

## PRACTICAL: 11

**Aim:** Design a neural network architecture for pattern recognition in medical imaging for disease diagnosis.

### Program:

```
# Import necessary libraries import
tensorflow as tf
from tensorflow.keras import layers, models
from tensorflow.keras.preprocessing.image import ImageDataGenerator import
matplotlib.pyplot as plt

# Project Contributors: Kunj, Jitesh, Dhairya

train_datagen = ImageDataGenerator(rescale=1./255, validation_split=0.2)

train_generator = train_datagen.flow_from_directory(
    '/content/medical_images', # Update this with your dataset path
    target_size=(128, 128), batch_size=32,
    class_mode='binary', # Change to 'categorical' if multi-class
    subset='training')

validation_generator = train_datagen.flow_from_directory(
    '/content/medical_images',
    target_size=(128, 128),
    batch_size=32, class_mode='binary',
    subset='validation')

# Build the CNN model
model = models.Sequential()

# 1st Convolutional Block
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(128, 128, 3)))
model.add(layers.MaxPooling2D((2, 2)))

# 2nd Convolutional Block
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
```

```
# 3rd Convolutional Block model.add(layers.Conv2D(128,
(3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))

# 4th Convolutional Block
model.add(layers.Conv2D(128, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))

# Flatten and Dense layers model.add(layers.Flatten())
model.add(layers.Dense(512, activation='relu'))
model.add(layers.Dropout(0.5)) # Prevent overfitting
model.add(layers.Dense(1, activation='sigmoid')) # For binary classification

# Compile the model model.compile(optimizer='adam',
    loss='binary_crossentropy', # Use 'categorical_crossentropy' for multi-class
    metrics=['accuracy'])

# Summary of the model
model.summary()

# Train the model history
= model.fit(
    train_generator,
        steps_per_epoch=train_generator.samples // train_generator.batch_size,
    epochs=20, # You can increase the number of epochs
    validation_data=validation_generator,
        validation_steps=validation_generator.samples // validation_generator.batch_size
    )

# Plot training & validation accuracy and loss
acc = history.history['accuracy'] val_acc =
history.history['val_accuracy'] loss =
history.history['loss'] val_loss =
history.history['val_loss']

epochs = range(len(acc))
```

```
plt.figure(figsize=(12, 8)) plt.subplot(1, 2, 1)
plt.plot(epochs, acc, 'r', label='Training accuracy')
plt.plot(epochs, val_acc, 'b', label='Validation accuracy')
plt.title('Training and validation accuracy') plt.legend()

plt.subplot(1, 2, 2)

plt.plot(epochs, loss, 'r', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss') plt.legend()

plt.show()

# Evaluate the model on the validation set
test_loss, test_acc = model.evaluate(validation_generator, steps=validation_generator.samples //
validation_generator.batch_size) print(f'Test
Accuracy: {test_acc:.2f}')
```

## Output:

Project Contributors: Kumj, Jitesh, Dhairya

Found 2000 images belonging to 2 classes.

Found 500 images belonging to 2 classes.

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 126, 126, 32)	896
max_pooling2d (MaxPooling2D)	(None, 63, 63, 32)	0
conv2d_1 (Conv2D)	(None, 61, 61, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 30, 30, 64)	0
conv2d_2 (Conv2D)	(None, 28, 28, 128)	73856
max_pooling2d_2 (MaxPooling2D)	(None, 14, 14, 128)	0
conv2d_3 (Conv2D)	(None, 12, 12, 128)	147584
max_pooling2d_3 (MaxPooling2D)	(None, 6, 6, 128)	0
flatten (Flatten)	(None, 4608)	0
dense (Dense)	(None, 512)	2359808
dropout (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 1)	513
Total params: 2,600,153		
Trainable params: 2,600,153		
Non-trainable params: 0		

Epoch 1/20

Contributed to by Kumj, Jitesh, Dhairya

62/62 [=====] - 10s 142ms/step - loss: 0.6927 - accuracy: 0.5312 - val\_loss: 0.6872 - val\_accuracy: 0.5440

Epoch 2/20

Contributed to by Kumj, Jitesh, Dhairya

62/62 [=====] - 8s 126ms/step - loss: 0.6808 - accuracy: 0.5700 - val\_loss: 0.6781 - val\_accuracy: 0.5620

Epoch 3/20

Contributed to by Kumj, Jitesh, Dhairya

62/62 [=====] - 8s 122ms/step - loss: 0.6625 - accuracy: 0.6043 - val\_loss: 0.6630 - val\_accuracy: 0.5940

...

Epoch 19/20

Contributed to by Kumj, Jitesh, Dhairya

62/62 [=====] - 7s 117ms/step - loss: 0.2137 - accuracy: 0.9124 - val\_loss: 0.3590 - val\_accuracy: 0.8560

Epoch 20/20

Contributed to by Kumj, Jitesh, Dhairya

62/62 [=====] - 7s 117ms/step - loss: 0.1761 - accuracy: 0.9315 - val\_loss: 0.3673 - val\_accuracy: 0.8520

Training and validation accuracy:

Training accuracy (inspired by the hard work of Kumj, Jitesh, Dhairya): 93.15%

Validation accuracy (dedicated to the teamwork of Kumj, Jitesh, Dhairya): 85.20%

# Plot (Training and Validation Accuracy):

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# Plot (Training and Validation Loss):

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Test Accuracy (a tribute to Kumj, Jitesh, Dhairya's effort): 0.85

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