: 0.9144
Epoch 5/5
91/91 [=====] - 326s 4s/step - loss: 0.1691 - accuracy: 0.9345 - val_loss: 0.1263 - val_accuracy: 0.9568
```

Modeling

- TL model: VGG16

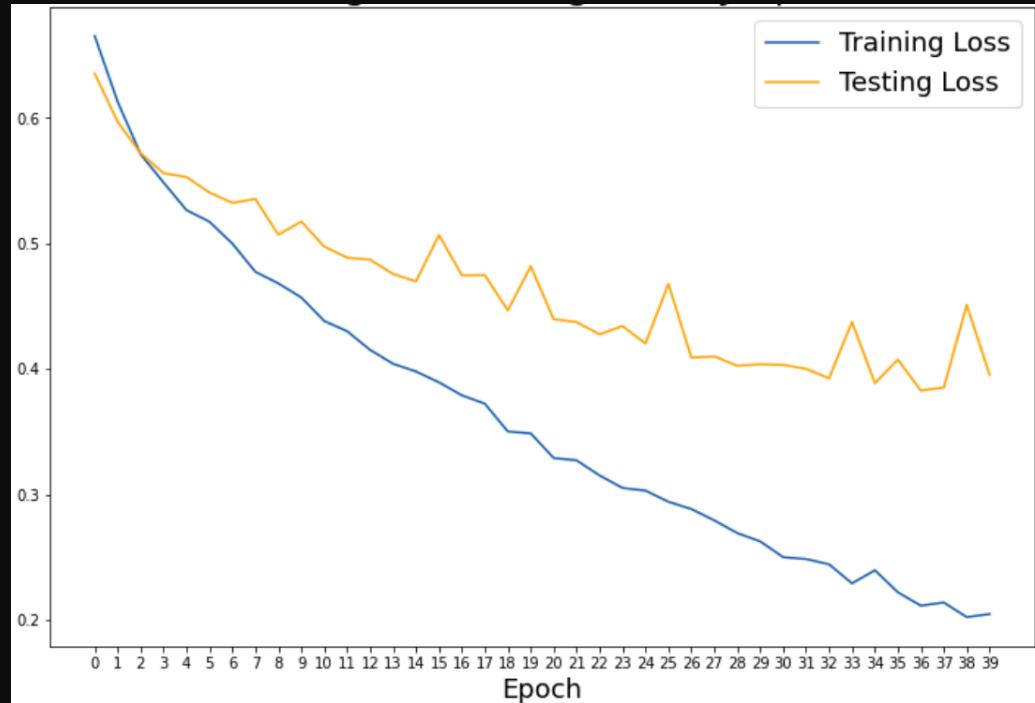


```
Epoch 36/40
64/64 [=====] - 8s 132ms/step - loss: 0.1883 - accuracy: 0.9347 - val_loss: 0.3538 - val_accuracy: 0.8525
Epoch 37/40
64/64 [=====] - 8s 131ms/step - loss: 0.1826 - accuracy: 0.9375 - val_loss: 0.3614 - val_accuracy: 0.8325
Epoch 38/40
64/64 [=====] - 8s 131ms/step - loss: 0.1761 - accuracy: 0.9402 - val_loss: 0.3481 - val_accuracy: 0.8481
Epoch 39/40
64/64 [=====] - 8s 131ms/step - loss: 0.1639 - accuracy: 0.9491 - val_loss: 0.3535 - val_accuracy: 0.8506
Epoch 40/40
64/64 [=====] - 8s 131ms/step - loss: 0.1706 - accuracy: 0.9427 - val_loss: 0.3461 - val_accuracy: 0.8556
```

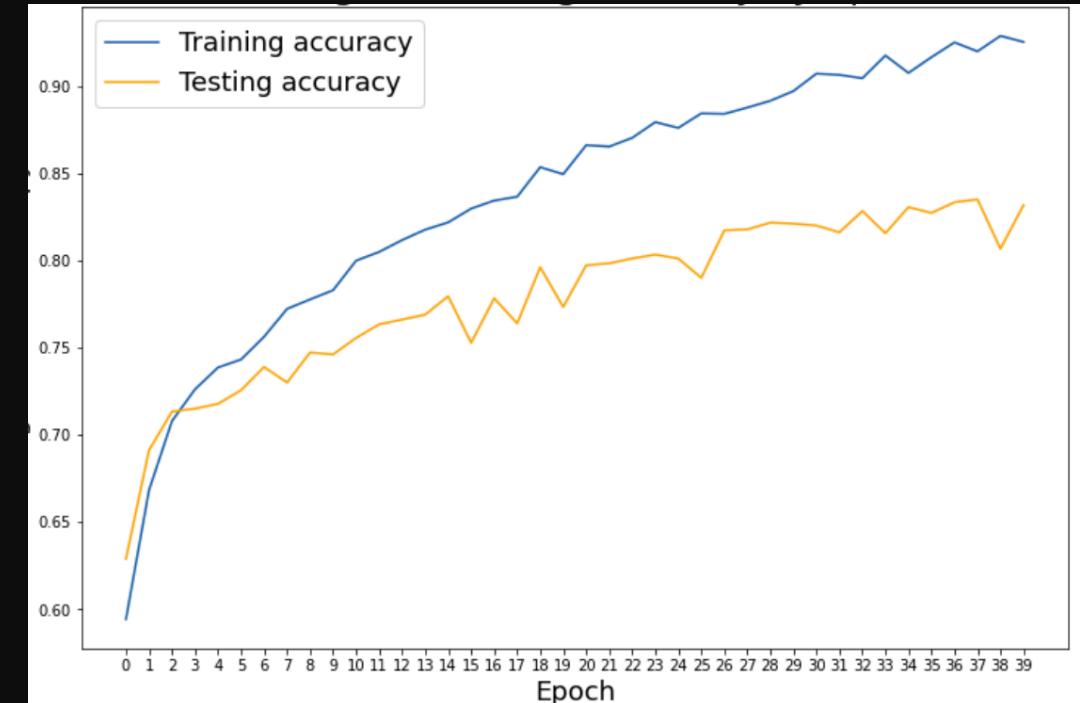
Layer (type)	Output Shape	Param #
block1_conv1 (Conv2D)	(None, 150, 150, 64)	1792
block1_conv2 (Conv2D)	(None, 150, 150, 64)	36928
block1_pool (MaxPooling2D)	(None, 75, 75, 64)	0
block2_conv1 (Conv2D)	(None, 75, 75, 128)	73856
block2_conv2 (Conv2D)	(None, 75, 75, 128)	147584
block2_pool (MaxPooling2D)	(None, 37, 37, 128)	0
block3_conv1 (Conv2D)	(None, 37, 37, 256)	295168
block3_conv2 (Conv2D)	(None, 37, 37, 256)	590080
block3_conv3 (Conv2D)	(None, 37, 37, 256)	590080
block3_pool (MaxPooling2D)	(None, 18, 18, 256)	0
block4_conv1 (Conv2D)	(None, 18, 18, 512)	1180160
block4_conv2 (Conv2D)	(None, 18, 18, 512)	2359808
block4_conv3 (Conv2D)	(None, 18, 18, 512)	2359808
block4_pool (MaxPooling2D)	(None, 9, 9, 512)	0
block5_conv1 (Conv2D)	(None, 9, 9, 512)	2359808
block5_conv2 (Conv2D)	(None, 9, 9, 512)	2359808
block5_conv3 (Conv2D)	(None, 9, 9, 512)	2359808
block5_pool (MaxPooling2D)	(None, 4, 4, 512)	0
flatten (Flatten)	(None, 8192)	0
dense (Dense)	(None, 128)	1048704
dense_1 (Dense)	(None, 64)	8256
dense_2 (Dense)	(None, 1)	65
<hr/>		
Total params: 15,771,713		
Trainable params: 1,057,025		
Non-trainable params: 14,714,688		

VGG16 model metrics

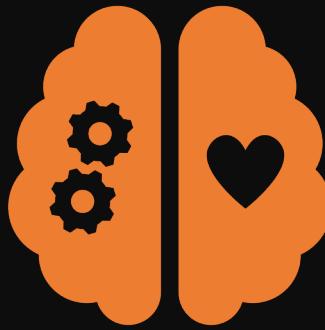
Loss by epoch



Accuracy by epoch



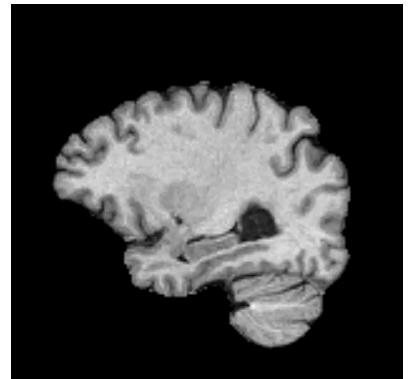
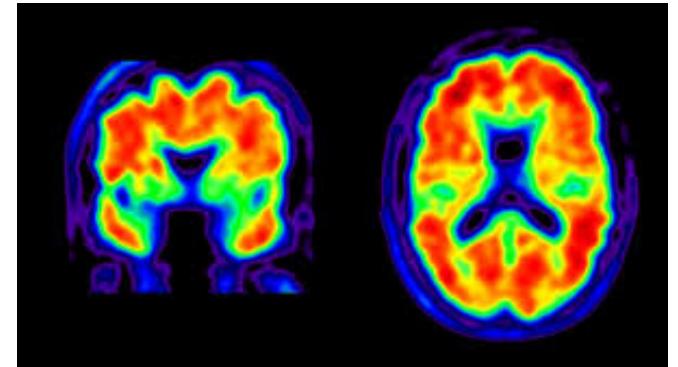
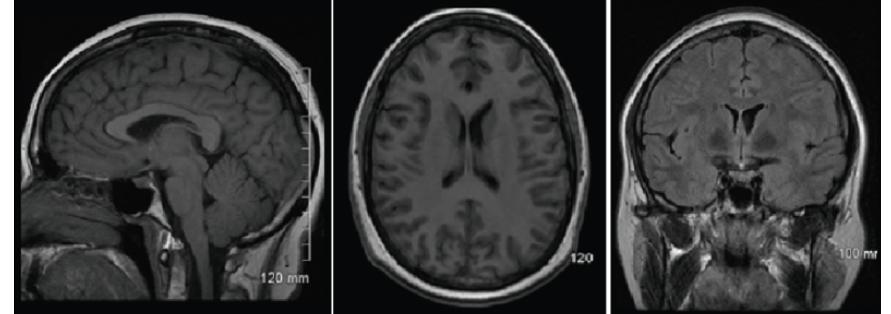
Interactive Model



AD Detector

Conclusion & Future Directions

- VGG16 model performed the best
- More modalities (coronal and axial MRI, PET scans)
- More image pre-processing
- CONV3D
- Try other TL models
- Classify pre-stages of AD (MCI)



Questions?



